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ABSTRACT

Introduction: Published estimates of Alzheimer's disease (AD) progression do not capture the full disease continuum. This study provides transition probabilities of individuals with amyloid- β (A β +) pathology across the disease continuum.

Methods: Patient-level longitudinal data from the National Alzheimer's Coordinating Center

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were used to estimate progression rates. Progression rates through five clinically defined AD stages— asymptomatic, mild cognitive impairment due to AD (MCI-AD), mild AD dementia, moderate AD dementia, severe AD dementia— and death were measured as transition probabilities. Rates were assessed in “incident” patients who recently entered the stage, controlling for covariates. Transition probabilities were generated from multinomial logit regression models that predicted an individual's health state as a function of health state at the previous visit and adjusted for time between initial and follow-up visits, age, sex, years of education, and concomitant symptomatic AD medications.

Results: Annual transition probabilities to more severe dementia stages for surviving incident A β + patients were as follows: asymptomatic to MCI-AD, 40.8%; MCI-AD to mild AD dementia or worse, 21.8%; mild AD dementia to moderate AD dementia or worse, 35.9%; moderate AD dementia to severe AD dementia, 28.6%. Transition probabilities to less severe dementia stages were: 5.3% annual reversion from MCI-AD to asymptomatic, 3.0% mild AD dementia to MCI-AD, 1.8% moderate AD dementia to mild AD dementia, and 1.3% for severe AD dementia to moderate AD dementia.

Conclusions: These transition probabilities reflect the full continuum of AD progression in A β + individuals and can be used to assess the impact of treatment on expected transitions.

Keywords: Alzheimer's disease; Beta amyloid; Dementia; Disease progression; Mild cognitive impairment; Transition probabilities; Economic modeling

Key Summary Points

Why carry out this study?

Alzheimer's disease (AD) imposes an enormous economic burden; therefore, valid economic models of disease progression are critical to inform decision-making for payers, the healthcare system, and society.

This study used patient-level longitudinal data from the National Alzheimer's Coordinating Center Uniform Data Set to estimate annual AD progression rates across the five clinically defined AD stages in individuals with biomarker-confirmed AD pathology.

What was learned from the study?

Annual transition probabilities to more severe AD stages were 40.8% (asymptomatic to mild cognitive impairment due to AD), 21.8% (mild cognitive impairment due to AD to mild AD dementia or worse), 35.9% (mild AD dementia to moderate AD dementia or worse), and 28.6% (moderate AD dementia to severe AD dementia).

Calculating transition probabilities using a population of patients with biomarker confirmation of AD, for which the full time spent in the prior stage of disease is accounted for, may increase the precision of estimates.

The estimates derived from this study directly inform the current understanding of AD progression and can aid in trial design, care planning, and benefit assessments of AD interventions that reduce progression rates.

INTRODUCTION

Alzheimer's disease (AD) is a neurodegenerative disorder in which individuals progress clinically from the asymptomatic stages of disease to mild cognitive impairment (MCI) and finally to the mild, moderate, and severe stages of dementia [1–3]. As the most common cause of dementia, AD accounts for approximately 60–80% of cases and affects > 5 million patients in the US and approximately 36 million people worldwide [4, 5]. The Alzheimer's Association estimates that, by 2050, the number of individuals aged > 65 years with AD in the US will increase to 13.8 million, and costs of AD and other dementias will reach \$1.1 trillion (in 2018 dollars) [4].

Because of the large and growing economic burden of AD, valid economic models of the disease are critical to inform decision-making for payers, the healthcare system, and society. These models require transition probability estimates that accurately reflect progression across the disease continuum. Without such estimates, models assessing the cost-effectiveness of new AD therapies risk over- or underestimating the true benefits and costs of such interventions.

Multiple estimates of AD clinical progression rates and transition probabilities across health states have been published, although these vary widely because of differences in data sources, patient sample selection criteria, and analytic methodology [6–11]. For example, estimates of the annual transition probability from severe AD dementia to death range from 48.0 [7] to 89.8% [8]. In addition, few studies have attempted to estimate transitions across the full range of disease severity. This is likely due to a paucity of information on relevant indicators of AD early in the disease, preventing estimates of early transition rates. Furthermore, these estimates often do not account for the duration of time in a particular state, yet the probability of transitioning to a more severe dementia state of AD likely depends on the amount of time spent in the prior state. To capture the full spectrum of an individual's disease progression, transition probabilities must be examined in "incident"

patients, individuals who recently entered the stage and for whom the full time spent in the prior stage is known.

The current literature lacks transition probability estimates that account for patient amyloid- β ($A\beta$) status in the brain. $A\beta$ is both a core biomarker of AD pathophysiology and a major target of disease-modifying therapies [12]. Thus, biomarker confirmation of disease is expected to improve models of disease progression compared with methods based solely on clinical diagnosis of AD and, in turn, improve clinical trial design and health care system planning. Although previous studies have examined the rate of progression in patients with known amyloid pathology and neurodegeneration, the full spectrum of disease was not captured [9, 10]. Furthermore, Davis et al. [11] examined transition probabilities across the full spectrum of disease severity; however, no biomarkers were assessed.

To address previous methodological limitations, the current study used patient-level longitudinal data from the National Alzheimer's Coordinating Center (NACC) Uniform Data Set (UDS) [13] to estimate annual progression rates in $A\beta$ -bearing ($A\beta+$) individuals, using a consistent methodology across the entire AD continuum.

METHODS

Data and Sample Selection

The NACC provided data for this study [13, 14]. The NACC UDS [13] was used to estimate the transition probabilities associated with AD. This data set contains information on nearly 30,000 patients with varying degrees of cognitive impairment collected from September 2005 through December 2017, from approximately 30 Alzheimer's Disease Research Centers across the US. The NACC conducts quarterly data freezes; therefore, the data used to estimate transition probabilities are specific to the data freeze in December 2017 and the eligibility criteria upon which the data file was created. Data

were gathered prospectively by clinicians, neuropsychologists, and other research personnel using standardized protocols and include information on sociodemographic characteristics, dementia history, neurological examination findings, functional status, neuropsychological test results, clinical diagnosis, and apolipoprotein E (ApoE) genotype. In addition, the NACC Neuropathology Data Set from this specific data freeze provides autopsy information for approximately 3000 patients. Because the NACC data were completely de-identified, this study did not qualify as Human Subjects Research under US Department of Health and Human Services Protection of Human Subjects regulations based on guidance from the University of Washington Human Subjects Division. To be included in the sample, patients must have had at least two visits observed in the NACC data. They must also have had either a diagnosis of normal cognition at all visits or a diagnosis of MCI, dementia, or AD at any visit. Determination of $A\beta$ status was required (discussed below). The unit of analysis was the patient visit. Patient visit-level observations were recorded in the NACC data on an approximately annual basis. Clinical input and NACC expertise were provided by Drs. Jeffrey Cummings and Soo Borson.

Amyloid Status

Amyloid status was constructed at the patient level based on information collected during the patient visits. In this study, $A\beta+$ was defined as having any one of the following indicated at a visit: (1) abnormally elevated amyloid on a positron emission tomography scan by visual read, (2) abnormally low amyloid per a cerebrospinal fluid assay, or (3) an autopsy result consistent with a patient having been $A\beta+$ during their lifetime, defined as frequent neocortical neuritic plaques and a Braak stage of V or VI. Patients were categorized as $A\beta+$ if they were observed to be $A\beta+$ at any visit within 10 years following their first visit assigned to the asymptomatic/preclinical, MCI due to AD (MCI-AD) or mild AD dementia health state. In

addition, a diagnosis of moderate or severe AD dementia was considered highly consistent with amyloid positivity by clinical estimation because the likelihood of A β positivity at these later stages is $\geq 90\%$ based on a meta-analysis of numerous studies [15, 16]. Individuals who were determined to be A β + represented 14.8% of the total population in this NACC data set (a large number of patients in the NACC database have not had amyloid measures of the type used in this study).

Measures

Health State Definitions

Visits were assigned to one of six health states listed here in ascending order of disease severity: (1) asymptomatic/preclinical, (2) MCI-AD, (3) mild AD dementia, (4) moderate AD dementia, (5) severe AD dementia, and (6) determined to have died since the last visit. Criteria for the first five health state definitions were based on the CDR[®] Dementia Staging Instrument. The Clinical Dementia Rating scale Sum of Boxes (CDR-SB) scores included in the NACC data were used to generate the CDR-based transition probabilities. CDR-SB values for asymptomatic, MCI-AD, mild AD dementia, moderate AD dementia, and severe AD dementia were 0, 0.5–4.0, 4.5–9.0, 9.5–15.5, and 16.0–18.0, respectively [17].

Patient death was captured via an NACC indicator variable (NACCDIED). This variable was derived from the autopsy data and the NACC Milestones form.

Covariates

Six patient visit-level covariates relevant to transition probabilities were identified in the NACC data, of which five were used in this analysis: (1) time elapsed between current and next visits (measured in years), (2) age at visit (measured in years), (3) patient sex (male, female), (4) years of education, and (5) concomitant symptomatic AD medications identified based on National Institute for Health and

Care Excellence guidance [18] (measured as a binary indicator based on the presence of any of the following medications at any visit: donepezil, galantamine, rivastigmine, memantine).

Additionally, ApoE status (positive vs. negative for the $\epsilon 4$ allele) was reviewed; specifically, ($\epsilon 4$, $\epsilon 4$), ($\epsilon 4$, $\epsilon 2$), and ($\epsilon 3$, $\epsilon 4$) are the genotypes defined as ApoE $\epsilon 4$ -positive.

Sample Construction

Construction of Populations Based on Observability of Entry into Health State

To explore the impact of potential limitations to duration of illness captured within the NACC data, patient visit-level observations were constructed based on whether entry into a health state was observed in the data set. These are referred to as “prevalent” and “incident” populations. The prevalent population did not restrict membership; all observations for patients with at least two visits were included. The incident population included only those in which a patient’s new entry into a health state was observed in the data. The incident population is a subset of the prevalent population, so individual patients could contribute visit observations to the calculation of multiple transition probabilities. An illustrative example of visits satisfying the criteria for incident is displayed in Fig. 1. The incident visit population consisted of visits for which an antecedent transition to that visit’s health state was observed in the data. For example, a patient’s first recorded visit may have been with a health state of MCI-AD, which would not be considered incident because there was no prior visit. If the patient’s second visit was for the health state of mild AD dementia, that visit was considered incident, because the patient was observed to have transitioned from MCI-AD to mild AD dementia. The analytic sample was further limited to patient visits for which none of the covariates was missing and an immediately preceding visit was observed.

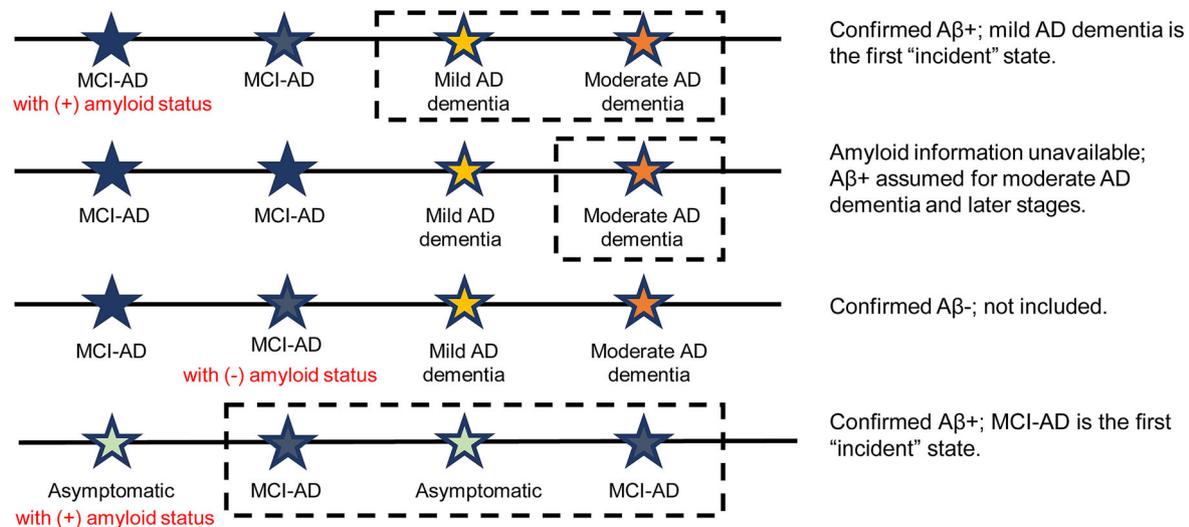


Fig. 1 Incident patient determination. Colors correspond to the labeled stage of disease. Observations that would be included are indicated in the dashed-line boxes.

A β amyloid- β , AD Alzheimer's disease, MCI-AD mild cognitive impairment due to Alzheimer's disease

Analytic Methods

Descriptive Analyses

Descriptive analyses were conducted to provide a clinical perspective of the patients in the analytic sample. The main patient characteristics of interest (age, sex, ApoE ϵ 4, CDR-SB, years of education, concomitant symptomatic AD medications, and comorbidity profile) were assessed. Measures were summarized with counts and proportions for categorical variables and means and standard deviations for continuous variables.

Prevalence was calculated for the following comorbidities at baseline: diabetes and cardiovascular conditions, including hypertension, hypercholesterolemia, angioplasty/endarterectomy/stent, atrial fibrillation, heart attack/cardiac arrest, cardiac bypass procedure, pacemaker, congestive heart failure, other cardiovascular disease, any cerebrovascular disease, stroke, or transient ischemic attack.

Model Estimation

The product of these analyses was a set of transition matrices describing the likelihood of moving from each initial AD health state to any other state in 1 year (including remaining in the

initial state). Each transition matrix was constructed from the data for a specific patient population in which consecutively observed pairs of visits (initial, follow-up) formed a potential state transition. Because the health states were exhaustive, the sum of transition probabilities from any initial state was 1.

Transition matrices were constructed in three steps. First, patient visit-level data were organized according to initial AD state (i.e., the state of the first observation in each pair). Next, a full-information maximum likelihood multinomial logit model was developed to predict the outcome state conditional on the initial state. This approach was used to account for censoring from lack of follow-up as well as simultaneous competing risks of transitions to other states. Adjusted models controlled for time between initial and follow-up visits, patient age, sex, and concomitant symptomatic AD medications. ApoE ϵ 4 was excluded because it was highly collinear with the other predictors in some populations, which led to failures of model convergence. Model estimation produced a transition equation for each pair (initial state–follow-up state) that relates the predictors to the probability of transitioning via the estimated coefficients. In the last step, the transition equations were in turn used to calculate

mean transition probabilities for each pair (initial state–follow-up state). For k transition states (and the base set to 1), the multinomial logit formula for calculating the probability that individual j transitions to the i th state (p_{ij}) is

$$\hat{p}_{ij} = \Pr(y_j = i) = \begin{cases} \frac{1}{1 + \sum_{m=2}^k \exp(x_j \beta_m)}, & \text{if } i = 1 \\ \frac{\exp(x_j \beta_i)}{1 + \sum_{m=2}^k \exp(x_j \beta_m)}, & \text{if } i > 1 \end{cases},$$

where y_j is individual j 's realized outcome, x_j is the row vector of observed values for individual j 's covariates, and β_m is the coefficient vector from the model predicting transition to state m . The simple mean of the predicted probabilities of transitioning to state i is calculated over the N_j individuals as

$$\bar{p}_i = \frac{\sum_{n=1}^{N_j} \hat{p}_{in}}{N_j}.$$

Four covariates were held at constant values for annual transition probability calculations, as follows: (1) time between visits: 365 days; (2) female: 0.5 (representing a 50/50 split in the distribution of patient sex); (3) age: 77 years; (4) years of education: 14 years. The fifth covariate, a binary indicator for whether the patient had concomitant symptomatic AD medications, was allowed to vary; the transition probability for each patient was calculated using that patient's own value for the concomitant medication covariate. Note that covariate values were informed by the data and that the estimated coefficients can generate transition probabilities for any appropriate covariate values.

RESULTS

Sample Characteristics

Characteristics of the sample are displayed by incident and prevalent populations (Table 1). The incident population consisted of 3291 unique patients, and the prevalent population

comprised 4370 patients. The total numbers of patient visit-level observations were 7027 and 11615 for the incident and prevalent populations, respectively. Tables S1 and S2 in the electronic supplementary materials provide the counts of observed transitions and unique patients by initial health state for the incident and prevalent populations, respectively. Most of the variables assessed showed little difference between these populations, except for CDR-SB health state. In the incident population, the percentages of the asymptomatic, MCI-AD, mild AD dementia, and moderate AD dementia patients were generally lower compared with patients in the prevalent population. This result was as expected because the incident population is captured in the prevalent population. Conversely, the percentages of patients in the incident population with severe AD dementia or those who had died were higher than those observed for patients in the prevalent population.

Transition Probability Estimates

In this study, annual transition probability estimates were generated for the incident and prevalent populations from a set of transition matrices that modeled the likelihood of moving from each initial AD health state to any other state. Table 2 provides a summary and comparison of the transition estimates across populations after adjustment for time between initial and follow-up visits, patient age, sex, and concomitant symptomatic AD medications. The table shows the percentage of patients in each state who progressed to a more severe dementia state over 12 months. In the incident population, the percentage of asymptomatic patients transitioning to more severe dementia disease states was noticeably higher (40.8%) than that observed for the prevalent population (23.4%).

The complete adjusted annual transition probabilities for the incident and prevalent populations are shown in Tables 3 and 4, respectively. Annual transition probabilities to more severe dementia stages for surviving patients with known amyloid status were 40.8% for transitioning from asymptomatic to MCI-

Table 1 Sample summary statistics among amyloid-restricted patients for both the incident and prevalent populations

Variable	Incident	Prevalent
Female, % (<i>N</i>)	51.4% (3610)	51.0% (5923)
Age		
Mean (SD)	78.69 (10.43)	77.72 (10.48)
Median (IQR)	80.00 (14.00)	79.00 (14.00)
Years of education		
Mean (SD)	14.84 (3.58)	14.73 (3.73)
Median (IQR)	16.00 (6.00)	16.00 (6.00)
Concomitant symptomatic AD medication use, % (<i>N</i>)	64.4% (4528)	64.5% (7487)
ApoE ϵ 4 genotype, % (<i>N</i>)		
ϵ 3, ϵ 3	32.5% (2285)	32.4% (3761)
ϵ 3, ϵ 4	35.6% (2505)	36.2% (4206)
ϵ 3, ϵ 2	4.6% (326)	4.0% (462)
ϵ 4, ϵ 4	11.5% (806)	11.0% (1282)
ϵ 4, ϵ 2	2.4% (170)	2.3% (266)
ϵ 2, ϵ 2	0.3% (20)	0.2% (27)
Missing	13.0% (915)	13.9% (1611)
CDR health state, % (<i>N</i>)		
Asymptomatic	0.7% (50)	2.9% (333)
MCI-AD	3.1% (219)	9.3% (1077)
Mild AD dementia	6.3% (441)	11.9% (1387)
Moderate AD dementia	23.5% (1648)	21.5% (2495)
Severe AD dementia	34.1% (2394)	30.6% (3553)
Died	32.4% (2275)	23.8% (2770)
Diabetes	13.7% (752)	13.6% (1288)
CVD risk factors	78.4% (4297)	76.6% (7257)
Count of CVD components		
0	64.7% (3547)	66.6% (6317)
1	19.4% (1064)	18.7% (1774)
2	9.2% (506)	8.8% (832)
3	4.3% (235)	3.9% (372)
4	1.5% (84)	1.3% (125)
5	0.6% (31)	0.5% (43)
6	0.3% (14)	0.2% (15)

Table 1 continued

Variable	Incident	Prevalent
Stroke/TIA	18% (972)	16% (1560)
Sample size (observations)	7027	11615

AD Alzheimer's disease, *ApoE* apolipoprotein E, *CDR* Clinical Dementia Rating, *CVD* cardiovascular disease, *IQR* interquartile range, *MCI-AD* mild cognitive impairment due to Alzheimer's disease, *TIA* transient ischemic attack
Sample reflects observations used in transition probability models

Table 2 Annual transition probabilities by current health state and sample populations

Sample variant	Annual probability of transitioning to a more severe dementia health state from current health state				
	Asymptomatic	MCI-AD	Mild AD dementia	Moderate AD dementia	Severe AD dementia ^a
Incident population	40.8%	26.5%	45.2%	59.8%	46.7%
Prevalent population	23.4%	36.9%	43.4%	55.5%	41.6%

AD Alzheimer's disease, *MCI-AD* mild cognitive impairment due to Alzheimer's disease

^a Transition from severe AD dementia to death

Table 3 Annual transition probabilities among the incident population with amyloid restriction

Initial state	Next state					
	Asymptomatic	MCI-AD	Mild AD dementia	Moderate AD dementia	Severe AD dementia	Died
Asymptomatic	59.2%	40.8%	0.0%	0.0%	0.0%	0.0%
MCI-AD	5.3%	68.2%	15.9%	5.7%	0.2%	4.7%
Mild AD dementia	0.0%	3.0%	51.8%	31.6%	4.3%	9.2%
Moderate AD dementia	0.0%	0.0%	1.8%	38.4%	28.6%	31.2%
Severe AD dementia	0.0%	0.0%	0.0%	1.3%	52.0%	46.7%

AD Alzheimer's disease, *MCI-AD* mild cognitive impairment due to Alzheimer's disease

Please see Table S1 in the electronic supplementary materials for counts of observed transitions and unique patients by initial state among the incident population with amyloid restriction

AD, 21.8% for MCI-AD to mild AD dementia or worse, 35.9% for mild AD dementia to moderate AD dementia or worse, and 28.6% for moderate AD dementia to severe AD dementia (Table 3). Annual transition probabilities to less severe dementia stages were 3.0% for mild AD dementia to MCI-AD, 1.8% for moderate AD

dementia to mild AD dementia, and 1.3% for severe AD dementia to moderate AD dementia (Table 3). Estimated annual transition probabilities in the prevalent population (Table 4) were similar to results from the incident population, although the percentage transitioning from MCI-AD to mild, moderate, or severe AD

Table 4 Annual transition probabilities among the prevalent population with amyloid restriction

Initial state	Next state					
	Asymptomatic	MCI-AD	Mild AD dementia	Moderate AD dementia	Severe AD dementia	Died
Asymptomatic	76.6%	21.8%	0.7%	0.0%	0.0%	0.9%
MCI-AD	2.2%	60.9%	30.7%	3.7%	0.6%	1.9%
Mild AD dementia	0.0%	2.6%	54.0%	33.6%	4.3%	5.6%
Moderate AD dementia	0.0%	0.0%	3.0%	41.5%	30.2%	25.3%
Severe AD dementia	0.0%	0.0%	0.0%	0.9%	57.5%	41.6%

AD Alzheimer's disease, *MCI-AD* mild cognitive impairment due to Alzheimer's disease

Please see Table S2 in the electronic supplementary materials for counts of observed transitions and unique patients by initial state among the prevalent population with amyloid restriction

dementia was higher (35.0% vs. 21.8%; Table 4). Note that some patients progressed by more than one stage in 1 year (e.g., MCI-AD to moderate or severe AD dementia). The transition probabilities from each stage to death are presented in Tables 3 and 4.

DISCUSSION

This study identified transition probabilities across the AD spectrum using a patient population with A β pathology in which the entire time spent in the prior disease stage before a transition was known (i.e., incident population). It was hypothesized that transition probabilities, calculated using an A β + confirmed population for which the full time spent in the prior health state was accounted for, would increase the precision of estimates. Analyzed in this study (data not shown) and compared with published transition probabilities in which A β status was unknown, transitions observed in A β + individuals were higher in the earlier stages of AD (i.e., MCI-AD and mild AD dementia) [11]; however, transition probabilities for the moderate AD group align with previously published estimates. One explanation for this observation is that, considering that AD is the most common cause of dementia [4], most individuals with moderate to severe AD

dementia are expected to have amyloid pathology [15, 16]. Therefore, the likelihood that our later-stage disease populations would deviate significantly from those previously published is low. Additionally, when using incident population criteria, lower transition estimates were generally observed for the transition from MCI-AD to mild AD dementia and from mild AD dementia to moderate AD dementia. Collectively, these results suggest the importance of amyloid confirmation in the earlier stages of AD (i.e., to distinguish between general MCI vs. MCI due to AD) when estimating transition probabilities and that using an incident population can increase the precision of estimates.

Because studies determining transition probabilities are typically conducted with cross-sectional data that do not account for patient history of disease (so-called prevalent patients), these estimates may not account for the duration of time spent in the initial state. However, the probability of transitioning to a more severe AD dementia state is highly dependent on how far the patient has progressed through the prior state when the progression is quantified. Therefore, in this study, transition estimates were calculated in patients who were known to be "incident" to the state in order to capture the full time spent in the disease stage. Restricting the sample to this incident population

decreased the probability of transition between MCI-AD and mild AD dementia and between mild AD dementia and moderate AD dementia. Transition probabilities calculated using all observations (i.e., the prevalent population) were expected to be biased toward higher transition rates compared with those from the incident population, because patients may have spent unobserved time in their current health state prior to entering the NACC data set. Thus, without observation of the full duration of time spent in the initial state, these patients would appear to be transitioning faster than those who are incident. However, a lower transition probability was not observed in the incident population for the transition from moderate to severe AD dementia. Because the symptoms of moderate to severe AD dementia are more apparent [4], it may be that patients in these disease stages were diagnosed closer to the time of disease onset and, therefore, most of the time spent in the initial state was captured. Alternatively, lower transition probabilities may not have been uniformly observed in the data because of incomplete sample selection criteria in which the time-in-state (i.e., incident population) hypothesis is only valid for a patient sample different from the one analyzed here, patient heterogeneity, and/or random chance. Additionally, sample sizes were smaller in the early stages of disease due to the determination of A β +

The annual probability of transitioning from an asymptomatic/preclinical stage of disease to MCI-AD in the incident population of this study (40.8%) was higher than expected. Because patients in the incident sample were known to be A β +, this finding may be a result of this population being biomarker-confirmed for AD pathology and, therefore, not representative of the general asymptomatic population (which may include patients who are not A β +) . Furthermore, for individuals in the asymptomatic stage, the definition of *incident* is not likely to be applicable. Rather, this type of transition would reflect someone on the cusp of an MCI diagnosis or in an early MCI stage. In this context, a prevalent population may be more appropriate for determining transition probabilities in an asymptomatic population.

Although reversion rates were low, transitions to less severe stages were observed. Day-to-day fluctuations in clinical manifestations of disease may account for this finding. Longer follow-up would be necessary to assess whether these individuals ultimately continued to decline.

Limitations

Limitations of the data set should be acknowledged. First, although NACC protocols are designed for annual use and longitudinal follow-up, intervals between observations are not precisely 1 year (based on NACC UDS data-collection protocol). Additionally, information on amyloid presence was available only for a small subset of the NACC patient sample (from 2015 onward), and that subset may not be generalizable. In order to maximize the usable sample, a collaboration with clinical experts was undertaken to develop an approach to identify A β + patients based on amyloid testing data that occurred up to 10 years after their first visit in the MCI-AD or mild AD dementia health state. The selection of A β + patients via autopsy data may also select for patients who progressed faster and died during the study period.

CONCLUSIONS

Precise and accurate transition probabilities are critical for economic modeling of AD and planning for future interventions. However, accurately capturing natural history data in AD presents a host of challenges, especially in the early stages of the disease. When modeling transition probabilities, the precision of clinical staging is imperfect, and the precision of time estimates is limited by the frequency of visits. Examining the entire continuum in one data set produces methodologically consistent estimates across the disease spectrum and is therefore a desirable approach to capturing transition probabilities in an A β + sample.

In this study, the NACC UDS sample was used to estimate transition probabilities between AD health states ranging across the disease continuum in individuals known to be

A β +, from asymptomatic to MCI-AD, mild AD dementia, moderate AD dementia, severe AD dementia, and death. This work directly informs the current understanding of progression along the AD continuum, and the estimates derived from this study can aid in trial design, care planning, and clinical and economic benefit assessments of AD interventions that reduce progression rates. The transition probabilities observed in this study, including a low reversion rate, should be further assessed in clinical settings. Furthermore, it will be important to test the consistency of these transition probabilities across countries and over time as the care pathway continues to develop.

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Data Availability. The data sets analyzed during the current study are not publicly available from the authors due to compliance with the NACC Data Use Agreement. The NACC data are publicly available from the NACC via their data request process: <https://naccdata.org/requesting-data/data-request-process>.

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