

Q-67

(a) To prove $y^T P y \geq 0$.

$$\Rightarrow y^T A^T A y$$

$$\because P = A^T A$$

$$\Rightarrow (A y)^T A y$$

$$\Rightarrow \|A y\|^2, \text{ is positive or zero}$$

Hence, we can say that $y^T P y \geq 0$

To prove $z^T Q z \geq 0$

$$\Rightarrow z^T A A^T z$$

$$\because Q = A A^T$$

$$\Rightarrow z^T A (z^T A)^T$$

$$\Rightarrow \|z^T A\|^2, \text{ is positive or zero}$$

Hence, we can say that $z^T Q z \geq 0$

Eigen values of P and Q are non-negative, as

$$P Y = \lambda Y \quad [\text{for any eigen values } \lambda \text{ \& \; eigen vector } Y]$$

Multiplying Both sides by Y^T

$$Y^T P Y = \lambda$$

And we have proved above that $Y^T P Y \geq 0$. Hence eigen values of P & Q are non-negative

Q-6).

(b) To prove: $Q(Au) = \lambda(Au)$

We know,

$Pu = \lambda u$, where λ is eigen value of P & u is eigen vector of P .

$APu = \lambda Au$, Multiplying both sides by A

$AA^T Au = \lambda Au$, $\because P = A^T A$

$Q(Au) = \lambda(Au)$, $\because Q = AA^T$

Hence proved, $Q(Au) = \lambda(Au)$ i.e. Au is an eigen vector of Q with eigen value λ

To prove: $P(A^T v) = \mu(A^T v)$

We know,

$Qv = \mu v$, where μ is eigen value of Q & v is eigen vector of Q .

$A^T Qv = \mu A^T v$, Multiplying both sides by A^T

$A^T A A^T v = \mu A^T v$, $\because Q = AA^T$

$P(A^T v) = \mu(A^T v)$, $\because P = A^T A$

Hence proved, $P(A^T v) = \mu(A^T v)$ i.e. $A^T v$ is an eigen vector of P with eigen value μ

Number of elements in u is n

Number of elements in v is m

(c) To prove $\gamma_i \geq 0$ such that $A u_i = \gamma_i v_i$

$$\text{Given } u_i \triangleq \frac{A^T v_i}{\|A^T v_i\|_2}$$

$$A \frac{A^T v_i}{\|A^T v_i\|_2} = \gamma_i v_i$$

$$v_i^T A A^T v_i = \gamma_i \cdot \|A^T v_i\|_2^2, \text{ multiplying both sides by } v_i^T$$

We have proved in part a that $v_i^T A A^T v_i \geq 0$

Hence,

$$\gamma_i \|A^T v_i\|_2^2 \geq 0$$

$$\gamma_i \geq 0 \quad \because \|A^T v_i\|_2^2 \geq 0$$

Hence proved.

(d) Given: $\rightarrow U = [u_1 | u_2 \dots | u_m]$, $V = [v_1 | v_2 \dots | v_n]$

To show, $A = U \Sigma V^T$

We know from part a,

$P = V \Sigma V^T$, where V contains eigen vector, Σ is a diagonal matrix contain eigen values of P .

$$P_{n \times n} = V_{n \times n} \Sigma_{n \times n} V_{n \times n}^T$$

Now Σ is a diagonal matrix,

$\therefore \Sigma_{n \times n} = T_{n \times m}^T \times T_{m \times n}$, where $T_{m \times n}$ contains all sq. root values of $\Sigma_{n \times n}$

Putting value of $\Sigma_{n \times n}$ in eq ①. $\therefore \Sigma$ contains all non-negative values, T will contain all non-negative values.

$$P = V (T^T T) V^T$$

$$P = V (T^T I T) V^T$$

$$P = V (T^T U^T U T) V^T$$

where $U^T U = I$

$$P = (U T V^T)^T (U T V^T)$$

$$A^T A = (U T V^T)^T (U T V^T)$$

Hence, $A = U T V^T$

Proved.