

Assignment 4

Question 6

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- (a) Given $P = A^T A$ ($n \times n$ matrix) and $Q = A A^T$ ($m \times m$ matrix), and y is a $n \times 1$ real-valued vector, we have (given the correct dimension of real-valued y):

$$\begin{aligned} y^T P y &= y^T A^T A y \\ &= (A y)^T (A y) \end{aligned}$$

Now let $A y = w$ is a $n \times 1$ vector. Hence, we have

$$\begin{aligned} (A y)^T (A y) &= w^T w = \|w\|^2 \geq 0 \\ y^T P y &\geq 0 \quad \forall y \end{aligned}$$

Similarly, for a real-valued $m \times 1$ vector z , we have (given the correct dimension of real-valued z):

$$\begin{aligned} z^T Q z &= z^T A A^T z \\ &= (A^T z)^T (A^T z) \end{aligned}$$

Now let $A^T z = u$ is a $m \times 1$ vector. Hence, we have

$$\begin{aligned} (A^T z)^T (A^T z) &= u^T u = \|u\|^2 \geq 0 \\ z^T Q z &\geq 0 \quad \forall z \end{aligned}$$

Proof of non-negative eigenvalues of P and Q : For P , let the non-zero eigen vector be v with corresponding eigenvalue λ :

$$P v = \lambda v$$

Pre-multiplying both sides by v^T , we get,

$$\begin{aligned} v^T P v &= \lambda v^T v = \lambda \|v\|^2 \\ \implies \lambda &= \frac{v^T P v}{\|v\|^2} \end{aligned}$$

Now, $\|v\|^2$ is always positive (by-definition). Also, $v^T P v \geq 0$ (proved earlier). Hence, value of λ is always non-negative. Since we made no assumptions for v or λ , the condition holds for all eigenvalues of P .

Similarly, for Q , let:

$$Q v = \lambda v$$

Pre-multiplying both sides by v^T , we get,

$$v^T Q v = \lambda v^T v = \lambda \|v\|^2$$

$$\implies \lambda = \frac{v^T Q v}{\|v\|^2}$$

Now, $\|v\|^2$ is always positive (by-definition). Also, $v^T Q v \geq 0$ (proved earlier). Hence, value of λ is always non-negative. Since we made no assumptions for v or λ , the condition holds for all eigenvalues of Q .

(b) u is an eigenvector of P with eigenvalue λ . Now,

$$A \in R^{m \times n} \implies A^T A = P \in R^{n \times n} \implies u \in R^n \quad \text{and}$$

$$P u = \lambda u$$

Pre-multiplying both sides with A , we get

$$A P u = \lambda A u$$

Using $P = A^T A$, we simplify this to,

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

Simplify $A A^T = Q$, and $A u = v \in R^m$ to get,

$$Q(A u) = Q v = \lambda(A u) = \lambda v$$

Hence, $A u$ is an eigenvector of Q with eigenvalue λ .

Similarly to prove for Q , we assume an eigenvector v with eigenvalue μ . Now,

$$A \in R^{m \times n} \implies A A^T = Q \in R^{m \times m} \implies v \in R^m \quad \text{and}$$

$$Q v = \mu v$$

Pre-multiplying both sides with A^T , we get

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

Using $Q = A A^T$, we simplify this to,

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

Simplify $A^T A = P$, and $A^T v = w \in R^n$ to get,

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

Hence, $A^T v$ is an eigenvector of P with eigenvalue μ .

- (c) We are given that v_i is an eigenvector of Q (with eigenvalue α_i). Which gives us the first equation

$$Qv_i = AA^T v_i = \alpha_i v_i \quad (i)$$

Now, given

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

we have

$$\begin{aligned} Au_i &= \frac{AA^T v_i}{\|A^T v_i\|_2} \\ \implies Au_i &= \frac{Qv_i}{\|A^T v_i\|_2} = \frac{\alpha_i v_i}{\|A^T v_i\|_2} = \left(\frac{\alpha_i}{\|A^T v_i\|_2} \right) v_i \end{aligned}$$

Substitute $\left(\frac{\alpha_i}{\|A^T v_i\|_2} \right) = \gamma_i$ to get the required equation.

$$Au_i = \gamma_i v_i$$

Now, α_i is non-negative since it's an eigenvalue of Q (proved earlier), and $\|A^T v_i\|_2$ is positive by definition, the value of γ_i is non-negative.

There exists a non-negative γ_i for which the given equation holds.

- (d) We know as a result that

$$u_i^T u_j = 0 \quad \text{for } i \neq j$$

and for

$$u_i^T u_i = \frac{(A^T v_i)^T (A^T v_i)}{\|A^T v_i\|^2} = 1$$

So, the matrix $V = [u_1 | u_2 | \dots | u_n]$ is orthonormal because $VV^T = V^T V = I_n$

Similarly, for $U = [v_1 | v_2 | \dots | v_m]$, $UU^T = U^T U = I_m$

because $v_i^T v_j = 0$ for $i \neq j$ and $v_i^T v_i = 1$ (because v_i and v_j are eigenvectors of P , and which can be assumed to be of unit length for consistency).

Given these results, the value of $U^T AV$

$$\begin{aligned} &= U^T AV = U^T A[u_1 | u_2 | \dots | u_n] \\ &= \begin{bmatrix} v_1^T \\ v_2^T \\ \vdots \\ v_m^T \end{bmatrix} A[u_1 | u_2 | \dots | u_n] \end{aligned}$$

From the previous result, we have $Au_i = \gamma_i v_i$

$$= \begin{bmatrix} v_1^T \\ v_2^T \\ \vdots \\ v_m^T \end{bmatrix} [\gamma_1 v_1 | \gamma_2 v_2 | \dots | \gamma_n v_n]$$

$$= [\Gamma_{ij}^T]_{m \times n}$$

Where

$$\Gamma_{ij} = v_i^T \gamma_j v_j = \begin{cases} 0 & \text{if } i \neq j \\ \gamma_j & \text{otherwise} \end{cases}$$

Hence, $U^T A V = \Gamma$ where T is a diagonal matrix, with i^{th} diagonal element $= \gamma_i$

$$\implies (U U^T) A (V V^T) = U \Gamma V^T$$

$$\implies A = U \Gamma V^T$$

This holds since U and V are orthonormal.

Hence, proved.