$m{A} \in \mathbb{R}^{m \times n}$  where  $m \leq n$ . Now,  $m{P} = m{A}^T m{A} \in \mathbb{R}^{n \times n}$  and  $m{Q} = m{A} m{A}^T \in \mathbb{R}^{m \times m}$ 

(a) Now for  $\mathbf{y} \in \mathbb{R}^n$ , we have

$$\begin{aligned} \boldsymbol{y}^T \boldsymbol{P} \boldsymbol{y} &= \boldsymbol{y}^T \boldsymbol{A}^T \boldsymbol{A} \boldsymbol{y} \\ &= (\boldsymbol{A} \boldsymbol{y})^T \boldsymbol{A} \boldsymbol{y} \\ &= \|\boldsymbol{A} \boldsymbol{y}\|_2^2 \end{aligned} \tag{1}$$
Hence  $\boldsymbol{y}^T \boldsymbol{P} \boldsymbol{y} \geq 0$   $(\|\cdot\| \geq 0)$ 

Similarly, for  $z \in \mathbb{R}^m$ , we have

$$z^{T}Qz = z^{T}AA^{T}z$$

$$= (A^{T}z)^{T}Az$$

$$= ||A^{T}z||_{2}^{2}$$
(2)

Hence  $\boldsymbol{z}^T \boldsymbol{Q} \boldsymbol{z} \geq 0$ 

Now, let  $\boldsymbol{u} \in \mathbb{R}^n$  be an eigenvector of  $\boldsymbol{P}$  with eigenvalue  $\lambda$ .

$$\lambda < 0 \Rightarrow \boldsymbol{u}^{T} \boldsymbol{P} \boldsymbol{u} = \boldsymbol{u}^{T} (\boldsymbol{P} \boldsymbol{u})$$

$$= \boldsymbol{u}^{T} (\lambda \boldsymbol{u})$$

$$= \lambda \boldsymbol{u}^{T} \boldsymbol{u}$$

$$= \lambda \|\boldsymbol{u}\|_{2}^{2}$$
Hence  $\lambda < 0 \Rightarrow \boldsymbol{u}^{T} \boldsymbol{P} \boldsymbol{u} < 0$ 
(3)

But  $\mathbf{y}^T \mathbf{P} \mathbf{y} \geq 0$  for any  $\mathbf{y} \in \mathbb{R}^n$ , hence the eigenvalues of  $\mathbf{P}$  are nonnegative.

Similarly, let  $v \in \mathbb{R}^m$  be an eigenvector of Q with eigenvalue  $\mu$ .

$$\mu < 0 \Rightarrow \mathbf{v}^{T} \mathbf{Q} \mathbf{v} = \mathbf{v}^{T} (\mathbf{Q} \mathbf{v})$$

$$= \mathbf{v}^{T} (\mu \mathbf{v})$$

$$= \mu \mathbf{v}^{T} \mathbf{v}$$

$$= \mu \|\mathbf{v}\|_{2}^{2}$$

$$(4)$$

Hence  $\mu < 0 \Rightarrow \boldsymbol{v}^T \boldsymbol{Q} \boldsymbol{v} < 0$ 

But  $v^T Q v \ge 0$  for any  $y \in \mathbb{R}^m$ , hence the eigenvalues of Q are also non-negative.

(b) Here,  $\boldsymbol{u}$  eigenvector of  $\boldsymbol{P}$  with eigenvalue  $\lambda$  and  $\boldsymbol{v}$  eigenvector of  $\boldsymbol{Q}$  with eigenvalue  $\mu$ . They have n and m elements respectively.

Now,

$$Q(Au) = AA^{T}(Au)$$

$$= A(A^{T}A)u$$

$$= A(Pu)$$

$$= A\lambda u \qquad (Pu = \lambda u)$$
Hence  $Q(Au) = \lambda(Au)$ 

Similarly,

$$P(\mathbf{A}^{T}\mathbf{v}) = \mathbf{A}^{T}\mathbf{A}(\mathbf{A}^{T}\mathbf{v})$$

$$= \mathbf{A}^{T}(\mathbf{A}\mathbf{A}^{T})\mathbf{v}$$

$$= \mathbf{A}^{T}(\mathbf{Q}\mathbf{v})$$

$$= \mathbf{A}^{T}\mu\mathbf{v} \qquad (\mathbf{Q}\mathbf{v} = \mu\mathbf{v})$$
Hence  $P(\mathbf{A}^{T}\mathbf{v}) = \mu(\mathbf{A}^{T}\mathbf{v})$ 

Hence, Au is an eigenvector of Q with eigenvalue  $\lambda$  and  $A^Tv$  is an eigenvector of P with eigenvalue  $\mu$ .

(c) Here,  $v_i$  eigenvector of Q with eigenvalue  $\mu$  and  $u_i \triangleq \frac{A^T v_i}{\|A^T v_i\|_2}$ Consider,

$$A\boldsymbol{u}_{i} = A \frac{A^{T}\boldsymbol{v}_{i}}{\|A^{T}\boldsymbol{v}_{i}\|_{2}}$$

$$= \frac{AA^{T}\boldsymbol{v}_{i}}{\|A^{T}\boldsymbol{v}_{i}\|_{2}}$$

$$= \frac{Q\boldsymbol{v}_{i}}{\|A^{T}\boldsymbol{v}_{i}\|_{2}}$$

$$= \frac{\mu\boldsymbol{v}_{i}}{\|A^{T}\boldsymbol{v}_{i}\|_{2}}$$

$$= \frac{\mu\boldsymbol{v}_{i}}{\|A^{T}\boldsymbol{v}_{i}\|_{2}}$$
(7)

Hence, there exists  $\gamma_i = \frac{\mu}{\|\mathbf{A}^T \mathbf{v}_i\|_2}$  such that  $\mathbf{A} \mathbf{u}_i = \gamma_i \mathbf{v}_i$ . Also  $\gamma_i$  is non-negative as the eigenvalue of  $\mathbf{Q}$  ( $\mu$ ) is non-negative as shown in (a).

(d) We know that P will have n eigenvectors and Q will have m eigenvectors. Since  $m \le n$ , the following holds

$$\mathbf{A}\mathbf{u}_{i} = \begin{cases} \gamma_{i}\mathbf{v}_{i} & \text{if } i \in \{1, 2, \dots, m\} \text{ (from (c))} \\ 0_{m} & \text{if } i \in \{m + 1, m + 2, \dots, n\} \end{cases}$$
(8)

Construct  $U = [v_1 \quad v_2 \quad \cdots \quad v_m], V = [u_1 \quad u_2 \quad \cdots \quad u_n]$  and  $\Gamma$  as a  $m \times n$  rectangular diagonal matrix with diagonal entries  $\gamma_i$  as shown below

$$\Gamma = \begin{bmatrix}
\gamma_1 & 0 & \cdots & 0 & 0 & \cdots & 0 \\
0 & \gamma_2 & \cdots & 0 & 0 & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & \gamma_m & 0 & \cdots & 0
\end{bmatrix}$$
(9)

Note that U, V are orthogonal matrices as  $u_i^T u_j = \delta_{ij}$ ,  $v_i^T v_j = \delta_{ij}$  (proved in the footnote 1 of the homework sheet). Now, 8 can also be written as

$$AV = U\Gamma$$

$$A = U\Gamma V^{T} \quad (V^{-1} = V^{T})$$
(10)

Equation 10 gives us the singular value decomposition of any matrix  $\boldsymbol{A}$ , where  $\boldsymbol{U}, \boldsymbol{V}$  are the corresponding orthogonal matrices and  $\boldsymbol{\Gamma}$  corresponds to  $\boldsymbol{\Sigma}$  matrix and  $\gamma_i$  indeed corresponds to singular values as they are also non-negative as proved in (c).