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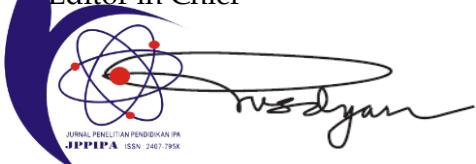
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Overall Equipment Effectiveness (OEE) Analysis as a Framework for Mapping Root Causes of Production Inefficiencies in Cream Filling Processes

Asmaul Fauziah¹, Endang Wahyu Fitriani ^{2*}, Agnes Nuniek Winantari³

Pharmacy Industry, Faculty of Pharmacy, University of Surabaya, Surabaya, 60293, Indonesia.

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Revised:

Accepted:

Published:

Corresponding Author:

Author Name:

Endang Wahyu Fitriani

endangwahyu@staff.ubaya.ac.id

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Abstract: This study aims to utilize Overall Equipment Effectiveness (OEE) analysis as a diagnostic framework to identify and map the root causes of performance losses in critical cream filling machinery within a cosmetics manufacturing company in East Java. The research was conducted over a three-month period using a descriptive quantitative method through direct observation of filling machine operations. The collected data included machine operating time, downtime, production output, and defective units. OEE was calculated based on its three core components – availability, performance, and quality. The results show that OEE values fluctuated significantly between 20.31% and 90.51%, with monthly averages of 71.12%, 80.62%, and 61.29%, respectively. While the quality component remained relatively stable at above 98%, substantial variations were observed in availability and performance due to machine downtime and reduced operating speeds. Further analysis using the Six Big Losses framework identified setup and adjustment time, and idling/minor stoppages as the dominant contributors to decreased machine effectiveness. These losses collectively accounted for the majority of production interruptions and performance deterioration throughout the observation period. The findings underscore the importance of systematically reducing the Six Big Losses to stabilize and improve OEE. Implementing preventive maintenance, minimizing setup time variability, and strengthening operational discipline are critical steps for achieving sustainable improvements in machine effectiveness and overall production efficiency.

Keywords: Overall Equipment Effectiveness (OEE); Availability; Performance; Quality; Six Big Losses; Filling Machine; Manufacturing Efficiency

Introduction

In modern cream manufacturing industries, production efficiency and equipment reliability play a decisive role in ensuring consistent output and maintaining product quality. Among various types of production equipment, the filling machine is considered one of the most critical assets because it directly determines packaging accuracy, product uniformity, and overall throughput. Any inefficiency in this machine can significantly disrupt the production flow, increase material waste, and ultimately affect customer satisfaction. As a result,

ensuring high performance in filling machinery is essential for achieving operational excellence and meeting the demands of competitive markets.

One of the most effective tools for evaluating equipment performance is Overall Equipment Effectiveness (OEE), a comprehensive metric that integrates availability, performance, and quality to assess how effectively production assets are utilized (Nakajima, 1988). In the context of cream filling operations, OEE serves not only as a performance indicator but also as a strategic decision-support tool

How to Cite:

Example: Susilawati, S., Doyan, A., Mulyadi, L., & Hakim, S. (2019). Growth of tin oxide thin film by aluminum and fluorine doping using spin coating Sol-Gel techniques. *Jurnal Penelitian Pendidikan IPA*, 1(1), 1-4. <https://doi.org/10.29303/jppipa.v1i1.264>

to identify inefficiencies that hinder production output.

The interpretation of OEE is closely connected to the Six Big Losses, a classification widely used in Total Productive Maintenance (TPM) to identify production losses that degrade manufacturing performance. These losses—equipment failures, setup and adjustment losses, idling and minor stops, reduced speed, process defects, and reduced yield—represent the primary sources of operational inefficiencies that impact the three pillars of OEE (Womack & Jones, 2003). In cream filling machines, these losses often manifest as intermittent breakdowns, inaccurate fill volumes, nozzle clogging, slow operating speeds, and increased reject rates.

Despite the availability of structured evaluation methods, many manufacturers still face challenges in analyzing the root causes behind reduced OEE in filling machinery. Factors such as inconsistent preventive maintenance, inadequate operator training, variations in cream viscosity, and improper machine calibration frequently contribute to recurring performance issues (Muchiri & Pintelon, 2008). Without systematic root cause analysis, these issues tend to accumulate over time, resulting in lower productivity and higher operational costs.

Given the critical role of filling machines in determining the overall production output and quality, performing a detailed root cause analysis becomes increasingly urgent. By identifying dominant losses and tracing them to their underlying causes, manufacturers can implement targeted improvements that enhance equipment reliability, reduce downtime, and improve production stability.

This study aims to conduct a comprehensive root cause analysis of the filling machine used in cream production by integrating the OEE framework and the Six Big Losses. The research seeks to provide actionable recommendations that not only address existing inefficiencies but also strengthen long-term maintenance and operational practices.

Anatomy and Working Principle of a Semi-Automatic Pneumatic Filling Machine

The semi-automatic pneumatic filling machine is a widely utilized piece of equipment in cream, paste, and viscous-product manufacturing due to its simplicity, accuracy, and suitability for small to medium-scale production environments. Anatomically, the machine consists of several key components that work cohesively to ensure precise volumetric dispensing. At the top of the system, a stainless-steel hopper serves as the primary reservoir that holds the cream or viscous material prior to

filling. Its conical design ensures continuous and uniform feeding of the product into the filling chamber through gravity-assisted flow.

Beneath the hopper lies the central volumetric filling mechanism, which is based on a piston-cylinder assembly. This mechanism is driven entirely by compressed air supplied through a pneumatic control system equipped with air regulators, solenoid valves, and flow control valves. These pneumatic components govern the motion of the piston during both the suction and dispensing phases. When compressed air actuates the pneumatic cylinder, the piston retracts to create negative pressure within the filling chamber, thereby drawing product from the hopper into the cylinder. In the subsequent phase, the pneumatic system reverses the airflow direction, pushing the piston forward to discharge the product through the filling nozzle. The nozzle itself is designed with an anti-drip system that ensures clean, controlled material transfer into the container.

The machine's operation is semi-automatic, typically activated by a foot pedal that triggers each filling cycle. This allows the operator to maintain control over the timing of the filling process while ensuring consistent dispensing volume determined by the adjustable piston stroke. The stainless-steel frame provides structural stability and supports hygienic operation, making the equipment compliant with food-grade and pharmaceutical manufacturing requirements.

From a performance standpoint, the pneumatic filling machine offers several advantages. Its volumetric piston mechanism ensures high accuracy and repeatability, while the pneumatic actuation provides smooth and reliable operation even for high-viscosity products. Additionally, the machine requires minimal energy consumption, as it operates without electric motors for the filling motion. Maintenance procedures are relatively simple because the filling components are easy to disassemble for cleaning, and the system has few wear-prone parts.

Overall, the anatomical structure and operating principles of the semi-automatic pneumatic filling machine make it an efficient, flexible, and economically viable solution for industries requiring consistent cream-filling performance. Its straightforward mechanism not only supports accurate and repeatable filling but also enables seamless integration with performance improvement tools such as Overall Equipment Effectiveness (OEE) analysis and root cause diagnostic methods within a continuous improvement framework. Furthermore, the machine has undergone formal qualification activities—including Installation Qualification (IQ)

and Operational Qualification (OQ)—which confirm that the equipment operates in accordance with predefined specifications. The completion of these qualification stages establishes the machine's technical reliability and regulatory compliance, thereby ensuring that subsequent OEE evaluation is based on a validated and properly functioning production asset.

established formulas as proposed by Ghafoorpoor Yazdi, Azizi, and Hashemipour (2018):

$$\text{Availability} = \frac{\text{Effective Working Time}}{\text{Total Working Time}} \times 100\% \quad (1)$$

Planned Production Time = Total scheduled production time - planned downtime.

Operating Time = Planned Production Time - unplanned downtime

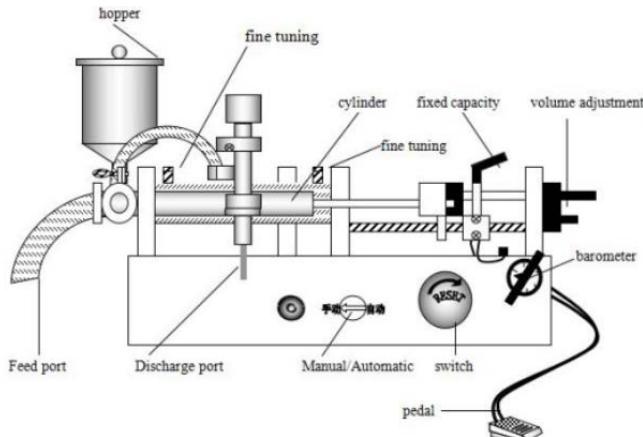


Figure 1. Semi- Automatic Pneumatic Filling Machine & Anatomy

Method

This study employs a descriptive quantitative approach aimed at measuring production efficiency through the calculation and analysis of Overall Equipment Effectiveness (OEE). The research was conducted at a cosmetics manufacturing company located in East Java, which utilizes a filling machine as a key component in its production process. The goal of this study is to evaluate the machine's performance, identify factors that contribute to efficiency losses, and determine opportunities for improvement based on OEE results.

Data were collected over a period of three consecutive months through direct observation of daily production activities. The collected data included Total Working Time, Production Idle Time, Effective Working Time, total units produced, the number of good units, and the number of defective units. These operational parameters served as the primary inputs for calculating OEE and diagnosing performance losses within the production system. The OEE calculation was conducted based on its three main components: availability, performance, and quality. Each component was computed using

$$\text{Performance} = \frac{\text{Actual Output}}{\text{Theoretical Output}} \times 100\% \quad (2)$$

Theoretical Output = Planned Production Time \times Ideal Cycle Rate

$$\text{Quality} = \frac{\text{Good Units}}{\text{Total Units Produced}} \times 100\% \quad (3)$$

The data analysis in this study was conducted to evaluate the machine's availability level and to identify the main factors contributing to downtime during a 13-week observation period. The data were obtained from machine operation and downtime records for each production week. The availability value was calculated by comparing the actual operating time with the planned operating time. Subsequently, all downtime data were categorized based on their causes, including waiting for raw materials, setup and adjustment, mechanical failure, production administration, and other contributing factors. The analysis was further supported by constructing a Pareto Diagram to determine the dominant factors contributing the most to production time losses.

Result and Discussion

The calculation of Overall Equipment Effectiveness (OEE) in this study was based on three main components: (1) availability, which represents the proportion of effective operating time to the planned production time; (2) performance, which measures the ratio between the actual output and the machine's theoretical capacity; and (3) quality, which is determined by the percentage of good products relative to the total output. The values of availability, performance, and quality were calculated using the formulas proposed by Ghafoorpoor Yazdi, Azizi, and Hashemipour (2018).

In the context of this study, the company employed an internal calculation approach that incorporates the terms Effective Working Time and Production Idle Time to reflect the actual production conditions. *Production Idle Time* includes all periods when the machine is not operating and not producing output, encompassing both planned downtime (e.g., routine maintenance activities) and unplanned downtime (e.g., technical failures or unexpected breakdowns). Therefore, in this study, *Production Idle Time* is treated as a combined representation of both planned and unplanned downtimes. This clarification is intended to ensure consistency between the company's operational terminology and the standard OEE components used in the analysis (Zelinka, Bobovský, & Bobrová, 2019).

Ideally, the achievement of Overall Equipment Effectiveness (OEE) is realized after the implementation of Total Productive Maintenance (TPM). Under such conditions, machine performance is expected to reach an availability rate above 90%, performance efficiency exceeding 95%, and a quality rate greater than 99%. When all three indicators are achieved, the overall OEE value can surpass the ideal benchmark of 85%, which is widely recognized as the global standard for production equipment effectiveness (Ariyah, 2022).

At PT X, located in East Java, OEE was used as a method to measure the effectiveness of machine performance by considering the three main indicators: availability, performance, and quality. The availability calculation was based on Effective Working Time and Production Idle Time, distinguishing the total machine hours into these two operational categories. *Effective Working Time* represents the machine's actual production time within one month, calculated as the total machine

working hours minus *Production Idle Time*. Meanwhile, *Production Idle Time* includes the time spent on setup, shutdown, maintenance, repair, handling of defective products, and process adjustments. This distinction enables the company to accurately assess machine utilization levels and identify sources of production time losses. The data on *Effective Working Time* and *Production Idle Time* collected during the period of January to March 2025 served as the basis for the OEE calculation. The results of the availability calculation are presented in Table 1.

Based on the calculation of availability for the production process over a 13-week period, a significant variation was observed from week to week. The highest availability value reached 97.50% in Weeks 8 and 9, indicating that the machine was almost fully utilized according to the planned working schedule, with very minimal idle time (only 60 minutes out of a total of 2400 minutes). This condition reflects effective scheduling and minimal operational disturbances.

Conversely, the lowest availability value occurred in Week 10, at only 30%, due to an extensive idle time of 1680 minutes out of a total of 2400 minutes. This result indicates that the machine was idle for most of the production period, leading to a drastic decline in production capacity.

Based on the analysis of downtime data collected during the observation period, a Pareto Diagram was constructed, as shown in Figure 1 & Table 2. This diagram illustrates the proportional distribution of machine downtime according to its main causes, which include setup and adjustment, mechanical failures, production administration, and other minor factors.

Based on the analysis presented in the Pareto Diagram, the primary contributor to downtime in the cream filling process was setup and adjustment activities, with a total duration of 3,631 minutes, accounting for 68% of the total downtime. This was followed by mechanical failures, amounting to 720 minutes or 14%, and production administration-related delays, which contributed 507 minutes or 10% of total downtime. The remaining 452 minutes (9%) were categorized as other minor disturbances. Collectively, setup & adjustment and mechanical failure accounted for more than 80% of total downtime, indicating that the decline in machine availability and OEE performance was primarily driven by process-related and technical factors that disrupted operational continuity.

Table 1. Calculation of Availability Values

Week	Total Working Time (Minutes)	Production Idle Time (Minutes)	Effective Working Time (Minutes)	Availability (%)
1	1140	420	720	63.16
2	2400	240	2160	90
3	2400	420	1980	82.50
4	2400	180	2520	93.33
5	1560	120	1440	92.31
6	2400	420	1980	82.50
7	2400	600	1800	75
8	2400	60	2340	97.50
9	2400	60	2340	97.50
10	2400	1680	720	30
11	2400	240	2160	90
12	2400	600	1800	75
13	2070	270	1800	86.96

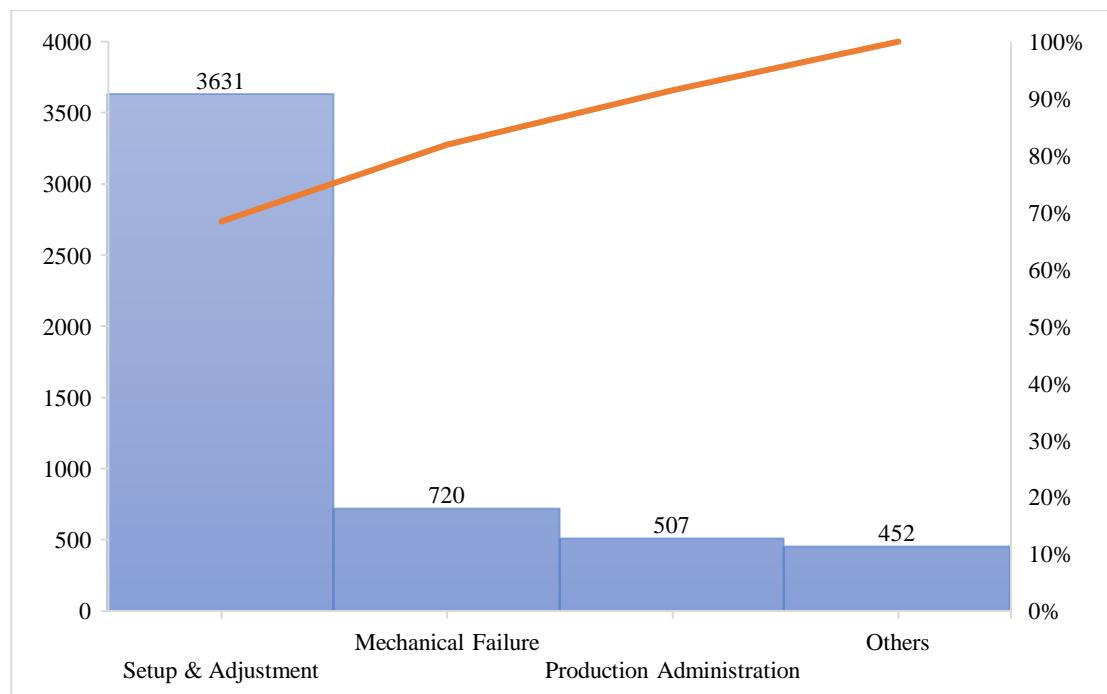


Figure 2. Pareto Diagram of Machine Downtime Causes

Table 2. Distribution of Downtime Causes, Duration, and Contribution Percentage

Downtime Cause	Duration (minutes)	Percentage (%)	Cumulative (%)
Setup & Adjustment	3,631	68%	68%
Mechanical Failure	720	14%	82%
Production Administration	507	10%	91%
Others	452	9%	100%
Total	5,310	100%	—

These findings are consistent with the Pareto 80/20 principle, which states that the majority of production losses typically stem from a small number

of dominant factors (Prasetyo & Sutopo, 2021). Therefore, improvement efforts should prioritize reducing setup and adjustment duration—

Table 2. Calculation of Performance Values

Week	Effective Working Time (Minutes)	Ideal Cycle Rate (unit/menit)	Theoretical Output (unit)	Actual Output (unit)	Performance (%)
1	720	4	2880	2115	73.44
2	2160	4	8640	7803	90.31
3	1980	4	7920	7412	93.59
4	2520	4	10080	7506	74.46
5	1440	4	5760	5203	90.33
6	1980	4	7920	7211	91.05
7	1800	4	7200	6705	93.13
8	2340	4	9360	8503	90.84
9	2340	4	9360	8711	93.07
10	720	4	2880	2051	71.22
11	2160	4	8640	7908	91.53
12	1800	4	7200	6603	91.71
13	1800	4	7200	6804	94.50

particularly through better standardization, operator training, and streamlined changeover procedures—and addressing recurring mechanical issues through enhanced preventive and predictive maintenance. This observation aligns with the perspective of Ghafoorpoor Yazdi, Azizi, and Hashemipour (2018), who identified setup time and mechanical disturbances as common root causes of low machine effectiveness in batch-based manufacturing environments. By implementing focused improvements in these areas, the company can significantly enhance Overall Equipment Effectiveness (OEE) and ensure more stable production performance.

In general, most weeks exhibited availability values ranging between 75% and 93%, which can be considered satisfactory. For instance, Week 2 (90%), Week 4 (93.33%), and Week 5 (92.31%) showed stable performance with relatively low idle time. However, several weeks experienced notable declines, such as Week 1 (63.16%) and Week 6 (82.50%), indicating potential areas for improvement in material planning, setup scheduling, maintenance practices, and production problem handling. This suggests that consistency remains a key challenge in maintaining stable production performance.

Significant fluctuations in availability require close attention, as this instability directly affects production throughput and cost efficiency. Overall, the analysis shows that while the production process can achieve high availability under certain conditions, there were also critical weeks with substantial idle time. Improvement efforts should therefore focus on reducing unplanned downtime, optimizing setup

duration, and enhancing machine reliability through preventive and predictive maintenance programs. Addressing these aspects is expected to improve availability stability, thereby contributing to higher overall OEE and lower production costs (COGS).

The performance analysis of the production process over the 13-week observation period revealed variations ranging from 71.22% to 94.50%. The highest performance value was achieved in Week 13, reaching 94.50%, where the machine produced 6,804 units out of a theoretical output of 7,200 units. This indicates that the production rate was operating very close to its ideal cycle speed. Conversely, the lowest performance values were recorded in Week 10 (71.22%) and Week 1 (73.44%), reflecting a significant reduction in production speed compared to the ideal standard. Such decreases are likely attributable to technical factors, such as equipment wear or operational issues that caused the machine to run below its optimal capacity (Tang, 2019).

Overall, most weeks showed performance values within the 90%–94% range, which can be classified as good according to OEE standards. However, the fluctuations observed in several weeks indicate inconsistency in production speed performance. These variations are closely associated with the dominant downtime contributors identified in the analysis, particularly setup and adjustment activities (68%), which often require operators to recalibrate machine settings, causing delays in achieving stable operating speeds. Additionally, mechanical failures (14%) may lead to temporary reductions in machine speed as operators attempt to maintain safe operation, while production

administration delays (10%) and other minor disturbances disrupt workflow continuity and affect machine rhythm.

To improve performance stability, the company

The quality values of the production process throughout the 13-week observation period remained relatively high, ranging from 92.67% to 99.94%. Most weeks recorded quality rates above 99%, indicating

Table 3. Calculation of Quality Values

Week	Total Output (unit)	Good Units (unit)	Defect Units (unit)	Quality (%)
1	2115	2105	10	99.53
2	7803	7798	5	99.94
3	7412	7402	10	99.87
4	7506	7490	16	99.79
5	5203	5100	103	98.02
6	7211	7100	111	98.46
7	6705	6685	20	99.70
8	8503	8483	20	99.76
9	8711	8688	23	99.74
10	2051	1950	101	95.08
11	7908	7853	55	99.30
12	6603	6423	180	97.27
13	6804	6305	499	92.67

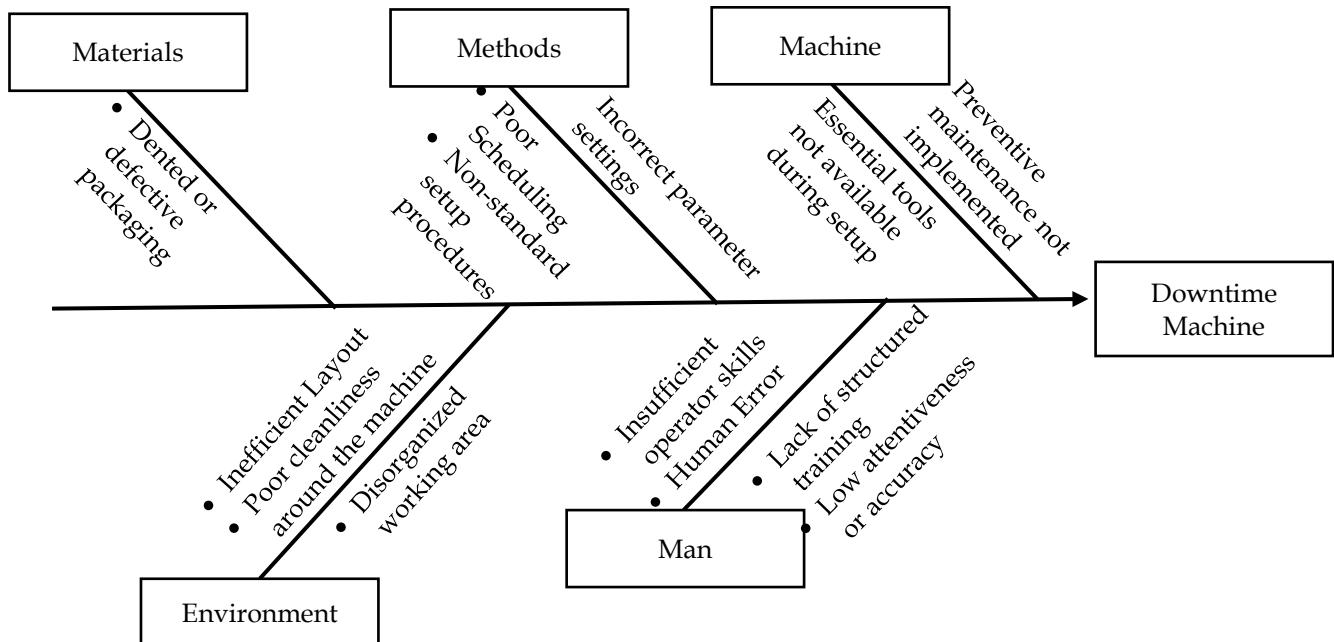


Figure 3. Fishbone Diagram

should ensure consistent implementation of scheduled maintenance, optimize process parameter settings during setup, and strengthen operator competency to maintain machine speed close to the ideal cycle rate (Annamalai & Suresh, 2019). These measures will help reduce speed-related losses, stabilize machine performance, and ultimately contribute to higher and more consistent Overall Equipment Effectiveness (OEE).

that the pneumatic filling machine generally produced low levels of defective products. The highest quality value occurred in Week 2 (99.94%), with only five defective units out of 7,803 units produced, while the lowest was found in Week 13 (92.67%) due to a substantial increase of 499 defective units out of 6,804 units. Several other weeks—such as Week 10 (95.08%) and Week 12 (97.27%)—also

showed noticeable quality declines, primarily due to a rise in defective outputs.

To understand the underlying causes that contributed to these quality deviations, an Ishikawa (Fishbone) Diagram was used as shown in Figure 2. The fishbone model systematically categorizes potential sources of problems into Materials, Methods, Machine, Man, and Environment, enabling a structured evaluation of factors that may influence machine downtime and product defects (Singh, Clements, & Sonwaney, 2018).

Based on the analysis using the Fishbone Diagram (cause-and-effect diagram), several key factors were identified as potential contributors to the decline in product quality during the homogenization process. These factors include Man, Machine, Material, Method, Measurement, and Environment.

setup caused prolonged adjustment time. Mechanical wear, misalignment of the filling cylinder, and leakage in pneumatic connections also contributed to unstable machine operation. These problems directly affect fill accuracy, sealing consistency, and machine cycle time, thereby influencing both performance and quality.

The fishbone analysis identified material conditions as a direct contributor to downtime. Dented or defective packaging frequently caused interruptions as the machine was unable to process deformities, forcing operators to stop production and manually replace units. Variability in packaging quality also increased the risk of misfeeds and product spillage. Inefficient layout of supporting materials—leading to longer searching and handling time—further contributed to minor stoppages.

Table 4. Calculation of OEE Values

Week	Availability (%)	Performance (%)	Quality (%)	OEE (%)	OEE / Month
1	63.16	73.44	99.53	46.17	Month 1 71.12%
2	90	90.31	99.94	81.23	
3	82.5	93.59	99.87	77.11	
4	93.33	74.46	99.79	69.35	
5	92.31	90.33	98.02	81.73	
6	82.5	91.05	98.46	73.96	Month 2 80.62 %
7	75	93.13	99.7	69.64	
8	97.5	90.84	99.76	88.36	
9	97.5	93.07	99.74	90.51	
10	30	71.22	95.08	20.31	Month 3 61.29 %
11	90	91.53	99.3	81.80	
12	75	91.71	97.27	66.90	
13	86.96	94.5	92.67	76.15	

From the human (Man) aspect related issues were among the prominent contributors to downtime and quality variation. Insufficient operator skills, human error, and a lack of structured training were frequently observed. Operators with limited experience may struggle to adjust pneumatic pressure, filling volume settings, or sealing parameters accurately, which can lead to inconsistent filling results or temporary machine stoppages. Low attentiveness or accuracy, particularly during long production cycles, may also increase the likelihood of improper tube alignment and misfeeds (Kamble, Gunasekaran, & Gawankar, 2018).

Machine-related causes played a critical role in downtime events. The most significant issues included the absence of preventive maintenance, leading to gradual performance degradation. In addition, essential tools that were not available during

Several downtime events were associated with method-related factors, especially poor scheduling and non-standard setup procedures. Inconsistent setup practices led to repeated adjustments during production, reducing machine availability. Incorrect parameter settings—identified as a major root cause—resulted in unstable filling volume, improper sealing, and the need for rework. These inconsistencies emphasize the lack of standardized operating procedures for machine setup and calibration.

Environmental factors also contributed to downtime. Inefficient layout around the machine, poor cleanliness, and a disorganized working area increased the time required for movement, inspection, or troubleshooting. Accumulated dust or spilled product residues could interfere with pneumatic components, increasing the likelihood of minor stoppages and reduced filling speed.

Overall, the Fishbone Diagram analysis indicates that downtime in the pneumatic filling machine is not caused by a single factor but by the interaction among human error, inadequate machine maintenance, substandard packaging materials, non-standardized methods, and a poorly organized working environment (Danese, Manfè, & Romano, 2018). Although the quality rate remained relatively high across most weeks, the identified issues highlight the need for improvements in operator training, standardization of setup procedures, and preventive maintenance practices to minimize downtime-related losses.

Ensuring consistency in raw material quality, maintaining a clean production environment, and implementing structured work arrangements are also crucial to stabilizing machine operations. Enhancing these factors is expected to further improve the reliability of the pneumatic filling process, reduce the rate of defective products, and ultimately strengthen overall production effectiveness (Durga Prasad & Radhakrishna, 2019).

The calculation results of Overall Equipment Effectiveness (OEE) for the 13-week production period showed a wide variation, ranging from 20.31% to 90.51%. This fluctuation was influenced by changes in the three OEE components: availability, performance, and quality. During the first month (Weeks 1-5), the average OEE was 71.12%, with the highest value achieved in Week 5 (81.73%) and the lowest in Week 1 (46.17%), primarily due to low availability (63.16%). In the second month (Weeks 6-9), the average OEE improved to 80.62%, reaching its peak in Week 9 (90.51%). This improvement was supported by high availability (97.5%) and consistently strong performance levels above 90%. However, in the third month (Weeks 10-13), the average OEE declined significantly to 61.29%. The sharp drop in Week 10 (20.31%) was the most notable, driven by a steep decrease in availability (30%) and a slight decline in quality (95.08%).

To better understand the causes behind these OEE fluctuations, the Six Big Losses framework was utilized to analyze performance losses across the observation period. Among the loss categories, equipment failure (breakdown losses) emerged as the most dominant factor affecting availability. This was evident in Weeks 1 and 10, where extended mechanical failures—including agitator motor malfunction and valve leakage—resulted in prolonged downtime and consequently low OEE values. These failures highlight the critical need for more structured preventive maintenance and periodic equipment inspection.

Setup and adjustment losses also played a significant role in reducing availability, particularly during production transitions such as material changeovers and sanitation procedures. Weeks 3, 6, and 11 showed noticeable increases in setup duration, indicating potential inconsistencies in preparation activities. Standardizing these processes through SMED-based improvements may help reduce variability and enhance production readiness.

In addition to major downtime events, idling and minor stops contributed to hidden performance losses, particularly in weeks where OEE fell below 70%. Short but frequent interruptions—such as raw material delays and operator coordination issues—accumulated to reduce effective machine utilization, suggesting the need for better material flow control and workflow synchronization.

While performance values remained relatively high across most weeks, reduced speed losses were observed during the third month due to variations in product viscosity, forcing operators to lower machine speed to maintain accuracy. This indicates the importance of raw material consistency and may warrant implementation of viscosity monitoring or pre-processing adjustments.

Regarding quality losses, process defects and reduced yield had a smaller but still noticeable impact. Weeks 10 and 12 showed slight declines in quality due to inconsistent homogenization, contamination risks, and temperature deviations, resulting in defective outputs and start-up rejects. Although average quality performance remained above 95%, these deviations still contributed to weekly OEE instability.

Overall, the OEE performance can be categorized as fairly good but inconsistent, as it did not consistently achieve the world-class standard of 85% (Nakajima, 1988). Only in Weeks 8 and 9 did the machine surpass this benchmark. The integrated Six Big Losses analysis confirms that breakdown losses, setup and adjustment losses, and reduced speed losses were the most influential contributors to efficiency variation throughout the 13-week observation period. Therefore, improvement efforts should focus on minimizing unplanned downtime through enhanced maintenance routines, optimizing setup procedures, reducing minor stops through better operational coordination, and strengthening process controls to minimize defects. Addressing these areas is essential for achieving stable and improved OEE performance across future production cycles.

Conclusion

The analysis revealed that the Overall Equipment Effectiveness (OEE) of the homogenizer machine fluctuated significantly across the 13-week observation period, ranging from 20.31% to 90.51%. The monthly average OEE values were 71.12%, 80.62%, and 61.29%, indicating that machine effectiveness was inconsistent and did not yet achieve the world-class benchmark of 85%. Although the quality component remained relatively stable at above 98%, the main performance issues were found in the availability and performance components, which were affected by frequent downtime and reduced operating speeds.

The evaluation of losses using the Six Big Losses framework provided a clearer understanding of the dominant sources of inefficiency. The analysis showed that breakdown losses, setup and adjustment losses, and idling/minor stoppages were the most influential contributors to reduced availability and performance. Waiting for raw materials and prolonged setup activities accounted for nearly 70% of total downtime, confirming that operational and material-flow disruptions were the primary barriers to achieving stable OEE performance. Reduced speed losses were also observed during weeks with lower production effectiveness, often triggered by variations in material characteristics that forced operators to lower the machine speed to maintain process stability.

These findings emphasize that improvements in OEE must focus on minimizing the dominant losses identified through the Six Big Losses analysis. Strengthening preventive and predictive maintenance programs, reducing setup time variability, improving production-material readiness, and enhancing standard operating discipline are essential to reduce availability losses. Additionally, ensuring consistency in raw material properties and enhancing operator responsiveness can help mitigate speed losses and minor stoppages.

Overall, the study demonstrates that optimizing OEE requires a systematic reduction of the Six Big Losses that significantly impact machine performance. By addressing the major sources of downtime and speed reduction, the homogenizer machine can achieve more stable and higher OEE values, supporting greater operational reliability and long-term production efficiency in the cosmetics manufacturing process.

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Author Contributions

Asmaul Fauziah contributed to data collection and manuscript writing. Endang Wahyu Fitriani & Agnes Nuniek Winantari provided critical review and final approval of the manuscript.

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Conflicts of Interest

Authors declare that no conflict of interest in this publication.

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