

10-702 Statistical Machine Learning: Practice Midterm

Problem 1. Regression

Let $X \in \mathbb{R}$ and

$$Y = \gamma X^2 + \epsilon \quad (1)$$

where $\mathbb{E}(\epsilon) = 0$.

- (a) Find an expression for the oracle linear predictor. In other words, find β_* such that $m(x) = \beta_* x$ minimizes the predictive risk.
- (b) We are given n observations $(X_1, Y_1), \dots, (X_n, Y_n)$ from (1). Give an estimator $\hat{\beta}_n$ for β_* and show that it is consistent.

Problem 2. Model Selection

Suppose we have the following data:

$$\begin{array}{c|ccccc} X & -2 & -1 & 0 & 1 & 2 \\ Y & 0 & 0 & 0 & 0 & 0 \end{array}$$

Consider two regression models:

$$\begin{array}{ll} \text{Model 1:} & Y_i = \beta_0 + \epsilon_i \\ \text{Model 2:} & Y_i = \beta_0 + \beta_1 X_i + \epsilon_i \end{array}$$

- (a) Use C_p to choose between these model. You may assume that $\sigma^2 = 1$. (Recall that the C_p penalty is $2|S|\sigma^2/n$.)
- (b) Fit the model $Y_i = \beta X_i + \epsilon_i$ using the lasso. That is, find $\hat{\beta}$ to minimize

$$\sum_i (Y_i - \beta X_i)^2 + \lambda |\beta|.$$

Problem 3. Convex Duality

Let $X_i \sim \text{Poisson}(\theta)$ be independent, with observations $\{X_1, X_2\} = \{5, 9\}$. Consider the optimization problem

$$\begin{array}{ll} \min_{\theta} & f(\theta) \\ \text{such that} & |\theta| \leq 6 \end{array}$$

where $f(\theta) = -\log f_\theta(5) - \log f_\theta(9)$ is the negative log-likelihood, where $f_\theta(n) = \frac{e^{-\theta} \theta^n}{n!}$.

- (a) What is the solution to this problem?
- (b) Write the Lagrangian.
- (c) Derive the dual problem.

Problem 4. *Convexity and Regularization*

Let Y be the random variable

$$Y = \mu + \epsilon$$

where $\epsilon \sim N(0, 1)$ and $\mu \in \mathbb{R}$ is a constant. Define $\hat{\mu}$ to be the value of μ that minimizes

$$M(\mu) = (Y - \mu)^2 + \lambda J(\mu)$$

where $\lambda > 0$. Consider three cases:

- 1. $J(\mu) = I(\mu \neq 0)$
- 2. $J(\mu) = |\mu|$
- 3. $J(\mu) = \mu^2$.

- (a) For which cases is $M(\mu)$ convex?
- (b) Find $\hat{\mu}$ for all three cases.

Problem 5. *Mixture Models*

Let $(Z_1, X_1, Y_1), \dots, (Z_n, X_n, Y_n)$ be generated as follows:

$$\begin{aligned} Z_i &\sim \text{Bernoulli}(p) \\ X_i &\sim \text{Uniform}(0, 1) \\ \epsilon_i &\sim N(0, \sigma^2) \\ Y_i &\sim \begin{cases} 5X_i + \epsilon_i & \text{if } Z_i = 0 \\ -5X_i + \epsilon_i & \text{if } Z_i = 1. \end{cases} \end{aligned}$$

- (a) Assume we do not observe the Z_i 's or ϵ_i 's. Write the distribution $f(x, y)$ of X and Y as a mixture.
- (b) Write down the likelihood function for p .
- (c) Find a consistent estimator of p that avoids using EM. Hint: find $\mathbb{E}(Y \mid X = x)$.

Problem 6. *Classification*

Suppose that $\mathbb{P}(Y = 1) = \mathbb{P}(Y = 0) = \frac{1}{2}$ and

$$\begin{aligned} X \mid Y = 0 &\sim N(0, 1) \\ X \mid Y = 1 &\sim \frac{1}{2}N(-5, 1) + \frac{1}{2}N(5, 1). \end{aligned}$$

Find an expression for the Bayes classifier.

Problem 7. *Graphical Models*

Let $X = (X_1, X_2, X_3, X_4, X_5)$ be a random vector distributed as

$$X \sim N(0, \Sigma)$$

where the covariance matrix Σ is given by

$$\Sigma = \frac{1}{15} \begin{pmatrix} 9 & -3 & -3 & -3 & -3 \\ -3 & 6 & 1 & 1 & 1 \\ -3 & 1 & 6 & 1 & 1 \\ -3 & 1 & 1 & 6 & 1 \\ -3 & 1 & 1 & 1 & 6 \end{pmatrix} \quad \text{with inverse} \quad \Sigma^{-1} = \begin{pmatrix} 3 & 1 & 1 & 1 & 1 \\ 1 & 3 & 0 & 0 & 0 \\ 1 & 0 & 3 & 0 & 0 \\ 1 & 0 & 0 & 3 & 0 \\ 1 & 0 & 0 & 0 & 3 \end{pmatrix}$$

- (a) What is the graph for X , viewed as a graphical model?
- (b) Which of the following independence statements are true?
 - (a) $X_2 \perp X_3 \mid X_1$
 - (b) $X_3 \perp X_4$
 - (c) $X_1 \perp X_3 \mid X_2$
 - (d) $X_1 \perp X_5$
- (c) List the local Markov properties for this graphical model.