

Course outline

How does an NPTEL online course work?

Week-0

Week-1

Week-2

Week-3

Week-4

Week-5

Lec 21- Least Squares (LS) Solution, Pseudo-Inverse Concept

Lec 22- Least Squares (LS) via Principle of Orthogonality, Projection Matrix, Properties

Lec 23- Application: Pseudo-Inverse and MIMO Zero Forcing (ZF) Receiver

Lec 24- Wireless Application: Multi-Antenna Channel Estimation

Lec 25- Machine Learning Application: Linear Regression

Quiz : Assignment-5

Feedback for Week 5

Solution-5

Week-6

Week-7

Week-8

Week-9

Week-10

Week-11

Week-12

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Assignment-5

The due date for submitting this assignment has passed.

Due on 2021-02-24, 23:59 IST.

As per our records you have not submitted this assignment.

 1) The pseudo-inverse of a tall full column rank matrix \mathbf{A} is given as

1 point

- $(\mathbf{A}\mathbf{A}^T)^{-1}\mathbf{A}^T$
 $\mathbf{A}^T(\mathbf{A}\mathbf{A}^T)^{-1}$
 $\mathbf{A}(\mathbf{A}\mathbf{A}^T)^{-1}$
 $(\mathbf{A}^T\mathbf{A})^{-1}\mathbf{A}^T$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $(\mathbf{A}^T\mathbf{A})^{-1}\mathbf{A}^T$

 2) Consider the least-squares (LS) problem $\min \|\bar{\mathbf{y}} - \mathbf{A}\bar{\mathbf{x}}\|^2$, where \mathbf{A} is a tall matrix with full column rank. The solution $\hat{\mathbf{x}}$ for this LS problem is

1 point

- $\bar{\mathbf{y}}(\mathbf{A}\mathbf{A}^T)^{-1}\mathbf{A}^T$
 $\bar{\mathbf{y}}^T\mathbf{A}^T(\mathbf{A}\mathbf{A}^T)^{-1}$
 $(\mathbf{A}^T\mathbf{A})^{-1}\mathbf{A}^T\bar{\mathbf{y}}$
 $\mathbf{A}(\mathbf{A}\mathbf{A}^T)^{-1}\bar{\mathbf{y}}$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $(\mathbf{A}^T\mathbf{A})^{-1}\mathbf{A}^T\bar{\mathbf{y}}$

3) The least-squares (LS) principle can be used for

1 point

- Only channel estimation in wireless systems
 Only linear regression in machine learning
 Only polynomial fitting for mathematical modelling
 All of these applications

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
All of these applications

 4) Consider the matrix \mathbf{A} below

1 point

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix}$$

 The pseudo-inverse of the matrix \mathbf{A} is

- $\frac{1}{4} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix}$
 $\frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$
 $\frac{1}{4} \begin{bmatrix} 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \end{bmatrix}$
 $\frac{1}{4} \begin{bmatrix} 1 & 1 \\ -1 & 1 \\ 1 & -1 \\ -1 & -1 \end{bmatrix}$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $\frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$

 5) Consider the linear system of equations $\bar{\mathbf{y}} = \mathbf{A}\bar{\mathbf{x}}$, where

1 point

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix}, \bar{\mathbf{y}} = \begin{bmatrix} 1 \\ -2 \\ 3 \\ 2 \end{bmatrix}$$

The least-squares (LS) solution for the system of equations is

- $\frac{1}{2} \begin{bmatrix} 1 \\ -2 \end{bmatrix}$
 $\frac{1}{2} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$
 $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$
 $\frac{1}{3} \begin{bmatrix} 2 \\ -3 \end{bmatrix}$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $\frac{1}{2} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$

6) Principal Component Analysis (PCA) is used in machine learning for

1 point

- Dimensionality reduction
 Signal Estimation
 Direction of Arrival determination
 Covariance classification

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
Dimensionality reduction

 7) Consider an Alamouti coded system with channel coefficients $h_1 = -2 - 3j$, $h_2 = 3 + 2j$
1 point

The effective channel matrix for this system is given as

- $\begin{bmatrix} -2 - 3j & 3 + 2j \\ 3 - 2j & 2 + 3j \end{bmatrix}$
 $\begin{bmatrix} -2 - 3j & 3 + 2j \\ 3 - 2j & 2 - 3j \end{bmatrix}$
 $\begin{bmatrix} -2 - 3j & 3 + 2j \\ -3 - 2j & 2 - 3j \end{bmatrix}$
 $\begin{bmatrix} -2 - 3j & 3 - 2j \\ 3 - 2j & 2 - 3j \end{bmatrix}$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $\begin{bmatrix} -2 - 3j & 3 + 2j \\ 3 - 2j & 2 - 3j \end{bmatrix}$

 8) Let the covariance estimate of the data vectors obtained during Principal Component Analysis (PCA) be denoted by \mathbf{R} . The vector $\bar{\mathbf{v}}_1$ corresponding to the direction of the largest principal component is given as

1 point

- Eigenvector of \mathbf{R} corresponding to the smallest eigenvalue
 Dominant vector in the nullspace of \mathbf{R}
 Eigenvector of \mathbf{R} corresponding to the largest eigenvalue
 Right singular vector of \mathbf{R} corresponding to any zero singular value

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
Eigenvector of \mathbf{R} corresponding to the largest eigenvalue

 9) The determinant of a unitary matrix \mathbf{U} satisfies the property

1 point

- $\det(\mathbf{U}) = \pm 1$
 $\det(\mathbf{U}) = 1$
 $\det(\mathbf{U}) \geq 0$
 $|\det(\mathbf{U})| = 1$

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
 $|\det(\mathbf{U})| = 1$

 10) Let the covariance matrix \mathbf{R} estimated during the Principal Component Analysis (PCA) be given as

1 point

$$\mathbf{R} = \begin{bmatrix} -2 & -2 \\ 2 & -2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ -1 & -1 \end{bmatrix}$$

 Let $\bar{\mathbf{v}}_1$ denote the unit-norm vector corresponding to the direction of the largest principal component extracted from the above covariance matrix. Then, the quantity $\bar{\mathbf{v}}_1^H \mathbf{R} \bar{\mathbf{v}}_1$ equals

- 6
 12
 16
 8

 No, the answer is incorrect.
Score: 0

 Accepted Answers:
16