

MAST30025: Linear Statistical Models

Assignment 1 S1 2021

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March 23, 2021

Question 1 Solution:

Part a:

$$A^2 = A^3$$

Suppose A is a square matrix is (real and) symmetric then its eigenvalues are all real, and its eigenvectors are orthogonal.

Theorem 2.3

Proof:

Take A to be a square matrix, $n \times n$. First we diagonalise A , i.e., find P such that.

$$= D = P^T A P$$

$$= \begin{bmatrix} \lambda_1 & \dots & 0 \\ \dots & \dots & \dots \\ 0 & \dots & \lambda_k \end{bmatrix}$$

where $\lambda_1, \lambda_2, \dots, \lambda_k$ are the eigenvalues of A .

Since P is orthogonal both P and P^T are non-singular,

$$r(P^T A P) = r(P^T A) = r(A)$$

Because $P^T A P$ is diagonal $r(P^T A P)$ is the number of non zero eigenvalues of A .

But we wanted to prove **Theorem 2.2**

that A any symmetric matrix is idempotent. Which has eigenvalues of $\lambda = 0$ or $\lambda = 1$.

The eigenvalues of idempotent matrices are always either

$$\lambda = 0 \text{ or } \lambda = 1.$$

$$A^2 = \lambda^2 x$$

Multiplying by A !!!

$$A^3 x = A^2 \lambda x = \lambda A^2 x = \lambda^3 x$$

$$(\lambda^3 - \lambda^2)x = 0$$

By definition, $x \neq 0$,

$$\lambda^3 - \lambda^2 = 0$$

$$\lambda^2(\lambda - 1) = 0$$

Therefore there are two values with eigenvalues of 0 and one eigenvalue of 1! satisfies this theorem that A is idempotent!

Part b:

$$A = A^3$$

$$A^3 x = A \lambda x = \lambda A x = \lambda^3 x$$

Using the same theorem from the previous it has eigenvalues of 0,1 and -1. Since we care that A has to be positive semi-definite. Which has an eigenvalue of -1. Which does not satisfy Theorem 2.2! A is not idempotent!

Question 2 Solution:

Theorem 2.4

There exists a matrix **P** which diagonalises A_1, \dots, A_m .

$$P^T A_i P = D_i$$

and

$$P^T A_j P = D_j$$

We take A_i and A_j to be $k \times k$ matrices first we diagonalizes A_i, A_j , i.e. find P such that,

$$D_i = P^T A_i P = \begin{bmatrix} \lambda_1 & \dots & 0 \\ \dots & \lambda_2 & \dots \\ 0 & \dots & \lambda_k \end{bmatrix}$$

for $i = 1, \dots, k$

$$D_j = P^T A_j P = \begin{bmatrix} \lambda_1 & \dots & 0 \\ \dots & \lambda_2 & \dots \\ 0 & \dots & \lambda_m \end{bmatrix}$$

for $j = 1, \dots, m$

Proof:

$$P^T A_i A_j P = (P^T A_i P)(P^T A_j P) = (P^T A_j P)(P^T A_i P) = P^T A_j A_i P$$

Pre-multiply by P and post-multiply by P^T to get $A_i A_j = A_j A_i$.

Question 3 Solution:

Pre Proof Using Theorem 2.3

For any matrix A

$$r(A) = r(A^T) = r(A^T A) = \text{tr}(A)$$

$$A = \begin{bmatrix} | & | & \dots & | & \dots & | \\ a_1 & a_2 & \dots & a_p & \dots & a_n \\ | & | & \dots & | & \dots & | \end{bmatrix}$$

Given A matrix with dimensions n x p with p independent columns.

Let $x_1, x_2, x_3, \dots, x_k$ the basis for column space of A.

Definition of basis every column vector of A is a linear combination of the column vectors of x.

$$a_1 = b_1 x_1 + b_2 x_2 + \dots + b_k x_k$$

Definition of linear combination

where b is scalar

$$B = \begin{bmatrix} - & - & b_1 & - & - \\ - & - & b_2 & - & - \\ | & & & & \\ - & - & b_p & - & - \end{bmatrix}$$

$$\begin{bmatrix} | & | & \dots & | & \dots & | \\ a_1 & a_2 & \dots & a_p & \dots & a_n \\ | & | & \dots & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | & \dots & | \\ x_1 & x_2 & \dots & x_p & \dots & x_n \\ | & | & \dots & | & \dots & | \end{bmatrix} \begin{bmatrix} - & - & b_1 & - & - \\ - & - & b_2 & - & - \\ | & & & & \\ - & - & b_p & - & - \end{bmatrix}$$

$$A = XB$$

$$A^T = (XB)^T = B^T X^T$$

$$r(A) \leq r(A^T) \text{ or } r(A) \geq r(A^T) \text{ to satisfy!}$$

$$r(A) = r(A^T) = r(A^T A) = \text{tr}(A) = p$$

Since P is orthogonal both P and P^T are non-singular. Therefore we need to sum up the diagonal elements of $P^T A P$, so we need to sum up its trace!

$$r(A) = r(P^T A P) = \text{tr}(P^T A P) = \text{tr}(P P^T A) = \text{tr}(A) = p$$

Because D = $P^T A P$ is diagonal $r(P^T A P)$ is the number of nonzero values of A!

But A is idempotent so its takes eigenvalues between 0 or 1. To Prove Theorem 2.7! We need only the identity matrix to allow $A^T A$ to be positive definite!

Using Theorem 2.7

Proof (\Leftarrow) :

We want $A^T A$ to be symmetric

and have all the eigenvalues to be strictly positive to prove $A^T A$ is a positive definite matrix!

we know $r(A^T A) = p$ is a $p \times p$ matrix so it has to be a full rank matrix, p !

Let, $\lambda_1, \lambda_2, \dots, \lambda_p > 0$ be the eigenvalues of $A^T A$ for every x and for each eigenvalue has to have a value of 1.

for $z = P^T x = (z_1, \dots, z_p)^T$

$$x^T (A^T A) x = x^T P D P^T x = z^T D z = \sum_{i=1}^p z_i^2 \lambda_i$$

since $\lambda_i = 1$!

$$= \sum_{i=1}^p z_i^2$$

> 0

Thus $A^T A$ is positive definite as required!

Proof (\Rightarrow) :

Suppose $A^T A$ is positive definite let x_i be its normalised i -th eigenvector then,

$$x_i^T (A^T A) x_i = \lambda_i x_i^T x_i = \lambda_i$$

From theorem 2.3 we want $A^T A$ to be symmetric and idempotent. We want the eigenvalues to be 0 or 1. This case all of the eigenvalues must equal to 1.

$\lambda_i = 1 > 0$

So, the eigenvalues of $A^T A$ are strictly positive as required!!