

QUIZ : Linear Models

- 1) For any deterministic $\mu \in \mathbb{R}^p$ and any random variable $X \in \mathbb{R}^p$ express $Cov(X + \mu)$.
- 2) What is the orthogonal projection of a vector $\mathbf{y} \in \mathbb{R}^n$ over $\text{Vect}(\mathbf{1}_n)$, where $\mathbf{1}_n = (1, \dots, 1)^\top \in \mathbb{R}^n$?
- 3) For any matrix $A \in \mathbb{R}^{m \times p}$ and any random vector $X \in \mathbb{R}^p$, express $Cov(AX)$.
- 4) What is the bias the of $\frac{1}{n} \sum_{i=1}^n (y_i - \bar{y}_n)^2$ (\bar{y}_n is the empirical mean) for i.i.d y_i Gaussian variables centered and with variance σ^2 ?
- 5) Let y_1, \dots, y_n be random Gaussian variables i.i.d., centered with variance σ^2 . What is the quadratic risk of $\frac{1}{n} \sum_{i=1}^n (y_i - \bar{y}_n)^2$ as an estimator of σ^2 (\bar{y}_n is the empirical mean)?
- 6) What are the vectors $\mathbf{y} \in \mathbb{R}^n$ such that $\text{var}_n(\mathbf{y}) = 0$ ($\text{var}_n(\text{where } \mathbf{y}) = \frac{1}{n} \sum_{i=1}^n (y_i - \bar{y}_n)^2$ is the empirical variance)?

- Least-squares notation : we write $\mathbf{y} = (y_1, \dots, y_n)^\top \in \mathbb{R}^n$ and $X \in \mathbb{R}^{n \times p}$, and

$$\hat{\boldsymbol{\theta}}^{\text{OLS}} \in \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 . \quad (1)$$

- 7) Let y_1, \dots, y_n and x_1, \dots, x_n be real numbers. Is the following function convex or concave?

$$\begin{aligned} \mathbb{R}^2 &\rightarrow \mathbb{R} \\ (\theta_0, \theta_1) &\mapsto \frac{1}{2} \sum_{i=1}^n (y_i + 3\theta_0 - \theta_1 x_i)^2 . \end{aligned}$$

- 8) Write a pseudo-code to perform the gradient descent algorithm for solving the least squares problem given in Eq. (1) (with input X, \mathbf{y} and α being the step size).
- 9) For any $X \in \mathbb{R}^{n \times p}$ express $\text{Ker}(X^\top X)$ in terms of $\text{Ker}(X)$.
- 10) Let $X \in \mathbb{R}^{n \times n}$ satisfies $X^\top X = \text{Id}_n$ (Id_n being the identity matrix). Can you provide a closed-form solution for the least squares and show it is unique?

- 11) Let $X \in \mathbb{R}^{n \times p}$ be a full (column) rank matrix. What is the covariance (matrix) of the least squares estimator (assuming that the noise model is as follows : $\varepsilon = \mathbf{y} - X\boldsymbol{\theta}^*$ is a centered random vector with covariance matrix $\sigma^2 \text{Id}_n$).
- 12) Express the pseudo inverse of X thanks to its SVD : $X = \sum_{i=1}^r s_i \mathbf{u}_i \mathbf{v}_i^\top$, with $r = \text{rg}(X)$ and $s_1 \geq \dots \geq s_r > 0$.
- 13) Give an explicit solution of the following problem :

$$\arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2} (\mathbf{y} - X\boldsymbol{\theta})^\top \Omega (\mathbf{y} - X\boldsymbol{\theta}) ,$$

for positive-definite matrix $\Omega = \text{diag}(w_1, \dots, w_n)$, in the case where X is full (column) rank.

- Ridge notation : for any $\lambda > 0$,

$$\hat{\boldsymbol{\theta}}^{\text{Ridge}, \lambda} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2n} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \frac{\lambda}{2} \|\boldsymbol{\theta}\|_2^2 .$$

- 14) Give a closed-form solution for the ridge estimator $\hat{\boldsymbol{\theta}}^{\text{Ridge}, \lambda}$ w.r.t. X, \mathbf{y} and λ .
- 15) Give a closed-form solution for the ridge estimator $\hat{\boldsymbol{\theta}}^{\text{Ridge}, \lambda}$ w.r.t. \mathbf{y} and λ when $X = \text{Id}_n$ (here $n = p$).
- 16) Assuming that the noise model is as follows : $\varepsilon = \mathbf{y} - X\boldsymbol{\theta}^*$ is a centered random vector with covariance matrix $\sigma^2 \text{Id}_n$. What is the covariance matrix of the Ridge estimator $\hat{\boldsymbol{\theta}}^{\text{Ridge}, \lambda}$?
- 17) Give a closed-form solution of the following problem w.r.t.
- $$\hat{\boldsymbol{\theta}}^{D, \lambda} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \lambda \|D\boldsymbol{\theta}\|_2^2 .$$
- 18) Give a closed-form solution of $\eta_\lambda(z) = \arg \min_{x \in \mathbb{R}} x \mapsto \frac{1}{2n} (z - x)^2 + \lambda |x|$ w.r.t. z , $\text{sign}(z)$ and the positive part function $(\cdot)_+$.
- 19) What is the sub-differential of the real function $x \mapsto \max(-2x, 0)$?
- 20) Provide the main step in the coordinate descent algorithm to solve the Elastic Net problem :

$$\hat{\boldsymbol{\theta}}^{\text{ENET}, \lambda, \alpha} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \left[\frac{1}{2} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \lambda \left(\alpha \|\boldsymbol{\theta}\|_1 + (1 - \alpha) \frac{\|\boldsymbol{\theta}\|_2^2}{2} \right) \right] .$$

- 21) Provide the main step in the coordinate descent algorithm to solve the Positive Lasso :

$$\hat{\boldsymbol{\theta}}^{\text{Lasso}+, \lambda} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}_+^p} \frac{1}{2} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \lambda \|\boldsymbol{\theta}\|_1 .$$

- 22) Let us assume one has a Lasso solver $\text{Lasso}(X, \mathbf{y}, \lambda)$ that solve the following problem

$$\hat{\boldsymbol{\theta}}^{\text{Lasso}, \lambda} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2n} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \lambda \|\boldsymbol{\theta}\|_1 .$$

What transformation on X can you perform to solve the following problem with positive w_1, \dots, w_p .

$$\hat{\boldsymbol{\theta}}^{\text{Lasso}, \lambda} = \arg \min_{\boldsymbol{\theta} \in \mathbb{R}^p} \frac{1}{2} \|\mathbf{y} - X\boldsymbol{\theta}\|_2^2 + \lambda \sum_{j=1}^p w_j |\theta_j| .$$

- 23) What is the solution of $\begin{cases} \max_{u \in \mathbb{R}^n, v \in \mathbb{R}^p} u^\top Xv \\ \text{s.c. } \|u\|_2^2 = 1 \text{ et } \|v\|_2^2 = 1 \end{cases}$?

- 24) For X_1, \dots, X_n i.i.d. with values in $\{0, 1\}$, propose a procedure to test the hypothesis $p = P(X_1 = 1) = 1/2$.

- 25) In the regression model, assuming that X is deterministic and that $\boldsymbol{\varepsilon} = \mathbf{y} - X\boldsymbol{\theta}^*$ is a Gaussian, centered, with covariance matrix $\sigma^2 \text{Id}_n$, what is the distribution of the least square $\hat{\boldsymbol{\theta}}^{\text{OLS}}$ (one could assume that X is full (column)rank here).

- 26) Let X_1, \dots, X_n be i.i.d Gaussian variables with (unknown) mean μ and known variance σ^2 , i.e., for all $i = 1, \dots, n$, $X_i \sim \mathcal{N}(\mu, \sigma^2)$. Propose a way to test the hypothesis $\mu = 1$.

- 27) Let X_1, \dots, X_n be independent Gaussian variables with (unknown) mean μ and known variance σ_i^2 , i.e., for all $i = 1, \dots, n$, $X_i \sim \mathcal{N}(\mu, \sigma_i^2)$. Propose a way to test the hypothesis $\mu = 1$.

- 28) Let X_1, \dots, X_n be i.i.d. random variables with expectation μ and variance σ^2 . Write a pseudo-code based on the bootstrap for testing if $\mu = 1$ or not.

- 29) Suppose that $Y \in \mathbb{R}^n$ and $X \in \mathbb{R}^{n \times p}$. Write a bootstrap procedure to estimate the variance of the ordinary least squares estimates $\hat{\beta}_k$ of the linear regression $Y \simeq X\beta$.