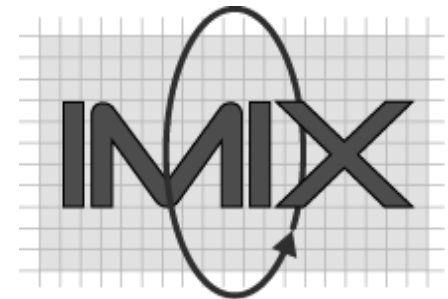


# Syntactic Contexts for Finding Semantically Similar Words



CLIN  
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# Motivation(1)

## Overview

Motivation

Distributional similarity

Similarity Measures

Data

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Results

Conclusion

- Extending Dutch EuroWordNet for:
  - Question Classification  
Welke tennisser ...? > person ques.
  - Answering 'which' questions  
Welk beroep heeft Renzo Piano?  
De **Italiaan** Renzo Piano is **architect**.



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## Motivation(2)

- Disambiguation

- Coordination:

(Chirac uit Frankrijk) en Blair.

Chirac uit (Frankrijk en Blair).

- PP-attachment

Hij at (mie) met stokjes.

Hij at (mie met stokjes).



# Distributional Similarity

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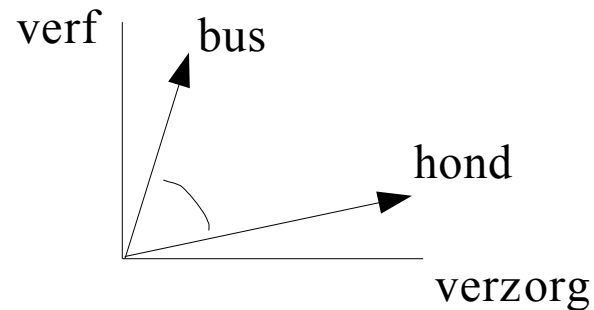
- Similar words share similar contexts
- Context = **Syntactic** context

	zie	verf	verzorg	laat_uit
bus	50	5	1	0
hond	56	1	5	8

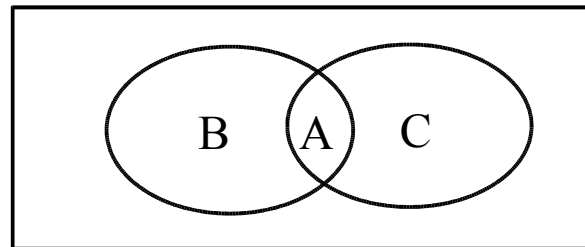


# Similarity Measures

- Cosine



- Dice



$$\frac{2A}{2A+B+C}$$

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# Weights

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- Mutual Information

- T-Test

	hebben	verdoen	geven	doordrijven
zin	500	0	400	18
tijd	560	10	600	0



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## Data

- 78 million words of parsed Dutch newspaper text

- Subject-Verb      `kat eet`
- Verb-Object      `voer kat`
- Adjective-Noun      `langharige kat`
- Coordination      `Bassie en Adriaan`
- Apposition      `de clown Bassie`
- PC      `begin_met werk`



# Extracted from data

Gram rel.	# types	# tuples
Coord	2.465.098	965.296
Subj	2.122.107	5.639.140
Adj	1.040.785	3.262.403
Obj	993.913	2.642.356
Appo	526.337	602.970
PC	389.139	770.631

Tuples, triples etc

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	ga_subj	geel_adj	neem_obj	Lassie_app
bus	4	9	8	0
hond	4	1	6	8

Cutoff:  
row > 10





# Example Output

(for combi obj/adj using MI and Dice)

## **Trainingspak**

bloes

tricot

tshirt

hemd

shirt

tenue

winterjas

bretel

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# Example Output

(for subject using TTest and Cosine)

## **Aanduiding**

quorum

skilruik

term

dashboardkastje

raambediening

Pep

godfather

gelijkwaardigheid

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# Example Output

(for coord. rel. using MI and Dice)

**Wim Kok**

Elco Brinkman

Thijs Wöltgens

Hans van Mierlo

Frits Bolkestein

Jacques Wallage

Relus ter Beek

Hedy d'Ancona

Felix Rottenberg

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**RUG**

**University of Groningen**



# Evaluation Framework

- Dutch EWN as gold standard
  - Advantages
    - No need for human evaluators, nor application
  - Disadvantages:
    - EWN contains no proper names
    - More than one way to measure distance in WordNet

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# Evaluation Framework

- 1000 words from EWN (random, freq > 10)

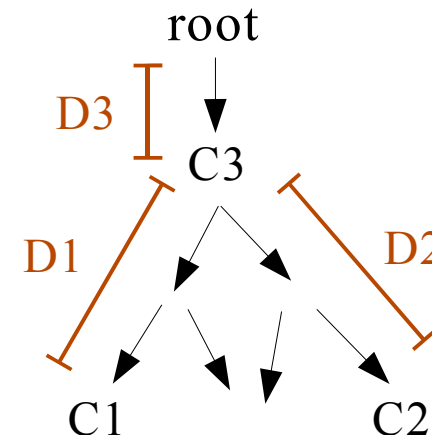
For each word

→ collect 100 most similar words

For each pair of words

→ measure similarity in EWN with Wu & Palmer's measure

$$\frac{2(D3)}{D1 + D2 + 2(D3)}$$



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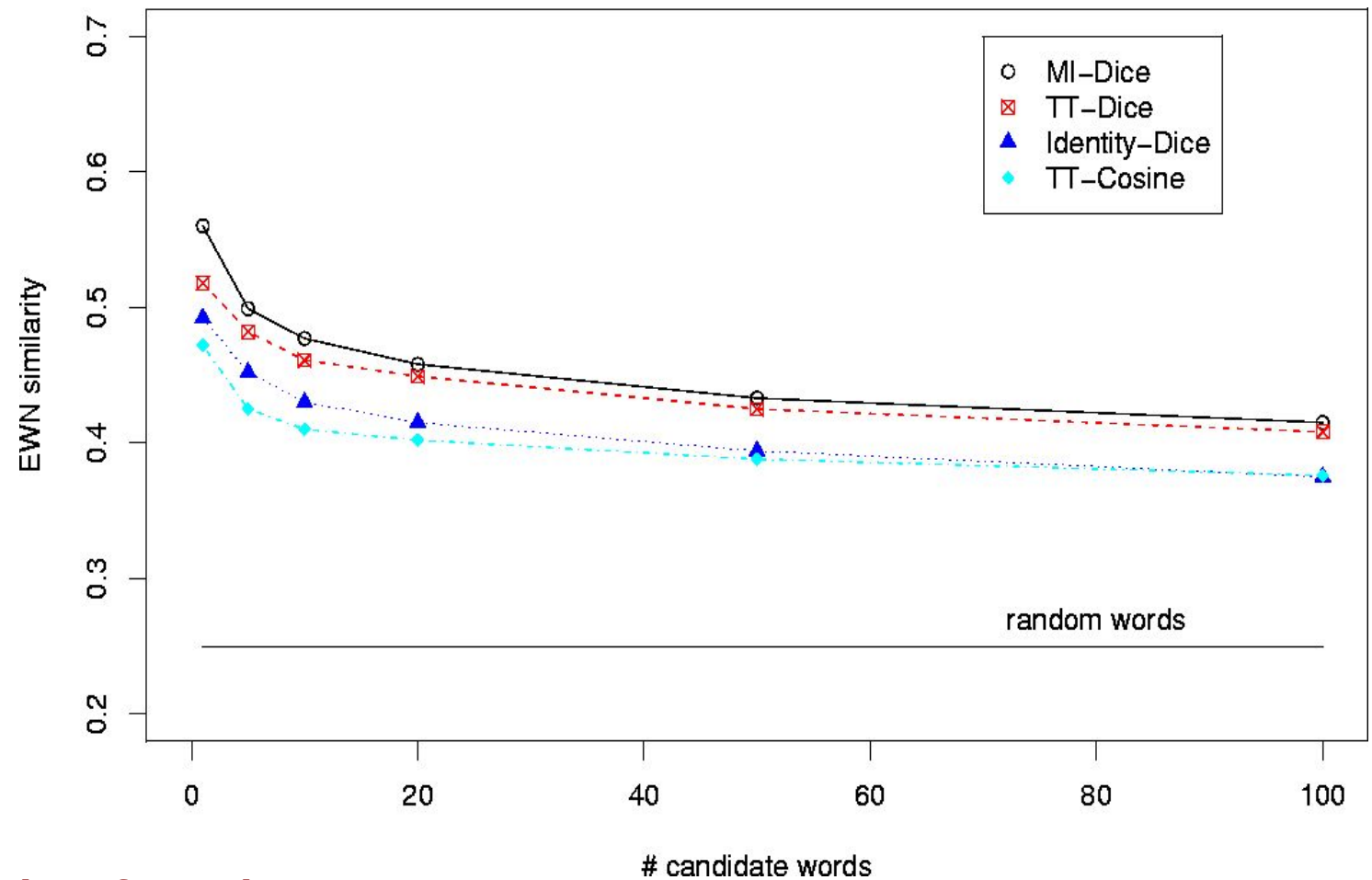
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## Results (object relation)





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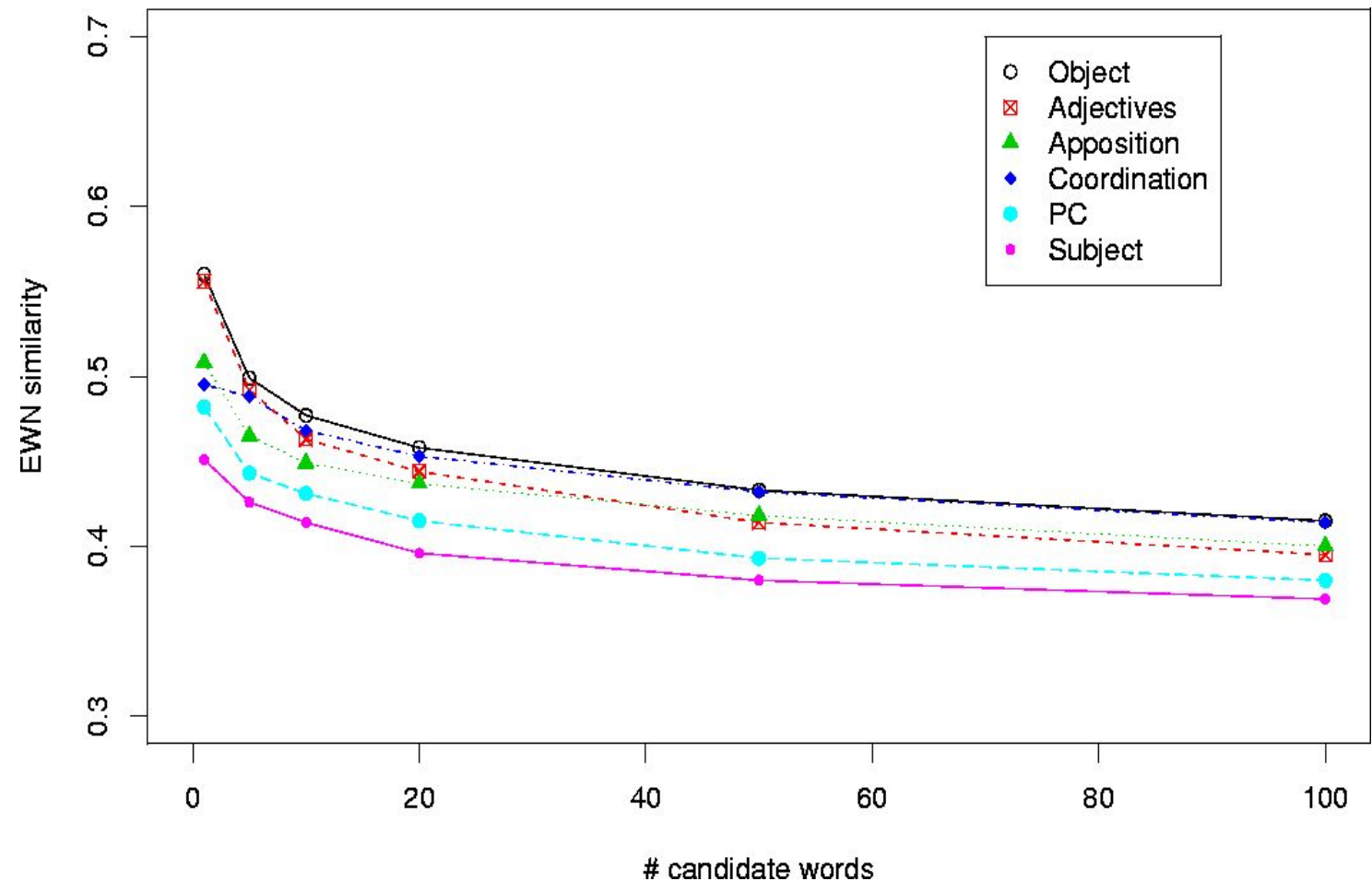
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## Results (diff. gram rels Dice MI)





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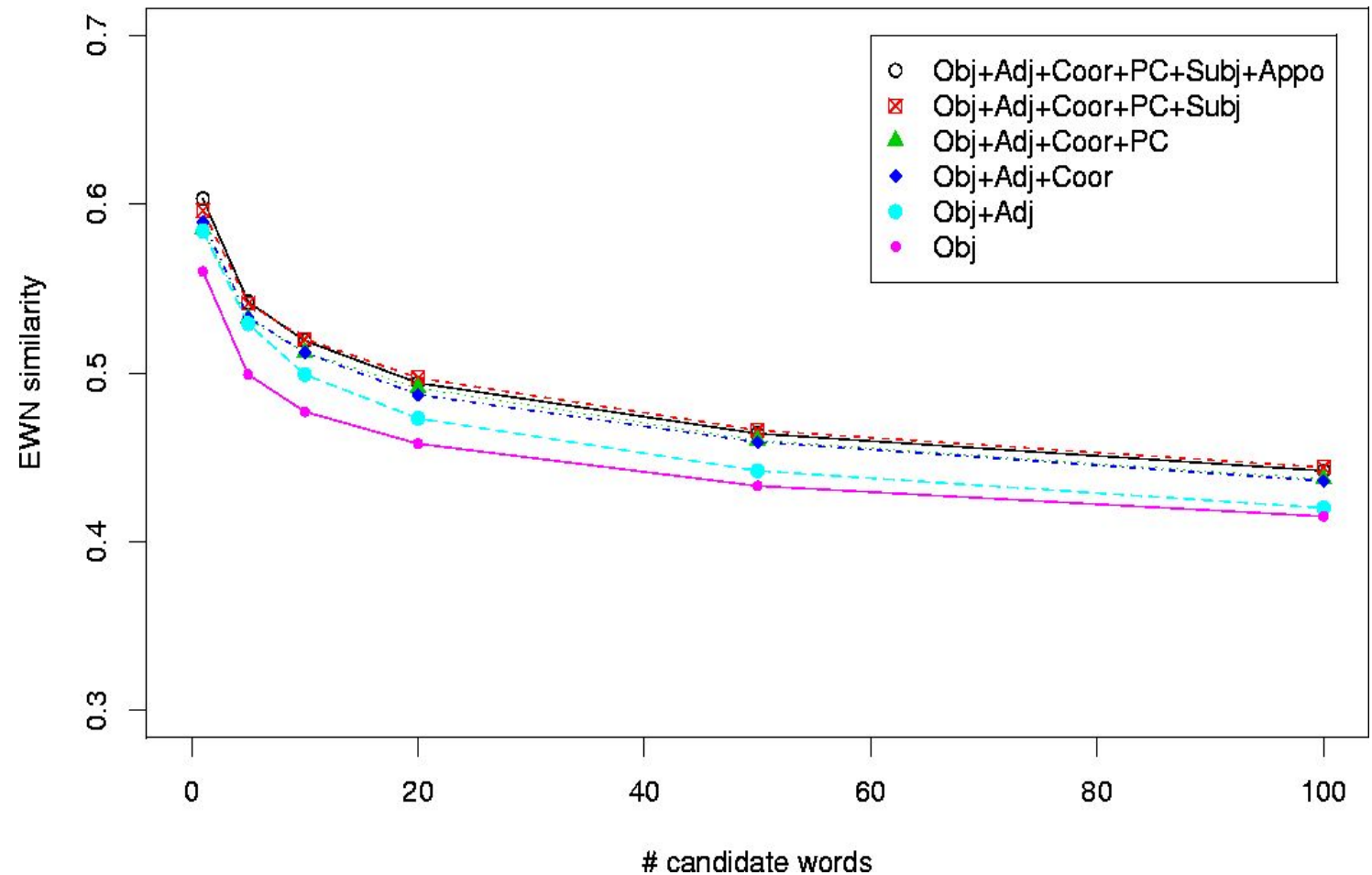
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## Results (combining gramrels)







# Recall

Gram rel	Recall (%)
Appo	11,2
PC	29,8
Obj	47,8
Adj	50,3
Coord	56,0
Subj	57,9

Gram rel	Recall (%)
O+A	62,3
O+A+C	72,9
O+A+C+P	74,5
O+A+C+P+S	78,8
All	78,9

- Recall is number of words found in our matrix
- On average 40 % of target's most similar words are not found in EWN

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## Conclusions

- Combination of Dice and MI gives best results.
- Object seems most suitable grammatical relation for this task, although evaluating on EWN is not fair for Coord + Appo
- Combining grammatical relations improves results.
- With 40 % of words missing in EWN room for improvement



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## Future Work

- From horizontal to vertical relations (hypernym structures) using
  - Clustering techniques
  - Patterns from encyclopaedia
  - Better similarity measures
- LSA, LDA to compress feature space, allow for generalization
- Get more insight into behaviour of different grammatical relations