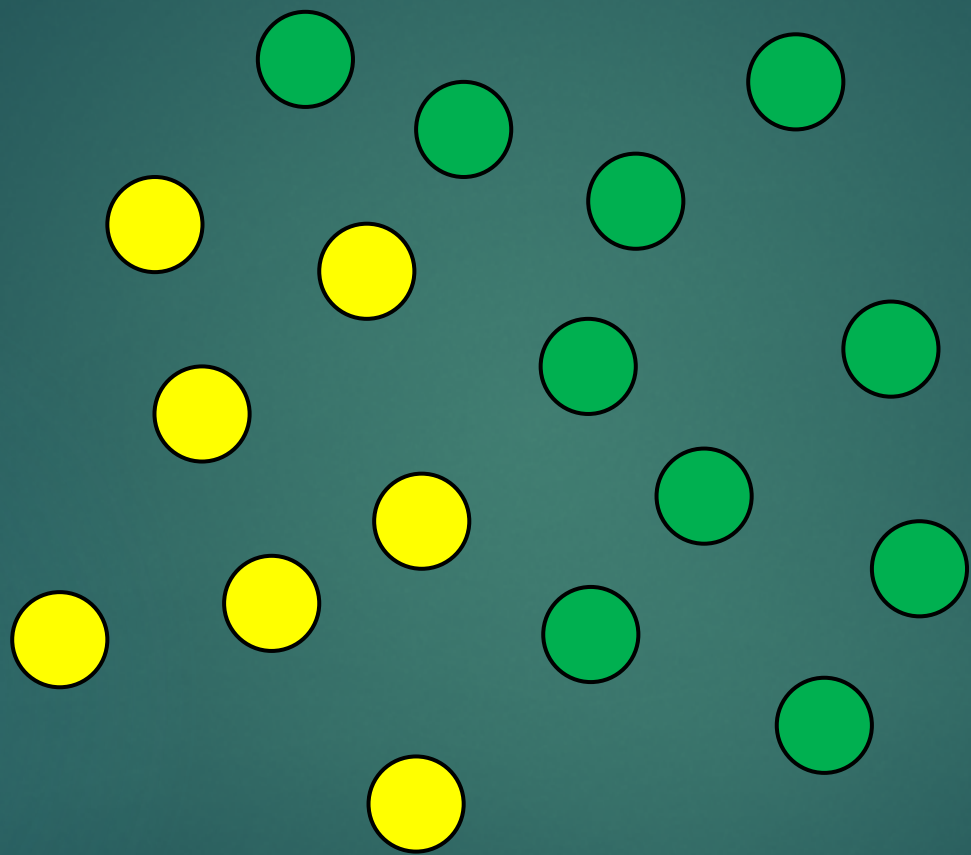


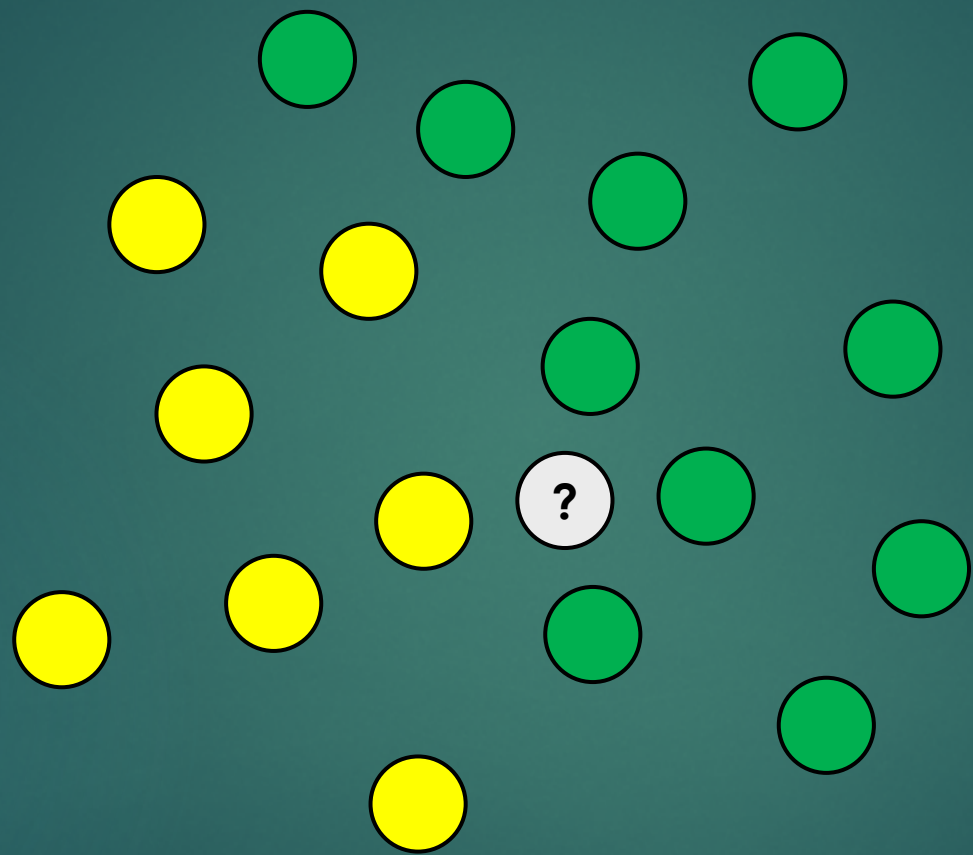
MACHINE LEARNING

NAIVE BAYES

Naive Bayes classifier

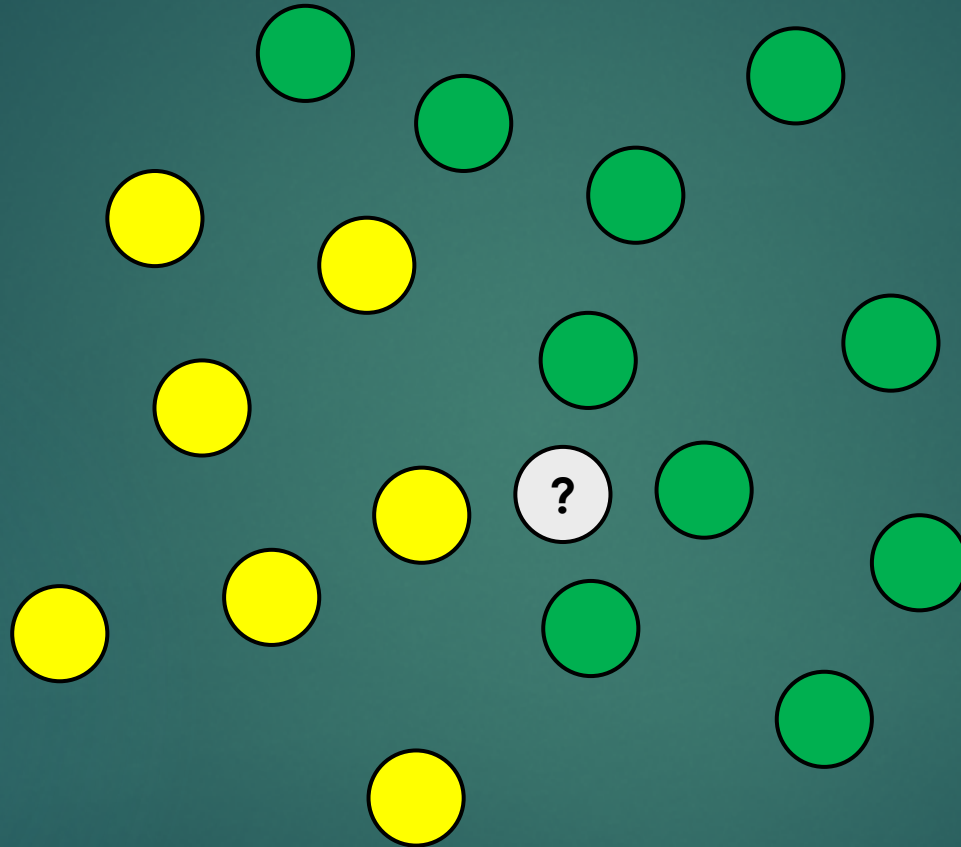
- ▶ Very efficient supervised learning algorithm
- ▶ It scales well even in high dimensions !!!
- ▶ It is able to compete with **SVM** or **random forest** classifiers
- ▶ It is able to make good predictions even when the training data is relatively small
- ▶ **Why is it naive?**
- ▶ The naive assumption is that every pair of features are independent





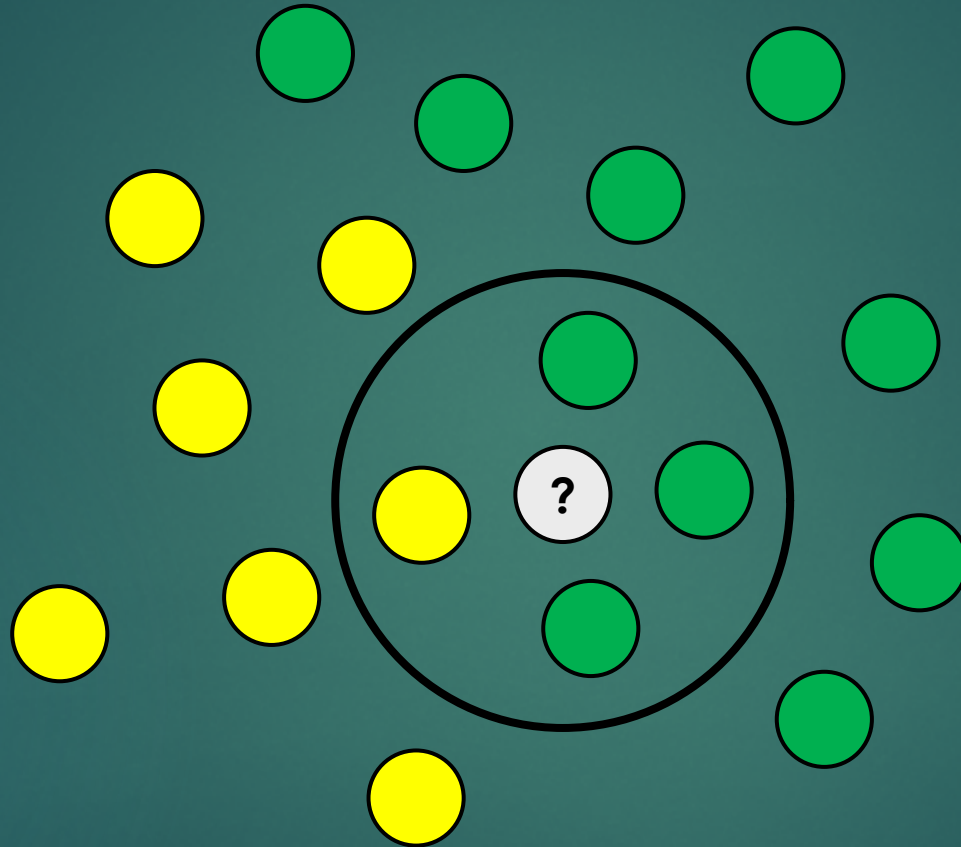
$$P(\text{yellow}) = \frac{7}{17}$$

$$P(\text{green}) = \frac{10}{17}$$



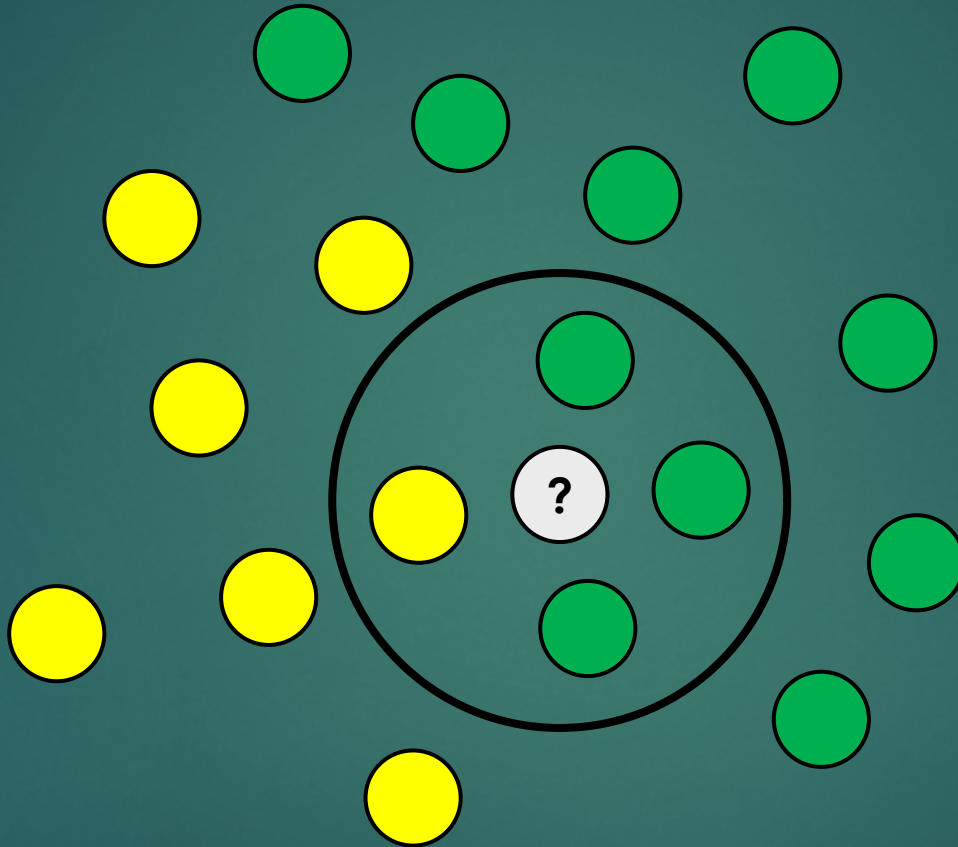
$$P(\text{yellow}) = \frac{7}{17}$$

$$P(\text{green}) = \frac{10}{17}$$



$$P(\text{yellow}) = \frac{7}{17}$$

$$P(\text{green}) = \frac{10}{17}$$

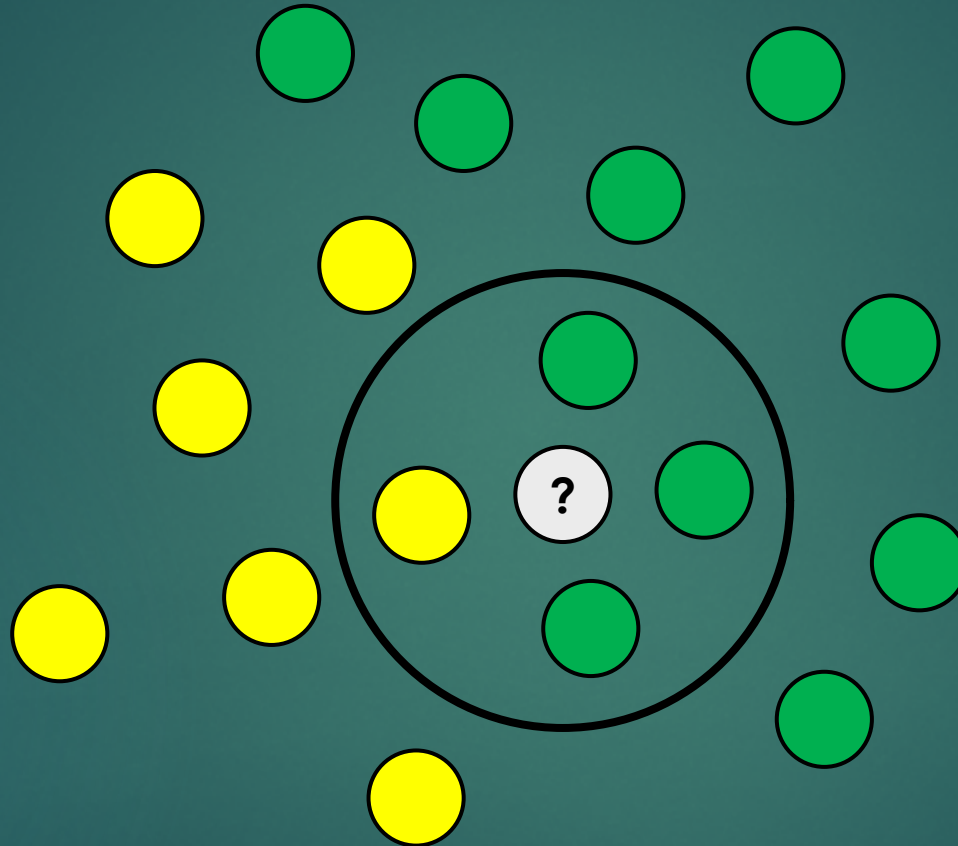


$$P'(? \mid \text{green}) = \frac{3}{10}$$

$$P'(? \mid \text{yellow}) = \frac{1}{7}$$

$$P(\text{yellow}) = \frac{7}{17}$$

$$P(\text{green}) = \frac{10}{17}$$



$$P'(? \mid \text{green}) = \frac{3}{10}$$

$$P'(? \mid \text{yellow}) = \frac{1}{7}$$

posterior probability

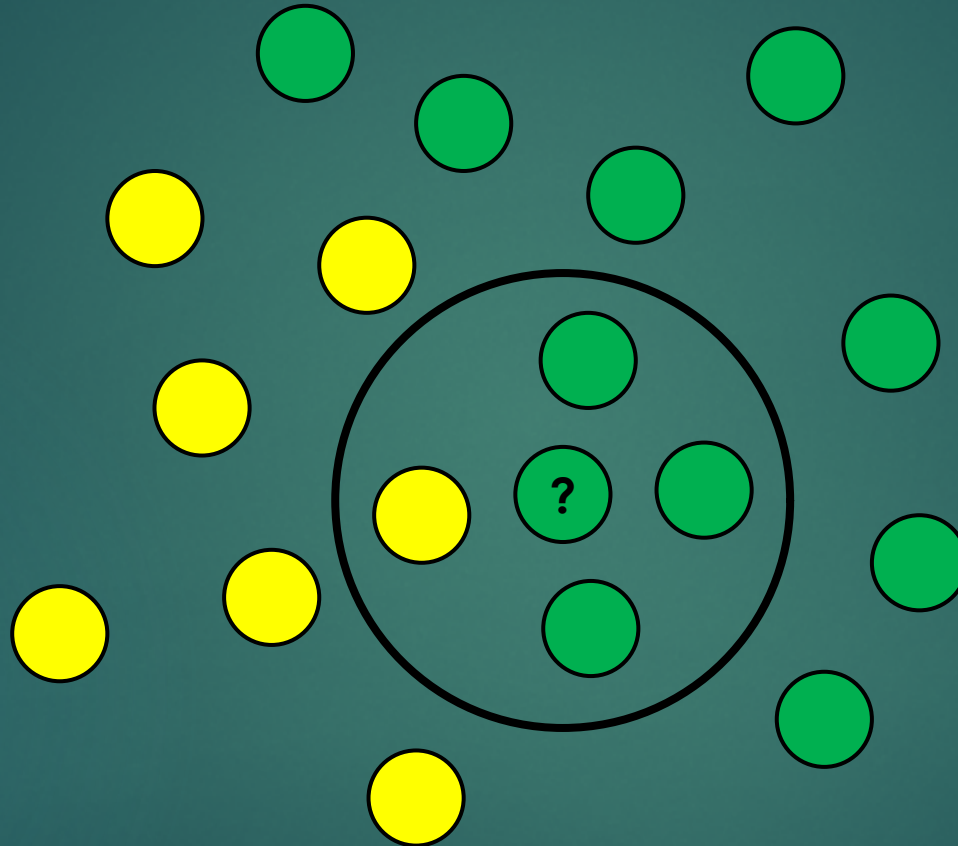
$$P''(? \text{ is green}) = P(\text{green}) * P'(? \mid \text{green}) = \frac{10}{17} * \frac{3}{10} = \frac{30}{170}$$

posterior probability

$$P''(? \text{ is yellow}) = P(\text{yellow}) * P'(? \mid \text{yellow}) = \frac{7}{17} * \frac{1}{7} = \frac{7}{119}$$

$$P(\text{yellow}) = \frac{7}{17}$$

$$P(\text{green}) = \frac{10}{17}$$



$$P'(? \mid \text{green}) = \frac{3}{10}$$

$$P'(? \mid \text{yellow}) = \frac{1}{7}$$

posterior probability

$$P''(? \text{ is green}) = P(\text{green}) * P'(? \mid \text{green}) = \frac{10}{17} * \frac{3}{10} = \frac{30}{170}$$

posterior probability

$$P''(? \text{ is yellow}) = P(\text{yellow}) * P'(? \mid \text{yellow}) = \frac{7}{17} * \frac{1}{7} = \frac{7}{119}$$

Pros and cons

ADVANTAGES

- relatively simple to understand
- it can be trained on small datasets as well
- it is a fast approach
- it is not sensitive to irrelevant features

DISADVANTAGES

- it assumes every feature is independent → it is not always true !!!