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## **Lecture: Vector Space Models**

- **Video:** Week Introduction 47 sec
- **Video:** Vector Space Models 2 min
- Reading: Vector Space Models 10 min
- Video: Word by Word and Word by Doc. 4 min
- Reading: Word by Word and Word by Doc.

  10 min
- Lab: Linear algebra in Python with Numpy
- **Video:** Euclidean Distance
- Reading: Euclidian Distance
   10 min
- (b) **Video:** Cosine Similarity: Intuition 2 min
- Reading: Cosine Similarity: Intuition
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- Video: Manipulating Words in Vector Spaces 3 min
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- E Lab: Manipulating word embeddings
- Video: Visualization and PCA 3 min
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   10 min
- Video: PCA Algorithm
  3 min
- Reading: PCA algorithm

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- E Lab: Another explanation about PCA
  1h
- Reading: The Rotation Matrix (Optional Reading)10 min

Lecture Notes (Optional)

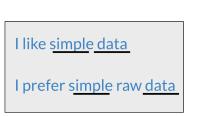
**Practice Quiz** 

**Assignment: Vector Space Models** 

# Word by Word and Word by Doc.

### Word by Word Design

We will start by exploring the word by word design. Assume that you are trying to come up with a vector that will represent a certain word. One possible design would be to create a matrix where each row and column corresponds to a word in your vocabulary. Then you can iterate over a document and see the number of times each word shows up next each other word. You can keep track of the number in the matrix. In the video I spoke about a parameter K. You can think of K as the bandwidth that decides whether two words are next to each other or not.



	simple	raw	like	1	
data	2	1	1	0	
					_

n

k=2

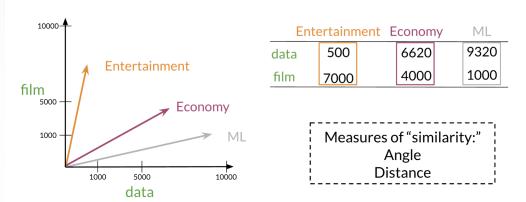
In the example above, you can see how we are keeping track of the number of times words occur together within a certain distance k. At the end, you can represent the word data, as a vector v = [2, 1, 1, 0].

### **Word by Document Design**

You can now apply the same concept and map words to documents. The rows could correspond to words and the columns to documents. The numbers in the matrix correspond to the number of times each word showed up in the document.

	Corpus			
	Entertainment	Economy	Machine Learning	
data	500	6620	9320	
film	7000	4000	1000	

You can represent the entertainment category, as a vector v=[500,7000]. You can then also compare categories as follows by doing a simple plot.



Later this week, you will see how you can use the angle between two vectors to measure similarity.