

Word Vectors

Princeton AI4ALL: NLP Group

How should we represent words?

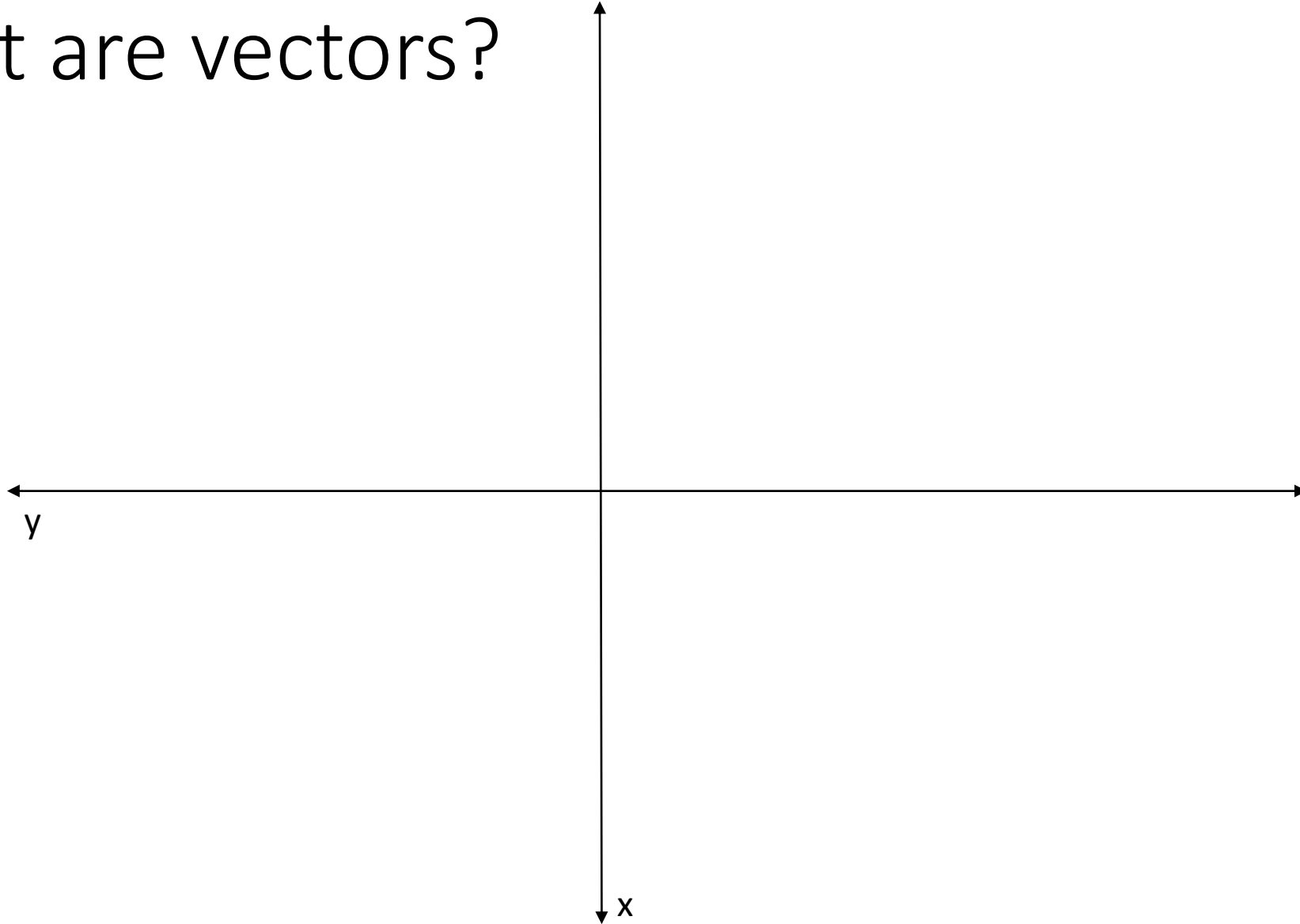
To communicate this information to computers...

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quantitatively represent words,

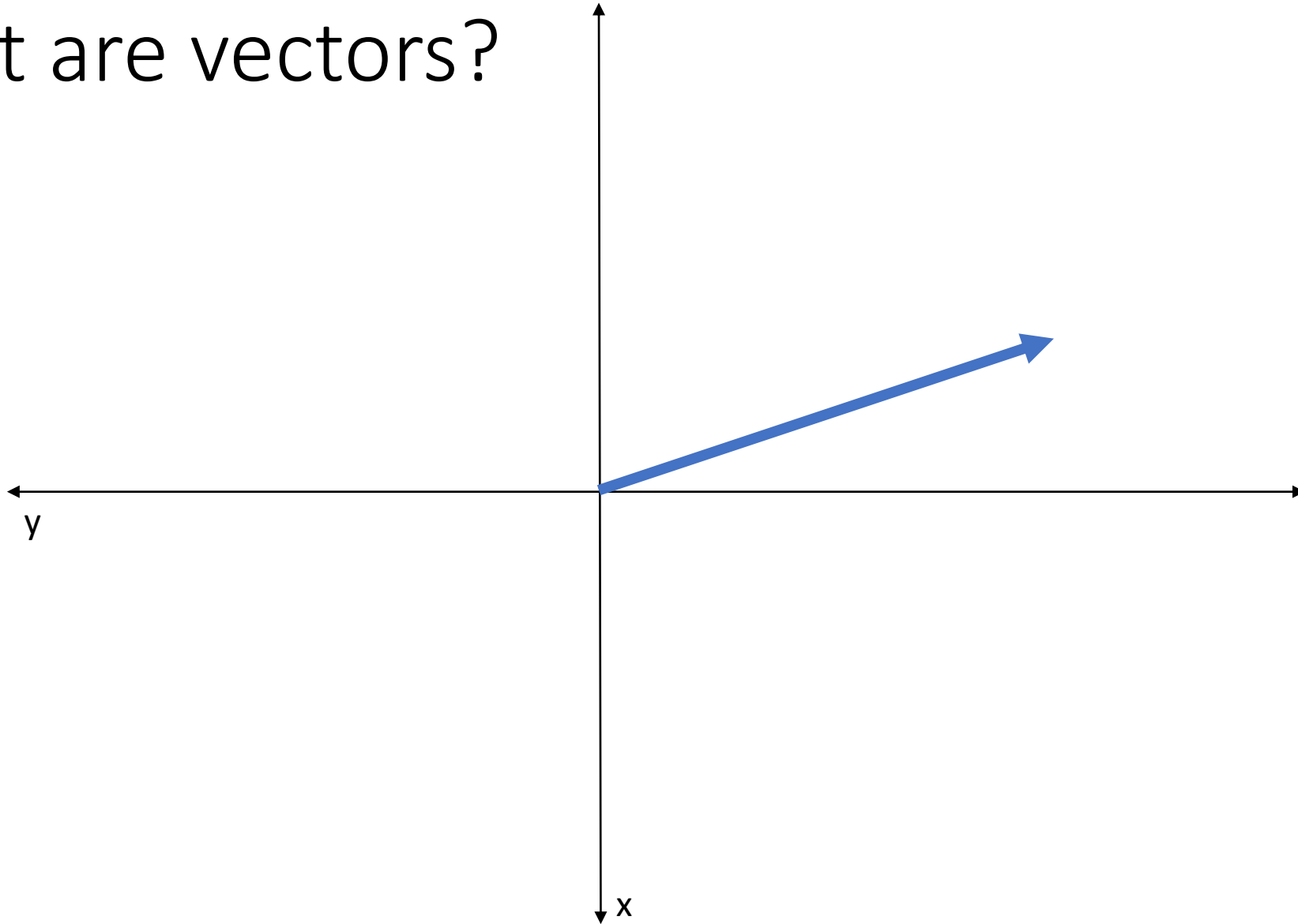
To communicate this information to computers...
quantitatively represent words,
where more similar words have similar
representations

What are word vectors?

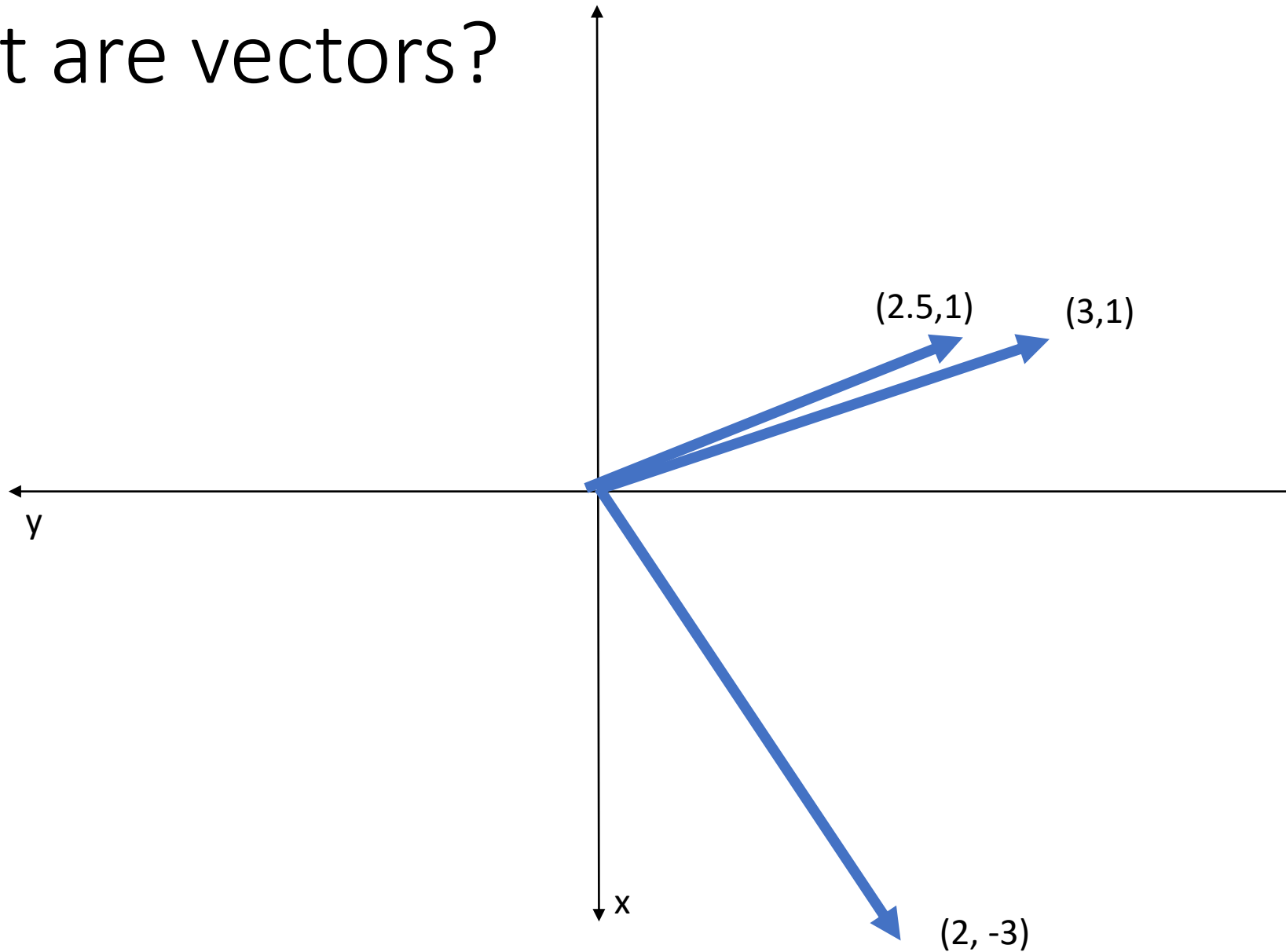
What are vectors?



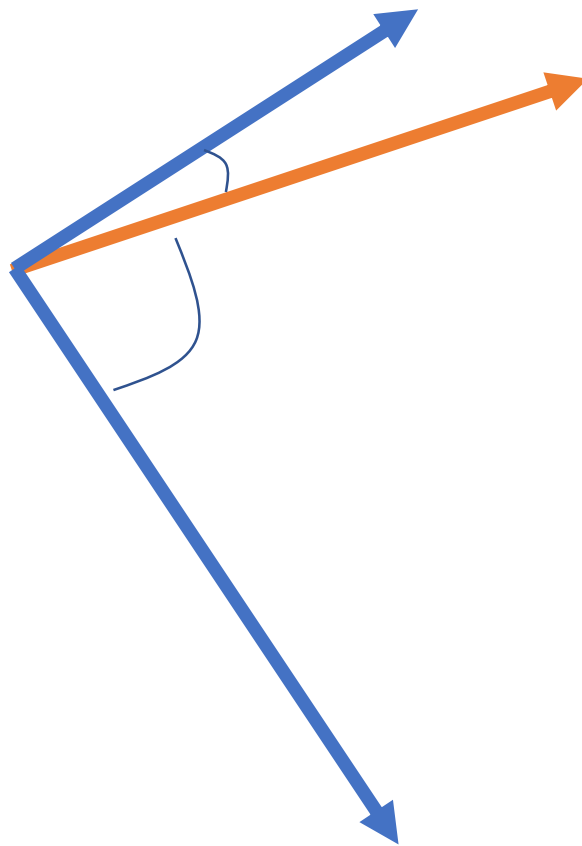
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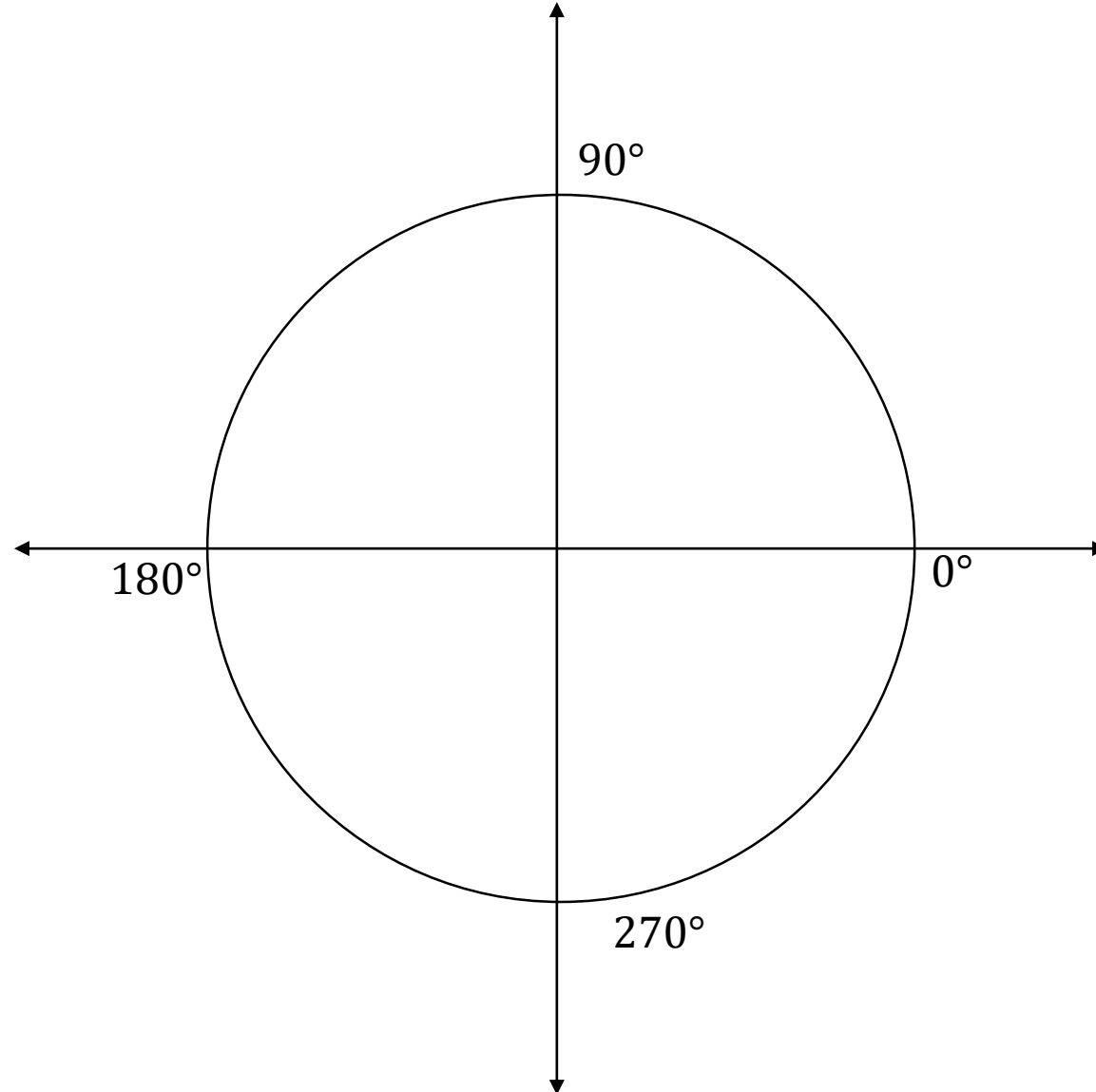
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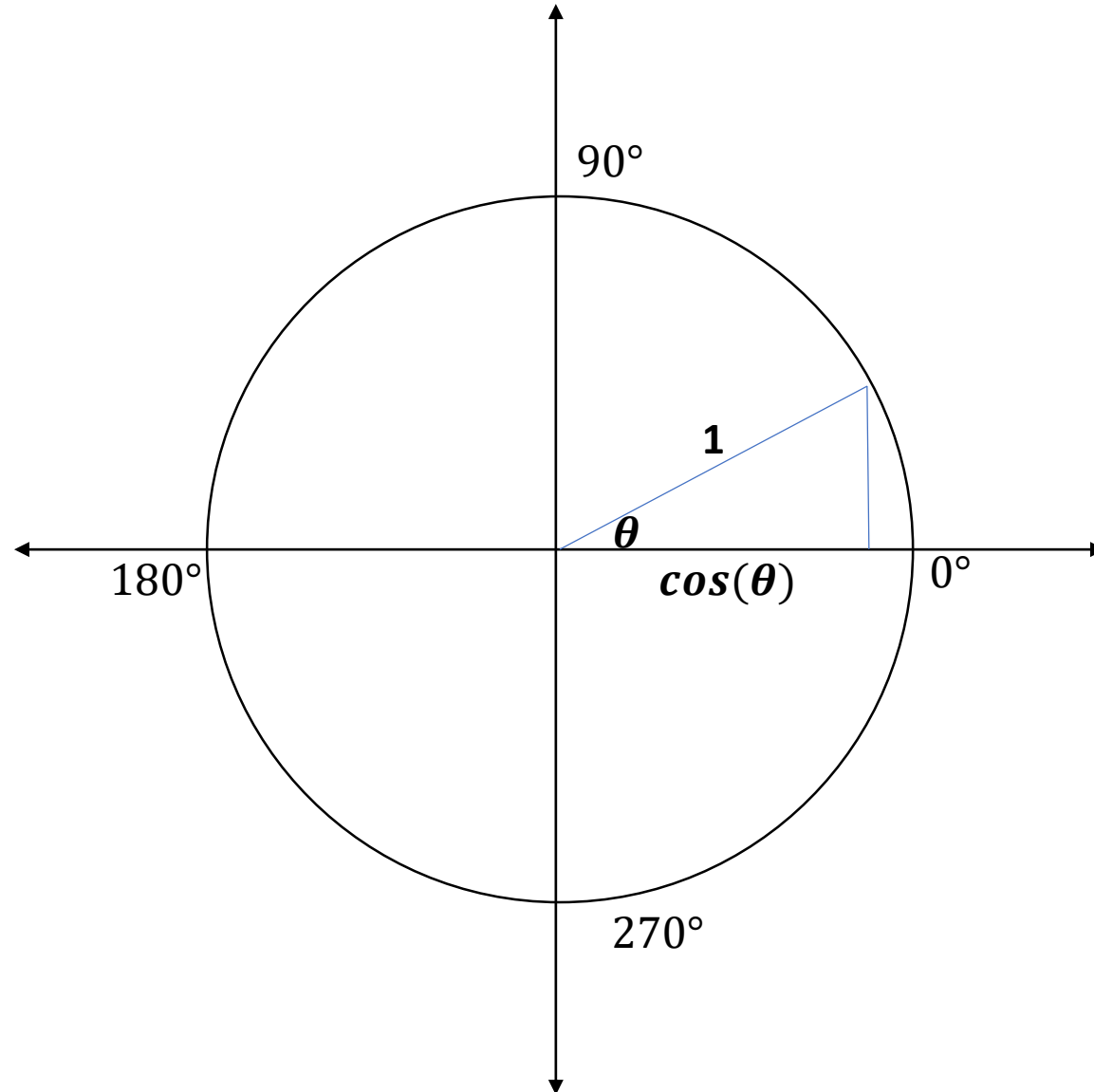
How do we measure similarity between vectors?



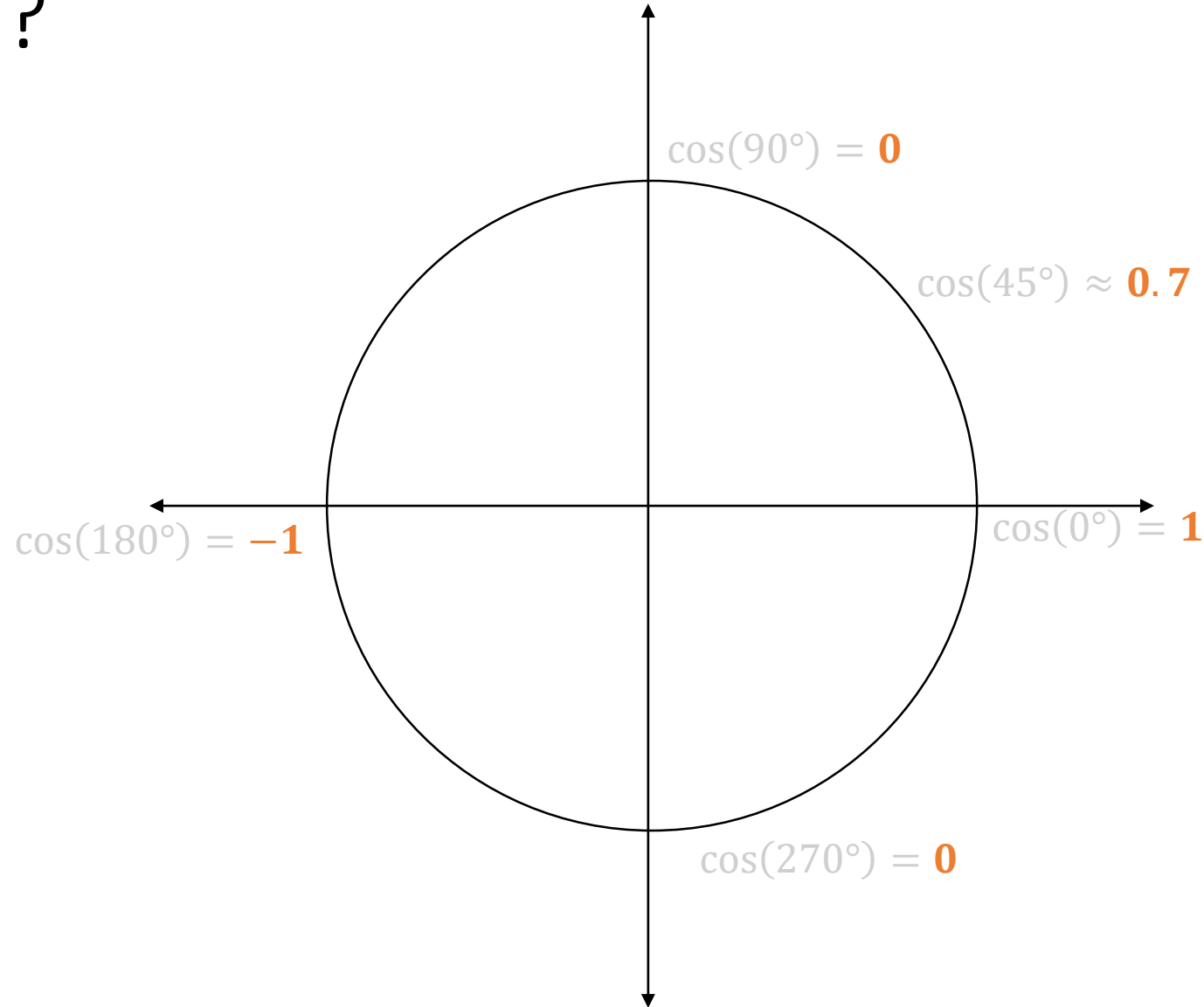
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How do we measure similarity between vectors?



How do we measure similarity between vectors?

Use the cosine similarity between vectors

What if we want more information?

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Vectors in 2 dimensions

3	2.5	2
1	1	-3

Vectors in 6 dimensions

3	2.5
1	1
-2	0.5
4	-2
1	9
0.7	1

What if we want more information?

Vectors in 6 dimensions

3	2.5
1	1
-2	0.5
4	-2
1	9
0.7	1

$$\text{cosine similarity (a, b)} = \frac{a \cdot b}{||a|| \cdot ||b||}$$

How do we create word vectors?

How do we know if words are similar?

How do we know if words are similar?

PRINCETON UNIVERSITY

WordNet

A Lexical Database for English

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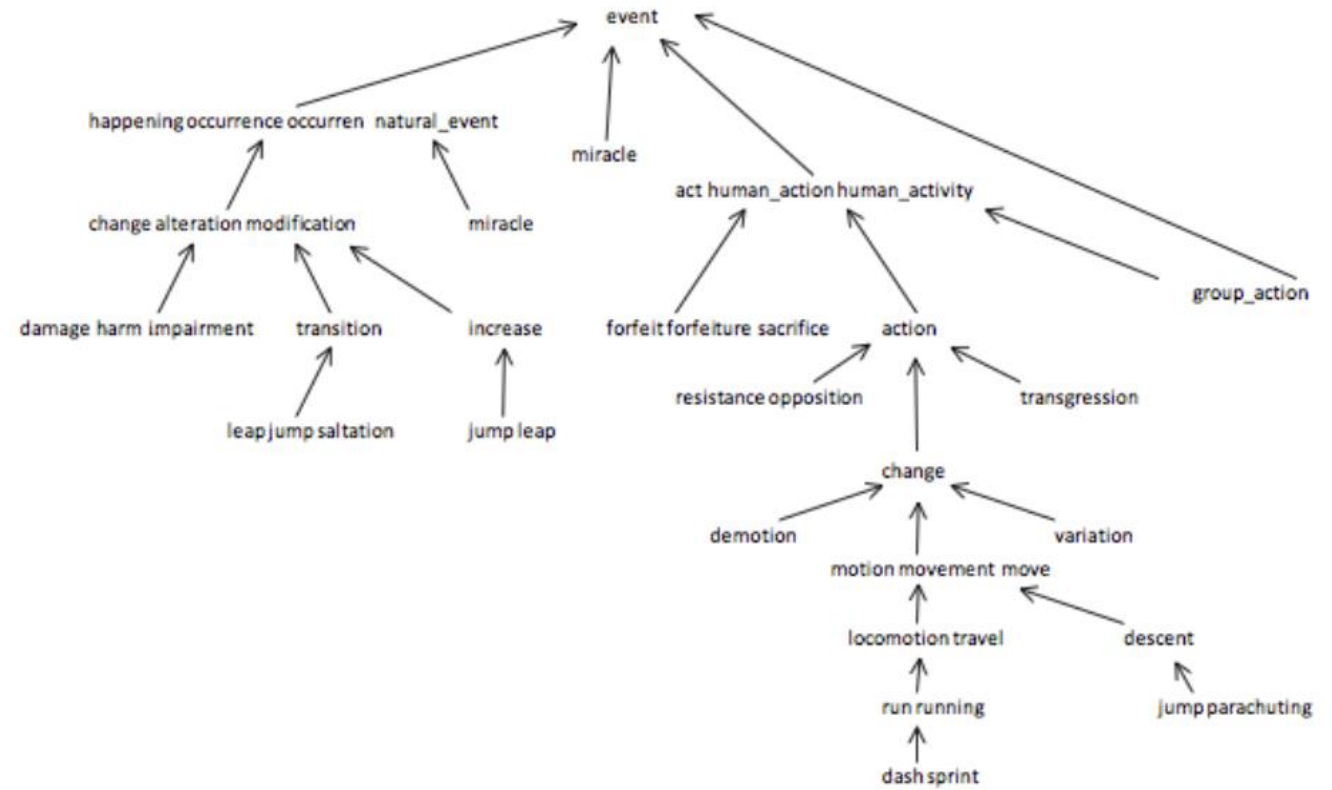
What is WordNet?

Any opinions, findings, and conclusions or recommendations expressed in this material are those of the creators of WordNet and do not necessarily reflect the views of any funding agency or Princeton University.

When writing a paper or producing a software application, tool, or interface based on WordNet, it is necessary to properly [cite the source](#). Citation figures are critical to WordNet funding.

About WordNet

WordNet® is a large lexical database of English. Nouns, verbs, adjectives and adverbs are grouped into sets of cognitive synonyms (synsets), each expressing a distinct concept. Synsets are interlinked by means of conceptual-semantic and lexical relations. The resulting network of meaningfully related words and concepts can be navigated with the [browser](#). WordNet is also freely and publicly available for [download](#). WordNet's structure makes it a useful tool for computational linguistics and natural language processing.



<https://wordnet.princeton.edu/>

<https://www.cs.princeton.edu/courses/archive/fall12/cos226/assignments/wordnet.html>

How do we know if words are similar?

Natural language processing

From Wikipedia, the free encyclopedia

This article is about language processing by computers. For the processing of language by the human brain, see [Language processing in the brain](#).

Natural language processing (NLP) is an area of [computer science](#) and [artificial intelligence](#) concerned with the interactions between computers and human (natural) languages, in particular how to program computers to process and analyze large amounts of [natural language](#) data.

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Distributional Hypothesis

“You shall know a word by the company it keeps”
- John Rupert Firth (1957)

How do we find the distribution of nearby words?

1. Count cooccurrence within a window of size n
2. Create a vector cooccurrences

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1. Count cooccurrence within a window of size n
2. Create a vector cooccurrences

Any problems with this?

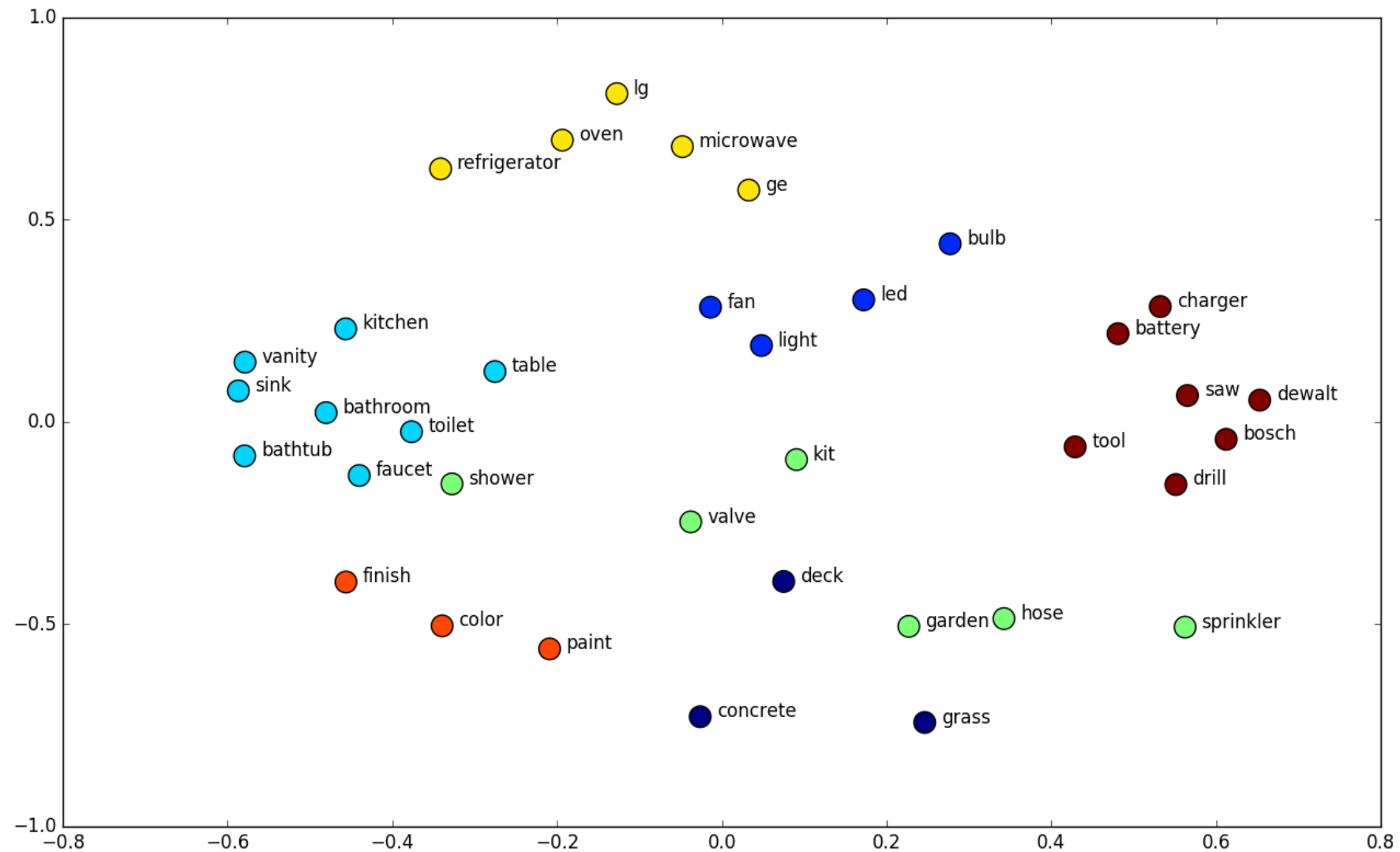
Some words are more informative than others...

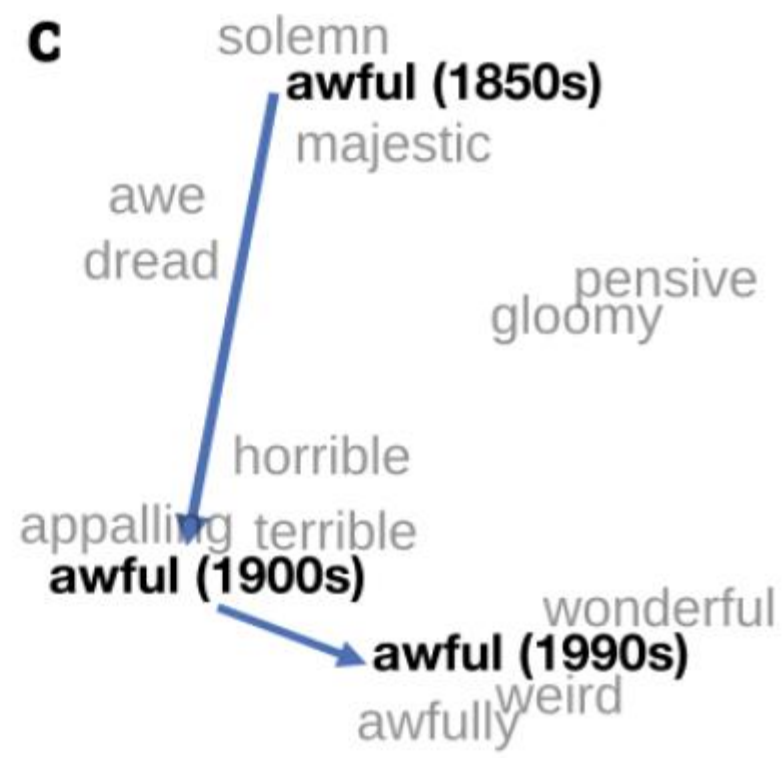
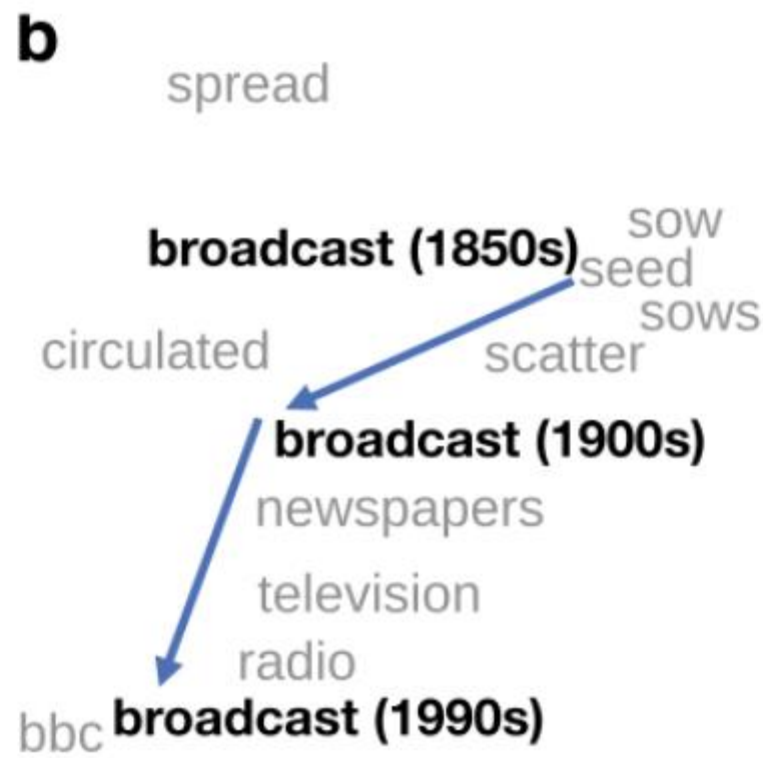
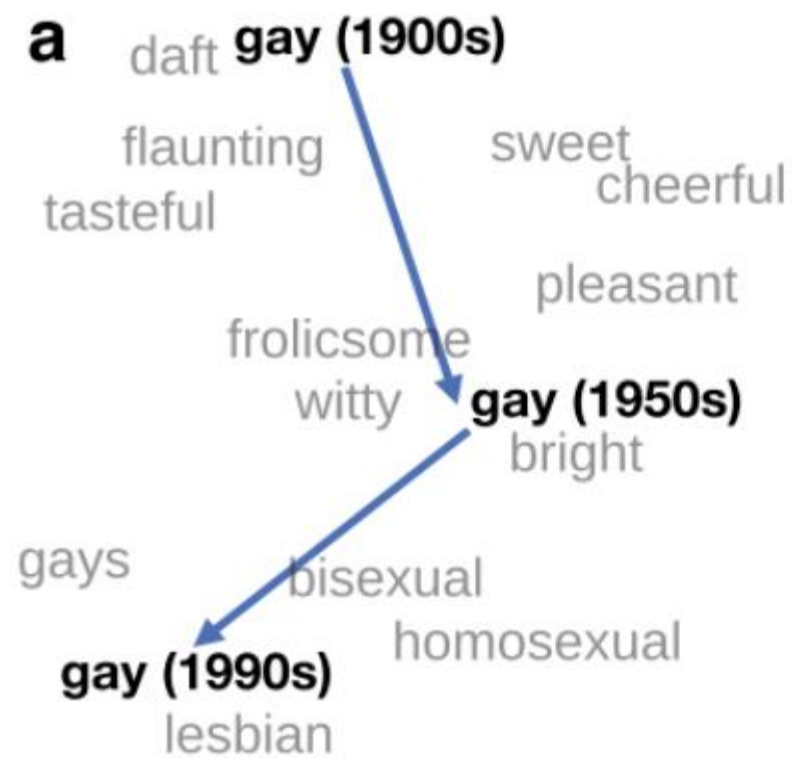
Pointwise Mutual Information (PMI)

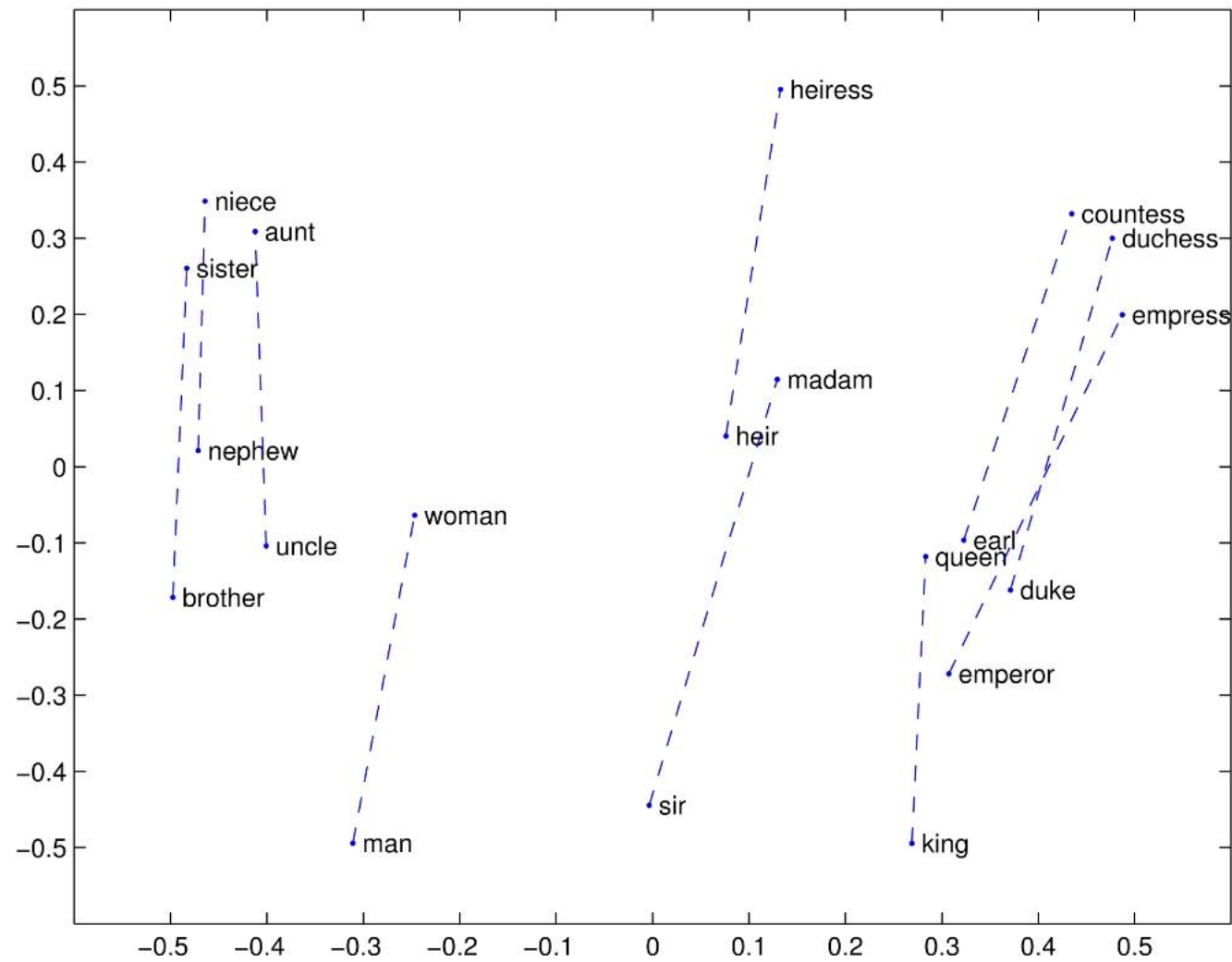
$$PMI(a, b) = \log \frac{P(a, b)}{P(a)P(b)}$$

How can we represent sentences
given word embeddings?

How can we use word vectors?







Man is to Computer Programmer

Tolga Bolukbasi¹, Kai-Wei Chang², James Zou², Venkatesh Saligrama^{1,2}, Adam Kalai²

¹Boston University, 8 Saint Mary's Street, Boston, MA

²Microsoft Research New England, 1 Memorial Drive, Cambridge, MA

tolgab@bu.edu, kw@kwchang.net, jamesyzou@gmail.com, srv@bu.edu, adam.kalai@microsoft.com

$$\vec{\text{man}} - \vec{\text{woman}} \approx \vec{\text{king}} - \vec{\text{queen}},$$

<https://papers.nips.cc/paper/6228-man-is-to-computer-programmer-as-woman-is-to-homemaker-debiasing-word-embeddings.pdf>

Man is to Computer Programmer as Woman is to Homemaker? Debiasing Word Embeddings

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$$\vec{\text{man}} - \vec{\text{woman}} \approx \vec{\text{computer programmer}} - \vec{\text{homemaker}}.$$

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Extreme <i>she</i>	Extreme <i>he</i>			
1. homemaker	1. maestro	sewing-carpentry nurse-surgeon blond-burly giggle-chuckle sassy-snappy volleyball-football queen-king waitress-waiter	Gender stereotype <i>she-he</i> analogies	
2. nurse	2. skipper		registered nurse-physician	housewife-shopkeeper
3. receptionist	3. protege		interior designer-architect	softball-baseball
4. librarian	4. philosopher		feminism-conservatism	cosmetics-pharmaceuticals
5. socialite	5. captain		vocalist-guitarist	petite-lanky
6. hairdresser	6. architect		diva-superstar	charming-affable
7. nanny	7. financier		cupcakes-pizzas	lovely-brilliant
8. bookkeeper	8. warrior		Gender appropriate <i>she-he</i> analogies	
9. stylist	9. broadcaster		sister-brother	mother-father
10. housekeeper	10. magician		ovarian cancer-prostate cancer	convent-monastery

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