

MATH-GA.2711 Machine Learning & Computational Statistics

Homework 3

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Due: February 16, 2026

Instruction

This homework is to be done individually. No collaboration and/or code sharing permitted.

Questions

1. The classical linear regression model.

- (a) Show that $\mathbf{P} = \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$ is a symmetric and orthogonal projection from $\mathbb{R}^n \rightarrow \mathcal{R}(\mathbf{X})$.
- (b) Using NumPy generate a random full-rank design matrix $\mathbf{X} \in \mathbb{R}^{n \times p}$ (e.g. $n = 50, p = 5$), form \mathbf{P} as above. Compute and report: (i) $\|\mathbf{P} - \mathbf{P}'\|_F$ (ii) $\|\mathbf{P}^2 - \mathbf{P}\|_F$ (iii) for a random \mathbf{y} , verify $\mathbf{X}'(\mathbf{y} - \mathbf{P}\mathbf{y}) \approx 0$ Also report $\text{rank}(\mathbf{P})$ via `np.linalg.matrix_rank` and compare to p .
- (c) Can the R^2 from a classical linear regression be negative? If yes, provide an example. If no, prove it is ≥ 0 .
- (d) Show that for a symmetric matrix \mathbf{A} the following identity holds

$$\nabla_{\beta} \beta' \mathbf{A} \beta = 2\mathbf{A}\beta.$$

- (e) Explain the difference between fixed and random regressors.
- (f) Simulate in NumPy two scenarios for \mathbf{X} with $n = 200, p = 3$: (i) *fixed* \mathbf{X} : draw once and hold it fixed across B Monte Carlo repetitions, (ii) *random* \mathbf{X} : redraw \mathbf{X} each repetition. Generate $\mathbf{y} = \mathbf{X}\beta + \varepsilon$ with $\varepsilon \sim \mathcal{N}(0, \sigma^2 I)$, compute $\hat{\beta}$ each time, and compare: (a) the empirical $\text{Var}(\hat{\beta} \mid \mathbf{X})$ in case (i) to $\sigma^2(\mathbf{X}'\mathbf{X})^{-1}$, (b) the unconditional variance across repetitions in case (ii). Summarize what changes conceptually between conditioning on \mathbf{X} and averaging over \mathbf{X} .

2. Finite-sample properties of OLS.

- (a) List the assumptions under which s^2 is an unbiased estimator of σ^2 . Then prove that s^2 is unbiased under these assumptions.
- (b) Fix $n = 80, p = 5$ and in NumPy draw a random full-rank \mathbf{X} and random β . Set $\sigma^2 = 2$. For $B = 5000$ repetitions, simulate $\varepsilon \sim \mathcal{N}(0, \sigma^2 I)$, set $\mathbf{y} = \mathbf{X}\beta + \varepsilon$, compute $\hat{\beta}$, $\hat{\varepsilon} = \mathbf{y} - \mathbf{X}\hat{\beta}$ and s^2 . Demonstrate s^2 is unbiased estimator of σ^2 . What would need to change in the above setup to create a biased estimator? Demonstrate numerically.
- (c) List the assumptions under which

$$\text{var}[\hat{\beta}|\mathbf{X}] = \sigma^2(\mathbf{X}'\mathbf{X})^{-1}. \quad (1)$$

Then show this assertion.

- (d) Assume the classical linear regressions assumptions (I)-(IV) hold. Show

$$\text{cov}[\hat{\beta}, \mathbf{r}|\mathbf{X}] = 0. \quad (2)$$

- (e) For fixed \mathbf{X} , simulate B datasets in NumPy as above, and for each compute $\hat{\beta}^{(b)}, \hat{\mathbf{y}}^{(b)}, \mathbf{r}^{(b)}$. Estimate the cross-covariance matrix between $\hat{\beta}$ and \mathbf{r} :

$$\hat{C} = \frac{1}{B-1} \sum_{b=1}^B (\hat{\beta}^{(b)} - \bar{\hat{\beta}})(\mathbf{r}^{(b)} - \bar{\mathbf{r}})' \in \mathbb{R}^{p \times n}.$$

Report $\|\hat{C}\|_F$ (or $\max |\hat{C}_{ij}|$) and verify it shrinks as B increases.

3. Hypothesis testing.

- (a) (i) Describe what is a linear hypothesis in the context of the classical linear regression model

$$y = \mathbf{x}'\boldsymbol{\beta} + \varepsilon, \quad (3)$$

where $y \in \mathbb{R}$ and $\varepsilon \in \mathbb{R}$ are random variables, $\mathbf{x} \in \mathbb{R}^p$ is a random vector and $\boldsymbol{\beta} \in \mathbb{R}^p$ are the parameters. (ii) How do you test this hypothesis using the F -ratio?

- (b) Using (a) describe how you would test

$$H_0 : \beta_2 = \dots = \beta_p = 0. \quad (4)$$

- (c) List the assumptions under which the F -ratio

$$F|\mathbf{X}, H_0 \sim F_{(d, n-p)}. \quad (5)$$

Then show this assertion.