



# Latent Dirichlet Allocation (LDA)



# LDA

- A statistical model for discovering the abstract *topics* in ***topic modeling***.

## What is topic modeling?

Topic modeling is a method for ***unsupervised*** classification of documents, similar to clustering on numeric data, which finds some natural groups of items (topics) even when we're not sure what we're looking for.



# Why topic modeling?

Topic modeling provides methods for automatically organizing, understanding, searching, and summarizing large electronic archives.

It can help with the following:

- discovering the hidden themes in the collection.
- classifying the documents into the discovered themes.
- using the classification to organize/summarize/search the documents.



# LDA

- It is one of the most popular **topic modeling methods**.
- Each document is made up of various words, and each topic also has various words belonging to it.
- The aim of LDA is to find topics a document belongs to, based on the words in it.



# Example:

- Let's say we have 2 topics that can be classified as *CAT\_related* and *DOG\_related*.
- A topic has probabilities for each word, so words such as *milk*, *meow*, and *kitten*, will have a higher probability in the *CAT\_related* topic than in the *DOG\_related* one.
- The *DOG\_related* topic, likewise, will have high probabilities for words such as *puppy*, *bark*, and *bone*.
- If we have a document containing the following sentences:
- “*Dogs* like to *chew* on *bones* and fetch sticks”.
- “*Puppies* drink *milk*.”
- “Both like to *bark*.”



- We can easily say it belongs to topic *DOG\_related* because it contains words *such as Dogs, bones, puppies, and bark.*
- Even though it contains the word *milk* which belongs to the topic *CAT\_related*, the document belongs to *DOG\_related as more words match with it.*



# How does LDA work?

There are 2 parts in LDA:

- The ***words that belong to a document***, that we already know.
- The ***words that belong to a topic*** or the probability of words belonging into a topic, that we need to calculate.



# Algorithm to find the latter

- Go through each document and randomly assign each word in the document to one of  $k$  topics ( $k$  is chosen beforehand).
- For each document  $d$ , go through each word  $w$  and compute :
- **$p(\text{topic } t \mid \text{document } d)$** : the **proportion of words in document  $d$  that are assigned to topic  $t$** . Tries to capture how many words belong to the topic  $t$  for a given document  $d$ . Excluding the current word.
- **$p(\text{word } w \mid \text{topic } t)$** : the **proportion of assignments to topic  $t$  over all documents that come from this word  $w$** . Tries to capture how many documents are in topic  $t$  because of word  $w$ .





- Update the probability for the word  $w$  belonging to topic  $t$ , as

$$p(\text{word } w \text{ with topic } t) = p(\text{topic } t \mid \text{document } d) * p(\text{word } w \mid \text{topic } t)$$