

homework 1

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1 1

1.1 1

Lemma 1.1. *Suppose \mathbf{A} with eigenvalues λ_i is symmetric, then*

$$\text{eig}(\mathbf{I} + c\mathbf{A}) = 1 + c\lambda_i$$

$$\text{eig}(\mathbf{A} - c\mathbf{I}) = \lambda_i - c$$

Proof. Note

$$|\mathbf{I} + c\mathbf{A} - \lambda\mathbf{I}| = |\mathbf{I} + c\lambda_i\mathbf{I} - \lambda\mathbf{I}| = 0 \Rightarrow \lambda = 1 + c\lambda_i$$

the other one can be proved similarly. □

Lemma 1.2. Suppose $\mathbf{A} \in \mathbb{R}^{n \times m}$ and $\mathbf{B} \in \mathbb{R}^{m \times n}$, then

$$\text{eig}(\mathbf{AB}) = \text{eig}(\mathbf{BA})$$

Proof. Suppose $\lambda_i \in \text{eig}(\mathbf{AB})$, then

$$\mathbf{ABx} = \lambda_i \mathbf{x} \Rightarrow \mathbf{BABx} = \lambda_i \mathbf{Bx} \Rightarrow \lambda_i \in \text{eig}(\mathbf{BA})$$

and the other direction follows from symmetry. □

Then we have following corollary:

Corollary 1.1. Suppose \mathbf{x} is a n -vector, then

$$\text{eig}(\mathbf{xx}') = \underbrace{0, 0, \dots, 0}_{n-1}, \mathbf{x}'\mathbf{x}$$

Note

$$\mathbf{A} = \mathbf{I} - \rho(-\mathbf{I} + \mathbf{ee}')$$

Where \mathbf{e} is p all-one vector and \mathbf{ee}' has one eigenvalue of p and $p - 1$ eigenvalues of 0, then \mathbf{A} has $p - 1$ eigenvalues of $1 - \rho$ and one $1 + (p - 1)\rho$ and thus

$$|\mathbf{A}| = (1 - \rho)^{p-1} [1 + (p - 1)\rho]$$

1.2 2

Let $|A| = 0$, we have $(1 - \rho) = 0$ or $1 + (p - 1)\rho = 0$ and thus

$$\rho = \begin{cases} 1 \\ -\frac{1}{p-1} \end{cases}$$

1.3 3

Already done in 1.

2 2

Note eigenvalues of $c\mathbf{A}$ is $c\lambda_i$. Plug $\rho = .5$ and $p = 3$, we find

$$\text{eig}(\mathbf{A}) = 4, 1, 1$$

and the corresponding eigenvectors are:

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}, \mathbf{v}_3 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

3 3

3.1 1

$$\begin{aligned} \sum_{i=1}^n (x_i - \bar{x})^2 &= (\mathbf{x} - \bar{x}\mathbf{e})'(\mathbf{x} - \bar{x}\mathbf{e}) \\ &= \left(\mathbf{x} - \frac{\mathbf{e}'\mathbf{x}}{\mathbf{e}'\mathbf{e}}\mathbf{e} \right)' \left(\mathbf{x} - \frac{\mathbf{e}'\mathbf{x}}{\mathbf{e}'\mathbf{e}}\mathbf{e} \right) \\ &= \mathbf{x}'(\mathbf{I} - \mathbf{P}_e)'(\mathbf{I} - \mathbf{P}_e)\mathbf{x} \\ &= \mathbf{x}'(\mathbf{I} - \mathbf{P}_e)\mathbf{x} \end{aligned}$$

where $\mathbf{P}_e = \frac{\mathbf{e}\mathbf{e}'}{\mathbf{e}'\mathbf{e}}$ and thus

$$\mathbf{A} = \mathbf{I} - \mathbf{P}_e$$

3.2 2

It's symmetric and idempotent and thus a projection. In fact we have

Lemma 3.1. *If \mathbf{P} is a projection matrix, so is $\mathbf{I} - \mathbf{P}$*

Proof. Symmetric follows from both \mathbf{I} and \mathbf{P} is symmetric and idempotent follows from

$$(\mathbf{I} - \mathbf{P})^2 = (\mathbf{I}^2 + \mathbf{P}^2 - \mathbf{PI} - \mathbf{IP}) = \mathbf{I} - \mathbf{P}$$

□

3.3 3

Nonnegative definite as $\sum_{i=1}^n (x_i - \bar{x})^2 \geq 0$ clearly.

3.4 4

$$\begin{aligned} \text{rank}(\mathbf{I} - \mathbf{P}_e) &= \text{tr}(\mathbf{I} - \mathbf{P}_e) \\ &= \text{tr}(\mathbf{I}) - \text{tr}(\mathbf{P}_e) \\ &= n - \text{rank}(\mathbf{P}_e) \\ &= n - 1 \end{aligned}$$

4 4

By 1 we have find it's eigenvalues are $1 + \rho$ and $1 - \rho$, thus the corresponding eigenvectors is

$$v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, v_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

and thus

$$\mathbf{A} = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 - \rho & \\ & 1 + \rho \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix}$$

Similarly,

$$\mathbf{A}^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{1-\rho} & \\ & \frac{1}{\rho+1} \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 1 \\ 1 & 1 \end{bmatrix}$$

and we find that $\frac{1}{\text{eig}(\mathbf{A})} = \text{eig}(\mathbf{A}^{-1})$.

5 5

Note

$$\text{eig}(\mathbf{A}) = \{6 \pm \sqrt{33}, 10\}$$

thus the maximum is $6 + \sqrt{33}$ and minimum is $6 - \sqrt{33}$.

6 6

6.1 1

Note $\frac{\partial \mathbf{x}'\mathbf{A}\mathbf{x}}{\partial \mathbf{x}} = (\mathbf{A} + \mathbf{A}')\mathbf{x}$ for general \mathbf{A} and \mathbf{x} :

$$\frac{\partial \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{x}}}{\partial \mathbf{x}} \propto 2\mathbf{A}\mathbf{x}(\mathbf{x}'\mathbf{x}) - \mathbf{x}'\mathbf{A}\mathbf{x}2\mathbf{x} = 0 \Rightarrow \mathbf{A}\mathbf{x} = \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{x}}\mathbf{x}$$

that implies the extreme value occurs when \mathbf{x} is the eigenvectors and the eigenvalues is the value of $\frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{x}}$. Thus

$$\begin{aligned} \max_{x \neq 0} \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{x}} &= \lambda_1 \\ \min_{x \neq 0} \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{x}} &= \lambda_p \end{aligned}$$

and holds when $\mathbf{x} = \mathbf{t}_1$ and $\mathbf{x} = \mathbf{t}_p$ respectively.

6.2 2

Lemma 6.1. *If \mathbf{A} is positive definite then*

$$2\mathbf{x}'\mathbf{A}\mathbf{x} = \mathbf{x}'(\mathbf{A} + \mathbf{A}')\mathbf{x}$$

Proof. As $\mathbf{x}'\mathbf{A}\mathbf{x}$ is a scalar, then $\mathbf{x}'\mathbf{A}\mathbf{x} = \mathbf{x}'\mathbf{A}'\mathbf{x}$ thus $\mathbf{x}'(\mathbf{A} - \mathbf{A}')\mathbf{x} = \mathbf{0}$

$$\begin{aligned}\mathbf{x}'(\mathbf{A} + \mathbf{A}')\mathbf{x} &= \mathbf{x}'[2\mathbf{A} - (\mathbf{A} - \mathbf{A}')]\mathbf{x} \\ &= 2\mathbf{x}'\mathbf{A}\mathbf{x}\end{aligned}$$

□

By lemma 6.1:

$$\begin{aligned}\frac{\partial \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{B}\mathbf{x}}}{\partial \mathbf{x}} &\propto 2\mathbf{A}\mathbf{x}(\mathbf{x}'\mathbf{B}\mathbf{x}) - \mathbf{x}'\mathbf{A}\mathbf{x}(\mathbf{B} + \mathbf{B}')\mathbf{x} \\ &= 2\mathbf{A}\mathbf{x}(\mathbf{x}'\mathbf{B}\mathbf{x}) - \mathbf{x}'\mathbf{A}\mathbf{x}2\mathbf{B}\mathbf{x} \\ &\Rightarrow \mathbf{A}\mathbf{x} = \frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{B}\mathbf{x}}\mathbf{B}\mathbf{x}\end{aligned}$$

thus \mathbf{x} is eigenvectors of $\mathbf{B}^{-1}\mathbf{A}$ and correspond to $\frac{\mathbf{x}'\mathbf{A}\mathbf{x}}{\mathbf{x}'\mathbf{B}\mathbf{x}}$ and the claim follows easily.