Abstract: Dementia is one of the most disabling and burdensome health conditions worldwide. In this study, we identified new potential risk factors for dementia from nationwide longitudinal population-based data by using Bayesian statistics.

We first tested the consistency of the results obtained using Bayesian statistics with those obtained using classical frequentist probability for 4 recognized risk factors for dementia, namely severe head injury, depression, diabetes mellitus, and vascular diseases. Then, we used Bayesian statistics to verify 2 new potential risk factors for dementia, namely hearing loss and senile cataract, determined from the Taiwan’s National Health Insurance Research Database.

We included a total of 6546 (6.0%) patients diagnosed with dementia. We observed older age, female sex, and lower income as independent risk factors for dementia. Moreover, we verified the 4 recognized risk factors for dementia in the older Taiwanese population; their odds ratios (ORs) ranged from 3.469 to 1.207. Furthermore, we observed that hearing loss (OR = 1.577) and senile cataract (OR = 1.549) were associated with an increased risk of dementia.

We found that the results obtained using Bayesian statistics for assessing risk factors for dementia, such as head injury, depression, DM, and vascular diseases, were consistent with those obtained using classical frequentist probability. Moreover, hearing loss and senile cataract were found to be potential risk factors for dementia in the older Taiwanese population. Bayesian statistics could help clinicians explore other potential risk factors for dementia and for developing appropriate treatment strategies for these patients.
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MATERIALS AND METHODS

Data Sources and Ethical Consideration

Since its implementation in 1995, the Taiwan’s National Health Insurance (NHI) program has been providing comprehensive, unified, and universal healthcare services to approximately 99% of the Taiwanese population.11 We used one of the subsets of the Taiwan’s NHIRD, developed by the NHI program using 1995 to 2010 data and consisting of 1 million patients (approximately 5% of the total Taiwanese population) randomly selected in 2010. The NHIRD contains data on patients’ demographics, diagnoses, medication types, prescription dates, and dosages and durations of drug supply.

All data used for the present study are available after others apply to the Center for Biomedical Resources of National Health Research Institutes in Taiwan and should be submitted to the executive committee (for more information please you refer to http://nhird.nhri.org.tw/en/). Furthermore, this study protocol was also approved by an institutional review board (KMUHIRB-SV (II)-20150007), and informed consent was waived because of the use of previously stored deidentified medical information from the NHIRD.

TABLE 1. Sociodemographic Characteristics of the Enrolled Elders, Stratified by Presence or Absence of Dementia From 1995 to 2010 (N = 109,709)

| Patient characteristics                        | Patients With Dementia (n = 6546) | Patients Without Dementia (n = 103,163) | P       |
|-----------------------------------------------|----------------------------------|----------------------------------------|---------|
| Age (≥65, y), mean (SD)                       | 80.2 (2.2)                       | 74.3 (1.3)                             | <0.001  |
| Age group (y), n (%)                          |                                  |                                        |         |
| Young–old (65–74)                             | 1522 (2.5)                       | 58,422 (97.5)                         | <0.001  |
| Old–old (75–84)                               | 3121 (8.1)                       | 35,226 (91.9)                         |         |
| Oldest–old (≥85)                              | 1903 (16.7)                      | 9515 (83.3)                           |         |
| Gender, n (%)                                 |                                  |                                        | <0.001  |
| Male                                          | 2848 (5.4)                       | 49,749 (94.6)                         |         |
| Female                                        | 3698 (6.5)                       | 53,414 (93.5)                         |         |
| Geographic region of Taiwan, n (%)            |                                  |                                        | 0.241   |
| Central                                       | 1155 (5.8)                       | 18,878 (94.2)                         |         |
| Northern                                      | 3013 (6.0)                       | 46,941 (94.0)                         |         |
| Southern                                      | 2164 (5.9)                       | 34,327 (94.1)                         |         |
| Eastern                                       | 216 (6.7)                        | 3017 (93.3)                           |         |
| Urban level, n (%)                            |                                  |                                        | 0.215   |
| Urban                                         | 3990 (6.1)                       | 61,374 (93.9)                         |         |
| Suburban                                      | 274 (5.3)                        | 4853 (94.7)                           |         |
| Rural                                         | 2282 (5.8)                       | 36,936 (94.2)                         |         |
| Monthly income, NTS, n (%)                    |                                  |                                        | <0.001  |
| Low (0–15,000)                                | 2846 (9.6)                       | 26,933 (90.4)                         |         |
| Median (15,000–29,999)                        | 3147 (5.2)                       | 57,149 (94.8)                         |         |
| High (≥30,000)                                | 553 (2.8)                        | 19,082 (97.3)                         |         |

SD = standard deviation.

Study Design and Sampling

The study group consisted of patients with dementia diagnosed on the basis of the International Classification of Disease, Ninth Revision, Clinical Modification (ICD-9-CM) diagnostic criteria (ICD-9-CM codes 290, 294.1–294.2, 331.0, A210, A213, and A222) between March 1995 and December 2010. To enhance diagnostic validity, we only selected patients who had inpatient diagnosis files with primary or secondary diagnosis of dementia or outpatient diagnosis files with at least three consistent diagnosis of dementia.12 We assigned their first visit for the diagnosis of dementia as their index date.

Potential Risk Factors Associated in Patients With Dementia

We assessed patients with dementia before or on their index date to ascertain their histories of severe head injury (ICD-9-CM codes 431, 430, 800–804, 850–852, 853.1–853.2, 854.0–854.1, A290, A291, and A470), depression (ICD-9-CM codes 296.2–296.3, 296.82, 293.83, 300.4, and 311), DM (ICD-9-CM codes 249–250, 648.01, 648.02, 588.1, 357.2, and A181), vascular diseases (ICD-9-CM codes 325, 430–459, 410, 411.0, 411.1, 411.81, 411.89, 412, 413.1, 413.9, 414.00–414.07, 414.10–414.12, 414.19, 414.2–414.4, 414.8, 414.9, A279, A291-A294, and A299), senile cataract (ICD-9-CM codes 366 and A231), and hearing loss (ICD-9-CM codes 389 and A241) (see Table e-1 in the Appendix; http://links.lww.com/MD/B10).13–16 In addition, we examined the inpatient and outpatient diagnosis files of patients without dementia between 1995 and 2010 to ascertain their histories of severe head injury, depression, DM, vascular diseases, senile cataract, and hearing loss. Sociodemographic data of enrolled patients was also recorded, including age, gender, geographic region, urban level, and monthly income (see Table 1).
Statistical Analysis

We derived the odds ratio (OR) by using the Bayesian approach.\textsuperscript{17} We attempt to learn about the unknown distribution from given data, to make some inferences about the certain properties of the distribution, and to determine the relative likelihood that each possible distribution is actually the correct one. Suppose that proportion \( \theta \) of patients with dementia in a population in the presence of a risk factor is unknown, and let the prior distribution assigned to \( \theta \) be a uniform distribution at the interval \((0, 1)\); that is, the prior p.d.f. \( \xi(0) = 1 \) for \( 0 < \theta < 1 \).

This “informationless” prior assignment is for the sake of objective purpose and reducing computing cost. Suppose there is a given random sample of \( n \) persons who all have a risk factor, and for \( i = 1, 2, 3, \ldots, n \), let \( X_i = 1 \) if the \( i \)th person has dementia and \( X_i = 0 \) otherwise. Then, \( X_1, X_2, X_3, \ldots, X_n \) form \( n \) Bernoulli trials with parameter \( \theta \). We can determine the posterior p.d.f. of \( \theta \). The p.f. of each observation \( X_i \) is

\[
f(x|\theta) = \theta^x(1 - \theta)^{1-x} \quad \text{for} \quad x = 0, 1.
\]

Otherwise \( f(x|\theta) = 0 \).

Hence, if \( y = x_1 + x_2 + x_3 + \ldots + x_n \), the joint p.f. of \( X_1, X_2, X_3, \ldots, X_n \) can be written in the following form for \( x_i = 0 \) or 1 \((i = 1, 2, 3, \ldots, n)\):

\[
f_n(x|\theta) = \theta^y(1 - \theta)^{n-y}
\]

Since the prior p.d.f. \( \xi(0) = 1 \) for \( 0 < \theta < 1 \), it follows that

\[
f_n(x|\theta)\xi(0) = \theta^y(1 - \theta)^{n-y}
\]

It is a beta distribution with parameters \( \alpha = y + 1 \) and \( \beta = n - y + 1 \). Therefore, for \( 0 < \theta < 1 \),

\[
\xi(\theta|x) = \frac{\Gamma(n+2)}{\Gamma(y+1)\Gamma(n-y+1)}\theta^y(1 - \theta)^{n-y}
\]

That is, the above derived result illustrates that the posterior distribution of the combination of uniform prior distribution and the likelihood function of Bernoulli trials is Beta distribution. The mean of \( \theta \) given the observation \( x \) is \( \mathbb{E}(\theta|x) = \frac{y}{n} \), which is also the estimate of \( \theta \) based on the squared error loss function that is to be derived later. Let \( \theta^* \) be the estimate of \( \theta \); then, the OR is equal to \( \frac{\theta^*}{1 - \theta^*} \). Similarly, we can get the OR of cases in a population in the absence of a risk factor, when the squared error loss function is used, the Bayes estimator determined using specific observed values \( x_1, x_2, x_3, \ldots, x_n \) and using the squared error loss function, which is the most commonly used loss function in estimation problem, \( L(\theta, a) = (\theta - a)^2 \). Bayes estimate \( \delta^*(x) \) for any observed value of \( x \) will be the value of a for which the expectation \( \mathbb{E}(\theta-a)^2|x] \) is minimum. This expectation is minimum when \( a \) is chosen to be equal to the mean \( \mathbb{E}(\theta|x) \) of the posterior distribution; in other words, when the squared error loss function is used, the Bayes estimator is \( \delta^*(X) = \mathbb{E}(\theta|X) \). Therefore, we can use the mean of posterior probability distribution as the estimate.

We used the aforementioned Bayesian inference—given the evidence of 1 million population, sampled from the NHIRD—to determine the posterior probabilities of \( \theta \). The posterior distribution is insensitive to variations in the prior distribution; this insensitivity rapidly increases with the sample size.\textsuperscript{18} This is the reason why we do not use a strong subjective prior and just choose an informationless prior. Furthermore, this can also reduce the computation cost. The probability calculus does more than explaining how states of belief (i.e., the proportion \( \theta \) of dementia in a population in the presence of a risk factor) decompose into prior and empirical (i.e., the NHIRD) elements; it also measures the probability \( \theta \) relative to the NHIRD. The prior distribution of \( \theta \) is a uniform distribution of subjective prior probabilities over the range \((0, 1)\). Thus, using the Bayes’ theorem and the NHIRD sample, we calculated the corresponding posterior probability to be a beta distribution as stated earlier. Thus, the posterior probability is concentrated in a narrow region even if the prior probability curve is spread out, illustrating the considerable initial uncertainty regarding the parameter value.

RESULTS

Demographic Characteristics of the Study Population

Of the 1,000,000 patients screened, we included 109,709 (11.0%) patients aged \( \geq 65 \) years. A total of 6546 (6.0%) patients were diagnosed with dementia. Table 1 shows the distribution of sociodemographic characteristics of the study population. Patients with dementia were significantly older than those without dementia (80.2 years vs 74.3 years, \( P < 0.001 \)). Moreover, the percentage of patients with dementia in the young—old (65–74 years), old—old (75–84 years), and old—est—old (\( \geq 85 \) years) groups was significantly increased by 2.5%, 8.1%, and 16.7%, respectively (\( P < 0.001 \)). The percentage of women with dementia was significantly higher than that of men with dementia (\( n = 3698, 6.5\% \) vs 2848, 5.4%; \( P < 0.001 \)); furthermore, patients with high income had significantly lower incidence of dementia than in those with lower income (2.8%, 5.2%, and 9.6%; \( P < 0.001 \)).

Risk Factors for Dementia

Regarding the 4 recognized risk factors for dementia, ORs in the older population with a history of severe head injury, depression, DM, and vascular diseases were 1.701, 2.637, 1.207, and 3.469, respectively (Table 2); ORs for these risk factors were higher in men with dementia than in women with dementia except for DM. Furthermore, we tested ORs of other potential risk factors for dementia, including senile cataract and hearing loss. We observed that hearing loss (OR = 1.577) and senile cataract (OR = 1.492) were associated with an increased risk of dementia in the older population. In terms of sex-based differences, ORs of these risk factors were generally higher in men with dementia than in women with dementia (Table 2).

DISCUSSION

To the best of our knowledge, this is the first nationwide population-based study identifying new potential risk factors for dementia using Bayesian statistics. Overall, older age, female sex, and lower income were identified as independent risk factors for dementia in this study. By using Bayesian statistics to assess the older population in Taiwan, we reidentified the 4 recognized risk factors for dementia, namely vascular diseases, depression, severe head injury, and DM; their ORs
that in other developed countries. The risk of dementia for population from 1995 to 2010 was approximately 6%, similar to data demonstrated that the prevalence of dementia in the older population exceeded 10% of the total population in 2005. Our results showed that approximately 11% of the population was aged ≥65 years in 2010. The prevalence of dementia in the older population in Taiwan from 1995 to 1998 was 3.2%, which is lower than that in other developed countries. Thus, our data demonstrated that the prevalence of dementia in the older population from 1995 to 2010 was approximately 6%, similar to that in other developed countries. The risk of dementia for an individual can vary with area, population, and diagnostic criteria because various sociodemographic risk factors vary, expect for age and educational level. A review on Asian populations, including that of Taiwan, has showed that older age, female sex, and lower socioeconomic status were the risk factors for dementia in the East. These results are similar to those observed in our study.

Increasing evidence supports that increased risks of dementia are associated with its treatable medical comorbidities, including severe head injury, depression, diabetes mellitus, and vascular diseases. An analysis of 15 case–control studies from 1991 to 2001 concluded that a history of head injury is a significant risk factor for Alzheimer disease (AD, OR = 1.58), and OR for this risk factors is higher in men than in women. However, the 95% CI does not report a measure’s statistical significance. It would be inappropriate to interpret an OR with a large CI as indicating lack of association between the risk factor and dementia. Because the frequentist interpretation of probability considers long-run frequencies rather than data at hand, it is impossible to assign a probability to a hypothesis, given the data. Frequentist hypothesis tests do not technically test any hypothesis, given the data; instead, the test the data, given the hypothesis. The Bayesians approximation) and still permits exact probability statements. Bayesian inference handles uncertainty better than frequentist approximation (as opposed to asymptotic approximation) and still permits exact probability statements.

Table 2. Risk Factors (Odds Ratios) for Dementia by Age and Gender

| Age | Severe Head Injury | Depression | Diabetes Mellitus | Vascular Diseases |
|-----|-------------------|------------|------------------|------------------|
| ≥65 | 1.701             | 2.637      | 1.207            | 3.469            |
| Male| 1.795             | 2.848      | 1.121            | 3.967            |
| Female| 1.658            | 2.455      | 1.250            | 3.135            |
| <65 |                   |            |                  |                  |

Sample size that is too large or too small is a potential issue in classical frequentist hypothesis test. In the case of a too large sample size, hypothesis testing becomes biased in favor of rejecting the null hypothesis. When a sample size is too small, the frequentist reliance on asymptotic approximations becomes inappropriate and the estimates become biased. By contrast, Bayesian inference does not force an artificial dichotomy between a null and alternative hypothesis, allows exact probability statements about any hypothesis (ie, 0 here), and is not biased due to a large sample size (eg, 1 million sample). Regarding a small sample size, Bayesian inference is also unbiased due to exact estimation (as opposed to asymptotic approximation) and still permits exact probability statements. Bayesian inference handles uncertainty better than frequentist
inference, and the probability intervals of parameters are appropriately wide. In addition, the findings from the experimental result shown in Table 2 suggest that Bayesian inference can provide a similar trend as frequentist inference and thus can be helpful in assessing risks for dementia. Furthermore, because all observed data are used to compute posterior probability distribution, any bias from the data sampling, which is often criticized in frequentist hypothesis test, can be avoided. Furthermore, Bayesian approach is helpful in assessing the effect of the coexistence of multiple risk factors, which has not been discussed yet in research studies by using the analysis method of classical frequentist hypothesis test.

Strengths and Limitations

The main strength of our study is the use of nationwide population-based data that accurately represent the general population. Majority of the events could be traced and referral bias is minimized because the Taiwan’s NHII program is a single-payer, mandatory health insurance with affordable payments and services covering approximately 99% of the Taiwanese population. In addition, one of the major benefits of the Bayesian method is its ability to incorporate prior information. Most traditional risk factor assessment approaches use prior information by illustrating the levels or ranges of individual parameters in sensitivity analysis, the Bayesian method analyzed historical data sets or refers expert and domain knowledge to determine what is known about biological parameters and processes. While other risk factor assessment approaches can be more susceptible to biases than a prospective design; this study employed a retrospective study design, which tends to all observed data are used to compute posterior probability distribution, any bias from the data sampling, which is often criticized in frequentist hypothesis test, can be avoided. Furthermore, Bayesian approach is helpful in assessing the effect of the coexistence of multiple risk factors, which has not been discussed yet in research studies by using the analysis method of classical frequentist hypothesis test.

The main strength of our study is the use of nationwide population-based data that accurately represent the general population. Majority of the events could be traced and referral bias is minimized because the Taiwan’s NHII program is a single-payer, mandatory health insurance with affordable payments and services covering approximately 99% of the Taiwanese population. In addition, one of the major benefits of the Bayesian method is its ability to incorporate prior information. Most traditional risk factor assessment methods do not use any of the quantitative information that could be gathered from historical experiences with other risk factors and thus treat each risk factor assessment as a new and independent problem. However, it is extremely computationally intensive to apply Bayes theorem to complex models. It often takes days to conduct defensible decision analyses for assessments using computers. We tested consistency between results obtained using classical frequentist probability and Bayesian statistics in the 4 recognized risk factors for dementia (head injury, depression, DM, and vascular disease). Moreover, Bayesian statistics could effectively help clinicians and researchers to explore more potential risk factors for diseases like dementia using big data such as the Taiwan’s NHIRD. Moreover, our results may provide a good representation of risk factors for patients with dementia in ethnic Chinese populations. Furthermore, the diagnoses of dementia and its comorbidities were reliable because these health insurance claims are scrutinized by medical reimbursement specialists and subject to peer review.

Nevertheless, there are some limitations of our study. First, the Taiwan’s NHIRD does not include detailed information on biochemistry data, body mass index, clinical severity of diseases, family history, and lifestyle, which may be potential risk factors or comorbidities for dementia. As an example, APOE gene is recognized as a substantial factor in the majority of patients with dementia, but this biochemistry data are unavailable in the NHIRD. Similarly, lifestyle-related dementia risk factors, including physical activity, dietary habits, cigarette smoking, and alcohol consumption, were not available. Second, this study employed a retrospective study design, which tends to be more susceptible to biases than a prospective design; therefore, we avoided 2 possible major biases related to sample problem (selecting random patients from nationwide data) and recall problem (analyzing national health care records). Finally, all administrative databases are subject to possible coding errors or undercoding.

CONCLUSION

The findings of this study concur with our 2 hypotheses. First, our results of the 4 recognized risk factors for dementia, namely severe head injury, depression, DM, and vascular diseases, are consistent with both classical frequentist probability and Bayesian statistics. Second, hearing loss and senile cataract were found to be potential risk factors for dementia in the older Taiwanese population. Bayesian statistics could help clinicians to explore more risk factors for illnesses, such as dementia, to develop appropriate treatment strategies for these patients.

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