

## Task 1 (logistic regression - generative vs discriminative approach)

### a) Generative approach

- Generate a sample from  $\mathcal{N}(m_1, \Sigma)$  ( $Y = 1$ ) and  $\mathcal{N}(m_2, \Sigma)$  ( $Y = -1$ )<sup>1</sup>. Does the distribution of  $P(Y = 1|X = x)$  correspond to a logistic model?
- If the distribution of  $P(Y = 1|X = x)$  corresponds to a logistic model, find the formulas for the parameters of the logistic model (the coefficients and the intercept). Compare obtained parameters  $\beta$  with the estimated  $\hat{\beta}$ .

**Simulations details:** each sample from a normal distribution consists of  $n = 500$  observations (in total we have 1000 observations),  $m_1 = (1, 1)$ ,  $m_2 = (0, 0)$ , and

$$\Sigma = \begin{pmatrix} 1 & -0.5 \\ -0.5 & 1 \end{pmatrix}.$$

Assume that  $\pi = P(Y = 1) = 0.5$ .

### b) Discriminative approach

- Generate a sample from a gaussian mixture

$$\frac{1}{2}\mathcal{N}(m_1, \Sigma) + \frac{1}{2}\mathcal{N}(m_2, \Sigma).$$

Next, generate  $y_i$  for  $i = 1, 2, \dots, n$  from Bernoulli distribution with probability of success  $1/(1 + \exp(-\beta_0 - \beta x_i))$ , where  $\beta_0 = -2$  and  $\beta = (2, 2)$ . Fit a logistic model and compare  $\beta$  with the estimated  $\hat{\beta}$ .

- Draw a scatterplot of data generated in a) and in b). Compare the distributions. What distinguishes the generative approach from the discriminative approach?

- c) Generate  $n_1$  observations from the distribution  $\text{Exp}(1)^2$  (corresponding to  $Y = 1$ ) and  $n_2$  observations from a certain distribution such that  $P(Y = 1|X = x)$  corresponds to a logistic model (you may fix  $n_1 = 1000$ ,  $n_2 = 2000$  and  $\beta_1 = 1$ ). Fit the model.

Q1. What is the distribution of  $f_{X|Y=-1}$ ?

Q2. Once  $\pi$  is fixed, do we have freedom to choose an intercept in the logistic model thus constructed?

Q3. If  $\pi$  is not given, is the intercept in the logistic model uniquely determined?

If that is doable, given  $\beta_1$  and  $\pi$  compute  $\beta_0$ .

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<sup>1</sup>Recall:  $\phi_{\mu, \Sigma}(x) = \frac{1}{(2\pi)^{p/2} |\Sigma|^{1/2}} \exp(-\frac{1}{2}(x - \mu)' \Sigma^{-1}(x - \mu))$

<sup>2</sup>Recall:  $\phi_{\lambda}(x) = \lambda \exp(-\lambda x) \mathbb{I}(x \geq 0)$