

Modelling Cross-Situational Learning on Full Sentences in Few Shots with Simple RNNs

Xavier Hinaut, Subba Reddy Oota, Alexandre Variengien, Frédéric Alexandre

▶ To cite this version:

Xavier Hinaut, Subba Reddy Oota, Alexandre Variengien, Frédéric Alexandre. Modelling Cross-Situational Learning on Full Sentences in Few Shots with Simple RNNs. Proceedings of the 46th Annual Conference of the Cognitive Science Society, Jul 2024, Rotterdam, Netherlands. hal-04906122

HAL Id: hal-04906122 https://hal.science/hal-04906122v1

Submitted on 22 Jan 2025

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Proceedings of the Annual Meeting of the Cognitive Science Society

Title

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Permalink

https://escholarship.org/uc/item/0665s4vw

Journal

Proceedings of the Annual Meeting of the Cognitive Science Society, 46(0)

Authors

Hinaut, Xavier OOTA, SUBBA REDDY Variengien, Alexandre et al.

Publication Date

2024

Peer reviewed

Modelling Cross-Situational Learning on Full Sentences in Few Shots with Simple RNNs

Xavier Hinaut

Inria, Bordeaux, France

SUBBA REDDY OOTA

Inria, Bordeaux, France

Alexandre Variengien

Inria, Bordeaux, France

Frederic Alexandre

Inria-Research, Bordeaux, France, France

Abstract

How do children bootstrap language through noisy supervision? Most prior works focused on tracking co-occurrences between individual words and referents. We model cross-situational learning (CSL) at sentence level with few (1000) training examples. We compare reservoir computing (RC) and LSTMs on three datasets including complex robotic commands. For most experiments, reservoirs yield superior performance over LSTMs. Surprisingly, reservoirs demonstrate robust generalization when increasing vocabulary size: the error grows slowly. On the contrary, LSTMs are not robust: the number of hidden units needs to be dramatically increased to follow up vocabulary size increase, which is questionable from a biological or cognitive perspective. This suggests that that random projections used in RC helps to bootstrap generalization quickly. To our knowledge, this is a new result in developmental learning modelling. We analyse the evolution of internal representations during training of both recurrent networks and suggest why reservoir generalization seems more efficient.