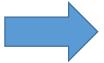


## Supervised classification

# A classification problem

You have a bottle of wine whose label is missing.





Which winery is it from, 1, 2, or 3?

Solve this problem using visual and chemical features of the wine.

# wine\_data.txt

#### Training set obtained from 130 bottles

Winery 1: 43 bottles

Winery 2: 51 bottles

Winery 3: 36 bottles

For each bottle, 13 features:

'Alcohol', 'Malic acid', 'Ash', 'Alcalinity of ash', 'Magnesium',

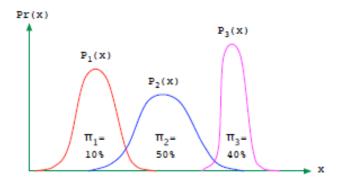
'Total phenols', 'Flavanoids', 'Nonflavanoid phenols',

'Proanthocyanins',

'Color intensity', 'Hue', 'OD280/OD315 of diluted wines',

'Proline'

Also, a separate test set of 48 labeled points.



For any data point  $x \in \mathcal{X}$  and any candidate label j,

$$\Pr(y = j | x) = \frac{\Pr(y = j) \Pr(x | y = j)}{\Pr(x)} = \frac{\pi_j P_j(x)}{\Pr(x)}$$

Optimal prediction: the class j with largest  $\pi_i P_i(x)$ .

### Wzór Bayesa [edytuj|edytuj kod]

Twierdzenie (wzór) Bayesa w swej podstawowej formie mówi, że<sup>[1]</sup>

$$P(A \mid B) = \frac{P(B \mid A) P(A)}{P(B)},$$

gdzie A i B są zdarzeniami oraz  $\mathsf{P}(B)>0$ , przy czym

- $P(A \mid B)$  oznacza prawdopodobieństwo warunkowe, tj. prawdopodobieństwo zajścia zdarzenia A, o ile zajdzie zdarzenie B.
- $\mathsf{P}(B \mid A)$  oznacza prawdopodobieństwo zajścia zdarzenia B, o ile zajdzie zdarzenie A.

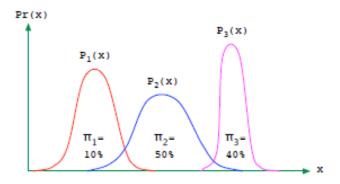
Training set of 130 bottles:

- Winery 1: 43 bottles, winery 2: 51 bottles, winery 3: 36 bottles
- For each bottle, 13 features: 'Alcohol', 'Malic acid', 'Ash',
  'Alcalinity of ash', 'Magnesium', 'Total phenols', 'Flavanoids',
  'Nonflavanoid phenols', 'Proanthocyanins', 'Color intensity',
  'Hue', 'OD280/OD315 of diluted wines', 'Proline'

#### Class weights:

$$\pi_1 = 43/130 = 0.33, \quad \pi_2 = 51/130 = 0.39, \quad \pi_3 = 36/130 = 0.28$$

Need distributions  $P_1$ ,  $P_2$ ,  $P_3$ , one per class. Base these on a single feature: 'Alcohol'.



For any data point  $x \in \mathcal{X}$  and any candidate label j,

$$\Pr(y = j | x) = \frac{\Pr(y = j)\Pr(x | y = j)}{\Pr(x)} = \frac{\pi_j P_j(x)}{\Pr(x)}$$

Optimal prediction: the class j with largest  $\pi_i P_i(x)$ .

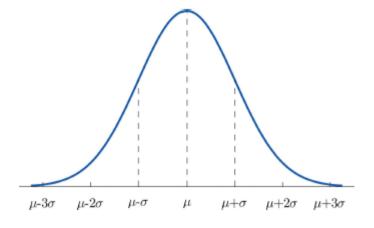
Training set of 130 bottles:

- Winery 1: 43 bottles, winery 2: 51 bottles, winery 3: 36 bottles
- For each bottle, 13 features: 'Alcohol', 'Malic acid', 'Ash',
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  'Nonflavanoid phenols', 'Proanthocyanins', 'Color intensity',
  'Hue', 'OD280/OD315 of diluted wines', 'Proline'

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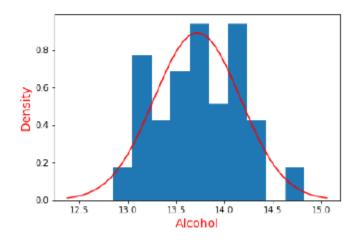


The Gaussian  $N(\mu, \sigma^2)$  has mean  $\mu$ , variance  $\sigma^2$ , and density function

$$p(x) = \frac{1}{(2\pi\sigma^2)^{1/2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right).$$

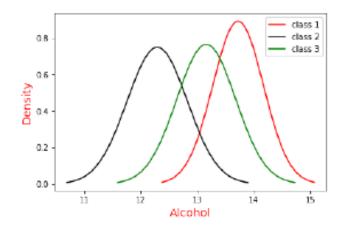
# The distribution for winery 1

Single feature: 'Alcohol'



Mean  $\mu=$  13.72, Standard deviation  $\sigma=$  0.44 (variance 0.20)

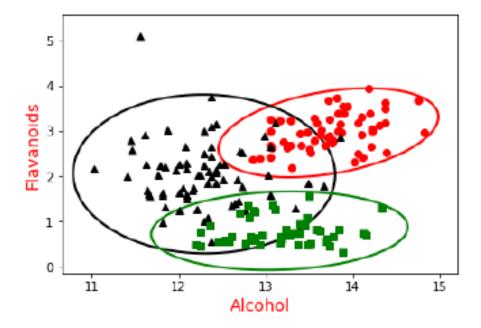
# winary 3

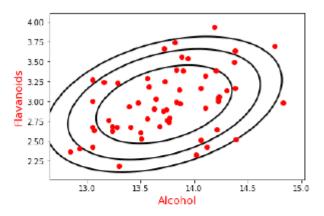


- $\pi_1 = 0.33$ ,  $P_1 = N(13.7, 0.20)$
- $\pi_2 = 0.39$ ,  $P_2 = N(12.3, 0.28)$
- $\pi_3 = 0.28$ ,  $P_3 = N(13.2, 0.27)$

To classify x: Pick the j with highest  $\pi_j P_j(x)$ 

### 2 features





Model class 1 by a bivariate Gaussian, parametrized by:

mean 
$$\mu = \begin{pmatrix} 13.7 \\ 3.0 \end{pmatrix}$$
 and covariance matrix  $\Sigma = \begin{pmatrix} 0.20 & 0.06 \\ 0.06 & 0.12 \end{pmatrix}$ 

- Mean  $(\mu_1,\mu_2)\in\mathbb{R}^2$ , where  $\mu_1=\mathbb{E}(X_1)$  and  $\mu_2=\mathbb{E}(X_2)$
- Covariance matrix  $\Sigma = \left[ egin{array}{ccc} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{array} \right]$

Density 
$$p(x_1, x_2) = \frac{1}{2\pi |\Sigma|^{1/2}} \exp\left(-\frac{1}{2} \begin{bmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{bmatrix}^T \Sigma^{-1} \begin{bmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{bmatrix}\right)$$