Review

New Perspectives on the Aging Lexicon

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The field of cognitive aging has seen considerable advances in describing the linguistic and semantic changes that happen during the adult life span to uncover the structure of the mental lexicon (i.e., the mental repository of lexical and conceptual representations). Nevertheless, there is still debate concerning the sources of these changes, including the role of environmental exposure and several cognitive mechanisms associated with learning, representation, and retrieval of information. We review the current status of research in this field and outline a framework that promises to assess the contribution of both ecological and psychological aspects to the aging lexicon.

Cognitive Aging and the Mental Lexicon

There is consensus in the cognitive sciences that human development extends well beyond childhood and adolescence, and there has been remarkable empirical progress in the field of cognitive aging in past decades [1]. Nevertheless, the role of environmental and cognitive factors in age-related changes in the structure and processing of lexical and semantic representations (see Glossary) is still under debate. For example, age-related memory decline is commonly attributed to a decline in cognitive abilities [2,3], yet some researchers have proposed that massive exposure to language over the course of one’s life leads to knowledge gains that may contribute to, if not fully account for, age-related memory deficits [4–6]. We argue that to resolve such debates we require an interdisciplinary approach that captures how information exposure across adulthood may change the way that we acquire, represent, and recall information. We summarize recent developments in the field (Figure 1, Table 1) and propose a conceptual framework (Figure 2, Key Figure) and associated research agenda that argues for combining ecological analyses, formal modeling, and large-scale empirical studies to shed light on the contents, structure, and neural basis of the aging mental lexicon in both health and disease.

Mental Lexicon: Aging and Cognitive Performance

The mental lexicon can be thought of as a repository of lexical and conceptual representations, composed of organized networks of semantic, phonological, orthographic, morphological, and other types of information [7]. The cognitive sciences have provided considerable knowledge about the computational (Box 1; [8–11]) and neural basis (Box 2; [12,13]) of lexical and semantic cognition, and there has been considerable interest in how such aspects of cognition change across adulthood and aging [14,15].

Past work on the aging lexicon emphasized the amount of information acquired across the life span (e.g., vocabulary gains across adulthood; [15]); however, new evaluations using graph-based approaches suggest that both quantity and structural aspects of representations differ between individuals [16] and change across the life span [17–19]. Such insights were gathered, for example, from a large-scale analysis of free association data from thousands of individuals [17], ranging from 10 to 84 years of age, using networks with words as nodes and edges defined by the strength of shared associations (Figure 1). The analyses suggest that older adults’ semantic networks are less connected (i.e., the words in the network have lower average degrees), less organized (i.e., the words in the network have a lower average local clustering coefficient),
and less efficient (i.e., the shortest path length between any two words in the network is greater relative to those of younger adults) ([17–19]; Figure 1).

Crucially, evidence is also mounting that lexical and semantic structure is crucial to understanding individual cognitive performance in a variety of domains ([7,20–22]; for a review, see [8]). For instance, low clustering in semantic networks, a measure of the extent to which nodes in a network tend to cluster together, has been linked to poorer performance in cued recall of words [23]. Table 1 provides an overview of work that has linked different aspects of semantic network structure to cognitive performance. It suggests that uncovering the structural characteristics of networks may be useful to describe and perhaps predict cognitive performance of older individuals or distinguish between normal and pathological aging [24–26].

Although evidence is mounting concerning the links between aging and semantic structure and potential importance of lexical and semantic structure for cognitive performance, we have yet to gain a full understanding of the sources and mechanisms of these changes. Crucially, a variety of likely candidates have been proposed in the literature, including environmental factors, such as the cumulative nature of information exposure across the life span, and a suite of cognitive mechanisms, such as those concerning learning, representation, and retrieval of information. In what follows, we review past evidence for the role of such factors and discuss the need to assess the relative contribution of each in order to understand the aging lexicon.
A Framework for Understanding the Aging Lexicon

We introduce a novel framework to help us discuss a number of mechanisms that have been linked to age differences in the mental lexicon. Our framework spans both ecological and psychological aspects and consists of four components (Figure 2): (i) the physical, social, and linguistic environment; (ii) the learning processes that build up a mental representation; (iii) the structure of the mental representation itself; and (iv) the processes of manipulating or retrieving information from the representation. Although our illustration may suggest a unidirectional information cascade from the environment to retrieval, our framework does not preclude a dynamic flow, with later components influencing earlier ones. For example, pronunciation tends to change with age, likely as a result of continued experience and efforts to optimize discrimination between words [27,28], and these perceptual/motor changes can be seen as influences on the linguistic environment of those exposed to the language of older speakers. In what follows, we review past evidence concerning each of these components below.

Environment

Cumulative Exposure

Over the course of a lifetime, an average European attends about 10.9 years of schooling [29], watches more than 100,000 hours of TV [30], works 10 different jobs [31], and is part of a countless number of conversations with family, friends, and coworkers. These experiences are the fundamental basis for learning and shaping an individual’s mental representations [32]. Some have argued that older adults can be considered experts in a general sense [33] in that they possess different memory representations because they have been exposed to more environmental input overall, and these have important implications for cognition [4,6,22]. Consistent with Heaps’ law, which states that the number of word types grows with the amount of linguistic input [34], both simulation [35] and empirical work suggest individuals’ vocabulary increases continuously across the life span [15,36]. Moreover, computational models of lifelong word and association learning have been shown to successfully account for performance declines in older adults relative to younger adults, for instance, in word-pair learning [22] and recognition [4], suggesting that the exposure to different amounts of information alone could account for age differences in word-pair memory performance [6].

Different Environments

Older and younger adults differ not only in quantity of experience but also in its content. Younger and older adults differ in occupational status [37], social networks [38], and their use of the Internet and social media [39]. These differences in experience further contribute to shaping the contents of younger and older adults’ lexical and semantic representations [40]. Regrettably, the extent to which differences in the amount and content of information exposed to younger and older adults determines their lexical and semantic representations and cognitive performance remains largely unexplored.

We should note the ecological approaches emphasized above do not logically exclude the contribution of additional mechanisms to age-related differences in cognitive performance, including age-related differences in learning and other factors that we review below. Given the body of knowledge concerning the biology of age-related cognitive decline [1] it is unlikely that ecological explanations alone provide a full understanding of differences in younger and older adults’ mental lexicon. Nevertheless, the results above show that it would be naïve to neglect the role of ecological factors in models of the aging lexicon and that it remains to be tested to what extent additional psychological factors are needed to account for age differences in linguistic and semantic cognition.
Learning

Sensory Constraints

Sensory acuity declines with age [41,42] and differences in cognitive performance, including the ability to learn new associations, have been linked to changes in sensory acuity [43]. Proponents of the information degradation hypothesis have argued that degraded perceptual inputs can lead to errors in perceptual processing, which in turn may affect nonperceptual, higher-order cognitive processes [44]. However, changes in learning and cognitive performance are found even when controlling for sensory limitations during testing [45], implying that age differences in sensory acuity are more likely to reflect general senescent alterations in the aging brain rather than simply sensory deficits in the processing of training and assessment stimuli. Nevertheless, whether specific impairments (e.g., hearing) represent direct contributors to age differences in the aging lexicon remains largely unknown.

Attention/Encoding Failures

Older adults suffer from difficulties in sustaining attention across an encoding episode [2] and in encoding associations between words [46]. As a consequence, a generally held position is that learning depends on executive or cognitive control abilities that are impaired in older adults [47]. Given the important role of cognitive control structures in the processing of linguistic and semantic information, it is likely that age differences in cognitive control play a central role in information acquisition [13], for instance, by impacting how old adults can focus on the relevant and suppress irrelevant information during the learning episode [2].

Prior Knowledge

The encoding of new information is also moderated by an individual’s pre-existing knowledge [48], such as knowledge accumulated over the life span [32,33,49]. For instance, new associations with words that occur overall frequently in the environment and that already possess strong associations with other words are more difficult to form than are associations with infrequent words [6,22]. Experiences consistent with pre-existing larger schemata in semantic memory have been found to consolidate faster into a long-lasting memory trace relative to inconsistent ones [50,51]. Along these lines, older adults have been found to encode new material more efficiently than younger adults but only when the information is encapsulated in a context that is natural for the respective material, for instance, when a target word was placed within a meaningful sentence [52,53]. These results imply that older adults’ exposure to past environments can also have an indirect influence on the mental lexicon by impacting how new information is encoded. Overall, age-related differences in encoding are likely important drivers of differences between younger and older adult’s cognitive performance. Nevertheless, their mediating role in shaping the structure of the aging lexicon is still largely unexplored.

Representation

Decay

A longstanding hypothesis is that memory traces are subject to passive, gradual decay as a result of not using the particular trace [54,55]. Although decay accounts have been widely abandoned in memory research in favor of accounts focused on interference [56], the notion of passive decay has led to successful accounts of, in particular, pathological, age-related changes in mental representations. For instance, degrading the connection strength between words in an associative network could account for the increased semantic priming in patients with Alzheimer’s disease (AD) [57]. Similarly, lesioning specific representational loci in a connectionist model could account for the behavior of patients with semantic dementia in both semantic and lexical tasks [58]. The notion of weakening connection strength lies at the heart of another representation-based account of age differences in cognitive performance. The so-called transmission deficit hypothesis [59] posits that as connections between nodes weaken with age, the transmission of
activation between semantic and lexical word representations is especially affected. This progressive weakening is thought to produce states of semantic activation without lexical or phonological activation, resulting in a feeling of knowing without being able to actually pronounce a word, commonly known as a tip-of-the-tongue state.

Consolidation
Consolidation refers to the process in which an item in memory is transformed into a long-term form taking place both at the level of the synapse (synaptic consolidation) and the brain system (systems consolidation; [60]). Whereas the former works on relatively small timescales, the latter is believed to be ongoing for months or even years [51], altering not only where but also how memories are represented in the brain, including the transformation of episodic representations to more semantic ones [61,62]. That is, it has been argued that systems consolidation involves an active, well-organized decay process that systematically removes selective memories to produce sparser and more efficient memory representations [54]. Although its role across very long timescales, years to decades, is mostly unexplored [51], consolidation does represent a promising alternative for phenomena attributed to passive decay and, generally, a plausible neurophysiological mechanism for age-related changes in the mental lexicon.

Retrieval
Cognitive Control
Models of memory and language typically view the productions of the cognitive system not as direct readouts of internal representations but rather the result of a response mechanism that operates on them [63]. This mechanism is thought to involve cognitive control and retrieval.
strategies. Cognitive control is conceptually related to working memory capacity [64] and, generally, refers to an executive ability that is needed to actively maintain relevant information and inhibit external and internal distractors [65]. Cognitive control is thought to mediate retrieval from memory by reducing interference and enhancing focus on currently activated, task-relevant representations [65,66]. Older adults typically exhibit lower cognitive control resulting in poorer memory retrieval performance in, for instance, verbal fluency or episodic memory tasks [67,68].

Search Strategies
Search in memory refers to the systematic, goal-directed foraging of memory representations [69] and is often modeled as a strategic combination of sustained, focused attention to local areas of the representation (e.g., a particular semantic category) and (random) global switches to distant areas of the representation [9,70,71]. Applications of this modeling approach to verbal fluency tasks have found older adults to exhibit shorter periods of local search than younger adults, which has been attributed to reduced levels of cognitive control [72].

Search strategies and cognitive control do not concern the question of age-related differences in lexical representations directly, but they are nonetheless important; they represent the link between representations and behavior that must be understood to be able to make inferences about the representations underlying observable behavior [8,18,73,74]. Behavior is inevitably determined by both representation and retrieval mechanism, and both are powerful explanations making it difficult to attribute the source of a particular age-related difference unequivocally to either one. This is a major challenge insofar as theoretical and empirical work has suggested age-related differences in both of these components. This has led, for instance, to very different accounts of age-related pathologies for similar types of behavior: Studies have found that semantic cognition of patients with semantic dementia and semantic aphasia could be best accounted for by changes in a controlled retrieval process [66], whereas that of patients with AD had

| Network property | Empirical links to lexical and semantic cognition | Refs |
|------------------|-----------------------------------------------|------|
| Centrality       | Words with higher centrality in associative networks are retrieved more often as the first responses in letter fluency tasks (using PageRank) and are identified faster as a word (rather than a nonword) in a lexical decision task (using PageRank and node degree). | [101,102] |
| Neighborhood     | Words with many phonological or orthographic neighbors (or large neighborhood sizes) are more difficult to identify in spoken word recognition, are produced faster in a naming task, are more frequently involved in tip-of-the-tongue phenomena, and are subject to stronger inhibitory priming. Words with many semantic or associative neighbors are less likely to be remembered in a free recall task and cued recall tasks, trigger lower feelings of knowing, and are more likely to be accepted in new word combinations. Words with high phonological clustering are more difficult to identify in spoken word recognition and lexical decision tasks whereas high associative clustering are remembered better in a cued recall task. | [103–106] [23,107–109] [23,110] |
| Distance         | Words with short semantic or associative distance are judged as more semantically related, remembered better in paired-associate learning tasks, retrieved closer to each other in free recall or verbal fluency tasks, produce stronger priming effects in naming tasks, and lead to faster sentence verification and recognition. Words with low phonological or orthographic distance produce stronger priming effects. | [22,48,70,102,111–116] [117,118] |
| Large-scale structure | Shorter average distances between words in a network are assumed to facilitate the exchange of information exchange and have been empirically linked to creativity. Weak average connections between semantic and phonological representations of words are assumed to drive tip-of-the-tongue occurrences. | [73,119–121] [5,59] |
|                  | Associative schemata facilitate new learning, but also false-memory. | [33,50] |
previously been successfully attributed to representational decay [57]. The difficulty with disentangling representation and process has recently been addressed explicitly in an exchange of papers centering on the nature of search in a verbal fluency paradigm [9,74], which culminated in two insights. First, representations created from behavioral data, such as free associations, can contain signals of the retrieval processes involved in producing the behavioral data. Second, understanding the contribution of each component requires independent sources of data, which are seldom available.

All Together Now: Integrative and Interdisciplinary Approaches to Understanding the Aging Lexicon
Extant explanations of age differences in the mental lexicon and their behavioral consequences have typically relied on only a subset of the four components described above, environment,
learning, representation, and retrieval (Figure 2). For example, whereas some studies focused on the impact of cumulative experience to account for, for instance, paired-associate learning (environment; [6,22]), others considered damage to internal representations and controlled retrieval processes to account for semantic deficits [66], and yet others relied on a combination of attentional deficits (learning) and retrieval processes to account for age-related memory change [2].

Modeling approaches that encompass all four components as sources of age differences are lacking. Ideally, a full account of the aging lexicon should consider all four components to assess
whether age differences can arise from each component independently, their cumulative action, or dynamic interactions among them. Modeling accounts omitting some of the components risk falsely attributing age differences to the subset of evaluated components, when their joint action is more likely.

The goal for future research should be to develop a more integrative formal account of the aging lexicon spanning all four components. To this end, we propose three steps for future research. First, we hope to see researchers build models that integrate ecological and cognitive accounts of age differences in the mental lexicon. Second, the field should deploy large-scale studies that investigate individual and age differences for several indicators of linguistic and semantic cognition to constrain these models. Third, we hope to see increased use of neuroimaging techniques to derive more detailed signals of the contribution of different cognitive components, such as learning, representation, and retrieval. In what follows, we outline a few steps in these directions.

Past research has modeled semantic cognition assuming that representational structure is shared among both younger and older adults [9,71]. Such approaches have favored accounts of aging in the mental lexicon focusing on cognitive aspects [66,71], rather than on the role of the environment, because such frameworks do not capture the impact of environmental exposure on individual and age differences in mental representations. As reviewed above, several results suggest that it is now essential to consider the role of environmental factors. Fortunately, tools to account for the influence of the environment are readily available. Researchers can now choose from a variety of off-the-shelf learning models that turn a continuous stream of environment input, typically large amounts of digitized text, into distributed representation of words and concepts ([10,75]; Box 1). Recent research has demonstrated that varying the amount of text used for training such models can produce some behavioral patterns that are often otherwise attributed to cognitive decline [6,22]; however, the impact of qualitative differences in the environments remains unexplored. One reason for this is the lack of ambitious context-aware cross-sectional and longitudinal projects that could provide a characterization of the language environments of younger and older adults’ over time. Although unprecedented large amounts of contextualized text and speech data are becoming available with the digital revolution [72,76], few of these datasets differentiate age groups or individuals. Thus, one of the challenges for future research is to create age-annotated language corpora. These efforts will need to include measurements of nonlinguistic sensorial information, such as pictures or videos of real-world scenes [77,78], if we are to distinguish the relevance of linguistic versus other types of input to learning and semantic representation. Another challenge is to complement existing learning models to account for the changes in learning arising from the accumulation of knowledge and cognitive and sensory development [6,22].

Representations that are created by training learning models using age-specific language environments have the potential to account for many age differences in cognitive function. To further dissociate the contribution of the four components, large-scale studies that capture a clear set of diverse empirical benchmarks are required. There is a recent trend to conduct so-called mega-studies in the domain of memory and language; that is, studies involving the collection of behavioral data on a large number of linguistic stimuli – now typically in the order of tens of thousands [79,80]. However, some of these resources on the linguistic environment require considerable effort to collect and do not often focus on age differences (Box 3). Future studies may want to seriously consider individual and age differences and capture multiple outcomes across both laboratory and naturalistic settings [81,82] from the same individuals because these can be linked to different aspects of linguistic and semantic performance that give insight into the learning, representation, and retrieval components [83]. Crucially, researchers should be aware that naturalistic settings can provide increased room for the use of compensatory strategies and contextual cues that older adults use to optimize linguistic performance.
New natural language processing techniques are making large amounts of richly annotated data increasingly available [76]. Currently, a description of single individual’s linguistic environment is still challenging as large-scale corpora of written language derived from newspapers and online media has remained mostly aggregate and anonymous and may not be representative of an individual’s natural environment. The advent of Internet-based resources and individual tracking is a promising avenue to address these issues [122].

Child-directed speech is already available through the CHILDES corpus [123], whereas adult speech across the life span is covered in a variety of corpora such as the Switch-board I corpus [124]. Written corpora for children based on children books have also been collected in various languages [125,126]. Written corpora for adults are more comprehensive than those of children, albeit annotations are often incomplete. For example, the widely used British National Corpus (BNC; [127]) corpus contains information about author age for only 26% of the sources.

### Measuring and Modeling the Mental Lexicon across the Life Span

Mega-studies have become common to sample large amounts of lexical and semantic knowledge from individuals [79]. In most cases they have not directly targeted questions about age differences and do not sample individuals across the full adult life span. In turn, new modeling resources are becoming increasingly available and will facilitate and spur on future computational modeling of aging in linguistic and semantic cognition including prominently open-source software for learning representations [10] and simulating retrieval [128].

Comprehensive datasets on vocabulary development are increasingly available. Wordbank contains measurements of vocabulary in early life derived from over 75,000 children in 29 languages [129]. Also available are extensive measures of vocabulary size and prevalence obtained from hundreds to thousands of adults in various languages [130,131]. Semantic knowledge can be assessed from word association norms such as the Small World of Words project, which currently includes age-annotated word association corpora derived from adult Dutch and English speakers [93,102]. Other mega-studies that cover the life span have focused mainly on behavioral measures. These includes age-annotated lexicon projects (naming and lexical decision reaction times) in a variety of languages (see [131] for an overview).

and seem to contribute to differential age-related patterns of results between laboratory and naturalistic settings [81,82,84].

Moving forward, one particular challenge will be to distinguish the contribution of representational differences from the retrieval processes that operate on these representations. Past computational modeling approaches have found it challenging to separate these components [74]. Although it remains to be seen whether these issues can be addressed using computational modeling, neuroimaging approaches represent a promising source of data for dissociation [13]. For example, there has been some progress in using data-driven methods to provide a map of the neural representation of semantic information [85], and future work could use such techniques to quantify age differences in such representations to assess the degree of longitudinal change across individuals’ life span. These neuroimaging techniques may also be used to distinguish or compare the neural representations of linguistic and non-linguistic stimuli [86,87]. Finally, there is significant promise in linking neural biomarkers of age-related decline to both cognitive control and representational aspects of linguistic and semantic cognition (e.g., gray-matter density [88]; functional connectivity [85]; AD-specific biomarkers [89]) to understand the contribution of different neural structures and processes to age differences in the mental lexicon.

### Concluding Remarks

This review suggests that it is important to consider several explanations to understanding the development of the mental lexicon across the life span, including environmental exposure, as well as age-related changes in learning, representation, and information retrieval. As a consequence, future work in this field will require interdisciplinary teams with expertise in linguistics, computational modeling, neuroimaging approaches represent a promising source of data for dissociation [13]. For example, there has been some progress in using data-driven methods to provide a map of the neural representation of semantic information [85], and future work could use such techniques to quantify age differences in such representations to assess the degree of longitudinal change across individuals’ life span. These neuroimaging techniques may also be used to distinguish or compare the neural representations of linguistic and non-linguistic stimuli [86,87]. Finally, there is significant promise in linking neural biomarkers of age-related decline to both cognitive control and representational aspects of linguistic and semantic cognition (e.g., gray-matter density [88]; functional connectivity [85]; AD-specific biomarkers [89]) to understand the contribution of different neural structures and processes to age differences in the mental lexicon.

### Outstanding Questions

To what extent can purely environmental explanations account for reported age differences in lexical and semantic cognition? Are representational deficits necessary to account for differences in normal and pathological aging?

What is the level of dynamic interaction among the four components of our framework (environment, learning, representation, retrieval)? For example, to what degree do chronic differences in mnemonic retrieval strategies change mental representations? Do age-related changes in the structure of mental representations change the linguistic environment that is used as input by other speakers?

To what extent are different types of representation models, such as network-based, connectionist, or distributional models able to predict and explain the same underlying effects and account for the age differences observed in linguistic and semantic cognition?

How can we build on existing corpora or develop new resources to measure the most important properties of individuals’ linguistic environments? Can we annotate existing corpora to include age and sociodemographic information to investigate aspects of the aging lexicon? Is it feasible to deploy mega-studies to capture linguistic, sociodemographic, and biological properties of individuals longitudinally over periods of decades?

How can we integrate the results of different tasks and paradigms, such as reading, language comprehension, and production that may provide contradicting evidence concerning the role of specific mechanisms?

Which neuroimaging methods and analyses can provide the best ways to distinguish learning and search processes from representational deficits in the aging lexicon?

How can or should changes in motivation and goals that direct the cognitive system be captured in models of the aging lexicon? Past research and our overview primarily focused on the integrity and efficiency of information processing, without considering changes in motivation and goals.
modeling, psychological measurement and testing, and neuroscience, that can simultaneously tackle a description of individuals’ linguistic ecologies as well as the cognitive representations and processes that build on an individuals’ lifelong experiences (see Outstanding Questions).

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Resources
https://CRAN.R-project.org/package=text2vec
https://CRAN.R-project.org/package=memnet
https://CRAN.R-project.org/package=memnet
https://aginglexicon.github.io/ for additional resources

References
1. Lindenberger, U. (2014) Human cognitive aging: correct the fortune? Science 346, 572–578.
2. Healey, M.K. and Kahana, M.J. (2016) A four-component model of age-related memory change. Psychol. Rev. 123, 25–69.
3. Salthouse, T.A. (2010) Selective review of cognitive aging. J. Int. Neuropsychol. Soc. 16, 754–760.
4. Buchter, N.E.G. and Reder, L.M. (2007) Modeling age-related memory deficits: a two-parameter solution. Psychol. Aging 22, 104–121.
5. Dahlgren, D.J. (1998) Impact of knowledge and age on tip-of-the-tongue rates. Exp. Aging Res. 24, 139–153.
6. Ramscar, M. et al. (2014) The myth of cognitive decline: non-linear dynamics of lifelong learning. Top. Cogn. Sci. 6, 5–42.
7. Milin, P. et al. (2017) Discrimination in lexical decision. Psychol. One 12, e0171935.
8. Siew, C.S.Q. et al. (2018) Cognitive network science: A review of research on cognition through the lens of network representations, processes, and dynamics. Complexity Published online December 28, 2018. https://doi.org/10.1155/2018/948481.
9. Abbott, J.T. et al. (2015) Random walks on semantic networks can resemble optimal foraging. Psychol. Rev. 122, 558–569.
10. Baayen, R.H. et al. (2019) The discriminative lexicon: a unified computational model for the lexicon and lexical processing in comprehension and production grounded not in (de)composition but in linear discriminative learning. Complexity. 498581.
11. Jones, M.N. et al. (2015) Models of semantic memory. In The Oxford Handbook of Mathematical and Computational Psychology (Burns, C., ed.) pp. 222–254, Oxford University Press.
12. Hickok, G. and Poeppel, D. (2007) The cortical organization of speech processing. Nat. Rev. Neurosci. 8, 393–402.
13. Ramson, M.A. et al. (2017) The neural and computational bases of semantic cognition. Nat. Rev. Neurosci. 18, 42–55.
14. Coher-Shikora, E.R. and Balota, D. (2018) Visual word recognition across the adult lifespan. Psychol. Aging 31, 489–502.
15. Verhaeghen, P. (2020) Aging and vocabulary score: a meta-analysis. Psychol. Aging 18, 332–339.
16. Morea, A.S. et al. (2013) Mapping the structure of semantic memory. Cogn. Sci. 37, 125–145.
17. Dubossarsky, H. et al. (2017) Quantifying the structure of free association networks across the life span. Dev. Psychol. 53, 1560–1570.
18. Wulf, D.U. et al. (2019) Structural differences in the semantic networks of younger and older adults. PsycholSci Published online October 28, 2019. http://doi.org/10.1177/0956797619873572.
19. Zorites, M. et al. (2014) Graph analysis of semantic word association among children, adults, and the elderly. PLoS One 9, 90–99.
20. Brysbaert, M. et al. (2018) The word frequency effect in word processing: an updated review. Curr. Dir. Psychol. Sci. 27, 45–50.
21. Kenett, Y.N. et al. (2018) Flexibility of thought in high creative individuals represented by perfusion analysis. Proc. Natl. Acad. Sci. U. S. A. 115, 867–872.
22. Ramscar, M. et al. (2017) The mismeasurement of mind: Life-span changes in paired-associate learning scores reflect the “cost” of learning, not cognitive decline. Psychol. Sci. 28, 1171–1179.
23. Nelson, D.L. et al. (1993) Implicit memory: effects of network size and interconnectivity on cued recall. J. Exp. Psychol. Learn. Mem. Cogn. 19, 747–764.
24. Palkhovtov, S.V. and Hemmny, L.S. (2014) A computational linguistic measure of clustering behavior on semantic verbal fluency task predicts risk of future dementia in the Nun Study. Cortex 55, 97–106.
25. Zena, J.C. and Austerweil, J.L. (2018) Estimating semantic networks of groups and individuals from fluency data. Comput. Brain Behav. 1, 36–56.
26. Zena, J.C. and Austerweil, J.L. (2019) Analyzing knowledge retrieval impairments associated with Alzheimer’s disease using network analyses. Complexity 2019, 2019.
27. Gahi, S. and Baayen, R.H. (2019) Twenty-eight years of vowels: Tracking phonetic variation through young to middle age adulthood. J. Phon. 74, 42–54.
28. Baayen, R.H. et al. (2015) The Ecclesiastes principle in language change. In The Changing English Language: Psycholinguistic Perspectives (Kundt, M., et al., eds), pp. 21–48, Cambridge University Press.
29. Barro, R.J. and Lee, J.W. (2013) A new data set of educational attainment in the world, 1950–2010. J. Dev. Econ. 104, 184–198.
30. Anon, Average time spent watching television daily in European countries in 2016 and 2017 (in minutes). Statista.
31. European Political Strategy Center (2017) European Political Strategy Center – European countries in 2016 and 2017 (in minutes). Statista. http://www.statista.com/statistics/422719/tv-daily-viewing-time-europe/.
32. European Political Strategy Center (2017) 10 Trends Transforming Education as We Know It. European Commission.
33. Yee, E. and Thompson-Schill, S.L. (2016) Putting concepts into context. Psychon. Bull. Rev. 23, 1015–1027.
34. Umanath, S. and Marsh, E.J. (2014) Understanding how prior knowledge influences memory in older adults. Perspect. Psychol. Sci. 9, 408–428.
35. Heaps, H.S. (1978) Information Retrieval: Computational and Theoretical Aspects. Academic Press.
36. Montag, J.L. et al. (2018) Quantity and diversity: simulating early word learning environments. Cogn. Sci. 42, 375–412.
37. Brysbaert, M. et al. (2016) How many words do we know? Practical estimates of vocabulary size dependent on word definition, the degree of language input and the participant’s age. Front. Psychol. 7, 1116.
38. Fisher, G.G. et al. (2014) Mental work demands, retirement, and longitudinal trajectories of cognitive functioning. J. Occup. Health Psychol. 19, 231–242.
39. Wrons, C. et al. (2013) Social network changes and life events across the life span: a meta-analysis. Psychol. Bull. 139, 53–80.
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39. Zickuhr, K. and Madden, M. (2012) Older adults and internet use. Pew Research Center Internet & Technology, hdl:www.pewinternet.org/2012/05/older-adults-and-internet-use/

40. Johns, B.T. et al. (2019) Using exponential optimization to build lexical representations. Psychon. Bull. Rev. 26, 103–126

41. Stevenson, R.A. et al. (2013) Deficits in auditory speech perception in normal aging emerge at the level of whole-word recognition. Neurobiol. Aging 36, 283–291

42. Watson, G.R. (2001) Low vision in the geriatric population: rehabilitation and management. J. Am. Geriatr. Soc. 49, 317–330

43. Naveh-Benjamin, M. and Kilb, A. (2014) Age-related differences in associative memory: The role of sensory decline. Psychol. Aging 29, 672–693

44. Monge, Z.A. and Madden, D.J. (2016) Linking cognitive and sensory and cognitive performance in old age: Not due to sensory and cognitive performance differences operating during cognitive assessment. Psychol. Aging 31, 766–783

45. Hasher, L. (2004) Inhibitory control, circadian arousal, and age. In Attention and Performance. Attention and Performance XVII: Cognitive Regulation of Performance: Interaction of Theory and Application (Gopher, D. and Koriat, A., eds), pp. 653–675. Taylor & Francis Group

46. Brod, G. et al. (2013) The influence of prior knowledge on memory: a developmental cognitive neuroscience perspective. Front. Behav. Neurosci. 7, 139

47. Tse, D. et al. (2007) Schemas and memory consolidation. Science 316, 76–92

48. Duba, Y. et al. (2015) The consolidation and transformation of memory. Neuro 88, 20–32

49. Matzen, L.E. and Benjamin, A.S. (2013) Older and wiser: older adults’ episodic word memory benefits from sentence study contexts. Psychol. Aging 28, 754–767

50. Niv, Y. et al. (2015) Reinforcement learning in multidimensional environments relies on attention mechanisms. J. Neurosci. 35, 8145–8157

51. Oberauer, K. and Lewandowsky, S. (2008) Forgetting in immediate serial recall: decay, temporal distinctiveness, or interference? Psychol. Rev. 115, 544–576

52. Borge-Holthoefer, J. et al. (2011) Modeling abnormal priming in Alzheimer’s patients with a free association network. PLoS One 6, e22651

53. Dilks, K. et al. (2008) A single-system account of semantic and lexical deficits in five semantic dementia patients. Cogn. Neurosci. 25, 136–164

54. Burke, D.M. et al. (1991) On the tip of the tongue: what causes word finding failures in young and old adults? J. Mem. Lang. 30, 542–579

55. Duda, Y. (2012) The restless engrain: consolidations never end. Annu. Rev. Neurosci. 35, 227–247

56. Nasel, L. and Moscovitch, M. (1997) Memory consolidation, retrograde amnesia and the hippocampal complex. Curr. Opin. Neurobiol. 7, 217–227

57. Winou, G. and Moscovitch, M. (2011) Memory transformation and systems consolidation. J. Int. Neuropsychol. Soc. 17, 756–780

58. Rajajamakers, J.G. and Shrift, R.M. (1981) Search of associative memory. Psychol. Rev. 88, 90–134

59. Braver, T.S. (2012) The variable nature of cognitive control: a dual mechanisms framework. Trends Cogn. Sci. 16, 106–113

60. Unsworth, N. et al. (2014) Working memory and fluid intelligence: capacity, attention control, and secondary memory retrieval. Cogn. Sci. 37, 1–26

61. Hofman, P. et al. (2018) Concepts, control, and context: a connectionist account of normal and disordered semantic cognition. Psychol. Rev. 125, 263–328

62. Balota, D.A. et al. (1999) Verbal and false memory in healthy older adults and in Dementia of the Alzheimer Type. Cogn. Neuropsychol. 16, 361–384

63. Unsworth, N. et al. (2013) Working memory capacity and retrieval from long-term memory: The role of controlled search. Mem. Cogn. 41, 242–254

64. Hills, T.T. et al. (2015) Exploration versus exploitation in space, mind, and society. Trends Cogn. Sci. 19, 46–54

65. Wulf, D.U. et al. (2013) Wormholes in memory: is memory one representation or many? In Proceedings of the 35th Annual Conference of the Cognitive Science Society (M. Khun et al., eds), pp. 3817–3823. Cognitive Science Society, Austin, TX

66. Hills, T.T. et al. (2012) Optimal foraging in semantic memory. Psychol. Rev. 119, 431–440

67. Hills, T.T. et al. (2013) Mechanisms of age-related decline in memory search across the adult life span. Dev. Psychol. 49, 2396

68. Bullmore, E. and Sporns, O. (2012) The economy of brain network organization. Nat. Rev. Neurosci. 13, 306–314

69. Jones, M.N. et al. (2015) Hidden processes in structural representations: a reply to Abbott, Austerweil, and Griffiths (2015). Psychol. Rev. 122, 570–574

70. Jones, M.N. and Miewhart, D.J. (2007) Representing word meaning and order information in a composite holographic lexicon. Psychol. Rev. 114, 37–57

71. Hirschberg, J. and Manning, C.D. (2015) Advances in natural language processing. Science 349, 261–266

72. Borch, A.M. et al. (2017) The challenge of abstract concepts. Psychol. Bull. 143, 263–292

73. Sadeghi, Z. et al. (2015) You shall know an object by the company it keeps: an investigation of semantic representations derived from object co-occurrence in visual scenes. Neuropsychologia 70, 52–61

74. Koulers, E. and Balota, D.A. (2015) Megastudies, crowdsourcing, and large datasets in psycholinguistics: an overview of recent developments. Q. J. Exp. Psychol. 68, 1457–1469

75. Hills, T.T. et al. (2015) The strong connection between sensory perceptual decline in healthy aging and Alzheimer’s patients with a free association network. PLoS One 6, e22651

76. Dilks, K. et al. (2008) A single-system account of semantic and lexical deficits in five semantic dementia patients. Cogn. Neurosci. 25, 136–164

77. Burke, D.M. et al. (1991) On the tip of the tongue: what causes word finding failures in young and old adults? J. Mem. Lang. 30, 542–579

78. Duda, Y. (2012) The restless engrain: consolidations never end. Annu. Rev. Neurosci. 35, 227–247

79. Nasel, L. and Moscovitch, M. (1997) Memory consolidation, retrograde amnesia and the hippocampal complex. Curr. Opin. Neurobiol. 7, 217–227

80. Winou, G. and Moscovitch, M. (2011) Memory transformation and systems consolidation. J. Int. Neuropsychol. Soc. 17, 756–780

81. Rajajamakers, J.G. and Shrift, R.M. (1981) Search of associative memory. Psychol. Rev. 88, 90–134

82. Braver, T.S. (2012) The variable nature of cognitive control: a dual mechanisms framework. Trends Cogn. Sci. 16, 106–113
90. Collins, A.M. and Quillian, M.R. (1969) Retrieval time from semantic memory. J. Verbal Learning Verbal Behav. 8, 240–247
91. Siew, C.S. (2013) Community structure in the phonological network. Front. Psychol. 4, 553
92. Siew, C.S. and Vitevitch, M.S. (2016) Spoken word recognition and serial recall of words from components in the phonological network. J. Exp. Psychol. Learn. Mem. Cogn. 42, 394–410
93. De Deyne, S. et al. (2019) The “Small World of Words”: English word association norms for over 12,000 cue words. Behav. Res. Methods 51, 1031–1036
94. Nelson, D.L. et al. (2004) The University of South Florida free association, rhyme, and word fragment norms. Behav. Res. Methods Instrum. Comput. 36, 402–407
95. Stella, M. et al. (2018) Multiple-model of mental lexicon reveals explosive learning in humans. Sci. Rep. 8, 2259
96. Rumelhart, D.E. et al. (1986) Learning representations by back-propagating errors. Nature 323, 533–536
97. Farah, M.J. and McDermott, J.L. (1991) A computational model of semantic memory impairment: modality specificity and emergent category specificity. J. Exp. Psychol. Gen. 120, 339–357
98. Landauer, T.K. and Dumais, S.T. (1997) A solution to Plato’s problem: the latent semantic analysis theory of acquisition, induction, and representation of knowledge. Psychol. Rev. 104, 211–240
99. Hoffman, P. and Morcom, A.M. (2018) Age-related changes in the neural networks supporting semantic cognition: a meta-analysis of 47 functional neuroimaging studies. Neurosci. Biobehav. Rev. 84, 134–150
100. Nyberg, L. et al. (2010) Longitudinal evidence for diminished frontal cortex age in aging. Proc. Natl. Acad. Sci. U. S. A. 107, 22682–22686
101. Griffiths, T.L. et al. (2007) Google and the mind: predicting fluency with PageRank. Psychol. Sci. 18, 1069–1076
102. De Deyne, S. et al. (2013) Better explanations of lexical and semantic cognition using networks derived from continued rather than single-word associations. Behav. Res. Methods 45, 480–488
103. Sommers, M.S. (1996) The structural organization of the mental lexicon and its contribution to age-related declines in spoken-word recognition. Psychol. Aging 11, 333–341
104. Gordon, J.K. and Kurczek, J.C. (2014) The ageing neighbourhood: phonological density in naming. Lang. Cogn. Neurosci. 29, 320–344
105. Vitevitch, M.S. and Sommers, M.S. (2003) The facilitative influence of phonological similarity and neighborhood frequency in speech production in younger and older adults. Mem. Cogn. 31, 491–504
106. Dufour, S. and Peereeman, R. (2003) Inhibitory priming effects in auditory word recognition. When the target’s competitors conflict with the prime word. Cognition 88, 33–44
107. Nelson, D.L. and Schreiber, T.A. (1999) Word concreteness and word structure as independent determinants of recall. J. Mem. Lang. 31, 237–260
108. Schreiber, T.A. and Nelson, D.L. (1998) The relation between feelings of knowing and the number of neighboring concepts linked to the test cue. Mem. Cogn. 26, 869–888
109. Vecchi, E.M. et al. (2017) Spicy adjectives and nominal donkeys: capturing semantic deviation using compositionality in distributional spaces. Cogn. Sci. 41, 102–136
110. Chan, K.Y. and Vitevitch, M.S. (2009) The influence of the phonological neighborhood clustering coefficient on spoken word recognition. J. Exp. Psychol. Hum. Percept. Perform. 35, 1934–1949
111. Kenett, Y.N. et al. (2017) The semantic distance task: quantifying semantic distance with semantic network path length. J. Exp. Psychol. Learn. Mem. Cogn. 43, 1470–1486
112. Morton, N.W. and Polyn, S.M. (2016) A predictive framework for evaluating models of semantic organization in free recall. J. Mem. Lang. 86, 119–140
113. Bousfield, W.A. (1963) The occurrence of clustering in the recall of randomly arranged associates. J. Gen. Psychol. 49, 229–240
114. Sperber, R.D. et al. (1979) Semantic priming effects on picture and word processing. Mem. Cogn. 7, 339–345
115. Collins, A.M. and Loftus, E.F. (1975) A spreading activation theory of semantic processing. Psychol. Rev. 82, 407–428
116. Anderson, J.R. (1974) Retrieval of propositional information from long-term memory. Cogn. Sci. 6, 451–474
117. Vitevitch, M.S. and Sommers, M.S. (2003) The facilitative in word recognition using compositionality in distributional spaces. Nature 423, 536–537
118. Vitevitch, M.S. and Sommers, M.S. (2003) The facilitative in word recognition using compositionality in distributional spaces. Nature 423, 536–537
119. Vitevitch, M.S. and Sommers, M.S. (2003) The facilitative in word recognition using compositionality in distributional spaces. Nature 423, 536–537
120. Vitevitch, M.S. and Sommers, M.S. (2003) The facilitative in word recognition using compositionality in distributional spaces. Nature 423, 536–537
121. Kenett, Y.N. and Faust, M. (2019) A semantic network cartography of the creative mind. Trends Cogn. Sci. 23, 274–276
122. Nguyen, D. et al. (2013, June) “How old do you think I am?” A study of language and age in Twitter. In Seventh International AAAI Conference on Weblogs and Social Media
123. Godfrey, J. (2000) The CHILDES Project: Tools for Analyzing Talk: The Database (3rd edn), Lawrence Erlbaum Associates
124. Godfrey, J. et al. (1992) SWITCHBOARD: Telephone speech corpus for research and development. In IEEE International Conference on Acoustics, Speech, and Signal Processing (vol. 1), pp. 517–520, Institute of Electrical and Electronics Engineers
125. Masterson, J. et al. (2019) Children’s printed word database: Continuities and changes over time in children’s early reading vocabulary. Br. J. Psychol. 101, 221–242
126. Schröter, P. and Schroeder, S. (2017) The Developmental Lexicon Project: a behavioral database to investigate visual word recognition across the lifespan. Behav. Res. Methods 49, 2183–2203
127. Leech, G. (1992) 100 million words of English: the British National Corpus. Lang. Res. 28, 1–13
128. Siew, C.S. et al. (2019) spreading activation in a network. Behav. Res. Methods 51, 920–929
129. Frank, M.C. et al. (2017) Wordbank: an open repository for developmental vocabulary data. J. Child Lang. 44, 677–694
130. Keuleers, E. et al. (2015) Word knowledge in the crowd: measuring vocabulary size and word prevalence in a massive online experiment. Q. J. Exp. Psychol. 68, 1685–1692
131. Brysbaert, M. et al. (2019) Word prevalence norms for 62,000 English lemmas. Behav. Res. Methods 51, 467–479