

DIP Assignment 5: Question 4

October 22, 2015

1 Question

Consider a matrix \mathbf{A} of size $m \times n$. Define $\mathbf{P} = \mathbf{A}^T \mathbf{A}$ and $\mathbf{Q} = \mathbf{A} \mathbf{A}^T$. (Note: all matrices, vectors and scalars involved in this question are real-valued).

1. Prove that for any vector \mathbf{y} with appropriate number of elements, we have $\mathbf{y}^t \mathbf{P} \mathbf{y} \geq 0$. Similarly show that $\mathbf{z}^t \mathbf{Q} \mathbf{z} \geq 0$ for a vector \mathbf{z} with appropriate number of elements. Why are the eigenvalues of \mathbf{P} and \mathbf{Q} non-negative?

1.1 Solution

Substituting for P in $\mathbf{y}^t \mathbf{P} \mathbf{y}$, we get

$$\begin{aligned}\mathbf{y}^t \mathbf{A}^T \mathbf{A} \mathbf{y} \\ &= (\mathbf{A} \mathbf{y})^t \mathbf{A} \mathbf{y} \\ &= \mathbf{b}^t \mathbf{b}\end{aligned}$$

where $b = \mathbf{A} \mathbf{y}$. Now, dimension of b is $m \times 1$. So, $\mathbf{b}^t \mathbf{b}$ is a scalar which is nothing but inner product of b with itself, i.e. $\langle b, b \rangle$.

Thus, $\mathbf{y}^t \mathbf{P} \mathbf{y}$ is sum of square of elements of $\mathbf{A} \mathbf{y}$, which is ought to be non-negative. So,

$$\mathbf{y}^t \mathbf{P} \mathbf{y} \geq 0$$

On similar lines, $\mathbf{z}^t \mathbf{Q} \mathbf{z}$ can be expanded as

$$\begin{aligned}\mathbf{z}^t \mathbf{A} \mathbf{A}^T \mathbf{z} \\ &= (\mathbf{A}^T \mathbf{z})^t \mathbf{A}^T \mathbf{z} \\ &= \mathbf{c}^t \mathbf{c}\end{aligned}$$

where $c = \mathbf{A}^T \mathbf{z}$. Now, dimension of c is $n \times 1$. So, $\mathbf{c}^t \mathbf{c}$ is a scalar. Thus, $\mathbf{z}^t \mathbf{Q} \mathbf{z}$ is sum of square of elements of $\mathbf{A}^T \mathbf{z}$, which is ought to be non-negative. So,

$$\mathbf{z}^t \mathbf{Q} \mathbf{z} \geq 0$$

Say, \mathbf{u} is an eigenvector of \mathbf{P} with eigenvalue λ , then,

$$\mathbf{P} \mathbf{u} = \lambda \mathbf{u}$$

Pre-multiplying both sides by \mathbf{u}^T , we have

$$\mathbf{u}^T \mathbf{P} \mathbf{u} = \mathbf{u}^T \lambda \mathbf{u}$$

$$\mathbf{u}^T \mathbf{P} \mathbf{u} = \lambda \mathbf{u}^T \mathbf{u}$$

Now, from the result proved above, LHS is always non-negative. $\mathbf{u}^T \mathbf{u}$ is a scalar which is sum of square of elements of \mathbf{u} , and so is non-negative. Thus,

$$\lambda = \frac{\mathbf{u}^T \mathbf{P} \mathbf{u}}{\mathbf{u}^T \mathbf{u}}$$

λ is a product of two non-negative quantities and so is non-negative.

Following the similar arguments for eigenvalues and eigenvectors of \mathbf{Q} and using the result for \mathbf{Q} proved above, we can prove that eigenvalues of \mathbf{Q} are always non-negative.

2. If \mathbf{u} is an eigenvector of \mathbf{P} with eigenvalue λ , show that $\mathbf{A} \mathbf{u}$ is an eigenvector of \mathbf{Q} with eigenvalue λ . If \mathbf{v} is an eigenvector of \mathbf{Q} with eigenvalue μ , show that $\mathbf{A}^T \mathbf{v}$ is an eigenvector of \mathbf{P} with eigenvalue μ . What will be the number of elements in \mathbf{u} and \mathbf{v} ?

1.2 Solution

We have,

$$\mathbf{P} \mathbf{u} = \lambda \mathbf{u}$$

$$\mathbf{A}^T \mathbf{A} \mathbf{u} = \lambda \mathbf{u}$$

Pre-multiplying both sides by \mathbf{A} , we get

$$\mathbf{A} \mathbf{A}^T \mathbf{A} \mathbf{u} = \mathbf{A} \lambda \mathbf{u}$$

replacing $\mathbf{A} \mathbf{A}^T$ by \mathbf{Q}

$$\mathbf{Q} \mathbf{A} \mathbf{u} = \lambda \mathbf{A} \mathbf{u}$$

where, dimension of $\mathbf{A}\mathbf{u}$ is $m \times 1$. Thus, $\mathbf{A}\mathbf{u}$ is an eigenvector of \mathbf{Q} with corresponding eigenvalue as λ

On similar lines,

$$\begin{aligned}\mathbf{Q}\mathbf{v} &= \mu\mathbf{v} \\ \mathbf{A}\mathbf{A}^T\mathbf{v} &= \mu\mathbf{v}\end{aligned}$$

Pre-multiplying both sides by \mathbf{A}^T , we get

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

replacing $\mathbf{A}^T\mathbf{A}$ by \mathbf{P}

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

where, dimension of $\mathbf{A}^T\mathbf{v}$ is $n \times 1$. Thus, $\mathbf{A}^T\mathbf{v}$ is an eigenvector of \mathbf{P} with corresponding eigenvalue as μ .

\mathbf{u} is an eigenvector of $n \times n$ matrix \mathbf{P} and so has n elements.

\mathbf{v} is an eigenvector of $m \times m$ matrix \mathbf{Q} and so has m elements.

3. If \mathbf{v}_i is an eigenvector of \mathbf{Q} and we define $\mathbf{u}_i = \frac{\mathbf{A}^T\mathbf{v}_i}{\|\mathbf{A}^T\mathbf{v}_i\|}$. Then prove that there will exist some real, non-negative γ_i such that $\mathbf{A}\mathbf{u}_i = \gamma_i\mathbf{v}_i$.

1.3 Solution

$$\mathbf{u}_i = \frac{\mathbf{A}^T\mathbf{v}_i}{\|\mathbf{A}^T\mathbf{v}_i\|}$$

So,

$$\mathbf{A}\mathbf{u}_i = \mathbf{A} \frac{\mathbf{A}^T\mathbf{v}_i}{\|\mathbf{A}^T\mathbf{v}_i\|}$$

replacing $\mathbf{A}\mathbf{A}^T$ by \mathbf{Q}

$$\mathbf{A}\mathbf{u}_i = \mathbf{Q} \frac{\mathbf{v}_i}{\|\mathbf{A}^T\mathbf{v}_i\|}$$

Now, as \mathbf{v}_i is an eigenvector of \mathbf{Q} , let λ_i be the corresponding eigenvalue. We now have,

$$\mathbf{A}\mathbf{u}_i = \frac{\lambda_i}{\|\mathbf{A}^T\mathbf{v}_i\|} \mathbf{v}_i$$

Now, λ_i , being eigenvalue of \mathbf{Q} is non-negative as proved in part 1; and so is $\|\mathbf{A}^T \mathbf{v}_i\|$. So, replacing $\frac{\lambda_i}{\|\mathbf{A}^T \mathbf{v}_i\|}$ by some non-negative and real γ_i , we have

$$\mathbf{A} \mathbf{u}_i = \gamma_i \mathbf{v}_i$$

4. It can be shown that $\mathbf{u}_i^T \mathbf{u}_j = 0$ for $i \neq j$ and likewise $\mathbf{v}_i^T \mathbf{v}_j = 0$ for $i \neq j$ for correspondingly distinct eigenvalues. Now, define $\mathbf{U} = [\mathbf{v}_1 | \mathbf{v}_2 | \mathbf{v}_3 | \dots | \mathbf{v}_m]$ and $\mathbf{V} = [\mathbf{u}_1 | \mathbf{u}_2 | \mathbf{u}_3 | \dots | \mathbf{u}_n]$. Now show that $\mathbf{A} = \mathbf{U} \mathbf{\Gamma} \mathbf{V}^T$ where $\mathbf{\Gamma}$ is a diagonal matrix containing the non-negative values $\gamma_1, \gamma_2, \dots, \gamma_n$. With this, you have just established the existence of the singular value decomposition of any matrix \mathbf{A} . This is a key result in linear algebra and it is widely used in image processing, computer vision, computer graphics, statistics, machine learning, numerical analysis, natural language processing and data mining. [5 + 5 + 5 + 5 = 20 points]

1.4 Solution

Let us first consider $\mathbf{U} \mathbf{\Gamma}$ with the matrices as defined in the question. We have

$$\mathbf{U} \mathbf{\Gamma} = [\mathbf{v}_1 | \mathbf{v}_2 | \dots | \mathbf{v}_m] \begin{pmatrix} \gamma_1 & 0 & \dots & 0 \\ 0 & \gamma_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \gamma_m \end{pmatrix}$$

which can also be written as

$$\mathbf{U} \mathbf{\Gamma} = [\gamma_1 \mathbf{v}_1 | \gamma_2 \mathbf{v}_2 | \dots | \gamma_m \mathbf{v}_m]$$

As proved in the previous part, replacing $\gamma_i \mathbf{v}_i$ by $\mathbf{A} \mathbf{u}_i$, we have

$$\mathbf{U} \mathbf{\Gamma} = [\mathbf{A} \mathbf{u}_1 | \mathbf{A} \mathbf{u}_2 | \dots | \mathbf{A} \mathbf{u}_n]$$

$$\mathbf{U} \mathbf{\Gamma} = \mathbf{A} [\mathbf{u}_1 | \mathbf{u}_2 | \dots | \mathbf{u}_n]$$

Substituting this in $\mathbf{U} \mathbf{\Gamma} \mathbf{V}^T$,

$$\mathbf{U} \mathbf{\Gamma} \mathbf{V}^T = \mathbf{A} [\mathbf{u}_1 | \mathbf{u}_2 | \dots | \mathbf{u}_n] [\mathbf{u}_1 | \mathbf{u}_2 | \dots | \mathbf{u}_n]^T$$

$$\mathbf{U} \mathbf{\Gamma} \mathbf{V}^T = \mathbf{A} \mathbf{V} \mathbf{V}^T$$

Now $\mathbf{V} \mathbf{V}^T$ will be \mathbf{I}_n because orthogonality between all the eigenvectors and assuming that the eigenvectors are normalized, all eigenvectors will be orthonormal. Basically using,

$$\mathbf{u}_i^T \mathbf{u}_j = 0 \text{ for } i \neq j$$

and

$$\mathbf{u}_i^T \mathbf{u}_j = 1 \text{ for } i = j$$

So, all diagonal elements will be 1 and rest all the elements will be zero. Thus,

$$\mathbf{U}\mathbf{V}^T = \mathbf{A}$$

Thus proved.