

## Introduction

- 2018 was a breakthrough year in NLP. Transfer learning.

## BERT

- BERT (Bidirectional Encoder Representations from Transformers). Google.
- You can either use these models to extract high quality language features from your text data, or you can fine-tune these models on a specific task (classification, entity recognition, question answering, etc.) with your own data to produce state of the art predictions.
- if you want to match customer questions or searches against already answered questions or well documented searches, these representations will help you accurately retrieve results matching the customer's intent and contextual meaning, even if there's no keyword or phrase overlap.
- BERT offers an advantage over models like Word2Vec, because while each word has a fixed representation under Word2Vec regardless of the context within which the word appears, BERT produces word representations that are dynamically informed by the words around them. For example, given two sentences:

“The man was accused of robbing a bank.” “The man went fishing by the bank of the river.”

Word2Vec would produce the same word embedding for the word “bank” in both sentences, while under BERT the word embedding for “bank” would be different for each sentence.



Filters





Filters





Filters



Train only this layer





2 types of WE:

- Frequency based Embedding
  - Count Vector
  - TF-IDF Vector
  - Co-occurrence vector ( Similar words tend to occur together and will have similar context for example)

	He	She	lazy	boy	Neeraj	person
D1	1	1	2	1	0	0
D2	0	0	1	0	1	1

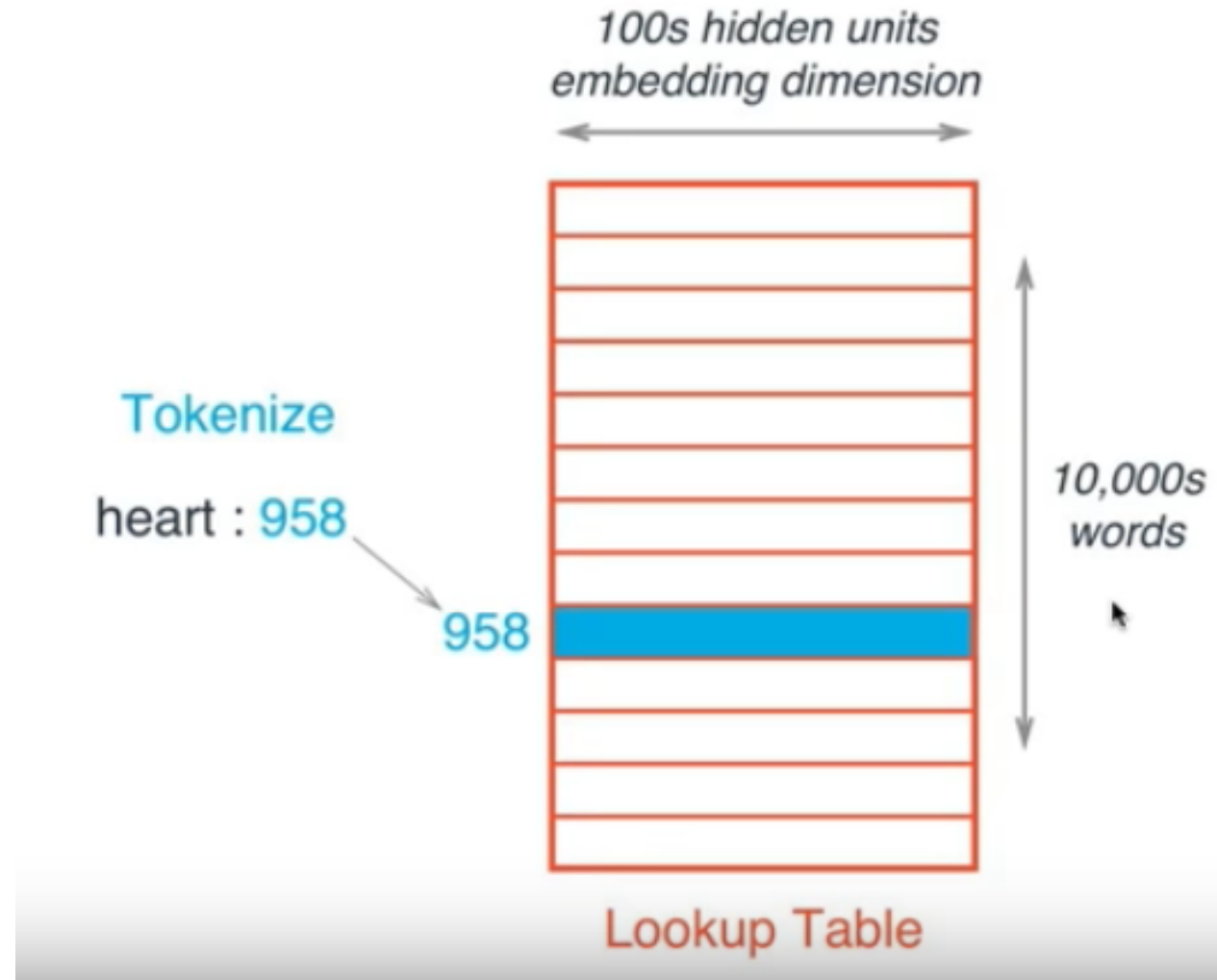
Quick Brown Fox Jump Over The Lazy Dog

The green words are a 2 (around) context window for the word 'Fox' and for calculating the co-occurrence only these words will be counted.

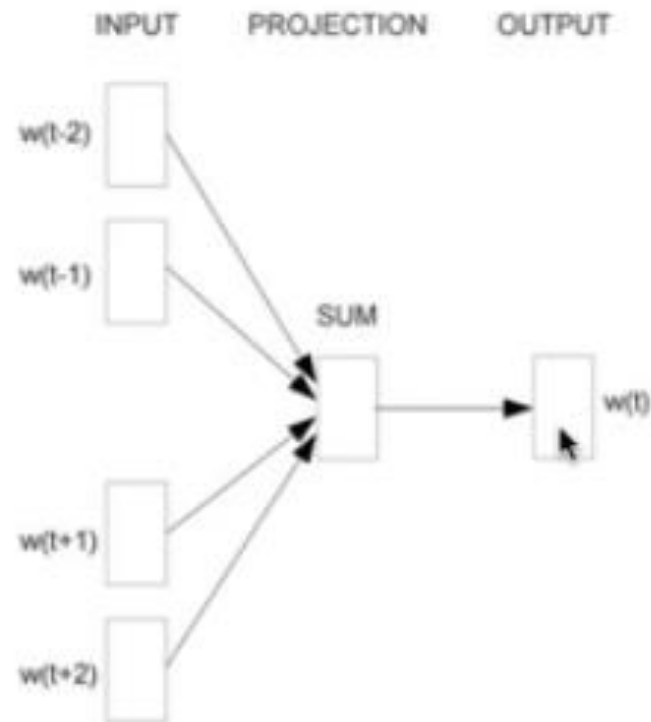
- Prediction based Embedding

- Prediction based Embedding

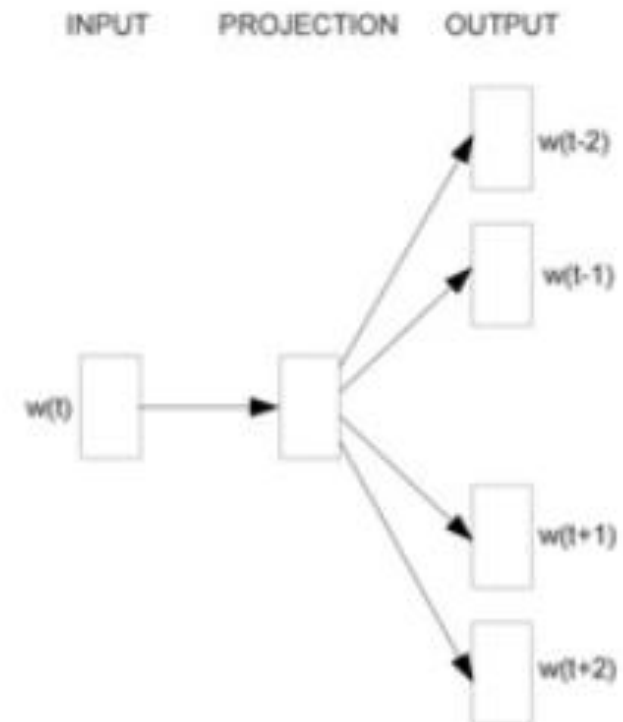
Not only for words. Whenever you have a massive number of classes



- Prediction based Embedding
- Word2vec
- CBOW
    - Input is the context around your word
  - Skip-gram
    - The other way round



**CBOW**



**Skip-gram**



- Analyzing BERT
  - Understand Tokenization in BERT
  - Run BERT & understand the output
  - Word & sentence vectors
  - Check similarity between vectors

Scalar   Vector   Matrix   Tensor

1

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$

$\begin{bmatrix} \begin{bmatrix} 1 & 2 \end{bmatrix} & \begin{bmatrix} 3 & 2 \end{bmatrix} \\ \begin{bmatrix} 1 & 7 \end{bmatrix} & \begin{bmatrix} 5 & 4 \end{bmatrix} \end{bmatrix}$

[buff.ly/2FSYAQK](https://buff.ly/2FSYAQK)

Bibliografía:

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