

Running Head: Computational Mechanisms of Development

Computational Mechanisms of Development?  
Connectionism and Bilingual Lexical Representation

Ping Li<sup>1</sup> and Xiaowei Zhao<sup>2</sup>

<sup>1</sup> The Hong Kong Polytechnic University

<sup>2</sup> Emmanuel College

Address for correspondence:

**Ping Li**, The Hong Kong Polytechnic University, 11 Yuk Choi Rd, Hung Hom, Hong Kong

[ping2.li@polyu.edu.hk](mailto:ping2.li@polyu.edu.hk)

## Computational Mechanisms of Development?

### Connectionism and Bilingual Lexical Representation

The Ontogenesis Model (OM) of Bordag, Gor and Opitz (2021) is a good example of applying computational thinking to the study of key issues in bilingual lexical representation. With the introduction of two critical computational concepts, namely Multidimensionality and Fuzziness, this model aims at explaining bilingual lexical representation through a theoretical framework of ‘ontogenesis’. A key feature of the model that distinguishes itself from other bilingual computation-minded models such as BIA and BIA+, as the authors argue, is the focus on development of lexical representation (hence ontogenesis in the name of the model). This focus is particularly important, and since the earliest days of BIA and BIA+ models, there has been an urgent need for the field to move from ‘proficient bilingual model’ to ‘developmental bilingual model’ (Li, 2002).

While the OM provides a good illustration of the complexity and the dynamic nature of bilingual lexical representation, it falls short of delivering the key promise on accounting for the *developmental* processes and underlying principles of bilingual lexical representation. In our view, there are significant gaps with the current formulation of the OM model as either a theoretical or an analytic framework. First and foremost, it largely dismisses the significant amount of work in the past two decades that has been devoted to address precisely the same questions that OM asks, the connectionist models of bilingual representation (see a few examples in Thomas, 1997; French, 1998; Li & Farkas, 2002; Lewy & Grosjean, 2008; Zhao & Li, 2010, 2013; Peñaloza, Grasemann, Dekhtyar, Miikkulainen, & Kiran, 2019; along with Shook & Marian, 2013; Kiran, Graesman, Sandberg, & Miikkulainen, 2013 and other articles in the 2013 special issue published by this journal). Bordag et al. write, “other frameworks, e.g.,

connectionist models, model non-optimal representations via, e.g., non-final weights and optimize them via re-weighting of connections due to new input..., fuzziness in the OM refers to imprecise lexical encoding due to a broad range of linguistic and cognitive factors and the learning conditions.” This statement, on the one hand, does not do justice to the significant amount of work inspired by connectionism that has indeed incorporated ‘the broad range of linguistic and cognitive factors and learning conditions’, and on the other, reflects a misunderstanding of what the weight updating and optimization in connectionist networks really are. As Shirai’s (2018) recent synthesis demonstrates (see especially Chapters 3 and 4), connectionist bilingual lexical representations have indeed attempted to incorporate a wide range of linguistic and cognitive factors, including the ones not even discussed in the OM model such as working memory and its impact on individual difference (see also Ellis, 2003; Wen, Biedroń, & Skehan, 2017). Further, weight updating and optimization are mechanisms used by the brain’s neural circuit to accomplish the process of learning and development, and they are based on realistic biological principles that provide the necessary mechanisms for a cognitively plausible computational account, which brings us to the next point.

Second, in defending connectionism and its success in modeling developmental L2 lexical representation, we should also point out that OM lacks precisely the kind of computational mechanistic account provided by connectionist models in disentangling the complex interactions among the key linguistic and cognitive factors (see Grant, Legault, & Li, 2019; Li, 2013; Li & Zhao, 2017 for discussion). Indeed, the BIA and BIA+ models are based on connectionist-like mechanisms (e.g., the original IA models by McClelland & Rumelhart, 1981 were precursors to connectionism; see Li & Zhao, 2020). Rather than dismissing the connectionist architecture, the OM should be able to benefit from integrating computational

mechanisms of connectionist models. The OM aims at modeling three key dimensions of linguistic domains, mappings, and networks (including IntraNetwork and InterNetwork), but it is unclear, unlike in connectionist models, how these dimensions can be actually modeled and implemented computationally, and what plausible mechanisms are to be deployed in the modeling enterprise so that the authors and others can verify and test. As an example, Figure 3 of Bordag et al. nicely illustrates the progressively enlarged network connections, but the OM provides no quantitative methods to actually model such progression or developmental changes. In our view, these dimensions match well with the computational architecture of the connectionism-based DevLex-II (Li, Zhao & MacWhinney, 2007; Zhao & Li, 2010), a multi-layer neural network model with three connected self-organizing maps representing basic linguistic contents (phonology, semantic, and the articulatory sequence) of the bilingual lexicon. The DevLex model was originally designed as a *developmental* model for L1 (Li & Farkas, 2004; Li et al., 2007) and DevLex-II is focused on the development of bilingual lexicon, having the same goal as the OM. Through computational mechanisms such as self-organization (SOM; Kohonen, 2001) and Hebbian Learning (Hebb, 1949), DevLex-II can explicitly model the development of bilingual lexical representation and empirical patterns of priming (Zhao & Li, 2013), key aspects that the OM is designed for explaining. Connectionist models, including DevLex-II and others starting from the late 1990's (e.g., Thomas, 1997; French, 1998), have aimed to provide the kind of computational mechanisms that the OM currently lacks, particularly with regard to such concepts as the word association networks within (OM's IntraNetwork) and between languages (OM's InterNetwork).

Finally, a crucial concept of the OM is its Fuzzy Lexical Representation (FLR) account. This view is highly consistent with what has been proposed by many connectionist language

development models, in both L1 (e.g. Plunkett & Marchman, 1996; Li et al, 2007) and L2 (Hernandez, Li, & MacWhinney, 2005; Li & Zhao, 2013; Zhao & Li, 2021). Specifically, Hernandez et al. (2005; see Fig. 1) proposed the concept of ‘parasitism’, according to which factors such age of acquisition, proficiency, and in particular competition/interaction between L1 and L2, are responsible for fuzzy L2 lexical-semantic representations which in turn lead to inaccurate lexical comprehension and production in both semantic and phonological domains. This leads to the question of why OM chooses to ignore L1-L2 interaction, given that the dynamic interaction of L2 with L1 is a core process that drives the outcome fuzzy L2 representation, based on both theoretical perspectives and neurocognitive evidence (e.g., Claussenius-Kalman, Hernandez, & Li, 2021; Zhang, Yang, Wang & Li, 2020).

Despite our defending connectionism in this commentary, we believe that OM provides a good platform for further discussion and investigation of core issues in bilingual lexical representation, but the integration of connectionist principles into its architecture could enhance its plausibility and generalizability.

### References

- Bordag, D, Gor, K and Opitz, A (2021) Ontogenesis Model of the L2 Lexical Representation. *Bilingualism: Language and Cognition*, 1-17. doi:10.1017/S1366728921000250
- Claussenius-Kalman, H, Hernandez, A and Li, P (2021) Expertise, Ecosystem, and Emergentism: Dynamic developmental bilingualism. *Brain and Language*, in press.
- Ellis, NC (2003) Constructions, chunking, and connectionism: The emergence of second language structure. In Doughty, CJ and Long, MH (Eds.), *The handbook of second language acquisition*, Malden, MA: Blackwell Publishing Ltd pp. 63-103.

- French, RM (1998) A simple recurrent network model of bilingual memory. In M. A. Gernsbacher & S. J. Derry (Eds.), *Proceedings of the 20th Annual Conference of the Cognitive Science Society* (pp. 368-373). Mahwah, NJ: Erlbaum.
- Grant, A, Legault, J and Li, P (2019) What do bilingual models tell us about the neurocognition of multiple languages? In Schwieter, J (ed.), *The handbook of the neuroscience of multilingualism*. Wiley-Blackwell, pp.48-74.
- Hernandez, A, Li, P and MacWhinney, B (2005) The emergence of competing modules in bilingualism. *Trends in Cognitive Sciences*, 9(5), 220-225.
- Hebb, D (1949). *The organization of behavior: A neuropsychological theory*. New York, NY: Wiley.
- Kiran, S, Graesman, U, Sandberg, C and Miikkulainen, R (2013) A computational account of bilingual aphasia rehabilitation. *Bilingualism: Language and Cognition*, 16 (Special Issue), 325-342. doi:10.1017/S1366728912000533
- Kohonen, T (2001) *Self-organizing maps (3rd ed.)*. Berlin: Springer.
- Lewy, N and Grosjean, F (2008) The Lewy and Grosjean BIMOLA model. In Grosjean, F (ed.), *Studying bilinguals*. Oxford: Oxford University Press, pp. 201–210.
- Li, P (2002) Bilingualism is in dire need of formal models. *Bilingualism: Language and Cognition*, 5(3), 213. doi:10.1017/S1366728902253018
- Li, P (2013) Computational modeling of bilingualism: How can models tell us more about the bilingual mind? *Bilingualism: Language and Cognition*, 16(2), 241-245.  
doi:10.1017/S1366728913000059

- Li, P and Farkas, I (2002) A self-organizing connectionist model of bilingual processing. In Heredia, R and Altarriba, J (Eds.), *Bilingual sentence processing*. North-Holland: Elsevier Science Publisher, pp.59-85.
- Li, P and Zhao, X (2017) Computational modeling. In de Groot, AMB and Hagoort, P (Eds.), *Research methods in psycholinguistics and the neurobiology of language: A practical guide*. Hoboken, NJ: John Wiley & Sons, Inc, pp. 208-229.
- Li, P and Zhao, X (2020) Connectionism. In Aronoff, M (Ed.), *Oxford bibliographies online (Linguistics)*. Oxford University Press. ([www.oxfordbibliographies.com](http://www.oxfordbibliographies.com); originally published 2012, last updated 2020)
- Li, P, Zhao, X and MacWhinney, B (2007) Dynamic self-organization and early lexical development in children. *Cognitive Science: A Multidisciplinary Journal*, 31, 581-612.
- McClelland, J and Rumelhart, D (1981). An interactive activation model of context effects in letter perception: Part 1. An account of basic findings. *Psychological Review*, 88, 375-407.
- Peñaloza, C, Grasemann, U, Dekhtyar, M, Miikkulainen, R and Kiran, S (2019) BiLex: A computational approach to the effects of age of acquisition and language exposure on bilingual lexical access. *Brain and Language*, 195, 104643.  
<https://doi.org/10.1016/j.bandl.2019.104643>
- Plunkett, K and Marchman, VA (1996) Learning from a connectionist model of the acquisition of the English past tense. *Cognition*, 61(3), 299–308. [https://doi.org/10.1016/s0010-0277\(96\)00721-4](https://doi.org/10.1016/s0010-0277(96)00721-4)
- Shook, A and Marian, V (2013) The bilingual language interaction network for comprehension of speech. *Bilingualism: Language and Cognition*, 16(2), 304-324.

Shirai, Y (2018) *Connectionism and second language acquisition*. Routledge.

<https://doi.org/10.4324/9780203118085>

Thomas, MSC (1997) *Connectionist networks and knowledge representation: The case of bilingual lexical processing*. PhD thesis, Oxford University.

Wen, Z, Biedroń, A and Skehan, P (2017). Foreign language aptitude theory: Yesterday, today and tomorrow. *Language Teaching*, 50(1), 1-31. doi:10.1017/S0261444816000276

Zhang, X, Yang, J, Wang, R and Li, P (2020) A neuroimaging study of semantic representation in first and second languages. *Language, Cognition, and Neuroscience*, 35, 1223-1238.

<https://doi.org/10.1080/23273798.2020.1738509>

Zhao, X and Li, P (2010) Bilingual lexical interactions in an unsupervised neural network model. *International Journal of Bilingual Education and Bilingualism*. 13, 505-524.

Zhao, X and Li, P (2013) Simulating cross-language priming with a dynamic computational model of the lexicon. *Bilingualism: Language and Cognition*, 16, 288-303.

Zhao, X and Li, P (2021) Fuzzy or Clear? A Computational Approach Towards Dynamic L2 Lexical-Semantic Representation. *Frontiers in Communication*. Ms. Under review