DISTRIBUTIONAL WORD SIMILARITY David Kauchak CS1*5*9 Fall 2014

Admin

Assignment 5 out





Word similarity

Four general categories

- Character-based
- turned vs. truned
- cognates (night, nacht, nicht, natt, nat, noc, noch)
 Semantic web-based (e.g. WordNet)

Dictionary-based

- Distributional similarity-based
- similar words occur in similar contexts

| Dictionary-based similarity | | | | |
|-----------------------------|---|--|--|--|
| Word | Dictionary blurb | | | |
| aardvark | a large, nocturnal, burrowing mammal, Orycteropus afer, ofcentral and southern Africa, feeding on ants and termites andhaving a long, extensile tongue, strong claws, and long ears. | | | |
| beagle | One of a breed of small hounds having long ears, short legs, and a usually black, tan, and white coat. | | | |
| dog | Any carnivore of the family Canidae, having prominent canine teeth and, in the wild state, a long and slender muzzle, a deep-chested muscular body, a bushy tail, and large, erect ears. Compare canid. | | | |



Dictionary + WordNet

WordNet also includes a "gloss" similar to a dictionary definition

Other variants include the overlap of the word senses as well as those word senses that are related (e.g. hypernym, hyponym, etc.)

incorporates some of the path information as well
 Banerjee and Pedersen, 2003

Word similarity

og or fogdog.

16. 17. 18.

a word f

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Corpus-based

The ${\it Beagle}$ is a breed of small to medium-sized dog. A member of the Hound Group, it is similar in appearance to the Foxhound but smaller, with shorter leg

Beagles are intelligent, and are popular as pets because of their size, even temper, and lack of inherited health problems.

Dogs of similar size and purpose to the modern **Beagle** can be traced in Ancient Greece[2] back to around the 5th century BC.

From medieval times, **beagle** was used as a generic description for the smaller hounds, though these dogs differed considerably from the modern breed.

In the 1840s, a standard **Beagle** type was beginning to develop: the distinction between the North Country Beagle and Southern

Corpus-based: feature extraction

The **Beagle** is a breed of small to medium-sized dog. A member of the Hound Group, it is similar in appearance to the Foxhound but smaller, with shorter leg

We'd like to utilize or vector-based approach

How could we we create a vector from these occurrences?

collect word counts from all documents with the word in it

- collect word counts from all sentences with the word in it
- □ collect all word counts from all words within X words of the word
- collect all words counts from words in specific relationship: subjectobject, etc.

Word-context co-occurrence vectors

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| Web-I | based similarity | |
|-------|--|--|
| | Cocose Martine Informations Cocose | |







Another feature weighting

count how likely feature f_i and word w are to occur together incorporates co-occurrence

 \square but also incorporates how often w and f_i occur in other instances

sim(context_vector(dog), context_vector(beagle))

Does IDF capture this?

Not really. IDF only accounts for f_i regardless of w

Mutual information

A bit more probability $\textcircled{\odot}$

$$I(X,Y) = \sum_{x} \sum_{y} p(x,y) \log \frac{p(x,y)}{p(x)p(y)}$$

When will this be high and when will this be low?

Mutual information

A bit more probability $\textcircled{\odot}$

$$I(X,Y) = \sum_{x} \sum_{y} p(x,y) \log \frac{p(x,y)}{p(x)p(y)}$$

if \boldsymbol{x} and \boldsymbol{y} are independent (i.e. one occurring doesn't impact the other occurring) then:

p(x, y) =

Mutual information

A bit more probability 😊

$$I(X,Y) = \sum_{x} \sum_{y} p(x,y) \log \frac{p(x,y)}{p(x)p(y)}$$

if x and y are independent (i.e. one occurring doesn't impact the other occurring) then:

p(x, y) = p(x)p(y)

What does this do to the sum?





| PMI weighting | | | | | |
|---|---|-------------------------------|--|--|--|
| Mutual information is often used for feature selection in many problem areas | | | | | |
| PMI weighting weights co-occurrences based on their correlation (i.e. high PMI) | | | | | |
| the: is: a: breed: are: intelligent: and: to: modern: | $\begin{array}{c c} c & c & c & c \\ c & c & c \\ c$ | How do we calculate these? | | | |
| | | | | | |