## **Latent Dirichlet Allocation**

**CPSC 503 - Pedagogical Project Final Presentation** 

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### **Outline**

#### **Latent Dirichlet Allocation (LDA)**

- 1. What is LDA?
- 2. The Posterior Distribution for LDA

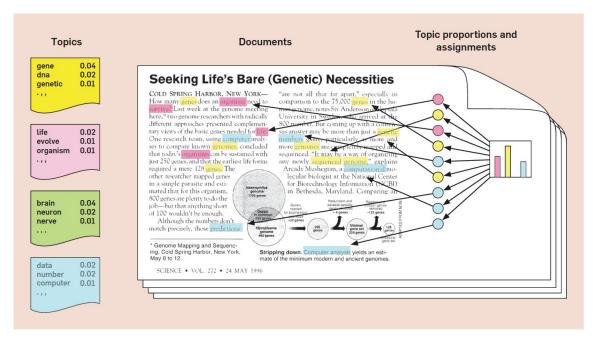
#### **Approximate Posterior Inference**

- 1. Gibbs Sampling
- 2. Variational Inference

## **Probabilistic Topic Modelling**

- 1. Treat data as observations that arise from a generative process that includes hidden variables
- 2. Infer the hidden structure using posterior inference  $\rightarrow$  topics
- Situate new data into the estimated model

### The intuition behind LDA

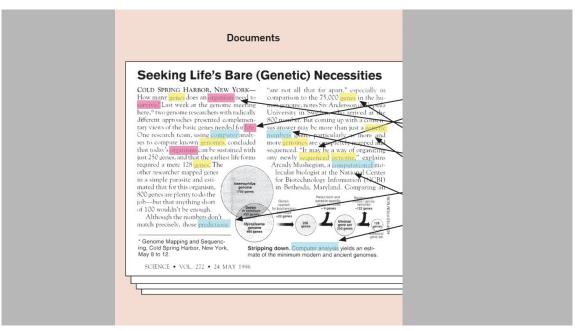


Each document is a random mixture of corpus-wide topics.

Each word is drawn from one of those topics.

Source: Blei, D. M. (2012). Probabilistic topic models. Communications of the ACM, 55(4), 77. https://doi.org/10.1145/2133806.2133826

#### The intuition behind LDA



In reality,

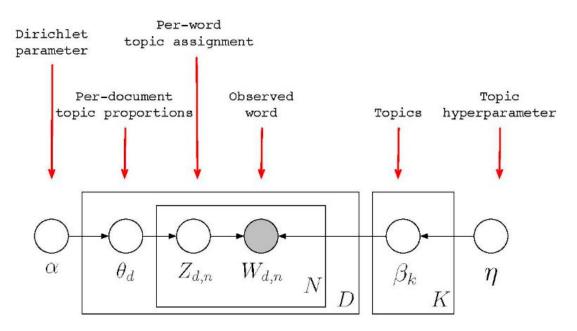
we only <u>observe the documents</u>

and

aim to infer the topic structure.

Source: Blei, D. M. (2012). Probabilistic topic models. Communications of the ACM, 55(4), 77. https://doi.org/10.1145/2133806.2133826

### **Graphical Model for LDA**



Source: Topic Model lecture by David Blei on <a href="http://videolectures.net/mlss09uk\_blei\_tm/">http://videolectures.net/mlss09uk\_blei\_tm/</a>

### **Approximate Posterior Inference**

There are various methods of algorithms for this purpose:

- Mean field variational methods (Blei et al., 2001, 2003)
- Expectation propagation (Minka and Lafferty, 2002)
- Collapsed Gibbs Sampling (Griffiths and Steyvers, 2002)
- Collapsed variational inference (Teh et al., 2006)

# **Applications in Informations Systems (IS)**

### **Evaluation of LDA**

### References