

Introduction to Machine Learning with Python

Supervised Learning: Naive Bayes and Tree based Algorithms

Dr. Süha Tuna
İTÜ Informatics Institute

The Models to be Covered

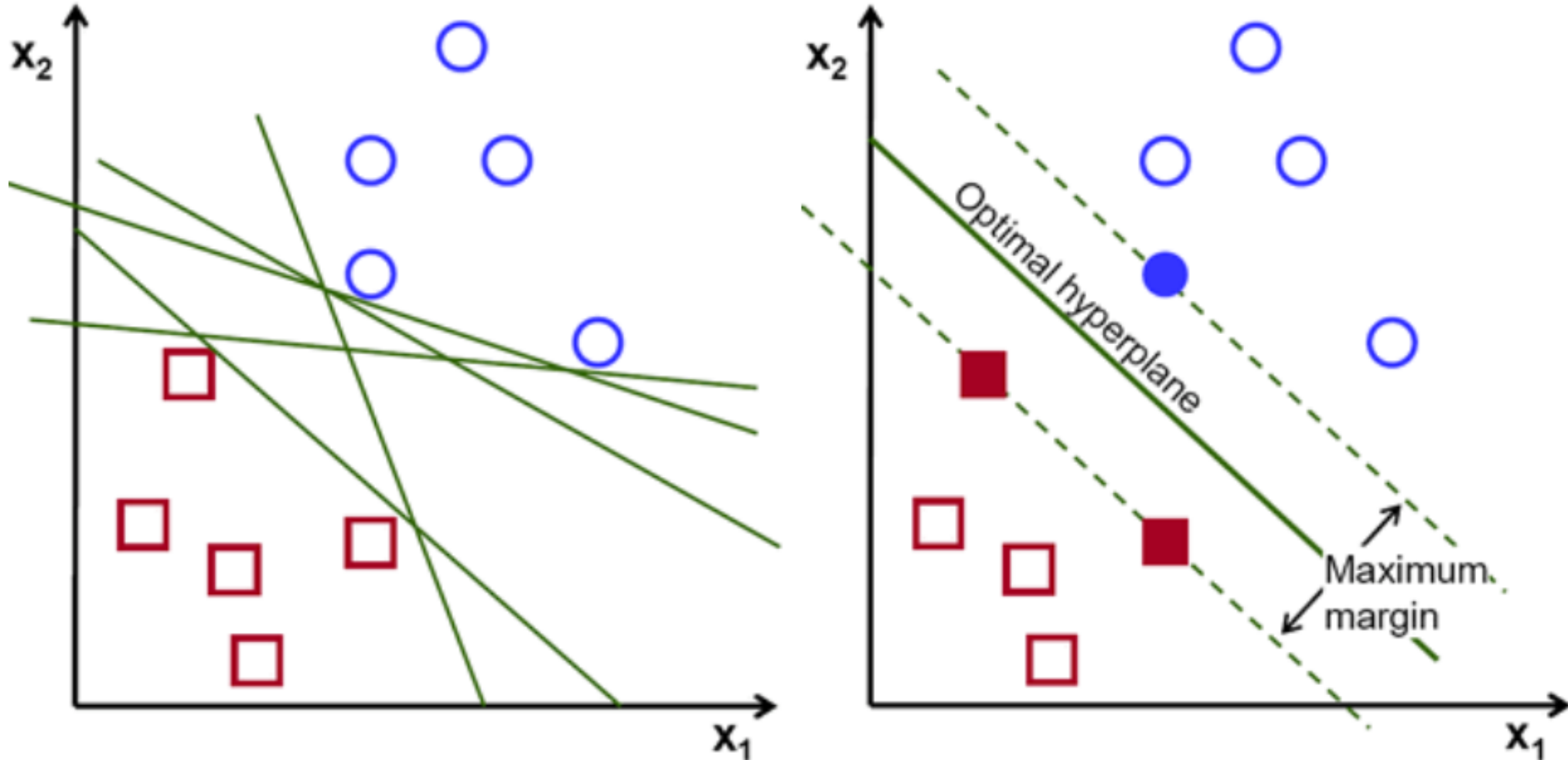
Linear Support
Vector
Classifier

Logistic
Regression

Decision Trees

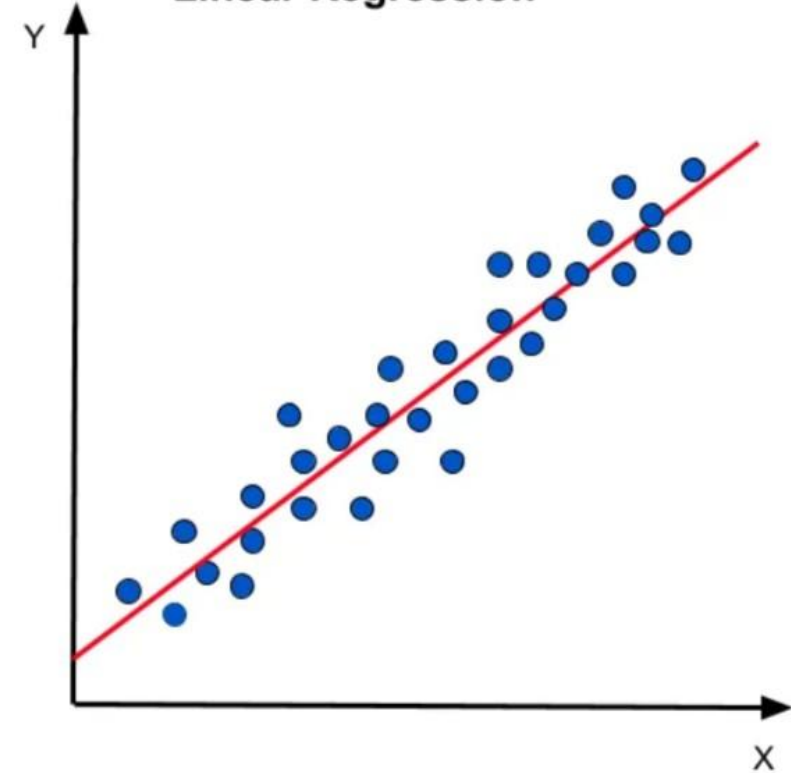
Naive Bayes

Linear Support Vector Classifier

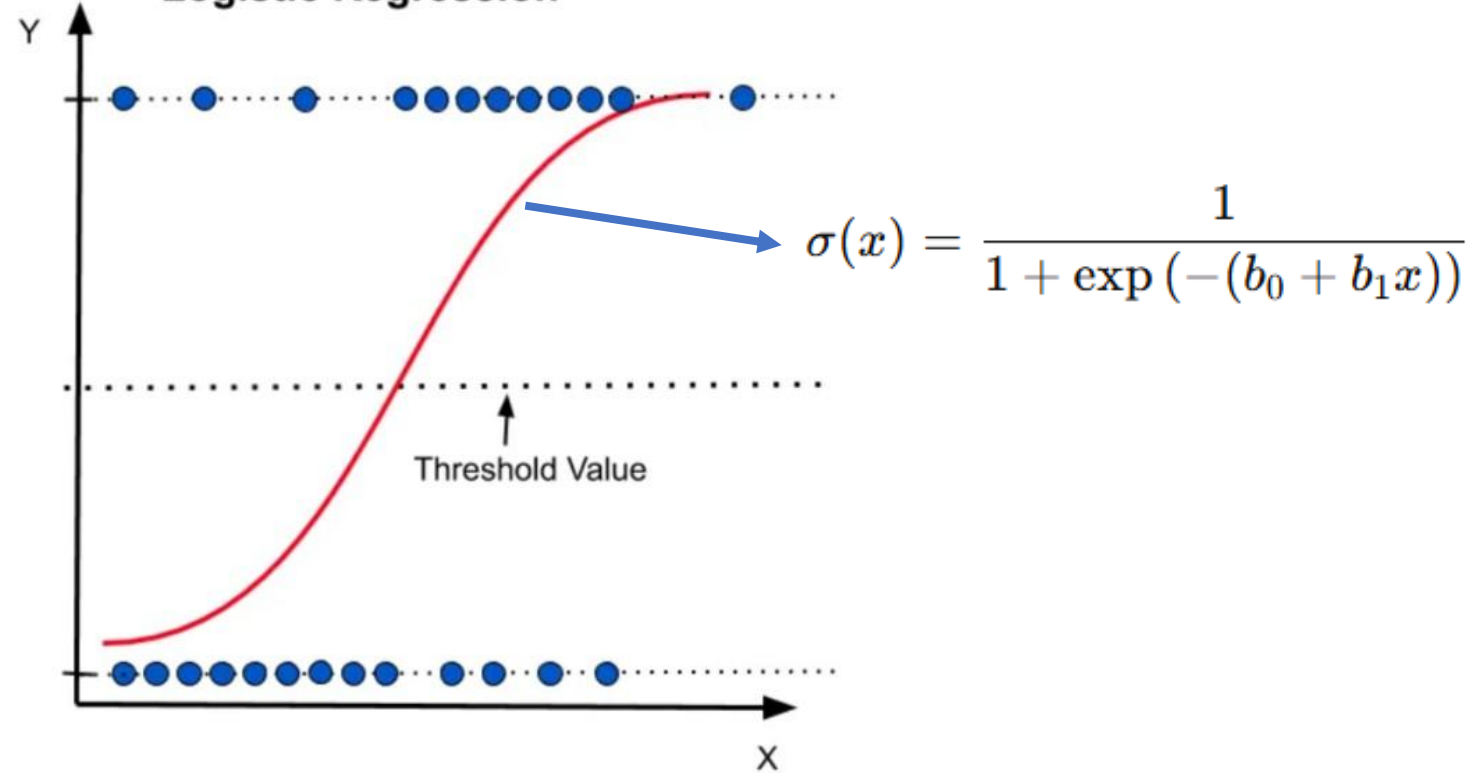


Logistic Regression

Linear Regression

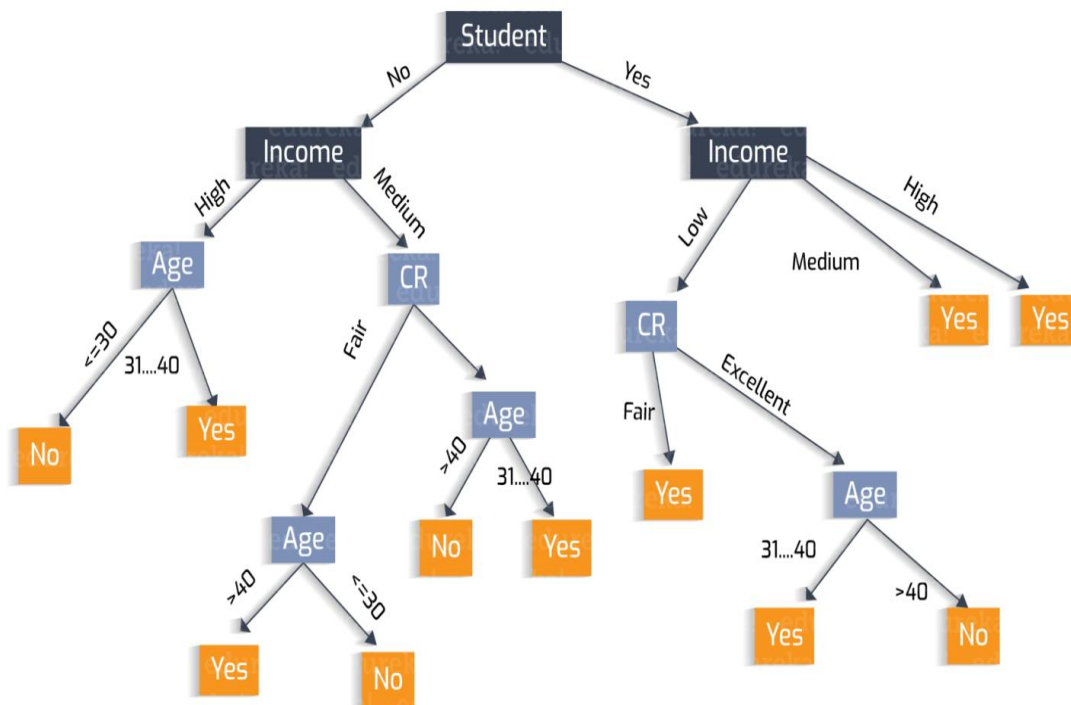


Logistic Regression



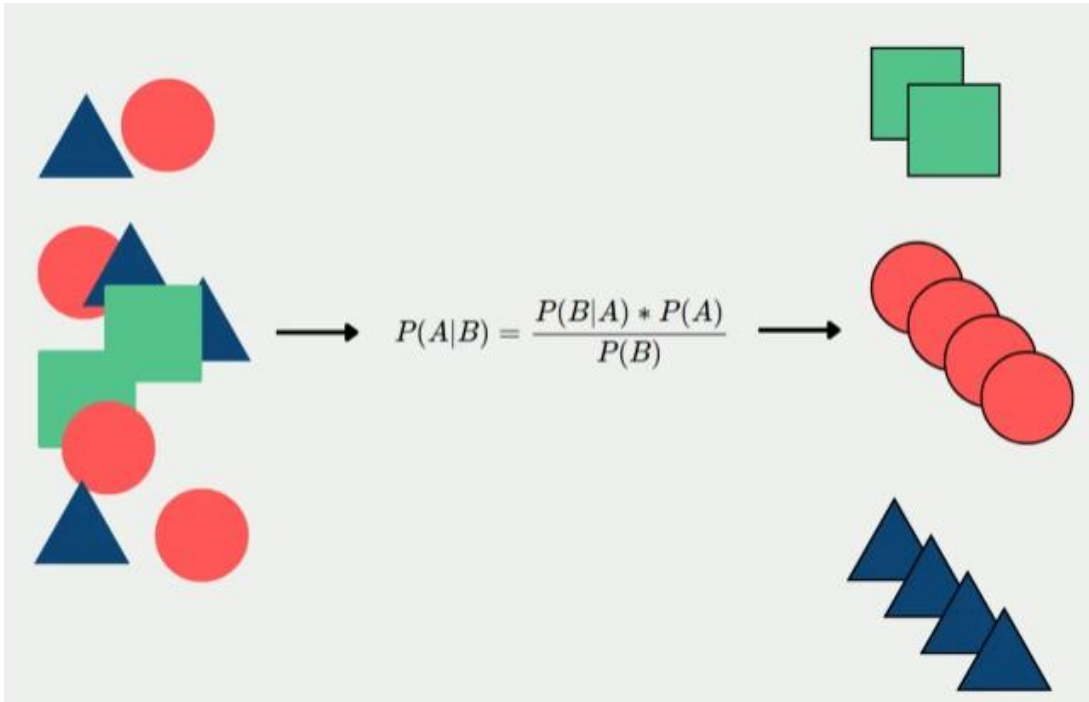
$$L_{BCE} = -\frac{1}{n} \sum_{i=1}^n (Y_i \cdot \log \hat{Y}_i + (1 - Y_i) \cdot \log (1 - \hat{Y}_i))$$

Decision Trees



Student	Income	CR	Age	Buys
No	High	-	<=30	No
No	High	-	31...40	Yes
No	Medium	Fair	>40	Yes
No	Medium	Fair	31...40	Yes
No	Medium	Fair	<=30	No
Yes	Low	Fair	-	Yes
Yes	Low	Excellent	>40	No
Yes	Low	Excellent	31...40	Yes
Yes	Medium	-	-	Yes
Yes	High	-	-	Yes
No	Medium	Fair	>40	? -> YES

Naive Bayes Classifier



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

Labels for the equation:

- Likelihood: $P(x | c)$
- Class Prior Probability: $P(c)$
- Posterior Probability: $P(c | x)$
- Predictor Prior Probability: $P(x)$

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

- $P(c|x)$ 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*.