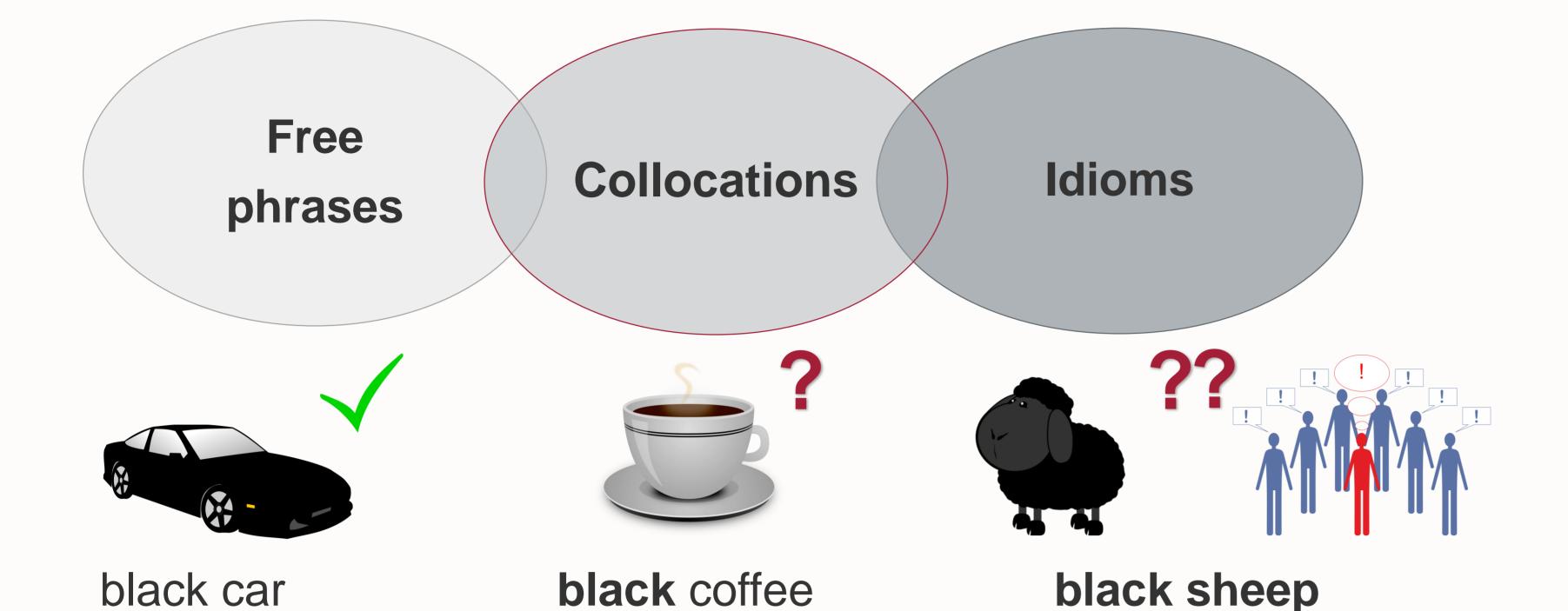






# Automatic Collocation Identification Using Word Embeddings

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#### **Collocations**

- syntactic relation
- partial semantic opacity
- one of the elements carries a special meaning found only in this combination
- constrained lexical selection
- high degree of statistical association

# **Association Measures (AMs)**

- ✓ extract frequency counts for the words from a corpus
- ✓ create ranked n-best lists according to the association measure scores

	f <sub>1</sub> adjective	f <sub>2</sub> noun	f adj+noun	Association (dice)
red army	880	160	22	0.04
red carpet	880	44	15	0.03
red rose	880	46	8	0.01
red dress	880	84	2	0.004

## Word Embeddings

n-dimensional, real-valued vectors co-occurrence frequencies of words (context):

context ->	bitter	black	fast	vector
coffee	2	7	0	[2,7,0]
tea	1	5	0	[1,5,0]
boat	0	1	6	[0,1,6]

He loved his black coffee, crackers, and an orange.

## Dataset (Evert 2008):

- German adj+noun
  lemma combinations
- annotated by lexicographers
- unique adjectives: 489unique nouns: 815

200 dimensions + AMs

300 dimensions + AMs

520 true groß+Liebe grün+Politik collocations 'great love' 'green politics' (categories 1&2)

732 non- groß+Park grün+Baum collocations 'big park' 'green tree'

# Collocation extraction: previous experiments

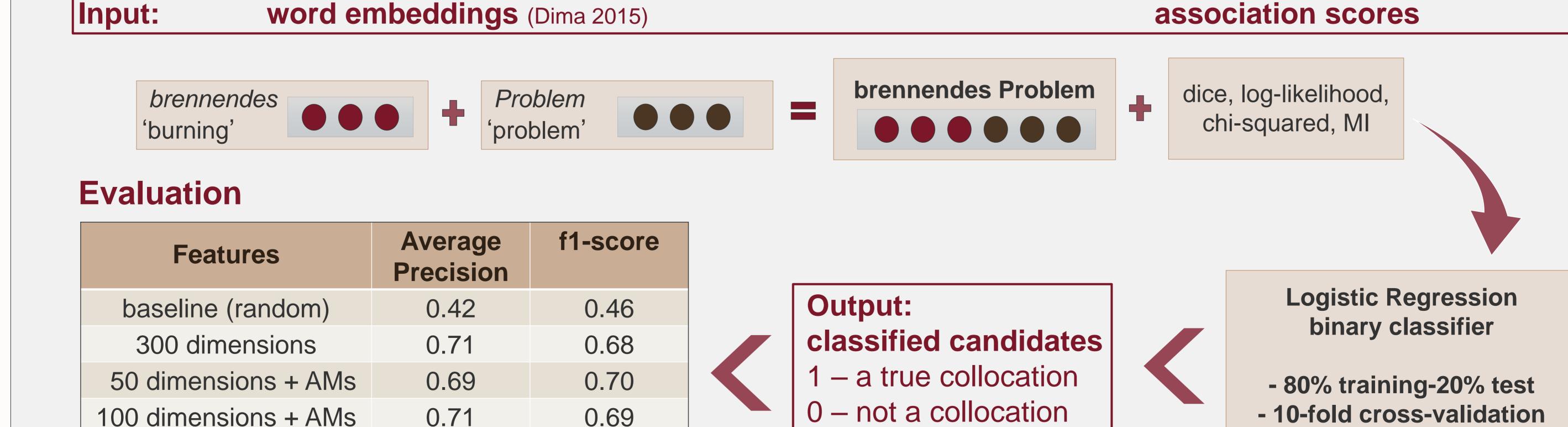
best AMs (Evert 2004, 2008):

- 1. dice
- 2. log-likelihood
- 3. chi-squared

best AMs and combining multiple AMs (Pecina 2008):

Method	Average Precision
Piatersky-Shapiro coefficient	0.63
LDA (multiple AMs)	0.61
GLM (multiple AMs)	0.60

#### Binary classification of collocation candidates



### REFERENCES

Dima, C. (2015). Reverse-engineering language: A study on the semantic compositionality of german compounds. In Proceedings of the 2015 Conference on Empirical Methods in Natural Language Processing, pages 1637–1642, Lisbon, Portugal. Association for Computational Linguistics.

0.69

0.70

0.71

0.72