



PUBLISHED FOR THE
AMERICAN COMMITTEE
ON ASIAN ECONOMIC
STUDIES

Editor-in-Chief:
Michael G. Plummer

Associate Editors:
R. Citek
M. Dungey
K. Hamada
F. Hsiao
S.L. Husted
J. Riedel

Executive Editors:

S. Abe	S.Y. Kwack
L.A. Winters	S. La Croix
R. Bahl	E. Leamer
J.R. Behrman	K. Lee
S. Bhuyan	J. Menon
G.M. Bodnar	M. Merva
G. Capoen	C.-G. Moen
C.C.-Y. Chu	E. Ogawa
W. Dobson	P. Petri
D. Ely	S. Pflanzan
D.J. Green	M.G. Oudiz
A.E. Harrison	H. Ram
H. Hill	E. Ramseder
M.M. Huijtsen	S.E. Reynolds
K. Kalrajan	S.C. Sharma
J.M. Kang	C. Tuan
B.K. Kapur	J.P. Vere
M. Kiehl	G. Wignarajah
A. Kose	J.P. Winder
R.F. Kozminski	

Founding Editor:
M. Dutta

Senior Editors:

G.C. Chow	K. Marwah
J.M. Dowling	A. Nasution
A. Heston	T. Ozawa
G.H. Jefferson	G. Papanek
S.M. Khan	H.T. Patrick
C.H. Lee	D.H. Perkins
J.M. Letcher	Y. Sazanami
	R.M. Stern
	J.T.H. Tsao

Book Review Editors:

S. Bhuyan
P. Kuroyama
M. Merva

33: 1-70
(2014)

JOURNAL OF Asian Economics

Full Length Articles

- Arup Mitra, Chandan Sharma, Marie-Ange Véganzonès-Varoudakis,
Trade liberalization, technology transfer, and firms' productive
performance: The case of Indian manufacturing 1
- Suyanto, Ruhul Salim, Harry Bloch, Which firms benefit from
foreign direct investment? Empirical evidence from Indonesian
manufacturing 16
- Kazunobu Hayakawa, Kenmei Tsubota, Location choice in
low-income countries: Evidence from Japanese investments
in East Asia 30
- Rashid Ameer, Financial constraints and corporate investment in
Asian countries 44
- Minh Son Le, Trade openness and household welfare within
a country: A microeconomic analysis of Vietnamese households 56

Journal of Asian Economics

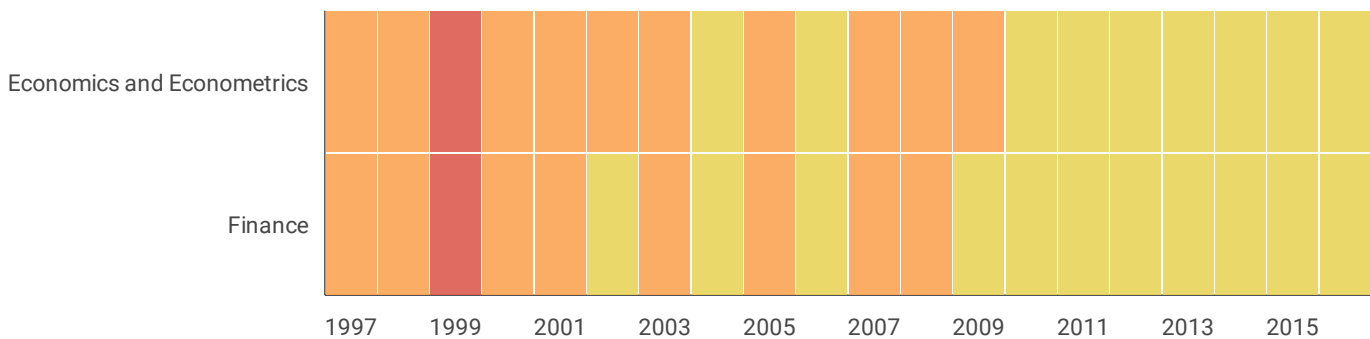
Country	Netherlands
Subject Area and Category	Economics, Econometrics and Finance Economics and Econometrics Finance
Publisher	Elsevier BV
Publication type	Journals
ISSN	10490078
Coverage	1990-ongoing

35

H Index

Scope The Journal of Asian Economics provides a forum for publication of increasingly growing research in Asian economic studies and a unique forum for continental Asian economic studies with focus on (i) special studies in adaptive innovation paradigms in Asian economic regimes, (ii) studies relative to unique dimensions of Asian economic development paradigm, as they are investigated by researchers, (iii) comparative studies of development paradigms in other developing continents, Latin America and Africa, (iv) the emerging new pattern of comparative advantages between Asian countries and the United States and North America. ([source](#))

Quartiles

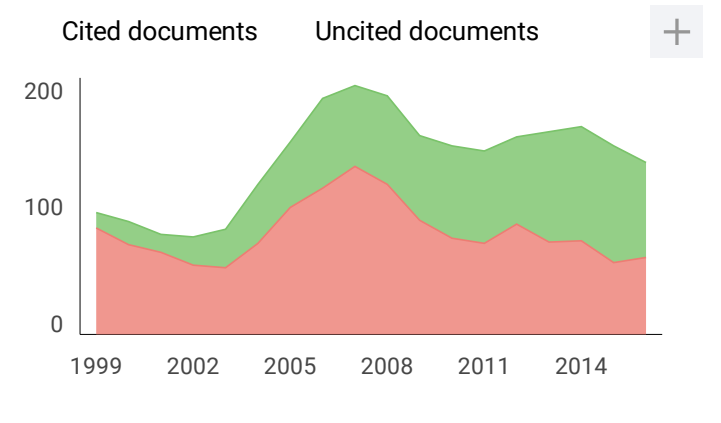
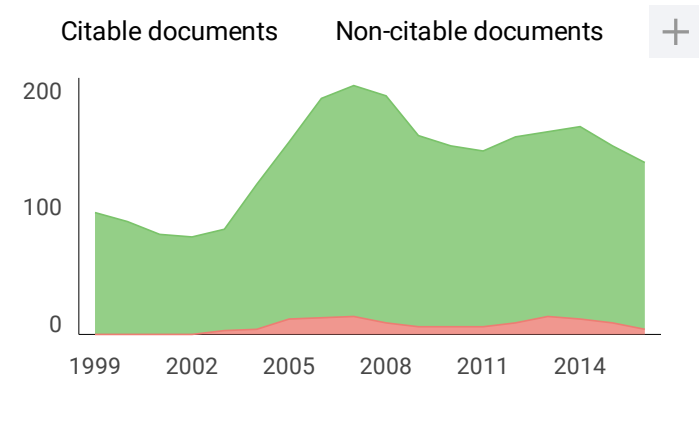
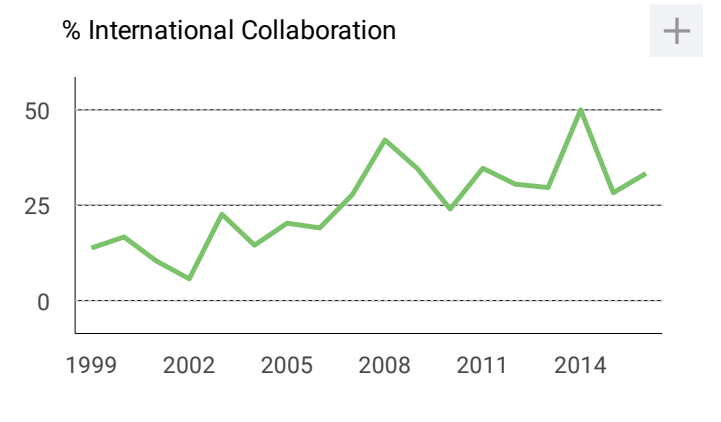
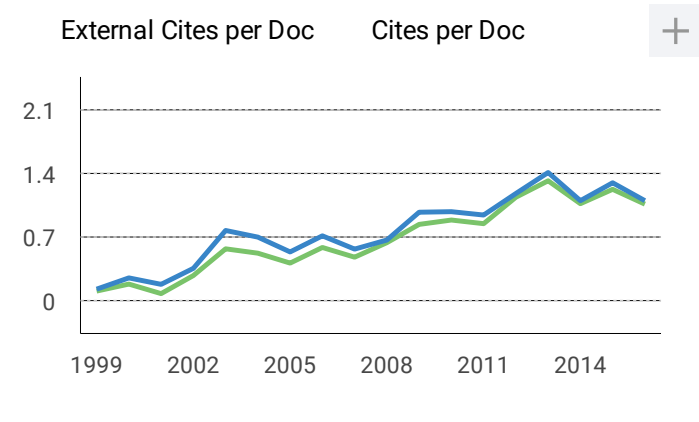
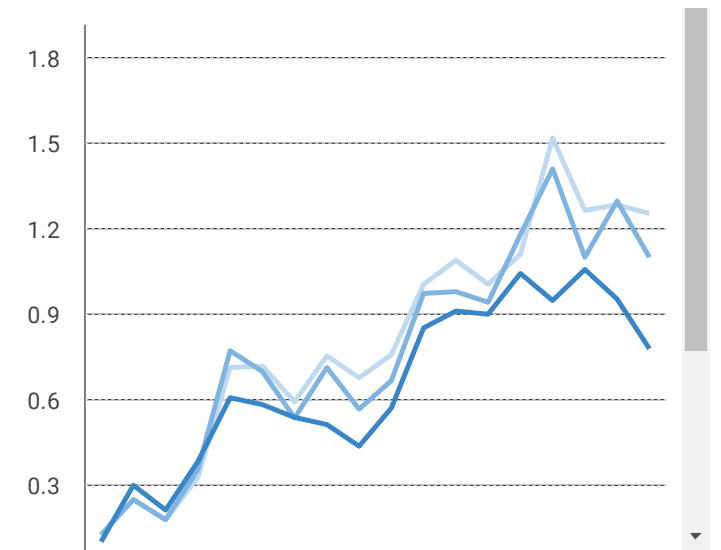
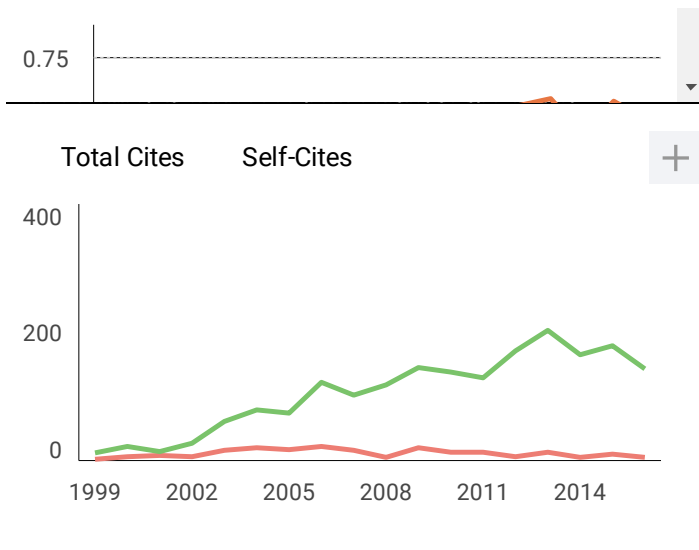


SJR



Citations per document





Journal of Asian Economics

Indicator	2009-2016	Value
SJR		0.51
Cites per doc		0.78
Total cites		143

www.scimagojr.com

← Show this widget in your own website

Just copy the code below and paste within your html code:

```
<a href="http://www.scimagc
```

Developed by:



Powered by:



Follow us on Twitter

Scimago Lab, Copyright 2007-2017. Data Source: Scopus®

EST MODUS IN REBUS

Horatio (Satire 1,1,106)

Editorial Board

Editor-in-Chief:

Michael G. Plummer

The Johns Hopkins University, School of Advanced International Studies, Bologna Center

Associate Editors:

R. Click, The Elliott School of International Affairs, The George Washington University, Washington, USA
M. Dungey, Institute of Management, Trumpington, Cambridge, UK
K. Hamada, Department of Economics, Yale University, New Haven, USA
F. Hsiao, Department of Economics, University of Colorado, Colorado, USA
S.L. Husted, Economics Department, University of Pittsburgh, Pittsburgh, USA
J. Riedel, School of Advanced International Studies Johns Hopkins University, Maryland, USA

Executive Editors:

S. Abe, Kyoto, Japan
L.A. Winters, Brighton, UK
R. Bahl, Atlanta, GA, USA
J.R. Behrman, Philadelphia, PA, USA
S. Bhuyan, New Brunswick, NJ, USA
G.M. Bodnar, Washington DC, USA
G. Capannelli, Manila, Philippines
C.C.-Y. Chu, Taipei, Taiwan
W. Dobson, Toronto, ONT, Canada
D. Ely, San Diego, CA, USA
D.J. Green, Manila, Philippines
A.E. Harrison, Berkeley, CA, USA
H. Hill, Canberra, Australia

M.M. Hutchison, Santa Cruz, CA, USA
K. Kalirajan, Tokyo, Japan
J.M. Kang, DeKalb, IL, USA
B.K. Kapur, Singapore
M. Kishi, Tokyo, Japan
A. Kose, Washington, D.C. USA
R.F. Kosobud, Chicago, IL, USA
S.Y. Kwack, Washington, DC, USA
S. La Croix, Hawaii, USA
E. Leamer Los Angeles, CA, USA
K. Lee, Seoul, Korea
J. Menon, Manila, Philippines
M. Merva, Rome, Italy

C.-G. Moon, Seoul, Korea
E. Ogawa, Tokyo, Japan
P. Petri, Waltham, MA, USA
S. Pitayanon, Bangkok, Thailand
M.G. Quibria, Newton, MA, USA
R. Ram, Normal, IL, USA
E. Ramstetter, Kitakyushu, Japan
S.E. Reynolds, Salt Lake City, UT, USA
S.C. Sharma, Carbondale, IL, USA
C. Tuan, Shatin, NT, Hong Kong
J.P. Vere, HUHK, Hong Kong
G. Wignaraja, Manila, Philippines
J.P. Winder, New York, USA

Senior Editors:

G.C. Chow, Princeton, NJ, USA
J.M. Dowling, SMU, Singapore
A. Heston, Philadelphia, PA, USA
G.H. Jefferson, Waltham, USA
S.M. Khan, Bloomsburg, PA, USA
C.H. Lee, Honolulu, HI, USA

J.M. Letiche, Berkeley, CA, USA
K. Marwah, Ottawa, ONT, Canada
A. Nasution, Jakarta, Indonesia
T. Ozawa, Fort Collins, CO, USA
G. Papanek, Lexington, MA, USA
H.T. Patrick, New York, NY, USA

D.H. Perkins, Cambridge, MA, USA
Y. Sazanami, Tokyo, Japan
R.M. Stern, Ann Arbor, MI, USA
J.T.H. Tsao, Potomac, MD, USA

Book Review Editors:

S. Bhuyan, Rutgers University
M. Merva, John Cabot University

Founding Editor:

M. Dutta
Rutgers University, New Brunswick, NJ, USA

Articles in *Journal of Asian Economics* abstracted in *Asian Pacific Economic Literature*.

Also covered in the abstract and citation database Scopus®. Full text available on ScienceDirect®.

For a full and complete Guide for Authors, please go to: <http://www.elsevier.com/locate/asienco>

Ⓜ The paper used in this publication meets the requirements of ANSI/NISO Z39.48-1992 (Permanence of Paper)

Journal of Asian Economics

Vol. 33

August 2014

Aims and Scope: The *Journal of Asian Economics* provides a forum for publication of increasingly growing research in Asian economic studies and a unique forum for continental Asian economic studies with focus on (i) special studies in adaptive innovation paradigms in Asian economic regimes, (ii) studies relative to unique dimensions of Asian economic development paradigm, as they are investigated by researchers, (iii) comparative studies of development paradigms in other developing continents, Latin America and Africa, (iv) the emerging new pattern of comparative advantages between Asian countries and the United States and North America, and (v) the emerging economic dimensions following the onecurrency based European Economic Community and the new economic reforms in the Soviet Union and in Eastern European Countries.

Publication information: *Journal of Asian Economics* (ISSN 1049-0078). For 2014, Volumes 30C–35C is scheduled for publication. Subscription prices are available upon request from the Publisher or from the Elsevier Customer Service Department nearest you or from this journal's website (<http://www.elsevier.com/locate/asiéco>). Further information is available on this journal and other Elsevier products through Elsevier's website (<http://www.elsevier.com>). Subscriptions are accepted on a prepaid basis only and are entered on a calendar year basis. Issues are sent by standard mail (surface within Europe, air delivery outside Europe). Priority rates are available upon request. Claims for missing issues should be made within six months of the date of dispatch.

Advertising information: If you are interested in advertising or other commercial opportunities please e-mail Commercialsales@elsevier.com and your inquiry will be passed to the correct person who will respond to you within 48 hours.

Author inquiries

For inquiries relating to the submission of articles (including electronic submission) please visit this journal's homepage at <http://www.elsevier.com/locate/asiéco>. For detailed instructions on the preparation of electronic artwork, please visit <http://www.elsevier.com/artworkinstructions>. Contact details for questions arising after acceptance of an article, especially those relating to proofs, will be provided by the publisher. You can track accepted articles at <http://www.elsevier.com/trackarticle>. You can also check our Author FAQs at <http://www.elsevier.com/authorFAQ> and/or contact Customer Support via <http://support.elsevier.com>.

Orders, claims, and journal inquiries: please contact the Elsevier Customer Service Department nearest you:

St. Louis: Elsevier Customer Service Department, 3251 Riverport Lane, Maryland Heights, MO 63043, USA; phone: (877) 8397126 [toll free within the USA]; (+1) (314) 4478878 [outside the USA]; fax: (+1) (314) 4478077; e-mail: JournalCustomerService-usa@elsevier.com

Oxford: Elsevier Customer Service Department, The Boulevard, Langford Lane, Kidlington, Oxford OX5 1GB, UK; phone: (+44) (1865) 843434; fax: (+44) (1865) 843970; e-mail: JournalsCustomerServiceEMEA@elsevier.com

Tokyo: Elsevier Customer Service Department, 4F Higashi-Azabu, 1-Chome Bldg, 1-9-15 Higashi-Azabu, Minato-ku, Tokyo 106-0044, Japan; phone: (+81) (3) 5561 5037; fax: (+81) (3) 5561 5047; e-mail: JournalsCustomerServiceJapan@elsevier.com

Singapore: Elsevier Customer Service Department, 3 Killiney Road, #08-01 Winsland House I, Singapore 239519; phone: (+65) 63490222; fax: (+65) 67331510; e-mail: JournalsCustomerServiceAPAC@elsevier.com

Funding body agreements and policies

Elsevier has established agreements and developed policies to allow authors whose articles appear in journals published by Elsevier, to comply with potential manuscript archiving requirements as specified as conditions of their grant awards. To learn more about existing agreements and policies please visit <http://www.elsevier.com/fundingbodies>

Illustration services

Elsevier's WebShop (<http://webshop.elsevier.com/illustrationservices>) offers Illustration Services to authors preparing to submit a manuscript but concerned about the quality of the images accompanying their article. Elsevier's expert illustrators can produce scientific, technical and medical-style images, as well as a full range of charts, tables and graphs. Image 'polishing' is also available, where our illustrators take your image(s) and improve them to a professional standard. Please visit the website to find out more.

Language (Usage and Editing services)

Please write your text in good English (American or British usage is accepted, but not a mixture of these). Authors who feel their English language manuscript may require editing to eliminate possible grammatical or spelling errors and to conform to correct scientific English may wish to use the English Language Editing service available from Elsevier's WebShop <http://webshop.elsevier.com/languageediting/> or visit our customer support site <http://support.elsevier.com> for more information.

Printed in the Netherlands

The American Committee on Asian Economic Studies (ACAES)

An Inter-university (Nonprofit) Program Founded in 1982

Executive Board: Life Members

Romeo M. Bautista, *International Food Policy Research Institute*; Gregory C. Chow, *Princeton University*; M. Dutta, *Rutgers University*; H. Peter Gray, *Rensselaer Polytechnic Institute at Troy, NY*; Richard Hooley, *University of Pittsburgh*; F. Tomasson Januzi, *University of Texas at Austin*; Gary H. Jefferson, *Brandeis University*; Lawrence R. Klein, *University of Pennsylvania*; Richard F. Kosobud, *University of Illinois at Chicago*; Lawrence B. Krause, *University of California at San Diego*; Lawrence J. Lau, *Stanford University*; Chung H. Lee, *University of Hawaii at Manoa*; Woo Bong Lee, *Bloomsburg University of Pennsylvania*; John M. Letiche, *University of California at Berkeley*; Edward J. Lincoln, *The Brookings Institution*; Kanta Marwah, *Carleton University*; James I. Nakamura, *Columbia University*; Walter C. Neale, *University of Tennessee*; Gustav F. Papanek, *Boston University*; Hugh T. Patrick, *Columbia University*; Dwight H. Perkins, *Harvard University*; James Riedel, *Johns Hopkins University*; George Rosen, *University of Illinois at Chicago*; Vernon W. Ruttan, *University of Minnesota*; Kazuo Sato, *Rutgers University*; Ryuzo Sato, *New York University*; T.N. Srinivasan, *Yale University*; Joseph J. Stern, *Harvard University*; Paul P. Streeten, *Boston University*; Vincent Su, *The City University of New York*; Anthony M. Tang, *Vanderbilt University*; James T.H. Tsao, *U.S. International Trade Commission*; Yien-I Tu, *University of Arkansas at Fayetteville*; Calla Wiemer, *University of Hawaii at Manoa*; Jang H. Yoo, *Korea Institute for International Economic Policy*.

M. DUTTA, PRESIDENT AND CHIEF EXECUTIVE OFFICER, 1982–2002

Executive Board: Term Members (Term Ends Dec. 3, 2012)

Edna E. Ehrlich, *Erich International Consulting*, New York, edna_ehrlich88@netzero.net; David Jay Green, *Asian Development Bank*, Manila, dgreen@adb.org; Frank Hsiao, *University of Colorado at Boulder*, hsiao@spot.colorado.edu; Steven L. Husted, *University of Pittsburgh*, husted1+@pitt.edu; Saleem Khan, *Bloomsburg University*, skhan@planetx.bloomu.edu; Sung Y. Kwack, *Howard University*, skwack@howard.edu; Hiro Lee, *Nagoya University*, Nagoya, Japan, hlee@rieb.kobe-u.ac.jp; Gene Gruver, *University of Pittsburgh*, gruver+@pitt.edu; Balwant Singh, *Bucknell University*, singh@bucknell.edu; Sumner LaCroix, *University of Hawaii*, lacroix@hawaii.edu; Keun Lee, *Seoul National University*, klee1012@plaza.snu.ac.kr; Eric Ramstetter, *International Centre for the Study of East Asian Development*, ramst@icsead.or.jp.

Board of Trustees: (2001–2012)

Jere Behrman, *University of Pennsylvania*, jbehrman@ssc.upenn.edu
Members: Richard Kosobud, kosobud@uic.edu
Calla Weimer, cjweimer@hotmail.com

Richard Hooley, President and Chief Executive Officer,
University of Pittsburgh, rhooley@pitt.edu
Jere Berrman, Vice President, ...
James Riedel, *Johns Hopkins University*,
jriedel@mail.jhuwash.jhu.edu

Executive Officers: (2007–2012):

Director: Michael G. Plummer, *The Johns Hopkins University*, SAIS-Bologna mplummer@jhbc.it; Vice-Director: Chung Lee, *University of Hawaii*, lchung@hawaii.edu; Assoc. Directors: Frank Hsiao, *University of Colorado*, frank.hsiao@colorado.edu; Richard Hooley, *University of Pittsburgh*, rhooley@pitt.edu; Steven L. Husted, *University of Pittsburgh*, husted1+@pitt.edu; Gary H. Jefferson, *Brandeis University*, jefferson@binah.cc.brandeis.edu; Treasurer: Stephen Husted, *University of Pittsburgh*, husted1+@pitt.edu

Nomination and Election Committee (2001–2012)

James Tsao (Chair), *George Washington University*, tsao@erols.com; John Letiche, *University of California (Berkeley)*, jletiche@econ.berkeley.edu;

Banker:

Citizens Bank, Pittsburgh, PA

© 2014 Elsevier Inc.

This journal and the individual contributions contained in it are protected under copyright, and the following terms and conditions apply to their use in addition to the terms of any Creative Commons or other user license that has been applied by the publisher to an individual article:

Photocopying

Single photocopies of single articles may be made for personal use as allowed by national copyright laws. Permission is not required for photocopying of articles published under the CC BY license nor for photocopying for non-commercial purposes in accordance with any other user license applied by the publisher. Permission of the publisher and payment of a fee is required for all other photocopying, including multiple or systematic copying, copying for advertising or promotional purposes, resale, and all forms of document delivery. Special rates are available for educational institutions that wish to make photocopies for non-profit educational classroom use.

Derivative Works

Users may reproduce tables of contents or prepare lists of articles including abstracts for internal circulation within their institutions or companies. Other than for articles published under the CC BY license, permission of the publisher is required for resale or distribution outside the subscribing institution or company.

For any subscribed articles or articles published under a CC BY-NC-ND license, permission of the publisher is required for all other derivative works, including compilations and translations.

Storage or Usage

Except as outlined above or as set out in the relevant user license, no part of this publication may be reproduced, stored in a retrieval system or transmitted in any form or by any means, electronic, mechanical, photocopying, recording or otherwise, without prior written permission of the publisher.

Permissions

For information on how to seek permission visit www.elsevier.com/permissions or call: (+44) 1865 843830 (UK) / (+1) 215 239 3804 (USA).

Author rights

Author(s) may have additional rights in their articles as set out in their agreement with the publisher (more information at <http://www.elsevier.com/authorsrights>).

Notice

No responsibility is assumed by the Publisher for any injury and/or damage to persons or property as a matter of products liability, negligence or otherwise, or from any use or operation of any methods, products, instructions or ideas contained in the material herein. Because of rapid advances in the medical sciences, in particular, independent verification of diagnoses and drug dosages should be made.

Although all advertising material is expected to conform to ethical (medical) standards, inclusion in this publication does not constitute a guarantee or endorsement of the quality or value of such product or of the claims made of it by its manufacturer.



Which firms benefit from foreign direct investment? Empirical evidence from Indonesian manufacturing

Suyanto^a, Ruhul Salim^{b,*}, Harry Bloch^b

^a Faculty of Economics, University of Surabaya, East Java, Indonesia

^b Curtin Business School, Curtin University, Perth, WA 6845, Australia

ARTICLE INFO

Article history:

Received 22 November 2012
Received in revised form 8 May 2014
Accepted 11 May 2014
Available online 19 May 2014

JEL classification:

D24
D29
F23

Keywords:

Foreign direct investment
Spillover effects
Technical efficiency
Stochastic production frontier
Indonesia

ABSTRACT

Despite growing concern regarding the productivity benefits of foreign direct investment (FDI), very few studies have been conducted on the impact of FDI on firm-level technical efficiency. This study helps fill this gap by empirically examining the spillover effects of FDI on the technical efficiency of Indonesian manufacturing firms. A panel data stochastic production frontier (SPF) method is applied to 3318 firms surveyed over the period 1988–2000. The results reveal evidence of positive FDI spillovers on technical efficiency. Interesting differences emerge however when the samples are divided into two efficiency levels. High-efficiency domestic firms receive negative spillovers, in general, while low-efficiency firms gain positive spillovers. These findings justify the hypothesis of efficiency gaps, that the larger is the efficiency gap between domestic and foreign firms the easier the former extracts spillover benefits from the latter.

© 2014 Elsevier Inc. All rights reserved.

1. Introduction

The spillover effects of foreign direct investment (FDI) have been a major concern for researchers and policy makers during the last two decades. A number of studies have examined the spillover effects of FDI on domestic firm productivity (Chakraborty & Nunnenkamp, 2008; Haddad & Harrison, 1993; Hu & Jefferson, 2002; Javorcik, 2004; Liang, 2007; Negara & Firdausy, 2011; Takii, 2005, 2011). These studies provide some useful insights regarding the evidence of the spillover benefits and offer some recommendations to maximize the benefits. However, most existing studies exclude technical efficiency and focus mainly on technology, ignoring that the FDI presence in host countries is the impetus for efficiency improvement through competition and demonstration effects (Wang & Blomstrom, 1992). A study of FDI spillover effects on firm-level technical efficiency is important to provide evidence as to whether the large amount of FDI inflows generate positive externalities to domestic firms through efficiency improvement, thus indicating whether the spillover hypothesis is justified in the context of technical efficiency. Such a study can explore to what extent FDI can induce efficiency spillovers, and which firm types really benefit from the spillovers.

* Corresponding author at: School of Economics & Finance, Curtin Business School, Curtin University, WA 6845, Australia. Tel.: +61 892664577; fax: +61 92663026.

E-mail address: Ruhul.Salim@cbs.curtin.edu.au (R. Salim).

Among the developing economies, Indonesia is particularly successful in attracting FDI. Net FDI inflows to Indonesia have risen more than 30 times since 1986, reaching a record level of US\$8.3 billion in 2008 (the [Central Bank of Indonesia, 2011](#)). However, there is a dearth of research on efficiency spillovers in Indonesia. Most empirical studies examine spillover effects under a framework of the long-run equilibrium production function, which assumes that firms are producing at a full efficiency level. Under this framework, the FDI spillovers on technical efficiency are not captured.

Two previous studies by the authors focus on technical efficiency using a stochastic production frontier framework for individual Indonesian manufacturing industries. [Suyanto, Salim, and Bloch \(2009\)](#) examine the pharmaceutical and chemical industries, while [Suyanto et al. \(2012\)](#) examine the electronic and garment industries. However, there are no studies providing comprehensive results for the whole Indonesian manufacturing sector using a stochastic framework.

A study by [Temenggung \(2007\)](#) examines the whole Indonesian manufacturing sector. Our current research differs from Temenggung in three important points. Firstly, Temenggung applies the ordinary least squared (OLS) regression method for panel data, which doesn't distinguish between fixed effects (FE) and random effects (FE). Secondly, the classical production function, employed in [Temenggung \(2007\)](#), assumes that all firms are fully efficient, so that the spillover effects of FDI reflect technological progress. In contrast, the current paper employs the stochastic production frontier, which relaxes the assumption of full efficiency of firms, so that both technological progress and efficiency improvement are examined. Thirdly, we calculate the scores of technical efficiency of each firm and estimates spillover effects separately for high-efficiency and low-efficiency firms, providing a useful insight into the differences in the ability of high-efficiency and low-efficiency firms in absorbing spillover effects from FDI.

This study contributes to the existing literature in several ways. Firstly, it examines the spillover hypothesis by focusing on technical efficiency, an important aspect that is often neglected in the previous studies. The adoption of a stochastic production frontier allows the authors to investigate the effects of FDI spillovers on firm-level technical efficiency. Secondly, this study covers a long series of surveyed firms, which includes also the period of the Asian crisis onwards. Thirdly, this study evaluates horizontal, backward, and forward spillovers of FDI. Most importantly, by examining the whole manufacturing sector, it is possible to identify characteristics of industries that affect the size of the technology and efficiency spillovers to domestic firms from FDI. In particular, we find evidence that the size of the technology gap between foreign and domestic firms is critical, with larger efficiency gaps associated with greater efficiency spillovers from FDI.

We proceed by reviewing the concept of spillover effects in the next section. We then discuss methodology and data. Empirical results are presented in Section 4 and the conclusions are given in the final section.

2. FDI, spillover effects, and technical efficiency: theoretical concept and empirical evidence

2.1. FDI and spillover effects

Foreign direct investment is believed to provide host countries with direct and indirect benefits. The direct benefits take the forms of new investments that boost national income, increase tax revenues, and provide new employment; whereas the indirect benefits are in the forms of externalities that are generated through non-market mechanisms to recipient economies and domestic firms within the economies ([Hymer, 1960](#)). These indirect benefits are commonly known as FDI spillovers.

The literature identifies at least three types of FDI spillovers. These are productivity spillovers, market-access spillovers, and pecuniary spillovers. Productivity spillovers are defined as the externalities from FDI that lead to increases in the productivity of domestic firms ([Aitken & Harrison, 1999](#)). Market-access spillovers exist when the presence of FDI generates an opportunity for domestic firms to access international markets ([Blomstrom & Kokko, 1998](#)). Pecuniary spillovers happen if the existence of FDI affects the profit functions of domestic firms through a reduction in costs or an increase in revenues ([Gorg & Strobl, 2005](#)).

Of the three types of FDI spillovers, productivity spillovers have been a particular concern among policy makers and researchers in the last two decades. Various incentives have been provided by policy makers to attract FDI and substantial efforts have been devoted by researchers to evaluate the productivity advantage. However, the empirical evidence is mixed at best. Some studies find evidence of positive productivity spillovers ([Caves, 1974](#); [Javorcik, 2004](#); [Kugler, 2006](#); [Schiff & Wang, 2008](#); [Temenggung, 2007](#)), but others discover nonexistent or even negative spillovers ([Aitken & Harrison, 1999](#); [Blalock & Gertler, 2008](#); [Djankov & Hoekman, 2000](#)). Thus, the relationship between FDI spillovers and firm productivity remains a controversial issue.

2.2. Spillover effects and firm-specific characteristics

Some researchers argue that the mixed evidence intuitively implies that the spillover effects are not an automatic consequence of the foreign presence in an economy, rather they depend significantly on the characteristics of firms in the industries ([Gorg & Greenaway, 2004](#); [Lipsey & Sjöholm, 2005](#); [Smeets, 2008](#)). One important characteristic of firms is the technology gap between foreign and domestic firms. In a study on UK manufacturing firms, [Griffith, Redding, and Simpson \(2002\)](#) find that the wider the technology gap the larger the FDI spillover effects that are obtained by domestic firms. This finding indicates a benefit of being less advanced in terms of technology, which supports the theoretical argument in [Findlay \(1978\)](#). A similar result is discovered also by [Castellani and Zanfei \(2003\)](#) for France and Spain, and by [Peri and Urban \(2006\)](#) for Italy and Germany.

Although there is an advantage in being less advanced, the technology gap should not be too wide (Wang & Blomstrom, 1992). A minimum level of technology is required for domestic firms to absorb the new technology from foreign firms. When the gap is too wide, the limited kind absorptive capacity of domestic firms may not permit assimilation the new technology (Glass & Saggi, 1998).

2.3. Technical efficiency gains from FDI spillovers

Earlier studies on FDI productivity spillovers focus on technology advantages (Gorg & Greenaway, 2004). The knowledge from foreign firms is regarded synonymously with technological knowledge, as this is consistent with the use of a conventional production function. Managerial and organizational knowledge that may lead to efficiency spillovers are not portrayed since firms are assumed to be producing at the long-run equilibrium with a full efficiency capacity. Thus, the productivity spillovers in these early studies are identically measured as technology spillovers.

More recent studies focus on both efficiency and technology advantages. In these studies, knowledge is defined broadly as product, process, managerial, and organizational knowledge. Hence, productivity spillovers lead to both technology and efficiency advantages. Unfortunately, studies that investigate efficiency advantages are not plentiful. In a study on Greek manufacturing firms, Dimelis and Lauri (2002) examine the effect of foreign equity shares on efficiency and find a positive relationship between these two variables. Also, Ghali and Rezgui (2008) analyze the Tunisian manufacturing sector and find that higher foreign share increase firm efficiency. Addressing the same issue but employing a different estimation method, our study investigates the efficiency spillovers in Indonesian manufacturing firms. We extend the studies by Dimelis and Lauri (2002) and Ghali and Rezgui (2008) by focusing on vertical spillovers as well as horizontal spillovers.

3. Methodology, data set, and variables

3.1. Methodology

There are two commonly used methods in measuring efficiencies and productivity at the firm level, namely data envelopment analysis (DEA) and stochastic frontier analysis (SFA).¹ Each of the two methods has its advantages and disadvantages, as explained below. The choice between these methods thus depends on the objective of the research, the type of firms in the chosen industry, and the nature of the data (Coelli, Rao, O'Donnell, & Battese, 2005; Olesen, Peterson, & Lovell, 1996).

DEA is a linear programming method that observes production possibilities using the technique of envelopment and measures efficiency as the distance to the frontier (Banker, Charnes, & Cooper, 1984; Charnes, Cooper, & Rhodes, 1978). This method has the primary advantage of being of a non-parametric nature and has the ability to handle multiple outputs and multiple inputs.² However, it has the disadvantage of producing biased estimates in the presence of measurement error and other statistical noise, as this method does not separate the stochastic random noise from the inefficiency effects (Schmidt, 1985). Hence, the estimation results under this method tend to be very sensitive to small changes in the data.

Alternatively, the stochastic frontier method is a regression-based method that assumes two separate unobserved error terms, one represents efficiency and the other represents statistical noise (Aigner, Lovell, & Schmidt, 1977; Meeusen & van den Broeck, 1977). It has a chief advantage in the ability to measure efficiency in the presence of statistical noise. However, this method is parametric and requires a specific functional form and distributional assumptions for the error terms (Coelli et al., 2005).

In this study the stochastic frontier method is applied to analyze the spillover effects from FDI. The one-stage stochastic production frontier (SPF) is used to estimate a production frontier and a technical inefficiency function simultaneously. As pointed out by Kumbhakar, Ghosh, and McGuckin (1991) and Wang and Schmidt (2002), the one-stage approach is preferable than the two-stage approach, as the latter exhibits at least two limitations in estimation that can lead to potentially severe bias. The first limitation is that technical efficiency might be correlated with the production inputs, which may cause inconsistent estimates of the production frontier. The second limitation is the OLS method in the second stage is inappropriate since technical efficiency distribution is assumed to be one-sided. Considering the advantages, the current study adopts the one-stage approach, following Battese and Coelli (1995).

The Battese–Coelli production frontier can be expressed as follows:

$$y_{it} = f(x_{it}, t; \beta) \exp(v_{it} - u_{it}) \quad (1)$$

and the inefficiency function may be written as:

$$u_{it} = z_{it} \delta + w_{it} \quad (2)$$

¹ Comprehensive reviews of the two methods are provided by Forsund et al. (1980), Bauer (1990), Bjurek et al. (1990), Bravo-Ureta and Pinheiro (1993), Greene (1993), Lovell (1993), and Coelli (1995).

² The non-parametric nature of DEA allows for measuring efficiency without imposing a specific functional form and a distributional assumption on data.

where y_{it} denotes the production of the i th firm ($i = 1, 2, \dots, N$) in the t th time period ($t = 1, 2, \dots, T$), x_{it} denotes a $(1 \times k)$ vector of explanatory variables, β represents the $(k \times 1)$ vector of parameters to be estimated, exp denotes exponential, v_{it} is the time specific and stochastic error, with $iid N(0, \sigma^2_v)$, and u_{it} represents technical inefficiency, which is assumed as a function of a $(1 \times j)$ vector of observable non-stochastic explanatory variables, z_{it} , and a $(j \times 1)$ vector of unknown parameters to be estimated, δ , and w_{it} is an unobservable random variable.

The parameters of Eqs. (1) and (2) are estimated using the maximum likelihood estimator (MLE) by following the three steps as explained in Coelli (1996). With simultaneous equation estimation, the MLE estimates are unbiased and efficient. The variance parameters of the Battese–Coelli's model are defined as $\sigma^2_s \equiv \sigma^2_v + \sigma^2_u$ and $\gamma \equiv \sigma^2_u / \sigma^2_s$.³

γ is an important parameter to decide whether there is technical inefficiency or not in the model. If the estimated value of γ is not statistically significant, there is no technical inefficiency and the results obtained from estimating Eq. (1) by ordinary least squares (OLS) would be efficient. In contrast, if the estimated value of γ is statistically significant, then there is technical inefficiency and Eqs. (1) and (2) should be estimated simultaneously.

The technical efficiency of the i th firm calculated from the Eqs. (1) and (2) is the ratio of observed output of the firm to its potential maximum output, which can be written as:

$$TE_{it} = \frac{y_{it}}{y_{it}^p} = \exp(-u_{it}) \quad (3)$$

Following Battese and Coelli (1988), the best estimator of the $\exp(-u_{it})$ is its conditional expectation, $E[\exp(-u_{it})]$, so technical efficiency can be written as:

$$TE_{it} = E[\exp(-u_{it})] \quad (4)$$

If it is assumed that the production frontier takes the form of a log-linear production function and there are four input variables (labour, capital, material, and energy) in the production process, the empirical model can be expressed in natural logarithms of variables as:

$$\ln Y_{it} = \beta_0 + \beta_1 \ln L_{it} + \beta_2 \ln K_{it} + \beta_3 \ln M_{it} + \beta_4 \ln E_{it} + \beta_5 T + \beta_6 \ln FDI_Sector + v_{it} - u_{it} \quad (5)$$

where Y is output, L is labour, K is capital, M is material, E is energy, T is a time-trend variable that increases by one for each year, FDI_Sector is a measure of FDI horizontal spillovers as explained in the next section and the other variables are as previously defined.

The inefficiency effect as a function of a set of FDI variables, a year dummy, an industry dummy, and a firm dummy can be written as:

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Sector_{jt} + \delta_3 FDI_Firm_{it} \times FDI_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (6)$$

where FDI_Firm is a dummy variable for foreign direct investment that takes a value of zero if a firm has no foreign ownership share and takes a value of one if a foreign firm has a positive share, FDI_Sector is as defined above, $Year$ is a year dummy variable, $Industry$ is an industry dummy and $Firm$ is a firm dummy. The interaction term of $FDI_Firm \times FDI_Sector$ is included in the inefficiency equation to estimate whether foreign and domestic firms benefit equally from the presence of a new foreign firm. A positive (negative) coefficient on the interaction term indicates less (more) efficiency gain for foreign firms than for domestic firms.

Eq. (6) is used to estimate the intra-industry spillovers, which capture the effects of foreign presence on the technical efficiency of firms in the same industry. The inter-industry spillovers are commonly estimated by replacing the horizontal-spillover variable (FDI_Sector) with vertical-spillover variables. The inefficiency function for the inter-industry spillovers can be expressed as:

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Downstream_Sector_{jt} + \delta_3 FDI_Firm_{it} * FDI_Downstream_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (7)$$

or

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Upstream_Sector_{jt} + \delta_3 FDI_Firm_{it} * FDI_Upstream_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (8)$$

where $FDI_Downstream_Sector$ is a proxy for spillover effects from foreign firms to foreign and domestic suppliers and $FDI_Upstream_Sector$ is a proxy for spillover effects from foreign firms to foreign and domestic buyers.

³ The complete derivation the log-likelihood function of the Battese–Coelli model and its related variance parameters are discussed in Battese and Coelli (1993).

3.2. Data and data set construction

The primary data for our study are the annual surveys of medium and large manufacturing establishments (*Survey Tahunan Statistik Industri* or *SI*) conducted by the Indonesian Central Board of Statistics (*Badan Pusat Statistik* or *BPS*). These annual surveys cover a wide range of information from each surveyed establishment. The basic information includes year of starting production, industrial classification, location, and the specific identification code. There is also information regarding ownership, which includes foreign and domestic ownership, and information related to production, such as gross output, number of workers in production and non-production, value of fixed capital, material usage, and energy consumption.

The annual surveys have been conducted since 1975 and the most recent available data relates to the year 2007. However, this study uses the data from 1988 to 2000. The year 1988 is chosen as a starting year since it is the first year that the replacement value of fixed assets, which is used as a measure for capital, is available. The year 2000 is selected as the last year because the BPS changed the specific identification code in 2001 to the new identification code (KIPN) without providing a concordance table to the previous used identification code (PSID). Efforts to match the observations in the years 2001–2005 to the years 1988–2000 using output values and labour do not yield consistent results. Therefore, the longest possible period for this study is 1988–2000.

In constructing a consistent data set, several adjustments are conducted. These include adjustment for industrial code, adjustment for variable definitions, cleaning for noise and typological errors, back casting missing values of capital, matching firms for a balanced panel, and choosing industries with foreign firms. The balanced panel data are preferable in this study due to two advantages: (1) it enables tracing the technical efficiency scores of each observed firm during the period of study; (2) it removes the influence of a firm that appears only in one or two years, while the period of estimation is for 13 years. The details of adjustments are presented in Appendix 1. After the adjustments, the final balanced panel of data consists of 3318 establishments with 43,225 observations.

To show the influence of the construction of the balanced panel dataset, the descriptive statistics of the related variables are calculated for the balanced panel data and for the original data before the adjustment process. The original data consist of establishments that do not report complete information on output, labour, capital, material, or energy. Therefore these establishments are not included in the calculation of the descriptive statistics for original data. Following Takii (2005), (1) 0.5 percent observations with the lowest values of output and 1.5 percent observations with the highest values of output are removed. After these deletions, the descriptive statistics for the original total data, as presented in Table 1, consists of 24,188 establishments for an unbalanced panel of 238,628 observations.

Table 1 shows that the minimum values of variables $\ln Y$, $\ln L$, $\ln K$, $\ln M$, $\ln E$ for the original data are lower if compared to the minimum values of those variables from the balanced panel. This makes sense as the balanced panel data removes some observations during the adjustment process. The maximum values of those variables are higher in the original data compared to those in balanced panel data. The mean values of these five variables are higher in the balanced panel data compared to those in original data, while the standard deviations of these five variables are lower in balanced panel when compared to those in original data.

For FDI_Firm , the minimum value is zero and the maximum value is one both for original data and the balanced panel data, because this variable is a dummy variable. Further, the minimum value and the maximum value of variables FDI_Sector ,

Table 1
Descriptive statistics for the original data and the balanced panel data.

	Original data ^a				Balanced panel data			
	Min	Max	Mean	SD	Min	Max	Mean	SD
<i>Production Frontier</i>								
$\ln Y$	6.461	20.980	12.514	2.256	6.591	20.761	13.964	2.006
$\ln L$	2.398	10.649	4.079	1.327	2.639	10.292	4.702	1.088
$\ln K$	4.105	23.398	12.308	2.268	4.220	23.106	13.152	2.245
$\ln M$	3.871	20.033	11.765	2.418	4.239	19.454	12.164	2.221
$\ln E$	1.791	16.583	9.377	2.221	1.882	15.836	9.587	2.077
FDI_Sector	0	1.492	0.208	0.218	0	1.492	0.234	0.209
<i>Inefficiency Function</i>								
FDI_Firm	0	1	0.064	0.273	0	1	0.072	0.258
FDI_Sector	0	1.492	0.208	0.218	0	1.492	0.234	0.209
$FDI_Downstream_Sector$	0.002	5.443	0.176	0.212	0.002	5.443	0.176	0.204
$FDI_Upstream_Sector$	0	0.921	0.160	0.181	0	0.921	0.160	0.174
Number of Establishments	24,188	24,188	24,188	24,188	3318	3318	3318	3318
Number of Observation	231,064	231,064	231,064	231,064	43,225	43,225	43,225	43,225

Source: Authors' calculations from the annual surveys of the Indonesian Central Board of Statistics (*Badan Pusat Statistik* or *BPS*).

Y = output, L = labour, K = capital, M = material and E = energy.

^a The original data in this table exclude: (1) the establishments that do not report information on output, labour, capital, material, or energy; (2) 1.5 percent observations with the lowest values of output and 1.5 percent observations the highest values of output.

FDI_Backward, and *FDI_Forward* are the same for original data and for the balanced panel, as the calculation of these inter-industry variables is based on all firms in the original data as in [Blalock and Gertler \(2008\)](#). The mean values of these three spillover variables are higher in the balanced panel compared to those in the original data, whereas the standard deviations are lower in balanced panel. From the descriptive statistics in [Table 1](#), the authors conclude that there is no substantial bias in the adjustment process since there is no substantial difference in the maximum value, minimum value, mean value, and standard deviation.

3.3. Measurement of variables

There are two sets of variables included in this study: production variables and inefficiency variables. The production variables consist of output, labour, capital, material, energy, time trend and *FDI_Sector*, while the inefficiency variables include FDI variables (*FDI_Firm*, *FDI_Sector*, *FDI_Upstream_Sector*, and *FDI_Downstream_Sector*), a year dummy, an industry dummy, and a firm dummy. The precise definition of each variable is given in Appendix 2.

In this study, gross output is used as the measure for output (y). It refers to the total value of output produced by a firm. The number of employees directly and indirectly engaged in production is used for the measure of labour (L). As a measure of capital (K), this study uses the replacement value of capital, while material (M) is measured using the total value of raw and intermediate materials and energy (E) is measured as the sum of electricity and fuel expenses.

FDI_Firm is measured by a dummy of foreign direct investment, which takes a value of one if a firm has a positive foreign ownership and takes a value of zero if otherwise. As a measure for the FDI horizontal spillovers, this study uses the share of foreign firm output to the total output at the five-digit ISIC sectoral level, which is expressed as in [Aitken and Harrison \(1999\)](#):

$$FDI_Sector_{jt} = \frac{\sum_{i \in j} FDI_Firm_{it} \times y_{it}}{\sum_{i \in j} y_{it}} \quad (9)$$

Eq. (9) captures the effect of FDI at the sectoral level on productivity at the firm level. It shows the spillover effects of foreign presence on domestic firms in the same five-digit ISIC industry.

Two alternative measures of FDI spillovers in this study are measures of inter-industry spillovers. The presence of foreign firms in certain five-digit ISIC industries may create productivity externalities for firms in upstream and downstream industries. This study measures the inter-industry spillovers by using variables that reflect the extent of backward and forward linkages between industries. Following [Javorcik \(2004\)](#), the measure for FDI spillovers from foreign firms in industries $k(k \neq j)$ that are being supplied by domestic firms in industries j is:

$$FDI_Downstream_Sector_{jt} = \sum_{k \neq j} \alpha_{jk} \times FDI_Sector_{kt} \quad (10)$$

where α_{jk} is the proportion of sector j 's output supplied to sector k , which is taken from the input–output (IO) matrix of four-digit industries.⁴ Similarly, the measure for FDI spillovers from foreign firms in industries m whose products are bought by domestic firms in industries n is:

$$FDI_Upstream_Sector_{mt} = \sum_{n \neq m} \gamma_{mn} \times FDI_Sector_{nt} \quad (11)$$

where γ_{mn} is the proportion of inputs purchased by industry n from industry m in total input sourced by industry n , which is taken from the input–output (IO) matrix of four-digit industries.

A time-trend variable is incorporated in the production function to measure technical change. The time-trend variable takes a value of one for the year 1988, a value of two for the year 1989, and so on. An industry dummy captures effects specific to a particular industry and has a value of one for an industry for an observation of that industry and a value of zero otherwise. A similar procedure is also applied to the firm dummy and year dummy variables.

4. Empirical results

We estimate a stochastic frontier estimation and first test for constant returns to scale to check whether the Cobb–Douglas production frontier is best suited to the data. Following the procedure of joint restriction test in [Baltagi \(2011, p. 80\)](#), the test of constant returns to scale is conducted under the null hypothesis that the sum of the estimated parameters (β_i) in

⁴ During the selected period in this study, there are four available IO matrixes, which were published in 1990, 1993, 1995, and 2000. This study uses these four input-output matrixes for calculating the backward coefficient α_{jk} . The following is the procedure for obtaining values of α_{jk} . Values of α_{jk} before and including 1990 are taken from the 1990 IO matrix. Values of α_{jk} for 1991 and 1992 are linearly interpolated from the 1990 and 1993 IO matrixes. Values of α_{jk} for 1993 are taken from the 1993 IO matrix. Values of α_{jk} for 1994 are calculated from the linear interpolation of the 1993 and 1995 IO matrixes. Values of α_{jk} for 1995 are taken from the 1995 IO matrix. Values of α_{jk} from 1996 to 1999 are linearly interpolated from the 1995 and the 2000 IO matrixes. Finally, values of α_{jk} for 2000 are taken from the 2000 IO matrix.

production frontier in Eq. (5) is equal to one. The regression sum of squares for unrestricted model (RSS_U) is 39,631.63, whereas the regression sum of squared for restricted model (RSS_R) is 25,549.50. The F -statistics is 392.52, suggesting that the null hypothesis is rejected. This result confirms that the Cobb–Douglas production frontier is not the best suited model for the stochastic frontier estimation. Rather, as the sum of the coefficients of the input variables is greater than one, the unrestricted model with variable returns to scale is appropriate and is used below

4.1. Intra-industry spillovers

We begin with estimation of intra-industry spillovers. Using Eqs. (5) and (6), the production frontier and the inefficiency function are estimated simultaneously for observing the effects of foreign investment on the production frontier and technical efficiency of firms. For the inefficiency function, the technical efficiency variable (u_{it}) is specified as a function of a foreign share dummy (FDI_Firm), the share of foreign firms' outputs over total outputs in the four-digit industry (FDI_Sector), and an interacting term between FDI_Firm and FDI_Sector . When foreign investment increases the firm's technical efficiency, the coefficient of FDI_Firm is negative.⁵ When technology spills over from firms with foreign direct investment to purely domestic firms in the same industry, the coefficient of FDI_Sector is negative. As for the interaction term, the sign of the coefficient shows whether or not foreign direct investment affects the firm's ability to benefit from spillovers originating from other foreign-owned firms in the same industry.

We estimate four alternative models in order to test the robustness of the estimated parameters. In the first model, a year dummy and an industry dummy are included in the inefficiency equation. The estimated parameters are presented in the Model (1) column of Table 2. The results from the production frontier show that the four input variables contribute positively and significantly to output, suggesting a positive elasticity of each input on output. There is also a positive and statistically significant coefficient of the time-trend variable indicating that technical change contributes positively to output. The positive and statistically significant coefficient of FDI_Sector suggests horizontal spillovers from intra-industry foreign direct investment increase the production frontier for all firms.

From the estimates of the inefficiency function, which is the main focus of this study, the coefficient of FDI_Firm is negative and highly significant, indicating that foreign direct investment decreases the firm's technical inefficiency. This suggests that firms with foreign ownership are, on average, more efficient than purely domestic firms. This finding confirms the argument in Caves (1971) and Dunning (1988) that foreign firms are more likely to operate on the production frontier. Furthermore, the negative and statistically significant estimate of FDI_Sector suggests that knowledge spills over from foreign-owned firms increases the technical efficiency of all firms in the industry. This result is in line with the argument in Wang and Blomstrom (1992) and findings in Ghali and Rezgui (2008). This result is also consistent with findings in Takii (2005), Temenggung (2007) and Blalock and Gertler (2008), which use different methods of analysis.

The positive significant estimate of interacting term means that, although the foreign-owned firms also benefit from the presence of other foreign investment in the industry, the benefit is smaller than for domestic firms. Given that the estimated coefficient of FDI_Firm and the estimated coefficient of FDI_Sector are negative and statistically significant, the positive coefficient of the interaction term means that $u_{it}/FDI_Firm = -0.5763 + 0.0330 \times FDI_Sector$ and that $u_{it}/FDI_Sector = -0.2224 + 0.0330 \times FDI_Firm$. As both FDI_Firm and FDI_Sector are each always less than or equal to one by construction, the net effect of FDI_Sector is negative for all foreign firms as well as domestic firms. However, the magnitude of the improvement in efficiency from having foreign firms in the industry is always greater for domestic firms than for foreign firms.

In addition, we conduct joint significance test (F -test) on the magnitude of spillovers for foreign establishments in order to check significance of the direct effect and the interacting effect of spillovers on foreign firms.⁶ The value of F -statistic is calculated from the log-likelihood value of the unrestricted model and the log likelihood value of the restricted model (when both the coefficient of FDI_Sector and the coefficient of interacting variable $FDI_Firm \times FDI_Sector$ equal to zero). The value of log likelihood for the unrestricted model is 7704.48, whereas the value of log likelihood for the restricted model is 7643.00, So that, the F -statistic is 13.22, which suggests that the unrestricted model (by including variables FDI_Sector and interacting variable $FDI_Firm \times FDI_Sector$) is the correct model and the two variables are jointly significant affecting spillovers on foreign establishments at 1% level.

The estimated coefficient of year dummy is not statistically significant, suggesting that on average there is no significant difference in technical inefficiency scores of firms across the sample years. The statistically significant estimated coefficient of industry dummy suggests that there is a significant difference in inefficiency scores across five-digit industries.

The highly significant estimate of gamma implicates that estimation of stochastic frontier should include an inefficiency effect. This finding provides the justification for the simultaneous estimation of stochastic production frontier and inefficiency equation. In other words, the model is appropriately representing the observed firms.

In the second model, industry dummies are replaced by firm dummies, in order to control for firm heterogeneity across the sample. The results are given in the Model (2) column of Table 2. The sign and significance of estimates are similar to

⁵ The dependent variable for the inefficiency function is technical inefficiency. The negative coefficient of FDI_Firm indicates that foreign investment decreases inefficiency, which implies an increase in the firm's efficiency.

⁶ We are grateful to one of the reviewers for suggesting this point.

Table 2
Estimating intra-industry spillovers.

Variables	Model (1)	Model (2)	Model (3)	Model (4)
<i>Production frontier</i>				
<i>lnL</i>	0.2227*** (0.0033)	0.2256*** (0.0031)	0.2197*** (0.0030)	0.2167*** (0.0031)
<i>lnK</i>	0.1018*** (0.0019)	0.1043*** (0.0017)	0.1023*** (0.0018)	0.1097*** (0.0012)
<i>lnM</i>	0.6263*** (0.0018)	0.6218*** (0.0018)	0.6223*** (0.0017)	0.6191*** (0.0022)
<i>lnE</i>	0.1128*** (0.0017)	0.1160*** (0.0017)	0.1165*** (0.0017)	0.1176*** (0.0016)
<i>T</i>	0.0007* (0.0005)	0.0039* (0.0006)	0.0066*** (0.0028)	0.0012*** (0.0003)
<i>FDI_Sector</i>	0.1224*** (0.0055)	0.2044*** (0.0065)	0.2687*** (0.0096)	0.1577*** (0.0065)
<i>Inefficiency function</i>				
<i>FDI_Firm</i>	-0.5763*** (0.0264)	-0.1550*** (0.0018)	-0.1960*** (0.0104)	-0.2362*** (0.0092)
<i>FDI_Sector</i>	-0.2224*** (0.0896)	-0.2000*** (0.0149)	-0.1780*** (0.0027)	-0.1819*** (0.0034)
<i>FDI_Firm × FDI_Sector</i>	0.0330*** (0.0028)	0.0460*** (0.0036)	0.1035*** (0.0184)	0.0673*** (0.0086)
Year Dummy	-0.0002 (0.0031)	-0.0010 (0.0009)	-0.0010 (0.0019)	-
Industry Dummy	-0.0039* (0.0008)	-	-	-
Firm Dummy	-	-0.0001** (0.0000) ^a	-	-
Sigma-squared	0.0416*** (0.0010)	0.0416*** (0.0005)	0.0413*** (0.0003)	0.0418*** (0.0003)
Gamma	0.0380*** (0.0038)	0.0224*** (0.0083)	0.0086*** (0.0002)	0.0151*** (0.0020)
Log-likelihood	7704.484	7759.086	7618.974	7572.755
Number of Observations	43,225	43,225	43,225	43,225

Source: Authors' calculations.

Notes: *Y* = output, *L* = labour, *K* = capital, *M* = material, *E* = energy, *T* = time trend. Standard errors are in parentheses.

^a The estimated standard error is 0.000009.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

those in the first model. The notable difference is only in the magnitude of the estimates. Focusing on the FDI variables, the magnitudes of coefficients are smaller in this second model compared to those in the first model. In other words, the inclusion of firm dummy and the exclusion of industry dummy in the second estimation (Model 2) results in a smaller effect of FDI spillovers on technical inefficiency. This is not surprising. Firm-specific effects are largely captured by the firm dummy, which removes a potential source of bias in the estimates of other coefficients. Notably, the results regarding the direction of spillover effects are the same as in the first model, as the coefficient of *FDI_Sector* is negative and statistically significant in both models.

For the third model, only a time dummy is included as a controlling variable in the estimation. The resulting estimates, which are presented in the Model (3) column, are very comparable with the results in Model (1) and Model (2). Similar findings are also observed in Model (4), when the time dummy, industry dummy and firm dummy are all excluded from estimation. The results from these four models confirm the robustness of the estimates of the positive spillovers from FDI on the technical efficiency of domestic firms.

4.2. Inter-industry spillovers

Besides the effects on domestic firms in the same industry, FDI can also generate spillovers on domestic firms in other industries. We estimate six models of the inter-industry spillovers, and the results of each model are presented in Table 3. The first three models are estimated on the full sample and the last three models are estimated on the sub-sample of only domestic firms. In the three full-sample models, the first model is to capture the simultaneous effect of the three spillover variables on technical inefficiency. The second and the third model focus on the individual effect of each of the vertical FDI spillovers (i.e. the downstream spillover and the upstream spillover). The same structure is also applied to the sub-sample of only domestic firms, with Model 4 captures the simultaneous effect of the three spillover variables, Model 5 captures the downstream effect only, and model 6 captures only the upstream effect.

In the first model (the first results column of Table 3), the three proxies of spillover variables are included in the estimations. The results show that the horizontal spillover variable (*FDI_Sector*) has a negative and statistically significant coefficient, suggesting that an increase in the share of foreign firm output decreases technical inefficiency across firms in the industry. Similarly, the spillovers from FDI in downstream industries also decrease inefficiency of suppliers, as demonstrated by the negative and highly significant coefficient of the backward spillover variable (*FDI_Downstream_Sector*). In addition, the coefficient of the forward spillover variable (*FDI_Upstream_Sector*) is negative and highly significant, indicating a negative relationship between FDI in supplier industries and the industry's own technical inefficiency. Although we employ a different methodology and use a different data set, the findings are similar to those in Liang (2007).

In the second and the third models (the second and the third columns of Table 3), the impacts of backward spillover variable and the forward spillover variable are estimated separately. In each model, the magnitude of the coefficient of the included spillovers variable is larger than in Model 1, but neither the sign nor the statistical significance of the coefficient changes. Clearly, there is multi-colinearity among the spillovers variables that makes the identification of separate effects

Table 3
Estimating inter-industry spillovers.

Variables	Full sample (1)	Full sample (2)	Full sample (3)	Domestic sample (4)	Domestic sample (5)	Domestic sample (6)
<i>Production frontier</i>						
<i>lnL</i>	0.2264 ^{***} (0.0030)	0.2209 ^{***} (0.0030)	0.2197 ^{***} (0.0029)	0.2258 ^{***} (0.0012)	0.2238 ^{***} (0.0033)	0.2256 ^{***} (0.0033)
<i>lnK</i>	0.1007 ^{***} (0.0018)	0.1023 ^{***} (0.0018)	0.1019 ^{***} (0.0018)	0.0986 ^{***} (0.0018)	0.0999 ^{***} (0.0022)	0.0981 ^{***} (0.0019)
<i>lnM</i>	0.6255 ^{***} (0.0018)	0.6271 ^{***} (0.0018)	0.6268 ^{***} (0.0017)	0.6225 ^{***} (0.0014)	0.6236 ^{***} (0.0020)	0.6229 ^{***} (0.0017)
<i>lnE</i>	0.1117 ^{***} (0.0017)	0.1144 ^{***} (0.00170)	0.1159 ^{***} (0.0016)	0.1217 ^{***} (0.0014)	0.1226 ^{***} (0.0018)	0.1227 ^{***} (0.0018)
<i>T</i>	0.0002 ^{**} (0.0000) ^a	0.0028 [*] (0.0013)	0.0004 ^{***} (0.0001)	0.0009 [*] (0.0006)	0.0021 ^{***} (0.0001)	0.0010 ^{***} (0.0002)
<i>FDI_Sector</i>	0.0375 ^{***} (0.0013)	0.0308 ^{***} (0.0038)	0.0217 ^{***} (0.0007)	0.0056 ^{***} (0.0007)	0.0572 ^{***} (0.0035)	0.0323 ^{***} 0.0064
<i>Inefficiency function</i>						
<i>FDI_Firm</i>	−0.2945 ^{***} (0.0137)	−0.3920 ^{***} (0.0393)	−0.1257 ^{***} (0.0130)	–	–	–
<i>FDI_Sector</i>	−0.1901 ^{***} (0.0061)	–	–	−0.2766 ^{***} (0.0275)	–	–
<i>FDI_Downstream_Sector</i>	−0.0216 ^{***} (0.0021)	−0.0715 ^{***} (0.0043)	–	−0.0279 ^{***} (0.0047)	−0.0548 ^{***} (0.0027)	–
<i>FDI_Upstream_Sector</i>	−0.0462 ^{***} (0.0060)	–	−0.1842 ^{***} (0.0097)	−0.0682 ^{***} (0.0175)	–	−0.3067 ^{***} (0.0214)
<i>Year Dummy</i>	−0.0018 [*] (0.0006)	−0.0050 [*] (0.0017)	−0.0017 ^{**} (0.0003)	0.0011 ^{***} (0.0002)	0.0046 ^{**} (0.0005)	0.0002 ^{***} (0.0010)
<i>Firm Dummy</i>	−0.0000 ^{b***} (0.0000) ^c	−0.0000 ^{d***} (0.0000) ^e	−0.0000 ^{f***} (0.0000) ^g	−0.0001 ^{***} (0.0000) ^h	−0.0001 ^{***} (0.0000) ⁱ	−0.0001 ^{***} (0.0000) ^j
<i>Sigma-squared</i>	0.0401 ^{***} (0.0003)	0.0416 ^{***} (0.0003)	0.0405 ^{***} (0.0003)	0.0411 ^{***} (0.0007)	0.0418 ^{***} (0.0001)	0.0405 ^{***} (0.0004)
<i>Gamma</i>	0.0194 ^{***} (0.0013)	0.0417 ^{***} (0.0040)	0.0124 ^{***} (0.0008)	0.0612 ^{***} (0.0111)	0.0709 ^{***} (0.0019)	0.0561 ^{***} (0.0045)
<i>Log-likelihood</i>	7849.487	7668.081	7750.109	8118.497	8001.479	8040.274
<i>Number of Observations</i>	43,225	43,225	43,225	40,042	40,042	40,042

Source: Authors' calculations.

Notes: *Y* = output, *L* = labour, *K* = capital, *M* = material, *E* = energy, *T* = Time trend. Actual estimates are ^a 0.00004, ^b 0.000034, ^c 0.0000017, ^d 0.000034, ^e 0.0000019, ^f 0.000034, ^g 0.0000014, ^h 0.0000024, ⁱ 0.000012, ^j 0.0000035. Standard errors are in parentheses.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

Table 4
Estimating intra-industry spillovers in high-efficiency and low-efficiency firms.

Variables	Full sample		Domestic sample	
	High-efficiency firms (1)	Low-efficiency firms (2)	High-efficiency firms (3)	Low-efficiency firms (4)
<i>Production frontier</i>				
<i>lnL</i>	0.2049*** (0.0047)	0.2258*** (0.0040)	0.2372*** (0.0018)	0.2012*** (0.0038)
<i>lnK</i>	0.1080*** (0.0032)	0.0985*** (0.0024)	0.1025*** (0.0024)	0.0911*** (0.0021)
<i>lnM</i>	0.6038*** (0.0023)	0.6634*** (0.0027)	0.5883*** (0.0036)	0.6900*** (0.0026)
<i>lnE</i>	0.1316*** (0.0027)	0.0835*** (0.0023)	0.1429*** (0.0013)	0.0791*** (0.0018)
<i>T</i>	0.0021** (0.0009)	0.0001* (0.0000) ^b	0.0022*** (0.0004)	0.0064*** (0.0003)
<i>FDI_Sector</i>	0.0940*** (0.0058)	0.0492** (0.0141)	0.0849*** (0.0032)	0.0727** (0.0133)
<i>Inefficiency function</i>				
<i>FDI_Firm</i>	-0.0617*** (0.0088)	-0.0096* (0.0063)	-	-
<i>FDI_Sector</i>	0.0742*** (0.0062)	-0.0556*** (0.0035)	0.0657*** (0.0038)	-0.0660*** (0.0115)
Year Dummy	0.0020 [†] (0.0014)	-0.0027*** (0.0007)	0.0029*** (0.0004)	0.0015*** (0.0001)
Firm Dummy	0.0001*** (0.0000) ^a	0.0001*** (0.0000) ^c	0.0001*** (0.0000) ^d	0.0000 ^e (0.0000) ^f
Sigma-squared	0.0425*** (0.0004)	0.0382*** (0.0004)	0.0414*** (0.0005)	0.0341*** (0.0006)
Gamma	0.0369*** (0.0043)	0.0151*** (0.0023)	0.0540*** (0.0036)	0.0746*** (0.0019)
Log-likelihood	3493.823	4697.164	3597.36	5417.533
Number of Observations	21,612	21,613	20,021	20,021

Source: Authors' calculations.

Notes: *Y* = output, *L* = labour, *K* = capital, *M* = material, *E* = energy and *T* = time trend Actual estimates are: ^a 0.0000042, ^b 0.000037 ^c 0.000005 ^d 0.0000076, ^e 0.000018, ^f 0.0000066. Standard errors are in parentheses.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

difficult. The coefficient of the *FDI_Downstream_Sector* being negative and statistically significant at the 1% level in both Model 1 and Model 2, indicates a robust finding that the foreign entry in a three-digit industry decreases the technical inefficiency of domestic suppliers (i.e. positive backward spillovers). Similarly, the negative and statistically significant coefficient of the *FDI_Upstream_Sector* in both Model 1 and Model 3 indicates a robust finding that the presence of foreign firms in a three-digit industry decreases the inefficiency of domestic buyers (i.e. positive forward spillovers).

To isolate the spillover effects on only domestic firms, we estimate the Models 1 through 3 for the sub-sample of only domestic firms. The estimation results are presented in the fourth through sixth result columns in Table 3. The results are similar to those for the full sample of firms in terms of the signs and significance of the coefficients. However, it is notable that the coefficients for the spillovers variables in the domestic firm sample are generally of larger magnitude than the corresponding coefficients for the full sample. This provides further evidence to support that from the results in Table 2 showing that spillovers from foreign firms are more beneficial for purely domestic owned firms than for firms with direct foreign investment.

Given the results from Table 3, we conclude that the spillover effects from FDI decrease technical inefficiency of domestic firms in upstream and downstream industries. These findings confirm the argument in Javorcik (2004) that a foreign presence in a domestic market may generate not only spillover effects on domestic firms in the same industry but also provide spillover benefits to domestic firms in the upstream and downstream industries.

4.3. Spillover effects and the level of technical efficiency

So far, the analysis pools together all firms with different levels of efficiency. It has advantage of showing the average effect of FDI spillovers on a firm's technical efficiency. However, it has a disadvantage in that the spillover effects are assumed to be uniform for all firms. Thus, the analysis does not clearly distinguish which firms gain the most spillover effect from FDI.

In this section, the analysis is extended to answer a question of whether the level of efficiency influences the ability of firms in absorbing spillover benefits. The firms are divided into two groups: firms with a high-efficiency level and those with a low-efficiency level. The procedure to group the firms is by sorting the firms from the one with the highest technical efficiency level to the firm with the lowest efficiency level, and then the sorted firms are divided into two. The upper half of the data is categorized as the high-efficiency firms and the lower half is the low-efficiency firms. The estimation results for these two groups of firms are presented in Table 4. We estimate results for the full sample of firms as well as for the sub-sample of only domestic firms.

Starting from the full sample estimations, the coefficient of *FDI_Firm* is negative and statistically significant both among high-efficiency firms (column 1 of Table 4) and among low-efficiency firms (column 2), suggesting that foreign-owned firms have a higher technical efficiency level in both groups of firms. The positive and significant coefficient of *FDI_Sector* demonstrates that spillovers at the industrial level increase the inefficiency of the firms (i.e. a negative efficiency spillover). In contrast, the low-efficiency firms experience a decrease in technical inefficiency when foreign firms are more important in the industry (i.e. a positive efficiency spillover), as indicated by a negative and highly significant coefficient of *FDI_Sector* (column 2).

The coefficients of *FDI_Sector* for the sub-sample of only domestic firms (columns 3 and 4) are of the same sign and significance as in the corresponding full sample estimation, but the magnitude of impact is somewhat lower in the domestic firm sub-sample. This suggests that FDI spillovers have smaller impact on domestic firms than on foreign firms in industries with large technology gaps.

The results in [Table 4](#) demonstrate that firms with different efficiency levels may receive different effects of FDI spillovers. High-efficiency firms tend to obtain negative spillover effects, while low-efficiency firms experience positive spillover effects. These findings confirm the argument that there is advantage from being less advanced in terms of efficiency in terms of benefitting from spillovers ([Glass & Saggi, 1998](#); [Wang & Blomstrom, 1992](#)) and are consistent with the results in [Griffith et al. \(2002\)](#), [Castellani and Zanfei \(2003\)](#), and [Peri and Urban \(2006\)](#).

5. Conclusion

This article empirically examines the spillover effects of FDI on firm technical efficiency in the Indonesian manufacturing sector for the period between 1988 and 2000. Using the framework of [Battese and Coelli's \(1995\)](#) stochastic production frontier, we find evidence of a positive spillover effect of FDI to firms in the same industry (competitors), firms in an upstream industry (suppliers), and firms in a downstream industry (buyers). The positive spillover effect is observed in both the estimation for the full sample of firms and the estimation for the sub-sample of only domestic firms. Notably, the effects on domestic firms are generally more powerful than on other foreign firms in the same industry.

An interesting finding emerges when the samples are divided into two groups based on the level of efficiency. It is found that the low-efficiency firms receive a positive spillover effect from FDI across firms in the same industry. In contrast, the high-efficiency firms obtain a negative spillover effect. These findings support the argument of the advantage for absorbing spillovers goes to firms that are less advanced in terms of efficiency.

Outcomes from this study provide support on policies that encourage FDI. On the basis of these findings, policy makers should continue providing an FDI-friendly environment in order to maximize the spillover gains. Additional incentives may be provided for foreign firms that are willing to transfer their knowledge to domestic firms, especially those domestic firms in upstream and downstream industries that do not directly compete with the foreign firm. Variations in incentives may need to be considered, with more focus on FDI in sectors where purely domestic firms have a low-efficiency level compared to firms with direct foreign investment.

Acknowledgements

Helpful comments from two anonymous referees, editor Professor Michael Plummer and associate editor Professor Frank Hsiao, are gratefully acknowledged, but the authors are responsible for any remaining errors or omissions.

Appendix 1. Adjustments for constructing a consistent panel data

The steps of adjustment for constructing a consistent panel data are described as follows:

Step 1: Adjustment for industrial code.

The BPS reclassified the industrial codes twice: in 1990 and in 1998. This study adjusts the industrial codes to the 1990 code (KKI-1990) in order to obtain a consistent industrial code for the observation years (1988–2000). This adjustment involves two phases. First, the data from 1988 to 1989 (which use KKI-1985) are adjusted to KKI-1990 using the establishment identification code and a special map provided by the BPS. Observations in 1988–1989 not observed in 1990–1998 are removed, since there is no code from KKI-1990 that could be assigned to these observations. This first phase of adjustment removes 1346 out of the original 29,340 establishments. Second, the data from 1998 to 2000 (which use KKI-1998) are adjusted to KKI-1990 by the following concordance table provided by the BPS. There are several concordance issues that arise during this second phase of adjustment, which include unmatched classifications and incomplete entries. An example of an incomplete entry is an observation recorded only with a two-, three-, or four-digit classification code. For dealing with this problem, only observations with four-digit classification codes are retained, while those with two- and three-digit classification codes are removed. The retained observations with four-digit codes are then assigned as five-digit codes using the establishment specific identification code. By doing so, all establishments in the 1988–2000 panel data have consistent and integrated classification codes. The total establishments removed after these industrial code adjustments are 3078 out of 29,340 establishments, which include those with Oil and Gas classification (ISIC 353 and 354) as these sub-sectors are not observed in the 1988 and 1989 surveys.

Step 2: Adjustment for the variable definitions.

In some years, the variable definitions provided by the BPS are not consistent, even though the variables are the same. The author compared the variable definitions in each year's survey questionnaires (which are provided by the BPS together with the SI data) and recalculated the inconsistent variables for obtaining consistent definitions throughout the selected period.

Step 3: Cleaning for noise and typographical errors.

This study applied several steps for data cleaning in order to minimize noises and typographical errors:

- a. Observations with zero or a negative value of output, labour, material, or energy have been removed. This removes around 4.5 percent of the total observations.
- b. If a firm reports a missing value for a particular variable in a given time but reports values in the year before and after, an interpolation is carried out to fill the gap. The interpolation for the missing data was not more than 1 percent of the total observations.
- c. Typographical errors (or key-punch errors) in the raw data are adjusted for consistency. For example, if in the raw data, foreign share in a firm for the whole of the selected period was typed as 100 percent, except for a certain year being typed as 0 percent, then the 0 percent share is adjusted to 100 percent.
- d. Observations that are considered as outliers are removed from the data set by following a procedure suggested by [Takai \(2005\)](#). First, observations are sorted from the lowest to the highest value of output. Second, 1.5 percent of the lowest values and 1.5 percent of the highest values are removed.

Step 4: Back-casting the missing values of capital.

In some years, the values of capital are missing for quite a large number of observations. To fill these gaps, this study follows the methodology introduced by [Vial \(2006\)](#).

Step 5: Matching firms for a balanced panel

A balanced panel data set is constructed for the selected period by matching firms based on the specific identification code (PSID). This study utilizes STATA10 software for the matching.

Step 6: Choosing Industries with Foreign Firms

Since the purpose of the study is to estimate the FDI spillovers, industries (at a five-digit level) without foreign firms are excluded from the balanced panel.

Step 7: All monetary variables (output, capital, material, and energy) are deflated using price indexes. The output and material values are deflated using the wholesale price index (for four-digit ISIC industries); the machinery price index is used for deflating the value of capital; the nominal values of energy, which are a sum of electricity and fuel expenditures, are deflated using the electricity price index and the fuel price index. All price indexes are at a constant price of the year 1993.

By following the steps of adjustment, the final balance panel data consists of 3318 establishments with 43,225 observations.

Appendix 2. Definitions of variables

Symbol	Category	Unit	Definitions
<i>Production frontier</i>			
<i>Y</i>	Output	Million of 1993 rupiah	Gross output, which is deflated using a wholesale price index of four-digit ISIC industries at a constant price of 1993
<i>L</i>	Labour	Number of workers	Total number of employees directly and indirectly engaged in production, which covers all workers, including technical, administration, marketing, storage, and clerical staffs, who work full-time or part-time, and also family members.
<i>K</i>	Capital	Million of 1993 rupiah	Replacement value of fixed assets, which is deflated using a wholesale price index for machinery of four-digit ISIC industries at a constant price of 1993.
<i>M</i>	Material	Million of 1993 rupiah	Total value of material used in production, which cover raw and intermediate materials, both domestically produced and imported deflated using a wholesale price index of four-digit ISIC industries at a constant price of 1993.
<i>E</i>	Energy	Million of 1993 rupiah	Total value of electricity and fuel used by a firm. The value of electricity is calculated from the electricity provided by the state energy company (<i>Perusahaan Listrik Negara</i> or PLN) and those provided by private power firms, and it is deflated using the wholesale electricity index at a constant price of 1993. The value of fuels are calculated from nine types of fuels, namely premium, solar, kerosene, coal, cokes, gas, firewood, lubricant, and other fuels, and it is deflated using the OECD price of fuels published by <i>DX for Windows</i> at the 1993 constant price.
<i>T</i>	Time trend		Take a value of one for 1988, value of two for 1989, and so on.
<i>FDI_Sector</i>	FDI Variable	Ratio	The share of foreign firms' output over total outputs in a five-digit industry, or can be expressed as in Eq. (5). This variable measures the intra-industry (or horizontal) spillovers.
<i>Inefficiency function</i>			
<i>FDI_Firm</i>	FDI variable	Binary (one or zero)	The FDI at the firm level, which takes a value of one if a firm has a positive foreign ownership and take a value of zero if otherwise.
<i>FDI_Sector</i>	FDI variable	Ratio	The share of foreign firms' output over total outputs in a five-digit industry, or can be expressed as in Eq. (5). This variable measures the intra-industry (or horizontal) spillovers.
<i>FDI_Downstream_Sector</i>	FDI variable	Ratio	Spillovers from foreign firms in industries k ($k \neq j$) that are being supplied by domestic firms in industries j is defined as in Eq. (6).

Appendix 1 (Continued)

Symbol	Category	Unit	Definitions
<i>FDI_Upstream_Sector</i>	FDI variable	Ratio	Spillovers from foreign firms in industries m ($m \neq n$) that sell their outputs to domestic firms in industries n is defined as in Eq. (6).
<i>Year</i>	Dummy variable		A year dummy, which takes a value of one for all observations for the year in question, and a value of zero for other years.
<i>Industry</i>	Dummy variable		An industry dummy, which has a value of one for all observations for the industry in question and a value of zero for other industries.
<i>Firm</i>	Dummy variable		A firm dummy, which has a value of one for all observations for the firm in question and a value of zero for every other.

References

- Aigner, D. J., Lovell, C. A. K., & Schmidt, P. (1977). Formulation and estimation of stochastic frontier production function models. *Journal of Econometrics*, 6(1), 21–37.
- Aitken, B. J., & Harrison, A. E. (1999). Do Domestic Firms Benefit from Direct Foreign Investment? Evidence from Venezuela. *The American Economic Review*, 89(3), 605–618.
- Baltagi, B. D. (2011). *Econometrics* (5th edition). Heidelberg: Springer.
- Banker, R. D., Charnes, A., & Cooper, W. W. (1984). Some models for estimating technical and scale inefficiency in data envelopment analysis. *Management Science*, 30(9), 1078–1092.
- Battese, G., & Coelli, T. J. (1988). Prediction of firm-level technical efficiencies with a generalized frontier production function and panel data. *Journal of Econometrics*, 38(3), 387–399.
- Battese, G. E., & Coelli, T. J. (1993). A Stochastic Frontier Production Function Incorporating a Model for Technical Inefficiency Effects. *Working Paper in Econometrics and Applied Statistics, Department of Economics University of New England*.
- Battese, G. E., & Coelli, T. J. (1995). A model for technical inefficiency effects in a stochastic frontier production function for panel data. *Empirical Economics*, 20(2), 325–332.
- Bauer, P. W. (1990). Recent developments in the econometric estimation of frontiers. *Journal of Econometrics*, 46(1–2), 39–56.
- Bjurek, H. L., Hjarmarsson, L., & Forsund, F. R. (1990). Deterministic parametric and nonparametric estimation in service production. *Journal of Econometrics*, 46(1–2), 213–227.
- Blalock, G., & Gertler, P. J. (2008). Welfare gain from foreign direct investment through technology transfer to local suppliers. *Journal of International Economics*, 74(2), 402–421.
- Blomstrom, M., & Kokko, A. (1998). Multinational corporation and spillovers. *Journal of Economic Surveys*, 12(2), 247–277.
- Bravo-Ureta, B. E., & Pinheiro, A. E. (1993). Efficiency analysis of developing country agriculture: A review of the frontier function literature. *Agricultural and Resource Economics Review*, 22(1), 88–101.
- Castellani, D., & Zanfei, A. (2003). Technology gaps, absorptive capacity and the impact of inward investments on productivity of European firms. *Economics of Innovation and New Technology*, 12(6), 555–576.
- Caves, R. E. (1971). International corporations: The industrial economics of foreign investment. *Economica*, 38(149), 1–27.
- Caves, R. E. (1974). Multinational firms, competition and productivity in host country markets. *Economica*, 41(162), 176–193.
- Central Bank of Indonesia (2011). *Economic and financial data for Indonesia*.
- Chakraborty, C., & Nunnenkamp, P. (2008). Economic reforms, FDI, and economic growth in India: A sector level analysis. *World Development*, 36(7), 1192–1212.
- Charnes, A., Cooper, W. W., & Rhodes, E. (1978). Measuring the efficiency of decision making units. *European Journal of Operational Research*, 2(6), 429–444.
- Coelli, T. J. (1995). Recent developments in frontier modelling and efficiency measurement. *Australian Journal of Agricultural Economics*, 39(3), 219–245.
- Coelli, T. J. (1996). A guide to frontier version 4.1: a computer program for stochastic frontier production and cost function estimation. *CEPA Working Paper No. 07/96 University of New England*.
- Coelli, T. J., Rao, D. S. P., O'Donnell, C. J., & Battese, G. E. (2005). *An introduction to efficiency and productivity analysis* (2nd ed.). New York: Springer.
- Dimelis, S., & Lauri, H. (2002). Foreign direct investment and efficiency benefits: A conditional quartile analysis. *Oxford Economic Papers*, 54(3), 449–469.
- Djankov, S., & Hoekman, B. (2000). Foreign investment and productivity growth in Czech enterprises. *World Bank Economic Review*, 14(1), 49–64.
- Dunning, J. (1988). *Multinational technology and competitiveness*. London: Allen & Unwin.
- Findlay, R. (1978). Relative backwardness, direct foreign investment, and the transfer of technology: A simple dynamic model. *Quarterly Journal of Economics*, 92(1), 1–16.
- Forsund, F. R. C. A. K., Lovell, P., & Schmidt (1980). A survey of frontier production functions and of their relationship to efficiency measurement. *Journal of Econometrics*, 13(1), 5–25.
- Ghali, S., & Rezgui, S. (2008). FDI contribution to technical efficiency in the Tunisian manufacturing sector. *ERF Working Paper Series No. 421*.
- Glass, A., & Saggi, K. (1998). International technology transfer and the technology gap. *Journal of Development Economics*, 55(2), 369–398.
- Gorg, H., & Greenaway, D. (2004). Much ado about nothing? Do domestic firms really benefit from foreign direct investment?. *The World Bank Research Observer*, 19(2), 171–197.
- Gorg, H., & Strobl, E. (2005). Spillovers from foreign firms through worker mobility: An empirical investigation. *Scandinavian Journal of Economics*, 107(4), 693–739.
- Greene, W. H. (1993). In H. O. Fried, C. A. K. Lovell, & S. S. Schmidt (Eds.), *The econometric approach to efficiency analysis. The measurement of productive efficiency: Techniques and applications*. New York: Oxford University Press.
- Griffith, R., Redding, S., & Simpson, H. (2002). Productivity Convergence and Foreign Ownership at the Establishment Level. *Institute Fiscal Studies Working Paper 22, London*.
- Haddad, M., & Harrison, A. E. (1993). Are there positive spillovers from foreign direct investment? Evidence from panel data for Morocco. *Journal of Development Economics*, 42(1), 51–74.
- Hu, A. G. Z., & Jefferson, G. H. (2002). FDI impact and spillover: Evidence from china's electronic and textile industries. *The World Economy*, 25(8), 1063–1076.
- Hymer, S. H. (1960). (PhD dissertation). In *The international operations of national firms: A study of direct foreign investment* (p. 1976). MIT, MA: MIT Press.
- Javorcik, B. S. (2004). Does foreign direct investment increase the productivity of domestic firms? In search of spillovers through backward linkages. *American Economic Review*, 94(3), 605–627.
- Kugler, M. (2006). Spillovers from foreign direct investment: Within or between industries? *Journal of Development Economics*, 80(2), 444–477.
- Kumbhakar, S. C., Ghosh, S., & McGuckin, J. T. (1991). A generalized production frontier approach for estimating determinants of inefficiency in US dairy farms. *Journal of Business and Economic Statistics*, 9(3), 279–286.
- Liang, F. H. (2007). Does foreign direct investment improve the productivities of domestic firms? Technology spillovers within and between industries. *Haas Berkeley Working Paper*. <http://www.faculty.haas.berkeley.edu/fenliang/research/spillover/FDIspillover.pdf> [accessed 20.07.07].
- Lipsey, R. E., & Sjöholm, F. (2005). In T. H. Moran, E. Graham, & M. Blomstrom (Eds.), *The impact of inward FDI on Host countries: Why such different answers? does foreign direct investment promote development?* (pp. 23–43). Washington, DC: Institute for International Economics and Center for Global Development.
- Lovell, C. A. K. (1993). In H. O. Fried, C. A. K. Lovell, & S. S. Schmidt (Eds.), *Production frontiers and productive efficiency. The measurement of productive efficiency: Techniques and applications*. New York: Oxford University Press.

- Meeusen, W., & van den Broeck, J. (1977). Efficiency estimation from Cobb–Douglas production function with composed error. *International Economic Review*, 18(2), 435–444.
- Negara, S. D., & Firdausy, C. M. (2011). In C. Sussangkarn, Y. C. Park, & S. J. Kang (Eds.), *The development of foreign direct investment and its impact on firms' productivity, employment and exports in Indonesia. Foreign direct investments in Asia*. London, UK: Routledge.
- Olesen, O. B., Peterson, N. C., & Lovell, C. A. K. (1996). Editors' introduction. *Journal of Productivity Analysis*, 7(2/3), 87–98.
- Peri, G., & Urban, D. (2006). Catching up to foreign technology? Evidence on the 'Veblen-Gerschenkron' effect of foreign investments. *Regional Science and Urban Economics*, 36(1), 72–98.
- Schiff, M., & Wang, Y. (2008). North–South and South–South trade-related technology diffusion: how important are they in improving TFP growth? *Journal of Development Studies*, 44(1), 49–59.
- Schmidt, P. (1985). Production frontier functions. *Econometric Reviews*, 4(2), 289–328.
- Smeets, R. A. (2008). Collecting the pieces of the FDI knowledge spillovers puzzle. *The World Bank Research Observer*, 23(2), 107–138.
- Suyanto, Salim, R., & Bloch, H. (2009). Does foreign direct investment lead to productivity spillovers? Firm level evidence from Indonesia. *World Development*, 37(12), 1861–1877.
- Suyanto, Bloch, H., & Salim, R. (2012). FDI spillovers and productivity growth in Indonesian garment and electronics manufacturing. *Journal of Development Studies* (in press).
- Takii, S. (2005). Productivity spillovers and characteristics of foreign multinational plants in Indonesian manufacturing 1990–1995. *Journal of Development Economics*, 76(2), 521–542.
- Takii, P. (2011). Do FDI spillovers vary among home economies? Evidence from Indonesian manufacturing. *Journal of Asian Economics*, 22(2), 152–163.
- Temenggung, D. (2007). Productivity spillovers from foreign direct investment: Indonesian manufacturing industry's experience 1975–2000, *mimeograph*, Australian National University, Canberra, Australia.
- Vial, V. (2006). New estimates of total factor productivity growth in Indonesian manufacturing. *Bulletin of Indonesian Economic Studies*, 42(3), 357–369.
- Wang, J. W., & Blomstrom, M. (1992). Foreign investment and technology transfer: A simple model. *European Economic Review*, 36(1), 137–155.
- Wang, H. J., & Schmidt, P. (2002). One-step and two-step estimation of the effects of exogenous variables on technical efficiency level. *Journal of Productivity Analysis*, 18(2), 129–144.

View Item: Which firms benefit from foreign direct investment? Empirical Evidence from Indonesian manufacturing

This item is in the repository with the URL
<http://repository.ubaya.ac.id/20659/>

Preview

Details

Actions

History

., Suyanto and Salim, Ruhul A. and Bloch, Harry (2014) *Which firms benefit from foreign direct investment? Empirical Evidence from Indonesian manufacturing*. Journal of Asian Economics, 33. pp. 16-29. ISSN 1049-0078



PDF

Suyanto journal asian economics August 2014.pdf - Accepted Version

[Download \(136Kb\)](#) | [Preview](#)

Official URL: <http://www.sciencedirect.com/science/article/pii/S...>

Abstract

Despite growing concern regarding the productivity benefits of foreign direct investment (FDI), very few studies has been conducted on the impact of FDI on firm-level technical efficiency. This study helps fill this gap by empirically examining the spillover effects of FDI on the technical efficiency of Indonesian manufacturing firms. A panel data stochastic production frontier (SPF) method is applied to 3,318 firms surveyed over the period 1988-2000. The results reveal evidence of positive FDI spillovers on technical efficiency. Interesting differences emerge however when the samples are divided into two efficiency levels. High-efficiency domestic firms receive negative spillovers, in general, while low-efficiency firms gain positive spillovers. These findings justify the hypothesis of efficiency gaps, that the larger the efficiency gap between domestic and foreign firms the easier the former extracts spillover benefits from the latter.

Item Type: Article

Uncontrolled Keywords: Foreign direct investment; Spillover effects; Technical efficiency; Stochastic production frontier;

Indonesia

Subjects: [H Social Sciences](#) > [H Social Sciences \(General\)](#)

Divisions: [Faculty of Business and Economic](#) > [Department of Economic](#)

Depositing User: Suyanto 3198

Date Deposited: 12 Sep 2014 08:44

Last Modified: 12 Sep 2014 08:44

URI: <http://repository.ubaya.ac.id/id/eprint/20659>

LEMBAR
HASIL PENILAIAN SEJAWAT SEBIDANG ATAU PEER REVIEW
KARYA ILMIAH : JURNAL ILMIAH

Judul Karya Ilmiah(artikel) : Which Firms Benefit from Foreign Direct Investment? Empirical Evidence from Indonesian Manufacturing

Jumlah Penulis : 3 Orang

Status Pengusul : Penulis Pertama

Identitas Jurnal Ilmiah :

- a. Nama Jurnal : Journal of Asian Economics
- b. Nomor ISSN : 1049-0078
- c. Vol, No, Bln, Thn : Vol. 33,
- d. Penerbit : Elsevier
- e. DOI Artikel : doi:10.1016/j.asieco.2014.05.003
- f. Alamat Web Jurnal : <http://www.sciencedirect.com/science/article/pii/S1049007814000244>
- g. Terindeks di : <http://www.scimagojr.com/journalsearch.php?q=22736&tip=sid&clean=0>

Kategori Publikasi Jurnal Ilmiah : Jurnal Ilmiah ~~Internasional~~ / Internasional Bereputasi Berdampak Faktor
 (beri \checkmark pada ketegori yang tepat)

Jurnal Ilmiah Nasional Terakreditasi

Jurnal Ilmiah Nasional / Nasional Terindeks di DOAJ, CABI, COPENICUS

Hasil Penilaian *Peer Review* :

No.	Komponen Yang Dinilai	Nilai Maksimal Jurnal Ilmiah			Nilai Akhir Yang Diperoleh ...(2)
		Internasional ...(1)	Nasional Terakreditasi	Nasional	
		<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	
1.	Kelengkapan unsur isi jurnal (10%)	4	---	---	$0,6 \times 4 = 2,4$
2.	Ruang lingkup dan kedalaman pembahasan (30%)	12	---	---	$0,6 \times 11 = 6,6$
3.	Kecukupan dan kemutakhiran data/infromasi dan metodologi (30%)	12	---	---	$0,6 \times 12 = 7,2$
4.	Kelengkapan unsur dan kualitas penerbit (30%)	12	---	---	$0,6 \times 12 = 7,2$
Total = (100%)		40	---	---	23,4
Nilai Pengusul=					

Catatan Penilaian Artikel oleh Reviewer:

Scopus coverage years: 1990-on going. CiteScore 1,07 and SJR 0,508. Journal Rank #199/533 untuk kategori Economics and Econometrics dan #68/216 untuk kategori Finance. Telah dilakukan cek similarity. Tulisan baik dengan penekanan pada keseluruhan perusahaan manufaktur di Indonesia yang disurvei oleh BPS. Kontribusi utama terletak pada cakupan data yang komprehensif dan analisis yang mencakup dampak dalam industri dan antar industri. Matriks input-output dipergunakan sebagai patokan evaluasi inter-industri spillovers.

Surabaya, 13 Mei 2016

Reviewer 1



Prof. Dr. R. Wilopo, Ak., CA, CFE

NIP / NPK ...**(3)** : 36940141

Unit Kerja ...**(4)** : STIE PERBANAS Surabaya

**LEMBAR
HASIL PENILAIAN SEJAWAT SEBIDANG ATAU PEER REVIEW
KARYA ILMIAH : JURNAL ILMIAH**

Judul Karya Ilmiah(artikel) : Which Firms Benefit from Foreign Direct Investment? Empirical Evidence from Indonesian Manufacturing

Jumlah Penulis : 3 Orang

Status Pengusul : Penulis Pertama

- Identitas Jurnal Ilmiah :
- a. Nama Jurnal : Journal of Asian Economics
 - b. Nomor ISSN : 1049-0078
 - c. Vol, No, Bln, Thn : Vol. 33,
 - d. Penerbit : Elsevier
 - e. DOI Artikel : doi:10.1016/j.asieco.2014.05.003
 - f. Alamat Web Jurnal : <http://www.sciencedirect.com/science/article/pii/S1049007814000244>
 - g. Terindeks di : <http://www.scimagojr.com/journalsearch.php?q=22736&tip=sid&clean=0>

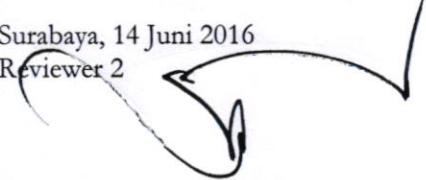
- Kategori Publikasi Jurnal Ilmiah : Jurnal Ilmiah Internasional / Internasional Bereputasi Berdampak Faktor (beri \checkmark pada ketegori yang tepat)
- Jurnal Ilmiah Nasional Terakreditasi
- Jurnal Ilmiah Nasional / Nasional Terindeks di DOAJ, CABI, COPERNICUS

Hasil Penilaian Peer Review :

No.	Komponen Yang Dinilai	Nilai Maksimal Jurnal Ilmiah			Nilai Akhir Yang Diperoleh ... (2)
		Internasional ... (1)	Nasional Terakreditasi	Nasional	
		<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	
1.	Kelengkapan unsur isi jurnal (10%)	4	---	---	0,6 x 4 = 2,4
2.	Ruang lingkup dan kedalaman pembahasan (30%)	12	---	---	0,6 x 11,5 = 6,9
3.	Kecukupan dan kemutakhiran data/infromasi dan metodologi (30%)	12	---	---	0,6 x 11,5 = 6,9
4.	Kelengkapan unsur dan kualitas penerbit (30%)	12	---	---	0,6 x 11,5 = 6,9
Total = (100%)		40	---	---	23,1
Nilai Pengusul=					

Catatan Penilaian Artikel oleh Reviewer:
 Cek web Scopus: journal coverage years 1990 – on going, Cite Score 2016 sebesar 1.07 dan Scimago Journal Ranking 0.508. Penerbit Elsevier berlokasi Kiddlington, Oxford, UK. Sudah similarity check dengan Turnitin. Kualitas tulisan baik. Referensi cukup up-to-date. Tulisan kemungkinan merupakan kelanjutan dari tulisan sebelumnya di Journal of Development Studies (2012) dengan pengembangan pada cakupan data observasi sejumlah 43.225 observasi, jauh lebih besar daripada penelitian sebelumnya. Inter-industri spillovers dimunculkan untuk memperlihatkan keterhubungan pengaruh antar industri. Kontribusi tambahan berupa adanya estimasi terhadap perusahaan dengan efisiensi tinggi dan perusahaan dengan efisiensi rendah.

Surabaya, 14 Juni 2016
 Reviewer 2



Prof. Dr. Munawar Ismail, SE, DEA
 NIP / NPK ... (3) : 19570212198401003
 Unit Kerja ... (4) : FEB Universitas Brawijaya

Which Firm Benefit From Foreign Direct Investment? Empirical Evidence From Indonesian Manufacturing

by 4 Suyanto

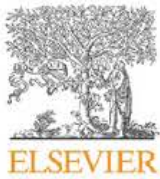
Submission date: 27-Mar-2018 05:14PM (UTC+0700)

Submission ID: 936926259

File name: Ill.1.C.1.5_asli.pdf (338.75K)

Word count: 11256

Character count: 59957



Contents lists available at ScienceDirect

Journal of Asian Economics



Which firms benefit from foreign direct investment? Empirical evidence from Indonesian manufacturing

Suyanto^a, Ruhul Salim^{b,*}, Harry Bloch^b^a Faculty of Economics, University of Surabaya, East Java, Indonesia^b Curtin Business School, Curtin University, Perth, WA 6845, Australia

ARTICLE INFO

Article history:

Received 22 November 2012

Received in revised form 8 May 2014

Accepted 11 May 2014

Available online 19 May 2014

JEL classification:

D24

D29

F23

Keywords:

Foreign direct investment

Spillover effects

Technical efficiency

Stochastic production frontier

Indonesia

ABSTRACT

Despite growing concern regarding the productivity benefits of foreign direct investment (FDI), very few studies have been conducted on the impact of FDI on firm-level technical efficiency. This study helps fill this gap by empirically examining the spillover effects of FDI on the technical efficiency of Indonesian manufacturing firms. A panel data stochastic production frontier (SPF) method is applied to 3318 firms surveyed over the period 1988–2000. The results reveal evidence of positive FDI spillovers on technical efficiency. Interesting differences emerge however when the samples are divided into two efficiency levels. High-efficiency domestic firms receive negative spillovers, in general, while low-efficiency firms gain positive spillovers. These findings justify the hypothesis of efficiency gaps, that the larger is the efficiency gap between domestic and foreign firms the easier the former extracts spillover benefits from the latter.

© 2014 Elsevier Inc. All rights reserved.

1. Introduction

The spillover effects of foreign direct investment (FDI) have been a major concern for researchers and policy makers during the last two decades. A number of studies have examined the spillover effects of FDI on domestic firm productivity (Chakraborty & Nunnenkamp, 2008; Haddad & Harrison, 1993; Hu & Jefferson, 2002; Javorcik, 2004; Liang, 2007; Negara & Firdausy, 2011; Takii, 2005, 2011). These studies provide some useful insights regarding the evidence of the spillover benefits and offer some recommendations to maximize the benefits. However, most existing studies exclude technical efficiency and focus mainly on technology, ignoring that the FDI presence in host countries is the impetus for efficiency improvement through competition and demonstration effects (Wang & Blomstrom, 1992). A study of FDI spillover effects on firm-level technical efficiency is important to provide evidence as to whether the large amount of FDI inflows generate positive externalities to domestic firms through efficiency improvement, thus indicating whether the spillover hypothesis is justified in the context of technical efficiency. Such a study can explore to what extent FDI can induce efficiency spillovers, and which firm types really benefit from the spillovers.

* Corresponding author at: School of Economics & Finance, Curtin Business School, Curtin University, WA 6845, Australia. Tel.: +61 892664577; fax: +61 92663026.

E-mail address: Ruhul.Salim@cbs.curtin.edu.au (R. Salim).

Among the developing economies, Indonesia is particularly successful in attracting FDI. Net FDI inflows to Indonesia have risen more than 30 times since 1986, reaching a record level of US\$8.3 billion in 2008 (the Central Bank of Indonesia, 2011). However, there is a dearth of research on efficiency spillovers in Indonesia. Most empirical studies examine spillover effects under a framework of the long-run equilibrium production function, which assumes that firms are producing at a full efficiency level. Under this framework, the FDI spillovers on technical efficiency are not captured.

Two previous studies by the authors focus on technical efficiency using a stochastic production frontier framework for individual Indonesian manufacturing industries. Suyanto, Salim, and Bloch (2009) examine the pharmaceutical and chemical industries, while Suyanto et al. (2012) examine the electronic and garment industries. However, there are no studies providing comprehensive results for the whole Indonesian manufacturing sector using a stochastic framework.

A study by Temenggung (2007) examines the whole Indonesian manufacturing sector. Our current research differs from Temenggung in three important points. Firstly, Temenggung applies the ordinary least squared (OLS) regression method for panel data, which doesn't distinguish between fixed effects (FE) and random effects (FE). Secondly, the classical production function, employed in Temenggung (2007), assumes that all firms are fully efficient, so that the spillover effects of FDI reflect technological progress. In contrast, the current paper employs the stochastic production frontier, which relaxes the assumption of full efficiency of firms, so that both technological progress and efficiency improvement are examined. Thirdly, we calculate the scores of technical efficiency of each firm and estimates spillover effects separately for high-efficiency and low-efficiency firms, providing a useful insight into the differences in the ability of high-efficiency and low-efficiency firms in absorbing spillover effects from FDI.

This study contributes to the existing literature in several ways. Firstly, it examines the spillover hypothesis by focusing on technical efficiency, an important aspect that is often neglected in the previous studies. The adoption of a stochastic production frontier allows the authors to investigate the effects of FDI spillovers on firm-level technical efficiency. Secondly, this study covers a long series of surveyed firms, which includes also the period of the Asian crisis onwards. Thirdly, this study evaluates horizontal, backward, and forward spillovers of FDI. Most importantly, by examining the whole manufacturing sector, it is possible to identify characteristics of industries that affect the size of the technology and efficiency spillovers to domestic firms from FDI. In particular, we find evidence that the size of the technology gap between foreign and domestic firms is critical, with larger efficiency gaps associated with greater efficiency spillovers from FDI.

We proceed by reviewing the concept of spillover effects in the next section. We then discuss methodology and data. Empirical results are presented in Section 4 and the conclusions are given in the final section.

2. FDI, spillover effects, and technical efficiency: theoretical concept and empirical evidence

2.1. FDI and spillover effects

Foreign direct investment is believed to provide host countries with direct and indirect benefits. The direct benefits take the forms of new investments that boost national income, increase tax revenues, and provide new employment; whereas the indirect benefits are in the forms of externalities that are generated through non-market mechanisms to recipient economies and domestic firms within the economies (Hymer, 1960). These indirect benefits are commonly known as FDI spillovers.

The literature identifies at least three types of FDI spillovers. These are productivity spillovers, market-access spillovers, and pecuniary spillovers. Productivity spillovers are defined as the externalities from FDI that lead to increases in the productivity of domestic firms (Aitken & Harrison, 1999). Market-access spillovers exist when the presence of FDI generates an opportunity for domestic firms to access international markets (Blomstrom & Kokko, 1998). Pecuniary spillovers happen if the existence of FDI affects the profit functions of domestic firms through a reduction in costs or an increase in revenues (Gorg & Strobl, 2005).

Of the three types of FDI spillovers, productivity spillovers have been a particular concern among policy makers and researchers in the last two decades. Various incentives have been provided by policy makers to attract FDI and substantial efforts have been devoted by researchers to evaluate the productivity advantage. However, the empirical evidence is mixed at best. Some studies find evidence of positive productivity spillovers (Caves, 1974; Javorcik, 2004; Kugler, 2006; Schiff & Wang, 2008; Temenggung, 2007), but others discover nonexistent or even negative spillovers (Aitken & Harrison, 1999; Blalock & Gertler, 2008; Djankov & Hoekman, 2000). Thus, the relationship between FDI spillovers and firm productivity remains a controversial issue.

2.2. Spillover effects and firm-specific characteristics

Some researchers argue that the mixed evidence intuitively implies that the spillover effects are not an automatic consequence of the foreign presence in an economy, rather they depend significantly on the characteristics of firms in the industries (Gorg & Greenaway, 2004; Lipsey & Sjöholm, 2005; Smeets, 2008). One important characteristic of firms is the technology gap between foreign and domestic firms. In a study on UK manufacturing firms, Griffith, Redding, and Simpson (2002) find that the wider the technology gap the larger the FDI spillover effects that are obtained by domestic firms. This finding indicates a benefit of being less advanced in terms of technology, which supports the theoretical argument in Findlay (1978). A similar result is discovered also by Castellani and Zanfei (2003) for France and Spain, and by Peri and Urban (2006) for Italy and Germany.

Although there is an advantage in being less advanced, the technology gap should not be too wide (Wang & Blomstrom, 1992). A minimum level of technology is required for domestic firms to absorb the new technology from foreign firms. When the gap is too wide, the limited kind absorptive capacity of domestic firms may not permit assimilation the new technology (Glass & Saggi, 1998).

2.3. Technical efficiency gains from FDI spillovers

Earlier studies on FDI productivity spillovers focus on technology advantages (Gorg & Greenaway, 2004). The knowledge from foreign firms is regarded synonymously with technological knowledge, as this is consistent with the use of a conventional production function. Managerial and organizational knowledge that may lead to efficiency spillovers are not portrayed since firms are assumed to be producing at the long-run equilibrium with a full efficiency capacity. Thus, the productivity spillovers in these early studies are identically measured as technology spillovers.

More recent studies focus on both efficiency and technology advantages. In these studies, knowledge is defined broadly as product, process, managerial, and organizational knowledge. Hence, productivity spillovers lead to both technology and efficiency advantages. Unfortunately, studies that investigate efficiency advantages are not plentiful. In a study on Greek manufacturing firms, Dimelis and Lauri (2002) examine the effect of foreign equity shares on efficiency and find a positive relationship between these two variables. Also, Ghali and Rezgui (2008) analyze the Tunisian manufacturing sector and find that higher foreign share increase firm efficiency. Addressing the same issue but employing a different estimation method, our study investigates the efficiency spillovers in Indonesian manufacturing firms. We extend the studies by Dimelis and Lauri (2002) and Ghali and Rezgui (2008) by focusing on vertical spillovers as well as horizontal spillovers.

3. Methodology, data set, and variables

3.1. Methodology

There are two commonly used methods in measuring efficiencies and productivity at the firm level, namely data envelopment analysis (DEA) and stochastic frontier analysis (SFA).¹ Each of the two methods has its advantages and disadvantages, as explained below. The choice between these methods thus depends on the objective of the research, the type of firms in the chosen industry, and the nature of the data (Coelli, Rao, O'Donnell, & Battese, 2005; Olesen, Peterson, & Lovell, 1996).

DEA is a linear programming method that observes production possibilities using the technique of envelopment and measures efficiency as the distance to the frontier (Banker, Charnes, & Cooper, 1984; Charnes, Cooper, & Rhodes, 1978). This method has the primary advantage of being of a non-parametric nature and has the ability to handle multiple outputs and multiple inputs.² However, it has the disadvantage of producing biased estimates in the presence of measurement error and other statistical noise, as this method does not separate the stochastic random noise from the inefficiency effects (Schmidt, 1985). Hence, the estimation results under this method tend to be very sensitive to small changes in the data.

Alternatively, the stochastic frontier method is a regression-based method that assumes two separate unobserved error terms, one represents efficiency and the other represents statistical noise (Aigner, Lovell, & Schmidt, 1977; Meeusen & van den Broeck, 1977). It has a chief advantage in the ability to measure efficiency in the presence of statistical noise. However, this method is parametric and requires a specific functional form and distributional assumptions for the error terms (Coelli et al., 2005).

In this study the stochastic frontier method is applied to analyze the spillover effects from FDI. The one-stage stochastic production frontier (SPF) is used to estimate a production frontier and a technical inefficiency function simultaneously. As pointed out by Kumbhakar, Ghosh, and McGuckin (1991) and Wang and Schmidt (2002), the one-stage approach is preferable than the two-stage approach, as the latter exhibits at least two limitations in estimation that can lead to potentially severe bias. The first limitation is that technical efficiency might be correlated with the production inputs, which may cause inconsistent estimates of the production frontier. The second limitation is the OLS method in the second stage is inappropriate since technical efficiency distribution is assumed to be one-sided. Considering the advantages, the current study adopts the one-stage approach, following Battese and Coelli (1995).

The Battese–Coelli production frontier can be expressed as follows:

$$y_{it} = f(x_{it}, t; \beta) \exp(v_{it} - u_{it}) \quad (1)$$

and the inefficiency function may be written as:

$$u_{it} = z_{it} \delta + w_{it} \quad (2)$$

¹ Comprehensive reviews of the two methods are provided by Forsund et al. (1980), Bauer (1990), Bjurek et al. (1990), Bravo-Ureta and Pinheiro (1993), Greene (1993), Lovell (1993), and Coelli (1995).

² The non-parametric nature of DEA allows for measuring efficiency without imposing a specific functional form and a distributional assumption on data.

where y_{it} denotes the production of the i th firm ($i = 1, 2, \dots, N$) in the t th time period ($t = 1, 2, \dots, T$), x_{it} denotes a $(1 \times k)$ vector of explanatory variables, β represents the $(k \times 1)$ vector of parameters to be estimated, \exp denotes exponential, v_{it} is the time specific and stochastic error, with $iid N(0, \sigma^2_v)$, and u_{it} represents technical inefficiency, which is assumed as a function of a $(1 \times j)$ vector of observable non-stochastic explanatory variables, z_{it} , and a $(j \times 1)$ vector of unknown parameters to be estimated, δ , and w_{it} is an unobservable random variable.

The parameters of Eqs. (1) and (2) are estimated using the maximum likelihood estimator (MLE) by following the three steps as explained in Coelli (1996). With simultaneous equation estimation, the MLE estimates are unbiased and efficient. The variance parameters of the Battese-Coelli's model are defined as $\sigma^2_s \equiv \sigma^2_v + \sigma^2_u$ and $\gamma \equiv \sigma^2_u / \sigma^2_s$.³

γ is an important parameter to decide whether there is technical inefficiency or not in the model. If the estimated value of γ is not statistically significant, there is no technical inefficiency and the results obtained from estimating Eq. (1) by ordinary least squares (OLS) would be efficient. In contrast, if the estimated value of γ is statistically significant, then there is technical inefficiency and Eqs. (1) and (2) should be estimated simultaneously.

The technical efficiency of the i th firm calculated from the Eqs. (1) and (2) is the ratio of observed output of the firm to its potential maximum output, which can be written as:

$$TE_{it} = \frac{y_{it}}{y_{it}^p} = \exp(-u_{it}) \quad (3)$$

Following Battese and Coelli (1988), the best estimator of the $\exp(-u_{it})$ is its conditional expectation, $E[\exp(-u_{it})]$, so technical efficiency can be written as:

$$TE_{it} = E[\exp(-u_{it})] \quad (4)$$

If it is assumed that the production frontier takes the form of a log-linear production function and there are four input variables (labour, capital, material, and energy) in the production process, the empirical model can be expressed in natural logarithms of variables as:

$$\ln Y_{it} = \beta_0 + \beta_1 \ln L_{it} + \beta_2 \ln K_{it} + \beta_3 \ln M_{it} + \beta_4 \ln E_{it} + \beta_5 T + \beta_6 \ln FDI_Sector + v_{it} - u_{it} \quad (5)$$

where Y is output, L is labour, K is capital, M is material, E is energy, T is a time-trend variable that increases by one for each year, FDI_Sector is a measure of FDI horizontal spillovers as explained in the next section and the other variables are as previously defined.

The inefficiency effect as a function of a set of FDI variables, a year dummy, an industry dummy, and a firm dummy can be written as:

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Sector_{jt} + \delta_3 FDI_Firm_{it} \times FDI_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (6)$$

where FDI_Firm is a dummy variable for foreign direct investment that takes a value of zero if a firm has no foreign ownership share and takes a value of one if a foreign firm has a positive share, FDI_Sector is as defined above, $Year$ is a year dummy variable, $Industry$ is an industry dummy and $Firm$ is a firm dummy. The interaction term of $FDI_Firm \times FDI_Sector$ is included in the inefficiency equation to estimate whether foreign and domestic firms benefit equally from the presence of a new foreign firm. A positive (negative) coefficient on the interaction term indicates less (more) efficiency gain for foreign firms than for domestic firms.

Eq. (6) is used to estimate the intra-industry spillovers, which capture the effects of foreign presence on the technical efficiency of firms in the same industry. The inter-industry spillovers are commonly estimated by replacing the horizontal-spillover variable (FDI_Sector) with vertical-spillover variables. The inefficiency function for the inter-industry spillovers can be expressed as:

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Downstream_Sector_{jt} + \delta_3 FDI_Firm_{it} * FDI_Downstream_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (7)$$

or

$$u_{it} = \delta_0 + \delta_1 FDI_Firm_{it} + \delta_2 FDI_Upstream_Sector_{jt} + \delta_3 FDI_Firm_{it} * FDI_Upstream_Sector_{jt} + \delta_4 Year + \delta_5 Industry + \delta_6 Firm + w_{it} \quad (8)$$

where $FDI_Downstream_Sector$ is a proxy for spillover effects from foreign firms to foreign and domestic suppliers and $FDI_Upstream_Sector$ is a proxy for spillover effects from foreign firms to foreign and domestic buyers.

³ The complete derivation the log-likelihood function of the Battese-Coelli model and its related variance parameters are discussed in Battese and Coelli (1993).

3.2. Data and data set construction

The primary data for our study are the annual surveys of medium and large manufacturing establishments (Survey Tahunan Statistik Industri or SI) conducted by the Indonesian Central Board of Statistics (*Badan Pusat Statistik* or BPS). These annual surveys cover a wide range of information from each surveyed establishment. The basic information includes year of starting production, industrial classification, location, and the specific identification code. There is also information regarding ownership, which includes foreign and domestic ownership, and information related to production, such as gross output, number of workers in production and non-production, value of fixed capital, material usage, and energy consumption.

The annual surveys have been conducted since 1975 and the most recent available data relates to the year 2007. However, this study uses the data from 1988 to 2000. The year 1988 is chosen as a starting year since it is the first year that the replacement value of fixed assets, which is used as a measure for capital, is available. The year 2000 is selected as the last year because the BPS changed the specific identification code in 2001 to the new identification code (KIPN) without providing a concordance table to the previous used identification code (PSID). Efforts to match the observations in the years 2001–2005 to the years 1988–2000 using output values and labour do not yield consistent results. Therefore, the longest possible period for this study is 1988–2000.

In constructing a consistent data set, several adjustments are conducted. These include adjustment for industrial code, adjustment for variable definitions, cleaning for noise and typological errors, back casting missing values of capital, matching firms for a balanced panel, and choosing industries with foreign firms. The balanced panel data are preferable in this study due to two advantages: (1) it enables tracing the technical efficiency scores of each observed firm during the period of study; (2) it removes the influence of a firm that appears only in one or two years, while the period of estimation is for 13 years. The details of adjustments are presented in Appendix 1. After the adjustments, the final balanced panel of data consists of 3318 establishments with 43,225 observations.

To show the influence of the construction of the balanced panel dataset, the descriptive statistics of the related variables are calculated for the balanced panel data and for the original data before the adjustment process. The original data consist of establishments that do not report complete information on output, labour, capital, material, or energy. Therefore these establishments are not included in the calculation of the descriptive statistics for original data. Following Takii (2005), (1) 0.5 percent observations with the lowest values of output and 1.5 percent observations with the highest values of output are removed. After these deletions, the descriptive statistics for the original total data, as presented in Table 1, consists of 24,188 establishments for an unbalanced panel of 238,628 observations.

Table 1 shows that the minimum values of variables $\ln Y$, $\ln L$, $\ln K$, $\ln M$, $\ln E$ for the original data are lower if compared to the minimum values of those variables from the balanced panel. This makes sense as the balanced panel data removes some observations during the adjustment process. The maximum values of those variables are higher in the original data compared to those in balanced panel data. The mean values of these five variables are higher in the balanced panel data compared to those in original data, while the standard deviations of these five variables are lower in balanced panel when compared to those in original data.

For FDI_Firm , the minimum value is zero and the maximum value is one both for original data and the balanced panel data, because this variable is a dummy variable. Further, the minimum value and the maximum value of variables FDI_Sector ,

137
Table 1

Descriptive statistics for the original data and the balanced panel data.

	Original data ^a				Balanced panel data			
	Min	Max	Mean	SD	Min	Max	Mean	SD
<i>Production Frontier</i>								
$\ln Y$	6.461	20.980	12.514	2.256	6.591	20.761	13.964	2.006
$\ln L$	2.398	10.649	4.079	1.327	2.639	10.292	4.702	1.088
$\ln K$	4.105	23.398	12.308	2.268	4.220	23.106	13.152	2.245
$\ln M$	3.871	20.033	11.765	2.418	4.239	19.454	12.164	2.221
$\ln E$	1.791	16.583	9.377	2.221	1.882	15.836	9.587	2.077
FDI_Sector	0	1.492	0.208	0.218	0	1.492	0.234	0.209
<i>Inefficiency Function</i>								
FDI_Firm	0	1	0.064	0.273	0	1	0.072	0.258
FDI_Sector	0	1.492	0.208	0.218	0	1.492	0.234	0.209
$FDI_Downstream_Sector$	0.002	5.443	0.176	0.212	0.002	5.443	0.176	0.204
$FDI_Upstream_Sector$	0	0.921	0.160	0.181	0	0.921	0.160	0.174
Number of Establishments	24,188	24,188	24,188	24,188	3318	3318	3318	3318
Number of Observation	231,064	231,064	231,064	231,064	43,225	43,225	43,225	43,225

Source: Authors' calculations from the annual surveys of the Indonesian Central Board of Statistics (*Badan Pusat Statistik* or BPS).

Y = output, L = labour, K = capital, M = material and E = energy.

^a The original data in this table exclude: (1) the establishments that do not report information on output, labour, capital, material, or energy; (2) 1.5 percent observations with the lowest values of output and 1.5 percent observations the highest values of output.

FDI_Backward, and *FDI_Forward* are the same for original data and for the balanced panel, as the calculation of these inter-industry variables is based on all firms in the original data as in [Blalock and Gertler \(2008\)](#). The mean values of these three spillover variables are higher in the balanced panel compared to those in the original data, whereas the standard deviations are lower in balanced panel. From the descriptive statistics in [Table 1](#), the authors conclude that there is no substantial bias in the adjustment process since there is no substantial difference in the maximum value, minimum value, mean value, and standard deviation.

3.3. Measurement of variables

There are two sets of variables included in this study: production variables and inefficiency variables. The production variables consist of output, labour, capital, material, energy, time trend and *FDI_Sector*, while the inefficiency variables include FDI variables (*FDI_Firm*, *FDI_Sector*, *FDI_Upstream_Sector*, and *FDI_Downstream_Sector*), a year dummy, an industry dummy, and a firm dummy. The precise definition of each variable is given in [Appendix 2](#).

In this study, gross output is used as the measure for output (y). It refers to the total value of output produced by a firm. The number of employees directly and indirectly engaged in production is used for the measure of labour (L). As a measure of capital (K), this study uses the replacement value of capital, while material (M) is measured using the total value of raw and intermediate materials and energy (E) is measured as the sum of electricity and fuel expenses.

FDI_Firm is measured by a dummy of foreign direct investment, which takes a value of one if a firm has a positive foreign ownership and takes a value of zero if otherwise. As a measure for the FDI horizontal spillovers, this study uses the share of foreign firm output to the total output at the five-digit ISIC sectoral level, which is expressed as in [Aitken and Harrison \(1999\)](#):

$$FDI_Sector_{jt} = \frac{\sum_{i \in j} FDI_Firm_{it} \times y_{it}}{\sum_{i \in j} y_{it}} \quad (9)$$

Eq. (9) captures the effect of FDI at the sectoral level on productivity at the firm level. It shows the spillover effects of foreign presence on domestic firms in the same five-digit ISIC industry.

Two alternative measures of FDI spillovers in this study are measures of inter-industry spillovers. The presence of foreign firms in certain five-digit ISIC industries may create productivity externalities for firms in upstream and downstream industries. This study measures the inter-industry spillovers by using variables that reflect the extent of backward and forward linkages between industries. Following [Javorcik \(2004\)](#), the measure for FDI spillovers from foreign firms in industries k ($k \neq j$) that are being supplied by domestic firms in industries j is:

$$FDI_Downstream_Sector_{jt} = \sum_{k | f^k \neq j} \alpha_{jk} \times FDI_Sector_{kt} \quad (10)$$

where α_{jk} is the proportion of sector j 's output supplied to sector k , which is taken from the input–output (IO) matrix of four-digit industries.⁴ Similarly, the measure for FDI spillovers from foreign firms in industries m whose products are bought by domestic firms in industries n is:

$$FDI_Upstream_Sector_{mt} = \sum_{n | f^n \neq m} \gamma_{mn} \times FDI_Sector_{nt} \quad (11)$$

where γ_{mn} is the proportion of inputs purchased by industry n from industry m in total input sourced by industry n , which is taken from the input–output (IO) matrix of four-digit industries.

A time-trend variable is incorporated in the production function to measure technical change. The time-trend variable takes a value of one for the year 1988, a value of two for the year 1989, and so on. An industry dummy captures effects specific to a particular industry and has a value of one for an industry for an observation of that industry and a value of zero otherwise. A similar procedure is also applied to the firm dummy and year dummy variables.

4. Empirical results

We estimate a stochastic frontier estimation and first test for constant returns to scale to check whether the Cobb–Douglas production frontier is best suited to the data. Following the procedure of joint restriction test in [Baltagi \(2011, p. 80\)](#), the test of constant returns to scale is conducted under the null hypothesis that the sum of the estimated parameters (β_i) in

⁴ During the selected period in this study, there are four available IO matrixes, which were published in 1990, 1993, 1995, and 2000. This study uses these four input–output matrixes for calculating the backward coefficient α_{jk} . The following is the procedure for obtaining values of α_{jk} . Values of α_{jk} before and including 1990 are taken from the 1990 IO matrix. Values of α_{jk} for 1991 and 1992 are linearly interpolated from the 1990 and 1993 IO matrixes. Values of α_{jk} for 1993 are taken from the 1993 IO matrix. Values of α_{jk} for 1994 are calculated from the linear interpolation of the 1993 and 1995 IO matrixes. Values of α_{jk} for 1995 are taken from the 1995 IO matrix. Values of α_{jk} from 1996 to 1999 are linearly interpolated from the 1995 and the 2000 IO matrixes. Finally, values of α_{jk} for 2000 are taken from the 2000 IO matrix.

production frontier in Eq. (5) is equal to one. The regression sum of squares for unrestricted model (RSS_U) is 39,631.63, whereas the regression sum of squared for restricted model (RSS_R) is 25,549.50. The F -statistics is 392.52, suggesting that the null hypothesis is rejected. This result confirms that the Cobb–Douglas production frontier is not the best suited model for the stochastic frontier estimation. Rather, as the sum of the coefficients of the input variables is greater than one, the unrestricted model with variable returns to scale is appropriate and is used below

4.1. Intra-industry spillovers

We begin with estimation of intra-industry spillovers. Using Eqs. (5) and (6), the production frontier and the inefficiency function are estimated simultaneously for observing the effects of foreign investment on the production frontier and technical efficiency of firms. For the inefficiency function, the technical efficiency variable (u_{it}) is specified as a function of a foreign share dummy (FDI_Firm), the share of foreign firms' outputs over total outputs in the four-digit industry (FDI_Sector), and an interacting term between FDI_Firm and FDI_Sector . When foreign investment increases the firm's technical efficiency, the coefficient of FDI_Firm is negative.⁵ When technology spills over from firms with foreign direct investment to purely domestic firms in the same industry, the coefficient of FDI_Sector is negative. As for the interaction term, the sign of the coefficient shows whether or not foreign direct investment affects the firm's ability to benefit from spillovers originating from other foreign-owned firms in the same industry.

We estimate four alternative models in order to test the robustness of the estimated parameters. In the first model, a year dummy and an industry dummy are included in the inefficiency equation. The estimated parameters are presented in the Model (1) column of Table 2. The results from the production frontier show that the four input variables contribute positively and significantly to output, suggesting a positive elasticity of each input on output. There is also a positive and statistically significant coefficient of the time-trend variable indicating that technical change contributes positively to output. The positive and statistically significant coefficient of FDI_Sector suggests horizontal spillovers from intra-industry foreign direct investment increase the production frontier for all firms.

From the estimates of the inefficiency function, which is the main focus of this study, the coefficient of FDI_Firm is negative and highly significant, indicating that foreign direct investment decreases the firm's technical inefficiency. This suggests that firms with foreign ownership are, on average, more efficient than purely domestic firms. This finding confirms the argument in Caves (1971) and Dunning (1988) that foreign firms are more likely to operate on the production frontier. Furthermore, the negative and statistically significant estimate of FDI_Sector suggests that knowledge spills over from foreign-owned firms increases the technical efficiency of all firms in the industry. This result is in line with the argument in Wang and Blomstrom (1992) and findings in Ghali and Rezgui (2008). This result is also consistent with findings in Takii (2005), Temenggung (2007) and Blalock and Gertler (2008), which use different methods of analysis.

The positive significant estimate of interacting term means that, although the foreign-owned firms also benefit from the presence of other foreign investment in the industry, the benefit is smaller than for domestic firms. Given that the estimated coefficient of FDI_Firm and the estimated coefficient of FDI_Sector are negative and statistically significant, the positive coefficient of the interaction term means that $u_{it}/FDI_Firm = -0.5763 + 0.0330 \times FDI_Sector$ and that $u_{it}/FDI_Sector = -0.2224 + 0.0330 \times FDI_Firm$. As both FDI_Firm and FDI_Sector are each always less than or equal to one by construction, the net effect of FDI_Sector is negative for all foreign firms as well as domestic firms. However, the magnitude of the improvement in efficiency from having foreign firms in the industry is always greater for domestic firms than for foreign firms.

In addition, we conduct joint significance test (F -test) on the magnitude of spillovers for foreign establishments in order to check significance of the direct effect and the interacting effect of spillovers on foreign firms.⁶ The value of F -statistic is calculated from the log-likelihood value of the unrestricted model and the log likelihood value of the restricted model (when both the coefficient of FDI_Sector and the coefficient of interacting variable $FDI_Firm \times FDI_Sector$ equal to zero). The value of log likelihood for the unrestricted model is 7704.48, whereas the value of log likelihood for the restricted model is 7643.00. So that, the F -statistic is 13.22, which suggests that the unrestricted model (by including variables FDI_Sector and interacting variable $FDI_Firm \times FDI_Sector$) is the correct model and the two variables are jointly significant affecting spillovers on foreign establishments at 1% level.

The estimated coefficient of year dummy is not statistically significant, suggesting that on average there is no significant difference in technical inefficiency scores of firms across the sample years. The statistically significant estimated coefficient of industry dummy suggests that there is a significant difference in inefficiency scores across five-digit industries.

The highly significant estimate of gamma implicates that estimation of stochastic frontier should include an inefficiency effect. This finding provides the justification for the simultaneous estimation of stochastic production frontier and inefficiency equation. In other words, the model is appropriately representing the observed firms.

In the second model, industry dummies are replaced by firm dummies, in order to control for firm heterogeneity across the sample. The results are given in the Model (2) column of Table 2. The sign and significance of estimates are similar to

⁵ The dependent variable for the inefficiency function is technical inefficiency. The negative coefficient of FDI_Firm indicates that foreign investment decreases inefficiency, which implies an increase in the firm's efficiency.

⁶ We are grateful to one of the reviewers for suggesting this point.

Table 2
Estimating intra-industry spillovers.

Variables	Model (1)	Model (2)	Model (3)	Model (4)
<i>Production frontier</i>				
<i>lnL</i>	0.2227*** (0.0033)	0.2256*** (0.0031)	0.2197*** (0.0030)	0.2167*** (0.0031)
<i>lnK</i>	0.1018*** (0.0019)	0.1043*** (0.0017)	0.1023*** (0.0018)	0.1097*** (0.0012)
<i>lnM</i>	0.6263*** (0.0018)	0.6218*** (0.0018)	0.6223*** (0.0017)	0.6191*** (0.0022)
<i>lnE</i>	0.1128*** (0.0017)	0.1160*** (0.0017)	0.1165*** (0.0017)	0.1176*** (0.0016)
<i>T</i>	0.0007* (0.0005)	0.0039* (0.0006)	0.0066*** (0.0028)	0.0012*** (0.0003)
<i>FDI_Sector</i>	0.1224*** (0.0055)	0.2044*** (0.0065)	0.2687*** (0.0096)	0.1577*** (0.0065)
<i>Inefficiency function</i>				
<i>FDI_Firm</i>	-0.5763*** (0.0264)	-0.1550*** (0.0018)	-0.1960*** (0.0104)	-0.2362*** (0.0092)
<i>FDI_Sector</i>	-0.2224*** (0.0896)	-0.2000*** (0.0149)	-0.1780*** (0.0027)	-0.1819*** (0.0034)
<i>FDI_Firm × FDI_Sector</i>	0.0330*** (0.0028)	0.0460*** (0.0036)	0.1035*** (0.0184)	0.0673*** (0.0086)
Year Dummy	-0.0002 (0.0031)	-0.0010 (0.0009)	-0.0010 (0.0019)	-
Industry Dummy	-0.0039* (0.0008)	-	-	-
Firm Dummy	-	-0.0001** (0.0000) ^a	-	-
Sigma-squared	0.0416*** (0.0010)	0.0416*** (0.0005)	0.0413*** (0.0003)	0.0418*** (0.0003)
Gamma	0.0380*** (0.0038)	0.0224*** (0.0083)	0.0086*** (0.0002)	0.0151*** (0.0020)
Log-likelihood	7704.484	7759.086	7618.974	7572.755
Number of Observations	43,225	43,225	43,225	43,225

Source: Authors' calculations.

Notes: *Y* = output, *L* = labour, *K* = capital, *M* = material, *E* = energy, *T* = time trend. Standard errors are in parentheses.

^a The estimated standard error is 0.000009.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

those in the first model. The notable difference is only in the magnitude of the estimates. Focusing on the FDI variables, the magnitudes of coefficients are smaller in this second model compared to those in the first model. In other words, the inclusion of firm dummy and the exclusion of industry dummy in the second estimation (Model 2) results in a smaller effect of FDI spillovers on technical inefficiency. This is not surprising. Firm-specific effects are largely captured by the firm dummy, which removes a potential source of bias in the estimates of other coefficients. Notably, the results regarding the direction of spillover effects are the same as in the first model, as the coefficient of *FDI_Sector* is negative and statistically significant in both models.

For the third model, only a time dummy is included as a controlling variable in the estimation. The resulting estimates, which are presented in the Model (3) column, are very comparable with the results in Model (1) and Model (2). Similar findings are also observed in Model (4), when the time dummy, industry dummy and firm dummy are all excluded from estimation. The results from these four models confirm the robustness of the estimates of the positive spillovers from FDI on the technical efficiency of domestic firms.

4.2. Inter-industry spillovers

Besides the effects on domestic firms in the same industry, FDI can also generate spillovers on domestic firms in other industries. We estimate six models of the inter-industry spillovers, and the results of each model are presented in Table 3. The first three models are estimated on the full sample and the last three models are estimated on the sub-sample of only domestic firms. In the three full-sample models, the first model is to capture the simultaneous effect of the three spillover variables on technical inefficiency. The second and the third model focus on the individual effect of each of the vertical FDI spillovers (i.e. the downstream spillover and the upstream spillover). The same structure is also applied to the sub-sample of only domestic firms, with Model 4 captures the simultaneous effect of the three spillover variables, Model 5 captures the downstream effect only, and model 6 captures only the upstream effect.

In the first model (the first results column of Table 3), the three proxies of spillover variables are included in the estimations. The results show that the horizontal spillover variable (*FDI_Sector*) has a negative and statistically significant coefficient, suggesting that an increase in the share of foreign firm output decreases technical inefficiency across firms in the industry. Similarly, the spillovers from FDI in downstream industries also decrease inefficiency of suppliers, as demonstrated by the negative and highly significant coefficient of the backward spillover variable (*FDI_Downstream_Sector*). In addition, the coefficient of the forward spillover variable (*FDI_Upstream_Sector*) is negative and highly significant, indicating a negative relationship between FDI in supplier industries and the industry's own technical inefficiency. Although we employ a different methodology and use a different data set, the findings are similar to those in Liang (2007).

In the second and the third models (the second and the third columns of Table 3), the impacts of backward spillover variable and the forward spillover variable are estimated separately. In each model, the magnitude of the coefficient of the included spillovers variable is larger than in Model 1, but neither the sign nor the statistical significance of the coefficient changes. Clearly, there is multi-collinearity among the spillovers variables that makes the identification of separate effects

Table 3
Estimating inter-industry spillovers.

Variables	Full sample (1)	Full sample (2)	Full sample (3)	Domestic sample (4)	Domestic sample (5)	Domestic sample (6)
<i>Production frontier</i>						
<i>lnL</i>	0.2264*** (0.0030)	0.2209*** (0.0030)	0.2197*** (0.0029)	0.2258*** (0.0012)	0.2238*** (0.0033)	0.2256*** (0.0033)
<i>lnK</i>	0.1007*** (0.0018)	0.1023*** (0.0018)	0.1019*** (0.0018)	0.0986*** (0.0018)	0.0999*** (0.0022)	0.0981*** (0.0019)
<i>lnM</i>	0.6255*** (0.0018)	0.6271*** (0.0018)	0.6268*** (0.0017)	0.6225*** (0.0014)	0.6236*** (0.0020)	0.6229*** (0.0017)
<i>lnE</i>	0.1117*** (0.0017)	0.1144*** (0.00170)	0.1159*** (0.0016)	0.1217*** (0.0014)	0.1226*** (0.0018)	0.1227*** (0.0018)
<i>T</i>	0.0002** (0.0000) ^a	0.0028* (0.0013)	0.0004*** (0.0001)	0.0009** (0.0006)	0.0021** (0.0001)	0.0010*** (0.0002)
<i>FDI_Sector</i>	0.0375*** (0.0013)	0.0308*** (0.0038)	0.0217*** (0.0007)	0.0056*** (0.0007)	0.0572*** (0.0035)	0.0323*** 0.0064
<i>Inefficiency function</i>						
<i>FDI_Firm</i>	-0.2945*** (0.0137)	-0.3920*** (0.0393)	-0.1257*** (0.0130)	-	-	-
<i>FDI_Sector</i>	-0.1901*** (0.0061)	-	-	-0.2766*** (0.0275)	-	-
<i>FDI_Downstream_Sector</i>	-0.0216*** (0.0021)	-0.0715*** (0.0043)	-	-0.0279*** (0.0047)	-0.0548*** (0.0027)	-
<i>FDI_Upstream_Sector</i>	-0.0462*** (0.0060)	-	-0.1842*** (0.0097)	-0.0682*** (0.0175)	-	-0.3067*** (0.0214)
<i>Year Dummy</i>	-0.0018* (0.0006)	-0.0050* (0.0017)	-0.0017* (0.0003)	0.0011*** (0.0002)	0.0046*** (0.0005)	0.0002*** (0.0010)
<i>Firm Dummy</i>	-0.0000 ^b (0.0000) ^c	-0.0000 ^d (0.0000) ^e	-0.0000 ^f (0.0000) ^g	-0.0001*** (0.0000) ^h	-0.0001*** (0.0000) ⁱ	-0.0001*** (0.0000) ^j
<i>Sigma-squared</i>	0.0401*** (0.0003)	0.0416*** (0.0003)	0.0405*** (0.0003)	0.0411*** (0.0007)	0.0418*** (0.0001)	0.0405*** (0.0004)
<i>Gamma</i>	0.0194*** (0.0013)	0.0417*** (0.0040)	0.0124*** (0.0008)	0.0612*** (0.0111)	0.0709*** (0.0019)	0.0561*** (0.0045)
<i>Log-likelihood</i>	7849.487	7668.081	7750.109	8118.497	8001.479	8040.274
<i>Number of Observations</i>	43,225	43,225	43,225	40,042	40,042	40,042

Source: Authors' calculations.

Notes: *Y* = output, *L* = labour, *K* = capital, *M* = material, *E* = energy, *T* = Time trend. Actual estimates are ^a 0.00004, ^b 0.000034, ^c 0.0000017, ^d 0.000034, ^e 0.0000019, ^f 0.000034, ^g 0.0000014, ^h 0.0000024, ⁱ 0.000012, ^j 0.0000035. Standard errors are in parentheses.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

Table 4
Estimating intra-industry spillovers in high-efficiency and low-efficiency firms.

Variables	Full sample		Domestic sample	
	High-efficiency firms (1)	Low-efficiency firms (2)	High-efficiency firms (3)	Low-efficiency firms (4)
<i>Production frontier</i>				
<i>lnL</i>	0.2049*** (0.0047)	0.2258*** (0.0040)	0.2372*** (0.0018)	0.2012*** (0.0038)
<i>lnK</i>	0.1080*** (0.0032)	0.0985*** (0.0024)	0.1025*** (0.0024)	0.0911*** (0.0021)
<i>lnM</i>	0.6038*** (0.0023)	0.6634*** (0.0027)	0.5883*** (0.0036)	0.6900*** (0.0026)
<i>lnE</i>	0.1316*** (0.0027)	0.0835*** (0.0023)	0.1429*** (0.0013)	0.0791*** (0.0018)
<i>T</i>	0.0021*** (0.0009)	0.0001*** (0.0000) ^b	0.0022*** (0.0004)	0.0054*** (0.0003)
<i>FDI_Sector</i>	0.0940*** (0.0058)	0.0492*** (0.0141)	0.0849*** (0.0032)	0.0727*** (0.0133)
<i>Inefficiency function</i>				
<i>FDI_Firm</i>	-0.0617*** (0.0088)	-0.0096* (0.0063)	-	-
<i>FDI_Sector</i>	0.0742*** (0.0062)	-0.0556*** (0.0035)	0.0657*** (0.0038)	-0.0660*** (0.0115)
Year Dummy	0.0020 (0.0014)	-0.0027*** (0.0007)	0.0029*** (0.0004)	0.0015*** (0.0001)
Firm Dummy	0.0001*** (0.0000) ^a	0.0001*** (0.0000) ^c	0.0001*** (0.0000) ^d	0.0000*** (0.0000) ^f
Sigma-squared	0.0425*** (0.0004)	0.0382*** (0.0004)	0.0414*** (0.0005)	0.0341*** (0.0006)
Gamma	0.0369*** (0.0043)	0.0151*** (0.0023)	0.0540*** (0.0036)	0.0746*** (0.0019)
Log-likelihood	3493.823	4697.164	3597.36	5417.533
Number of Observations	21,612	21,613	20,021	20,021

Source: Authors' calculations.

Notes: Y = output, L = labour, K = capital, M = material, E = energy and T = time trend Actual estimates are: ^a 0.0000042, ^b 0.000037 ^c 0.000005 ^d 0.0000076, ^e 0.000018, ^f 0.0000066. Standard errors are in parentheses.

* Significant at the 10% level.

** Significant at the 5% level.

*** Significant at the 1% level.

difficult. The coefficient of the *FDI_Downstream_Sector* being negative and statistically significant at the 1% level in both Model 1 and Model 2, indicates a robust finding that the foreign entry in a three-digit industry decreases the technical inefficiency of domestic suppliers (i.e. positive backward spillovers). Similarly, the negative and statistically significant coefficient of the *FDI_Upstream_Sector* in both Model 1 and Model 3 indicates a robust finding that the presence of foreign firms in a three-digit industry decreases the inefficiency of domestic buyers (i.e. positive forward spillovers).

To isolate the spillover effects on only domestic firms, we estimate the Models 1 through 3 for the sub-sample of only domestic firms. The estimation results are presented in the fourth through sixth result columns in Table 3. The results are similar to those for the full sample of firms in terms of the signs and significance of the coefficients. However, it is notable that the coefficients for the spillovers variables in the domestic firm sample are generally of larger magnitude than the corresponding coefficients for the full sample. This provides further evidence to support that from the results in Table 2 showing that spillovers from foreign firms are more beneficial for purely domestic owned firms than for firms with direct foreign investment.

Given the results from Table 3, we conclude that the spillover effects from FDI decrease technical inefficiency of domestic firms in upstream and downstream industries. These findings confirm the argument in Javorcik (2004) that a foreign presence in a domestic market may generate not only spillover effects on domestic firms in the same industry but also provide spillover benefits to domestic firms in the upstream and downstream industries.

4.3. Spillover effects and the level of technical efficiency

So far, the analysis pools together all firms with different levels of efficiency. It has advantage of showing the average effect of FDI spillovers on a firm's technical efficiency. However, it has a disadvantage in that the spillover effects are assumed to be uniform for all firms. Thus, the analysis does not clearly distinguish which firms gain the most spillover effect from FDI.

In this section, the analysis is extended to answer a question of whether the level of efficiency influences the ability of firms in absorbing spillover benefits. The firms are divided into two groups: firms with a high-efficiency level and those with a low-efficiency level. The procedure to group the firms is by sorting the firms from the one with the highest technical efficiency level to the firm with the lowest efficiency level, and then the sorted firms are divided into two. The upper half of the data is categorized as the high-efficiency firms and the lower half is the low-efficiency firms. The estimation results for these two groups of firms are presented in Table 4. We estimate results for the full sample of firms as well as for the sub-sample of only domestic firms.

Starting from the full sample estimations, the coefficient of *FDI_Firm* is negative and statistically significant both among high-efficiency firms (column 1 of Table 4) and among low-efficiency firms (column 2), suggesting that foreign-owned firms have a higher technical efficiency level in both groups of firms. The positive and significant coefficient of *FDI_Sector* demonstrates that spillovers at the industrial level increase the inefficiency of the firms (i.e. a negative efficiency spillover). In contrast, the low-efficiency firms experience a decrease in technical inefficiency when foreign firms are more important in the industry (i.e. a positive efficiency spillover), as indicated by a negative and highly significant coefficient of *FDI_Sector* (column 2).

The coefficients of *FDI_Sector* for the sub-sample of only domestic firms (columns 3 and 4) are of the same sign and significance as in the corresponding full sample estimation, but the magnitude of impact is somewhat lower in the domestic firm sub-sample. This suggests that FDI spillovers have smaller impact on domestic firms than on foreign firms in industries with large technology gaps.

The results in Table 4 demonstrate that firms with different efficiency levels may receive different effects of FDI spillovers. High-efficiency firms tend to obtain negative spillover effects, while low-efficiency firms experience positive spillover effects. These findings confirm the argument that there is advantage from being less advanced in terms of efficiency in terms of benefitting from spillovers (Glass & Saggi, 1998; Wang & Blomstrom, 1992) and are consistent with the results in Griffith et al. (2002), Castellani and Zanfei (2003), and Peri and Urban (2006).

5. Conclusion

This article empirically examines the spillover effects of FDI on firm technical efficiency in the Indonesian manufacturing sector for the period between 1988 and 2000. Using the framework of Battese and Coelli's (1995) stochastic production frontier, we find evidence of a positive spillover effect of FDI to firms in the same industry (competitors), firms in an upstream industry (suppliers), and firms in a downstream industry (buyers). The positive spillover effect is observed in both the estimation for the full sample of firms and the estimation for the sub-sample of only domestic firms. Notably, the effects on domestic firms are generally more powerful than on other foreign firms in the same industry.

An interesting finding emerges when the samples are divided into two groups based on the level of efficiency. It is found that the low-efficiency firms receive a positive spillover effect from FDI across firms in the same industry. In contrast, the high-efficiency firms obtain a negative spillover effect. These findings support the argument of the advantage for absorbing spillovers goes to firms that are less advanced in terms of efficiency.

Outcomes from this study provide support on policies that encourage FDI. On the basis of these findings, policy makers should continue providing an FDI-friendly environment in order to maximize the spillover gains. Additional incentives may be provided for foreign firms that are willing to transfer their knowledge to domestic firms, especially those domestic firms in upstream and downstream industries that do not directly compete with the foreign firm. Variations in incentives may need to be considered, with more focus on FDI in sectors where purely domestic firms have a low-efficiency level compared to firms with direct foreign investment.

Acknowledgements

Helpful comments from two anonymous referees, editor Professor Michael Plummer and associate editor Professor Frank Hsiao, are gratefully acknowledged, but the authors are responsible for any remaining errors or omissions.

Appendix 1. Adjustments for constructing a consistent panel data

The steps of adjustment for constructing a consistent panel data are described as follows:

Step 1: Adjustment for industrial code.

The BPS reclassified the industrial codes twice: in 1990 and in 1998. This study adjusts the industrial codes to the 1990 code (KKI-1990) in order to obtain a consistent industrial code for the observation years (1988–2000). This adjustment involves two phases. First, the data from 1988 to 1989 (which use KKI-1985) are adjusted to KKI-1990 using the establishment identification code and a special map provided by the BPS. Observations in 1988–1989 not observed in 1990–1998 are removed, since there is no code from KKI-1990 that could be assigned to these observations. This first phase of adjustment removes 1345 out of the original 29,340 establishments. Second, the data from 1998 to 2000 (which use KKI-1998) are adjusted to KKI-1990 by the following concordance table provided by the BPS. There are several concordance issues that arise during this second phase of adjustment, which include unmatched classifications and incomplete entries. An example of an incomplete entry is an observation recorded only with a two-, three-, or four-digit classification code. For dealing with this problem, only observations with four-digit classification codes are retained, while those with two- and three-digit classification codes are removed. The retained observations with four-digit codes are then assigned as five-digit codes using the establishment specific identification code. By doing so, all establishments in the 1988–2000 panel data have consistent and integrated classification codes. The total establishments removed after these industrial code adjustments are 3078 out of 29,340 establishments, which include those with Oil and Gas classification (ISIC 353 and 354) as these sub-sectors are not observed in the 1988 and 1989 surveys.

Step 2: Adjustment for the variable definitions.

In some years, the variable definitions provided by the BPS are not consistent, even though the variables are the same. The author compared the variable definitions in each year's survey questionnaires (which are provided by the BPS together with the SI data) and recalculated the inconsistent variables for obtaining consistent definitions throughout the selected period.

Step 3: Cleaning for noise and typographical errors.

This study applied several steps for data cleaning in order to minimize noises and typographical errors:

- Observations with zero or a negative value of output, labour, material, or energy have been removed. This removes around 4.5 percent of the total observations.
- If a firm reports a missing value for a particular variable in a given time but reports values in the year before and after, an interpolation is carried out to fill the gap. The interpolation for the missing data was not more than 1 percent of the total observations.
- Typographical errors (or key-punch errors) in the raw data are adjusted for consistency. For example, if in the raw data, foreign share in a firm for the whole of the selected period was typed as 100 percent, except for a certain year being typed as 0 percent, then the 0 percent share is adjusted to 100 percent.
- Observations that are considered as outliers are removed from the data set by following a procedure suggested by Takii (2005). First, observations are sorted from the lowest to the highest value of output. Second, 1.5 percent of the lowest values and 1.5 percent of the highest values are removed.

Step 4: Back-casting the missing values of capital.

In some years, the values of capital are missing for quite a large number of observations. To fill these gaps, this study follows the methodology introduced by Vial (2006).

Step 5: Matching firms for a balanced panel

A balanced panel data set is constructed for the selected period by matching firms based on the specific identification code (PSID). This study utilizes STATA10 software for the matching.

Step 6: Choosing Industries with Foreign Firms

Since the purpose of the study is to estimate the FDI spillovers, industries (at a five-digit level) without foreign firms are excluded from the balanced panel.

Step 7: All monetary variables (output, capital, material, and energy) are deflated using price indexes. The output and material values are deflated using the wholesale price index (for four-digit ISIC industries); the machinery price index is used for deflating the value of capital; the nominal values of energy, which are a sum of electricity and fuel expenditures, are deflated using the electricity price index and the fuel price index. All price indexes are at a constant price of the year 1993.

By following the steps of adjustment, the final balance panel data consists of 3318 establishments with 43,225 observations.

Appendix 2. Definitions of variables

Symbol	Category	Unit	Definitions
<i>Production frontier</i>			
<i>Y</i>	Output	Million of 1993 rupiah	Gross output, which is deflated using a wholesale price index of four-digit ISIC industries at a constant price of 1993
<i>L</i>	Labour	Number of workers	Total number of employees directly and indirectly engaged in production, which covers all workers, including technical, administration, marketing, storage, and clerical staffs, who work full-time or part-time, and also family members.
<i>K</i>	Capital	Million of 1993 rupiah	Replacement value of fixed assets, which is deflated using a wholesale price index for machinery of four-digit ISIC industries at a constant price of 1993.
<i>M</i>	Material	Million of 1993 rupiah	Total value of material used in production, which cover raw and intermediate materials, both domestically produced and imported deflated using a wholesale price index of four-digit ISIC industries at a constant price of 1993.
<i>E</i>	Energy	Million of 1993 rupiah	Total value of electricity and fuel used by a firm. The value of electricity is calculated from the electricity provided by the state energy company (<i>Perusahaan Listrik Negara</i> or PLN) and those provided by private power firms, and it is deflated using the wholesale electricity index at a constant price of 1993. The value of fuels are calculated from nine types of fuels, namely premium, solar, kerosene, coal, cokes, gas, firewood, lubricant, and other fuels, and it is deflated using the OECD price of fuels published by <i>DX for Windows</i> at the 1993 constant price.
<i>T</i>	Time trend		Take a value of one for 1988, value of two for 1989, and so on.
<i>FDI_Sector</i>	FDI Variable	Ratio	The share of foreign firms' output over total outputs in a five-digit industry, or can be expressed as in Eq. (5). This variable measures the intra-industry (or horizontal) spillovers.
<i>Inefficiency function</i>			
<i>FDI_Firm</i>	FDI variable	Binary (one or zero)	The FDI at the firm level, which takes a value of one if a firm has a positive foreign ownership and take a value of zero if otherwise.
<i>FDI_Sector</i>	FDI variable	Ratio	The share of foreign firms' output over total outputs in a five-digit industry, or can be expressed as in Eq. (5). This variable measures the intra-industry (or horizontal) spillovers.
<i>FDI_Downstream_Sector</i>	FDI variable	Ratio	Spillovers from foreign firms in industries k ($k \neq j$) that are being supplied by domestic firms in industries j is defined as in Eq. (6).

Appendix 1 (Continued)

Symbol	Category	Unit	Definitions
<i>FDI_Upstream_Sector</i>	FDI variable	Ratio	Spillovers from foreign firms in industries m ($m \neq n$) that sell their outputs to domestic firms in industries n is defined as in Eq. (6).
<i>Year</i>	Dummy variable		A year dummy, which takes a value of one for all observations for the year in question, and a value of zero for other years.
<i>Industry</i>	Dummy variable		An industry dummy, which has a value of one for all observations for the industry in question and a value of zero for other industries.
<i>Firm</i>	Dummy variable		A firm dummy, which has a value of one for all observations for the firm in question and a value of zero for every other.

References

- Aigner, D. J., Lovell, C. A. K., & Schmidt, P. (1977). Formulation and estimation of stochastic frontier production function models. *Journal of Econometrics*, 6(1), 21–37.
- Aitken, B. J., & Harrison, A. E. (1999). Do Domestic Firms Benefit from Direct Foreign Investment? Evidence from Venezuela. *The American Economic Review*, 89(3), 605–618.
- Baltagi, B. D. (2011). *Econometrics* (5th edition). Heidelberg: Springer.
- Banker, R. D., Charnes, A., & Cooper, W. W. (1984). Some models for estimating technical and scale inefficiency in data envelopment analysis. *Management Science*, 30(9), 1078–1092.
- Battese, G., & Coelli, T. J. (1988). Prediction of firm-level technical efficiencies with a generalized frontier production function and panel data. *Journal of Econometrics*, 38(3), 387–399.
- Battese, G. E., & Coelli, T. J. (1993). A Stochastic Frontier Production Function Incorporating a Model for Technical Inefficiency Effects. *Working Paper in Econometrics and Applied Statistics, Department of Economics University of New England*.
- Battese, G. E., & Coelli, T. J. (1995). A model for technical inefficiency effects in a stochastic frontier production function for panel data. *Empirical Economics*, 20(2), 325–332.
- Bauer, P. W. (1990). Recent developments in the econometric estimation of frontiers. *Journal of Econometrics*, 46(1–2), 39–56.
- Bjurek H. L., Hjarmarsson, L., & Forsund, F. R. (1990). Deterministic parametric and nonparametric estimation in service production. *Journal of Econometrics*, 46(1–2), 213–227.
- Blalock, G., & Gertler, P. J. (2008). Welfare gain from foreign direct investment through technology transfer to local suppliers. *Journal of International Economics*, 74(2), 402–421.
- Blomstrom, M., & Kokko, A. (1998). Multinational corporation and spillovers. *Journal of Economic Surveys*, 12(2), 247–277.
- Bravo-Ureta, B. E., & Pinheiro, A. E. (1993). Efficiency analysis of developing country agriculture: A review of the frontier function literature. *Agricultural and Resource Economics Review*, 22(1), 88–101.
- Castellani, D., & Zanfei, A. (2003). Technology gaps, absorptive capacity and the impact of inward investments on productivity of European firms. *Economics of Innovation and New Technology*, 12(6), 555–576.
- Caves, R. E. (1971). International corporations: The industrial economics of foreign investment. *Economica*, 38(149), 1–27.
- Caves, R. E. (1974). Multinational firms, competition and productivity in host country markets. *Economica*, 41(162), 176–193.
- Central Bank of Indonesia (2011). *Economic and financial data for Indonesia*.
- Chakraborty, C., & Nunnenkamp, P. (2008). Economic reforms, FDI, and economic growth in India: A sector level analysis. *World Development*, 36(7), 1192–1212.
- Charnes, A., Cooper, W. W., & Rhodes, E. (1978). Measuring the efficiency of decision making units. *European Journal of Operational Research*, 2(6), 429–444.
- Coelli, T. J. (1995). Recent developments in frontier modelling and efficiency measurement. *Australian Journal of Agricultural Economics*, 39(3), 219–245.
- Coelli, T. J. (1996). A guide to frontier version 4.1, a computer program for stochastic frontier production and cost function estimation. *CEPA Working Paper No. 07/96 University of New England*.
- Coelli, T. J., Rao, D. S. P., O'Donnell, C. J., & Battese, G. E. (2005). *An introduction to efficiency and productivity analysis* (2nd ed.). New York: Springer.
- Dimelis, S., & Lauri, H. (2002). Foreign direct investment and efficiency benefits: A conditional quartile analysis. *Oxford Economic Papers*, 54(3), 449–469.
- Djankov, S., & Hoekman, B. (2000). Foreign investment and productivity growth in Czech enterprises. *World Bank Economic Review*, 14(1), 49–64.
- Dunning, J. (1988). *Multinational technology and competitiveness*. London: Allen & Unwin.
- Findlay, R. (1978). Relative backwardness, direct foreign investment, and the transfer of technology: A simple dynamic model. *Quarterly Journal of Economics*, 92(1), 1–16.
- Forsund, F. R. C. A. K., Lovell, P., & Schmidt (1980). A survey of frontier production functions and of their relationship to efficiency measurement. *Journal of Econometrics*, 13(1), 5–25.
- Ghali, S., & Rezzoui, S. (2008). FDI contribution to technical efficiency in the Tunisian manufacturing sector. *ERF Working Paper Series No. 421*.
- Glass, A., & Saggi, K. (1998). International technology transfer and the technology gap. *Journal of Development Economics*, 55(2), 369–398.
- Gorg, H., & Greenaway, D. (2004). Much ado about nothing? Do domestic firms really benefit from foreign direct investment?. *The World Bank Research Observer*, 19(2), 171–197.
- Gorg, H., & Strobl, E. (2005). Spillovers from foreign firms through worker mobility: An empirical investigation. *Scandinavian Journal of Economics*, 107(4), 693–739.
- Greene, W. H. (1993). In H. O. Fried, C. A. K. Lovell, & S. S. Schmidt (Eds.), *The econometric approach to efficiency analysis. The measurement of productive efficiency: Techniques and applications*. New York: Oxford University Press.
- Griffith, R., Redding, S., & Simpson, H. (2002). Productivity Convergence and Foreign Ownership at the Establishment Level. *Institute Fiscal Studies Working Paper 22, London*.
- Haddad, M., & Harrison, A. E. (1993). Are there positive spillovers from foreign direct investment? Evidence from panel data for Morocco. *Journal of Development Economics*, 42(1), 51–74.
- Hu, A. G. Z., & Jefferson, G. H. (2002). FDI impact and spillover: Evidence from china's electronic and textile industries. *The World Economy*, 25(8), 1063–1076.
- Hyer, S. H. (1960). (PhD dissertation). In *The international operations of national firms: A study of direct foreign investment* (p. 1976). MIT, MA: MIT Press.
- Javorcik, B. S. (2004). Does foreign direct investment increase the productivity of domestic firms? In search of spillovers through backward linkages. *American Economic Review*, 94(3), 605–627.
- Kugler, M. (2006). Spillovers from foreign direct investment: Within or between industries? *Journal of Development Economics*, 80(2), 444–477.
- Kumbhakar, S. C., Ghosh, S., & McGuckin, J. T. (1991). A generalized production frontier approach for estimating determinants of inefficiency in US dairy farms. *Journal of Business and Economic Statistics*, 9(3), 279–286.
- Liang, F. H. (2007). Does foreign direct investment improve the productivities of domestic firms? Technology spillovers within and between industries. *Haas Berkeley Working Paper*. <http://www.faculty.haas.berkeley.edu/fenliang/research/spillover/FDIspillover.pdf> [accessed 20.07.07].
- Lipsey, R. E., & Sjöholm, F. (2005). In T. H. Moran, E. Graham, & M. Blomstrom (Eds.), *The impact of inward FDI on host countries: Why such different answers? does foreign direct investment promote development?* (pp. 23–43). Washington, DC: Institute for International Economics and Center for Global Development.
- Lovell, C. A. K. (1993). In H. O. Fried, C. A. K. Lovell, & S. S. Schmidt (Eds.), *Production frontiers and productive efficiency. The measurement of productive efficiency: Techniques and applications*. New York: Oxford University Press.

- Meeusen, W., & van den Broeck, J. (1977). Efficiency estimation from cobb-douglas production function with composed error. *International Economic Review*, 18(2), 435–444.
- Negara, S. D., & Firdausy, C. M. (2011). In C. Sussangkarn, Y. C. Park, & S. J. Kang (Eds.), *The development of foreign direct investment and its impact on firms' productivity, employment and exports in Indonesia. Foreign direct investments in Asia*. London, UK: Routledge.
- Olesen, O. B., Peterson, N. C., & Lovell, C. A. K. (1996). Editors' introduction. *Journal of Productivity Analysis*, 7(2/3), 87–98.
- Peri, G., & Urban, D. (2006). Catching up to foreign technology? Evidence on the 'Veblen-Gerschenkron' effect of foreign investments. *Regional Science and Urban Economics*, 36(1), 72–98.
- Schiff, M., & Wang, Y. (2008). North-South and South-South trade-related technology diffusion: how important are they in improving TFP growth? *Journal of Development Studies*, 44(1), 49–59.
- Schmidt, P. (1985). Production frontier functions. *Econometric Reviews*, 4(2), 289–328.
- Smeets, R. A. (2008). Collecting the pieces of the FDI knowledge spillovers puzzle. *The World Bank Research Observer*, 23(2), 107–138.
- Suyanto, Salim, R., & Bloch, H. (2009). Does foreign direct investment lead to productivity spillovers? Firm level evidence from Indonesia. *World Development*, 37(12), 1861–1877.
- Suyanto, Bloch, H., & Salim, R. (2012). FDI spillovers and productivity growth in Indonesian garment and electronics manufacturing. *Journal of Development Studies* (in press).
- Takii, S. (2005). Productivity spillovers and characteristics of foreign multinational plants in Indonesian manufacturing 1990–1995. *Journal of Development Economics*, 76(2), 521–542.
- Takii, P. (2011). Do FDI spillovers vary among home economies? Evidence from Indonesian manufacturing. *Journal of Asian Economics*, 22(2), 152–163.
- Temenggung, D. (2007) Productivity spillovers from foreign direct investment: Indonesian manufacturing industry's experience 1975–2000, mimeograph, Australian National University, Canberra, Australia.
- Vial, V. (2006). New estimates of total factor productivity growth in Indonesian manufacturing. *Bulletin of Indonesian Economic Studies*, 42(3), 357–369.
- Wang, J. W., & Blomstrom, M. (1992). Foreign investment and technology transfer: A simple model. *European Economic Review*, 36(1), 137–155.
- Wang, H. J., & Schmidt, P. (2002). One-step and two-step estimation of the effects of exogenous variables on technical efficiency level. *Journal of Productivity Analysis*, 18(2), 129–144.

Which Firm Benefit From Foreign Direct Investment? Empirical Evidence From Indonesian Manufacturing

GRADEMARK REPORT

FINAL GRADE

/100

GENERAL COMMENTS

Instructor

PAGE 1

PAGE 2

PAGE 3

PAGE 4

PAGE 5

PAGE 6

PAGE 7

PAGE 8

PAGE 9

PAGE 10

PAGE 11

PAGE 12

PAGE 13

PAGE 14

Which Firm Benefit From Foreign Direct Investment? Empirical Evidence From Indonesian Manufacturing

ORIGINALITY REPORT

35%

SIMILARITY INDEX

27%

INTERNET SOURCES

27%

PUBLICATIONS

18%

STUDENT PAPERS

PRIMARY SOURCES

1	onlinelibrary.wiley.com Internet Source	1%
2	www.lib.kobe-u.ac.jp Internet Source	1%
3	www.chinaglobaltrade.com Internet Source	1%
4	Submitted to Curtin University of Technology Student Paper	1%
5	Submitted to University of Economics Ho Chi Minh Student Paper	1%
6	web.cenet.org.cn Internet Source	1%
7	www-wds.worldbank.org Internet Source	1%
8	aut.researchgateway.ac.nz Internet Source	1%
9	Submitted to Utah Valley State College Student Paper	1%

10	repository.usfca.edu Internet Source	1%
11	Dyah Wulan Sari, Noor Aini Khalifah, Suyanto Suyanto. "The spillover effects of foreign direct investment on the firms' productivity performances", Journal of Productivity Analysis, 2016 Publication	1%
12	www.york.ac.uk Internet Source	1%
13	eprints.port.ac.uk Internet Source	1%
14	Submitted to University of Nottingham Student Paper	1%
15	dspace.lboro.ac.uk Internet Source	<1%
16	ageconsearch.umn.edu Internet Source	<1%
17	Submitted to Laureate Higher Education Group Student Paper	<1%
18	www.icsead.or.jp Internet Source	<1%
19	Abdul Wadud, Ben White. "Farm household efficiency in Bangladesh: a comparison of stochastic frontier and DEA methods", Applied Economics, 10/10/2000	<1%

20

sareb-journal.org

Internet Source

<1 %

21

Submitted to University of Glamorgan

Student Paper

<1 %

22

Submitted to University of Liverpool

Student Paper

<1 %

23

John Ruggiero. "A comparison of DEA and the stochastic frontier model using panel data", International Transactions in Operational Research, 5/2007

Publication

<1 %

24

Liu, Z.. "Foreign direct investment and technology spillovers: Theory and evidence", Journal of Development Economics, 200802

Publication

<1 %

25

aaltodoc.aalto.fi

Internet Source

<1 %

26

Xiaolan Fu. "Foreign Direct Investment and Managerial Knowledge Spillovers through the Diffusion of Management Practices : Managerial Knowledge Spillovers from FDI", Journal of Management Studies, 03/2012

Publication

<1 %

27

www.etui.org

Internet Source

<1 %

28

Salim, Ruhul A., Kamrul Hassan, and Sahar

Shafiei. "Renewable and non-renewable energy consumption and economic activities: Further evidence from OECD countries", *Energy Economics*, 2014.

Publication

<1%

29

Widodo, Wahyu, Ruhul Salim, and Harry Bloch. "The effects of agglomeration economies on technical efficiency of manufacturing firms: evidence from Indonesia", *Applied Economics*, 2015.

Publication

<1%

30

econstor.eu

Internet Source

<1%

31

Yu, C.. "The effects of exogenous variables in efficiency measurement-A monte carlo study", *European Journal of Operational Research*, 19980316

Publication

<1%

32

Submitted to University of Sydney

Student Paper

<1%

33

SOPHIA DIMELIS. "Spillovers from foreign direct investment and firm growth: technological, financial and market structure effects", *International Journal of the Economics of Business*, 2/1/2005

Publication

<1%

34

Chen, C.-F.. "Applying the stochastic frontier approach to measure hotel managerial

<1%

efficiency in Taiwan", Tourism Management, 200706

Publication

35

Laborda Castillo, Leopoldo, Daniel Sotelsek Salem, and Justo de Jorge Moreno. "Foreign Direct Investment and Productivity Spillovers: Firm-Level Evidence From Chilean Industrial Sector", Latin American Business Review, 2014.

Publication

<1 %

36

Submitted to Massey University

Student Paper

<1 %

37

www.macrothink.org

Internet Source

<1 %

38

mediatum.ub.tum.de

Internet Source

<1 %

39

theses.gla.ac.uk

Internet Source

<1 %

40

link.springer.com

Internet Source

<1 %

41

Beata Smarzynska Javorcik. "Does Foreign Direct Investment Increase the Productivity of Domestic Firms? In Search of Spillovers Through Backward Linkages", The American Economic Review, 06/2004

Publication

<1 %

42

Doan, Tinh, David Maré, and Kris Iyer.

<1 %

"Productivity spillovers from foreign direct investment in New Zealand", New Zealand Economic Papers, 2014.

Publication

43 Joshua Abor. "Foreign direct investment and firm productivity: evidence from firm-level data", Global Business and Economics Review, 2010

Publication

44 erf.org.eg <1 %

Internet Source

45 ir.nul.nagoya-u.ac.jp <1 %

Internet Source

46 researchonline.jcu.edu.au <1 %

Internet Source

47 www.ppsi.or.th <1 %

Internet Source

48 mro.massey.ac.nz <1 %

Internet Source

49 www.lahoreschoolofeconomics.edu.pk <1 %

Internet Source

50 www.wz.uw.edu.pl <1 %

Internet Source

51 Sirianni, Philip O'Hara, Michael. "Do actions speak as loud as words? commitments to "going green" on campus.(Report)", Contemporary Economic Policy, April 2014

Issue

Publication

52

74.220.207.160

Internet Source

<1%

53

Submitted to Kyungsung University

Student Paper

<1%

54

nordicom.statsbiblioteket.dk

Internet Source

<1%

55

www.suomenpankki.fi

Internet Source

<1%

56

www.australianwaterrecycling.com.au

Internet Source

<1%

57

d-nb.info

Internet Source

<1%

58

Yanrui Wu. "Technical Efficiency and Firm Attributes in the Chinese Iron and Steel Industry", International Review of Applied Economics, 5/1996

Publication

<1%

59

bora.nhh.no

Internet Source

<1%

60

Nitin Arora, Preeti Lohani. "Does foreign direct investment spillover total factor productivity growth? A study of Indian drugs and pharmaceutical industry", Benchmarking: An International Journal, 2017

Publication

<1%

61	om.ciheam.org Internet Source	<1 %
62	etda.libraries.psu.edu Internet Source	<1 %
63	www.researchgate.net Internet Source	<1 %
64	eprints.utas.edu.au Internet Source	<1 %
65	www.eria.org Internet Source	<1 %
66	eiit.org Internet Source	<1 %
67	Indian Growth and Development Review, Volume 5, Issue 2 (2012-09-29) Publication	<1 %
68	www.handels.gu.se Internet Source	<1 %
69	etd.uovs.ac.za Internet Source	<1 %
70	Xinpeng Xu, Yu Sheng. "Are FDI spillovers regional? Firm-level evidence from China", Journal of Asian Economics, 2012 Publication	<1 %
71	Nourzad, Farrokh. "Openness and the efficiency of FDI: a panel stochastic production frontier study.(Foreign Portfolio ",	<1 %

72

SUYANTO. "SOURCES OF PRODUCTIVITY GAINS FROM FDI IN INDONESIA: IS IT EFFICIENCY IMPROVEMENT OR TECHNOLOGICAL PROGRESS? :

productivity gains from fdi in indonesia", The Developing Economies, 12/2010

Publication

<1 %

73

Submitted to Monash University

Student Paper

<1 %

74

publications.polymtl.ca

Internet Source

<1 %

75

thesis.lib.nccu.edu.tw

Internet Source

<1 %

76

aria.org

Internet Source

<1 %

77

Miranti, Riyana, Alan Duncan, and Rebecca Cassells. "Revisiting the Impact of Consumption Growth and Inequality on Poverty in Indonesia during Decentralisation", Bulletin of Indonesian Economic Studies, 2014.

Publication

<1 %

78

Submitted to University of Bradford

Student Paper

<1 %

79

Alfons Oude Lansink. "Technical efficiency and CO₂ abatement policies in the Dutch glasshouse industry", *Agricultural Economics*, 2/2003

Publication

<1 %

80

Sabirianova Peter, Klara, Jan Svejnar, and Katherine Terrell. "Foreign Investment, Corporate Ownership, and Development: Are Firms in Emerging Markets Catching Up to the World Standard", *Review of Economics and Statistics*, 2012.

Publication

<1 %

81

Odd Bjarte Nilsen. "LEARNING-BY-DOING OR TECHNOLOGICAL LEAPFROGGING: PRODUCTION FRONTIERS AND EFFICIENCY MEASUREMENT IN NORWEGIAN SALMON AQUACULTURE", *Aquaculture Economics & Management*, 04/2010

Publication

<1 %

82

Submitted to Jacobs University, Bremen

Student Paper

<1 %

83

www.journalofscience.org

Internet Source

<1 %

84

www.oecd.org

Internet Source

<1 %

85

Meditari Accountancy Research, Volume 23, Issue 3 (2015)

<1 %

86 Suyanto, , and Ruhul Salim. "Foreign direct investment spillovers and technical efficiency in the Indonesian pharmaceutical sector: firm level evidence", Applied Economics, 2013. <1 %

Publication

87 boris.unibe.ch <1 %
Internet Source

88 mpra.ub.uni-muenchen.de <1 %
Internet Source

89 www.diva-portal.org <1 %
Internet Source

90 Submitted to CSU, San Jose State University <1 %
Student Paper

91 Marin, Anabel, and Martin Bell. "Technology spillovers from Foreign Direct Investment (FDI): the active role of MNC subsidiaries in Argentina in the 1990s", The Journal of Development Studies, 2006. <1 %

Publication

92 edoc.ub.uni-muenchen.de <1 %
Internet Source

93 d-scholarship.pitt.edu <1 %
Internet Source

94 www.adeimf.it <1 %
Internet Source

95	"Multinationals and Foreign Investment in Economic Development", Springer Nature, 2005 Publication	<1%
96	www.eta.libraries.psu.edu Internet Source	<1%
97	ajbe.info Internet Source	<1%
98	Leeds, Daniel M. Leeds, Michael A. Motom. "Are sunk costs irrelevant? Evidence from playing time in the National Basketball Association.", Economic Inquiry, April 2015 Issue Publication	<1%
99	Mohamad Ikhsan. "Total Factor Productivity Growth in Indonesian Manufacturing: A Stochastic Frontier Approach", Global Economic Review, 12/2007 Publication	<1%
100	wageindicator.org Internet Source	<1%
101	www.bath.ac.uk Internet Source	<1%
102	era.library.ualberta.ca Internet Source	<1%
103	imawesa.info Internet Source	<1%

104	Gerschewski, Stephan. "Do Local Firms Benefit from Foreign Direct Investment? An Analysis of Spillover Effects in Developing Countries", Asian Social Science, 2013. Publication	<1 %
105	repub.eur.nl Internet Source	<1 %
106	www.rsis.edu.sg Internet Source	<1 %
107	staff.cbs.dk Internet Source	<1 %
108	Lin, Yi Hsing, and Chun Fu Hong. "Estimating production technical efficiency for the US biopharmaceutical industry", International Journal of Biotechnology, 2012. Publication	<1 %
109	etheses.bham.ac.uk Internet Source	<1 %
110	etheses.whiterose.ac.uk Internet Source	<1 %
111	Climate Change Policies and Challenges in Indonesia, 2016. Publication	<1 %
112	dro.dur.ac.uk Internet Source	<1 %
113	chaire-eppp.org Internet Source	<1 %

114	eprints.soas.ac.uk Internet Source	<1 %
115	Wang, Yanling. "Exposure to FDI and new plant survival: evidence in Canada :", Canadian Journal of Economics/Revue canadienne d'économique, 2013. Publication	<1 %
116	www.fafo.no Internet Source	<1 %
117	ieeexplore.ieee.org Internet Source	<1 %
118	www.econstor.eu Internet Source	<1 %
119	Submitted to Bloomsbury Colleges Student Paper	<1 %
120	Lawson, L.G.. "Lameness, metabolic and digestive disorders, and technical efficiency in Danish dairy herds: a stochastic frontier production function approach", Livestock Production Science, 20041201 Publication	<1 %
121	geb.uni-giessen.de Internet Source	<1 %
122	www.merit.unu.edu Internet Source	<1 %
123	business.uow.edu.au Internet Source	<1 %

124	Hanson, G.H.. "Regional adjustment to trade liberalization", Regional Science and Urban Economics, 19980701 Publication	<1 %
125	www.igidr.ac.in Internet Source	<1 %
126	www.adbi.org Internet Source	<1 %
127	www.cepr.org Internet Source	<1 %
128	Demena, Binyam A., and Peter A. G. van Bergeijk. "A META-ANALYSIS OF FDI AND PRODUCTIVITY SPILLOVERS IN DEVELOPING COUNTRIES : A META-ANALYSIS OF FDI AND PRODUCTIVITY SPILLOVERS", Journal of Economic Surveys, 2016. Publication	<1 %
129	R. Jayasuriya. "Measuring and Explaining the Impact of Productive Efficiency on Economic Development", The World Bank Economic Review, 05/18/2005 Publication	<1 %
130	eprints-phd.biblio.unitn.it Internet Source	<1 %
131	eng.kea.ne.kr Internet Source	<1 %

132	ediss.uni-goettingen.de Internet Source	<1 %
133	deazone.com Internet Source	<1 %
134	ftp.iza.org Internet Source	<1 %
135	www.druid.dk Internet Source	<1 %
136	www.tjprc.org Internet Source	<1 %
137	ex.hhs.se Internet Source	<1 %
138	openresearch-repository.anu.edu.au Internet Source	<1 %
139	dspace.uah.es Internet Source	<1 %
140	aede.osu.edu Internet Source	<1 %
141	www.docstoc.com Internet Source	<1 %
142	issuu.com Internet Source	<1 %
143	Pooja Thakur, L.G. Burange. "An Analysis of Productivity Spillovers from Foreign Direct Investment in India's Services Sector",	<1 %

144 Sahu, Prithish Kumar Solarin, Sakiru Adebola. "[Does higher productivity and efficiency lead to spillover? Evidence from Indian manufacturing.\(\)](#)", [Journal of Developing Areas, Summer 2014 Issue](#)

Publication

<1%

145 Laura Alfaro. "[Multinational Activity in Emerging Markets: How and When Does Foreign Direct Investment Promote Growth?](#)", [Emerald, 2017](#)

Publication

<1%

146 Iršová, Zuzana, and Tomáš Havránek. "[Determinants of Horizontal Spillovers from FDI: Evidence from a Large Meta-Analysis](#)", [World Development, 2013](#).

Publication

<1%

147 Evis Sinani, Klaus E. Meyer. "[Spillovers of technology transfer from FDI: the case of Estonia](#)", [Journal of Comparative Economics, 2004](#)

Publication

<1%