World Journal of Psychiatry

World J Psychiatry 2022 August 19; 12(8): 1002-1114





Contents

Monthly Volume 12 Number 8 August 19, 2022

EDITORIAL

1002 Meeting employees where they are: The rise of workplace mental health services

Nov G, Shah RN

REVIEW

1004 Does COVID-19 related symptomatology indicate a transdiagnostic neuropsychiatric disorder? -Multidisciplinary implications

Goldstein Ferber S, Shoval G, Zalsman G, Weller A

ORIGINAL ARTICLE

Case Control Study

1016 Antidepressants combined with psychodrama improve the coping style and cognitive control network in patients with childhood trauma-associated major depressive disorder

Yu RQ, Tan H, Wang ED, Huang J, Wang PJ, Li XM, Zheng HH, Lv FJ, Hu H

Can the prediction model using regression with optimal scale improve the power to predict the 1031 Parkinson's dementia?

Byeon H

Observational Study

1044 Worldwide suicide mortality trends (2000-2019): A joinpoint regression analysis

Ilic M, Ilic I

1061 Peripartum depression and its predictors: A longitudinal observational hospital-based study

Hamed SA, Elwasify M, Abdelhafez M, Fawzy M

1076 Cross-sectional survey following a longitudinal study on mental health and insomnia of people with sporadic COVID-19

Li XJ, Guo TZ, Xie Y, Bao YP, Si JY, Li Z, Xiong YT, Li H, Li SX, Lu L, Wang XQ

1088 Fear of COVID-19 and emotional dysfunction problems: Intrusive, avoidance and hyperarousal stress as key mediators

Falcó R, Vidal-Arenas V, Ortet-Walker J, Marzo JC, Piqueras JA, PSICO-RECURSOS COVID-19 Study Group

LETTER TO THE EDITOR

1102 Difference between treatment-resistant schizophrenia and clozapine-resistant schizophrenia

Tseng PT, Chen MH, Liang CS

1105 Genetics of adult attachment and the endogenous opioid system

Troisi A

World Journal of Psychiatry

Contents

Monthly Volume 12 Number 8 August 19, 2022

1108 Cardiotoxicity of current antipsychotics: Newer antipsychotics or adjunct therapy? Liu Z, Zhang ML, Tang XR, Li XQ, Wang J, Li LL

1112 Underlying disease may increase mortality risk in users of atypical antipsychotics Li ZP, You YS, Wang JD, He LP



II

Contents

Monthly Volume 12 Number 8 August 19, 2022

ABOUT COVER

Editorial Board Member of World Journal of Psychiatry, Rajiv Gupta, MD, Director, Professor, Department of Psychiatry, Institute of Mental Health, Rohtak 124001, Haryana, India. rajivguptain2003@yahoo.co.in

AIMS AND SCOPE

The primary aim of World Journal of Psychiatry (WJP, World J Psychiatry) is to provide scholars and readers from various fields of psychiatry with a platform to publish high-quality basic and clinical research articles and communicate their research findings online.

WJP mainly publishes articles reporting research results and findings obtained in the field of psychiatry and covering a wide range of topics including adolescent psychiatry, biological psychiatry, child psychiatry, community psychiatry, ethnopsychology, psychoanalysis, psychosomatic medicine, etc.

INDEXING/ABSTRACTING

The WJP is now abstracted and indexed in Science Citation Index Expanded (SCIE, also known as SciSearch®), Current Contents/Clinical Medicine, Journal Citation Reports/Science Edition, PubMed, PubMed Central, Reference Citation Analysis, China National Knowledge Infrastructure, China Science and Technology Journal Database, and Superstar Journals Database. The 2022 Edition of Journal Citation Reports® cites the 2021 impact factor (IF) for WJP as 3.500; IF without journal self cites: 3.313; 5-year IF: 7.380; Journal Citation Indicator: 0.62; Ranking: 89 among 155 journals in psychiatry; and Quartile category: Q3.

RESPONSIBLE EDITORS FOR THIS ISSUE

Production Editor: Yu-Xi Chen, Production Department Director: Xu Guo; Editorial Office Director: Jia-Ping Yan.

NAME OF JOURNAL

World Journal of Psychiatry

ISSN 2220-3206 (online)

LAUNCH DATE

December 31, 2011

FREQUENCY

Monthly

EDITORS-IN-CHIEF

Rajesh R Tampi, Ting-Shao Zhu, Panteleimon Giannakopoulos

EDITORIAL BOARD MEMBERS

https://www.wjgnet.com/2220-3206/editorialboard.htm

PUBLICATION DATE

August 19, 2022

COPYRIGHT

© 2022 Baishideng Publishing Group Inc

INSTRUCTIONS TO AUTHORS

https://www.wjgnet.com/bpg/gerinfo/204

GUIDELINES FOR ETHICS DOCUMENTS

https://www.wignet.com/bpg/GerInfo/287

GUIDELINES FOR NON-NATIVE SPEAKERS OF ENGLISH

https://www.wjgnet.com/bpg/gerinfo/240

PUBLICATION ETHICS

https://www.wjgnet.com/bpg/GerInfo/288

PUBLICATION MISCONDUCT

https://www.wignet.com/bpg/gerinfo/208

ARTICLE PROCESSING CHARGE

https://www.wignet.com/bpg/gerinfo/242

STEPS FOR SUBMITTING MANUSCRIPTS

https://www.wjgnet.com/bpg/GerInfo/239

ONLINE SUBMISSION

https://www.f6publishing.com

© 2022 Baishideng Publishing Group Inc. All rights reserved. 7041 Koll Center Parkway, Suite 160, Pleasanton, CA 94566, USA E-mail: bpgoffice@wjgnet.com https://www.wjgnet.com



Submit a Manuscript: https://www.f6publishing.com

World J Psychiatry 2022 August 19; 12(8): 1031-1043

DOI: 10.5498/wjp.v12.i8.1031 ISSN 2220-3206 (online)

ORIGINAL ARTICLE

Case Control Study

Can the prediction model using regression with optimal scale improve the power to predict the Parkinson's dementia?

Haewon Byeon

Specialty type: Psychiatry

Provenance and peer review:

Unsolicited article; Externally peer reviewed.

Peer-review model: Single blind

Peer-review report's scientific quality classification

Grade A (Excellent): 0 Grade B (Very good): 0 Grade C (Good): C, C Grade D (Fair): 0 Grade E (Poor): 0

P-Reviewer: Dabbakuti JRKKK, India; Kujawski S, Poland

Received: March 16, 2022

Peer-review started: March 16, 2022 First decision: June 11, 2022 **Revised:** June 15, 2022 Accepted: July 11, 2022 Article in press: July 11, 2022

Published online: August 19, 2022



Haewon Byeon, Department of Medical Big Data, Inje University, Gimhae 50834, South Korea

Corresponding author: Haewon Byeon, DSc, PhD, Associate Professor, Director, Department of Medical Big Data, Inje University, No. 329, Gimhae 50834, South Korea. bhwpuma@naver.com

Abstract

BACKGROUND

Efficiently detecting Parkinson's disease (PD) with dementia (PDD) as soon as possible is an important issue in geriatric medicine.

AIM

To develop a model for predicting PDD based on various neuropsychological tests using data from a nationwide survey conducted by the Korean Centers for Disease Control and Prevention and to present baseline data for the early detection of PDD.

METHODS

This study comprised 289 patients who were 60 years or older with PD [110 with PDD and 179 Parkinson's Disease-Mild Cognitive Impairment (PD-MCI)]. Regression with optimal scaling (ROS) was used to identify independent relationships between the neuropsychological test results and PDD.

RESULTS

In the ROS analysis, Korean version of mini mental state ex-amination (MMSE) (KOREAN version of MMSE) (b = -0.52, SE = 0.16) and Hoehn and Yahr staging (b = 0.44, SE = 0.19) were significantly effective models for distinguishing PDD from PD-MCI (P < 0.05), even after adjusting for all of the Parkinson's motor symptom and neuropsychological test results. The optimal number of categories (scaling factors) for KOREAN version of MMSE and Hoehn and Yahr Scale was 10 and 7, respectively.

CONCLUSION

The results of this study suggest that among the various neuropsychological tests conducted, the optimal classification scores for KOREAN version of MMSE and Hoehn and Yahr Scale could be utilized as an effective screening test for the early discrimination of PDD from PD-MCI.

Key Words: Hoehn and Yahr staging; Optimal scale; Parkinson's dementia; Mini mental state ex-amination; Montreal Cognitive Assessment

©The Author(s) 2022. Published by Baishideng Publishing Group Inc. All rights reserved.

Core Tip: Although a general linear model can be constructed if all of the variables used in the analysis are numeric, it is difficult to fit the data when the variables are nominal. We build a regression model using the transform variables obtained by iteratively using alternating least squares to compute the optimal scaling. We developed a predictive model to discriminate Parkinson's disease with dementia from Parkinson's Disease-Mild Cognitive Impairment based on the results of nine neuropsychological tests and found that only Korean version of mini mental state examination and Hoehn and Yahr Scale could be successfully employed to this end.

Citation: Byeon H. Can the prediction model using regression with optimal scale improve the power to predict the Parkinson's dementia? World J Psychiatry 2022; 12(8): 1031-1043

URL: https://www.wjgnet.com/2220-3206/full/v12/i8/1031.htm

DOI: https://dx.doi.org/10.5498/wjp.v12.i8.1031

INTRODUCTION

As the longevity of the South Korean population increases, so does the proportion of advanced-aged individuals[1] along with the incidence of chronic degenerative diseases[1]. For this reason, the importance of prevention and early treatment of degenerative dis-eases in old age should be emphasized. Parkinson's disease (PD) is a representative neurodegenerative disease caused by damaged nerve cells that secrete dopamine in the sub-stantia nigra. However, researchers have paid less attention to PD than dementia or stroke in terms of health science because its incidence rate is only 1% in the older adult population (≥ 65 years old) and its prevalence rate is lower than for dementia or stroke. However, the number of PD patients is steadily increasing in the aged population. The Health Insurance Review and Assessment Service (HIRAS) (2019)[2] reported that the number of patients diagnosed with PD steadily increased from 61565 in 2010 to 100716 in 2018 and predicted that the number of PD patients will double in 2030 compared to 2005 at this rate. In particular, the number of older adults with PD is expected to increase even more in South Korea considering that by 2050, the proportion of the older adult population in South Korea will be 35.9%, the second-highest after Japan (40.1%)[3]. Consequently, the effective early detection of PD is an important topic in the field of geriatric medicine.

PD is a motor disorder comprising a combination of weakness, tremor, and rigidity. However, over the past 20 years, other symptoms including autonomic nerve disorder, affective and sensory disorders such as depressive disorders, and cognitive impairment have been reported[1-7]. Many previous studies [8-10] have reported that 20%-57% of patients develop mild cognitive impairment (MCI) within 5 years from the date of being diagnosed with PD. MCI refers to a state in which cognitive decline is observed without accompanying a decline in activities of daily living (ADL). It is a pre-clinical state of PD with dementia (PDD) and it is an intermediate stage from normal to PDD. Previous follow-up studies also have revealed that approximately 10% to 15% of MCI patients transited to dementia every year [7]. It means that they are highly vulnerable to dementia and it was much higher than the annual dementia incident rate of healthy older adults (65 years or older)[7]. It is the earliest stage of dementia that can be detected in clinical examination, and it is clinically very important because it is possible to maximize the therapeutic effect[7]. Neuropsychological screening battery, cognitive assessment, autonomic function, and other tests have been carried out to objectively assess the clinical status of PD accompanying MCI [11]. However, it is difficult to distinguish MCI from aging or mild dementia only using these screening tests[11]. To make it more challenging, it can be misdiagnosed with progressive supranuclear palsyparkinsonism (PSP-P) when a patient suffers from PD and cognitive deficit at the same time[12,13].

Compared to the United States and Europe, South Korea currently has insufficient epidemiological data on cognitive impairment in old age. Although community-based studies on PD conducted in South Korea have focused on patients in small and medium-sized cities, prediction models based on a nationwide epidemiological survey have not yet been developed [14-17]. Although a general linear model (GLM) for PD can be constructed if all of the variables used in the analysis are numeric, it is difficult to fit the data when the variables are ordinal or nominal. An alternative method to overcome this limitation is to build a regression model with an optimal scale (optimal regression).

Optimal scaling is based on the prediction theory (also known as the quantification theory) developed by considering how to quantify qualitative variables to enable optimal data analysis rather than simply ranking them and interpreting the results. Optimal scaling has been mainly used in social science fields such as psychology when proving causality is important[18-20]. However, it has only been used in a small number of studies in the cognitive science field. Identifying neuropsychological tests (*e.g.*, cognitive and de-pression tests) and Parkinson's motor symptom tests that are effective in discriminating PDD from PD-MCI by using regression with optimal scaling (ROS) and checking the optimal classification scores of the tests is clinically meaningful. However, it has only been used in a small number of studies in the cognitive science field. The objectives of the present study were to develop a model for predicting PDD based on various neuropsychological tests using data from a National Biobank of Korea data.

MATERIALS AND METHODS

Data source

Approval for the study was received from the Distribution Committee (No. KBN-2019-1327) and the Research Ethics Review Committee of the National Biobank of Korea under the Korean Centers for Disease Control and Prevention (No. KBN-2019-005). Epidemiologic data on patients with PD were collected from 14 tertiary care providers nationwide from January to December 2015 under the supervision of the Korean Centers for Disease Control and Prevention. PDD has been designated as idiopathic Parkinson's dis-ease according to the diagnostic criteria of the United Kingdom Parkinson's Disease Society Brain Bank[21]. The diagnostic criteria for probable PDD have been suggested by the Movement Disorder Society Task Force [22]. When causes of cognitive impairment other than PD (e.g., hydrocephalus and vascular Parkinsonism) were found in magnetic resonance imaging scans, the subject was excluded from the study (see Byeon[23] for more details). PD-MCI was diagnosed by neuropsychologists according to the criteria of the International Working Group on MCI[24]. Health surveys were conducted by using computer-assisted personal interviews. We analyzed the PD epidemiologic data comprising demographic information, any family history of PD, health-affecting behaviors (e.g., smoking), disease history (e.g., diabetes), and Parkinson's motor symptoms (e.g., rigidity) and neuropsychological characteristics (e.g., cognitive level). The variables and their values are reported in Table 1. Thus, data on 289 patients with PD (110 PDD and 179 PD-MCI) who were 60 years or older were used in the study.

Variable measurement

The label was defined as PDD confirmed by medical diagnosis. To understand the difference in the general characteristics of subjects according to the demographic variables (*e.g.*, age, sex, and education level), medical history (*e.g.*, hypertension), and family history [*e.g.*, PD and Alzheimer's disease (AD)].

Explanatory variables (neuropsychological tests) included scores from the Hoehn and Yahr (H&Y) staging[25], Global Clinical Dementia Rating (CDR)[26], Schwab and England Activities of Daily Living [27], the Korean Instrumental ADL (K-IADL)[28], the Unified PD Rating Scale (UPDRS) total[22], the UPDRS motor[22], the Korean Mini-Mental State Examination (KOREAN version of MMSE)[29], and the Korean-Montreal Cognitive Assessment (K-MoCA)[30]. Hoehn and Yahr Scale[25] is a screening test to determine the stage of PD and is measured by clinicians. The score ranges from 1 to 5, and a higher score indicates that the symptoms of PD are more severe.

CDR[26] is a screening test to determine the stage of dementia and is measured by clinicians. The possible outcomes are 0, 0.5, 1, 2, 3, 4, and 5 points, and a higher score means more severe dementia. Schwab and England ADL[27] is a screening test for physical impairment. It is evaluated by clinicians to measure indices regarding independent performance in the daily activities of PD patients. The score ranges from 0 to 100, and a higher score is interpreted as a lower functional impairment. K-IADL[28] is a cognitive screening test that measures skills and behaviors necessary for social life such as "money management" and "phone use". It consists of eleven items that can score between 0 and 3, and a higher score means higher functional impairment. UPDRS[22] is an overall evaluation scale for the symptoms of PD and consists of four segments (mentation/behavior/mood, ADL, motor examination, and dyskinesia). The test is conducted by a clinician, and a higher score is interpreted as a higher degree of disability. KOREAN version of MMSE[29] is a test for screening cognitive disorders such as dementia and consists of time orientation, spatial orientation, memory registration (input), calculation and attention, memory recall, and language items. The total score is 30 points, and the cut-off score is 24 points. A lower score indicates more severe cognitive impairment. K-MoCA[30] is a test for screening MCI. The total score is 30 points, and people with 22 points and above are interpreted as normal. A lower score is understood as more severe cognitive impairment.

Regression with optimal scale

If all the variables (*e.g.*, Independent variables, dependent variables, and confounding variables) used in the analysis are numeric variables (quantitative variables), the GLM can be used. However, if it is an ordinal or nominal variable, it is difficult to use the general linear regression model because these variable types do not meet the assumptions of the regression models and error terms. Therefore, analysis can be con-ducted by deriving an optimized linear regression equation of transformed variables

Tahl	م 1 N	lageuramant	of vari	ahlae

Factors	Measurement	Characteristics	
Demographic factors	Sex	Male or female	
	Age	60-74, 75+	
	Mainly used hand	Left hand or right hand	
	Education level	Middle school graduate and below or high school graduate and above	
Family history of the disease	Parkinson's disease; Alzheimer's disease	Yes or no; Yes or no	
Health behaviors	Pack-years (smoking)	Non-smoking, 1-20, 21-40, 41-60, or ≥ 61 pack-years	
	Coffee-drinking	Yes or no	
	Mean coffee intake per day (cups/d)	No, ≤ 1 , 2-3, or ≥ 4 cups	
	Coffee drinking period (yr)	No, ≤ 5 , 6-9, or ≥ 10	
Disease history	Carbon monoxide poisoning	Yes or no	
	Diabetes	Yes or no	
	Alcoholism	Yes or no	
	Hyperlipidemia	Yes or no	
	Traumatic brain injury	Yes or no	
	Hypertension	Yes or no	
Exercise characteristics related to Parkinson's disease	Tremor	Yes or no	
relatedmotor signs	Rigidity	Yes or no	
	Akinesia/bradykinesia	Yes or no	
	Postural instability	Yes or no	
	Late motor complications	Yes or no	
Neuropsychological test	K-MoCA	Continuous variable	
	K-MMSE	Continuous variable	
	Global CDR score	Continuous variable	
	Sum of boxes in CDR	Continuous variable	
	Hoehn and Yahr staging	Continuous variable	
	UPDRS (Total UPDRS score)	Continuous variable	
	UPDRS (Motor UPDRS score)	Continuous variable	
	K-IADL	Continuous variable	
	Schwab and England ADL	Continuous variable	

K-MoCA: Korean version of montreal cognitive assessment; K-MMSE: Korean version of mini mental state examination; CDR: Clinical dementia rating; UPDRS: Untitled parkinson disease rating; K-IADL: Korean version of instrumental activities of daily living; Schwab and England ADL: Schwab and England avtivities of daily living scale.

by repeatedly performing optimal scaling based on the alternating least squares method.

1034

It is a way to estimate parameters for the linear relationship between independent and dependent variables using data on each variable. The estimated general linear regression model is presented as follows[19]:

 $Yi = \alpha + \beta X_i + \varepsilon_i$

Yi = dependent variable

Xi = independent variable (Equation 1)

 $\varepsilon i = error \ term$

 α, β = parameter to estimate

When the assumptions for the error term, such as "the expected value of the error term shall be 0" and "it shall follow a normal distribution and all observations shall have the same variance", parameters are estimated by using the least-squares and other methods to determine the relationship between the independent and dependent variables. The least-squares method is used to obtain parameter estimates ($\alpha^{\hat{}}$ and $\beta^{\hat{}}$) that minimize the sum-of-squared residuals, where the residual ($\varepsilon_{-}i$) is equal to the difference between the actual observations ($Y_{-}i$) and the predicted values of the dependent variables (($Y_{-}i$) $^{\hat{}}$) (($Y_{-}i$) $^{\hat{}}$) = $\alpha^{\hat{}}$ + $\beta^{\hat{}}$ $X_{-}i$)).

In this study, ROS consisted of three stages. The first is the data transformation stage. After normalizing k categorical indicators for the nth variable by vectorizing them, all of the variables are treated as numeric variables. Subsequently, they are optimized repeatedly by using the calculated categorical quantification values and regression coefficients. The second stage is updating the categorical quantification vector by considering the scale level (*i.e.*, whether the variables are nominal, ordinal, or numeric) and calculating the regression coefficient vector. The third stage is to establish convergence by repeatedly calculating the categorical quantification vector and the regression coefficient vector until they satisfy the predetermined convergence condition[19].

ROS transforms each variable appropriately by considering its scale in the GLM. When dependent variable Y is transformed to $\theta(Y)$ and independent variable X to $\sigma(X)$, the resulting parameters are the intercept and slope of a GLM (linear regression) equation formed by minimizing the sum-of-squares (SSQ) of the error[19] as follows: $minSSQ[\theta(Y)-\beta\sigma(X)]$ (Equation 2).

The conversion variable has a standardization constraint. Minimizing the SSQ error is achieved by least-squares regressing the transformed variables [e.g., $\theta(Y)$, $\sigma_1(X_1)$, ..., $\sigma_n(X_n)$)]. The ROS analysis with standardization constraints is written as

$$minSSQ(\theta(Y) - \sum_{i=1}^{n} \beta_i \sigma_i(X_i))$$
 (Equation 3).

ROS was used to identify the independent relationship between each test and PDD. The analysis results were presented with a regression coefficient, odds ratio, 95% confidence interval (CI), quantification index, and standard error by bootstrap (n = 999). General characteristics of the subjects and the prevalence of PD were analyzed using the Chi-square test.

When independent significance was confirmed in the ROS, the Cochran-Armitage (CA) trend test was used to determine whether the p values had a linear trend based on the reference group as follows [31]:

$$T = \frac{\sum_{i=1}^{R} n_{i,1}(R_i - \bar{R})}{\sqrt{p_{i,1}(1 - p_{i,1})s^2}} s^2 = \sum_{i=1}^{R} n_{i,.}(R_i - \bar{R})^2 \text{ (Equation 4)}.$$

The analysis of ROS was conducted by using CatReg Software version 3.0 (the Data Theory Scaling System Group, Leiden, The Netherlands).

RESULTS

Characteristics of the participants and the prevalence of PD

The results of χ^2 tests show that age, handed, PD family history, gender, the highest level of education, AD family history, hypertension, traumatic brain injury history, stroke history, carbon monoxide poisoning history, hyperlipidemia, and diabetes were not significantly different between PDD and PD-MCI (Table 2). Therefore, the subjects in this study did not have statistically significant demographic or health differences between the groups.

Table 3 reports the data and Figure 1 shows a bag plot for visualizing the spread, location, outliers and skewness.

The neuropsychological test results of PD-MCI and PDD are compared (Table 4). As a result of the independent t-test, KOREAN version of MMSE, K-MoCA, Total UPDRS score, CDR (sum of boxes), K-IADL, Hoehn and Yahr staging, Motor UPDRS score, and Schwab and England ADL were not significantly different between PDD and PD-MCI (P < 0.05).

The analysis results of ROS

The analysis results of ROS are summarized in Table 5. Hoehn and Yahr Scale (b = 0.44, SE = 0.19) and KOREAN version of MMSE (b = -0.52, SE = 0.16) were significantly effective for distinguishing PDD from PD-MCI even after adjusting for all of test results (P < 0.05). The regression model was adjusted for demographic factors, family disease history, health-affecting behaviors, dis-ease history, Parkinson's motor symptoms, and neuropsychological test.

Quantification scores for KOREAN version of MMSE and Hoehn and Yahr Scale are reported in Tables 6 and 7, respectively, and presented in Figures 2 and 3, respectively. The results show that the optimal number of categories (scaling factors) for KOREAN version of MMSE and Hoehn and Yahr Scale was 10 and 7, respectively. The odds ratios (ORs) and 95%CIs for the optimal categories of KOREAN version of MMSE and Hoehn and Yahr Scale are reported in Table 8. When distinguishing PDD from PD-MCI, PD-MCI patients who had 23 or 24 points for KOREAN version of MMSE had a 4.5-fold higher risk of PDD than those who had 25 or higher. Moreover, those who scored 21 or 22, 19 or 20, 15 to 18, and 3 to 14 points had a 2.7-fold, 13.3-fold, 22.4-fold, and 55-fold higher risk of developing PDD, respectively, than those who had 25 or higher. The results of the CA Trend test show a significant relationship (*P* for Trend < 0.001) between the increase in OR and the KOREAN version of MMSE score (optimal categories score).

Table 2 General c	haracteristics of th	ha subjects based or	n Parkinson's dise	ase with dementia. n (%)
Table 2 General C	naraciensiics or ii	ile Subilects Daseu Ol	II Paikilisuli s uised	15e Willi Gelliellia. 11 (70)

Variables	PD-MCI (n = 179)	PDD (n = 110)	P value
Age			0.168
60-74	117 (65.0)	63 (35.0)	
75+	62 (56.9)	47 (43.1)	
Sex			0.550
Male	78 (63.9)	44 (36.1)	
Female	101 (60.5)	66 (39.5)	
Education level			0.072
Middle school graduate and below	110 (58.2)	79 (41.8)	
High school graduate and above	69 (69.0)	31 (31.0)	
Family history of the Parkinson's disease			0.600
No	144 (64.3)	80 (35.7)	
Yes	12 (70.6)	5 (29.4)	
Family history of the Alzheimer's disease			0.285
No	130 (63.4)	75 (36.6)	
Yes	8 (80.0)	2 (20.0)	
Carbon monoxide poisoning			0.743
No	158 (62.5)	95 (37.5)	
Yes	10 (66.7)	5 (33.3)	
Traumatic brain injury			0.277
No	158 (62.0)	97 (38.0)	
Yes	10 (76.9)	3 (23.1)	
Diabetes			0.508
No	144 (64.0)	81 (36.0)	
Yes	35 (59.3)	24 (40.7)	
Hypertension			0.304
No	110 (65.5)	58 (34.5)	
Yes	69 (59.5)	47 (40.5)	
Hyperlipidemia			0.220
No	155 (61.8)	96 (38.2)	
Yes	24 (72.7)	9 (27.3)	

DISCUSSION

In this study, KOREAN version of MMSE and Hoehn and Yahr Scale could independently differentiate PDD from PD-MCI even after adjusting for all of the PD's test results. Moreover, when the ROS (optimal classification scores) were calculated, the increase in OR according to all of the categories showed a significant proportional trend.

It is not easy to accurately detect and diagnose PSP-P by identifying the pattern of PD-MCI in PDD by using neuropsychological tests[26]. First, it is difficult to determine whether dementia is the cause of a patient's cognitive impairment symptoms[27] because patients with PD often take a variety of medications (e.g., anticholinergics, amantadine, anxiolytics, and sedatives) and can experience temporary cognitive decline or confusion (easily mistaken for dementia) as side effects of the medications[32]. Second, cognitive impairment can occur temporarily due to endocrine imbalance due to depression, electrolyte imbalance, and/or dehydration; systemic diseases; or infection[22]. Third, even if dementia is diagnosed, it is necessary to effectively differentiate it from other types of irreversible dementia such as Alzheimer's disease or, especially, dementia with Lewy bodies[22]. Hence, it is necessary to develop predictive models that can more efficiently discriminate PDD from PD-MCI as

Tahla 3 Ra	eulte of the	neuropsych	nological	l nrofiles

Results	K-MMSE	K-MoCA	Global CDR score	Sum of boxes in CDR	K-IADL	Total UPDRS	Motor UPDRS	H&Y staging	ADL
Mean	22.73	16.27	0.67	2.80	1.90	43.56	25.33	2.45	74.40
Standardized mean error	0.32	0.44	0.03	0.22	0.26	2.02	0.77	0.04	1.42
Standard deviation	5.51	6.33	0.56	3.49	4.08	23.77	12.59	0.78	18.30
Minimum	3	0	0	0	0	0.18	2.0	1.0	10
Maximum	30	27	4.0	25.0	28.0	130.00	74.0	5.0	100

K-MoCA: Korean version of montreal cognitive assessment; K-MMSE: Korean version of mini mental state examination; CDR: Clinical dementia rating; UPDRS: Untitled parkinson disease rating; H&Y staging: Hoehn and Yahr staging; K-IADL: Korean version of instrumental activities of daily living; Schwab and England ADL: Schwab and England avtivities of daily living scale.

Table 4 Result of the neuropsychological profiles based on Parkinson's disease with dementia, mean ± SD						
Variables	PD-MCI (n = 179)	PDD (n = 110)	P value			
K-MMSE	24.3 ± 3.4	18.8 ± 5.6	< 0.001			
K-MoCA	19.4 ± 4.9	11.9 ± 5.4	< 0.001			
CDR (sum of boxes)	1.6 ± 1.4	5.1 ± 4.9	< 0.001			
K-IADL	1.3 ± 2.9	3.0 ± 5.4	0.001			
UPDRS (Total UPDRS score)	36.4 ± 17.9	56.1 ± 27.2	< 0.001			
UPDRS (Motor UPDRS score)	22.6 ± 10.1	29.4 ± 14.6	< 0.001			
H&Y staging	2.2 ± 0.6	2.7 ± 0.8	0.001			
Schwab and England ADL	80.0 ± 14.4	65.6 ± 19.8	< 0.001			

K-MoCA: Korean version of montreal cognitive assessment; K-MMSE: Korean version of mini mental state examination; CDR: Clinical dementia rating; UPDRS: Untitled parkinson disease rating; H&Y staging: Hoehn and Yahr staging; K-IADL: Korean version of instrumental activities of daily living; Schwab and England ADL: Schwab and England avtivities of daily living scale.

> well as other types of dementia while simultaneously considering the results of several neuropsychological tests related to cognitive impairment.

> Nevertheless, in most of the previous studies, evaluating the predictive performance for PDD was conducted by comparing individual diagnostic performances in terms of accuracy and reliability [30,31, 33,34]. The results of the present study suggest that among the various neuropsychological tests examined, the optimal classification scores by MMSE-K and Hoehn and Yahr Scale show that these two tests could be utilized for effective early discrimination of PDD from PD-MCI. Moreover, they could be used to clinically determine whether PD-MCI patients will develop PDD or whether existing PDD patients are getting worse. Conducting these tests when a PD-MCI patient visits the hospital (or Public Health Center) for the first time provides baseline information and carrying them out sequentially at regular visits can be used to recognize clinically meaningful changes.

> Although it is very important to efficiently distinguish PDD from other diseases showing symptoms of PD as soon as possible, PD can only be accurately diagnosed through pathological examination with autopsy[6]. Dopamine transporter imaging has been reported as an effective test for diagnosing PDD at an early stage[35], but it is too expensive to be used as a screening test in the primary care setting. As a result, it is diagnosed through an interview on the symptoms of a patient and an examination of a specialist along with a cognitive screening test such as KOREAN version of MMSE in the clinical practice.

> However, Rizzo et al[36] reported that the misdiagnosis rate of dyskinesis was at least 20% even for neurologists with extensive experience in dyskinesias. Therefore, to accurately diagnose PD-MCI, a specialist must have a broad perspective to comprehensively consider the symptoms of a patient (e.g., resting tremor, bradypragia, postural changes, and gait abnormalities), living environment, presence of trauma, lifestyle, and occupation as well as the results of cognitive screening tests. Particularly, since cognitive issues and dyskinesias (e.g., bradypragia, resting tremor, and ankylosis) are slowly

> > 1037

Table 5 Results of regression with optimal scale						
Test	b	SE by boost 1	df	F	P value	
K-MMSE	-0.522	0.168	2	9.684	< 0.001	
KMoCA	-0.206	0.238	3	0.750	0.527	
CDR (Global CDR score)	0.127	0.269	1	0.222	0.639	
CDR (sum of boxes)	-0.271	0.412	3	0.431	0.732	
K-IADL	0.237	0.224	2	1.119	0.334	
UPDRS (Total UPDRS score)	0.433	0.444	3	0.949	0.423	
UPDRS (Motor UPDRS score)	-0.338	0.330	3	1.045	0.380	
H&Y staging	0.440	0.197	3	5.008	0.004	
Schwab and England ADL	0.353	0.333	2	1.123	0.333	

The regression model was adjusted for demographic factors, family disease history, health behaviors, disease history, Parkinson's disease-related motor signs and neuropsychological test. 1 SE by boost: Standard error by bootstrap (with n = 1000); K-MoCA: Korean version of montreal cognitive assessment; K-MMSE: Korean version of mini mental state examination; CDR: Clinical dementia rating; UPDRS: Untitled parkinson disease rating; H&Y staging; Hoehn and Yahr staging; K-IADL: Korean version of instrumental activities of daily living; Schwab and England ADL: Schwab and England avtivities of daily living scale.

Table 6 Quantification index of Korean version of mini mental state examination				
Category (point) Quantification index				
3-14	-1.260			
15-18	-1.198			
19-20	-1.013			
21-22	706			
23-24	320			
25	0.135			
26	0.656			
27	1.183			
28	1.508			
29-30	1.616			

Table 7 Quantification index of Hoehn and Yahr staging				
Category (point)	Quantification index			
1.0	-2.787			
1.5	-0.609			
2.0	-0.187			
2.5	-0.081			
3.0	0.151			
4.0	1.179			
5.0	2.167			

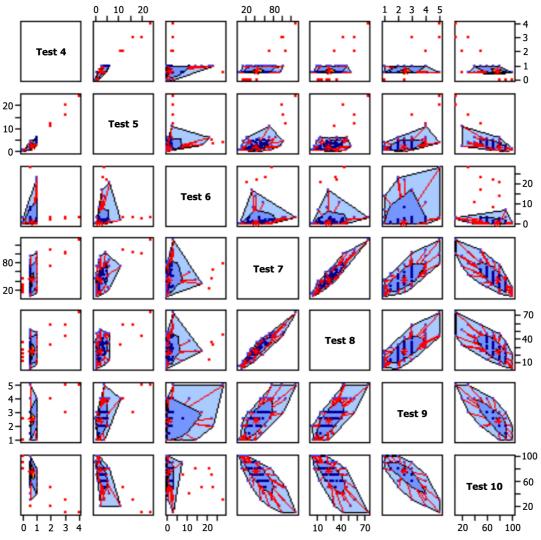
progressive cardinal symptoms, clinicians are more likely to rely on experience and the judgment of inexperienced clinicians may have low reliability.

It is believed that the analysis indices of this study can offer a range of information regarding the cognitive characteristics of the patient because they provide the optimal criteria for the screening test to distinguish PDD from PD-MCI. In particular, the optimal scale for early detection of PDD proposed in



Table 8 Optimal classification scores: odds ratios and 95% confidence interval								
Optimal classification scores	В	SE	Wald	P value	OR (95%CI)			
K-MMSE 25+ (Ref)			69.856	< 0.01				
23-24	1.499	0.473	10.035	0.002	4.478 (1.77-11.32)			
21-22	2.731	0.494	30.522	< 0.01	15.345 (5.82-40.43)			
19-20	2.587	0.549	22.195	< 0.01	13.294 (4.53-39.00)			
15-18	3.111	0.505	37.937	< 0.01	22.441 (8.33-60.39)			
3-14	4.008	0.799	25.185	< 0.01	55.020 (11.50-263.19)			
H&Y staging 1.0-2.5 (Ref)								
3.0-5.0	1.110	0.350	10.079	0.001	3.035 (1.52-6.02)			

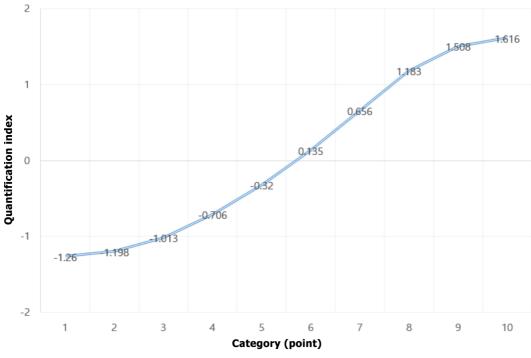
K-MMSE: Korean version of mini mental state examination; H&Y staging: Hoehn and Yahr staging.



DOI: 10.5498/wjp.v12.i8.1031 **Copyright** ©The Author(s) 2022.

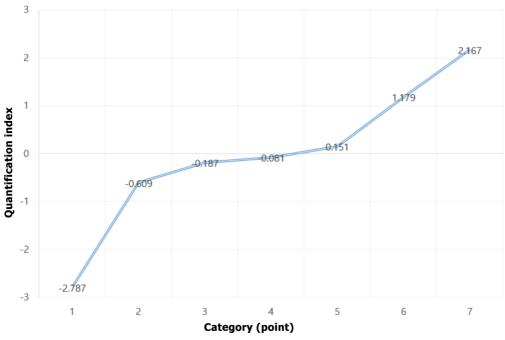
Figure 1 A bagplot that visualizes the location, spread, skewness, and outlier of the test results. Test 4 = Global Clinical Dementia Rating score; Test 5 = Sum of boxes in Clinical Dementia Rating; Test 6 = Korean Instrumental Activities of Daily Living; Test 7 = Unified PD Rating Scale (Total UPDRS score); Test 8 = Unified PD Rating Scale (Motor UPDRS score); Test 9 = Hoehn and Yahr staging; Test 10 = Schwab and England Activities of Daily Living.

1039



DOI: 10.5498/wjp.v12.i8.1031 **Copyright** ©The Author(s) 2022.

Figure 2 Quantification score graph among Korean Mini-Mental State Examination. Category 1 = 3-14 point; Category 2 = 15-18 point; Category 3 = 19-20 point; Category 4 = 21-22 point; Category 5 = 23-24 point; Category 6 = 25 point; Category 7 = 26 point; Category 8 = 27 point; Category 9 = 28 point; Category 10 = 29-30 point.



DOI: 10.5498/wjp.v12.i8.1031 **Copyright** ©The Author(s) 2022.

Figure 3 Quantification score graph among Hoehn and Yahr staging. Category 1 = 1.0 point; Category 2 = 1.5 point; Category 3 = 2.0 point; Category 4 = 2.5 point; Category 5 = 3.0 point; Category 6 = 4.0 point; Category 7=5.0 point.

this study is inexpensive, unlike dopamine transporter imaging and other methods, which have been proposed as efficient tests for early diagnosis of PDD but have limitations as screening tests due to space and cost. Moreover, the proposed scale can be utilized as a screening test simply in the primary medical setting without spatial restrictions. Consequently, it is believed that clinical application will be easy. Additional longitudinal studies are required to prove the effectiveness of the optimal scale for distinguishing PDD from PD-MCI proposed in this study.

This study had several limitations. First, although we used secondary data from a national survey, it is difficult to generalize the results of the study because the number of subjects was small due to the difficulties in diagnosing PD-MCI, which is not yet being actively screened for in PD patients. Second, we included patients taking medications such as dopaminergic drugs to treat PD, which can cause behavioral symptoms such as visual hallucinations that could influence the neuropsychological examination. Future studies are required to develop a model that can predict PDD from PD-MCI quickly while considering the administration of dopaminergic medication for PD. Third, the results of this study cannot be interpreted as a causal relationship because it was conducted using secondary data and the PD with Dementia Epidemiologic Data, the source data of this study, was designed as a crosssectional survey. Further longitudinal studies are needed to prove the causality of the results of this study. Fourth, the diagnosis of PSP-P was not distinguished in this study. Since the cognitive deficits in PD patients can be caused by PSP-P as well as PD-MIC, future studies are needed to exclude PSP-P in analysis.

CONCLUSION

We developed a predictive model to discriminate PDD from PD-MCI based on the results of nine neuropsychological tests and found that only KOREAN version of MMSE and Hoehn and Yahr Scale could be successfully employed to this end. For most efficiently discriminating PDD from PD-MCI, the optimal scaling factors for KOREAN version of MMSE and Hoehn and Yahr Scale were 10 and 7, respectively. We believe that our optimal scaling approach can be used to detect PDD in the early stages. Further longitudinal studies are required to confirm the performance of neuropsychological tests such as KOREAN version of MMSE and MoCA in predicting the progression of PD-MCI to PDD.

ARTICLE HIGHLIGHTS

Research background

It has been reported that Parkinson's disease (PD) with dementia (PDD) occurs frequently in people with PD.

Research motivation

The effective early detection of PD is an important topic in the field of geriatric medicine.

Research objectives

The aims of the present study were to develop a model for early detection of PDD based on neuropsychological testing.

Research methods

Data on 289 patients with PD [110 PDD and 179 Parkinson's Disease-Mild Cognitive Impairment (PD-MCI)] who were 60 years or older were used in the study. Regression with optimal scaling was used to identify independent relationships between the screening test results and PDD.

Research results

The Korean version of mini mental state examination (MMSE) (KOREAN version of MMSE) (b = -0.52, SE = 0.16) and Hoehn and Yahr scale (b = 0.44, SE = 0.19) were significantly effective models for distinguishing PDD from PD-MCI (P < 0.05), even after adjusting for all of the test results.

Research conclusions

The optimal number of categories (scaling factors) for KOREAN version of MMSE and Hoehn and Yahr Scale was 10 and 7, respectively.

Research perspectives

We believe that our optimal scaling approach can be used to detect PDD in the early stages.

ACKNOWLEDGEMENTS

The authors wish to thank the Korea CDC that provided the raw data for analysis.

1041

FOOTNOTES

Author contributions: Byeon H was designed the study, involved in data interpretation, preformed the statistical analysis, and assisted with writing the article.

Supported by Basic Science Research Program through the National Research Foundation of Korea (NRF) funded by the Ministry of Education, No. NRF-2018R1D1A1B07041091 and No. NRF-2021S1A5A8062526; and 2022 Development of Open-Lab based on 4P in the Southeast Zone.

Institutional review board statement: The study was conducted according to the guidelines of the Declaration of Helsinki, and approved by the Institutional Review Board of National Biobank of Korea under Korea Centers for Disease Control and Prevention (protocol code KBN-2019-1327).

Informed consent statement: All patients gave informed consent.

Conflict-of-interest statement: The author reports no relevant conflicts of interest for this article.

Data sharing statement: Technical appendix, statistical code from the corresponding author at bhwpuma@naver.com.

STROBE statement: The authors have read the STROBE Statement-checklist of items, and the manuscript was prepared and revised according to the STROBE Statement-checklist of items.

Open-Access: This article is an open-access article that was selected by an in-house editor and fully peer-reviewed by external reviewers. It is distributed in accordance with the Creative Commons Attribution NonCommercial (CC BY-NC 4.0) license, which permits others to distribute, remix, adapt, build upon this work non-commercially, and license their derivative works on different terms, provided the original work is properly cited and the use is noncommercial. See: https://creativecommons.org/Licenses/by-nc/4.0/

Country/Territory of origin: South Korea

ORCID number: Haewon Byeon 0000-0002-3363-390X.

S-Editor: Gao CC L-Editor: A P-Editor: Chen YX

REFERENCES

- Jay R, Jung SB, Park BH, Jeong BC, Seo SI, Jeon SS, Lee HM, Choi HY, Jeon HG. Compensatory structural and functional adaptation after radical nephrectomy for renal cell carcinoma according to preoperative stage of chronic kidney disease. Choi DK, Jung SB, Park BH, Jeong BC, Seo SI, Jeon SS, Lee HM, Choi HY, Jeon HG.J Urol. 2015 Oct;194(4):910-5. [Epub 2015 Apr 28]. doi: 10.1016/j.juro.2015.04.093. Urol Oncol 2017; 35: 118-119 [PMID: 28159494 DOI: 10.1016/j.urolonc.2016.12.024]
- 2 Health Insurance Review and Assessment Service. Regarding the role of the Health Insurance Review and Assessment Service that needs to be strengthened as the health care environment changes rapidly. HIRA Research 2021; 1: 98-102 [DOI: 10.52937/hira.21.1.1.98]
- Statistics Korea. Population projections for Korea (2017~2067), Statistics Korea: Daejeon, 2019
- 4 Gonzalez-Latapi P, Bayram E, Litvan I, Marras C. Cognitive Impairment in Parkinson's Disease: Epidemiology, Clinical Profile, Protective and Risk Factors. Behav Sci (Basel) 2021; 11 [PMID: 34068064 DOI: 10.3390/bs11050074]
- Rosca EC, Simu M. Parkinson's Disease-Cognitive Rating Scale for Evaluating Cognitive Impairment in Parkinson's Disease: A Systematic Review. Brain Sci 2020; 10 [PMID: 32854426 DOI: 10.3390/brainsci10090588]
- Vasconcellos LF, Pereira JS. Parkinson's disease dementia: Diagnostic criteria and risk factor review. J Clin Exp Neuropsychol 2015; 37: 988-993 [PMID: 26332178 DOI: 10.1080/13803395.2015.1073227]
- Aarsland D, Zaccai J, Brayne C. A systematic review of prevalence studies of dementia in Parkinson's disease. Mov Disord 2005; **20**: 1255-1263 [PMID: 16041803 DOI: 10.1002/mds.20527]
- Williams-Gray CH, Foltynie T, Brayne CE, Robbins TW, Barker RA. Evolution of cognitive dysfunction in an incident Parkinson's disease cohort. Brain 2007; 130: 1787-1798 [PMID: 17535834 DOI: 10.1093/brain/awm111]
- Janvin CC, Larsen JP, Aarsland D, Hugdahl K. Subtypes of mild cognitive impairment in Parkinson's disease: progression to dementia. Mov Disord 2006; 21: 1343-1349 [PMID: 16721732 DOI: 10.1002/mds.20974]
- Caviness JN, Driver-Dunckley E, Connor DJ, Sabbagh MN, Hentz JG, Noble B, Evidente VG, Shill HA, Adler CH. Defining mild cognitive impairment in Parkinson's disease. Mov Disord 2007; 22: 1272-1277 [PMID: 17415797 DOI: 10.1002/mds.21453]
- Byeon H. Application of Machine Learning Technique to Distinguish Parkinson's Disease Dementia and Alzheimer's Dementia: Predictive Power of Parkinson's Disease-Related Non-Motor Symptoms and Neuropsychological Profile. J Pers *Med* 2020; **10** [PMID: 32354187 DOI: 10.3390/jpm10020031]
- Necpál J, Borsek M, Jeleňová B. "Parkinson's disease" on the way to progressive supranuclear palsy: a review on PSP-



- parkinsonism. Neurol Sci 2021; 42: 4927-4936 [PMID: 34532773 DOI: 10.1007/s10072-021-05601-8]
- Alster P, Madetko N, Koziorowski D, Friedman A. Progressive Supranuclear Palsy-Parkinsonism Predominant (PSP-P)-A Clinical Challenge at the Boundaries of PSP and Parkinson's Disease (PD). Front Neurol 2020; 11: 180 [PMID: 32218768 DOI: 10.3389/fneur.2020.00180]
- 14 Pigott K, Rick J, Xie SX, Hurtig H, Chen-Plotkin A, Duda JE, Morley JF, Chahine LM, Dahodwala N, Akhtar RS, Siderowf A, Trojanowski JQ, Weintraub D. Longitudinal study of normal cognition in Parkinson disease. Neurology 2015; 85: 1276-1282 [PMID: 26362285 DOI: 10.1212/WNL.00000000000002001]
- Martinez-Martin P, Rodriguez-Blazquez C, Forjaz MJ, Frades-Payo B, Agüera-Ortiz L, Weintraub D, Riesco A, Kurtis MM, Chaudhuri KR. Neuropsychiatric symptoms and caregiver's burden in Parkinson's disease. Parkinsonism Relat Disord 2015; 21: 629-634 [PMID: 25892660 DOI: 10.1016/j.parkreldis.2015.03.024]
- Sun C, Armstrong MJ. Treatment of Parkinson's Disease with Cognitive Impairment: Current Approaches and Future Directions. Behav Sci (Basel) 2021; 11 [PMID: 33920698 DOI: 10.3390/bs11040054]
- Byeon H. Predicting the Severity of Parkinson's Disease Dementia by Assessing the Neuropsychiatric Symptoms with an SVM Regression Model. Int J Environ Res Public Health 2021; 18 [PMID: 33806474 DOI: 10.3390/ijerph18052551]
- Dusseldorp E, Meulman JJ. Prediction in medicine by integrating regression trees into regression analysis with optimal scaling. Methods Inf Med 2001; 40: 403-409 [PMID: 11776739 DOI: 10.1055/s-0038-1634200]
- Cleophas TJ, Zwinderman AH. Optimal scaling and automatic linear regression. Regression Analysis in Medical Research. Springer, Cham, 2018: 255-266
- Zhang D, Zhang C, Jiang P. Optimal scaling regression analysis for impact factors of SCI papers output. Zhonghua Yixue Keyan Guanli Zazhi 2018; 465-469
- Hughes AJ, Daniel SE, Kilford L, Lees AJ. Accuracy of clinical diagnosis of idiopathic Parkinson's disease: a clinicopathological study of 100 cases. J Neurol Neurosurg Psychiatry 1992; 55: 181-184 [PMID: 1564476 DOI: 10.1136/innp.55.3.1811
- Movement Disorder Society Task Force on Rating Scales for Parkinson's Disease. The Unified Parkinson's Disease Rating Scale (UPDRS): status and recommendations. Mov Disord 2003; 18: 738-750 [PMID: 12815652 DOI: 10.1002/mds.10473]
- Byeon H. Exploring the Predictors of Rapid Eye Movement Sleep Behavior Disorder for Parkinson's Disease Patients Using Classifier Ensemble. Healthcare (Basel) 2020; 8 [PMID: 32369941 DOI: 10.3390/healthcare8020121]
- Winblad B, Palmer K, Kivipelto M, Jelic V, Fratiglioni L, Wahlund LO, Nordberg A, Bäckman L, Albert M, Almkvist O, Arai H, Basun H, Blennow K, de Leon M, DeCarli C, Erkinjuntti T, Giacobini E, Graff C, Hardy J, Jack C, Jorm A, Ritchie K, van Duijn C, Visser P, Petersen RC. Mild cognitive impairment--beyond controversies, towards a consensus: report of the International Working Group on Mild Cognitive Impairment. J Intern Med 2004; 256: 240-246 [PMID: 15324367 DOI: 10.1111/j.1365-2796.2004.01380.xl
- Hoehn MM, Yahr MD. Parkinsonism: onset, progression and mortality. Neurology 1967; 17: 427-442 [PMID: 6067254 DOI: 10.1212/wnl.17.5.427]
- Choi SH, Na DL, Lee BH, Hahm DS, Jeong JH, Yoon SJ, Yoo KH, Ha CK, Han IW. Estimating the validity of the Korean version of expanded Clinical Dementia Rating (CDR) scale. J Korean Neurol Assoc 2001; 19: 585-591 [DOI: 10.1037/t63009-000]
- Gillingham FJ, Donaldson MC. Schwab and England activities of daily living. Third symposium of Parkinson's disease. E&S Livingstone: Edinburgh, Scotland, 1969
- Kang SJ, Choi SH, Lee BH, Kwon JC, Na DL, Han SH. The reliability and validity of the Korean Instrumental Activities of Daily Living (K-IADL). J Korean Neurol Assoc 2002; 20: 8-14 [DOI: 10.13029/jkaps.2013.19.1.1]
- Kang Y, Na DL, Hahn S. A validity study on the Korean Mini-Mental State Examination (K-MMSE) in dementia patients. J Korean Neurol Assoc 1997; 15: 300-308
- Kang Y, Park J, Yu KH, Lee BC. The validity of the Korean-Montreal Cognitive Assessment (K-MoCA) as a screening test for both MCI and VCI. Conference Abstract: The 20th Annual Rotman Research Institute Conference. The frontal lobes, 2010: 10
- 31 Wang DZ, Wang C, Shen CF, Zhang Y, Zhang H, Song GD, Xue XD, Xu ZL, Zhang S, Jiang GH. [Comparison of application of Cochran-Armitage trend test and linear regression analysis for rate trend analysis in epidemiology study]. Zhonghua Liu Xing Bing Xue Za Zhi 2017; **38**: 684-687 [PMID: 28651412 DOI: 10.3760/cma.j.issn.0254-6450.2017.05.026]
- Chin J, Park J, Yang SJ, Yeom J, Ahn Y, Baek MJ, Ryu HJ, Lee BH, Han NE, Ryu KH, Kang Y. Re-standardization of the Korean-Instrumental Activities of Daily Living (K-IADL): Clinical Usefulness for Various Neurodegenerative Diseases. Dement Neurocogn Disord 2018; 17: 11-22 [PMID: 30906387 DOI: 10.12779/dnd.2018.17.1.11]
- Van der Kooij AJ. Prediction accuracy and stability of regression with optimal scaling transformations. M.D. Thesis, Leiden University, Leiden. 2007. Available from: https://scholarly publications. universite it leiden. nl/access/item% 3A 2889259/view
- Dubois B, Burn D, Goetz C, Aarsland D, Brown RG, Broe GA, Dickson D, Duyckaerts C, Cummings J, Gauthier S, Korczyn A, Lees A, Levy R, Litvan I, Mizuno Y, McKeith IG, Olanow CW, Poewe W, Sampaio C, Tolosa E, Emre M. Diagnostic procedures for Parkinson's disease dementia: recommendations from the movement disorder society task force. Mov Disord 2007; 22: 2314-2324 [PMID: 18098298 DOI: 10.1002/mds.21844]
- Marshall V, Grosset D. Role of dopamine transporter imaging in routine clinical practice. Mov Disord 2003; 18: 1415-1423 [PMID: 14673877 DOI: 10.1002/mds.10592]

1043

Rizzo G, Copetti M, Arcuti S, Martino D, Fontana A, Logroscino G. Accuracy of clinical diagnosis of Parkinson disease: A systematic review and meta-analysis. Neurology 2016; 86: 566-576 [PMID: 26764028 DOI: 10.1212/WNL.0000000000002350]





Published by Baishideng Publishing Group Inc

7041 Koll Center Parkway, Suite 160, Pleasanton, CA 94566, USA

Telephone: +1-925-3991568

E-mail: bpgoffice@wjgnet.com

Help Desk: https://www.f6publishing.com/helpdesk

https://www.wjgnet.com

