



# NLP section 5

Text Classification



# Text Classification

- Text Classification is the task of assigning a label or class to a given text
- Why text classification?
  - 1- Spam/Not Spam
  - 2- Positive/Negative
  - 3- Fake/Real
  - 4- Bot/Human



## Multinomial Naive Bayes: Bayesian Theorem

$$P(A|B) = \frac{P(B|A) * P(A)}{P(B)}$$



## Multinomial Naive Bayes: Cont.

$$P(c | d) = \frac{P(d | c)P(c)}{P(d)}$$

$$c_{MAP} = \operatorname{argmax}_{c \in C} P(c | d) = \operatorname{argmax}_{c \in C} \frac{P(d | c)P(c)}{\cancel{P(d)}} = \operatorname{argmax}_{c \in C} P(x_1, x_2, \dots, x_n | c)P(c)$$

$$c_{NB} = \operatorname{argmax}_{c \in C} P(c_j) \prod_{x \in X} P(x | c) = \operatorname{argmax}_{c_j \in C} \left[ \log P(c_j) + \sum_{i \in \text{positions}} \log P(x_i | c_j) \right]$$

## Multinomial Naive Bayes: Cont.

$$\hat{P}(c_j) = \frac{\text{doccount}(C = c_j)}{N_{\text{doc}}}$$

$$\hat{P}(w_i | c_j) = \frac{\text{count}(w_i, c_j)}{\sum_{w \in V} \text{count}(w, c_j)} \xrightarrow{\text{Laplace for Smoothing}} \hat{P}(w_i | c) = \frac{\text{count}(w_i, c) + 1}{\sum_{w \in V} (\text{count}(w, c) + 1)}$$

$$\sum_{w \in V} (\text{count}(w, c) + 1) \quad \left( \sum_{w \in V} \text{count}(w, c) \right) + |V|$$



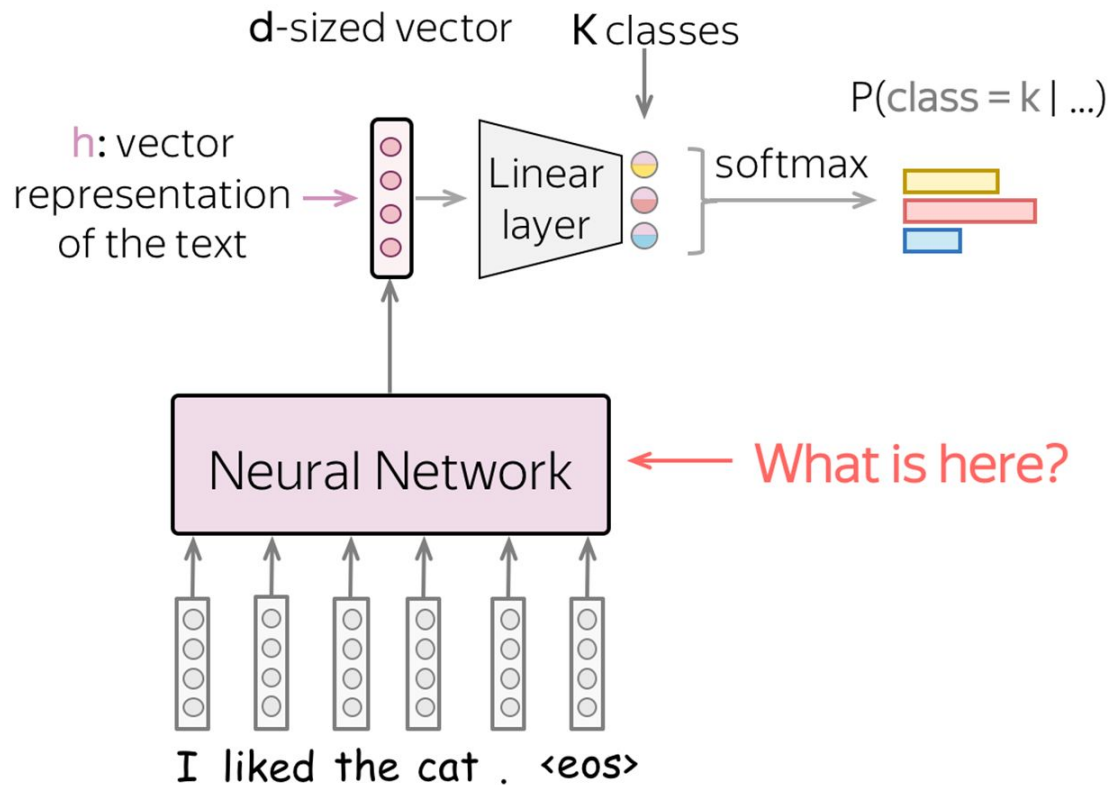
## Multinomial Naive Bayes: Example

Email	Class	Content
1	Spam	"buy cheap toys now"
2	Spam	"cheap replica watches buy"
3	Spam	"buy toys cheap"
4	Ham	"meeting schedule now"
5	Ham	"project meeting tomorrow"
6	Ham	"schedule project review"

Text to Classify:

- Schedule meeting review
- Buy cheap watches now

# Neural Network





# Code

- Multinomial Naive Bayes
- Neural Network
- Neural Network with GloVe