
TRAVAUX DIRIG  S N   1 : Bayes Classifier

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We work in the framework of binary classification. Throughout this tutorial, we consider a random descriptor X taking values in a measurable space $\mathcal{X} \subset \mathbb{R}^d$ ($d \in \mathbb{N}^*$) and a random label Y taking the value 0 or 1. The joint distribution of the vector (X, Y) is denoted by P , and the regression function (posterior probability) is

$$\eta : x \in \mathcal{X} \mapsto \mathbb{P}(Y = 1 \mid X = x) \in [0, 1].$$

EXERCICE 1. We consider $\mathcal{X} = [0, 1]$ and P such that :

- the conditional distribution of X given $Y = 0$ is $P_0 = \mathcal{U}([0, \theta])$, where $\theta \in]0, 1[$,
- the conditional distribution of X given $Y = 1$ is $P_1 = \mathcal{U}([0, 1])$,
- $p = \mathbb{P}(Y = 1) \in]0, 1[$.

For $x \in \mathcal{X}$, give $\eta(x)$ as a function of p and θ . Give the numerical expression for $\theta = 1/2$.

EXERCICE 2. We consider $\mathcal{X} = \mathbb{R}_+$ and P such that :

- the distribution of X is P_X on $\mathcal{X}$,
- the regression function is $\eta(x) = \frac{x}{x + \theta}$ for every $x \in \mathcal{X}$, where $\theta > 0$ is fixed.

Let h^* denote the Bayes classifier. Make the Bayes classifier explicit in this model. Then show that its 0-1 risk (its probability of error) is

$$L(h^*) = \int_{\mathcal{X}} \min(\eta(x), 1 - \eta(x)) dP_X(x).$$

Compute the Bayes risk when $P_X = \mathcal{U}([0, \alpha\theta])$ with $\alpha > 1$.

EXERCICE 3. Let $\omega(0), \omega(1) \geq 0$ be weights such that $\omega(0) + \omega(1) = 1$. We consider the weighted classification risk :

$$L_{\omega}(g) = \mathbb{E} \left(2\omega(Y) \cdot \mathbb{1}_{\{Y \neq g(X)\}} \right), \quad g : \mathcal{X} \rightarrow \{0, 1\}.$$

Give the Bayes classifier and the Bayes risk for this criterion. What is the interest of considering such a criterion?

EXERCICE 4. We consider $X = (T, U, V)$, where T, U, V are i.i.d. real-valued random variables with the standard exponential distribution. Set $Y = \mathbb{1}_{\{T+U+V < \theta\}}$, where $\theta \in \mathbb{R}_+$ is fixed. Compute the Bayes classifier $g^*(T, U)$ when V is unobserved, and then compute its 0-1 risk as a function of θ .

Bibliographic advice

Useful starting points for machine learning are listed below :

- Theoretical and focused on probabilistic aspects : [DGL97]
- Practical and utility-oriented : [HTF13]
- A recent book focused mainly on optimization : [SSBD14] (and, by the same author, on online learning : [SS12])
- Bayesian methods and graphical models : [MB12]

Références

- [DGL97] L. Devroye, L. Györfi, and G. Lugosi. *A Probabilistic Theory of Pattern Recognition*. Stochastic Modelling and Applied Probability. Springer New York, 1997. 2
- [HTF13] T. Hastie, R. Tibshirani, and J. Friedman. *The Elements of Statistical Learning : Data Mining, Inference, and Prediction*. Springer Series in Statistics. Springer New York, 2013. 2
- [MB12] K.P. Murphy and F. Bach. *Machine Learning : A Probabilistic Perspective*. Adaptive Computation and Machine Learning Series. MIT Press, 2012. 2
- [SS12] S. Shalev-Shwartz. *Online Learning and Online Convex Optimization*. Foundations and Trends in Machine Learning Series. Now Publishers, 2012. 2
- [SSBD14] S. Shalev-Shwartz and S. Ben-David. *Understanding Machine Learning : From Theory to Algorithms*. Cambridge University Press, 2014. 2