



## **Distributional Semantics**

Advanced Machine Learning for NLP

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SLIDES ADAPTED FROM YOAV GOLDBERG AND OMER LEVY

### Beyond word2vec

- word2vec is factorizing a word-context matrix.
- The content of this matrix affects the resulting similarities.
- word2vec allows you to specify a window size.
- But what about other types of contexts?
- Example: dependency contexts (Levy and Dagan, ACL 2014)

## Bag of Words (BoW) Context

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# Syntactic Dependency Context

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## Embedding Similarity with Different Contexts

Target Word	Bag of Words (k=5)	Dependencies
	Dumbledore	Sunnydale
	hallows	Collinwood
Hogwarts	half-blood	Calarts
(Harry Potter's school)	Malfoy	Greendale
	Snape	Millfield

Related to Harry Potter

Schools

## Embedding Similarity with Different Contexts

Target Word	Bag of Words (k=5)	Dependencies
	nondeterministic	Pauling
	non-deterministic	Hotelling
Turing	computability	Heting
(computer scientist)	deterministic	Lessing
	finite-state	Hamming

Related to computability

**Scientists** 

## Embedding Similarity with Different Contexts

Target Word	Bag of Words (k=5)	Dependencies
	singing	singing
	dance	rapping
dancing	dances	breakdancing
(dance gerund)	dancers	miming
	tap-dancing	busking

Related to dance

Gerunds

**Online Demo!** 

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- context: user who wrote the message
- ...
- the sky is the limit

#### Summary

#### **Distributional Semantics**

- Words in similar contexts have similar meanings.
- Represent a word by the contexts it appears in.
- But what is a context?

### Neural Models (word2vec)

- Represent each word as dense, low-dimensional vector.
- Same intuitions as in distributional vector-space models.
- Efficient to run, scales well, modest memory requirement.
- Dense vectors are convenient to work with.
- Still helpful to think of the context types.