

Measuring Semantic Coherence of a Conversation

Svitlana Vakulenko, Maarten de Rijke,
Michael Cochez, Vadim Savenkov, Axel Polleres



Semantic coherence

- An essential property of a conversation, “**continuity** of senses”



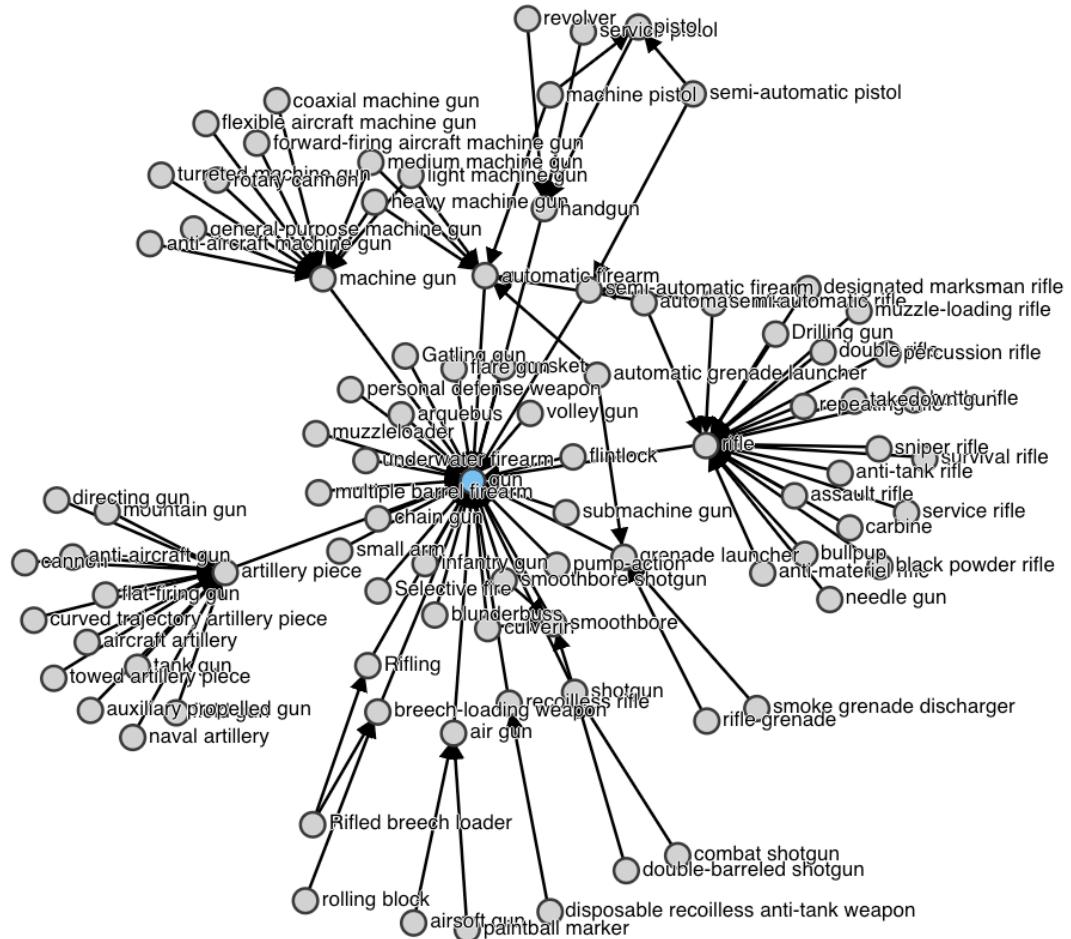
Research goal

- See if we can **detect** holes in conversations
- Evaluate existing **knowledge** models
- Propose an approach to **measure** these holes (incoherence)

- **Why:** dialogue system design, knowledge engineering

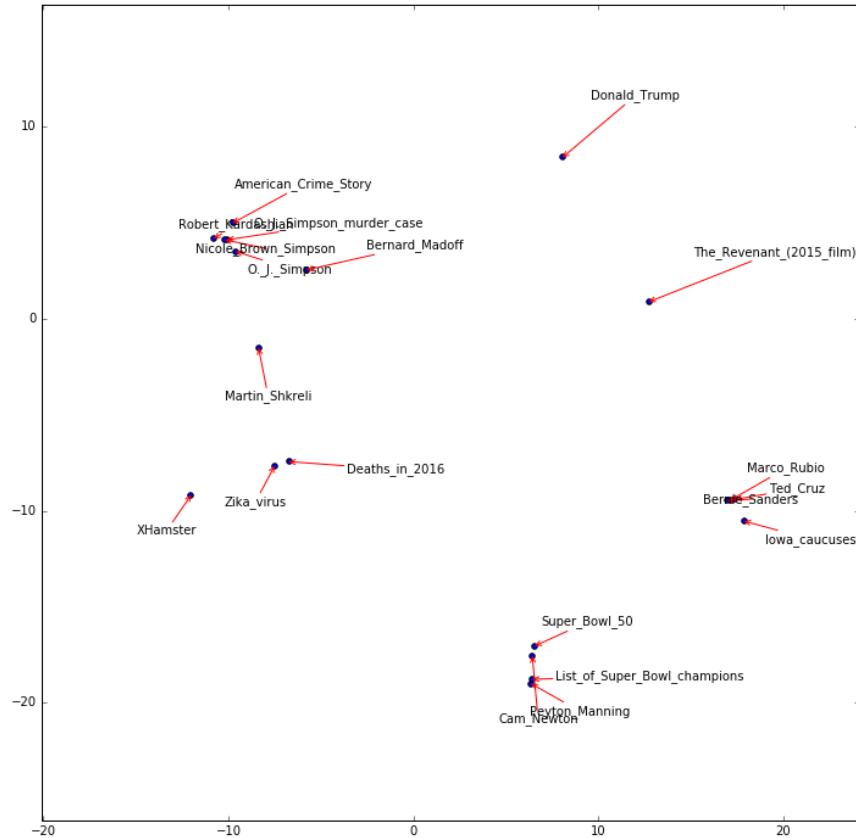
Semantic models

■ Knowledge Graphs



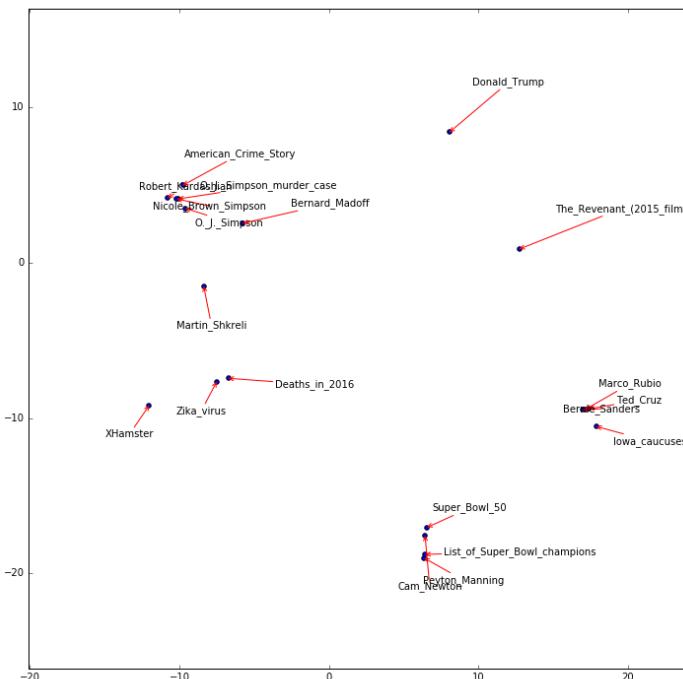
Semantic models

- Word embeddings

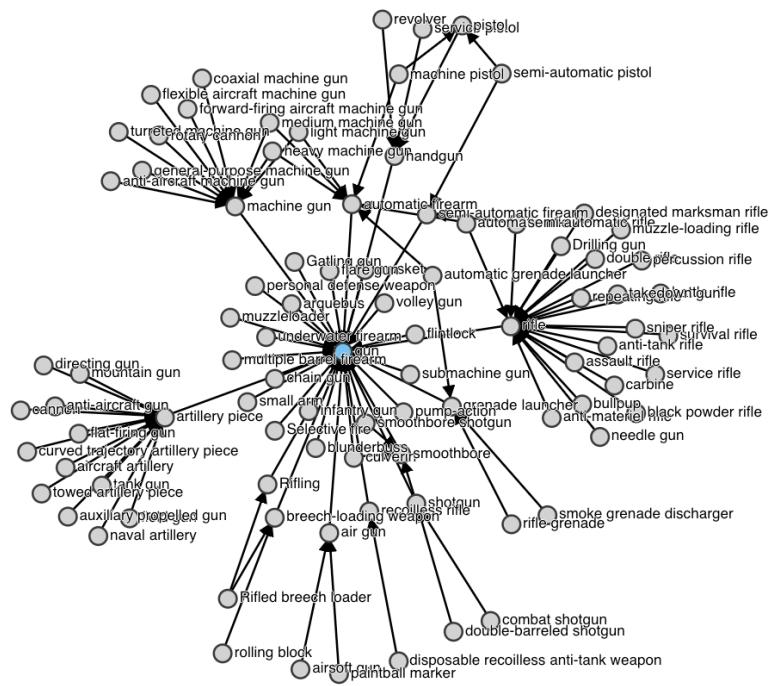


Semantic models

- Knowledge Graphs
- Word embeddings



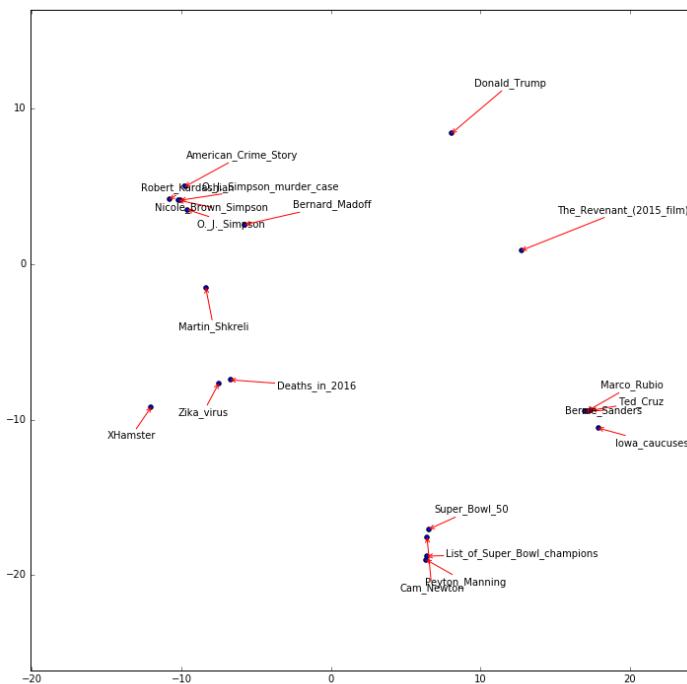
https://commons.wikimedia.org/wiki/File:2016_02_mini_embedding.png



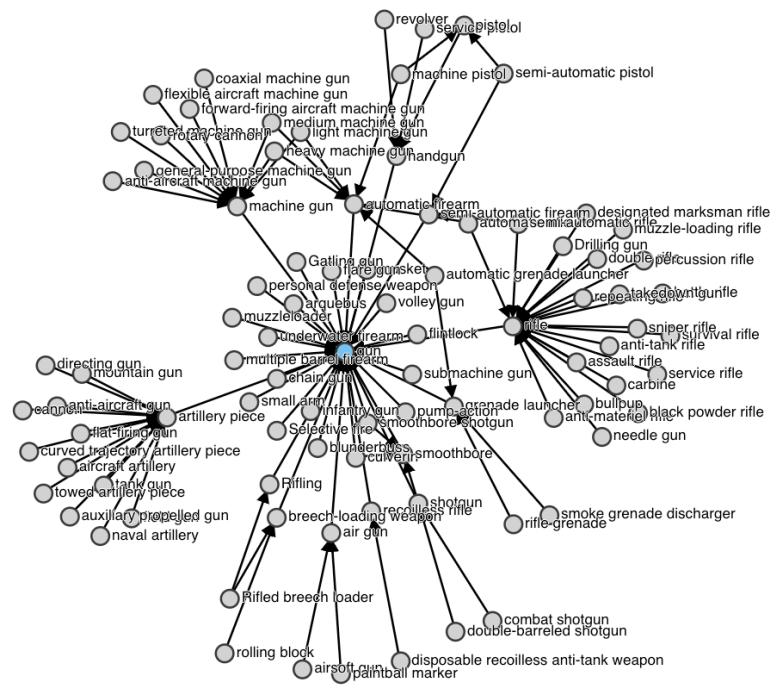
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Semantic models

- Knowledge Graphs
- Word embeddings
- Knowledge Graph embeddings



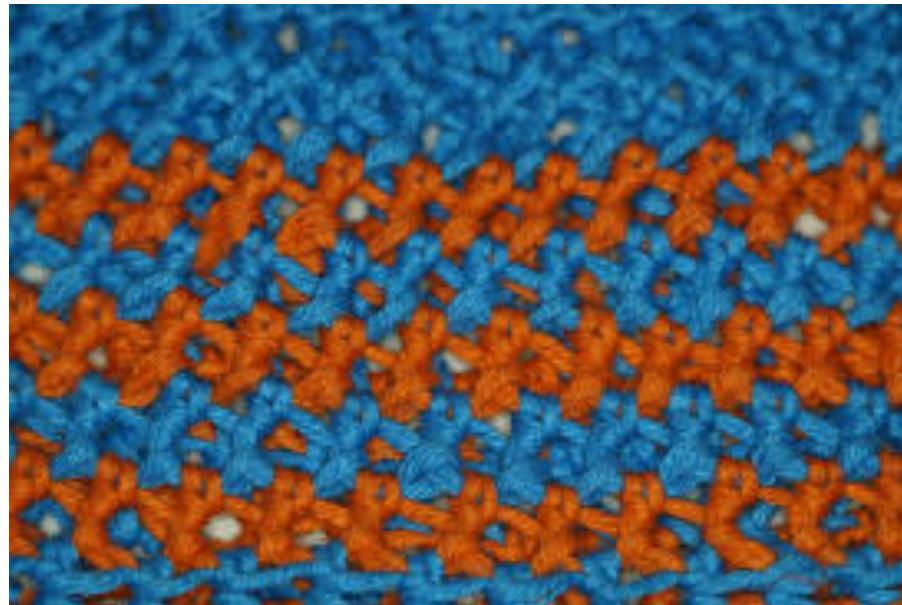
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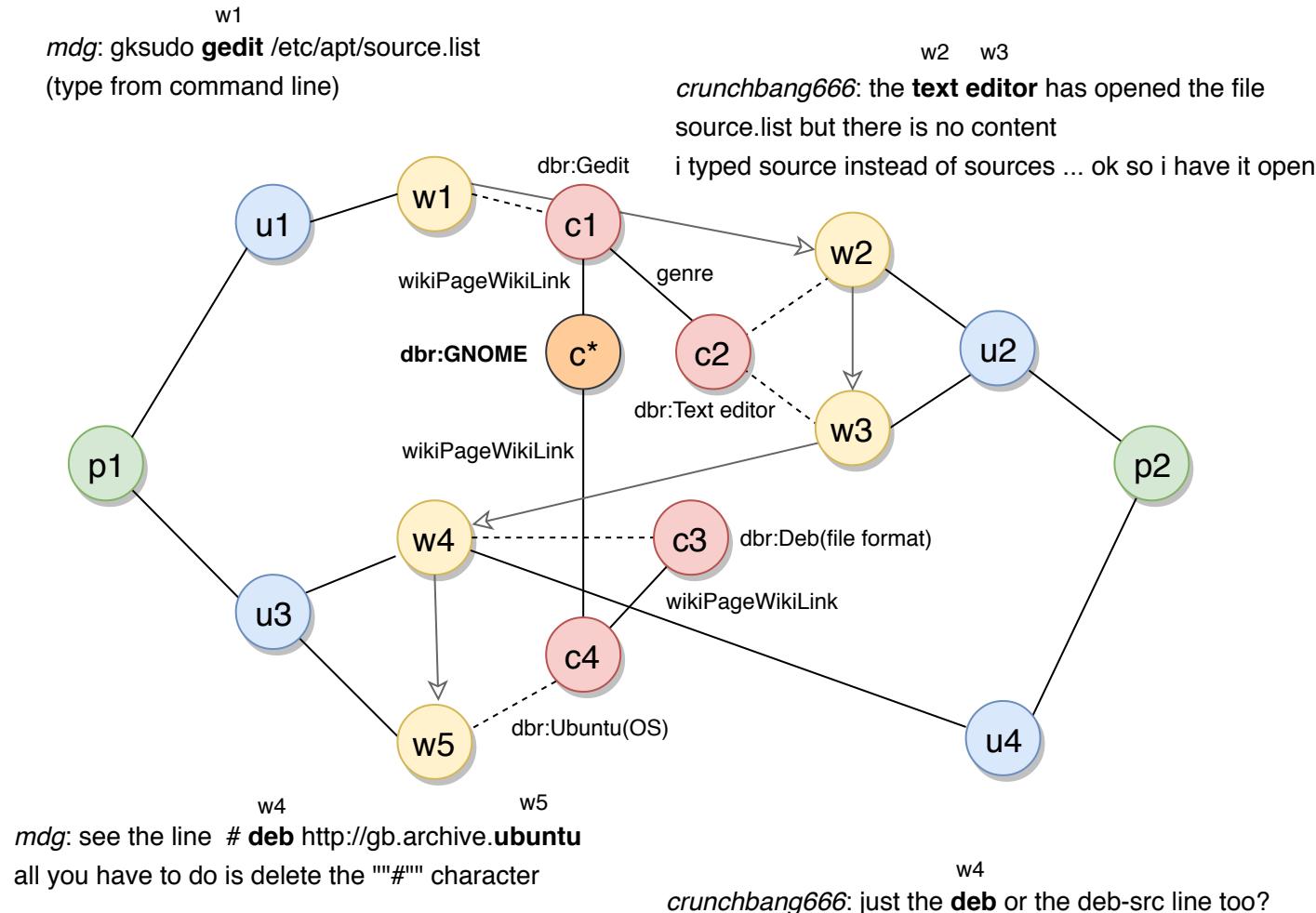
<https://commons.wikimedia.org/wiki/File:Wikidata-gun-ontology-2017-05-11.png>

Linking dialogue

- Take existing **knowledge** models
- See if we can **detect** holes in conversations through this models
- Propose an approach to **measure** these holes (incoherence)



Dialog graph



Experiments

- Ubuntu Dialogue Corpus
- DBpedia Spotlight API
- Knowledge Graphs: DBpedia+Wikidata HDT
- Knowledge Graph embeddings: rdf2vec, KGlove
- Word embeddings: word2vec, Glove



<https://github.com/rkadlec/ubuntu-ranking-dataset-creator>

<https://en.wikipedia.org/wiki/File:DBpediaSpotlight.jpg>

<https://en.wikipedia.org/wiki/Wikidata>

Subgraph induction



top-k shortest path

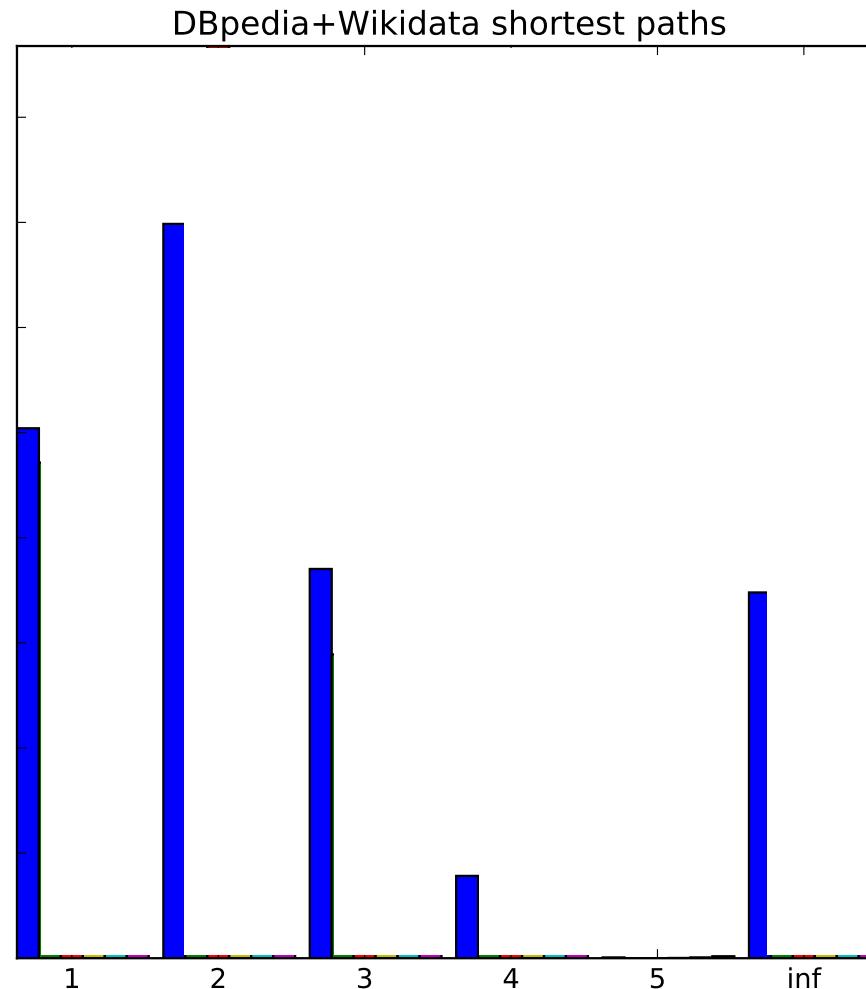
```
PREFIX ppf: <java:at.ac.wu.arqext.path.>
PREFIX dbr: <http://dbpedia.org/resource/>
SELECT * WHERE {
?X ppf:topk ("--source" dbr:Directory_service
              dbr:Gnome dbr:GNOME
              dbr:Desktop_environment
              "--target" dbr:Desktop_computer
              "--k" 5 "--maxlength" 9 "--timeout" 2000) }
```

Subgraph statistics

Table 1. The top 5 most common entities and relations in the Ubuntu dialogue dataset: mentioned entities – from linking dialogue utterances to DBpedia entities via Dbpedia Spotlight Web service; context entities and relations – from the shortest paths between the mentioned entities in DBpedia.

Top	Mentioned entities		Context entities		Relations	
	#	Label	Count	Label	Count	Label
1	Ubuntu(philosophy)	1605	Ubuntu(OS)	1058	wikiPageWikiLink	51014
2	Sudo	708	Linux	725	gold/hypernym	319
3	Booting	676	Microsoft_Windows	208	ontology/genre	178
4	APT(Debian)	405	FreeBSD	175	operatingSystem	140
5	Live_CD	314	Smartphone	171	rdf-schema#seeAlso	116

Shortest paths



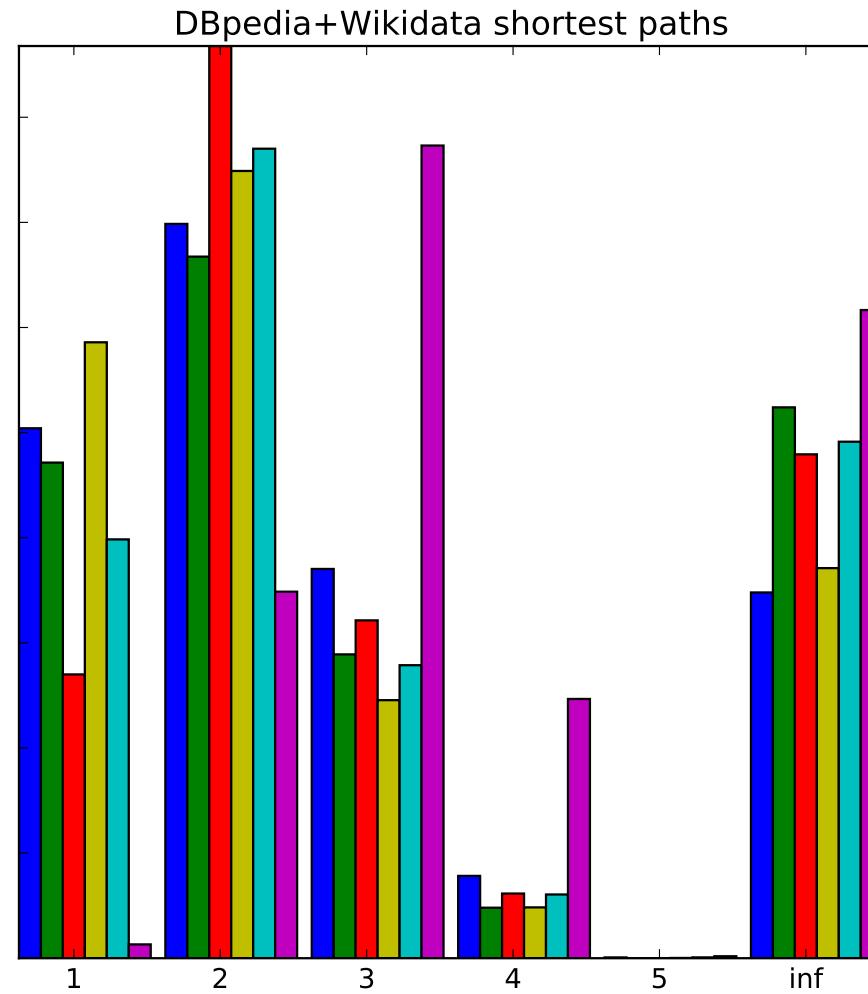
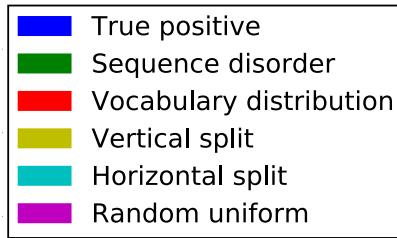
Negative sampling

- random uniform (RUF)
- vocabulary distribution (VoD)
- sequence disorder (SqD)
- horizontal split (HSp)
- vertical split (VSp)

Ubuntu
JFS
ext3
power
outage
fsck
ext2
partition
boot
dunno
OK

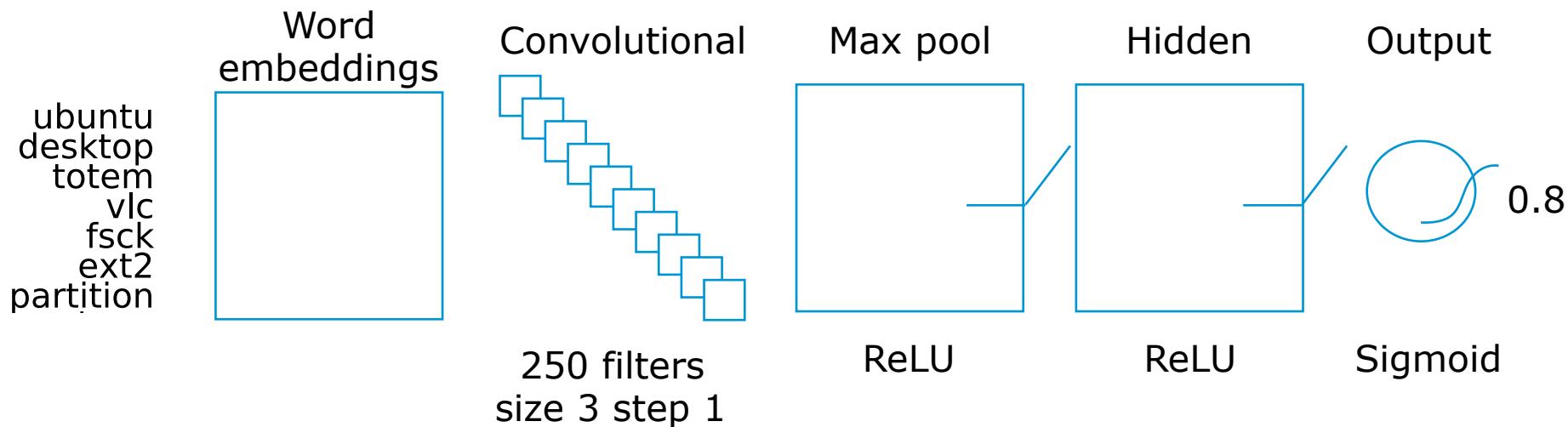
ubuntu
desktop
totem
vlc
fsck
ext2
partition
boot
dunno
OK

Shortest paths



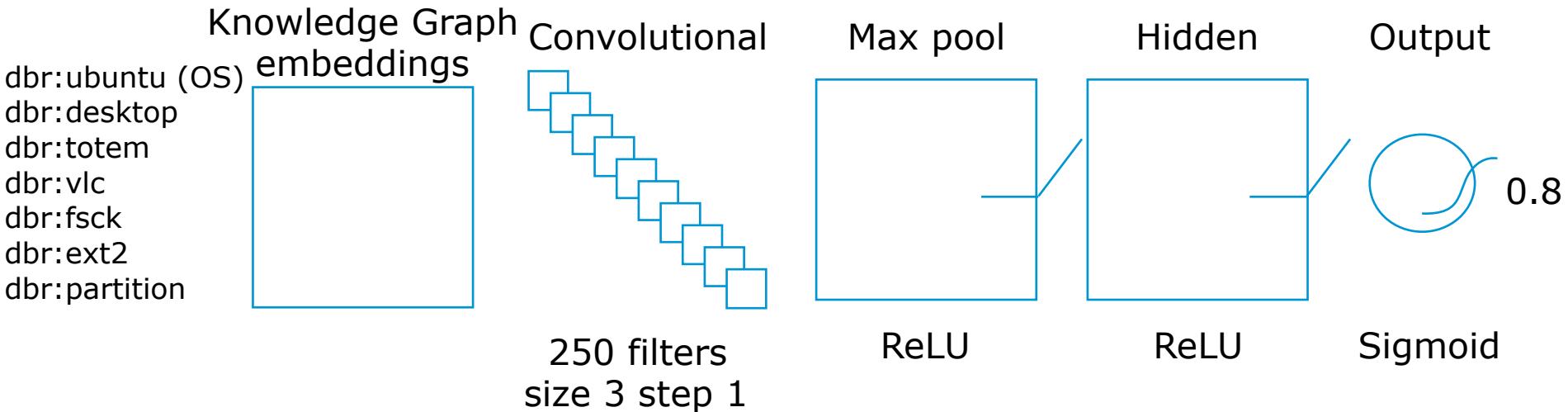
Binary classification

- Convolutional Neural Network (CNN)
- Input: sequence of words/entities
- Output: coherence score [0;1]



Binary classification

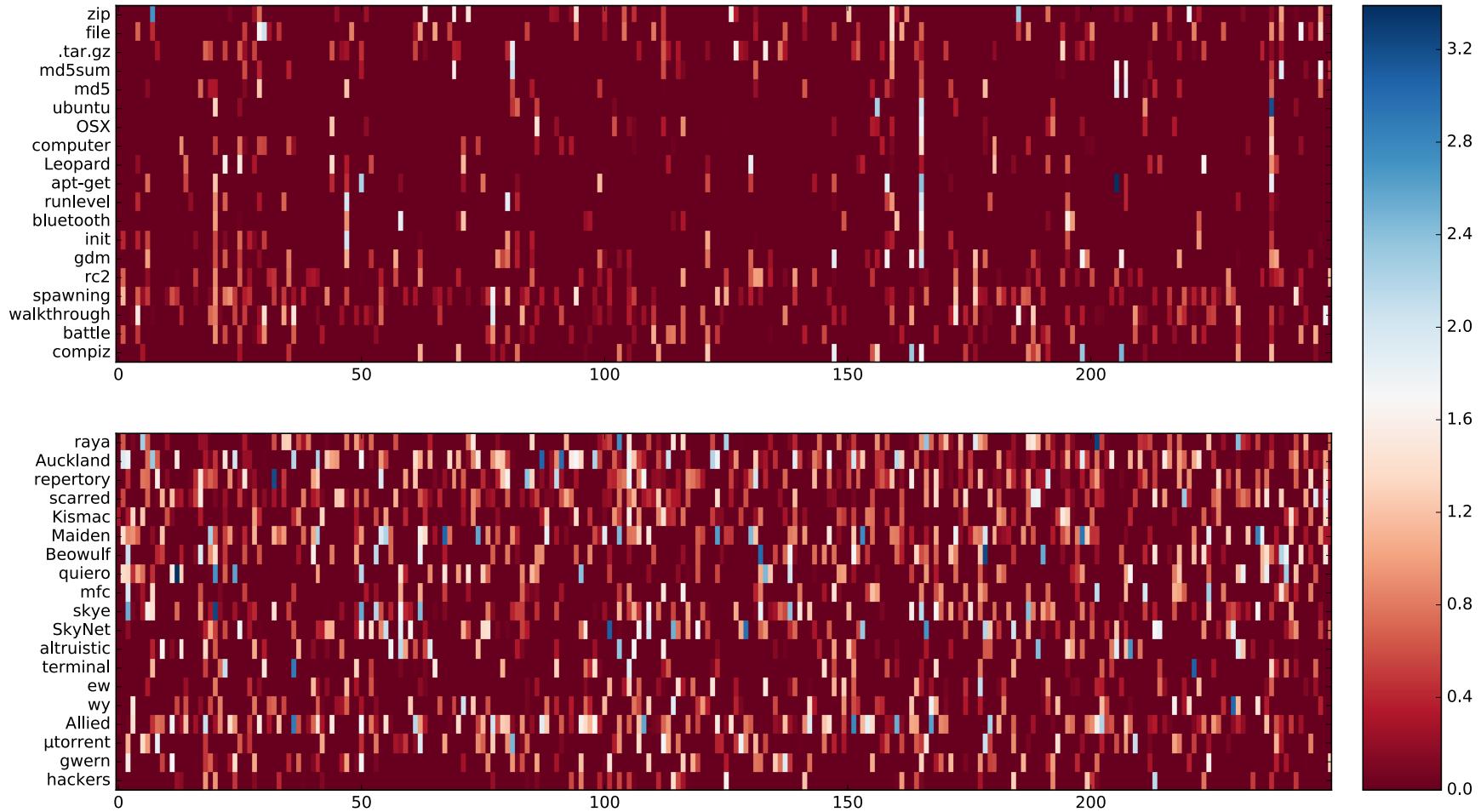
- Convolutional Neural Network (CNN)
- Input: sequence of words/entities
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Results

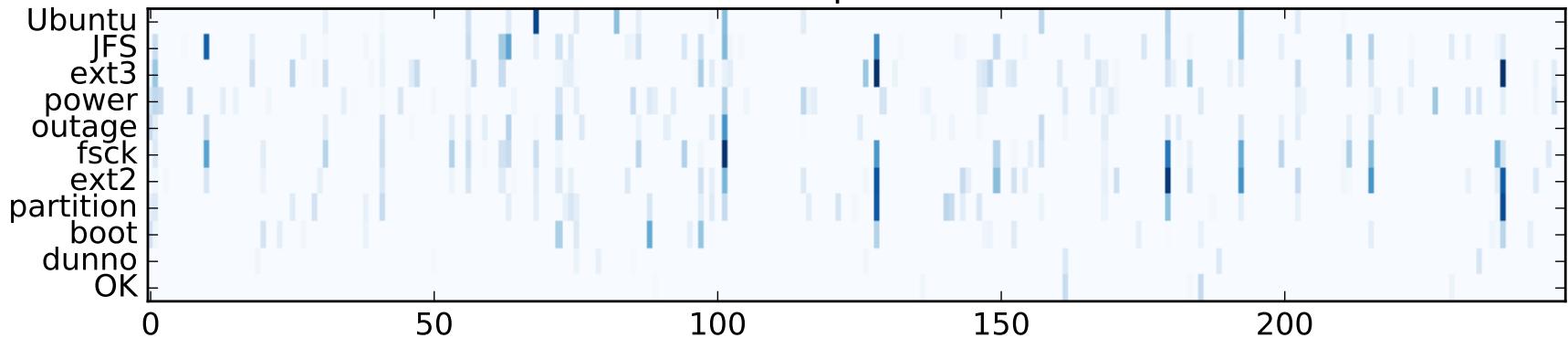
Embeddings	Data split	Accuracy											
		TPos	TNeg										Avg
			RUF	Avg	VoD	Avg	SqD	Avg	VSp	Avg	HSp	Avg	
Word2Vec	RUF	0.99	0.99	0.99	0.02	0.50	0.02	0.50	0.01	0.50	0.01	0.50	0.60
	VoD	0.89	0.62	0.75	0.90	0.89	0.53	0.71	0.18	0.54	0.20	0.54	0.69
	SqD	0.75	0.65	0.70	0.88	0.81	0.81	0.78	0.27	0.51	0.29	0.52	0.66
	VSp	0.59	0.50	0.55	0.82	0.71	0.41	0.50	0.59	0.59	0.61	0.60	0.59
	HSp	0.62	0.39	0.50	0.71	0.66	0.38	0.50	0.55	0.58	0.63	0.63	0.58
GloVe	RUF	0.99	0.99	0.99	0.00	0.50	0.01	0.50	0.00	0.50	0.00	0.50	0.60
	VoD	0.93	0.38	0.66	0.93	0.93	0.39	0.66	0.19	0.56	0.08	0.51	0.66
	SqD	0.76	0.71	0.73	0.91	0.84	0.82	0.79	0.16	0.46	0.15	0.45	0.66
	VSp	0.60	0.25	0.42	0.92	0.76	0.43	0.51	0.65	0.62	0.66	0.63	0.59
	HSp	0.71	0.34	0.52	0.81	0.76	0.30	0.50	0.55	0.63	0.66	0.68	0.62
rdf2vec PRS	RUF	0.98	0.99	0.99	0.02	0.50	0.02	0.50	0.02	0.50	0.01	0.50	0.60
	VoD	0.79	0.68	0.73	0.83	0.81	0.34	0.57	0.36	0.57	0.35	0.57	0.65
	SqD	0.59	0.48	0.54	0.72	0.66	0.67	0.63	0.43	0.51	0.40	0.50	0.56
rdf2vec PR	HSp	0.57	0.59	0.58	0.72	0.64	0.43	0.50	0.59	0.58	0.67	0.62	0.58
	RUF	0.92	0.97	0.94	0.11	0.51	0.09	0.50	0.08	0.50	0.07	0.50	0.59
KGloVe Uni	VoD	0.54	0.88	0.71	0.73	0.64	0.61	0.58	0.51	0.52	0.52	0.53	0.60
	SqD	0.55	0.62	0.58	0.64	0.59	0.63	0.59	0.47	0.51	0.45	0.50	0.56
KGloVe PrO	HSp	0.31	0.81	0.56	0.75	0.53	0.69	0.50	0.77	0.54	0.70	0.51	0.53
KGloVe PR	HSp	0.47	0.69	0.58	0.61	0.54	0.54	0.50	0.57	0.52	0.65	0.56	0.54

Random uniform

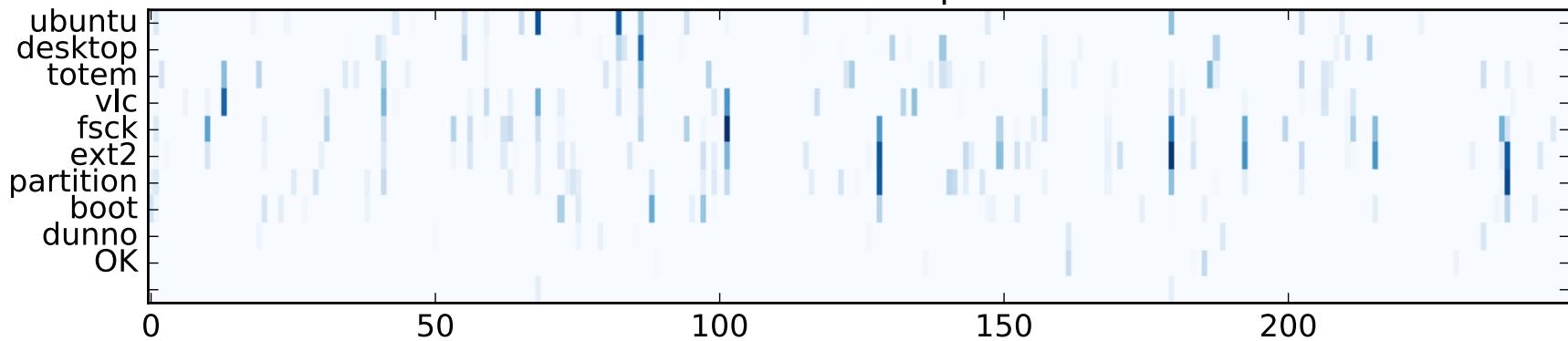


Horizontal split

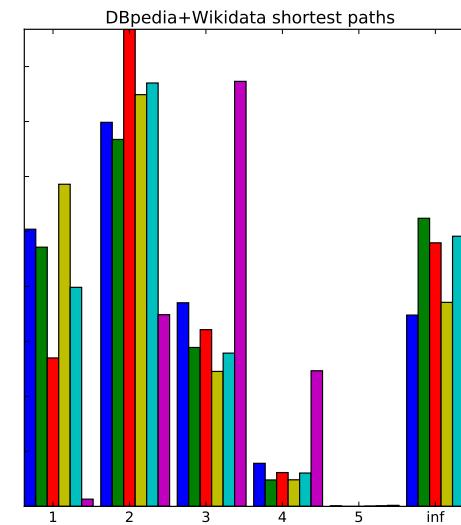
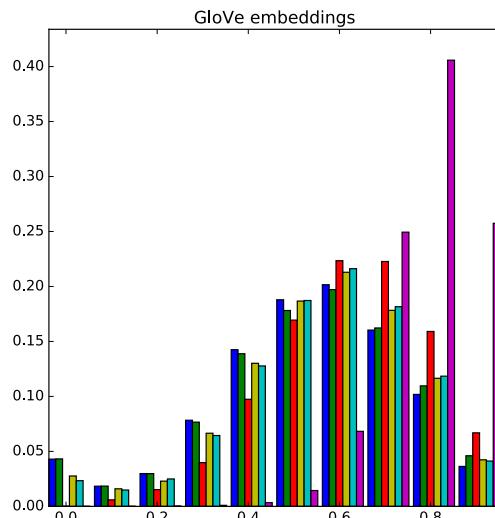
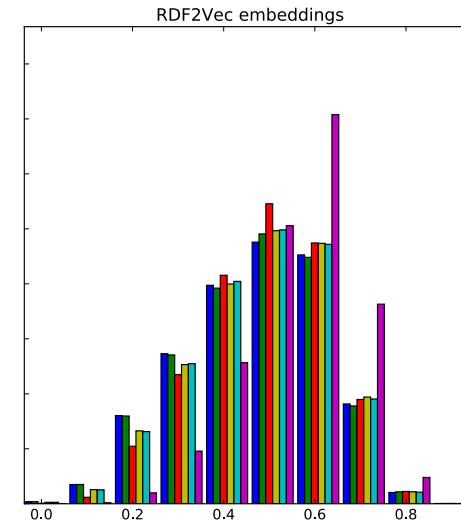
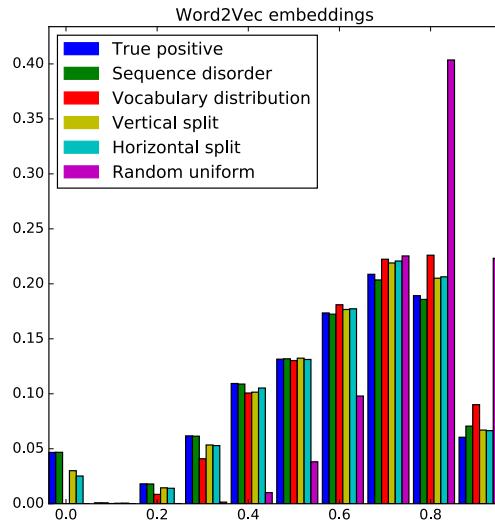
True positive



Horizontal split

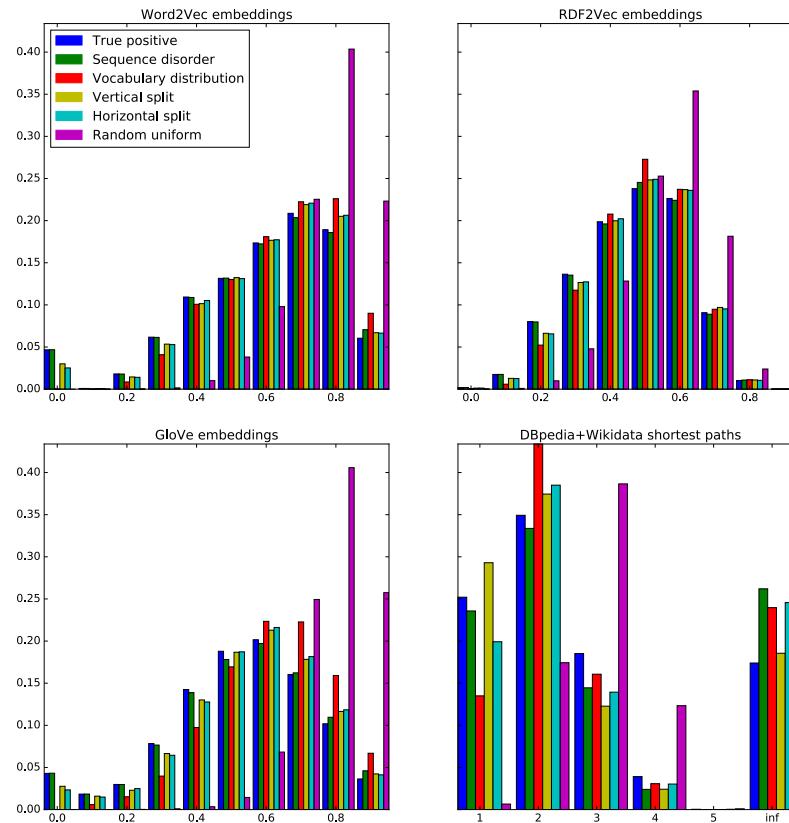


Semantic spaces



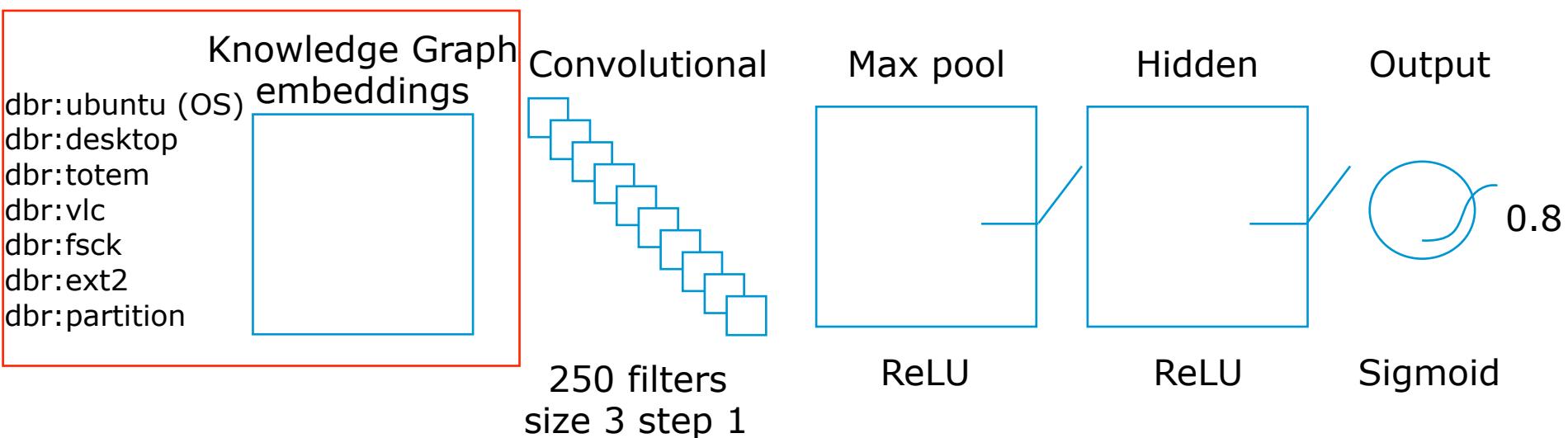
Conclusions and future work

- GloVe **word embeddings** show best performance
 - integrating **heterogenous** knowledge sources



Conclusions and future work

- **NEL** is a bottleneck for KG embeddings
 - **End-to-end** training (NEL NN-layer)



Conclusions and future work

- **Dialog graph embeddings**

