# Applying Many-facet Rasch Measurement to Evaluate the Modified Clock Drawing Test

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The *Clock Drawing Test* is usually used to evaluate prospective patients with cognitive decline as a screening tool. This test is well-known for having rapid administration and flexibility across different cultures. The objective of this study was to explore the psychometric properties of the C*lock Drawing Test* using the Many-Facet Rasch Measurement approach. We modified the 18-point clock drawing test, specifically adjusting the scale into three levels: poor, fair, and good. This study also involved the *Global Deterioration Scale* (GDS) as a screening tool to classify participants' cognitive decline status. The Many-Facet Rasch model was applied to analyze 9,045 rating sequences, three clinical psychologist raters, 208 participants, and 15 items. We found the psychometrics information of this modified test was sufficient with the summary of Rasch statistics along with rating scale, item, and rater difficulties analysis. Based on the Wright map, the majority of the items grouped in the middle level, along with the rating scale of this test provided category measure of (-1.73), 0.0, to 1.74. Item 5 "number spacing equal" was the most challenging item for the participants, while item 3, "number all the same (Roman/Arabic)" was the most unchallenging item. Moreover, items 1, 15, and 6 were indicated as misfit items. Rater agreement yielded at 74%. According to ROC analysis, this test effectively predicted participants with mild cognitive impairment and mild dementia based on GDS criteria. Similar studies are recommended to improve the usage of the many facets of Rasch approach in the screening process under many raters' circumstances.

Keywords: Many Facet Rasch Measurement, Clock Drawing Test, Cognitive Screening

There are various types of tests in neuropsychological assessment that are often used to assist in both diagnostic processes and screening purposes. The *Clock Drawing Test* (CDT) is a well-known measurement tool recognized for its rapid administration process (2-5 minutes), (Agrell & Dehlin, 1998; Cullen et al., 2007; Pinto & Peters, 2009). This tool is favored by practitioners due to its flexibility for cross-cultural use (Borson et al., 1999; Storey et al., 2002). Shulman et al. (1986) found the

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administration of this tool to be non-threatening and comfortable for elderly patients. Previous studies found that CDT was an efficient screening tool to evaluate cognitive abilities, (Hazan et al., 2018). CDT is also valued for its broad capacity to assess various cognitive dysfunctions, including visual ability, memory, visuospatial skills, planning, abstraction, concentration, understanding, and response processes, (Ismail et al., 2010; Shulman, 2000). Moreover, this test encompasses quantitative and qualitative tests, each with diverse administration techniques, (Lam et al., 1998; Manos & Wu, 1994; Mendez et al., 1992; Sunderland et al., 1989; Wolf-Klein et al., 1989).

Some studies discovered the validity of this test through its correlation with other instruments, for instance, the *Mini-Mental State Examination* (MMSE; Folstein et al., 1975) or the *Montreal Cognitive Assessment* (MoCA; Nasreddine et al., 2005). Initially, Shulman et al. (1993), who developed the oldest scoring system, found a significant correlation between Shulman’s scoring system and MMSE. However, there was some disagreement with that finding because a recent study found that this test did not correlate significantly with MMSE. Ilardi et al. (2020) found that several scoring methods, including Shulman’s version, did not strongly correlate with MMSE. Carneiro (2015), on the other hand, discovered that some scoring versions had a significant correlation with MMSE and MoCA's just in the visuospatial task domain section. Cahn-Weiner et al. (2003) stated that CDT is not designed for diagnosis due to its inability to precisely specify cognitive dysfunction. Several studies categorized CDT as a screening tool for patients with suspected dementia symptoms, (Babins et al., 2008; Kørner et al., 2012; Manos & Wu, 1994; Tabari & Amini, 2021).

Numerous previous validation studies have been conducted, encompassing various approaches such as the construction of new items, modification of existing ones, and even altering the scale. Several noteworthy findings have garnered attention from both researchers and practitioners. For instance, in a study by Ricci et al. (2016), the *Clock Drawing Test* (CDT) items were developed using a Likert scale. Their research used principal component analysis with some additional analyses. While their study successfully elucidated the psychometric properties of the test, it fell short of providing a comprehensive understanding of each item's difficulty level or its interaction with the participants. Similarly, Emek-Savaş et al. (2018) directed their analysis towards evaluating three scoring systems for the CDT. While their findings yielded valuable insights, more comprehensive information could be discovered by analyzing all the items individually. Furthermore, they used the Intraclass Correlation Coefficient (ICC) for reliability measurement in scenarios involving multiple raters and it may offer insights into the consistency of their assessments. However, this approach may encounter challenges in explaining the severity of discrepancies among raters, leaving us with the question of 'how large are these differences?'

Rasch measurement theory has been utilized across many fields for measurement tool development purposes (Alexandrowicz et al., 2018; Batchelder et al., 2020; Camargo & Henson, 2015; Franchignoni et al., 2011; Han & Li, 2015; Natanael, 2021; Park et al., 2021; Petrillo et al., 2015; Zahirah & Susanto, 2021). This method was selected for its capacity to conduct linear and objective measurements, as it treats the Likert scale as ordinal data (Boone, 2016; Sumintono, 2018). Essentially, this analysis aims to test the compatibility of the empirical data with the model fit (de Ayala, 2009). Beyond its logit transformations, this approach also emphasizes various features, including item-person difficulties, fit statistics, standard error, and point-measure correlation. Construct validity in this model can be assessed through Rasch residual principal component analysis or items' fit statistics, (Linacre, 2011). Moreover, researchers can determine the validity of the rating scale using Andrich threshold analysis (Chong et al., 2022).

The many-facets Rasch model (MFRM) was developed as an advanced technique in the measurement process, taking into consideration various facets (Linacre, 1994; Tavakol & Pinner, 2019). The primary focus of the MFRM is to establish fair measurement with bias estimation that may occur among facets during the evaluation process. Facets can encompass various elements within the evaluation setting, such as raters, places, and task variability (Bond & Fox, 2015). Previous studies have demonstrated that MFRM is commonly employed for evaluation in educational testing (Farlie et al., 2021; Gordon et al., 2021; Huebner & Skar, 2021; Uto, 2021). Nevertheless, this study seeks to extend the application of this method to clinical settings due to its capability to handle evaluations involving multiple raters, providing additional features compared to classical test theory.

To the best of the researchers' knowledge, no studies have endeavored to examine the psychometric qualities of the clock drawing test using the many-facet Rasch measurement. Consequently, in this study, researchers proposed employing a multifaceted Rasch measurement to thoroughly investigate and delineate the psychometric features of this test.

## METHOD

### Participants

This study received approval from the ethics committee of the University of Surabaya. The data collection process occurred in multiple elderly care facilities in Surabaya. All participants gave informed consent after receiving comprehensive details about the study, and individuals with sensory or motor impairments that could hinder their participation in data collection were screened. No initial diagnoses were made in this research; however, the researcher utilized the *Global Deterioration Scale* (GDS) to offer an initial overview of the participants' conditions.

A total of 208 volunteers, comprising 40.3% males and 59.6% females, participated in this study. Participants' ages ranged from 48 to 99 years, with a mean (*SD*) of 64.798 (8.815). All participants in this study had varying levels of education, ranging from 0 to 22 years, with a mean (*SD*) of 9.029 (4.937). GDS measurements were obtained through brief screenings conducted during clinical interviews with participants, assisted by caregivers. Subsequently, the researchers categorized the participants into five characteristics based on GDS criteria.

Table 1. Demographic Characteristics of Participants Based on GDS Criteria

| Characteristic | Frequency (%) |
|---|---|
| No Cognitive Decline | 33 (15.8%) |
| Age Associated Memory Impairment | 119 (57.2%) |
| Mild Cognitive Impairment | 40 (19.2%) |
| Mild Dementia | 13 (6.2%) |
| Moderate Dementia | 3 (1.4%) |

**Clock Drawing Test**

We utilized the 18-point scoring system of the *Clock Drawing Test* developed by Babins et al. (2008). This version represents an updated version of the scoring system by Rouleau et al., (1992), incorporating five items with combined response options (binary and ternary). Participants were instructed to draw a clock face along with the numbers, and the hands should point to 11:10. In this study, we modified the scoring technique into a continuous rating scale ranging from 1 to 3, with response options: 1 = poor, 2 = fair, 3 = good. The modified 18-point scoring technique in this study did not change either the content of the items or the instruction. The researchers transformed all of the sub-items from items 3 and 4 into individual items, standardizing the response options into a three-point rating scale. Consequently, the modified CDT using an 18-point scoring technique in this study comprised 15 items with equal response options.

**Global Deterioration Scale**

The *Global Deterioration Scale* (GDS) comprises a set of measurement tools developed by Reisberg et al. (1982), and designed for the rapid and efficient screening of patients suspected to have dementia. Additionally, this test has proven valuable for evaluating and monitoring the progression of patients diagnosed with Alzheimer's Disease (Eisdorfer et al., 1992). According to Choi et al. (2016), GDS is not a perfect measurement tool for further evaluating cognitive decline in patients, as compared to MMSE or CDR. However, in this study, the researchers employed the GDS as a screening tool to classify participants' cognitive decline status.

**Many Facets Rasch Measurement**

The data were analyzed using the many-facet Rasch measurement. the origins of this model can be traced back to Verlhust's differential equation, stemming from the Rasch model (Bock, 1997). The many-facets Rasch model is a transformation logistics-based equation, slightly distinct from Rasch measurement. This method considers facets as factors that may influence the evaluation process, encompassing various elements such as raters, places, or even task methods (Linacre, 1994). We employed the model represented by Equation 1, wherein $\theta_n$ represents the examinee's ability, $\beta_i$ describes the category difficulty of the test, $\alpha_j$ is the rater severity, and $\tau_k$ is the category coefficient (Eckes, 2014). For instance, if a rater is included as a facet, one of the main advantages of this model is its ability to identify bias facet in logit measurement.

Equation 1. Many Facet Rach Model (3 Facets)

$$In\left[\frac{p_{nijk}}{p_{nijk-1}}\right] = \theta_n - \beta_i - \alpha_j - \tau_k$$

We utilized FACET version 3.86.0 for data analysis. This software employs the unconditional joint maximum likelihood (UCON) approach, chosen for its independence, flexibility in handling missing data, and ability to analyze extensive datasets (Linacre, 2011). The iterative process in this analysis will adhere to convergence criteria standards, with a PROX iteration ($< 0.5$), while JMLE iteration will follow the standards by stopping at ($\leq 0.001$) and a maximum residual score ($\leq .1$) (Linacre, 2011).

**Procedure**

The procedure in this study was divided into two phases: data collection and rater assessment. The researchers involved three psychologists and research assistants throughout the process. After ensuring that all participants comprehended the informed consent, we provided them with blank A4 sheets of paper and writing tools. Subsequently, brief interviews were conducted with caregivers and participants to complete the GDS. Following this, participants were remotely instructed through digital devices, with the instructions displayed on the screen. The researchers then duplicated the participants' work, which was later subjected to the assessment phase by three raters, each conducting their evaluation independently.

## RESULTS

### Summary Rasch Statistics

A total of 3 raters x 15 items x 208 participants were involved in this study, resulting in 9,360 response sequences. Table 2 shows the summary statistics of the Rasch measurement. The global chi-square value of this test was $\chi 2$ (*d.f*) = 10,247.22553 (3.798) with a significance level of 0.6520. The chi-square analysis for all facets yielded significant results ($p \geq 0.00$), and the degree of freedom (*d.f*) was set as N-1, where N represents the total number of observations for each facet. Based on the unidimensionality test, this measurement tool exhibits a variance explained by the measure of 59.67%. In contrast to the Cronbach α reliability test, the separation and strata reliability measures indicate the

Table 2. Rasch Summary Statistics

| Statistics | Participants | Raters | Items |
|---|---|---|---|
| *M* (measure) | 0.95 | 0.00 | 0.00 |
| *S.D.* (measure) | 1.93 | 0.36 | 1.21 |
| *M* (S.E.) | 0.35 | 0.03 | 0.08 |
| Adj. True *SD* | 1.87 | 0.36 | 1.21 |
| $\chi 2$ | 3280.3 | 225.1 | 1976.2 |
| *df* | 207 | 2 | 14 |
| Strata | 5.80 | 14.51 | 19.01 |
| Separation | 0.94 | 0.99 | 0.99 |

length of the test and data variations for difficulty levels. This test particularly demonstrated high score separation (5.80), showcasing its capability to distinguish participants' abilities effectively. In some cases where an instrument fails to meet this criteria (< 2), proposed by Linacre (2011) additional relevant items may be necessary to extend the test. Moreover, with item separation (19.01), this test had a sufficient number of participants with various abilities.

The Wright map (Figure 1) consisted of all facets of this study (participants, raters, and items) in logit rulers with the same linearity. This visualization enables us to analyze the relationship between facets. Briefly, the majority of our participants are located on the upper side of the map, whereas the items tend to cluster in the middle area of the map.

Figure 1. Wright Map of All Facets

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+---------------------------------------------------+
|Measr|+participants|-raters|-items        |Scale|
|-----+-------------+-------+--------------+-----|
|   4 +             +       +              + (3) |
|     |             |       |              |     |
|     |             |       |              |     |
|     |             |       |              |     |
|     |             |       |              |     |
|   3 + ***         +       +              +     |
|     | **          |       |              |     |
|     | *           |       |              |     |
|     |             |       |              |     |
|     | ****        |       |              |     |
|   2 + ****        +       +              +     |
|     | *           |       |              |     |
|     | ***         |       |              |     |
|     | ************|       |              |     |
|     | ***         |       | 5            |     |
|   1 + **          +       + 1    2       +     |
|     | ***         |       | 12   4       | --- |
|     | *****       |       | 11   13   15 |     |
|     | ***         | 1     |              |     |
|     | **          |       | 10           |     |
*   0 * **          *       *              *  2  *
|     | *           | 2   3 | 14           |     |
|     | *           |       |              |     |
|     | ***         |       |              |     |
|     | *           |       | 9            | --- |
|  -1 + *           +       +              +     |
|     |             |       | 7    8       |     |
|     | *           |       |              |     |
|     |             |       | 6            |     |
|     | *           |       |              |     |
|  -2 + *           +       + 3            +     |
|     |             |       |              |     |
|     |             |       |              |     |
|     | **          |       |              |     |
|     |             |       |              |     |
|  -3 + ***         +       +              +     |
|     |             |       |              |     |
|     |             |       |              |     |
|     |             |       |              |     |
|     |             |       |              |     |
|  -4 + *           +       +              +     |
|     |             |       |              |     |
|     |             |       |              |     |
|     | **          |       |              |     |
|     |             |       |              |     |
|  -5 + ****        +       +              + (1) |
|-----+-------------+-------+--------------+-----|
|Measr| * = 1       |-raters|-items        |Scale|
+---------------------------------------------------+
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All raters in this study appear to have similar logit values, without many discrepancies among them. Further analysis of this map will be statistically presented in the next sections. Overall, from this figure the modified CDT 18-point items demonstrate the capability to cover participants at a moderate level of difficulty.

**Clock Drawing Test Psychometrics Properties**

All items in this study adhered to the standard fit test for Rasch modeling. However, researchers discovered that a few items exhibited poor infit-outfit statistics, exceeding the optimal threshold. Bond and Fox (2015) recommended both infit and outfit for MNSQ thresholds ideally should fall within the range of (0.5 - 1.5), while ZSTD should be within

Table 3. Items Analysis

| Code | Item | *M* | Infit MS | Infit *Z* | Outfit MS | Outfit *Z* |
|---|---|---|---|---|---|---|
| 5 | Number spacing equal (1,2,4,5,7,8,10,11) | 1.66 | 0.68 | -7.3 | 0.93 | -0.49 |
| 4 | Number spacing equal (3,6,9, 12) | 0.97 | 0.86 | -2.81 | 1.04 | 0.36 |
| 2 | Center | 0.95 | 0.84 | -3.32 | 1.02 | 0.23 |
| 1 | Contour integrity of the clock face | 0.9 | 0.8 | -4.09 | 2.62 | 9 |
| 12 | Size difference of the hands is respected (minute and longer) | 0.88 | 1.35 | 6.11 | 1.23 | 2.04 |
| 11 | Minute hand is towards correct number | 0.68 | 1.21 | 3.68 | 1.01 | 0.09 |
| 10 | Hour hand is towards correct number | 0.53 | 1.2 | 3.29 | 1.05 | 0.48 |
| 15 | Gestalt | 0.37 | 0.55 | -9 | 0.54 | -5.61 |
| 13 | Arrows are drawn | 0.2 | 1.1 | 1.53 | 0.96 | -0.35 |
| 14 | Hands are joined or within 12 mm (1/2") of joining | -0.01 | 0.86 | -2.25 | 0.83 | -1.73 |
| 9 | Clock has two recognizable hands | -0.47 | 0.98 | -0.27 | 0.79 | -1.97 |
| 7 | No missing or added numbers | -0.62 | 1.48 | 5.32 | 1.17 | 1.4 |
| 8 | Numbers clockwise and correct sequence | -1.42 | 1.16 | 1.51 | 0.7 | -1.94 |
| 6 | Number inside circle | -1.92 | 1.92 | 6.25 | 2.2 | 4.25 |
| 3 | Numbers all the same (Roman/Arabic) | -2.68 | 1.31 | 2 | 0.93 | -0.13 |

the range of +2 to – 2. Item number 1 (infit Z = -4.92, outfit MS = 2.62, outfit Z = 9), 15 (infit Z = -9, outfit Z = -5.61), and 6 (infit MS = 1.92, infit Z = 6.25, outfit MS = 2.2, outfit Z = 4.25) were identified as the most unfit items of this test, as they exceeded the recommended thresholds. In contrast, item numbers 3, 8, and 13 demonstrated good fit with the unidimensional model.

The specific standard error of this test ranged from .06 - .15, with point-measure correlation ranging from 0.58 to 0.72. According to Linacre (2011), a point-measure correlation of ≥ .4 is recommended as the ideal standard for discriminating between participants with high and low abilities. Specifically, item numbers 15, 8, and 9 had the highest point-measure correlation in this test, with values of 0.75 (15), 0.74 (8), and 0.73 (9) respectively.

Item numbers 5 (1.66), 4 (0.97), and 2 (0.95) posed the greatest challenges in this test, suggesting that many participants struggled to meet the assessment standards set by the three raters for these items. Conversely, researchers found that item numbers 8 (-1.42), 6 (-1.92), and 3 (-2.68) were the easiest for participants.

Table 4. Rating Scale Analysis

| Response | Quality Control | | | Rasch-Andrich Thresholds | | Expectation | |
|---|---|---|---|---|---|---|---|
| | Avg. Meas. | Exp. Meas. | Outfit MNSQ | Measure | *S.E.* | Cat. | -0.5 |
| 1 (poor) | -1.17 | -1.30 | 1.6 | | - | -1.73 | - |
| 2 (fair) | 0.38 | 0.60 | 0.8 | -0.42 | 0.04 | 0.00 | -0.97 |
| 3 (good) | 2.13 | 2.08 | 1.1 | 0.42 | 0.03 | 1.74 | 0.99 |

As researchers modified the scoring system into Likert scale, the validity test for the rating scale demonstrated that all the responses in this test were normally functioning. Participants or raters in this study did not experience confusion in recognizing the response. Based on Table 4, the Andrich thresholds and category expectations respectively exhibited consistent steps: none, -0.42, 0.42, and -1.73, 0.00, 1.74.

**Rater Analysis**

As part of the evaluation process, researchers also conducted analysis for the raters. All raters in this study exhibited a standard error ranging from 0.03 to 0.04 and a point-measure correlation ranging from 0.68 to 0.70, with both infit and outfit ZSTD values failing to meet the fit model criteria. Rater 1 demonstrated the most precise fit statistics, while Rater 2 appeared to be less fitted or underfit, and Rater 3 exhibited a higher degree of overfit compared to others. According to Table 5, researchers observed that rater 3 was the most stringent and selective in assigning scores, while rater 1 was more lenient compared to others in evaluating participants' test results. The inter-rater agreement value was 74%, indicating that these raters had a 26% disagreement rate with each other.

Table 5. Rater Analysis

| Raters | Measure | Infit MS | Infit *Z* | Outfit MS | Outfit *Z* |
|---|---|---|---|---|---|
| Rater 1 | -0.37 | 1.04 | 1.23 | 1.19 | 2.57 |
| Rater 2 | 0.02 | 1.06 | 2.29 | 1.26 | 3.85 |
| Rater 3 | 0.35 | 0.93 | -2.67 | 0.94 | -0.91 |

**Prediction Toward GDS Criteria**

Additionally, researchers conducted ROC analysis to assess the predictive ability of the modified 18-point CDT logit scores concerning the *Global Deterioration Scale* (GDS) score. Table 6 indicates that this test version was highly optimal in predicting cognitive impairment and mild dementia among participants based on GDS criteria, with sensitivity and specificity ranging from 0.80 to 0.89 and 0.87 to 0.92, respectively.

Table 6. ROC analysis

| Category | Age Associated Memory Impairment | Mild Cognitive Impairment | Mild Dementia | Moderate Dementia |
|---|---|---|---|---|
| Significance | 0.003 | < 0.001 | < 0.001 | 0.919 |
| Accuracy | 0.783 | 0.808 | 0.891 | 0.917 |
| AUC | 0.701 | 0.879 | 0.929 | 0.525 |
| Sensitivity | 0.992 | 0.800 | 0.970 | 0.000 |
| Specificity | 0.030 | 0.818 | 0.692 | 1.000 |

## DISCUSSION

We employed an 18-point scoring system developed by Babins et al. (2008) with modifications to the response using a continuous rating scale. In this study, Rasch measurement explained 59.67% of the variance. However, Linacre (2011) emphasized the necessity of conducting Rasch residual principal component analysis to investigate new dimension probability. Additionally, the researchers did not solely rely on this indication to assess the validity of this test, as fit statistics are also valuable in confirming unidimensionality. Item numbers 1, 15, and 6 were identified as the most misfit items of this test. Furthermore, participants with unexpected responses were more likely to play a significant role in the item's outfit measurement, as it is more sensitive to outliers and extreme values (Brentari & Golia, 2007; Engelhard, 1992). Nevertheless, this study did not eliminate any items. According to Abdaziz et al. (2014), the elimination of items requires other fulfilled conditions, such as point-measure correlation (≥ 0.4), in addition to infit and outfit tests.

The most challenging item to reach agreement on was item 5, which required participants to estimate the distance between numbers on the clock (1, 2, 4, 5, 7, 8, 10, 11) evenly. According to Tranel et al. (2008), participants who failed in tasks involved spatial ability were found to have lesions in their parietal lobe, specifically in the supramarginal gyrus. Talwar et al. (2019) also found that decreased activity in the parietal and bilateral occipital lobes is associated with poor performance in drawing a clock on the CDT in general. The findings of this study revealed that the item measuring the overall gestalt of a clock (item 15) was the most reliable for discriminating between high and low-ability participants, with a point-measure correlation of 0.75. This finding contradicted those of Bennasar et al. (2013), who found that the length of the clock's hands (item 12) was the most reliable criterion for differentiating between participants with cognitive issues and those without the spatial ability to analyze and plan while drawing a clock.

On this modified CDT-18 points, we modified the scoring system into a rating scale model. The previous scoring system included combined responses ranging from 0 to 1 and 0 to 2. In some studies outside the neurocognitive field, a continuous scale has been proven to yield robust and valid results compared to the binary response (Markon et al., 2011; Munson et al., 2017). Additionally, some items in this test might be better suited to binary responses, such as number 7 (“no missing or added number”), where the answer is either ‘yes’ or ‘no’. Nevertheless, we researchers aimed to elicit an in-between response by introducing the middle option for participants to choose. This addition is intended to facilitate a gradient of responses to a question as simple as “How many numbers are missing/added compared to others?” Furthermore, the rating scale analysis performed well, indicating that this scoring system did not cause any confusion among the raters. Previous studies have affirmed that a simpler scoring system is preferable because a rigid system may introduce limitations to the test's ability to capture subtle errors (Borson et al., 1999). Additionally, it has been found to decrease the test’s ability to perform rapid screening (Mainland et al., 2014). Therefore, maintaining the simplicity and ease of execution of this test, without compromising its quality, is essential.

Linacre (2022) emphasized the importance of declaring the rater type in an evaluation process, as this is associated with the assumption of rater agreement. In this study, it was assumed that raters followed screening processes with rigid agreement. However, the inter-rater agreement results in this study (74%) did not meet the researchers' expectations, as the anticipated agreement was less than 90%. Notably, Rater 3 was identified as the most stringent examiner, exhibiting a 0.33 logit difference from Rater 2. While these discrepancies may be relatively small, for practical purposes, this information is valuable for evaluating raters’ behavior during the screening process. For instance, it provides insights into whether some raters display extremely high strictness or leniency.

Lastly, in predicting the test using participants' logit scores from CDT, we observed that this test demonstrated better accuracy, yielding significant results in correctly identifying participants with normal cognitive function and those with mild cognitive impairment and mild dementia based on GDS criteria. On the other hand, Chiu et al. (2008) discovered that the CDT-18 points scoring method was optimum for discriminating between normal participants and those with mild cognitive dementia. Despite its inability to provide specific information about participants’ cognitive decline, it proved effective in swiftly distinguishing between normal and suspected patients with cognitive decline as a prior information before further assessment.

We acknowledge the limitations of this study, as it identified 603 empirical bias terms in the rater-participant interaction out of a total of 9,045 terms. Additionally, the sample size of raters in this study was relatively small. Therefore, it is recommended that this study be replicated with a larger number of raters, and efforts be made to reduce empirical bias in the rater-participant interaction. Furthermore, this study presents notable implications that encourage researchers to apply MFRM to other screening tests. It is hoped that this will provide a more robust understanding of test construction and its application in clinical settings.

## Conclusion

The *Clock Drawing Test-18-point* revised version demonstrated overall decent validity and reliability test based on the Many-Facets Rasch model perspective. However, it exhibited some misfits in specific items, namely, items 1, 15, and 6. While items 5, 4, and 2 proved to be highly challenging, items 8, 6, and 3 were more accessible. The researchers also identified items 15, 8, and 9 as the best items for distinguishing between lower and higher-ability participants. The strata reliability indicated that this test effectively covered the diversity of participants' abilities in this study. According to the ROC analysis, the test is accurate as a screener to distinguish between healthy participants and those with mild cognitive impairment and mild dementia based on GDS criteria as preliminary data before the actual diagnostic test. However, all raters in this test exhibited inter-rater agreement below the reference standard (< 90%).

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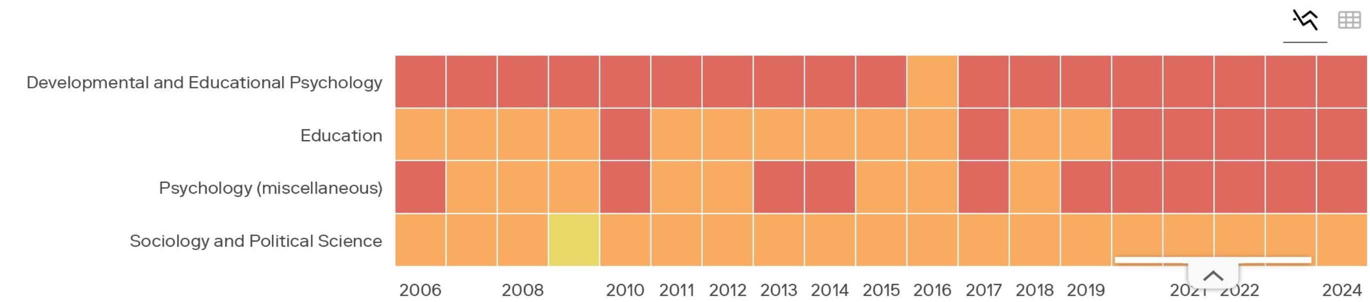


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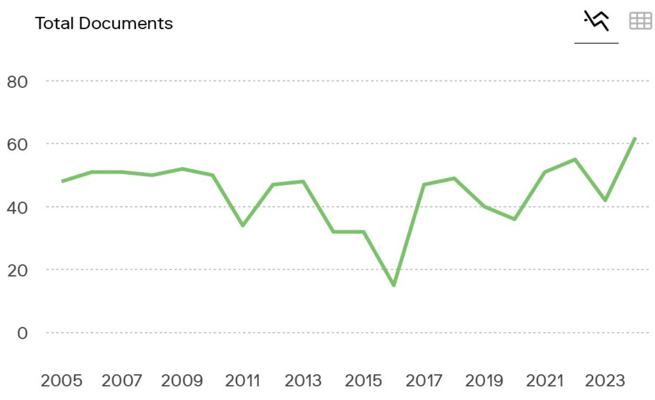
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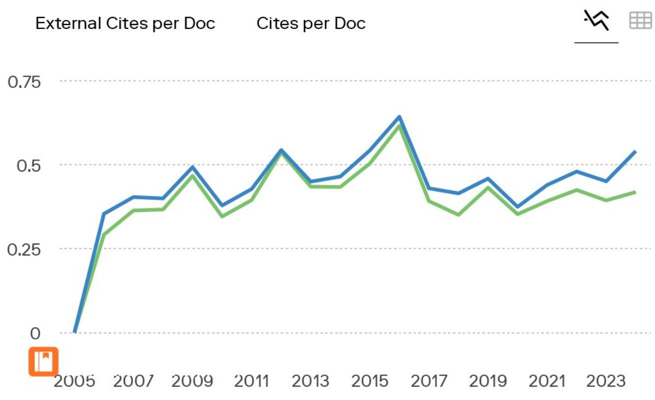
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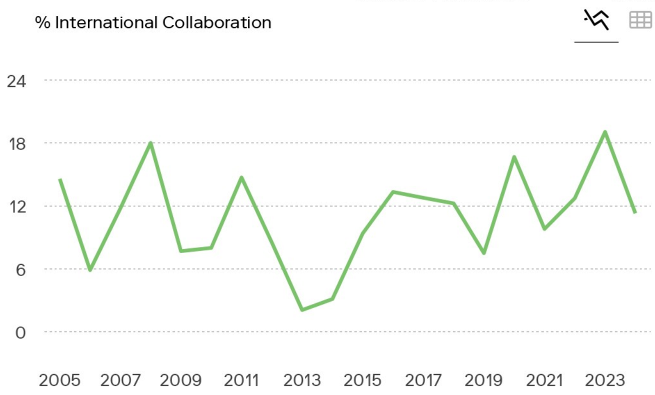
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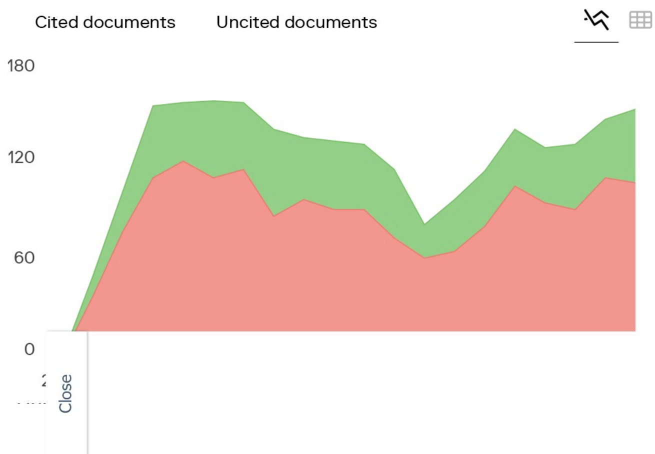
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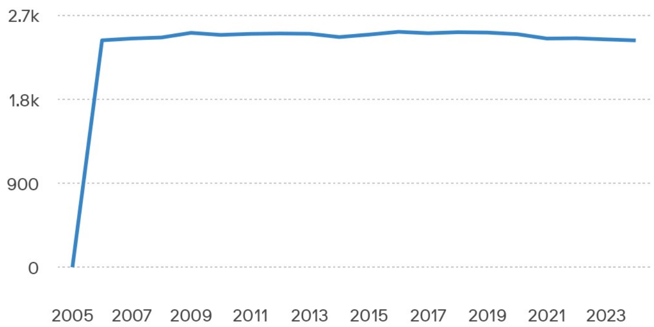


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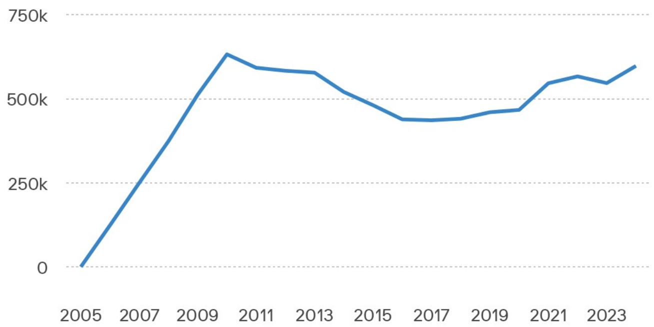


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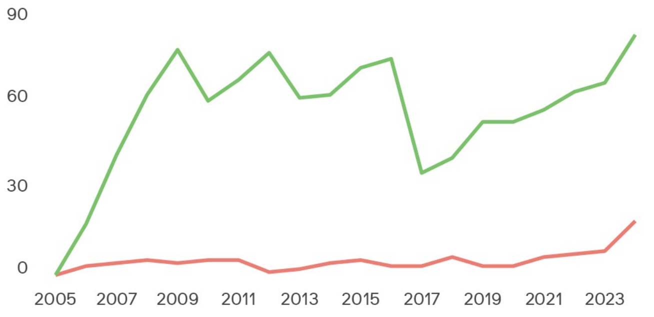


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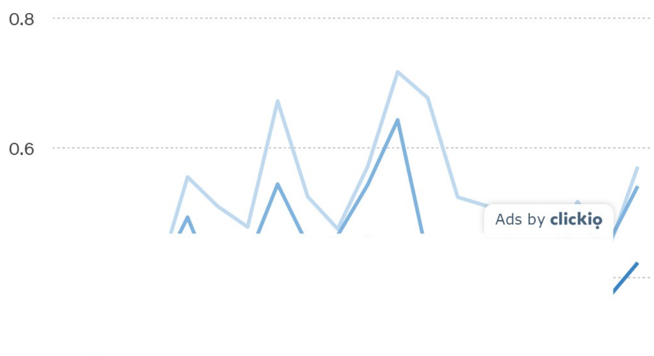

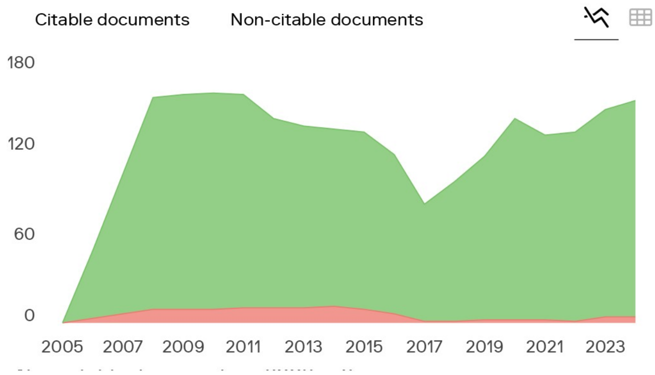


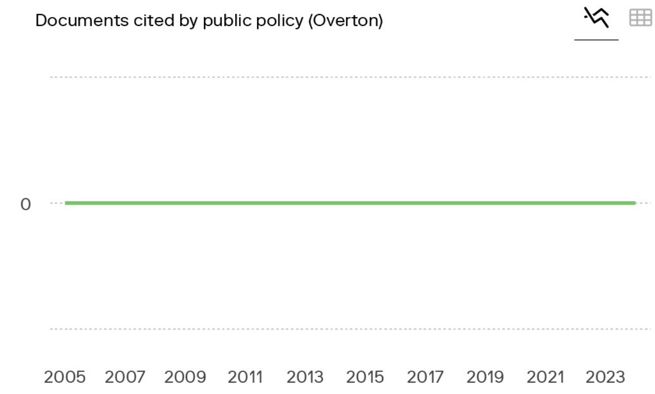


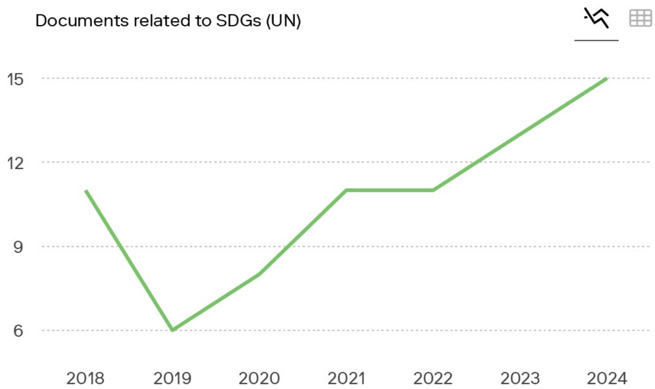



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