

# Naive Bayes' classifier

- Assumption 1: The text and label in one document does not affect those of another.
- Assumption 2: Words in a sentence are independent, conditioned on the class label

- Taken from Eisenstein, 2019, Chp 2

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**Algorithm 1** Generative process for the Naïve Bayes classification model

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**for** Instance  $i \in \{1, 2, \dots, M\}$  **do**:

    Draw the label  $y^{(i)} \sim \text{Categorical}(\boldsymbol{\mu})$ ;

    Draw the word counts  $\mathbf{x}^{(i)} \mid y^{(i)} \sim \text{Multinomial}(\boldsymbol{\phi}_{y^{(i)}})$ .

chain rule  $P(\mathbf{x}, y) = P(\mathbf{x} \mid y)P(y)$

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where  $\boldsymbol{\mu} = [\mu_1, \dots, \mu_K]$  label probability

$\boldsymbol{\phi} = [\phi_1, \dots, \phi_N]$  word probability

$$P_{mult}(\mathbf{x} \mid y; \boldsymbol{\phi}) = B(\mathbf{x}) \prod_{j=1}^N \phi_j^{\mathbf{x}_j} \quad B(\mathbf{x}) = \frac{(\sum_{j=1}^N \mathbf{x}_j)!}{\prod_{j=1}^N \mathbf{x}_j!}$$

# Naive Bayes' classifier prediction step

$$\begin{aligned}\hat{y} &= \operatorname{argmax}_y \log P(\mathbf{x}, y; \boldsymbol{\mu}, \boldsymbol{\phi}) \\ &= \operatorname{argmax}_y \log P(\mathbf{x} | y; \boldsymbol{\mu}, \boldsymbol{\phi}) + \log P(y; \boldsymbol{\mu}) \\ &= \operatorname{argmax}_y \log \left[ B(\mathbf{x}) \prod_{j=1}^N \phi_j^{\mathbf{x}_j} \right] + \log \mu_y \\ &= \operatorname{argmax}_y \log B(\mathbf{x}) + \sum_{j=1}^N x_j \log \phi_{y,j} + \log \mu_y \\ &= \operatorname{argmax}_y \sum_{j=1}^N x_j \log \phi_{y,j} + \log \mu_y\end{aligned}$$