



# BAYESIAN LEARNING: NAIVE BAYES

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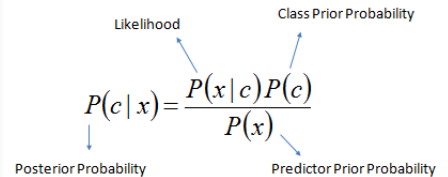
The materials are compiled from the following resources:

- <https://github.com/joaquinvanschoren/ML-course>
- [https://www.cse.iitk.ac.in/users/piyush/courses/ml\\_autumn16/ML.html](https://www.cse.iitk.ac.in/users/piyush/courses/ml_autumn16/ML.html)
- <http://sli.ics.uci.edu/Classes/2015W-273a>

## NAIVE BAYES BASICS

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- Predict the probability that a point belongs to each class, using Bayes' Theorem, assuming that the features are independent from each other.
- Very fast. They work by only extracting statistics from each feature.



The diagram shows the Bayes' Theorem formula: 
$$P(c|x) = \frac{P(x|c)P(c)}{P(x)}$$
 with four blue arrows pointing from labels to the formula components: 'Likelihood' points to  $P(x|c)$ , 'Class Prior Probability' points to  $P(c)$ , 'Posterior Probability' points to  $P(c|x)$ , and 'Predictor Prior Probability' points to  $P(x)$ .

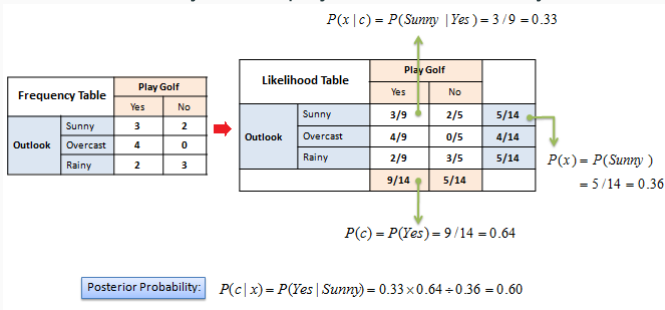
- $P(c|x)$  is the posterior probability of class (target) given predictor (attribute).
- $P(c)$  is the prior probability of class.
- $P(x|c)$  is the likelihood which is the probability of predictor given class.
- $P(x)$  is the prior probability of predictor (evidence or marginal likelihood).
- Naive Bayes assumes that all features are conditionally independent from each other, in which case:

$$P(\mathbf{x}|c) = P(x_1|c) \times P(x_2|c) \times \dots \times P(x_n|c)$$

- $M \ll 2^N$  parameters
- $y \in 2\{\text{spam}, \text{notspam}\}$
- $X$  = observed words in email
  - Ex: ["the" ... "probabilistic" ... "lottery" ...]
  - "1" if word appears; "0" if not
- 1000's of possible words:  $2^{1000}$  parameters?
- # of atoms in the universe:  $\gg 2^{270}$
- Model words given email type as independent
- Some words more likely for spam ("lottery")
- Some more likely for real ("Tugas kuliah")
- Only 1000's of parameters now

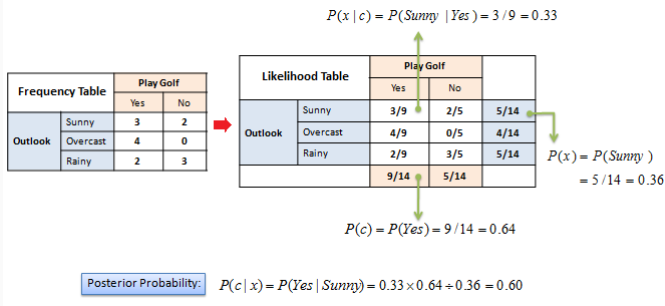
- True or not? Players will play if weather is sunny.

- True or not? Players will play if weather is sunny.





- True or not? Players will play if weather is sunny.



- Compute the posterior for every class and predict the class with highest probability

## GAUSSIAN NAIVE BAYES

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- Computes mean  $\mu_c$  and standard deviation  $\sigma_c$  of the feature values per class
- It then fits a Gaussian distribution around the mean

$$p(x = v \mid c) = \frac{1}{\sqrt{2\pi\sigma_c^2}} e^{-\frac{(v-\mu_c)^2}{2\sigma_c^2}}$$

- Prediction are made using Bayes' theorem, by computing the joint probability given all features

$$p(c \mid \mathbf{x}) = \frac{p(c) p(\mathbf{x} \mid c)}{p(\mathbf{x})}$$

- BernoulliNB
  - Assumes binary data
  - Feature statistics: Number of non-zero entries per class
- MultinomialNB
  - Assumes count data
  - Feature statistics: Average value per class

Mostly used for text classification (bag-of-words data)