

Topic modelling only works if you have the right document collection

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Is the Neutrality of Technology Possible or Desirable?

How neutral are Topic Models ?

Topic modeling (Latent Dirichlet Allocation)

Goal: Determine the common themes or topics in a large text collection

- **Unsupervised**
 - Topic labels are not given
 - The number of topics needs to be pre-specified

Topic modeling: Generative probabilistic model

Assumption: documents are generated from an underlying, (latent) topic distribution and each document is generated from a mixture of these topics that each have a different proportion in the document.

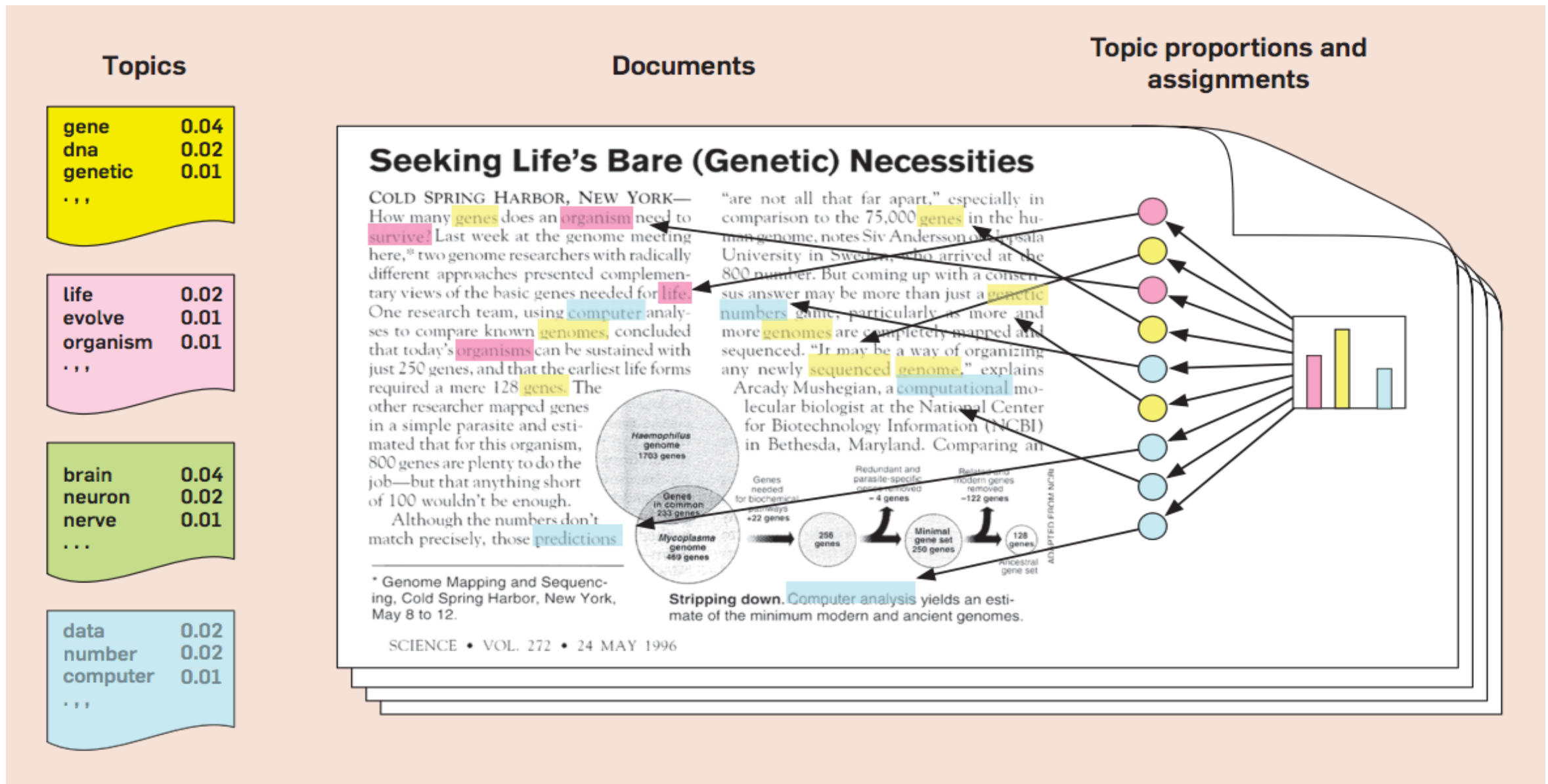
Topics defined as a distribution over words (a fixed vocabulary).

LDA uses an iterative process to estimate this underlying distribution based on the observed words in the text.

This model reflects the intuition that documents contain multiple topics.

Each document exhibits the topics in different proportion.

Generative model



picture from David Blei (2012)

Dream analysis

Despite the various research fields that study the meaning and purpose of dreams, such as psychiatry, psychology, neuroscience, and religious studies, a definitive explanation of the purpose of dreams is still lacking.

Psychologists and social scientists have studied dream content with quantitative methods for decades, working with the hypothesis that dreams reveal psychological information about the dreamer.

Continuity hypothesis: the content of dreams reflects a persons' daily life and personal concerns (Domhoff, 1996)

75--80\% of dream content relates to everyday settings, characters, and activities

topic modeling on dreambank.net (30k dream reports)

Random selection of the 50 topics:

44 money pay get give buy bank bill machine change

37 bathroom water toilet shower use clean bath floor sink

25 class school teacher students high test room classroom college

42 room door house see window open apartment go living

5 road hill tree see walking snow mountain trees people

28 love feel kiss make happy want man other hug

35 say says do see go man woman comes get

48 said did went came got told started saw looked asked

→ clear support for the **continuity hypothesis** as they reflect daily life events, characters and settings.

Guidelines for using LDA

- LDA only works on **large** unstructured document collections (no matter how long each document is)
- **document length:**
 - **too short:** LDA does not work
 - **too long:** useless (fraction will give the same result)
- **number of topics:** too many leads to inefficiency /no-convergence
- LDA works best if:
 - **topics** consist of a few high probable words
 - **documents** consist of only a small set of high probable topics
 - (and rest has low probability)
- be aware: each run with LDA will result in slightly different topics (random start)

(Tang et al, 2014)

How neutral are Topic Models?

Neutral:

- Unsupervised
- Reflect only the content collection and its word frequencies
- up to user to determine the value of the produced topics

But:

- Collection needs to adhere to certain prerequisites
- The assumptions and algorithmic implementation determine the usefulness of this method

Thank you for your attention!

References

- Blei, David M. "Probabilistic topic models." *Communications of the ACM* 55.4 (2012): 77-84.
- Tang, J., Meng, Z., Nguyen, X., Mei, Q., & Zhang, M. Understanding the Limiting Factors of Topic Modeling via Posterior Contraction Analysis. In: *ICML*. 2014. p. 190-198.