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Findings from a social media listening study on Chinese patients with Alzheimer's disease: a content analysis

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ABSTRACT

Background Social media listening is a new approach for gathering insights from social media platforms about users' experiences. This approach has not been applied to analyse discussions about Alzheimer's disease (AD) in China.

Aims We aimed to leverage multisource Chinese data to gain deeper insights into the current state of the daily management of Chinese patients with AD and the burdens faced by their caregivers.

Methods We searched nine mainstream public online platforms in China from September 2010 to March 2024. Natural language processing tools were used to identify patients and caregivers, and categorise patients by disease stage for further analysis. We analysed the current state of patient daily management, including diagnosis and treatment, choice of treatment scenarios, patient safety and caregiver concerns.

Results A total of 1211 patients with AD (66% female, 82% aged 60-90) and 756 caregivers for patients with AD were identified from 107 556 online sources. Most patients were derived from online consultation platforms (43%), followed by bulletin board system platforms (24%). Among the patients categorised into specific disease stages (n=382), 42% were in the moderate stage. The most frequent diagnostic tools included medical history (97%) and symptoms (84%). Treatment options for patients with AD primarily included cholinesterase inhibitors, N-methyl-D-aspartate receptor antagonists and antipsychotics. Both quantitative and qualitative analysis of patients who experienced wandering (n=92) indicated a higher incidence of wandering during the moderate stage of the disease. Most caregivers were family members, with their primary concerns focusing on disease management and treatment (90%), followed by daily life care (37%) and psychosocial support (25%).

Conclusions Online platform data provide a broad spectrum of real-world insights into individuals affected by AD in China. This study enhances our understanding of the experiences of patients with AD and their caregivers, providing guidance for developing personalised interventions, providing advice for caregivers and improving care for patients with AD.

WHAT IS ALREADY KNOWN ON THIS TOPIC

International social media data have been used to generate insights into users' experiences and to analyse discussions about Alzheimer's disease (AD); however, there is a lack of analysis on Chinese websites.

WHAT THIS STUDY ADDS

Our research aimed to leverage multisource Chinese data to gain deeper insights into the current state of the daily management of Chinese patients with AD and the burdens faced by their caregivers.

HOW THIS STUDY MIGHT AFFECT RESEARCH, PRACTICE OR POLICY

- This research innovatively used online data to explore the experiences and challenges faced by patients with AD and their caregivers.
- ⇒ The findings resonated with previous studies on diagnostic methods and treatment choices, addressing a research gap by quantifying incidents of wandering among patients with AD.

INTRODUCTION

From 1980 to 2020, the population of individuals aged 60 years and above in China surged, rising from 6.9% to 18.7% of adults. This profound ageing of China's population has been paralleled by a persistent and alarming increase in the incidence of Alzheimer's disease (AD). The China AD Report states that in 2019, 13.1 million people in China were affected by AD and related dementia, with a higher prevalence and mortality rate than the global averages, highlighting the severity of the issue. However, AD prevention and management in China face numerous challenges. First, patients with AD experience suboptimal diagnostic and treatment rates, with a particularly alarming underdiagnosis rate exceeding 90% in rural areas.³ Furthermore, a large proportion of individuals with dementia in China remain untreated with pharmacological interventions. ⁴ Additionally,

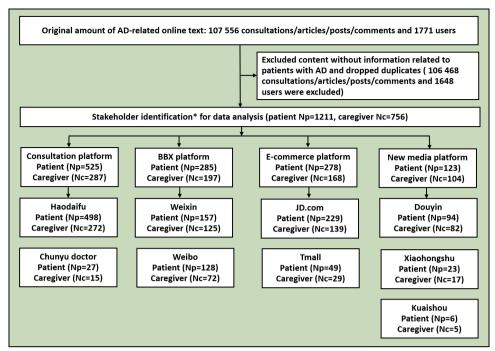


Figure 1 Flowchart of stakeholder identification and platform type from which online text was extracted. *Patient information was identified first, and caregiver information was then extracted from the patient-related text based on whether the caregiver perspective exists. AD, Alzheimer's disease; BBS, bulletin board system.

current treatments are insufficient to achieve satisfactory clinical outcomes or halt disease progression. Access to professional care is limited, and most patients with AD rely on informal home-based care.⁵ The lack of effective treatments and inadequate care systems place a significant financial and psychological burden on caregivers, affecting their well-being and the quality of care they provide.⁶

Social media listening (SML) is a new approach to harnessing information derived from social media platforms to generate insights into users' experiences. In addressing AD prevention and management challenges in China, SML has emerged as a promising tool, leveraging web data to aid dementia care by generating and analysing vast datasets, and offering real-time patient perspectives.⁸ SML streamlines the time-consuming process of patient recruitment, accelerates research and minimises recall biases. ⁹ By capturing diverse experiences not typically reflected in clinical trials or patient preference studies, SML ensures a more comprehensive understanding of dementia care. 10 This unbiased assessment of patient perspectives complements traditional research methods, fostering effective, patient-centred strategies. While previous studies have focused on patients with AD on international platforms like YouTube, Twitter and Facebook, these are less accessible in China. 11 12 However, the widespread access to the internet and mobile applications allows patients in China to easily obtain medical advice and consultations, regardless of their geographical location. As the integration of blockchain technology and artificial intelligence progresses, natural language processing (NLP) can classify the mood and tone in text

to help identify and alleviate psychological stress experienced by patients with AD, enhance their mood and reduce loneliness, offering real-world insights into their burdens. ¹²

To date, existing literature on SML for Chinese patients with AD is sparse. This lack of research represents a significant gap in our understanding of this vulnerable population, as their voices and perspectives are crucial for developing effective healthcare strategies. By analysing the experiences of Chinese patients with AD and their caregivers, we aimed to gain insights into diagnosis, treatment and management features including caregiver experiences and issues such as wandering, to inform AD market development. We also seek to use advanced NLP technology, such as large language models (LLMs), to understand the feelings and unmet needs of Chinese patients with AD.

METHODS

Data source and collection

We collected multisource data from various major online platforms in China, including online consultation platforms (Haodaifu, Chunyu Doctor), e-commerce platforms (Tmall, JD.com), bulletin board system (BBS) platforms (Weibo, WeChat) and social media platforms (Douyin, Kuaishou, Xiaohongshu) (figure 1). We used AD-related keywords, such as "Alzheimer" (阿尔茨海默), "AD", "Alzheimer Disease", "Senile dementia" (老年痴呆), "dementia" (认知症), "cognitive impairment" (认知障碍), "NMDA receptor antagonists" (NMDA 受体拮抗剂), "glutamate receptor antagonists" (谷氨酸受



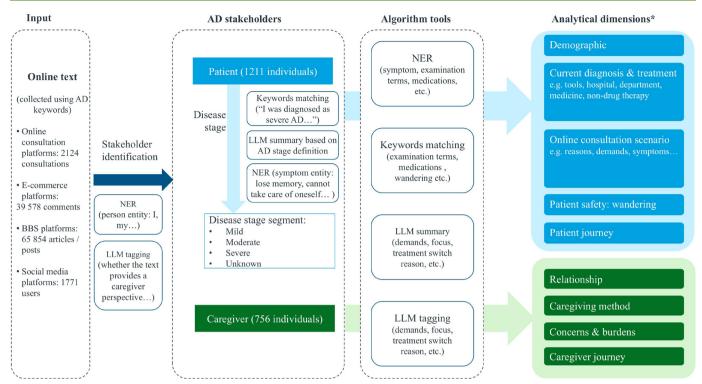


Figure 2 Overall workflow and algorithm tools used in the study. *Experts will conduct manual checks on the analytical results. AD, Alzheimer's disease; BBS, bulletin board system; NER, named-entity recognition; LLMs, large language models.

体拮抗剂), "cholinesterase inhibitors" (胆碱酯酶抑制剂), to collect information. Data collection spans from September 2010 to March 2024, with adjustments made for the platforms' available history. Considering some platforms became publicly accessible from 2016 (BBS) or 2017 (e-commerce), and some social media sites emerged in recent years, we collected all historical data from these newer social media sites.

Ethical considerations

All data used and presented in this study were obtained from publicly accessible sources without accessing password-protected information. Patients' privacy was respected, and online content was anonymised in compliance with data privacy obligations. No individual patient data requiring consent has been presented.

Data extraction and analysis

Given the colloquial language used by patients, traditional methods are challenging. Therefore, we used various NLP tools to process and analyse text (figure 2). Named-entity recognition (NER) is a subtask of NLP that involves identifying and classifying named entities in text, such as names of people, organisations and locations. In this study, we used the 'knowledge mining' interface from PaddleNLP (https://github.com/PaddlePaddle/PaddleNLP) to find self-defined types of entities from text. For example, NER can identify and categorise symptoms, medication names, dosages and patient outcomes. It segments sentences into words and determines their types, allowing us to select necessary labels for each analytical dimension to examine the original content mentioned

by patients, such as identifying 'AD' under the 'disease category' or 'donepezil' under the 'drug category'. In this study, we extracted entities including people, roles, symptoms, medicine and other related items from the text. Meanwhile, we used pkuseg (https://github.com/ lancopku/pkuseg-python), a commonly used tokeniser for Chinese, for multidomain word segmentation. LLMs are computational models known for their capacity in general-purpose language generation and NLP tasks. 13 In this study, we used generative pretrained transformer 4 (GPT-4) (https://openai.com/gpt-4), a renowned LLM developed by OpenAI for two main tasks: (1) summarising language from different stakeholders to understand their requirements and focus during diagnosis and treatment, diagnostic tools, reasons for online consultation, etc and (2) deriving tagging systems based on the content, and then applying the labels to categorise stakeholders' requirements, focus, triggers for treatment switch, and so on. When applying GPT-4, prompt engineering was used to optimise results. 14 Tactics included providing accurate background information of the disease, and one-shot/ few-shot prompting. Here is an example of a drug treatment prompt: 'Identify AD-related drug treatments. Pay attention to detail and make independent judgements. Summarise any AD drug treatments mentioned. If none are mentioned, return: no relevant information.' This demonstrates how we use LLM's summarisation and extraction capabilities to achieve our goals.

Stakeholder identification

Considering platform diversity, we employed tailored methods to identify patients. The methodology is shown



Demographics	Number of patients, n (%)
Gender (n=884)	
Female	582 (66)
Male	302 (34)
Province* (n=363)	
Zhejiang	35 (10)
Guangdong	33 (9)
Beijing	31 (9)
Shandong	26 (7)
Shanghai	23 (6)
Jiangsu	22 (6)
Hebei	20 (6)
Hubei	20 (6)
Hunan	18 (5)
Sichuan	15 (4)
Patient's age (years)† (n=528)	
<50	6 (1)
[50, 60)	44 (8)
[60, 70)	124 (23)
[70, 80)	153 (29)
[80, 90)	156 (30)
[90, 100)	44 (8)
≥100	1 (0)
Disease stage (n=382)	
Mild	90 (24)
Moderate	162 (42)
Severe	130 (34)
Duration of AD (n=489)	
<1 year	79 (16)
1 year≤n<5 years	171 (35)
5 years≤n<10 years	107 (22)
10 years≤n<20 years	50 (10)
≥20 years	2 (0)
Unknown‡	80 (16)

^{*}A total of 363 patients provided self-reported geographical information; provinces with the top 10 patient count are presented.

in online supplemental figure 1. For online consultation platforms, we used two steps to check whether a consultation was an AD-related case. First, if the consultation contained keywords like "diagnosed", "diagnosis", "suffering from" or "affected by" followed by AD-related terms, it was deemed a clear AD diagnosis. Additionally, if AD-related keywords were followed by 'X years', it indicated a history of AD. Second, we extracted entities including diseases, symptoms, examination terms and

medications. For AD-related entities, GPT-4 was used to confirm whether it was an AD case. For e-commerce platforms, we filtered AD-related product comments by searching for AD-related keywords or identifying comments that contained AD-related keywords. Patients were then defined by deduplicating the comments based on username, product, comment and location (if available). For BBS and social media platforms, if person entities and AD-related keywords existed in the context (for BBS) or username and introduction (for social media), the article/post/user would be included in the patient pool. Once the patient pool was established, LLM was used to identify whether a caregiver's perspective was present. Therefore, the expression of patient information in this paper includes both the patient's own expression and the caregiver's paraphrasing. To validate the above methodology, a dedicated team of trained experts meticulously reviewed the patient pool. In the case of discrepancies, the expert assessment was prioritised.

Patient segmentation

To understand patients' perspectives at different disease stages, we categorised them into three groups: mild, moderate and severe, based on the severity of clinical symptoms affecting daily life, 15 using a three-step process (online supplemental figure 2). First, we extracted selfreported descriptions. For example, if the content stated that the patient was diagnosed with severe AD, they would be categorised into the severe group. Second, for those who did not mention their disease stage, we leveraged GPT-4 to categorise them based on the descriptions of AD stages (online supplemental table 1). Third, for cases that cannot be recognised by GPT-4, we did the segmentation based on patients' behaviours and symptoms. In detail, each text was converted into a list of tokens using pkuseg tokeniser, and symptoms were extracted for expert review to determine the disease stage.

RESULTS

Patient cohort and demographics

Overall, we identified 1211 patients from nine online platforms (figure 1), mainly from online consultation platforms (43%), followed by BBS platforms (24%), e-commerce platforms (23%) and new media platforms (10%). Demographics are shown in table 1. Among 884 patients with gender information, the majority were female. Most identified patients were from Zhejiang, Guangdong and Beijing. Most (82%) were aged 60–90 years. Of the 382 patients with clear disease stages, 42% were classified as having moderate AD, followed by severe AD (34%) and mild AD (24%). Most patients had illness durations of 1–5 years or 5–10 years.

Current diagnosis and treatment

A total of 622 patients mentioned diagnostic tools (online supplemental table 2), with the most common being medical history (97%) and symptoms (84%), followed by

[†]Patient age refers to the age of the patient at the time of consultation or publication of content.

^{‡&#}x27;Unknown' duration of AD means either of the two situations: (1) 20 patients filled out the form as 'several years of AD' but did not specify the exact duration; (2) 57 patients selected the duration of AD as 'more than half a year' on Haodaifu, the longest available option, without specifying the duration. AD. Alzheimer's disease.

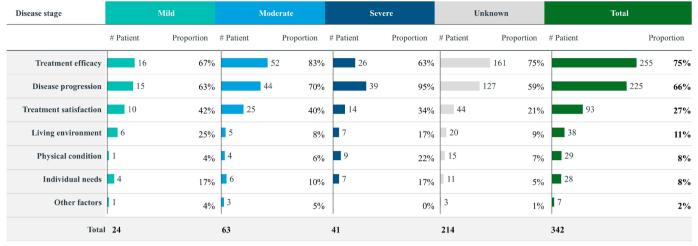


Figure 3 Reasons for patients' treatment switch in different disease stages. 'Other factors' refers to the patient mentioning changes in treatment without specifying the reasons.

physical examinations (41%) and imaging (23%). Cognitive and neurological examinations were less frequently reported. Among patients with moderate and severe AD, other professional examinations, such as blood tests, peripheral blood biomarker tests and cerebrospinal fluid tests, were common. Among 252 patients who sought online consultations, most visited tertiary hospitals in first-tier or new first-tier cities, primarily in the neurology department (60%), with others visiting psychiatry, geriatrics and other specialties.

Treatment for AD can be categorised into pharmacological and non-pharmacological approaches. Among 511 patients who mentioned pharmacotherapy, the most common medications were cholinesterase inhibitors (58%), N-methyl-D-aspartate (NMDA) receptor antagonists (43%) and antipsychotic drugs (29%) (online supplemental figure 3). Regarding pharmacological treatment concerns, patients were primarily focused on efficacy (76%). Compared with patients with mild and severe AD, those with moderate AD were more concerned about comorbidities, while patients with severe AD were significantly more focused on drug accessibility. Overall, treatment modifications were mainly driven by therapeutic efficacy (75%) and disease progression (66%) (figure 3). Among patients with mild and moderate AD, inadequate efficacy, side effects and the availability of new drugs were the main reasons for changes in treatment related to efficacy, while for patients with severe AD, disease progression, such as deterioration and the onset of new symptoms, was the predominant factor. Nonpharmacological treatments were used across all disease stages, including physical exercise (56%), cognitive training (27%) and dietary adjustments (25%). Among 602 patients who mentioned either pharmacological or non-pharmacological treatments, 38% used combination therapy, 26% used only one drug and 21% used a drug alongside non-pharmacological treatment. The most common combination therapy was memantine with physical exercise (online supplemental figure 4).

Patient safety: wandering as an example

A total of 92 patients were reported to have experienced wandering (table 2). Most were aged between 60 and 90 years, with a slight female preponderance. The majority were observed at the moderate stage (39%), and the typical duration of AD was 1–5 years or 5–10 years. Wandering occurred mostly once (43%), followed by multiple times (29%). Triggers were often unknown, but hallucinations/delusions may contribute. Patients who were able to go out alone were more prone to wandering, especially in unfamiliar environments. Caregivers noted the activities of 30 patients during these episodes of disappearance, with most patients being lost while trying to return home. Others were found walking, buying food or hallucinating.

Caregiver self-portrait analysis

Caregivers play an essential role in supporting patients with AD, who are often unable to live independently. In this study, we identified 756 AD caregivers from online platforms, predominantly the children of the patients. Throughout AD progression, home-based care remains the primary mode of caregiving (87%). Especially in the mild stage, 97% of caregivers choose home-based care. However, as the disease progresses, there is increasing adoption of community and institutional care, reflecting the increasing complexity of care needs and the necessity for more specialised care provisions. Among 705 caregivers, 90% focused on disease management, 37% on daily life care and 25% on psychosocial support (online supplemental table 3). As the disease progresses, caregivers' focus shifts from disease management and treatment towards an increased emphasis on meeting the evolving demands for daily life care, disease progression monitoring and prognosis assessment (figure 4). This shift indicates the increasing demands on caregivers to manage patients' daily life needs. They also face challenges related to self-care, fatigue and the need for medical or rehabilitation services. Family and community support become crucial as the disease advances.

Table 2 Demographic characteristics of the patients with AD who experienced wandering

Demographic	Number of patients, n (%)
Age (years) (n=92)	
[50, 60)	4 (4)
[60, 70)	16 (17)
[70, 80)	15 (16)
[80, 90)	19 (21)
[90, 100)	3 (3)
Unknown*	35 (38)
Gender (n=92)	
Female	48 (52)
Male	38 (41)
Unknown*	6 (7)
Disease severity (n=92)	
Mild	9 (10)
Moderate	36 (39)
Severe	16 (17)
Unknown*	31 (34)
Duration of AD (n=92)	
<1 year	4 (4)
1 year≤n<5 years	17 (18)
5 years≤n<10 years	15 (16)
10 years≤n<20 years	5 (5)
Unknown*	51 (55)
Frequency of wandering (n=92)	
One time	40 (43)
Multiple times	27 (29)
Unknown*	25 (27)
Scenarios of wandering (n=92)	
Unknown	62 (67)
Returning home	10 (11)
Walking/Strolling	5 (5)
Looking for someone	3 (3)
Shopping for groceries	3 (3)
Others†	9 (10)

[&]quot;'Unknown' indicates the lack of specific information about the patient who experienced wandering as mentioned in the context.

DISCUSSION Main findings

To our knowledge, this is the first study to use NER and LLM tools to analyse over a decade of online data (2010–2024) on SML regarding Chinese patients with AD and their caregivers. The strengths of this study include the utilisation of various types of online platforms and cutting-edge LLM methods. By incorporating some of the most popular social platforms in China, this study ensures broad coverage and enhances the generalisability of its findings. This study captures prevalent dementia themes on social media while providing detailed

insights into specific aspects of the condition. The NLP tools and GPT-4 used are currently stable and well-established, ¹⁶ ¹⁷ fulfilling our research's text processing needs.

Consistent with previous studies, patients with AD identified in this analysis were predominantly women (66%), ¹⁸ and most were aged 70–90 years (59%). Among 382 patients with confirmed disease stages, moderate AD was the most frequently reported (42%), with a reported disease duration of 1–5 years or 5–10 years. Social media insights discussions shed light on evolving diagnostic methods for AD, from traditional medical history, symptoms and imaging to new biomarker techniques for early and accurate AD diagnosis.

Our findings on AD treatment choices aligned with prior studies, with cholinesterase inhibitors, NMDA receptor antagonists and antipsychotics being the most commonly used therapies. ¹⁹ These treatments are likely preferred for their potential to improve cognitive function and slow disease progression. Treatment adjustments were primarily influenced by efficacy and disease progression, reflecting patient concerns and priorities across different disease stages. Patients placed significant emphasis on drug efficacy (75%), reflecting their keen interest in treatment effectiveness, which is crucial for managing the progressive nature of AD and its impact on quality of life. Patients with moderate AD expressed greater concerns due to more challenging clinical symptoms, ²⁰ while those with severe AD prioritised drug accessibility, possibly reflecting the physical, logistical or financial challenges in obtaining necessary medications. Non-pharmacological interventions like physical exercise, cognitive training and dietary modifications^{21 22} play a significant role in managing AD and are widely adopted across all disease stages. Most patients chose a combination of drug and non-pharmacological treatment, with memantine and physical exercise being common choices, highlighting the acceptance and advantages of a comprehensive treatment strategy for AD management.

A key focus of this study was the analysis of patients with AD who experienced getting lost, which occurred more frequently during the moderate stage due to episodic memory impairment and cognitive decline. Patients with mild AD often live independently, while those with severe AD, who have limited mobility and are frequently bedridden, have a lower risk of getting lost.²³ In fact, episodes of getting lost can occur at any stage of dementia, often leaving family caregivers unprepared.²⁴ Early symptoms of AD often include spatial disorientation, typically emerging within 2 years after the clinical onset and increasing the risk of getting lost. 25 As a globally prevalent issue, alarming statistics indicate that up to 70% of patients with dementia experience at least one episode of getting lost.²⁶ Our study revealed that 43% of patients with AD reported one lost incident, while 29% experienced multiple episodes, addressing a critical research gap in understanding the prevalence of getting lost among patients with AD. Following an initial incident, patients often adapt by staying within familiar areas, avoiding unfamiliar places and limiting outdoor activities within the community.²⁶ Due to this behavioural adaptation and increased caregiver vigilance, measures like closer monitoring, tracking devices and

^{†&#}x27;Others' include activities such as getting a haircut, taking the bus, riding a bike, etc.

AD, Alzheimer's disease.

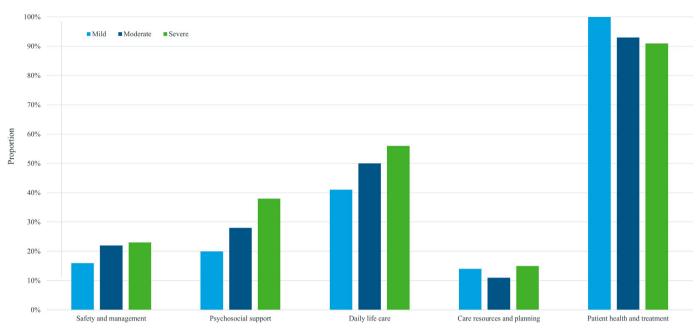


Figure 4 Caregiver topics of concern.

restrictions on unsupervised outings are taken. ^{25–27} Consequently, a significant portion of patients experience only one lost incident. However, it is crucial to recognise that AD progression can still lead to lost episodes even within familiar environments. ^{28 29} Furthermore, not all patients with AD and their caregivers implement proactive measures to forestall the recurrence of getting lost. ²⁷ The inadequate preventive strategies adopted by some caregivers contribute substantially to the elevated risk of repeated lost incidents. One study found that after a 2.5-year follow-up, 40% of patients with initial episodes experienced new lost events, highlighting the ongoing challenge. ²⁵

Similar to previous studies, ¹⁸ ¹⁹ ²² most caregivers in our analysis were adult children, while spouses accounted for <10%, contrasting with studies showing spousal caregivers are more active online. ³⁰ This discrepancy may be due to differences in the specific social forums. A longitudinal study examining both quantitative and qualitative aspects of caregiver discussions on social platforms suggested high engagement in activities of daily living. ¹¹ As disease advances, family and community support becomes increasingly important.

Limitations

Limitations exist in this study. First, it relies on online data, which introduces selection bias, favouring internet-proficient, urban and younger individuals, limiting generalisability. This constraint echoes similar limitations in previous research, where online data often fail to capture the full diversity of demographic profiles. Furthermore, despite efforts to include all patients with AD on digital platforms, inadvertent exclusions persist, challenging the validity and reliability of the study's conclusions. In addition, the unstructured nature of free-text data may lead to inaccuracies and misinterpretation, and hinder trend analysis over time. Finally, distinguishing between patient and caregiver perspectives is challenging, potentially introducing bias due to limited observation.

Implications

This study provides a large sample of patients with AD and their caregivers, contributing to the understanding of their real-world experiences in China. The study aligns with previous research emphasising the importance of large-scale, real-world data for complex diseases like AD. While offering a snapshot of social media insights, future research could track and predict key terms for early AD diagnosis. Additionally, enhancing engagement analysis using NLP on user interactions in public forums deserves future exploration.

Contributors NZ and YZ: web search, statistical analyses and manuscript writing. WC, JX, JL and HL: content summary and manual classification. XX and RR: default keywords and clinical analyses. JG and GW: study design, project management and manuscript revision. GW is the guarantor of this paper. All authors contributed to the article and approved the submitted version.

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Competing interests YZ is an employee of Lundbeck (Beijing) Pharmaceutical Consulting. Other authors declare that they have no competing interests.

Patient consent for publication Not applicable.

Ethics approval This study was reviewed by the Research Ethics Committee of Renji Hospital affiliated with Shanghai Jiao Tong University School of Medicine and deemed exempt from review as it did not meet the definition of 'human or animal subject research'.

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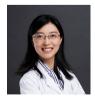
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REFERENCES

- 1 Zhang H-G, Fan F, Zhong B-L, et al. Relationship between left-behind status and cognitive function in older Chinese adults: a prospective 3-year cohort study. Gen Psychiatr 2023;36:e101054.
- 2 Ren R, Qi J, Lin S, et al. The China Alzheimer report 2022. Gen Psychiatr 2022;35:e100751.
- 3 Wang H, Xie H, Qu Q, et al. The continuum of care for dementia: needs, resources and practice in China. J Glob Health 2019:9:020321.
- 4 Jia J, Zuo X, Jia X-F, et al. Diagnosis and treatment of dementia in neurology outpatient departments of general hospitals in China. Alzheimers Dement 2016;12:446–53.
- 5 Zhao W, Jones C, Wu M-LW, et al. Healthcare professionals' dementia knowledge and attitudes towards dementia care and family carers' perceptions of dementia care in China: an integrative review. J Clin Nurs 2022;31:1753–75.
- 6 Chiao CY, Wu HS, Hsiao CY. Caregiver burden for informal caregivers of patients with dementia: a systematic review. *Int Nurs Rev* 2015;62:340–50.
- 7 Cook N, Mullins A, Gautam R, et al. Evaluating patient experiences in dry eye disease through social media listening research. Ophthalmol Ther 2019;8:407–20.
- 8 Schmidt AL, Rodriguez-Esteban R, Gottowik J, et al. Applications of quantitative social media listening to patient-centric drug development. *Drug Discov Today* 2022;27:1523–30.
- 9 Humphrey L, Willgoss T, Trigg A, et al. A comparison of three methods to generate a conceptual understanding of a disease based on the patients' perspective. J Patient Rep Outcomes 2017;1:9.
- 10 Farrar M, Lundt L, Franey E, et al. Patient perspective of tardive dyskinesia: results from a social media listening study. BMC Psychiatry 2021;21:94.
- 11 Bachmann P. Caregivers' experience of caring for a family member with Alzheimer's disease: a content analysis of longitudinal social media communication. *Int J Environ Res Public Health* 2020;17:4412.

- 12 Tahami Monfared AA, Stern Y, Doogan S, et al. Stakeholder insights in Alzheimer's disease: natural language processing of social media conversations. J Alzheimers Dis 2022;89:695–708.
- 13 Luo R, Xu J, Zhang Y, et al. Pkuseg: a toolkit for multi-domain chinese word segmentation. arXiv 2019;190611455.
- 14 Meskó B. Prompt engineering as an important emerging skill for medical professionals: tutorial. J Med Internet Res 2023;25:e50638.
- 15 2023 Alzheimer's disease facts and figures. Alzheimers Dement 2023;19:1598–695.
- 16 Yeung A, laboni A, Rochon E, et al. Correlating natural language processing and automated speech analysis with clinician assessment to quantify speech-language changes in mild cognitive impairment and Alzheimer's dementia. Alzheimers Res Ther 2021;13:109.
- 17 Chen Z, Zhang H, Yang X, et al. Assess the documentation of cognitive tests and biomarkers in electronic health records via natural language processing for Alzheimer's disease and related dementias. Int J Med Inform 2023;170:104973.
- 18 Tahami Monfared AA, Stern Y, Doogan S, et al. Understanding barriers along the patient journey in Alzheimer's disease using social media data. Neurol Ther 2023;12:899–918.
- 19 Isaacson RS, Seifan A, Haddox CL, et al. Using social media to disseminate education about Alzheimer's prevention & treatment: a pilot study on Alzheimer's universe (www.AlzU.org). J Commun Healthc 2018;11:106–13.
- 20 Soria Lopez JA, González HM, Léger GC. Alzheimer's disease. *Handb Clin Neurol* 2019;167:231–55.
- 21 Yamasaki T. Preventive strategies for cognitive decline and dementia: benefits of aerobic physical activity, especially open-skill exercise. *Brain Sci* 2023;13:521.
- 22 Parums DV. A review of the current status of disease-modifying therapies and prevention of Alzheimer's disease. *Med Sci Monit* 2024;30:e945091.
- 23 Zvěřová M. Clinical aspects of Alzheimer's disease. Clin Biochem 2019;72:3–6.
- 24 Li S-H, Wu S-FV, Liu C-Y, et al. Experiences of family caregivers taking care getting lost of persons with dementia: a qualitative study. BMC Psychiatry 2024;24:452.
- 25 Pai MC, Lee CC. The incidence and recurrence of getting lost in community-dwelling people with Alzheimer's disease: a two and a half-year follow-up. PLoS One 2016;11:e0155480.
- 26 Puthusseryppady V, Morrissey S, Aung MH, et al. Using GPS tracking to investigate outdoor navigation patterns in patients with Alzheimer disease: cross-sectional study. JMIR Aging 2022;5:e28222.
- 27 Williams S, Sammut A, Blaxland W. Boundary crossing and boundary violation by service providers and carers in dementia care. *Int* Psychogeriatr 2006;18:565–7.
- 28 Rosenbaum RS, Gao F, Richards B, et al. 'Where to?' remote memory for spatial relations and landmark identity in former taxi drivers with Alzheimer's disease and encephalitis. J Cogn Neurosci 2005;17:446–62.
- 29 Algase DL, Moore DH, Vandeweerd C, et al. Mapping the maze of terms and definitions in dementia-related wandering. Aging Ment Health 2007;11:686–98.
- 30 Ni C, Song Q, Malin B, et al. Examining online behaviors of adultchild and spousal caregivers for people living with Alzheimer disease or related dementias: comparative study in an open online community. J Med Internet Res 2023;25:e48193.



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