

# Latent Dirichlet Allocation



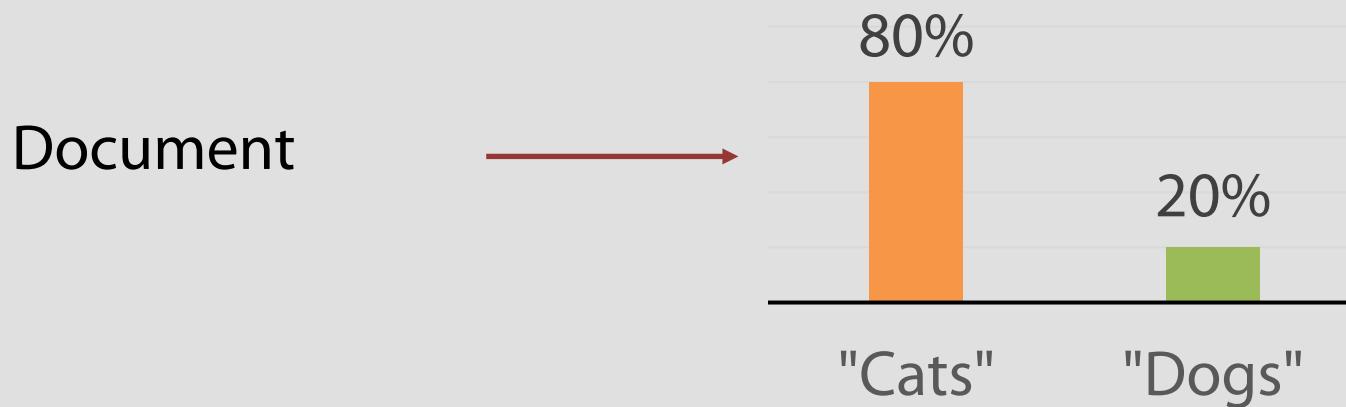
# Topics

Document is a distribution over topics



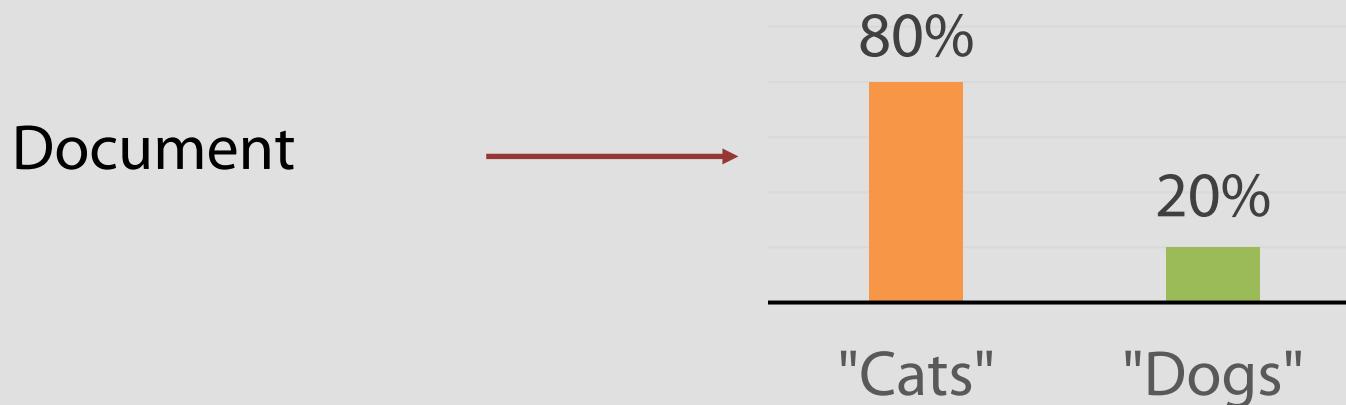
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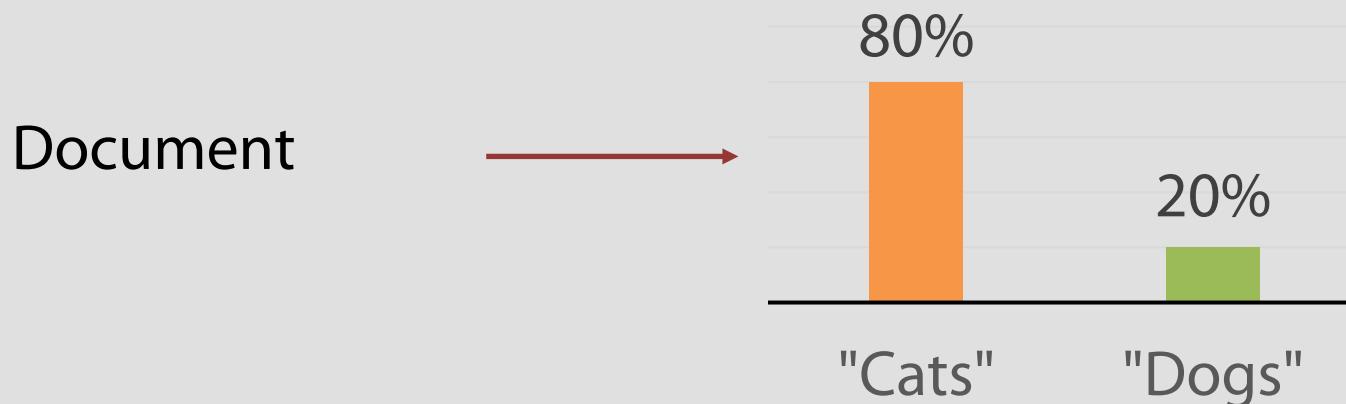


Topic is a distribution over words

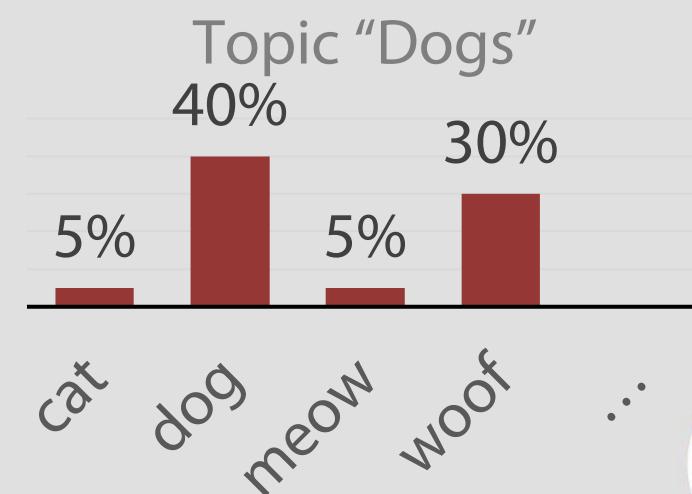
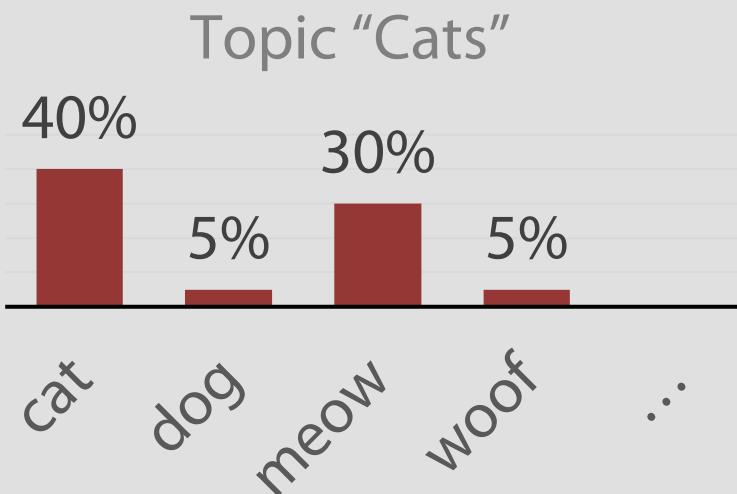


# Topics

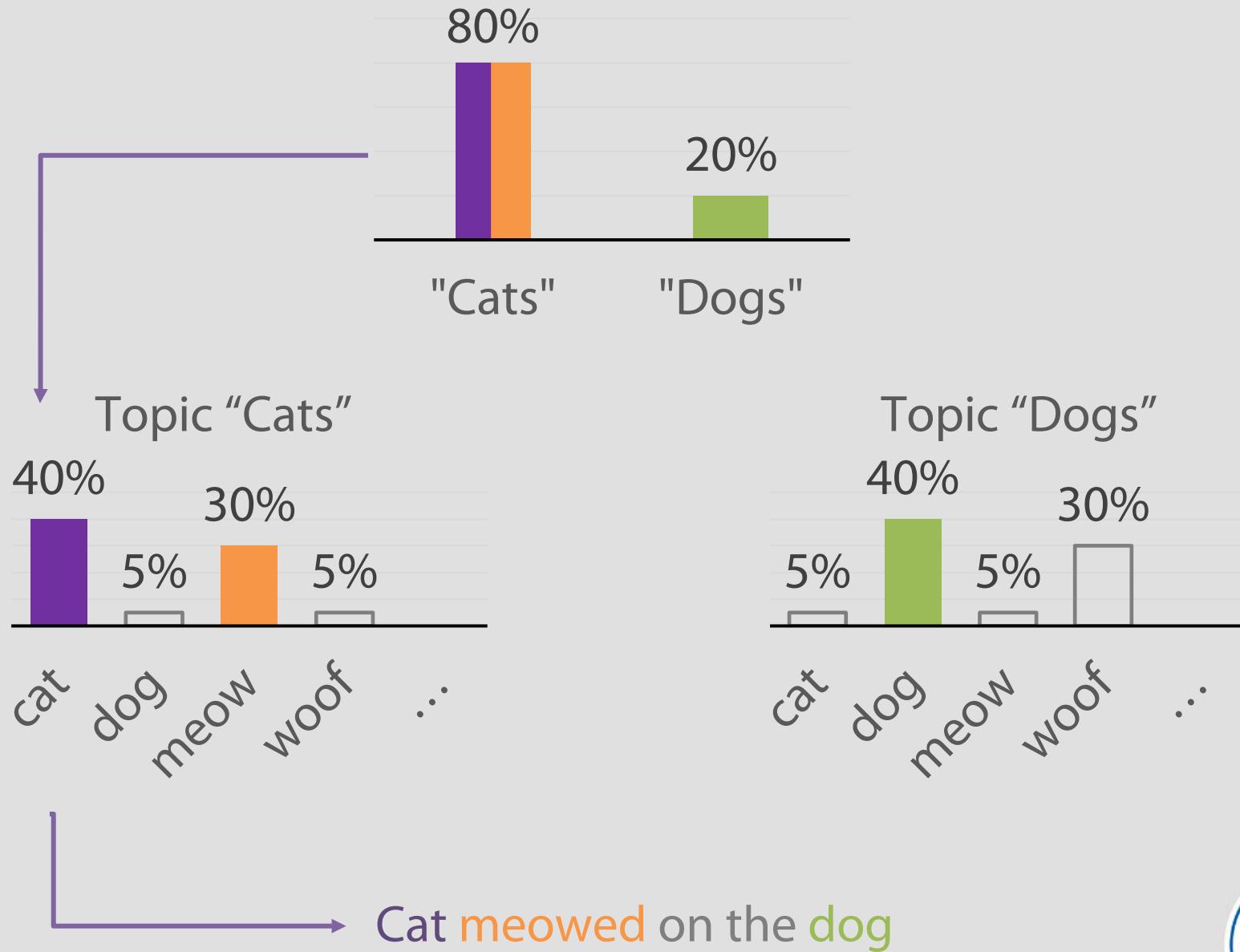
Document is a distribution over topics



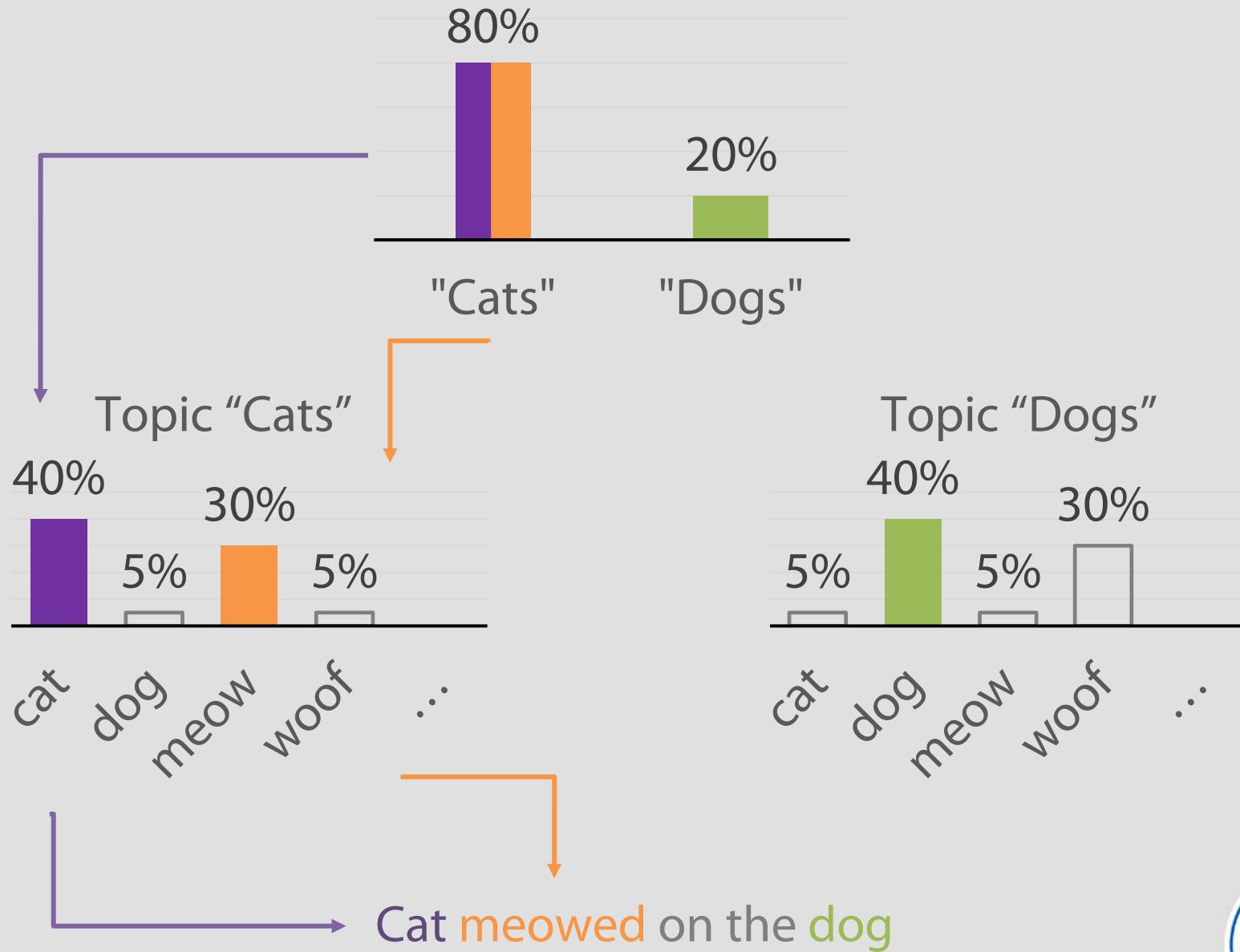
Topic is a distribution over words



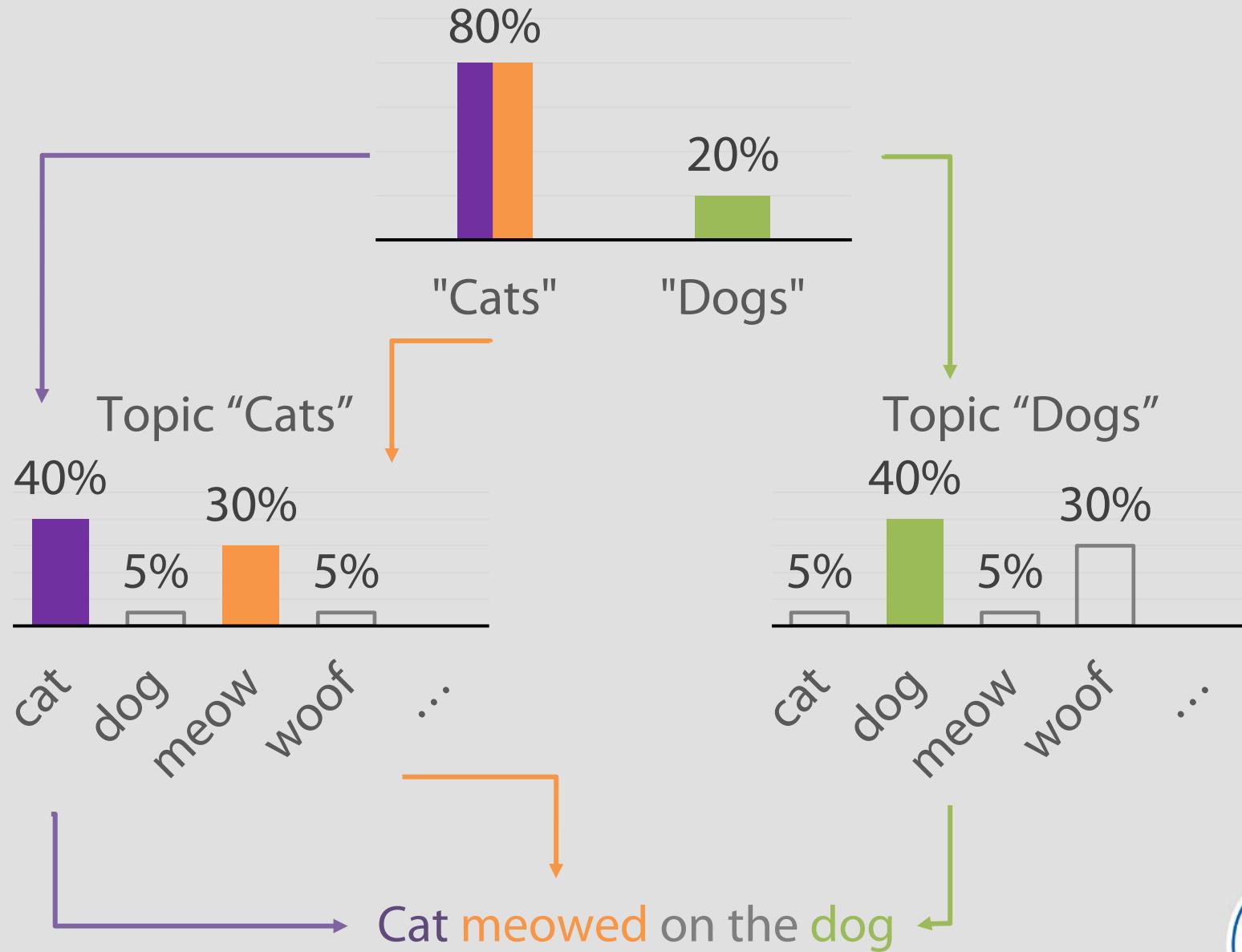
# Text generation



# Text generation



# Text generation



# Model

Distribution  
over topics

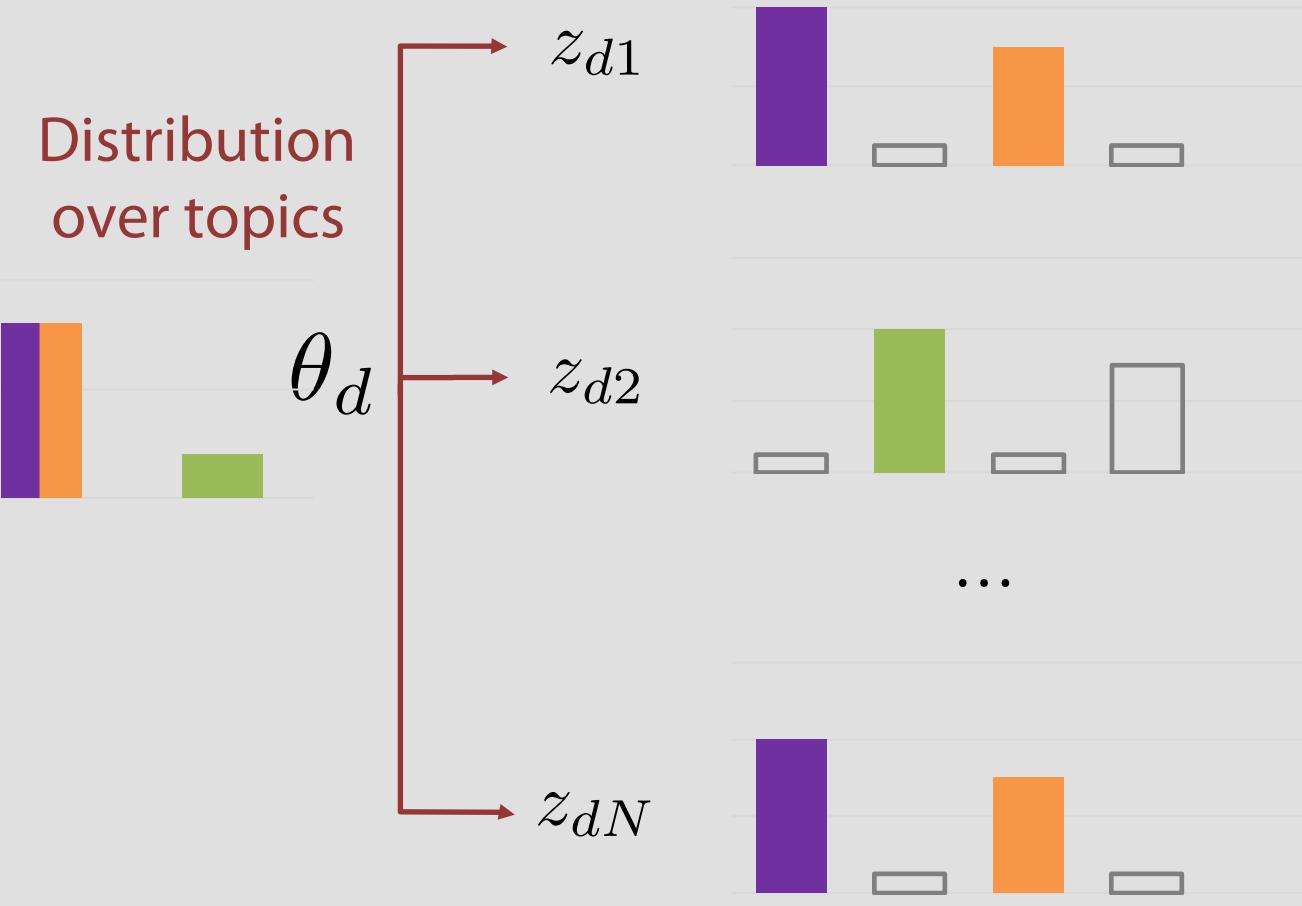


$\theta_d$



# Model

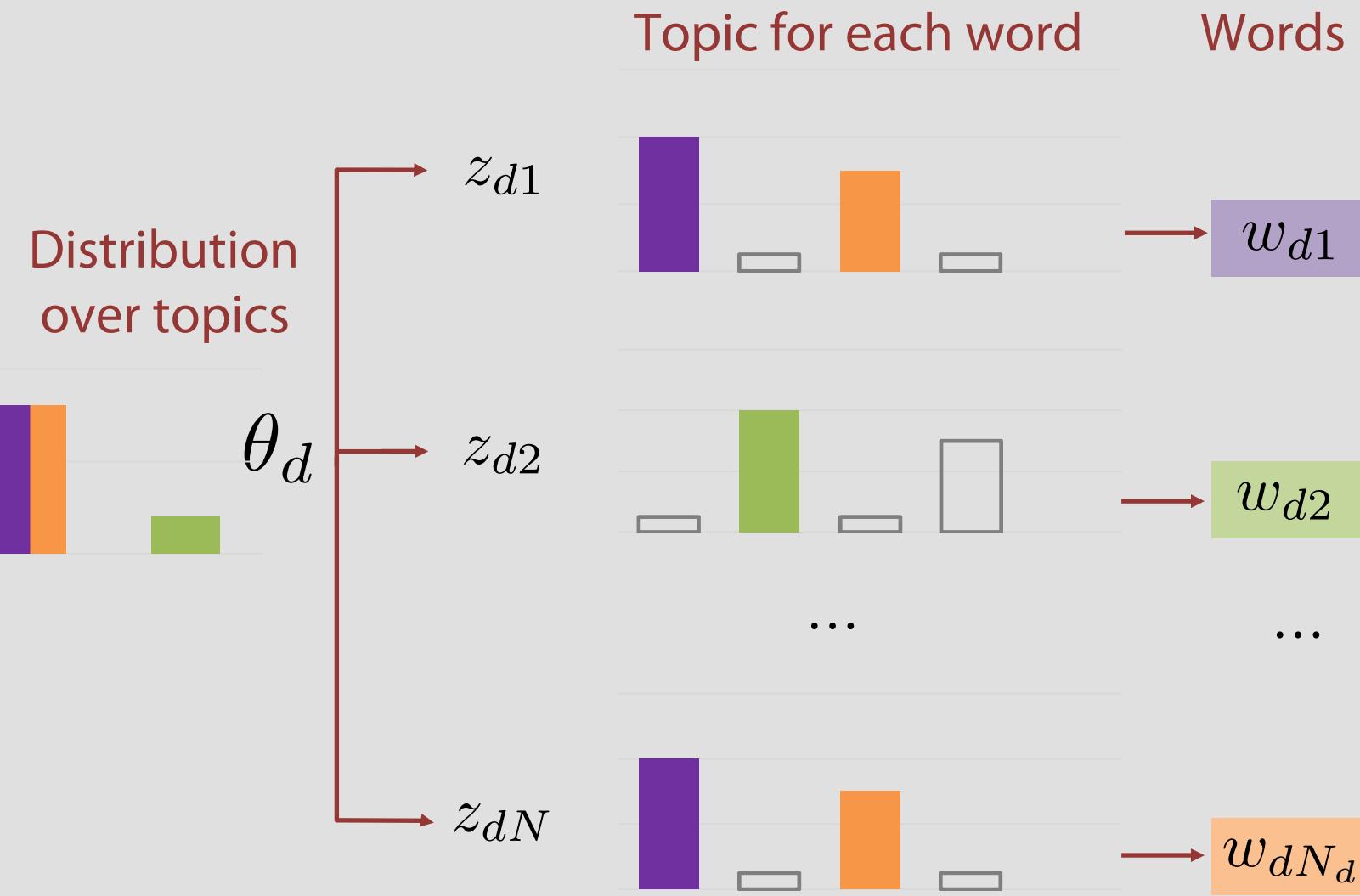
Topic for each word



$$z_{dn} \in \{1 .. T\}$$



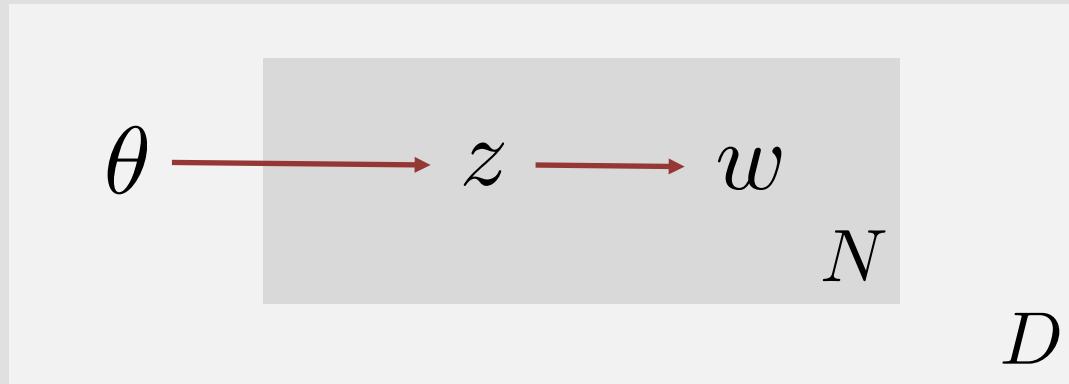
# Model



$$z_{dn} \in \{1..T\} \quad w_{dn} \in \{1..V\}$$



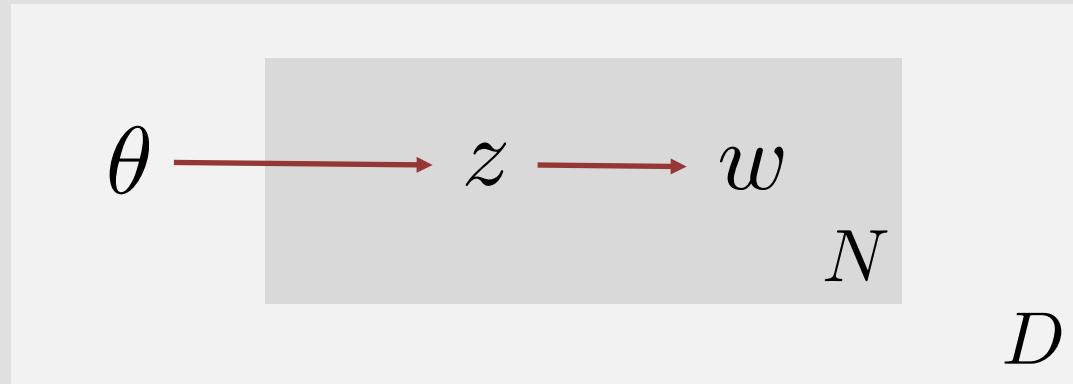
# LDA Model



$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$



# LDA Model

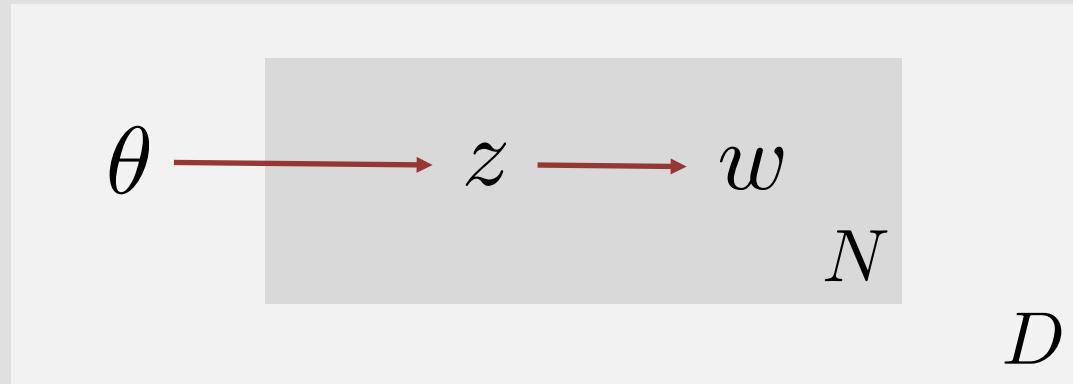


$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

for each document



# LDA Model

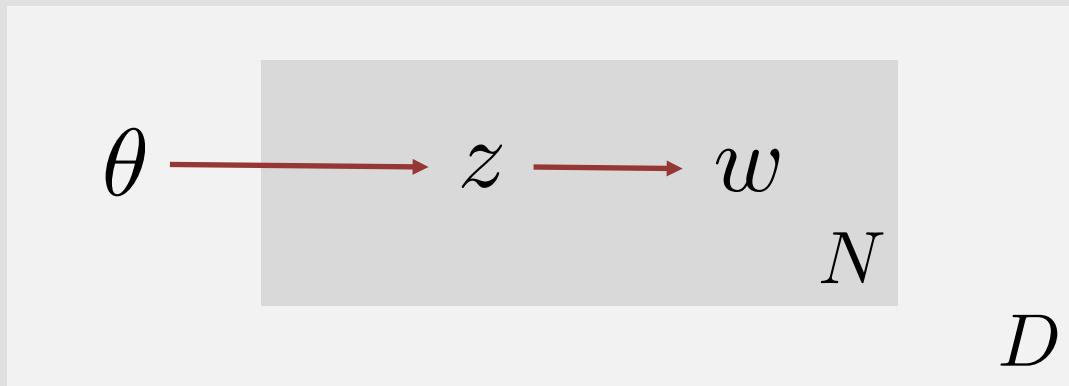


$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

for each document   ↑   ↑ generate topic probabilities



# LDA Model



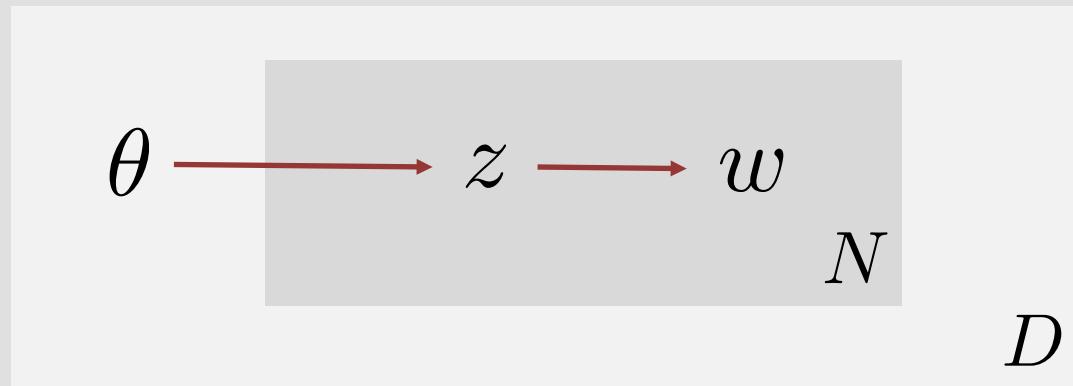
for each word

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

for each document      generate topic probabilities



# LDA Model

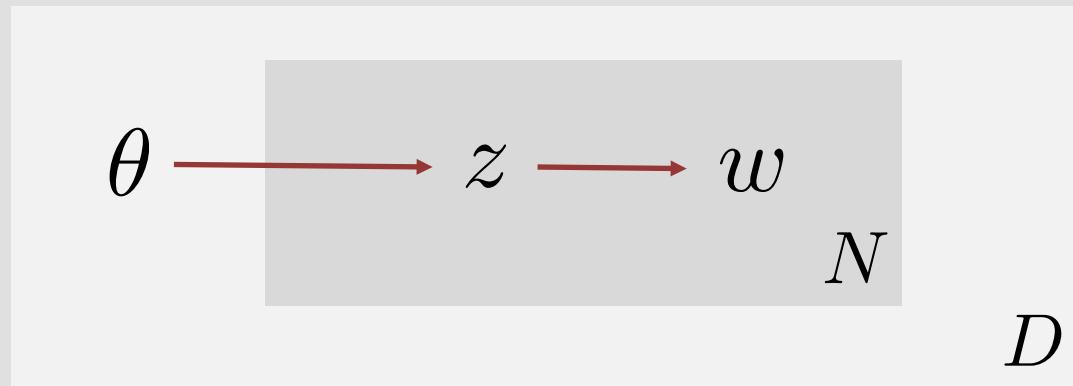


for each word      select topic

$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$

for each document      generate topic probabilities

# LDA Model



for each word      select topic

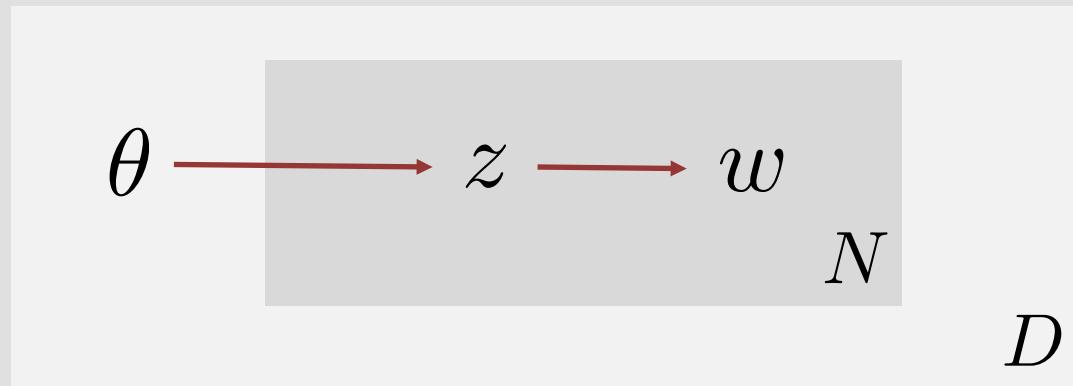
for each document      generate topic probabilities

select word from topic

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$



# LDA Model



for each word      select topic      select word from topic

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

for each document      generate topic probabilities



# LDA Model

$$p(\textcolor{red}{W}, \textcolor{green}{Z}, \textcolor{blue}{\Theta}) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$



# LDA Model

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

$$p(\theta_d) \sim \text{Dir}(\alpha)$$



# LDA Model

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

$$p(\theta_d) \sim \text{Dir}(\alpha)$$

$$p(z_{dn} | \theta_d) = \theta_{dz_{dn}}$$



# LDA Model

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

$$p(\theta_d) \sim \text{Dir}(\alpha)$$

$$p(z_{dn} | \theta_d) = \theta_{dz_{dn}}$$

$$p(w_{dn} | z_{dn}) = \Phi_{z_{dn} w_{dn}}$$



# LDA Model

$$p(W, Z, \Theta) = \prod_{d=1}^D p(\theta_d) \prod_{n=1}^{N_d} p(z_{dn} | \theta_d) p(w_{dn} | z_{dn})$$

$$p(\theta_d) \sim \text{Dir}(\alpha)$$

Constraints:

$$p(z_{dn} | \theta_d) = \theta_{dz_{dn}}$$

$$\Phi_{tw} \geq 0$$

$$p(w_{dn} | z_{dn}) = \Phi_{z_{dn} w_{dn}} \longleftarrow$$

$$\sum_w \Phi_{tw} = 1$$



# LDA Model

**Known:**  $W$  data

**Unknown:**  $\Phi$  parameters, distribution over words for each topic

**Unknown:**  $Z$  latent variables, topic of each word

**Unknown:**  $\Theta$  latent variables, distribution over topics for each document



# ТЕХНИЧЕСКИЙ СЛАЙД (15 мин на доску)

- ВЫВОД ФОРМУЛ VAR. ЕМ НА ДОСКЕ

