

# Word2vec

**Representing word meanings in form of large vectors**

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# Main Idea

*“You shall know a word by the company it keeps” : J.R. Firth 1957*

Words that occurs in *similar contexts* tend to have *similar meanings*

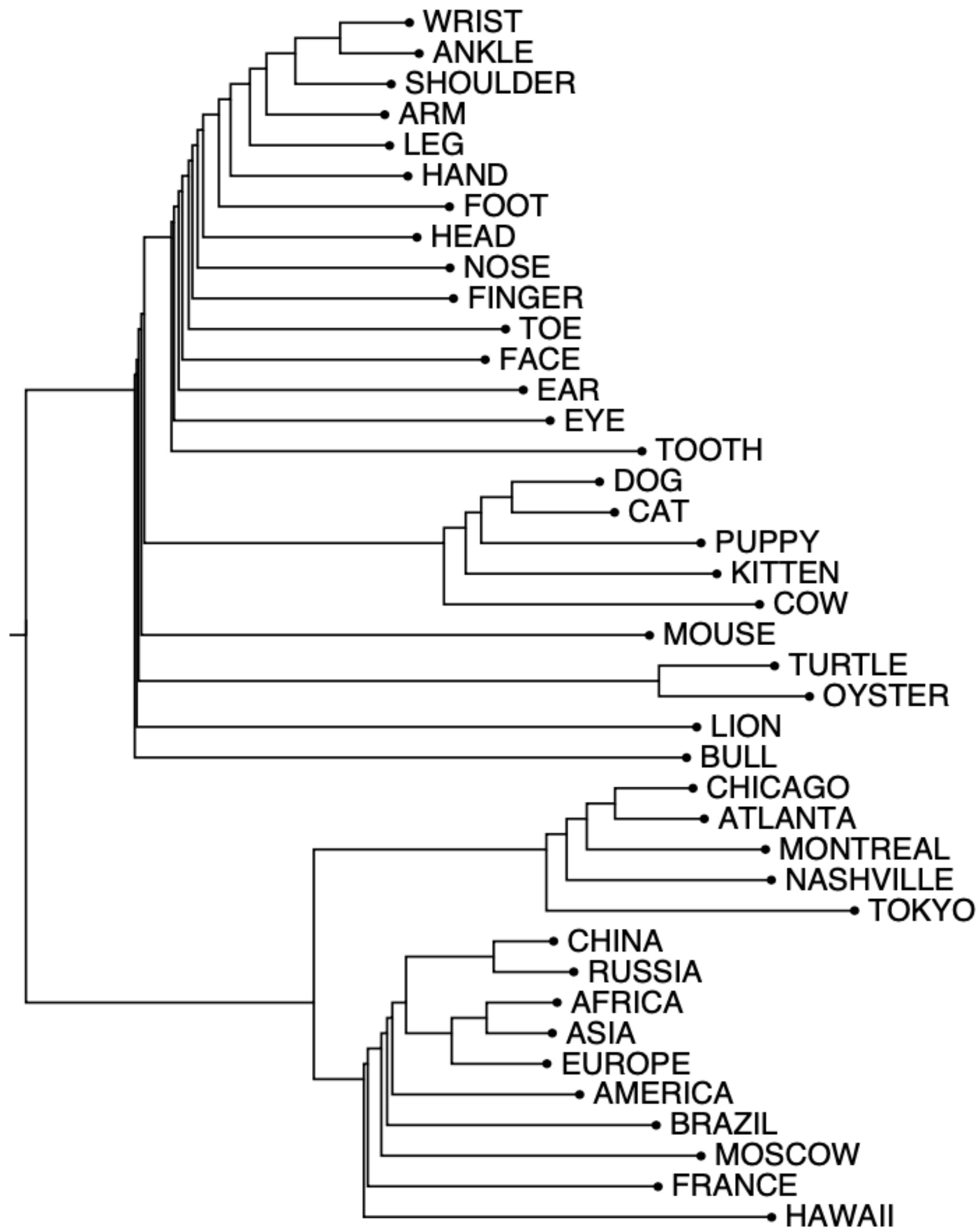
...government debt problems turning into **banking** crisis as happened in 2009...

...saying the Europe needs unified **banking** regulation to replace the hodgepodge...

...India has just given its **banking** system a shot in the arm...



These context words represent **banking**



Words are represented as vectors

Similar words are closer in vector space

Similarity can be computed using  
normalised dot product (cosine similarity)  
between any two vectors (words)

$$\text{similarity}(\vec{Arm}, \vec{Leg}) = \frac{\vec{Arm} \cdot \vec{Leg}}{|\vec{Arm}| |\vec{Leg}|}$$

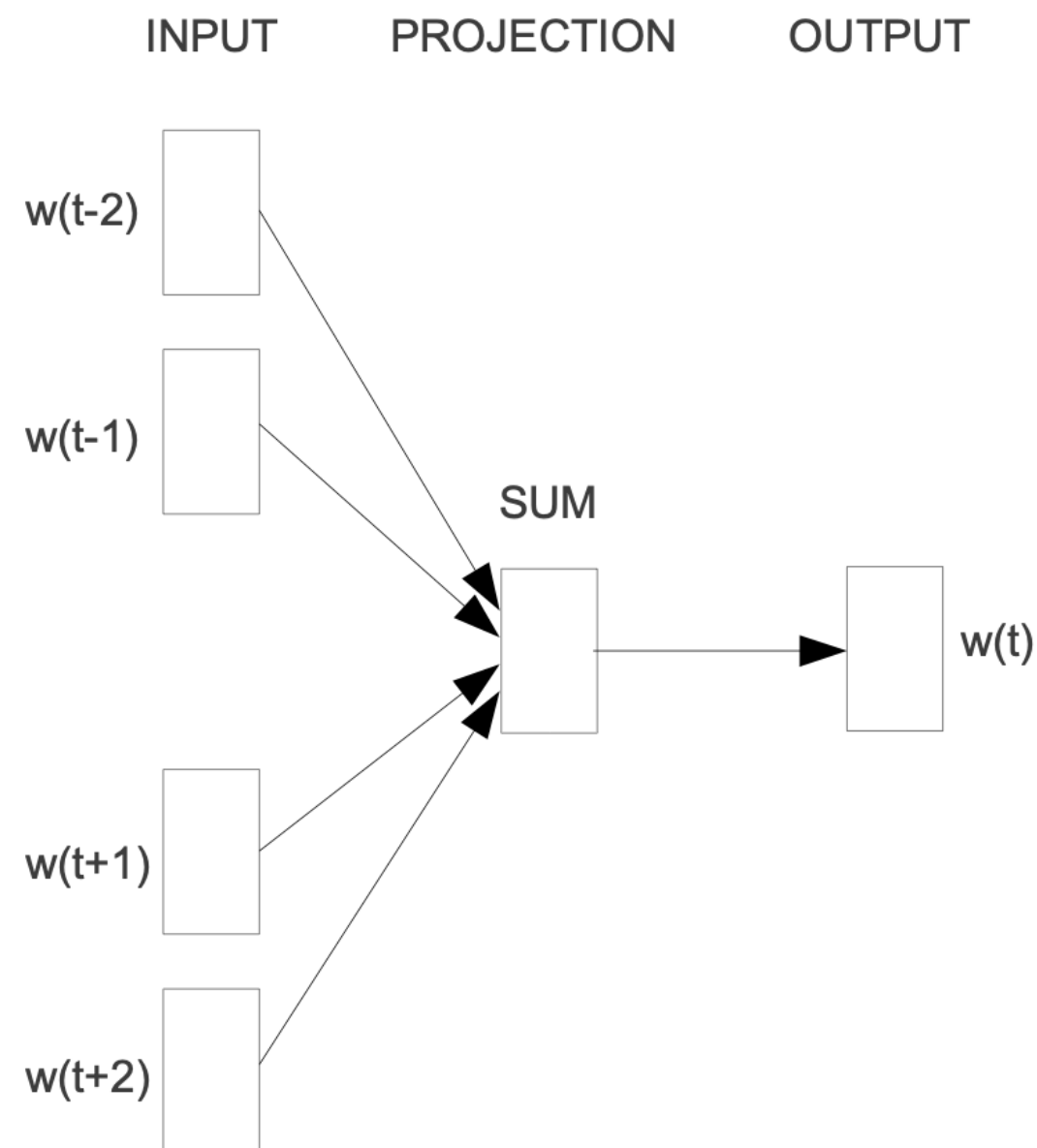
# Two methods

## Continuous bag of words

Use the context and predict the word in the middle

More intuitive

One prediction

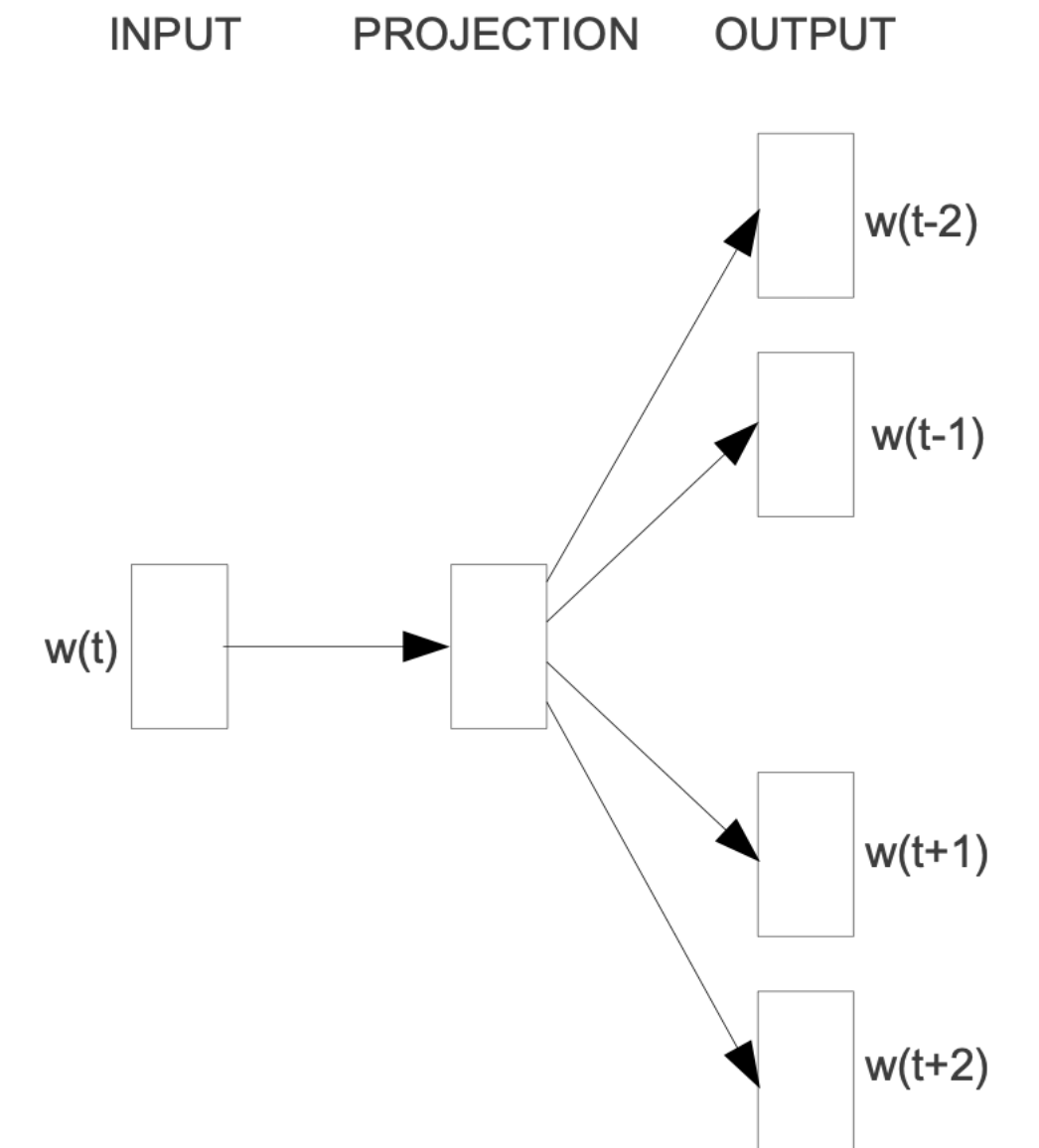


## Skip N-gram

Use the word in the middle to predict the context words

Commonly used because it works better for rarer words

Multiple predictions



# Methodology

Represent each word as a  $d$  (*around 300*) dimensional vector

Represent each context word as a  $d$  dimensional vector

Initialise all vectors to random weights

Arrange them into two matrices  $W$  (word) and  $C$  (context word)

# Cost Function

$$\log p(c|w; \theta) = \frac{\exp v_c \cdot v_w}{\sum_{c' \in C} \exp v_{c'} \cdot v_w}$$

*Predict context word(s) from focus word*

## Objective Function

$$\arg \max_{\theta} \sum_{(w,c) \in D} \log p(c|w) = \sum_{(w,c) \in D} \left[ \log \exp v_c \cdot v_w - \log \sum_{c'} \exp v_{c'} \cdot v_w \right]$$

## Negative Sampling Distribution

$$p^{NS}(w) = \frac{f(w)^{\frac{3}{4}}}{\sum_{w'} f(w)^{\frac{3}{4}}}$$

# Some Insights

Why two different representations of the same word (word and context)?

What about the words that have multiple meanings?

Why  $d=300$ ?

## GitHub Repo Link

<https://github.com/Sahil1776/NLP-TeachBack-Assignment>

# References

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