

IN2110: Språkteknologiske metoder
Ord: Leksikalsk semantikk og ordvektorer

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- ▶ **Evaluation** of classifiers.
- ▶ Unsupervised machine learning for class discovery: **Clustering**.
- ▶ **k-means** clustering.



- ▶ Focus on *words* rather than documents.
- ▶ **Distributional models** of word meaning (lexical semantics).
- ▶ Vector Semantics.
- ▶ Words and Vectors.
- ▶ Example tasks for evaluating word vectors.



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What do we want a word meaning model to do for us?

We want a model of word meaning to tell us (e.g.):

- ▶ words have similar meanings (*cat* is similar to *dog*).
- ▶ words are antonyms (*cold* is the opposite of *hot*).
- ▶ words have positive or negative connotations (*happy* and *sad* respectively).
- ▶ the meaning of *sell*, *pay*, *buy* are different perspectives on the same underlying purchasing event.

lemma (citation form)

Dictionary

mouse (N)

1. any of numerous small rodents...
2. a hand-operated device that controls a cursor...

sense

definition

What do words mean?



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mice? = wordform

What do words mean?



- ▶ A **sense** or “concept” is the meaning component of a word.
- ▶ Important component of word meaning = **relationships** between word senses.



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- ▶ Synonymy between **senses**: “a word sense whose meaning is (nearly) identical to a sense of another word”. Chapter 6, p.3.
- ▶ Synonymy between **words** (more formal): “two words are synonyms if they are substitutable one for the other in any sentence without changing the *truth conditions* of the sentence, the situations in which the sentence would be true.” Chapter 6, p.3.



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- ▶ No examples of **perfect** synonymy.
- ▶ The Linguistic Principle of Contrast:
 - ▶ Difference in **form** → difference in **meaning**.
 - ▶ *water & H₂O*: H₂O appropriate in scientific contexts, inappropriate in hiking guide.



Senses that are **opposites** with respect to **one** feature of meaning.
Otherwise, very similar!

- ▶ *dark/light* , *short/long* , *fast/slow* , *rise/fall* , *hot/cold* , *up/down* , *in/out*

More formally, antonyms can:

- ▶ define a **binary opposition**, or be at **opposite ends** of a scale
 - ▶ long/short , fast/slow
- ▶ be **reversives**:
 - ▶ rise/fall, up/down



How to automatically distinguish synonyms from antonyms ?



Words don't have many synonyms, but have lots of **similar** words.

From synonymy to similarity:

- ▶ relations between **senses** → relations between **words**.

Words with similar meanings. Not synonyms, but sharing some element of meaning:

Examples

- ▶ alligator, crocodile
- ▶ love, affection
- ▶ cat, dog



Word **relatedness** (or word **association**): words are related if they do not share features, but commonly “participate” in a shared event.

Similarity VS Relatedness

- ▶ car , bicycle: **similar**.
- ▶ car , gasoline: **related**, **not** similar.

Word Relatedness

- ▶ car , tyre
- ▶ car , motorway
- ▶ car , crash
- ▶ coffee , cup



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Words can be related in any way, e.g. via a **semantic frame** or **semantic field**.

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How to get values for word similarities?

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SimLex-999 dataset (Hill et al., 2015):

- ▶ gold standard resource for evaluating distributional semantic models.
- ▶ quantifies similarity rather than relatedness.

word1	word2	similarity
vanish	disappear	9.8
behave	obey	7.3
belief	impression	5.95
muscle	bone	3.65
modest	flexible	0.98
hole	agreement	0.3



Words that:

- ▶ cover a particular semantic domain.
- ▶ bear structured relations with each other.

Examples

- ▶ **hospitals:** surgeon, scalpel, nurse, anesthetic, hospital
- ▶ **restaurants:** waiter, menu, plate, food, chef
- ▶ **houses:** door, roof, kitchen, family, bed



Closely related to semantic fields.

“A **semantic frame** is a set of words that denote **perspectives** or **participants** in a particular type of event.” Chapter 6, p.4.

Frames have semantic roles, and words in a sentence can take on these roles.

- ▶ buy, sell, pay.
- ▶ buyer, seller, goods, money.

Semantic frames makes it possible for systems to recognize paraphrases.

- ▶ Sam bought the book from Ling.
- ▶ Ling sold the book to Sam.



Word senses can be related **taxonomically**.

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hyponyms and **hypernyms**:

- ▶ a word (sense) is a **hyponym** of another word (sense) if the first is more specific, denoting a subclass of the other.
- ▶ a word (sense) is a **hypernym** of another word (sense) if the first one is more general.

Examples

- ▶ car is a **hyponym** of vehicle.
- ▶ mango is a **hyponym** of fruit.
- ▶ vehicle is a **hypernym** of car.
- ▶ fruit is a **hypernym** of mango.

Hypernymy can be defined in terms of **entailment**:

- ▶ sense A is a hyponym of B if everything that is A is also B.
- ▶ being A entails being B.

Hyponymy and hypernymy are **transitive**: A hyponym of B, B hyponym of C, then A is hyponym of C.

Hyponyms and hypernyms structure is the **IS-A** hierarchy: A IS-A B, or B **subsumes** A.

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Hyponym and hypernym too similar: easily confused.

The words **subordinate** and **superordinate** are used instead.

Superordinate	vehicle	fruit	furniture
Subordinate	car	mango	chair



Words have **affective meanings** or **connotations**.

Connotations are aspects of a word's meaning related to a writer/reader's emotions, sentiment, opinions, or evaluations.

- ▶ positive **connotations** (happy).
- ▶ negative **connotations** (sad).
- ▶ positive **evaluation** (great, love) – sentiment.
- ▶ negative **evaluation** (terrible, hate) – sentiment.

In affective meaning (Osgood et al., 1957) – words vary along 3 dimensions:

- ▶ **valence**, **arousal**, and **dominance** represented by numbers.
- ▶ word meaning can be represented as a vector, a list of numbers, point in a dimensional space.



Concepts or word senses

- ▶ Have a complex many-to-many association with words (homonymy, multiple senses)
- ▶ Have relations with each other
 - ▶ Synonymy , Antonymy, Similarity, Relatedness, Superordinate/subordinate, Connotation

How to build a computational model that successfully deals with the different aspects of word meaning?

How to define word meaning for a computational model?



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- ▶ “The meaning of a word is its use in the language” (Ludwig Wittgenstein, 1953, PI 43).
- ▶ The linguistic **distributionalists** Joos (1950), Harris (1954), Firth (1957):
 - ▶ words are defined by their **environments** or distributions (the words around them).
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 - ▶ a word's distribution is the set of **contexts** in which it occurs: the **neighboring words** or **grammatical environment**.
- ▶ “If A and B have almost identical environments we say that they are synonyms.” (Zellig Harris, 1954).

→ two words that occur in very similar distributions (context, similar words) tend to have the same meaning.

What does ongchoi mean?



Suppose you see these sentences:

- ▶ Ongchoi is delicious **sautéed with garlic**.
- ▶ Ongchoi is superb **over rice**.
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And you've also seen these:

- ▶ ...spinach **sautéed with garlic over rice**.
- ▶ Chard stems and **leaves** are **delicious**.
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Conclusion:

- ▶ Ongchoi is a leafy green like spinach, chard (bladbete), or collard greens (en type grønnkål).

How to do it computationally?



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- ▶ count words in the context of *ongchoi*:
 - ▶ find words like *sauteed*, *eaten*, *garlic*.
 - ▶ these words occur around *spinach/collard green*.

→ there is a similarity between *ongchoi* and *spinach/collard green*.



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Vector semantics combines two intuitions:

- ▶ **distributional** intuition.
- ▶ **vector** intuition (Osgood et al., 1957, slide 22)

Model of word meaning



Build a model of meaning focused on similarity:

- ▶ Each word = a vector.
- ▶ Similar words are “nearby in space”.





A word vector is called an “**embedding**” (embedded into a space).

The **standard** way to represent meaning in NLP.

Fine-grained model of meaning for similarity:

We focus on 2 kinds of embeddings:

- ▶ Tf-idf:
 - ▶ Sparse vectors and common baseline model.
 - ▶ Words are represented by a simple function of the counts of nearby words.
- ▶ Word2vec (next week with Fredrik):
 - ▶ Dense vectors.
 - ▶ Representation is created by training a classifier to distinguish nearby and far-away words.

Vectors and documents VS Words and vectors



Distributional models of meaning (vectors) generally based on a **co-occurrence** matrix.

Term-document matrix VS **word-word matrix**:

	As You Like It	Twelfth Night	Julius Caesar	Henry V
battle	1	0	7	13
good	114	80	62	89
fool	36	58	1	4
wit	20	15	2	3

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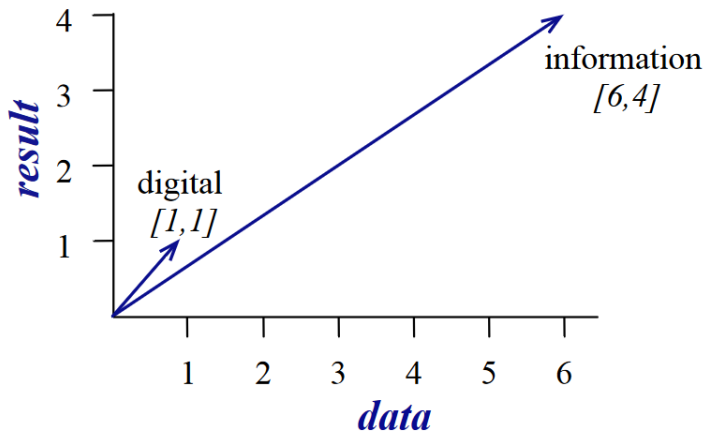
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	aardvark	...	computer	data	pinch	result	sugar
apricot	0	...	0	0	1	0	1
pineapple	0	...	0	0	1	0	1
digital	0	...	2	1	0	1	0
infomation	0	...	1	6	0	4	0





Based on all we've seen so far, how to compute word similarity?
(Eivind will tell you more about it!)