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Trajectories of care of community-dwelling people living with dementia: a multidimensional state sequence analysis

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Abstract

Background The type and level of healthcare services required to address the needs of persons living with dementia fluctuate over disease progression. Thus, their trajectories of care (the sequence of healthcare use over time) may vary significantly. We aimed to (1) propose a typology of trajectories of care among community-dwelling people living with dementia; (2) describe and compare their characteristics according to their respective trajectories; and (3) evaluate the association between trajectories membership, socioeconomic factors, and self-perceived health.

Methods This is an observational study using the data of the innovative Care Trajectories -Enriched Data (TorSaDE) cohort, a linkage between five waves of the Canadian Community Health Survey (CCHS), and health administrative data from the Quebec provincial health-insurance board. We analyzed data from 690 community-dwelling persons living with dementia who participated in at least one cycle of the CCHS (the date of the last CCHS completion is the index date). Trajectories of care were defined as sequences of healthcare use in the two years preceding the index date, using the following information: 1) Type of care units consulted (Hospitalization, Emergency department, Outpatient clinic, Primary care clinic); 2) Type of healthcare care professionals consulted (Geriatrician/psychiatrist/neurologist, Other specialists, Family physician).

Results Three distinct types of trajectories describe healthcare use in persons with dementia: 1) low healthcare use (n = 377; 54.6%); 2) high primary care use (n = 154; 22.3%); 3) high overall healthcare use (n = 159; 23.0%). Group 3 membership was associated with living in urban areas, a poorer perceived health status and higher comorbidity.

Conclusion Further understanding how subgroups of patients use healthcare services over time could help highlight fragility areas in the allocation of care resources and implement best practices, especially in the context of resource shortage.

Keywords Dementia, Alzheimer's disease, Care trajectory, State sequence analysis, Healthcare services

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Background

Dementia is considered a primary health concern, as individuals living with such conditions often experience severe and persistent disabilities, participation restrictions, and high formal and informal care needs. The World Health Organisation recognizes dementia as a public health priority [1]. Accordingly, many North American jurisdictions have adopted large-scale policies oriented towards improving health equity, access to healthcare, and quality of life for people living with dementia and their care partners [2, 3].

Yet, despite significant improvements in recent years, various barriers to optimal care remain. Indeed lack of care coordination, suboptimal transition processes, and poor quality of care are still reported [2, 4], which may lead to adverse outcomes, such as higher rates of avoidable hospitalization and emergency departments (ED) visits, as well as institutionalization [5, 6]. Indeed, people living with dementia experience diverse patterns of healthcare use.

Dementia is characterized by three stages (early, middle, and advanced stages), where the type and level of services required to address the needs of individuals may fluctuate significantly [7]. Throughout disease progression, these needs are further complexified by higher rates of multimorbidity and polypharmacy [8, 9]. Moreover, socioeconomic determinants further exacerbate these difficulties in accessing and navigating the healthcare system, as socioeconomic deprivation is associated with more ED visits and urgent healthcare needs and negatively impacts long-term survival [10].

Thus, service improvement requires a better understanding of the complex dynamics underlying care use in populations living with dementia. This could be gained by describing trajectories of care (ToCs), i.e., the sequence of healthcare use over time [11].

Some studies have advanced our knowledge of ToCs in people living with dementia [12]. For example, a study conducted in the United States by Rahman et al. [13] aimed to describe survival and ToCs among individuals from rural and urban areas. Their results suggest that, after diagnosis, individuals from rural areas spend more time in nursing homes and receive less home health care than their urban counterparts [13]. In Canada, Bronskill et al. [14] aimed to determine the long-term ToCs by a general group of individuals with dementia as they remain in the community over time. Amongst others, they observed higher rates of health service use, longterm care placement, and mortality among individuals living with dementia compared with individuals without the condition [14]. Building on these previous studies and to bridge remaining knowledge gaps, we sought to examine multidimensional ToCs (according to the healthcare settings and the professionals consulted) while accounting for population heterogeneity [12].

We used a state sequence analysis (SSA) approach and data from the Care Trajectories—Enriched Data (Tor-SaDE) cohort, a novel health administrative and survey data linkage. SSA is an innovative statistical method that enables the visualization of complex longitudinal patterns [15]. Its use in healthcare research has recently increased, and to our knowledge, this is the first study to apply SSA to examine ToC in people living with dementia [12]. We operationalized the "6W" multidimensional model of ToCs in six dimensions: patients, with their individual and clinical attributes (*who*), in responding to their illness condition and care needs (*why*) will seek healthcare services from different categories of care providers (*which*) in care units and settings (*where*) where they will receive tests and treatments (*what*) at specific times (*when*) [11].

Specifically, we aimed to (1) propose a multidimensional typology of ToCs among community-dwelling people living with dementia; (2) describe and compare the characteristics of people living with dementia according to their respective ToC; and (3) evaluate the association between ToC membership, socioeconomic factors, and self-perceived health.

Methods

Design and data sources

This is a retrospective cohort study in the province of Quebec, Canada. The study was approved by the Commission d'accès à l'information, l'Institut de statistique du Québec, and McGill's Research Ethics Board.

We gathered study data from the Care Trajectories-Enriched Data (TorSaDE) cohort (n = 102,148 distinct participants and available from 1996 to 2016), a linkage of data from two complementary sources: the Canadian Community Health Survey (CCHS) and health administrative data from the provincial health insurance boards (Régie de l'assurance maladie du Québec: RAMQ). This novel data linkage overcomes the perennial challenge that health administrative data tend to lack individuallevel socioeconomic characteristics and patient-reported outcomes, such as self-rated health. The full description of the cohort is available elsewhere [16]. Briefly, the CCHS is a cross-sectional survey providing in-depth self-reported information on health status, healthcare utilization, health indicators, and sociodemographic characteristics. Its sampling base represents about 97% of the Canadian population aged 12 years or older. The Tor-SaDE cohort includes all Quebec (Canada) respondents who participated in at least one of the five cycles (2007-08, 2009–10, 2011–12, 2013–14, 2015–16) of the CCHS and agreed to their data to be shared (sharing consent rate of about 96% for all CCHS cycles).

In Quebec, RAMQ provides universal health insurance coverage to all residents. Coverage includes services provided in ED, hospitals, and medical clinics, including primary health care clinics. The RAMQ administrative health register gives access to an extensive range of variables, including (1) patient demographic information (date of birth, date of death if applicable, postal code); (2) medical services register (data on medical services provided by Quebec fee-for-service physicians, including diagnoses coded according to the International Classification of Diseases 9 (ICD-9)); (3) pharmaceutical services (data on pharmacy-claimed drugs, for patients covered by the public drug insurance plan – about 85% of people 65 years old and over); (4) MED-ECHO registry (information on hospitalization, including diagnosis coded in ICD-10); (5) I-CLSC database (available since April 2012 and containing information about care services provided in local community service centers (CLSC)).

Studied population

The cohort was composed of all community-dwelling Quebec citizens who: (1) participated in at least one cycle of the CCHS — index date — (if an individual participated in more than one cycle, we kept the most recent); (2) were 65 years or older at the time of the CCHS completion; 3) had a dementia condition at the time of the CCHS completion. Patients permanently institutionalized in long-term care settings are excluded from the CCHS coverage and were thus not included in our study [16].

We used the RAMQ data to identify dementia cases using a validated algorithm (sensitivity of 79.3%, specificity of 99.1%) [17]. The definition of dementia used to develop the algorithm includes Alzheimer's disease, vascular dementia, dementia in other diseases classified elsewhere (frontotemporal dementia, idiopathic normal pressure hydrocephalus), and unspecified dementia (senile dementia, presenile dementia). The ICD-9 and 10 codes used are: ICD-9 (46.1, 290.0, 290.1, 290.2, 290.3, 290.4, 294.x, 331.0, 331.1, 331.5, 331.82); ICD-10 (F00.x, F01.x, F02.x, F03.x, G30.x). Individuals were considered as having dementia if they met a least one of the following criteria: (1) the presence of a primary or secondary diagnosis code for dementia during hospitalization (in MED-ECHO); (2) the presence of three dementia diagnosis codes registered at least 30 days apart in two years (in the medical services register); (3) a prescription claimed for an Alzheimer's disease-related drug in the pharmaceutical services register (cholinesterase inhibitor subclass, including donepezil, galantamine, rivastigmine, tacrine, and memantine). The date of dementia identification was when the first criteria became positive.

Variables

Trajectories of care

ToCs of our study population (Dimension *who*) were defined as sequences of healthcare utilization in time (dimension *when*) in the two years before the index date (date of the CCHS completion). Our ToCs were defined using the following information: 1) Dimension *where* — Care units consulted (1. Hospital, 2. ED, 3. Outpatient clinic, 4. Primary care clinic); 2) Dimension *which* — Healthcare care professionals consulted (1. Geriatrician/psychiatrist/neurologist, 2. Other specialists, 3. General practitioner). As the I-CLSC database was not available for our whole study period, we could not consider data relative to services provided in the community by nurses and other professionals in the ToCs development.

Demographics and clinical characteristics (dimension Who)

Variables identified in health administrative data Age sex, residential neighborhood characteristics (metropolitan area: \geq 100,000 inhabitants; small town: 10,000– 100,000 inhabitants; rural: <10,000 inhabitants) were considered at the index date. Second, health conditions were identified in the two years preceding the index date, based on two criteria: 1) two diagnosis codes for the same condition entered in the medical services register on two different dates or (2) one primary or secondary diagnosis code for the condition during hospitalization (MED-ECHO), if applicable (See Additional file 1). The comorbidity index selected is the Combined Comorbidity Index of Charlson and Elixhauser [18].

We calculated the monthly average number of specific claimed drugs two years before the index date. We computed if the individuals claimed a prescription (yes/no) of potentially inappropriate medications (benzodiazepine, anticholinergic, antipsychotic, and opioid) to bring complementary information. We estimated the maximum proportion of healthcare visits to the same provider in the two years preceding the index date to inform on care continuity in primary care settings.

Variables identified in the CCHS The following variables were identified at the index date and categorized as follows: household size (living alone or not), individual annual income (\leq 20,000\$, > 20,000\$), education level (\leq secondary, > post-secondary), perceived health status (poor/fair, good/very good/excellent), and having a regular physician (yes/no).

Statistical analysis

Using SSA terminology, a ToC corresponds to a sequence during which successive 'states' (value taken by a variable at a specific time) are encountered. The general principle of SSA is to compare sequences (i.e., individual ToCs of patients) with respect to the order in which their component states appear [19].

To propose a typology of ToCs of people living with dementia and to suit our multidimensional approach, we followed the six steps of a modified SSA approach [15], as proposed in the work of Vanasse et al. [20].

In step one, we developed ToC for people living with dementia two years before the CCHS completion (index date) and chose 'months' as the time unit. In step two, for each month over the two-year observation period and for all patients, we defined the dimension-specific states in priority order. Dimension where: 1) Hospital; 2) ED; 3) Outpatient clinic; 4) Primary care clinic; 5) No such healthcare services. Dimension which: 1) Geriatrician/psychiatrist/neurologist; 2) Other specialists; 3) General practitioner: 4) No such healthcare services. According to the dimension-specific states, and for each patient, we created a ToC sequence consisting of 24 states (for the 24 months preceding the index date). In step three and four, we used the dynamic hamming distance (DHD) alignment algorithm in TraMineR to derive a pooled distance matrix. [21] In step five, we used a hierarchical cluster analysis on the matrix with Ward's criterion to create homogeneous ToC types. The optimal number of ToC types was based on statistical criteria (dendrogram and inertia curve), parsimony, interpretability, and clinical judgment [22]. In step six, we used chronograms (also referred to as state distribution plots) to show a general distribution pattern of states at each time unit and a representation of the mean duration in each state.

We described the individual's characteristics by ToC types. The differences between subgroups were tested using ANOVA (for continuous variables) and chi-square tests (for categorical variables), with a significance level of 0.05.

Finally, we used a multinomial logistic regression model to test the association between group membership (ToC type), socioeconomic factors, and perceived health while accounting for comorbidity. The final model report odds ratios with the associated 95% confidence intervals. The two \times two matrices and the variance inflation factor showed no multicollinearity issues.

SSA was performed using the TraMineR package in R [23]. All other analyses were performed using SAS Enterprise Guide 8.3 (SAS Institute, Cary, NC).

Results

The study cohort included 690 community-dwelling individuals living with dementia. Our SSA approach revealed three groups, or ToC types, composed of individuals characterized by (1) *low healthcare use* (n = 377; 54.6%); (2) *high primary care use* (n = 154; 22.3%); and (3) *high overall healthcare use* (n = 159; 23.0%).

Figure 1 presents the chronogram for each time unit point (24 months) for each group and dimension (*where* and *which*), while Fig. 2 illustrates the mean number of days spent in each care setting for the *where* dimension.

Table 1 presents individual characteristics and shows that those characteristics differ according to group membership. Individuals in group 1 (low healthcare use) had the lowest proportion of overall healthcare use, with a mean of 6.7 primary-care visits over two years (SD = 16.0). They are notably characterized by being older (mean = 81.7; SD = 6.6), having a better perceived health status (n = 137; 36.4%), and having the lowest comorbidity index (mean = 3.3; SD = 3.0). Individuals in group 2 (high primary care use) had the most primary care visits over two years (mean = 14.2; SD = 14.2). This group mainly consisted of females (n=111; 72.1%) and individuals most likely to live in rural areas (n = 55; 35.7%). The proportion of all their health conditions and healthcare use lies between groups 1 and 3. Individuals from group 3 (high overall healthcare use) reported the highest healthcare use, with a higher mean number of days spent in hospitals (mean = 35.6; SD = 50.5), EDs (mean = 4.4; SD = 4.6),and outpatient clinics (mean = 14.9; SD = 31.2) during the two-year period. Individuals in this group were less likely to be female (n = 77; 48.4%), as well as being more likely to live in metropolitan areas (n = 99; 62.3%) and to report a poor/fair perceived health status (n = 94; 59.2%). They also presented with a higher comorbidity index (mean = 7.1; SD = 4.4), mean number of monthly medications (mean = 21.4; SD = 8.6), reported more common (n=61; 38.4%) and severe (n=36; 22.6%)mental health disorders, and experienced a higher death rate in the two years after the index date (n = 33; 20.8%).

Figure 3 presents the multinomial logistic regression results aiming to evaluate the association between ToC type membership, socioeconomic factors, and self-perceived health while accounting for comorbidity (with group 1 being the reference group). Group 3 membership was associated with younger age, living in urban areas, a poorer perceived health status, and a higher comorbidity index. A higher comorbidity index was the only variable significantly associated with group 2 membership.

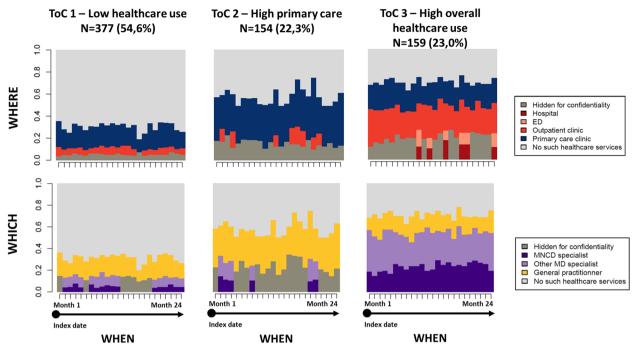


Fig. 1 State distribution plot of ToCs of persons living with dementia, by dimension.

The category 'hidden for confidentiality' groups together the categories not meeting the confidentiality data standards, due to an insufficient number of observations. Abbreviations: ToC, Trajectory of care; ED, Emergency department; MD, Doctor of Medicine

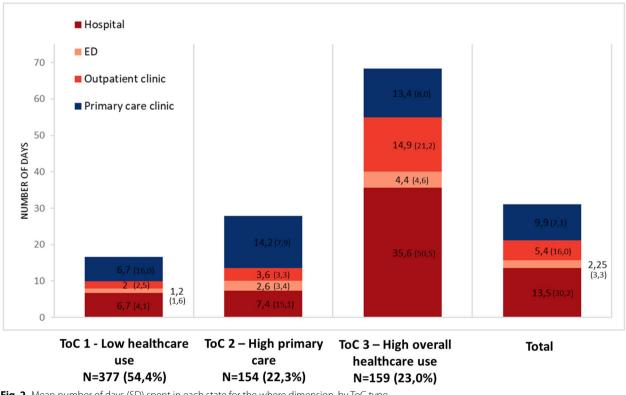


Fig. 2 Mean number of days (SD) spent in each state for the where dimension, by ToC type.

Abbreviations: ED, Emergency department; SD, Standard deviation; ToC, Trajectory of care

Variables	All individuals N=690	1—Low healthcare use <i>n</i> = 377 (54.6%)	2- High primary care use <i>n</i> = 154 (22.3%)	3- High overall healthcare use <i>n</i> = 159 (23.0%)	<i>p</i> -value ^a
Socioeconomic					
Age ^b	80.8 ± 6.7	81.7 ± 6.6	80.6 ± 6.6	79.2 ± 6.8	0.0004
Sex (female)	435 (63.0)	247 (65.5)	111 (72.1)	77 (48.4)	<.0001
Living alone	309 (44.8)	166 (44.0)	72 (46.8)	71 (44.7)	0.8484
Individual annual income more then 20,000\$	203 (39.3)	174 (63.3)	80 (52.0)	60 (48.8)	0.0060
Education level Post-secondary or higher	265 (40.2)	138 (38.4)	51 (34.7)	76 (49.4)	0.0215
Residential area					
Metropolitan	353 (51.2)	181 (48.0)	73 (47.4)	99 (62.3)	0.0177
Small town	142 (20.6)	90 (23.9)	26 (16.9)	26 (16.9)	
Rural	195 (28.3)	106 (28.1)	55 (35.7)	34 (21.4)	
Perceived health status (Poor/Fair)	294 (42.7)	137 (36.4)	63 (40.9)	94 (59.1)	<.0001
Health conditions					
Months since dementia identification ^b	36.9 ± 35.2	39.3 ± 34.8	35.6 ± 38.2	32.3 ± 32.6	0.1001
Dementia identified during the index period (yes)	182 (26.4)	90 (23.8)	45 (29.2)	47 (29.56)	0.1203
Comorbidity index ^b	3.7 ± 3.4	3.3 ± 3.0	4.4 ± 3.2	7.1 ± 4.4	<.0001
Comorbidity index					
0–3	353 (51.2)	246 (65.3)	72 (46.8)	35 (22.0)	<.0001
4 and over	337 (48.8)	131 (34.8)	82 (53.3)	124 (78.0)	
High blood pressure	304 (44.1)	120 (31.8)	78 (50.7)	106 (66.7)	<.0001
COPD	120 (17.4)	45 (12.0)	29 (18.8)	46 (28.9)	<.0001
Diabetes	138 (20.0)	58 (15.4)	31 (20.1)	49 (30.8)	0.0002
Heart disease	225 (32.61)	74 (19.6)	54 (35.1)	97 (61.0)	<.0001
Cancer	40 (5.8)	Less than 15	Less than 15	25 (15.7)	<.0001
Common mental-health disorder	155 (22.5)	50 (13.3)	44 (28.6)	61 (38.4)	<.0001
Severe mental-health disorder	76 (11.0)	21 (5.6)	19 (12.3)	36 (22.6)	<.0001
Healthcare use					
Regular primary care provider	508 (95.9)	269 (94.4)	116 (95.9)	123 (99.2)	0.0813
Primary care provider continuity of care ^b	0.90 ± 0.18	0.914 ± 0.164	0.89 ± 0.19	0.869±0.201	0.0309
Death within 2 years after index date	80 (11.6)	34 (9.0)	Less than 15	33 (20.8)	0.0002
Long-term care admission within 2 years after the index date	92 (13.33)	53 (14.06)	20 (12.99)	19 (11.95)	0.7982
Medication					
Average number of monthly medications ^b	12.1 ± 6.7	11.8±5.8	16.0 ± 7.46	21.4±8.6	<.0001
Benzodiazepine	276 (42.1)	127 (35.8)	70 (46.7)	79 (52.7)	0.0009
Anticholinergic	478 (73.0)	231 (65.1)	111 (74.0)	136 (90.7)	<.0001
Antipsychotic	145 (22.1)	7 (21.7)	31 (20.7)	37 (24.7)	0.6750
Opioid	115 (17.6)	38 (10.7)	28 (18.7)	49 (32.7)	<.0001

Table 1 Characteristics of individuals, according to their respective ToC type

Abbreviations: COPD chronic obstructive pulmonary disease, SD standard deviation

^a Differences between ToC type were tested using ANOVA (continuous variables) and Chi-square test (categorical variables)

 $^{\rm b}\,$ Continuous variable reported using mean $\pm\,$ SD, all other variables are reported using n(%)

Discussion

We aimed to identify the ToCs of people living with dementia in the community and to evaluate the association of socioeconomic factors and self-perceived health with ToC membership. Three groups emerged: (1) *low*

healthcare use; (2) high primary care use; and (3) high overall healthcare use. The results from our regression model show that lower perceived health and specific socioeconomic factors (younger age, living in urban areas) were associated with high overall healthcare use (relatively to low healthcare use). A higher comorbidity

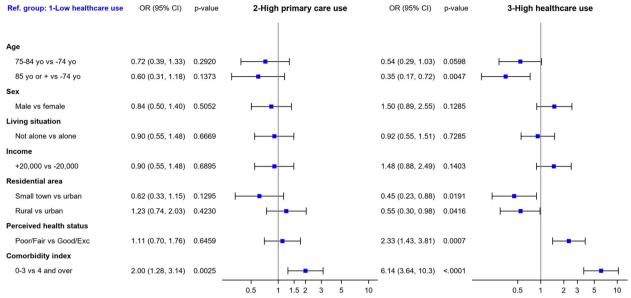


Fig. 3 Association between group membership, socioeconomic factors, and self-perceived health.

Abbreviations: OR, Odd ratio; CI, Confidence interval; Yo, Years old; Exc, Excellent

index was also associated with membership in groups 2 and 3.

These results contribute to an emerging literature on ToCs for persons living with dementia [12]. Nevertheless, comparing our results with existing studies is challenging. Indeed, their approach often encompasses a global population or is oriented toward other services such as long-term care placement or end-of-life years [14, 24, 25].

Our results suggest a significant variation in healthcare use within our population and are supported by scientific literature. First, the characteristics of group 1 (low healthcare use) could indicate that most individuals could manage their health conditions with minimal healthcare services and visits to general practitioners. Indeed, this group had few primary care visits to these professionals, while the yearly rate for the overall population with dementia could be around 7 to 10 visits to general practitioners [26]. Their lower comorbidity rates could explain this limited healthcare use, as higher comorbidity is associated with more healthcare use and costs [27, 28]. The patterns of healthcare use in this group allow us to hypothesize that this group gathers the less severe dementia cases, as functional and cognitive deficits are also associated with higher rates of healthcare use, including hospitalizations [27, 29].

In parallel, group 2 (*high primary care use*) comprised individuals with higher comorbidity and the highest number of visits to primary care clinics. Studies have shown that those living with dementia present higher patterns of primary care use. This is especially true around critical periods when healthcare needs could be more significant, such as diagnosis and end-of-life years [28, 30]. Thus, the rate of consultations with cognition specialists in this group may indicate higher severity of dementia (lying between groups 1 and 3), mainly managed in primary care settings. In Canada, guidelines also recommend that the treatment and care of patients with dementia be placed under the responsibility of family medicine groups (FMGs), with the collaboration of secondary and tertiary care for complex cases [31].

Group 3 (*high overall healthcare use*) comprises individuals with the highest comorbidity and overall use of healthcare services. Individuals with dementia and high comorbidity tend to have higher polypharmacy and healthcare use [32]. This group's higher rates of severe mental health conditions could further complicate dementia management. Indeed, persons living with dementia and presenting a psychiatric comorbidity are more likely to use medical inpatient services than those without psychiatric comorbidities [33].

Our second objective was to evaluate the association between ToC membership, socioeconomic factors, and self-perceived health while accounting for comorbidity. Mainly, the result shows that living in urban areas, lower perceived health status is associated with group 3 membership, while higher comorbidity is associated with membership to groups 2 and 3. This last result is not surprising, as comorbidity is associated with higher healthcare use in the population with and without dementia [34].

Regarding socioeconomic factors, studies showed that individuals living with dementia in urban areas have more ED visits and hospital use than their rural and micropolitan counterparts. This might be related to service availability and a regional propensity to seek such types of care [13, 35]. These patterns of healthcare use, however, put the individuals at higher risk of adverse effects, notably as longer lengths of stay in EDs and hospital environments predispose patients living with dementia to more hospital harm (including delirium and higher mortality) [35]. Lastly, membership in group 3 was associated with lower self-perceived health. Self-rated health is widely used for assessing health in older adults. It is negatively influenced by comorbidity, polypharmacy, poor mental health, functional outcomes, and healthcare use [36, 37], which is consistent with our results.

Implications for practice and research

Persons living with dementia present with a range of medical and cognitive needs, underscoring that models of care ought to incorporate this medical complexity to improve the management of the condition [38].

For example, the high healthcare use characterizing individuals from group 3 shows they could benefit from intensive interventions to ensure coordination and communication between the various services they already use. Innovative approaches to care delivery, including a dedicated navigator to support the patient throughout the care trajectory, could have a positive impact [39, 40], as the number of transitions tends to decrease with better use of services and support [41].

Polypharmacy was present in all our groups, including group 1, with less healthcare use and low comorbidity. This calls for educating healthcare professionals on medication management, such as with the help of initiatives such as *Choosing Wisely*, which advocates for avoiding unnecessary medical treatments [42]. For complex individuals with a higher level of polypharmacy, such as in group 3, an interdisciplinary approach engaging pharmacists could also be considered. In primary care teams, medication reviews led by a pharmacist can minimize the adverse effects of polypharmacy and improve medication management, including in vulnerable geriatric populations [43, 44].

The use of a ToC approach is also congruent with a care pathway management approach, which describes the optimal course and service to be provided to the right person, in the proper sequence, at the right time, in the right place, and with the best results while focusing on the user's experience [45]. The research on ToCs is of significant value to care planning and organization. It

illuminates how real-life ToCs differ from the ideal ones and how they vary concerning inequities [46]. For example, care trajectories can help contextualize the timing and nature of different care transitions and identify the types of services an individual may require based on their characteristics.

However, formal healthcare use may not reflect actual healthcare needs of people living with dementia. To fully understand an individual's care trajectory, one must consider their use of other healthcare services, such as home care and informal care. In any three ToC, individuals may present unmet needs in many spheres, including safety, daily activities, and managing cognitive symptoms [47]. Considering the trajectory of cognitive decline is also crucial for services planning, as it may significantly impact healthcare use and costs throughout the disease [48].

For future research, the development of ToCs should be disaggregated within specific geographic areas to inform service planning for individuals living with dementia. Our groups could also be further described and understood regarding key aspects (including the severity of dementia, frailty index, intensity, and type of home-care services used) and moments (i.e., end-of-life years, time of diagnosis). Future research incorporating an overall measure of deprivation could also give insight into how ToCs and needs vary within specific geographical areas.

Strength and limitations

First, our study uses an innovative cohort in Quebec (the TorSaDe cohort), which links survey and health administrative data. We identified ToCs using a robust SSA approach, which allows the development of homogeneous ToC types to better capture patterns of healthcare use within a complex population. Our multidimensional approach allowed the inclusion of nine states, partitioned into two distinct sequence dimensions, facilitating interpretation and visualization [49, 50].

However, data limitations prevented us from considering important variables related to patient clinical attributes (e.g., dementia severity). Also, our analysis was limited by the exclusion of nursing and communitybased services. Specifically, individuals using lower levels of formal healthcare services, such as those in ToC 1, may still present a high level of healthcare use that was not captured in our analysis.

Furthermore, the choice of our index period (two years before the index date of the CCHS) limited the inclusion of people with more severe dementia (as about 26.4% were diagnosed during the index period), in end-of-life situations or near death and might have introduced a bias in favor of healthier individuals (possibly compounding the bias inherent with considering community-dwelling persons with dementia, as access to long-term care facilities in Quebec is scarce and limited to the most severe cases). This choice was consecutive to sample size and the respect of data confidentiality standards. Lastly, we could not consider immigration status in our regression model for confidentiality purposes related to low-frequency cell counts. To overcome these limitations, an updated database could provide access to a more extensive sample. A mixed methods approach could also be employed to consider patients' clinical attributes. This approach would involve combining qualitative and quantitative methods to capture a more comprehensive understanding of the complex interactions between patient characteristics and healthcare utilization.

Conclusions

This is the first study to propose a comprehension assessment of shared patterns of healthcare use among persons living with dementia in the community, which is a complex and heterogeneous population. Understanding how subgroups of patients use healthcare services over time helps highlight areas of fragility in allocating geriatriccare resources to implement best practices and in the context of resource shortage.

Abbreviations

CCHS	Canadian Community Health Survey
DHD	Dynamic hamming distance
ED	Emergency department
ICD	International Classification of Diseases
RAMQ	Régie de l'assurance maladie du Québec
SSA	State sequence analysis
TorSaDE	Care Trajectories—Enriched Data
ToCs	Trajectories of care

Supplementary Information

The online version contains supplementary material available at https://doi.org/10.1186/s12877-023-03926-x.

Additional file 1. International Classification of Diseases (ICD) codes used for the measures of chronic conditions.

Additional file 2. Past and current members of the TORSADE Cohort Working Group.

Authors' information

NA

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Authors' contributions

All authors have contributed to developing the design and method. All authors made substantial contributions to the article conception and design. I. Dufour was involved in analysis and interpretation of data. All authors revised the article for intellectual content and gave their approval for the final version.

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Availability of data and materials

The data from the TorSaDE cohort are not publicly available and are under strict security and confidentiality regulations. Please contact the corresponding author for more information.

Declarations

Ethics approval and consent to participate

This study was performed according to the Helsinki Declaration, and informed consent was obtained from all individuals participants included in the TORSADE cohort. The study received ethics approval through the Commission d'accès à l'information, l'Institut de statistique du Québec, and McGill's Research Ethics Board.

Consent for publication

Not applicable.

Competing interests

The authors declare no competing interests.

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References

- World Health Organization [WHO]. Global action plan on the public health response to dementia 2017–2025. 2017. https://apps.who.int/iris/ rest/bitstreams/1092215/retrieve. Accessed on 15 Jul 2022.
- Agence de la santé publique du Canada. Une stratégie sur la démence pour la Canada: Ensemble, nous y aspirons. 2019. https://www.canada. ca/content/dam/phac-aspc/documents/services/publications/diseasesconditions/dementia-highlights-canadian-chronic-disease-surveillance/ dementia-highlights-canadian-chronic-disease-surveillance. Accessed on 15 Jul 2022.
- U.S. Department of Health and Human Services. National Plan to AddressAlzheimer's Disease.2020. https://aspe.hhs.gov/sites/default/files/ migrated_legacy_files//197726/NatlPlan2020, Accessed on 15 Jul 2022.
- Bergman H, Ministère de la santé et des services sociaux. Relever le défi de la maladie d'Alzheimer et des maladies apparentées : Une vision centrée sur la personne, l'humanisme et l'excellence. Gouvernement du

Québec, 2009. https://publications.msss.gouv.qc.ca/msss/fichiers/2009/ 09-829-01W.pdf. Accessed on 15 Jul 2022.

- Godard-Sebillotte C, et al. Avoidable hospitalizations in persons with dementia: a population-wide descriptive study (2000–2015). Can Geriatr J. 2021;24(3):209–21.
- Prince M, Comas-Herrera A, Knapp M, Guerchet M, Karagiannidou M. Improving healthcare for people living with dementia: coverage, quality and costs now and in the future. Alzheimer's Disease International. 2016. https://www.alz.co.uk/research/worldreport-2016. Accessed 15 Jul 2022.
- Alzheimer's association. Alzheimer's and dementia: Stages of Alzheimer's. 2023. https://www.alz.org/alzheimers-dementia/stages. Accessed 16 Feb 2023.
- Amjad H, Carmichael D, Austin AM, Chang C-H, Bynum JPW. Continuity of care and health care utilization in older adults with dementia in fee-forservice medicare. JAMA Intern Med. 2016;176(9):1371–8.
- Griffith LE, Gruneir A, Fisher K, et al. Patterns of health service use in community living older adults with dementia and comorbid conditions: a population-based retrospective cohort study in Ontario. Canada BMC Geriatr. 2016;16(1):177–177.
- Watson J, Giebel C, Green M, Darlington-Pollock F, Akpan A. Use of routine and cohort data globally in exploring dementia care pathways and inequalities: a systematic review. Int J Geriatr Psychiatry. 2021;36(2):252–70.
- Vanasse A, Courteau M, Ethier JF. The "6W" multidimensional model of care trajectories for patients with chronic ambulatory care sensitive conditions and hospital readmissions. Public Health. 2018;157:53–61.
- Dufour I, Bolster-Foucault C, Vedel I, Quesnel-Vallée A. Examining the concept of healthcare trajectories in older adults with major neurocognitive disorder using the '6w' multidimensional model of care trajectories: a mixed methods systematic review protocol. Sci Nurs Health Pract. 2021;4(2):1–110.
- Rahman M, White EM, Thomas KS, Jutkowitz E. Assessment of rural-urban differences in health care use and survival among medicare beneficiaries with alzheimer disease and related dementia. JAMA netw open. 2020;3(10): e2022111.
- 14. Bronskill SE, Maclagan LC, Walker JD, et al. Trajectories of health system use and survival for community-dwelling persons with dementia: a cohort study. BMJ Open. 2020;10(7): e037485.
- Robette N. Explorer et décrire les parcours de vie: les typologies de trajectoires: CEPED; 2011. p. 86. (halshs-01016125). https://halshs.archi ves-ouvertes.fr/halshs-01016125. Accessed 28 Jan 2022.
- Vanasse A, Chiu YM, Courteau J, et al. Cohort profile: the care trajectoriesenriched data (TorSaDE) cohort. Int J Epidemiol. 2020. https://doi.org/10. 1093/ije/dyaa167.
- Jaakkimainen RL, Bronskill SE, Tierney MC, et al. Identification of physician-diagnosed alzheimer's disease and related dementias in populationbased administrative data: a validation study using family physicians' electronic medical records. J Alzheimers Dis. 2016;54(1):337–49.
- Simard M, Sirois C, Candas B. Validation of the combined comorbidity index of Charlson and Elixhauser to predict 30-day mortality across ICD-9 and ICD-10. Med Care. 2018;56(5):441–7.
- Vanoli J, Nava CR, Airoldi C, Ucciero A, Salvi V, Barone-Adesi F. Use of State Sequence Analysis in Pharmacoepidemiology: A Tutorial. Int J Environ Res Public Health. 2021;18(24):13398.
- Vanasse A, Courteau J, Courteau M, et al. Healthcare utilization after a first hospitalization for COPD: a new approach of State Sequence Analysis based on the "6W" multidimensional model of care trajectories. BMC Health Serv Res. 2020;20(1):177.
- Gabadinho A, Ritschard, G., Müller, NS., Studer, M. Analyzing and Visualizing State Sequences in R with TraMineR. 2011; http://www.jstatsoft.org/ v40/i04/. Accessed 28 Jan 2022.
- Nylund KL, Asparouhov T, Muthén BO. Deciding on the number of classes in latent class analysis and growth mixture modeling: A Monte Carlo simulation study. Struct Equ Modeling. 2007;14(4):535–69.
- 23. R Development Core Team. R: A language and environment for statistical computing. Vienna R Foundation for Statistical Computing; 2011.
- Aaltonen M, Forma L, Pulkki J, Raitanen J, Rissanen P, Jylha M. Changes in older people's care profiles during the last 2 years of life, 1996–1998 and 2011–2013: a retrospective nationwide study in Finland. BMJ Open. 2017;7(11): e015130.

- Karmel R, Gibson D, Anderson P, Wells Y, Duckett S. Care trajectories through community and residential aged care services: disease effects. Ageing Soc. 2012;32(8):1428–45.
- Weber SR, Pirraglia PA, Kunik ME. Use of services by community-dwelling patients with dementia: a systematic review. Am J Alzheimers Dis Other Demen. 2011;26(3):195–204.
- Zhu CW, Cosentino S, Ornstein K, Gu Y, Andrews H, Stern Y. Use and cost of hospitalization in dementia: longitudinal results from a communitybased study. Int J Geriatr Psychiatry. 2015;30(8):833–41.
- Eisele M, van den Bussche H, Koller D, et al. Utilization patterns of ambulatory medical care before and after the diagnosis of dementia in Germany – results of a case-control study. Dement Geriatr Cogn Disord. 2010;29(6):475–83.
- Zhu CW, Cosentino S, Ornstein KA, Gu Y, Andrews H, Stern Y. Interactive effects of dementia severity and comorbidities on medicare expenditures. J Alzheimers Dis. 2017;57(1):305–15.
- Schwarzkopf L, Hao Y, Holle R, Graessel E. Health care service utilization of dementia patients before and after institutionalization: a claims data analysis. Dement Geriatr Cogn Dis Extra. 2014;4(2):195–208.
- 31 Ismail Z, Black SE, Camicioli R, et al. Recommendations of the Canadian consensus conference on the diagnosis and treatment of dementia. Alzheimer's Dementia. 2020;16(8):1182.
- Kaczynski A, Michalowsky B, Eichler T, et al. Comorbidity in dementia diseases and associated health care resources utilization and cost. J Alzheimers Dis. 2019;68(2):635–46.
- Kunik ME, Snow AL, Molinari VA, et al. Health care utilization in dementia patients with psychiatric comorbidity. Gerontologist. 2003;43(1):86–91.
- Dufour I, Dubuc N, Chouinard M-C, Chiu Y, Courteau J, Hudon C. Profiles of frequent geriatric users of emergency departments: a latent class analysis. J Am Geriatr Soc. 2021;69(3):753–61.
- 35. Canadian Institute for Health Information. Dementia in Canada: Digital report delivers CIHI's first comprehensive look at this complex illness and its effects on seniors, caregivers and health systems. 2018. https://www. cihi.ca/en/dementia-in-canada. Accessed on 15 Jul 2022.
- Machón M, Vergara I, Dorronsoro M, Vrotsou K, Larrañaga I. Self-perceived health in functionally independent older people: associated factors. BMC Geriatr. 2016;16:66.
- 37. Nelis SM, Wu Y-T, Matthews FE, et al. The impact of co-morbidity on the quality of life of people with dementia: findings from the IDEAL study. Age Ageing. 2019;48(3):361–7.
- Schubert CC, Boustani M, Callahan CM, et al. Comorbidity profile of dementia patients in primary care: are they sicker? J Am Geriatr Soc. 2006;54(1):104–9.
- Dohan D, Schrag D. Using navigators to improve care of underserved patients: current practices and approaches. Cancer. 2005;104(4):848–55.
- Bernstein A, Harrison KL, Dulaney S, et al. The Role of Care Navigators Working with People with Dementia and Their Caregivers. J Alzheimers Dis. 2019;71(1):45–55.
- Parker KJ, Hickman LD, Phillips JL, Ferguson C. Interventions to optimise transitional care coordination for older people living with dementia and concomitant multimorbidity and their caregivers: A systematic review. Contemp Nurse. 2020;56:1–29.
- 42. Born KT, Huynh W. Levinson. Reflecting on choosing wisely Canada at five years: accomplishments, challenges and opportunities for reducing overuse and improving quality. Healthcare Papers. 2019;18(1):9–17.
- 43. Khera S, Abbasi M, Dabravolskaj J, Sadowski CA, Yua H, Chevalier B. Appropriateness of medications in older adults living with frailty: impact of a pharmacist-led structured medication review process in primary care. J Prim Care Community Health. 2019;10:1–8.
- 44. Nguyen TA, Gilmartin-Thomas J, Tan ECK, et al. The impact of pharmacist interventions on quality use of medicines, quality of life, and health outcomes in people with dementia and/or cognitive impairment: A systematic review. J Alzheimers Dis. 2019;71(1):83–96.
- 45. Thiebaut G-C, Lavoie C, Labrecque-Pégora-ro S, Maillet L. Pour des trajectoires de soins et services performantes : Un outillage pour l'analyse et la mesure. 2020. https://www.iuplsss.ca/clients/SanteEstrie/Sous-sites/Centr es_de_recherche/IUPLSSS/transfert_connaissances/outils-publications/ Pour_des_trajectoires_de_soins_et_de_services_performantes_Un_ outillage_pour_l%E2%80%99analyse_et_la_mesure.pdf. Accessed on 15 Jul 2022.

- 46 Pinaire J, Azé J, Bringay S, Landais P. Patient healthcare trajectory. An essential monitoring tool: a systematic review. Health Inf Sci Syst. 2017;5(1):1–1.
- Black BS, et al. Unmet needs in community-living persons with dementia are common, often non-medical and related to patient and caregiver characteristics. Int Psychogeriatr. 2019;31(11):1643–54.
- Evenden D, et al. Computer simulation of dementia care demand heterogeneity using hybrid simulation methods: improving populationlevel modelling with individual patient decline trajectories. Public Health. 2020;186:197–203.
- 49. Fasang AE, Liao TF. Visualizing sequences in the social sciences: relative frequency sequence plots. Sociol Methods Res. 2014;43(4):643–76.
- Piccarreta R, Lior O. Exploring sequences: a graphical tool based on multidimensional scaling. J R Stat Soc Ser A Stat Soc. 2010;173(1):165–84.

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