

Understanding the role of linguistic distributional knowledge in cognition: A systematic comparison of tasks, models and parameters

Cai Wingfield

c.wingfield@lancaster.ac.uk

Louise Connell

l.connell@lancaster.ac.uk

@drlouiseconnell



Download PDF copy

Lancaster University



European Research Council
Established by the European Commission



This project has received funding from the European Research Council (ERC) under the European Union's Horizon 2020 research and innovation programme under grant agreement No. 852448.

Linguistic distributional models (LDMs) learn associations between words from occurrences in language, and model how they can aid in cognitive tasks.

LDM research has recently focussed on certain neural-network models, optimising for similarity-driven tasks.

Do these recommendations generalise to modelling conceptual processing as a whole?

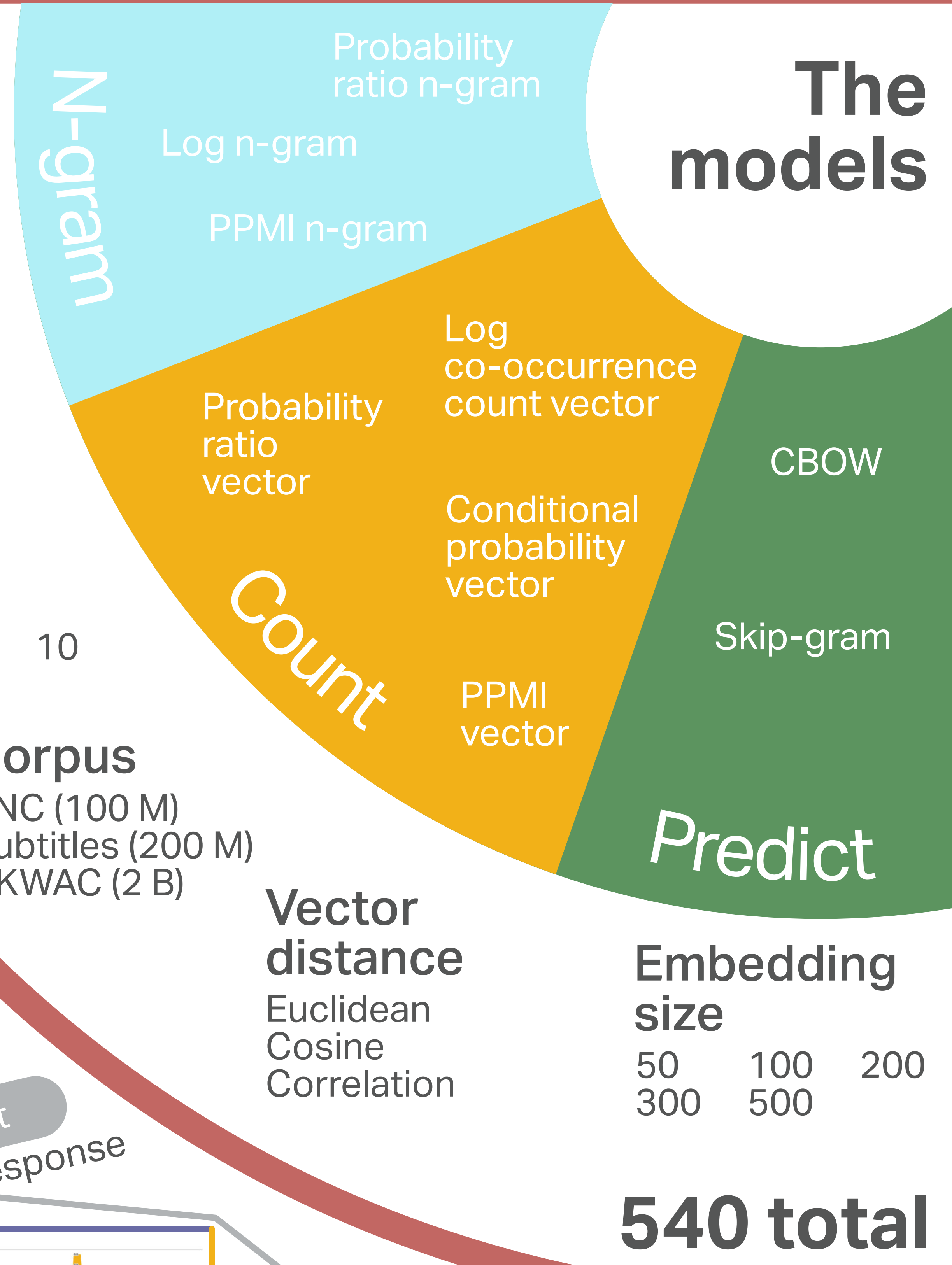
This study comprises a systematic comparison of LDMs on a wide range of tasks.

We ask: is there a one-size-fits-all LDM which works for general cognitive processing, and how performance varies by task features?

Using Bayesian model comparisons, we make recommendations as to the optimal LDM, corpus type and parameters for tasks with particular features.

We separately trained each LDM on each corpus with each parameter setting.

Each LDM was evaluated against each task dataset, with Bayesian comparison to a baseline model.



Context window

1 3 5 10

Corpus

BNC (100 M)
Subtitles (200 M)
UKWAC (2 B)

Vector distance

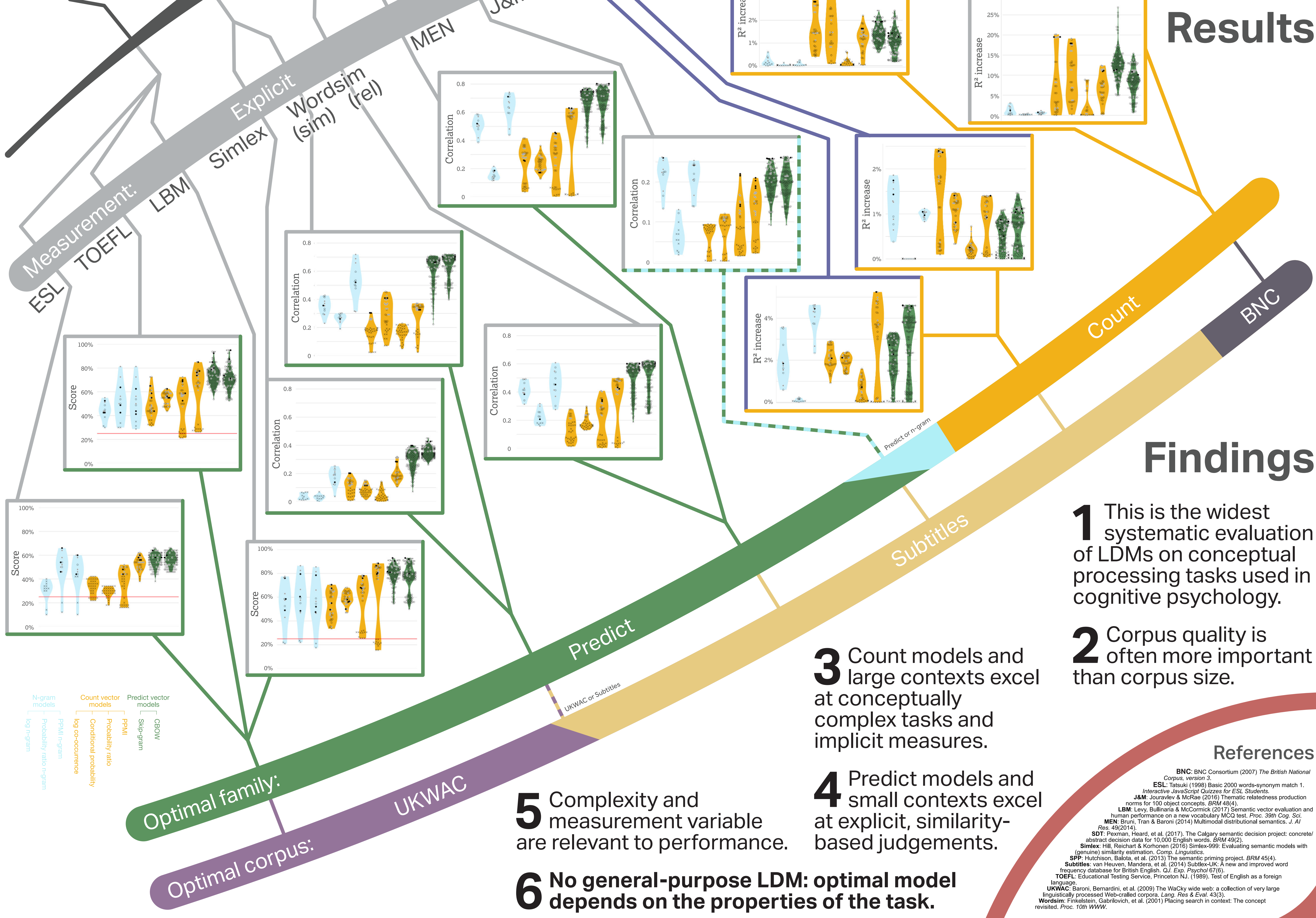
Euclidean
Cosine
Correlation

Embedding size

50 100 200
300 500

540 total

Task datasets



Results

Findings

- 1 This is the widest systematic evaluation of LDMs on a collection of conceptual processing tasks used in cognitive psychology.
- 2 Corpus quality is often more important than corpus size.
- 3 Count models and large contexts excel at conceptually complex tasks and implicit measures.
- 4 Predict models and small contexts excel at explicit, similarity-based judgements.
- 5 Complexity and measurement variable are relevant to performance.
- 6 No general-purpose LDM: optimal model depends on the properties of the task.

References

BNC: BNC Consortium (2007) *The British National Corpus*, version 3.
 ESL: Tateski (1998) Basic 2000 words-synonym match 1.
 Interactive JavaScript Quizzes for ESL Students.
 J&M: Journeaire & McRae (2016) Thematic relatedness production norms for 100 object concepts. *BRM* 48(4).
 LBM: Levy, Bullinaria & McCormick (2017) Semantic vector evaluation and human performance on a new vocabulary MCQ test. *Proc. 39th Cog. Sci. Conf.* 492(2014).
 MEN: Brunj, Tran & Baroni (2014) Multinomial distributional semantics. *J. AI Res.* 49(2014).
 SDT: Pexman, Heard, et al. (2017). The Calgary semantic decision project: concrete/abstract decision data for 10,000 English words. *BRM* 49(2).
 SPP: Hutchinson, Sakota, et al. (2015) The semantic priming project. *BRM* 45(4).
 Simlex: Hill, Reichart & Korhonen (2016) Simlex-999: Evaluating semantic models with (genuine) similarity estimation. *Comp. Linguistics*.
 Subtitles: van Heuven, Mandera, et al. (2014) *Sublex-UK: A new and improved word frequency database for British English*. *Q. J. Exp. Psychol.* 67(6).
 TOEFL: Educational Testing Service, Princeton NJ, (1989). *Test of English as a foreign language*.
 UKWAC: Baroni, Bernardini, et al. (2009) *The WaCy wide web: a collection of very large linguistically processed Web-crawled corpora*. *Lang. Res. & Eval.* 43(3).
 Wordsim: Finkelestein, Gabrionovich, et al. (2001) Placing search in context: The concept revisited. *Proc. 10th WWW*.