



# Automatic animacy classification for Dutch

Jelke Bloem, Gosse Bouma



## Noun animacy

- › Sentience of the referent

sister – participant – carpenter – dude – northener

cat – angel – dragon – bacteria

oak – robot – community – government

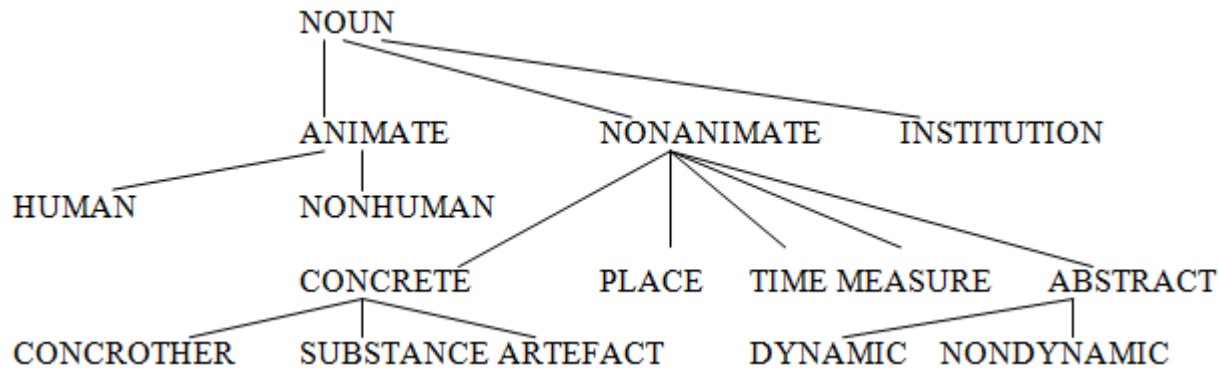
fountain – second – observation – music – fog

- › Animacy hierarchy



# Animacy hierarchy

- › Humans (OConnor et al., 2004)
- › Other animates  
ORGANIZATIONS, ANIMALS, INTELLIGENT MACHINES, VEHICLES.
- › Inanimates  
CONCRETE INANIMATE, NON-CONCRETE INANIMATE, PLACE, TIME



- › HUMAN > NONHUMAN > NONANIMATE



## Animacy and grammaticality

- › The spoon **which** is on the table is mine.
  - › \* **The man which** is sitting on the table is my friend.
  - › \* The spoon **who** is on the table is mine.
  - › **The man who** is sitting on the table is my friend.
- who** refers to **ANIMATE**, **which** refers to **INANIMATE**
- › Cut-off point



## Animacy and sentence processing

- › Dative alternation

(Bresnan et al. 2007)

She gave a push to the car. (prepositional dative)

She gave the car a push. (double object)

She gave a toy to **the child**. (prepositional dative)

She gave **the child** a toy. (double object)

- › For inanimate recipients, the double object construction is used more often



# Automatic animacy classification goals

- › Corpus annotation
- › Use in language technology
  - e.g. Automatic translation:  
*De man die op de tafel zat*  
**die** = that, which, **who**, those, these  
**The man who** sat on the table
  - Anaphora resolution (Orasan and Evans, 2007)  
The tree fell on **the man**. **He** survived.
  - Better parsing (Øvrelid, 2009)



# Animacy in natural language processing

- › Few animacy resources are available (Zaenen, 2004)
- › Therefore, few tools make use of animacy
- › A few animacy classifiers were made, none for Dutch
- › Dutch resources:
  - Cornetto (lexical-semantic database)
  - Lassy Large (automatically annotated corpus),  
1.5 billion words



## Animacy classification task

- › For any noun, decide whether it refers to a human, nonhuman animate or inanimate entity
- › Classification features
  - World knowledge?
  - Morphology?
  - Context?





# Animacy classification task

- › World knowledge (Orasan and Evans, 2007)
  - Lexical-semantic database (WordNet)  
**poet -> writer -> communicator -> person**  
wikipedian -> ???
  
- › Morphology (Baker and Brew, 2008)
  - 诗 人
  - poetry **person**
  - or: Case marking



## Context features

(Øvrelid, 2009)

- › Animates prefer the agent role & subject position
- › Inanimates prefer the patient role & object position
- › Genitive case
  - das Haus meines **Vaters**
- › Reflexive
  - **The teacher** hurt **himself**



## Context features

- › Lexical association features: Verbs

The **doctor** **thought** John was right.

The banana **thought** John was right.

- › Adjectives

The **lazy** **thief**.

The **lazy** hurricane.



## Animacy: Data

### › Word lemmas and their animacy from Cornetto

```
<noun animacy="nonanimate">gevoel</noun>  
<noun animacy="nonanimate">IJsselmeer</noun>  
<noun animacy="nonanimate">noord</noun>  
<noun animacy="nonanimate">paasei</noun>  
<noun animacy="human">doctor</noun>  
<noun animacy="human">Engelsman</noun>  
<noun animacy="human">roker</noun>  
<noun animacy="human">symfonieorkest</noun>  
<noun animacy="nonhuman">fuchsia</noun>  
<noun animacy="nonhuman">pony</noun>  
<noun animacy="nonhuman">yeti</noun>
```

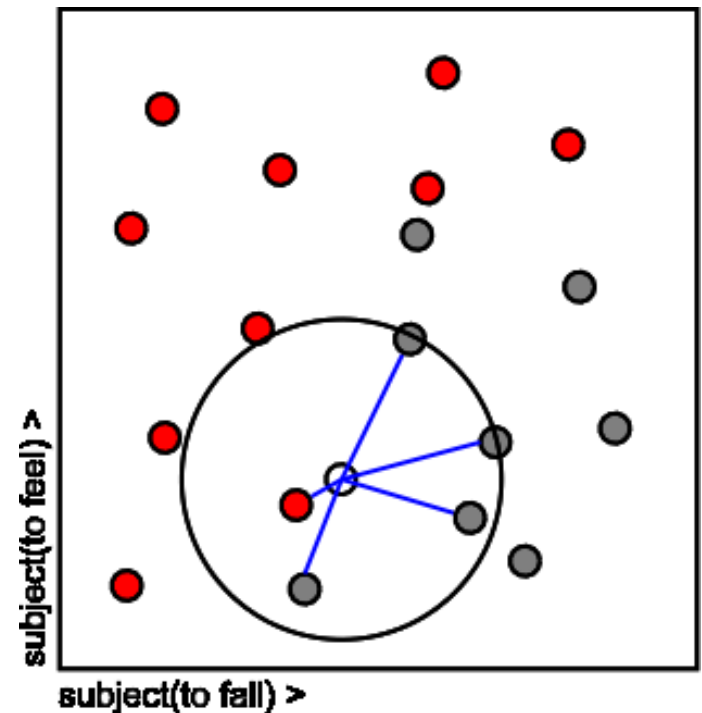
### › Verb-argument relations from Lassy Large corpus

```
85#blijf|intransitive|su#gevoel  
298#ontsta|intransitive|su#gevoel  
1#schrijf|transitive|su#gevoel  
8#rest|intransitive|su#gevoel  
7#ontdek|transitive|su#Engelsman  
4#ontwerp|transitive|su#Engelsman  
3#overschat|transitive|su#Engelsman
```



# Classification procedure

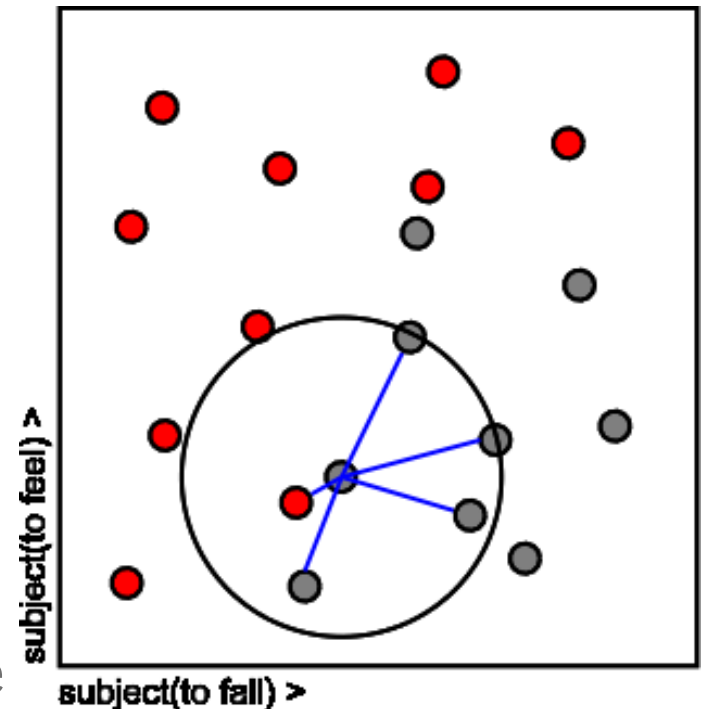
- › K-nearest neighbor (TiMBL)
- › Each noun is a feature vector
- › Classify new instances based on most similar (nearest) noun in multidimensional feature space





## Classification procedure

- › K-nearest neighbor (TiMBL)
- › Each noun is a feature vector
- › Classify new instances based on most similar (nearest) noun in multidimensional feature space
- › 4 of 5 neighbors are inanimate
  - The inanimate class is assigned





## Feature values: Association strength

- › Noun-verb association

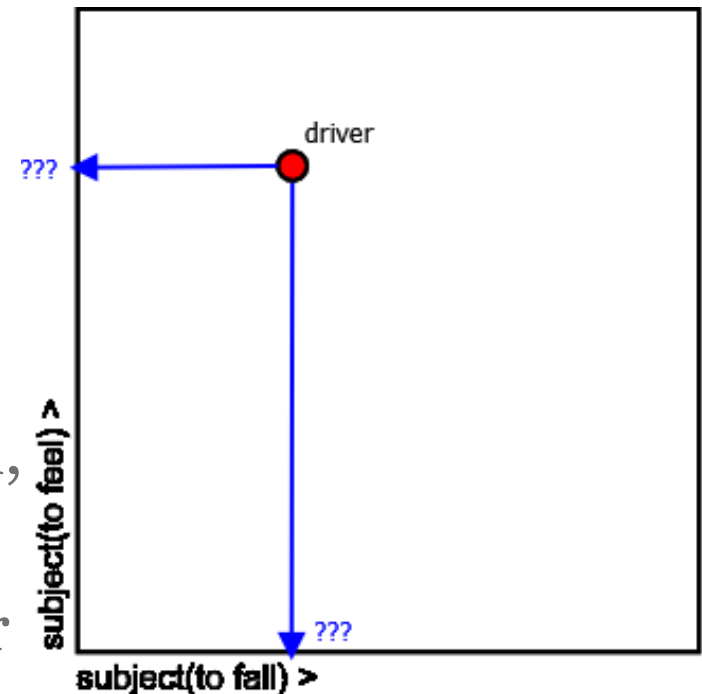
Subj (*ben*) : bestuurder **125**

Subj (*rij*) : bestuurder **12**

- › Which is more interesting?

- › Pointwise Mutual Information,  
Fisher's Exact Test

Feature-noun pairs that co-occur  
more often than would be  
expected by chance





# Association strength

“gevoel” (*feeling*, inanimate) subject relation strength (Fisher’s)

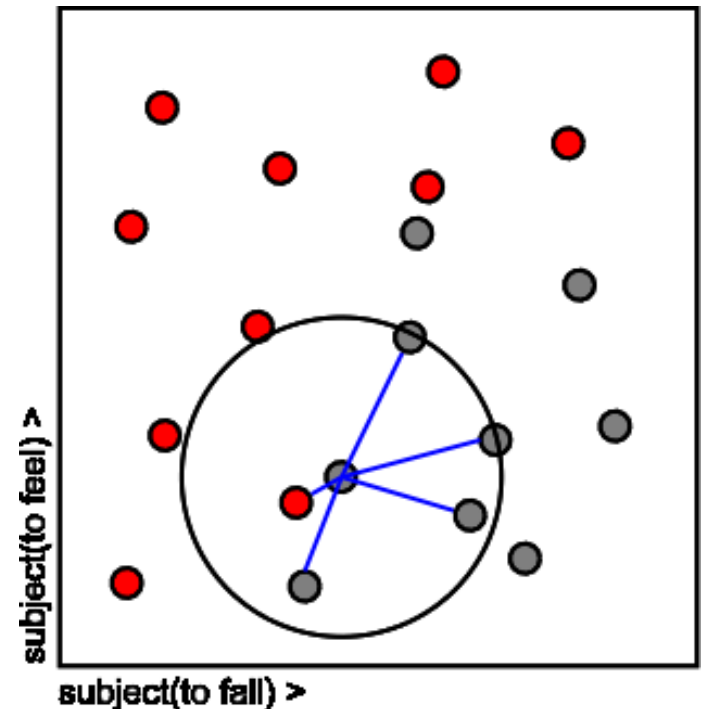
0.0000000000000000	ontsta	<i>arise</i>
0.0000000000000830	heb	<i>have</i>
0.0000000000002380	speel	<i>play</i>
0.0000000000501125	ben	<i>be</i>
0.000000003404273	zeg	<i>say</i>
0.731409478841741	krijg	<i>get</i>
0.823487761949459	spreek	<i>speak</i>
0.853510038160385	neem	<i>take</i>
0.902189553992116	ken	<i>know</i>
1.0000000000002866	schrijf	<i>write</i>





## Classification procedure

- › K-nearest neighbor (TiMBL)
- › Feature values are association scores
- › Evaluate by classifying unseen nouns according to these features





## Results: Features

Features	Accuracy
Baseline	80.92%
Object/subject ratio	81.09%
Verb subject relations	91.06%
Verb object relations	91.20%
Adjective relations	88.91%
<b>Subj+Obj+Adj</b>	<b>93.34%</b>

Baseline: Classify everything as the majority class

Ten-fold cross validation accuracy scores



## Noun frequency

- › Classifying low-frequency nouns is generally more difficult

Frequency cutoff	Baseline	Accuracy	Number of nouns
>0	76.68%	83.39%	30.950
>1	78.16%	90.27%	16.454
>10	80.92%	<b>93.34%</b>	12.168
>100	84.00%	91.22%	6.276
>1000	88.99%	88.62%	1.671



## Results: Class confusion

- › Classification errors

<b>Predicted -&gt;</b>	<b>Human</b>	<b>Nonhuman</b>	<b>Nonanimate</b>
<b>Human</b>	151	0	24
<b>Nonhuman</b>	0	2	53
<b>Nonanimate</b>	1	3	982

- › NONHUMAN class is only chosen correctly twice!



## Results: Two-way classification

### › Human/Nonhuman

Features	Accuracy
Baseline	85.57%
All	<b>98.03%</b>

Pred ->	Human	NonH
Human	152	23
NonH	1	1040

### › Animate/Inanimate

Features	Accuracy
Baseline	80.92%
All	<b>92.52%</b>

Pred ->	Anim.	Inanim.
Anim.	155	75
Inanim.	16	970



## Discussion

- › Can classify over 90% of Dutch nouns correctly
  - The Cornetto “nonhuman animate” class cannot be classified well
- › Corpus creation/annotation
- › Applications (parser, anaphora resolution)
  
- › Token-based instead of lemma-based (DutchSemCor)?
- › Reduce resource requirements
  - Incorporate morphology, seed set



## References

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- M.C. O'Connor, A. Anttila, V. Fong, and J. Maling. Differential possessor expression in English: Re-evaluating animacy and topicality effects. In *Annual Meeting of the Linguistic Society of America*, January, pages 9–11, 2004.
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- A. Zaenen, J. Carletta, G. Garretson, J. Bresnan, A. Koontz-Garboden, T. Nikitina, M.C. O'Connor, and T. Wasow. Animacy Encoding in English: why and how. In *Proceedings of the 2004 ACL Workshop on Discourse Annotation*, pages 118–125. Association for Computational Linguistics, 2004.
- Cornetto lexical-semantic database: <http://www2.let.vu.nl/oz/cltl/cornetto/>
- Lassy Large corpus: <http://www.let.rug.nl/vannoord/Lassy/>



# Questions?





Thank you for your attention



# Classes

## Human

Brabander  
Eerste-Kamerlid  
afstammeling  
begeleidingsteam  
drieling  
ex-burgemeester  
geallieerden  
haantje-de-voorste  
juf  
oermens  
racist  
tachtiger

## Nonhuman

ANWB  
appelboom  
brandweer  
cycloop  
dienstensector  
embryo  
familie  
ijsbergsla  
maatjesharing  
microbe  
olifant  
snackbar  
vrouwenrechten

## Inanimate

Groningen  
Koninginnedag  
appel  
belastingkantoor  
compassie  
friettent  
gebarentaal  
keel  
orkaan  
robot  
sneltrain  
terrorisme  
zeewier



## Russian case marking

*pervogo (acc=gen) studenta (acc=gen)*  
first student

Fraser and Corbett (1995)

‘the first student’

*pervyj (acc=nom) zakon (acc=nom)*  
first law

‘the first law’



## Dutch Wh-clefts

- a. *Wat ik leuk vind, is die tafel*(GEN=COMM,-ANIMATE)  
what i like, is that table
  - b. *Wat ik leuk vind, is dat huis*(GEN=NEUT,-ANIMATE)  
what i like, is that house
  - c. *Wie ik leuk vind, is dat kind*(GEN=NEUT,+ANIMATE)  
who i like, is that child
  - d. *Wie ik leuk vind, is die vrouw*(GEN=COMM,+ANIMATE)  
who i like, is that woman
- › Found no good counter-examples in corpus search



## Dutch quantifier suffixes

(de Swart et al., 2008)

*De studenten hebben beide<sup>\*</sup>(-n) het boek gelezen.*

the students have both the book read

‘The students have both read the book.’

*De boeken werden beide<sup>\*</sup>(-n) door de studenten gelezen.*

the books were both by the students read

‘Both books were read by the students.’

› In written Dutch



## Fisher's Exact Test: Contingency table

- The Fisher's exact test is calculated using tables
- Totals are fixed

The noun “gevoel” (*feeling*) as a subject of the verb “ontstaan” (*to start, to arise*)

	gevoel	¬gevoel	Row totals
ontstaan	<b>298</b>	5927	<b>6225</b>
¬ontstaan	405	111952	112357
Column totals	<b>703</b>	117879	<b>118582</b>

$p < 0.00001$



## Dependence and independence

- The p-value can go both ways: Association strength

The noun “gevoel” (*feeling*) as a subject of the verb “schrijven” (*to write*)

	gevoel	¬gevoel	Row totals
schrijven	<b>1</b>	299	<b>300</b>
¬schrijven	702	117578	118282
Column totals	<b>703</b>	117879	<b>118582</b>

$p > 0.99999$



## Association strength

- › This p-value can be used as a measure of association strength
- › A low value indicates a strong association, a high value indicates none
- › Because the totals are fixed, you cannot compare p-values from samples of different sizes





## Fisher's exact test Hypothesis

- ›  $H_0$ : The noun  $x$  and the verb  $y$  are independent in subject relations
- ›  $H_1$ : The noun  $x$  occurs as a subject of the verb  $y$  more often than would be expected by chance



## Calculating the value

- › The p-value expresses the total probability of the observed distribution (table) and all the more extreme ones

	gevoel	¬gevoel
ontstaan	<b>298</b>	5927
¬ontstaan	405	111952

	gevoel	¬gevoel
ontstaan	<b>300</b>	5925
¬ontstaan	403	111950

	gevoel	¬gevoel
ontstaan	<b>299</b>	5926
¬ontstaan	404	111951

	gevoel	¬gevoel
ontstaan	<b>301</b>	5924
¬ontstaan	402	111949



## Calculating the value

	gevoel	¬gevoel	totals
ontstaan	298	5927	6225
¬ontstaan	405	111952	112357
totals	703	117879	118582

$$\text{> } P(n) = \frac{6225! * 112357! * 703! * 117879!}{298! * 5927! * 405! * 111952! * 118582!}$$

$$\text{> } P(n + 1) = \frac{6225! * 112357! * 703! * 117879!}{299! * 5926! * 404! * 111951! * 118582!}$$

> etc

$$\text{> } p = P(n) + P(n + 1) + P(n + 2) + \dots$$

> A and B are associated more strongly than would be expected by chance ( $\alpha = 0.001$ )



# Association measure evaluation

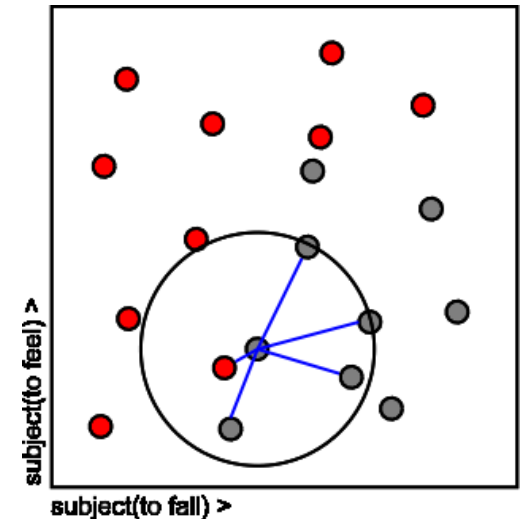
Measure of association	Correctly classified
Pointwise Mutual Information	<b>93.33%</b>
Fisher's Exact Test	91.37%
Frequency	90.96%
None (Baseline)	80.92%



# Number of features

# Wrapped Progressive Sampling (van den Bosch, 2004)

- › TiMBL has many parameters:
  - Nr. of nearest neighbours
  - Feature vector distance measure
  - Neighbour weighting
  - Feature weighting
- › Wrapped Progressive Sampling can automatically converge to the optimal parameters for the data set





# Appendix references

Van den Bosch, A. (2004). Wrapped progressive sampling search for optimizing learning algorithm parameters. In R. Verbrugge, N. Taatgen, and L. Schomaker (Eds.), *Proceedings of the 16th Belgian-Dutch Conference on Artificial Intelligence*, Groningen, The Netherlands

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N.M. Fraser and G.G. Corbett. Gender, animacy, and declensional class assignment: a unified account for Russian. *Yearbook of morphology*, 1994:123–150, 1995.