

## Exercise 3 : expected value of empirical risk for ols

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### 1 Question 1: Expected Value over $\epsilon$

We need to show:

$$E[R_n(\hat{\theta})] = E_{\epsilon} \left[ \frac{1}{n} \|I_n - X(X^T X)^{-1} X^T \epsilon\|^2 \right]$$

We know that:

$$R_n(\hat{\theta}) = \frac{1}{n} \|y - X\hat{\theta}\|^2 = \frac{1}{n} \|(I_n - X(X^T X)^{-1} X^T) \epsilon\|^2$$

So we can do:

$$E[R_n(\hat{\theta})] = E_{\epsilon} \left[ \frac{1}{n} \|(I_n - X(X^T X)^{-1} X^T) \epsilon\|^2 \right]$$

### 2 Question 2: Trace Identity

We need to show:

$$\sum_{(i,j) \in [1,n]^2} A_{ij}^2 = \text{tr}(A^T A)$$

By definition:

$$\sum_{(i,j) \in [1,n]^2} A_{ij}^2 = \sum_{i=1}^n \sum_{j=1}^n A_{ij}^2$$

The trace of  $A^T A$  is:

$$\text{tr}(A^T A) = \sum_{i=1}^n (A^T A)_{ii}$$

Each diagonal element of  $A^T A$  is the dot product of the  $i$ -th row of  $A^T$  and the  $i$ -th column of  $A$ :

$$(A^T A)_{ii} = \sum_{j=1}^n (A^T)_{ij} A_{ji} = \sum_{j=1}^n A_{ji} A_{ji} = \sum_{j=1}^n A_{ij}^2$$

Adding all  $i$ , we get:

$$\text{tr}(A^T A) = \sum_{i=1}^n \sum_{j=1}^n A_{ij}^2$$

So:

$$\sum_{(i,j) \in [1,n]^2} A_{ij}^2 = \text{tr}(A^T A)$$

### 3 Question 3: Expected Norm of $A\epsilon$

We need to show:

$$E_\epsilon \left[ \frac{1}{n} \|A\epsilon\|^2 \right] = \frac{\sigma^2}{n} \text{tr}(A^T A)$$

First, let's do  $E_\epsilon [\|A\epsilon\|^2]$ . Since  $\epsilon \sim \mathcal{N}(0, \sigma^2 I_n)$ , we have:

$$E_\epsilon [\epsilon^T A^T A \epsilon] = \sigma^2 \text{tr}(A^T A)$$

So:

$$E_\epsilon [\|A\epsilon\|^2] = E_\epsilon [\epsilon^T A^T A \epsilon] = \sigma^2 \text{tr}(A^T A)$$

Dividing by  $n$ ,

$$E_\epsilon \left[ \frac{1}{n} \|A\epsilon\|^2 \right] = \frac{\sigma^2}{n} \text{tr}(A^T A)$$

### 4 Question 4: Idempotency of $A$

We need to show:

$$A^T A = A$$

where  $A = I_n - X(X^T X)^{-1} X^T$ .

First, note that  $A$  is symmetric:

$$A^T = (I_n - X(X^T X)^{-1} X^T)^T = I_n - X(X^T X)^{-1} X^T = A$$

To show idempotency:

$$A^2 = A$$

$$A^2 = (I_n - X(X^T X)^{-1} X^T)(I_n - X(X^T X)^{-1} X^T)$$

$$A^2 = I_n - 2X(X^T X)^{-1} X^T + X(X^T X)^{-1} X^T X(X^T X)^{-1} X^T$$

Using  $X^T X(X^T X)^{-1} = I_d$ ,

$$A^2 = I_n - 2X(X^T X)^{-1} X^T + X(X^T X)^{-1} I_d(X^T X)^{-1} X^T$$

$$A^2 = I_n - 2X(X^T X)^{-1} X^T + X(X^T X)^{-1} X^T$$

$$A^2 = I_n - X(X^T X)^{-1} X^T$$

$$A^2 = A$$

## 5 Question 5: Conclusion

Using results from Questions 1-4:

$$E[R_n(\hat{\theta})] = E_\epsilon \left[ \frac{1}{n} \|I_n - X(X^T X)^{-1} X^T \epsilon\|^2 \right]$$

From Question 3:

$$E_\epsilon \left[ \frac{1}{n} \|A\epsilon\|^2 \right] = \frac{\sigma^2}{n} \text{tr}(A^T A)$$

From Question 4,  $A^T A = A$ . So:

$$E_\epsilon \left[ \frac{1}{n} \|A\epsilon\|^2 \right] = \frac{\sigma^2}{n} \text{tr}(A)$$

The trace of  $A = I_n - X(X^T X)^{-1} X^T$  is:

$$\text{tr}(A) = \text{tr}(I_n) - \text{tr}(X(X^T X)^{-1} X^T) = n - d$$

So:

$$E_\epsilon \left[ \frac{1}{n} \|A\epsilon\|^2 \right] = \frac{\sigma^2}{n} (n - d)$$

So:

$$E[R_n(\hat{\theta})] = \frac{n - d}{n} \sigma^2$$

## 6 Question 6: Expected Value of $\frac{\|y - X\hat{\theta}\|_2^2}{n - d}$

In the context of linear regression with the OLS estimator  $\hat{\theta} = (X^T X)^{-1} X^T y$ ,

we want to find the expected value of  $\frac{\|y - X\hat{\theta}\|_2^2}{n - d}$ .

Residual Calculation:

$$y - X\hat{\theta} = y - X(X^T X)^{-1} X^T y = (I_n - X(X^T X)^{-1} X^T) y$$

Norm Squared of Residuals:

$$\|y - X\hat{\theta}\|_2^2 = \|(I_n - X(X^T X)^{-1} X^T) y\|_2^2$$

Expected Value: By the properties of OLS and assuming  $\epsilon \sim \mathcal{N}(0, \sigma^2 I_n)$ , we know:

$$E[\|y - X\hat{\theta}\|_2^2] = \sigma^2 \text{tr}(I_n - X(X^T X)^{-1} X^T)$$

$$E[\|y - X\hat{\theta}\|_2^2] = \sigma^2 (n - \text{tr}(X(X^T X)^{-1} X^T))$$

Expected Value of  $\frac{\|y - X\hat{\theta}\|_2^2}{n - d}$ :

$$E \left[ \frac{\|y - X\hat{\theta}\|_2^2}{n - d} \right] = \frac{\sigma^2 (n - \text{tr}(X(X^T X)^{-1} X^T))}{n - d}$$