Influence of Lifestyles on Mild Cognitive Impairment: A Decision Tree Model Study

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Zongqiu Wang^{1,*}

Jiwen Hou^{2,*}

Yu Shi³

Qiaowen Tan¹

Lin Peng¹

Zhiying Deng¹

Zhihong Wang¹

Zongjun Guo¹

¹Department of Geriatrics, The Affiliated Hospital of Qingdao University, Qingdao, People's Republic of China; ²Department of Geriatrics, The Affiliated Hospital of Chengdu University, Chengdu, People's Republic of China; ³Department of Critical Medicine, Weihai Central Hospital, Weihai, People's Republic of China

*These authors contributed equally to this work

Objective: To explore the effects of different lifestyle choices on mild cognitive impairment (MCI) and to establish a decision tree model to analyse their predictive significance on the incidence of MCI.

Methods: Study participants were recruited from geriatric and physical examination centres from October 2015 to October 2019: 330 MCI patients and 295 normal cognitive (NC) patients. Cognitive function was evaluated by the Mini-Mental State Examination Scale (MMSE) and Clinical Dementia Scale (CDR), while the Barthel Index (BI) was used to evaluate life ability. Statistical analysis included the χ^2 test, logistic regression, and decision tree. The ROC curve was drawn to evaluate the predictive ability of the decision tree model.

Results: Logistic regression analysis showed that low education, living alone, smoking, and a high-fat diet were risk factors for MCI, while young age, tea drinking, afternoon naps, social engagement, and hobbies were protective factors for MCI. Social engagement, a high-fat diet, hobbies, living condition, tea drinking, and smoking entered all nodes of the decision tree model, with social engagement as the root node variable. The importance of predictive variables in the decision tree model showed social engagement, a high-fat diet, tea drinking, hobbies, living condition, and smoking as 33.57%, 27.74%, 22.14%, 11.94%, 4.61%, and 0%, respectively. The area under the ROC curve predicted by the decision tree model was 0.827 (95% CI: 0.795~0.856).

Conclusion: The decision tree model has good predictive ability. MCI was closely related to lifestyle; social engagement was the most important factor in predicting the occurrence of MCI.

Keywords: lifestyle, behaviours habit, mild cognitive impairment, influencing factors, decision tree model

Introduction

Mild cognitive impairment (MCI) is a common age-related cognitive disorder that is a transitional stage between normal aging and dementia.¹ According to Petersen's survey in 2018,² the prevalence rate of MCI in people aged 60 to 64 years in the United States was 6.7%. For those over 80 years, the prevalence rate of MCI reached 25.2%, and the cumulative incidence of dementia in those with MCI older than 65 years after a follow-up for two years was 14.9%. Therefore, MCI is considered to be the precursory stage of dementia; early identification of and intervention with the factors affecting the occurrence and development of MCI may play an important role in its prevention.^{3,4}

A decision tree is a simple and powerful machine learning model that utilises any information obtained to find the best classification index of data samples. These

Correspondence: Zongjun Guo
Department of Geriatrics, The Affiliated Hospital of Qingdao University, No. 16 Jiangsu Road, Qingdao, Shandong 266003, People's Republic of China
Email guozj@163.com

classification indexes are the nodes of the decision tree, which then grow to form the tree structure.⁵ At present, the decision tree model has been successfully applied to research on public health and health behaviour.⁶ Not only can it determine the influencing factors of a disease and reflect the interaction among them,^{7,8} but it can also be used to predict the disease.^{9–12} Therefore, a variety of factors affecting MCI can be comprehensively analysed. Previous studies have found that lifestyle (such as drinking habits, smoking, and exercise), age, and educational attainment are related to the occurrence of MCI.^{13–15} However, there are relatively few studies on the interaction of different influencing factors on MCI. Therefore, this study aimed, through the decision tree method, to produce a model that predicts the interactions of the influencing factors and MCI in order to intervene with or delay its occurrence and development.

Methods

Research Object

The inclusion criteria of the MCI group were as follows: age ≥ 60 years old and meets the core clinical criteria of MCI (i.e. cognitive problems, at least one cognitive impairment, independence of daily function, and absence of dementia).¹⁶ The inclusion criteria of the normal cognitive (NC) group were as follows: age ≥ 60 years old, no symptoms or signs of cognitive decline or impairment were found during physical examination, and the two evaluation tools for cognitive function were normal. The exclusion criteria included presence of other important neurological diseases other than MCI (such as Parkinson's disease, epilepsy, and cerebral infarction);¹⁷ presence of clinically advanced or unstable diseases (such as advanced tumours and severe infections); a history of schizophrenia, mania, or depression;¹⁸ and the inability to cooperate with the completion of the questionnaire. From October 2015 to October 2019, 625 cases in geriatric and physical examination centres met the above criteria and were collected: 330 cases in the MCI group and 295 cases in the NC group. All patients were informed of the purpose of this study and provided consent. For patients with obvious cognitive impairment, we have obtained the consent of their families. The informed verbal consent process was approved by the Medical Ethics Committee of the Affiliated Hospital of Qingdao University (QYFYWZLL25697) and complied with the Declaration of Helsinki.

Introduction of Evaluation Scale

Mini-Mental State Examination (MMSE)

The MMSE scale consists of 30 questions relating to orientation, memory, attention, reading construction, and comprehension with a total possible score of 30 points. According to the level of education, the criteria for judging cognitive impairment are as follows: illiteracy group ≤ 19 points, primary school group ≤ 22 points, and junior high school and above group ≤ 26 points. MMSE is one of the methods to judge whether the cognitive function is normal or not and can not replace the clinical diagnosis. In addition, it contributes to the early detection of dementia patients and reduces missed diagnosis.¹⁹

Clinical Dementia Rating Scale (CDR)

CDR scores were obtained by questioning patients and people with close relationships with them (such as family members and caregivers) and assessing each of the six cognitive categories (memory, orientation, judgment and problem solving, community affairs, family and hobbies, and personal care). The scoring standard was five points, no dementia = 0, suspected dementia = 0.5, mild dementia = 1, moderate dementia = 2, and severe dementia = 3.²⁰ In this study, this scale was used to distinguish mild cognitive impairment from dementia.

Ability of Daily Living Assessment

In this study, the widely used Barthel Index (BI) scale was adopted to evaluate the activities of daily living (ADL), which mainly include eating, bathing, grooming, dressing, controlling defecation, controlling urination, going to the toilet, bed and chair transfer, walking on the ground for 45 minutes, and walking up and down stairs, with a total possible score of 100 points. The higher the score, the better the independence. Based on the total possible score, ADL was divided into the following ranges: 100 points for normal activities of daily living, 60 points for good and basic self-care, 59–40 points for moderate dysfunction and need of help, 39–20 points for severe dysfunction and life dependence, and below 20 points for complete disability and complete life dependence.²¹ This study followed the diagnostic criteria of MCI, with BI index ≥ 60 points, indicating independence of daily life.

Research methods

The researchers used standardised language and unified scale score standards to obtain medical history and

Table 1 MCI Studies Variable Assignment

Research Variables	Assignment Situation
Cognitive state	MCI=1, NC=2
Gender	Male=1, female=2
Age group (years)	60–64=1, 65–69=2, 70–74=3, 75–79=4, ≥80=5
Educational level	Illiteracy =1, primary school =2, middle school =3, high school =4, university =5
Occupation	Mental labour =1, physical labour =2
Living condition	Live alone =1, live with family =2
Drinking	Yes = 1, no = 2
Smoking	Yes = 1, no = 2
Drink tea	Yes = 1, no = 2
Sleep time (hours)	<7=1, 7–9=2, >9=3
Nap	Yes = 1, no = 2
Physical exercise	Yes = 1, no = 2
Social engagement	Yes = 1, no = 2
Hobbies	Yes = 1, no = 2
High-fat diet	Yes = 1, no = 2

perform a physical examination, using MMSE, CDR, and ADL evaluation and collecting demographic, lifestyle and behaviour habits, and other data. The data were then double-checked to ensure accuracy and assign values to the research variables (Table 1).

Statistical Analysis

The data were analysed using IBM SPSS 25.0 software. χ^2 test was used for single-factor analysis, binary logistic regression analysis was used to analyse the influencing factors, and $P < 0.05$ was considered statistically significant. The decision tree model was established by using IBM SPSS Modeler 14.1 software decision tree C5.0 algorithm; MedCalc 15.2.2 software was used to draw the ROC curve.

Results

Characteristics of General Demographic Data

This study enrolled 625 elderly people classified either in the MCI group (330 cases) or the NC group (295 cases). There were 301 males and 324 females. The constituent ratio of those in MCI group relative to their counterparts in the NC group was as follows: according to sex, 55% male and 51% female; according to age, 41% 60–64 years old, 52% 65–69 years, 57% 70–74 years, 68% 75–79 years, and 62% over 80 years; according to educational attainment, 61% illiterate, 59% primary school,

50% junior high school, 51% senior high school, and 12% above college; according to occupation, 51% mental work and 54% manual work; and according to living situation, 84% living alone and 48% living with family (Table 2).

Single Factor Analysis on the Influence of Lifestyle on MCI

There were significant differences ($P < 0.005$) between the MCI and NC groups in all of the following aspects: age ($P < 0.0001$), educational attainment ($P < 0.0001$), living condition ($P < 0.0001$), smoking habits ($P = 0.016$), tea drinkers ($P < 0.0001$), afternoon naps ($P < 0.0001$), physical exercise ($P < 0.0001$), social engagement ($P < 0.0001$), hobbies ($P < 0.0001$), and a high-fat diet ($P < 0.0001$). Among them, the proportion of participants with MCI was lower with regard to young age, tea drinking, afternoon naps, physical exercise, social engagement, and hobbies. The elderly with a low level of education, who lived alone, who smoked, or who had a high-fat diet had a higher proportion of MCI (Table 2).

Binary Logistic Regression Analysis on the Influence of Lifestyle on MCI

Taking MCI as the dependent variable and the statistically significant factors in the single factor analysis as independent variables, binary logistic regression analysis showed that age, education, living condition, smoking habits, tea drinking, afternoon naps, social engagement, hobbies, and a high-fat diet were the influencing factors of MCI. Among them, illiteracy (OR=12.248, 95% CI: 3.088–48.571), an educational attainment of primary school (OR=8.975, 95% CI: 2.482–32.452), junior high school (OR=6.445, 95% CI: 1.749–23.743), or senior high school (OR=6.386, 95% CI: 1.640–24.873), living alone (OR=5.026, 95% CI: 2.204–11.463), smoking (OR=2.702, 95% CI: 1.674–4.362), and a high-fat diet (OR=2.955, 95% CI: 1.892–4.615) were risk factors for the occurrence of MCI. However, being in the age range of 60–64 years (OR=0.23, 95% CI: 0.096–0.572) or 65–69 years (OR=0.309, 95% CI: 0.127–0.754), drinking tea (OR=0.285, 95% CI: 0.182–0.447), taking afternoon naps (OR=0.566, 95% CI: 0.350–0.916), taking part in social engagement (OR=0.145, 95% CI: 0.091–0.233), and having hobbies (OR=0.421, 95% CI: 0.264–0.671) were the protective factors for the occurrence of MCI (Table 3).

Table 2 Single Factor Analysis of the Influence of Lifestyle on MCI

Research Variables	Category	NC	MCI	Total	MCI Constituent Ratio (%)	χ^2	P
Gender	Male	137	164	301	55	0.662	0.416
	Female	158	166	324	51		
Age group (years)	60–64	112	79	191	41	21.083	<0.001
	65–69	78	83	161	52		
	70–74	57	76	133	57		
	75–79	28	60	88	68		
	≥80	20	32	52	62		
Educational level	Illiteracy	27	43	70	61	33.547	<0.001
	Primary school	115	163	278	59		
	Middle school	69	70	139	50		
	High school	48	49	97	51		
	University	36	5	41	12		
Occupation	Mental labour	121	127	248	51	0.417	0.518
	Physical labour	174	203	377	54		
Living condition	Live alone	13	68	81	84	36.235	<0.001
	Live with family	282	262	544	48		
Drinking	Yes	94	97	191	51	0.443	0.508
	No	201	233	434	54		
Smoking	Yes	79	118	197	60	5.816	0.016
	No	216	212	428	50		
Drink tea	Yes	213	117	330	35	84.405	<0.001
	No	82	213	295	72		
Sleep time (hours)	<7h	154	175	329	53	0.272	0.873
	7–9h	123	138	261	53		
	>9h	18	17	35	49		
Nap	Yes	215	194	409	47	13.680	<0.001
	No	80	136	216	63		
Physical exercise	Yes	213	170	383	44	28.096	<0.001
	No	82	160	242	66		
Social engagement	Yes	178	74	252	29	93.051	<0.001
	No	117	256	373	69		
Hobbies	Yes	240	164	404	41	68.302	<0.001
	No	55	166	221	75		
High-fat diet	Yes	101	236	337	70	87.116	<0.001
	No	194	94	288	33		

Decision Tree Model of the Influence of Lifestyle on MCI

The statistically significant factors in the univariate analysis were selected for the decision tree analysis. Selecting social engagement, a high-fat diet, hobbies, living condition, tea drinking, and smoking habits as each node of the decision tree model and social

engagement as the root node variable, the results showed that the probability of MCI in people who had social engagement was 29.365% (74 cases), lower than those without social engagement (68.633%; 256 cases). Among the participants who had social engagement, the probability of MCI among people with a high-fat diet was 53.782% (64 cases), higher than that of those with a non-

Table 3 Binary Logistic Regression Analysis of the Influence of Lifestyle on MCI

Research Variables	β	Walds	P	OR (95% CI)
60–64 years	−1.450	10.184	0.001	0.235 (0.096–0.572)
65–69 years	−1.174	6.653	0.010	0.309 (0.127–0.754)
70–74 years	−0.751	2.647	0.104	0.472 (0.191–1.166)
75–79 years	−0.092	0.034	0.854	0.912 (0.343–2.428)
Illiteracy	2.505	12.704	0.000	12.248 (3.088–48.571)
Primary school	2.194	11.199	0.001	8.975 (2.482–32.452)
Middle school	1.863	7.843	0.005	6.445 (1.749–23.743)
High school	1.854	7.144	0.008	6.386 (1.640–24.873)
Live alone	1.615	14.732	0.000	5.026 (2.204–11.463)
Smoking	0.994	16.544	0.000	2.702 (1.674–4.362)
Drink tea	−1.254	29.943	0.000	0.285 (0.182–0.447)
Nap	−0.569	5.379	0.020	0.566 (0.350–0.916)
Physical exercise	−0.091	0.155	0.694	0.913 (0.581–1.435)
Social engagement	−1.928	64.496	0.000	0.145 (0.091–0.233)
Hobbies	−0.865	13.227	0.000	0.421 (0.264–0.671)
High-fat diet	1.084	22.694	0.000	2.955 (1.892–4.615)

high-fat diet (7.519%; 10 cases). Among the participants who had social engagement and a high-fat diet, the probability of MCI of those with hobbies was 35.211% (25 cases), lower than that of those without hobbies (81.250%; 39 cases). Among participants who had social engagement, a high-fat diet, and hobbies, the probability of MCI among people living alone was 77.778% (7 cases), higher than those living with their families (29.032%; 18 cases). In addition, among the people who did not have social engagement, the probability of MCI among tea drinkers was 49.153% (87 cases), lower than that of those who did not drink tea (86.224%; 169 cases). Among those who had no social engagement and drank tea, the probability of MCI among those living with their families was 45.679% (74 cases), lower than that of participants living alone (86.667%; 13 cases). Among the people who had no social engagement, drank tea, and lived with their families, the probability of MCI among smokers was 60% (36 cases), lower than that of non-smokers (37.225%; 38 cases) (Figure 1).

The Importance of Prediction Variables in the Decision Tree Model

The importance of each node prediction variable to the construction of the model was different in the decision tree model, while the importance of social engagement as the root node variable was 33.57%. In the leaf nodes, the importance of a high-fat diet, tea drinking, hobbies, living conditions, and smoking were 27.74%, 22.14%, 11.94%, 4.61%, and 0%, respectively (Figure 2).

Prediction of ROC Curve of MCI Occurrence by Decision Tree Model

Taking the prediction of the decision tree model as the test variable and the real MCI data as the state variable to draw the ROC curve, it was concluded that the AUC of the decision tree model for predicting the occurrence of MCI was 0.827 (95% CI: 0.795–0.856), $P < 0.001$. This shows that the prediction effect of the decision tree model was good (Figure 3).

Discussion

In this study, logistic regression analysis showed that low education, living alone, smoking, and a high-fat diet were risk factors for MCI, while a young age, tea drinking, afternoon naps, social engagement, and hobbies were protective factors for MCI. These results were basically consistent with previous studies.^{22–24} With social engagement, a high-fat diet, hobbies, living conditions, tea drinking, and smoking entering each node of the decision tree model, social engagement was the root node variable and the most important predictor of MCI. The probability of MCI in the elderly with social engagement was significantly lower than that in those without social engagement. The results of a large cross-border study involving 19,832 people over a period of six years showed that social engagement plays an important role in cognitive function and successful aging, but the relationship between social engagement and cognitive function is not yet clear.²⁵ The possible mechanism is that

Figure I Decision tree model of the influence of lifestyles on MCI. This tree is generated by the software through 10 cross-validations. Each box is numbered sequentially. The blue icon represents MCI, and the red icon represents NC. Each box has its own percentage and specific number of examples. Below each box is the influence factor of the next classification; at the top of each box is the corresponding category of influence factor.

Notes: MCI is mild cognitive impairment, NC is normal cognitive function.

Figure 2 Importance of predictive variables in decision tree model. This figure shows the importance of each influence factor to the result prediction in the decision tree model. As a root node variable, social engagement is the most important predictive variable.

Figure 3 ROC curve of MCI occurrence predicted by decision tree model. The broken line in the figure represents the ability of the decision tree model to predict the occurrence of MCI. The AUC was 0.827, $P < 0.001$.

environmental complexity is an important factor in cognitive reserve; social engagement enables MCI patients to better compensate for brain changes, such as aging, cognitive impairment, and AD by activating and strengthening neurobiological pathways.²⁶

In people who have social engagement, those who eat a high-fat diet have a high probability of developing MCI; this may be due to a high-fat diet leading to hypercholesterolemia, increased oxidative stress, and an increased concentration of the amyloid protein in the hippocampus, resulting in a decline in cognitive function.²⁷ For example,

a 30-year retrospective cohort study showed that middle-aged hypercholesterolemia increases the risk of dementia in the elderly by 27%.²⁸ In people with social engagement and a high-fat diet, those with hobbies have a low probability of developing MCI, and studies have shown that hobbies (such as reading, music, and chess) can make related neural networks more effective or more malleable, thus delaying the onset of dementia and cognitive deterioration.²⁹ A study of 6586 subjects followed up with over five years found that the risk of MCI in the elderly who read regularly decreased by 41%, and the protective effect on the cognitive function of the well-educated elderly was more obvious.³⁰ Among the participants with social engagement, a high-fat diet, and hobbies, those living alone had a high probability of MCI. Chinese scholars have shown that the low psychological elasticity of people living alone is an important factor leading to the occurrence and development of MCI.³¹

In the subjects without social engagement, tea drinkers have a low probability of developing MCI. Studies have shown that tea polyphenols can resist oxidative stress and protect nerve cells.³² Among participants who had no social engagement, drank tea, and lived with their families, smokers had a higher probability of developing MCI. A study in Rotterdam showed that smoking increased the risk of cognitive decline (OR=1.47, 95% CI: 1.18–1.86) after an average of 7.1 years of follow-up in elderly smokers.³³ A large number of toxic substances in cigarettes (such as carbon monoxide and nitrosamines) accumulate free radicals and damage nerve cells, resulting in cognitive decline.³⁴

In the results of the decision tree model of this study, we also found that people with hobbies and non-high-fat diets have the lowest probability of MCI (7.519%), while those who have no hobbies, drink tea, and live alone have the highest probability of MCI (86.667%), which is similar to the probability of those with MCI who do not have hobbies and do not drink tea. It can be seen that the interaction of protective factors can minimise the incidence of MCI, while the interaction of risk factors can increase the probability of MCI.

Therefore, the decision tree model can intuitively and clearly show the influence and probability of each factor entering the model on MCI and can be read down along each main line, with a high degree of readability. The AUC of the ROC curve of MCI predicted by the decision tree model is 0.827 (95% CI: 0.795–0.856). It is a good method to analyse the influencing factors of clinical

disease and predict the probability of disease occurrence. The limitation of this study is that the lifestyle should be further refined, such as the time, intensity and type of physical exercise, the type of tea being consumed, and the frequency of alcohol consumption. In addition, more samples should be included to increase the reliability of the model.

Conclusion

This study shows that MCI is closely related to a variety of lifestyles, among which social communication is the most important factor to predict the occurrence of MCI. Other influencing factors such as high-fat diet, tea drinking, hobbies, and living condition also have a certain impact on the occurrence of MCI. The decision tree model method has a good ability to predict MCI, and the interaction between factors can have a further impact on the occurrence of MCI. Therefore, individual lifestyle intervention can be carried out for the elderly in order to reduce the incidence of MCI.

Highlights

We found that low education, living alone, smoking, and high-fat diet were risk factors for MCI, while low age, tea drinking, afternoon naps, social engagement, and hobbies were protective factors for MCI.

The decision tree model has good predictive ability, with social engagement as the most important factor in predicting the occurrence of MCI. Other factors such as a high-fat diet, tea drinking, hobbies, and living condition can also predict the occurrence of MCI.

The combination of medical research and data mining technology provides new ideas for future research.

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Disclosure

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References

- Petersen RC, Smith GE, Waring SC, et al. Mild cognitive impairment: clinical characterization and outcome. *Arch Neurol*. 1999;56(3):303–308. doi:10.1001/archneur.56.3.303
- Peterson RC, Lepez O, Armstrong MJ, et al. Practice guideline update summary: mild cognitive impairment: report of the guideline development, dissemination, and implementation subcommittee of the American Academy of Neurology. *Neurol*. 2018;90(3):126–135. doi:10.1212/WNL.0000000000004826
- Langa KM, Levine DA. The diagnosis and management of mild cognitive impairment: a clinical review. *JAMA*. 2014;312(23):2551–2561. doi:10.1001/jama.2014.13806
- Gonzalez HM, Tarraf W, Schneiderman N, et al. Prevalence and correlates of mild cognitive impairment among diverse Hispanics/Latinos: study of Latinos-investigation of neurocognitive aging results. *Alzheimers Dement*. 2019;15(12):1507–1515. doi:10.1016/j.jalz.2019.08.202
- Kuhn M, Johnson K. *Applied Predictive Modeling*. Vol. 13. Nonlinear classification models; 2013:329–367.
- Lemen SC, Roy J, Clark MA, Friedmann PD, Rakowski W. Classification and regression tree analysis in public health: methodological review and comparison with logistic regression. *Ann Behav Med*. 2003;26(3):172–181. doi:10.1207/S15324796ABM2603_02
- Tufféry S. *Data Mining and Statistics for Decision Making*. 1st ed. John Wiley & Sons, Ltd; 2011.
- Yong HH, Karmakar C, Borland R, Kusmaker S, Fuller-Tyszkiewicz M, Yearwood J. Identifying smoker subgroups with high versus low smoking cessation attempt probability: a decision tree analysis approach. *Addict Behav*. 2020;103:106258. doi:10.1016/j.addbeh.2019.106258
- Li H, Wu TT, Yang DL, et al. Decision tree model for predicting in-hospital cardiac arrest among patients admitted with acute coronary syndrome. *Clin Cardiol*. 2019;42(11):1087–1093. doi:10.1002/clc.23255
- Pei D, Zhang C, Quan Y, Guo Q. Identification of potential type II diabetes in a Chinese population with a sensitive decision tree approach. *J Diabetes Res*. 2019;2019:1–7. doi:10.1155/2019/4248218
- Brims FJ, Meniawy TM, Duffus I, et al. A novel clinical prediction model for prognosis in malignant pleural mesothelioma using decision tree analysis. *J Thorac Oncol*. 2016;11(4):573–582. doi:10.1016/j.jtho.2015.12.108
- Wang X, Guo ZJ, Zhang WQ, et al. Study on the decision tree model for risk factors of vascular cognitive impairment. *Chin J Behav Med Brain Sci*. 2017;26(6):534–538.
- Solfrizzi V, D'Introno A, Colacicco AM, et al. Lifestyle-related factors, alcohol consumption, and mild cognitive impairment. *J Am Geriatr Soc*. 2007;55(10):1679–1681. doi:10.1111/j.1532-5415.2007.01313.x
- Wingbermuehle R, Wen KX, Wolters FJ, Ikram MA, Bos D. Smoking, APOE genotype, and cognitive decline: the Rotterdam study. *J Alzheimers D*. 2017;57(4):1191–1195. doi:10.3233/JAD-170063
- Falck RS, Davis JC, Best JR, Crockett RA, Liu-Ambrose T. Impact of exercise training on physical and cognitive function among older adults: a systematic review and meta-analysis. *Neurobiol Aging*. 2019;79:119–130. doi:10.1016/j.neurobiolaging.2019.03.007
- Albert M, DeKosky S, Dickson D, et al. The diagnosis of mild cognitive impairment due to Alzheimer's disease: recommendations from the National Institute on Aging-Alzheimer's association workgroups on diagnostic guidelines for Alzheimer's Dement. 2011;7(3):270–279. doi:10.1016/j.jalz.2011.03.008
- Hunter SW, Divine A, Frengopoulos C, Montero Odasso M. A framework for secondary cognitive and motor tasks in dual-task gait testing in people with mild cognitive impairment. *BMC Geriatr*. 2018;18(1):202. doi:10.1186/s12877-018-0894-0
- Snitz BE, Wang T, Cloonan YK, et al. Risk of progression from subjective cognitive decline to mild cognitive impairment: the role of study setting. *Alzheimers Dement*. 2018;14(6):734–742. doi:10.1016/j.jalz.2017.12.003

19. Zhang ZX, Hong X, Li H, et al. The mini-mental state examination in the Chinese residents population aged 55 years and over in the urban and rural areas of Beijing. *Chin J Neurol.* 1999;32(3):149–153.
20. Morris JC. The Clinical Dementia Rating (CDR): current version and scoring rules. *Neurol.* 1993;43(11):2412–2414. doi:10.1212/WNL.43.11.2412-a
21. Mahoney FI, Barthel DW. Functional evaluation: the barthel index. *Md State Med J.* 1965;14:61–65.
22. Su XN, Hua QZ, Zhang L, Li NN, Cheng JH, Zhang LP. Influencing factors of mild cognitive impairment of seniors in communities of Xi'an. *J Nurs.* 2013;11:10–13.
23. Fu C, Yang F, Mao ZF. Mild cognitive impairment and its influencing factors of empty nest elderly in social district of Wuhan. *Chin J Gerontology.* 2018;38(23):220–222.
24. Li X, Ma C, Zhang J, et al. Prevalence of and potential risk factors for mild cognitive impairment in community-dwelling residents of Beijing. *J Am Geriatr Soc.* 2013;61(12):2111–2119. doi:10.1111/jgs.12552
25. Bourassa KJ, Memel M, Woolverton C, Sbarra DA. Social participation predicts cognitive functioning in aging adults over time: comparisons with physical health, depression, and physical activity. *Aging Ment Health.* 2017;21(2):133–146. doi:10.1080/13607863.2015.1081152
26. Stern Y, Barulli D. Cognitive reserve. *Neuropsychologia.* 2009;47(10):2015–2028. doi:10.1016/j.neuropsychologia.2009.03.004
27. Gamba P, Testa G, Gargiulo S, et al. Oxidized cholesterol as the driving force behind the development of alzheimer's disease. *Front Aging Neurosci.* 2015;7:119. doi:10.3389/fnagi.2015.00119
28. Whitmer RA, Sidney S, Selby J, Johnston SC, Yaffe K. Midlife cardiovascular risk factors and risk of dementia in late life. *Neurology.* 2005;64(2):277–281. doi:10.1212/01.WNL.0000149519.47454.F2
29. Stern Y. Cognitive reserve in ageing and alzheimer's disease. *Lancet Neurol.* 2012;11(11):1006–1012. doi:10.1016/S1474-4422(12)70191-6
30. Zhu XY, Qiu CX, Zeng Y, Li J. Leisure activities, education, and cognitive impairment in Chinese older adults: a population-based longitudinal study. *Int Psychogeriatr.* 2017;29(5):727–739. doi:10.1177/S1041610216001769
31. Xie LQ, Deng YL, Tang Y, Zhou J. The status quo of resilience in empty-nest elderly with mild cognitive impairment in the rural area and path analysis on the influencing factors. *Chin J Behav Med Brain Sci.* 2015;24(1):66–68.
32. Qi GY, Mi YS, Wang YW, et al. Neuroprotective action of tea polyphenols on oxidative stress-induced apoptosis through the activation of the TrkB/CREB/BDNF pathway and Keap1/Nrf2 signaling pathway in SH-SY5Y cells and mice brain. *Food Funct.* 2017;8(12):4421–4432. doi:10.1039/C7FO00991G
33. Reitz C, den Heijer T, van Duijn C, Hofman A, Breteler MM. Relation between smoking and risk of dementia and alzheimer disease: the Rotterdam Study. *Neurol.* 2007;69(10):998–1005. doi:10.1212/01.wnl.0000271395.29695.9a
34. Lobo V, Patil A, Phatak A, Chandra N. Free radicals, antioxidants and functional foods: impact on human health. *Phcog Rev.* 2010;4(8):118–126. doi:10.4103/0973-7847.70902

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