

$$6) (a) y^T P y = y^T A^T A y \\ = (A y)^T (A y)$$

$A \Rightarrow (m \times n)$ matrix

$y \Rightarrow (n \times 1)$ vector

So, $A y \Rightarrow (m \times 1)$ vector

The ^{square of the} L-2 norm of a vector z is defined as $z^T z$

$$\text{Here } z = A y$$

$$\Rightarrow y^T P y = \text{L-2 norm of } (A y)$$

From the property of L-2-norms,

$$\text{L-2-norm} \geq 0$$

$$\Rightarrow (A y)^T (A y) \geq 0$$

$$\text{Similarly, } z^T Q z = \cancel{z^T A} z^T A A^T z \\ = (A^T z)^T (A^T z) \\ = \text{L-2-norm of } (A^T z) \\ \geq 0$$

Now, consider the eigen values λ of P

$$P v = \lambda v$$

$$\Rightarrow v^T P v = \lambda v^T v$$

$$\Rightarrow \|A v\|_2^2 = \lambda \|v\|_2^2$$

$$\text{as } \|A v\|_2^2 \geq 0 \text{ \& } \|v\|_2^2 \geq 0, \lambda \geq 0$$

\therefore Eigen values are non negative

Similarly, consider an eigen value- λ of Q ,

$$Qv = \lambda v$$

$$\Rightarrow v^T Q v = \lambda (v^T v)$$

$$\Rightarrow \|A^T v\|_2^2 = \lambda \|v\|_2^2$$

By the same argument, ^{as before} eigen values of Q are non-negative

(b) $Au = \lambda u$

$$A^T A u = \lambda u$$

$$\Rightarrow A(A^T A u) = \lambda (A u)$$

using associative property,

$$(A A^T)(A u) = \lambda (A u)$$

$$\textcircled{\bullet} Q(A u) = \lambda (A u)$$

So, $A u$ is an eigen vector of Q with eigen value λ .

Now,

$$Qv = \mu v$$

$$A A^T v = \mu v$$

$$A^T (A A^T v) = \mu A^T v$$

$\textcircled{\bullet}$ using associative property,

$$(A^T A)(A^T v) = \mu (A^T v)$$

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

So, $A^T V$ is an eigenvector of P with eigenvalue μ .
Now, coming to the no. of elements,

$$A \Rightarrow (m \times n) \text{ matrix}$$

$$u \Rightarrow (y \times 1) \text{ vector}$$

for Au to be defined, $y=n$ is required

So, $u \Rightarrow (n \times 1) \text{ vector}$ - It has 'n' elements

Similarly $A^T \Rightarrow (n \times m) \text{ matrix}$

$$v \Rightarrow (x \times 1) \text{ vector}$$

$x=m$ for $A^T v$ to be defined

So, $v \Rightarrow (m \times 1) \text{ vector}$ - It has 'm' elements

(c) $Qv_i = \lambda_i v_i$

$$\Rightarrow A A^T v_i = \lambda_i v_i$$

$$\Rightarrow A (A^T v_i) = \lambda_i v_i$$

$$\Rightarrow A \frac{(A^T v_i)}{\|A^T v_i\|_2} = \left(\frac{\lambda_i}{\|A^T v_i\|_2} \right) v_i$$

$$\Rightarrow A u_i = \left(\frac{\lambda_i}{\|A^T v_i\|_2} \right) v_i$$

Taking $\gamma_i = \frac{\lambda_i}{\|A^T v_i\|_2}$, we show there exists γ_i such that

$$A u_i = \gamma_i v_i$$

$$(d) \quad P_{ii} = \delta_{ii} \Rightarrow Q_{ii} = \delta_{ii} v_i$$

$$A v_i = \gamma_i v_i$$

$$\Rightarrow AV = A[u_1 | u_2 | \dots | u_m]$$

Block matrix multiplication
is valid as sizes of A & u_i
match & is applied

$$= [A u_1 | A u_2 | A u_3 | \dots | A u_m]$$

$$= [\gamma_1 v_1 | \gamma_2 v_2 | \dots | \gamma_m v_m]$$

$$= \underbrace{[v_1 | v_2 | \dots | v_m]}_U \underbrace{\begin{bmatrix} \gamma_1 & 0 & \dots & 0 \\ 0 & \gamma_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \gamma_m \end{bmatrix}}_T$$

$$AV = UT$$

$$\Rightarrow A(VV^T) = UT V^T$$

$$VV^T = [u_1 | u_2 | \dots | u_m] \begin{bmatrix} u_1^T \\ u_2^T \\ \vdots \\ u_m^T \end{bmatrix} = I_n \text{ (from lecture slides)}$$

$$\Rightarrow A I_n = UT V^T$$

~~using block matrix multiplication~~

$$\therefore A = UT V^T$$