

STAT35

Discussion 2

①

Def: Let A be symmetric matrix. A is positive definite if $x^T A x > 0$ for every non-zero vector x .

$x^T A x :=$ a quadratic form

Exercise

Show that $A = \begin{bmatrix} 1 & -2 \\ -2 & 6 \end{bmatrix}$ is positive definite (p.d.)

$$\begin{aligned} (x_1 \ x_2) \begin{pmatrix} 1 & -2 \\ -2 & 6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= (x_1 \ x_2) \begin{pmatrix} x_1 - 2x_2 \\ -2x_1 + 6x_2 \end{pmatrix} \\ &= x_1^2 - 2x_1x_2 - 2x_1x_2 + 6x_2^2 \\ &= x_1^2 - 4x_1x_2 + 6x_2^2 \end{aligned}$$

Key is to rewrite RHS as a sum of squares

(i) $(x_1 - 2x_2)^2 = x_1^2 - 4x_1x_2 + 4x_2^2 > 0$

(ii) add $2x_2^2$ and we have

$$x^T A x = (x_1 - 2x_2)^2 + 2x_2^2 > 0$$

Also $A^T = \begin{pmatrix} 1 & -2 \\ -2 & 6 \end{pmatrix}$ is symmetric

so A is p.d.

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The Spectral Decomposition

Def: Every symmetric matrix has the factorization $A = P\Lambda P'$ with real eigenvalues in Λ and orthonormal eigenvectors in P .

Note: $A = \sum_{j=1}^n \lambda_j e_j e_j'$ where

e_j is an eigenvector of unit length with corresponding eigenvalue λ_j .

Exercise

Derive the spectral decomposition for

$$A = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$$

Step 1 Find eigenvalues.

$|A - \lambda I| = 0$ the characteristic equation

$$\begin{vmatrix} 2-\lambda & -1 \\ -1 & 2-\lambda \end{vmatrix} = (2-\lambda)^2 - 1 = \lambda^2 - 4\lambda + 3 = 0$$

$$(\lambda - 1)(\lambda - 3) = 0 \Rightarrow \lambda_1 = 1, \lambda_2 = 3$$

Note: Since $\lambda_1, \lambda_2 > 0$ A is also p.d.

You can verify this by definition application.

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Step 2 Find eigenvectors

$$\lambda_1 = 1: (A - \lambda_1 I)x = 0 \quad \text{solve for } x$$

Plug in for $\lambda_1 = 1$

$$\begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$x_1 - x_2 = 0 \quad \text{take } x_1 = 1 \Rightarrow x_2 = 1$$

$$x_2 - x_1 = 0$$

$$\underline{x}' = (1 \ 1) \longrightarrow \text{unit length: } \sqrt{x'x} = \sqrt{2}$$

$$e'_1 = \left(\frac{1}{\sqrt{2}} \quad \frac{1}{\sqrt{2}} \right)$$

$$\lambda_2 = 3:$$

$$\begin{pmatrix} -1 & -1 \\ -1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0$$

$$-x_1 - x_2 = 0 \quad \text{take } x_1 = 1 \Rightarrow x_2 = -1$$

$$\underline{x}' = (1 \ -1) \longrightarrow \text{unit length}$$

$$e'_2 = \left(\frac{1}{\sqrt{2}} \quad -\frac{1}{\sqrt{2}} \right)$$

(4)

Step 3 Express the decomposition

$$\begin{aligned}
 A &= \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix} \\
 &= 1 \times \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} + 3 \times \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{pmatrix}
 \end{aligned}$$

What would be the square root matrix in this case?

$$A = P \Lambda^{1/2} P' = \sum_{j=1}^2 \sqrt{\lambda_j} e_j e_j'$$

(4)

Application of Positive Definite Matrices

For positive definite matrix A ,
the graph of $x^T A x = c^2$
is an ellipse (hyperellipsoid for $p > 2$).

Exercise 2.17

Claim: Every eigenvalue of a
 $n \times n$ p.d. matrix A is positive.

Proof

$$Ae = \lambda e$$

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

$$= \lambda (e_1^2 + e_2^2 + \dots + e_p^2)$$

By p.d. of A $e^T A e > 0$ for non-zero e ,
and $e^T e > 0 \Rightarrow \lambda > 0$.

Remark: There is a strong geometric
reason why the eigenvalues are positive,
so we turn our attention back to the
ellipse to investigate.

(5)

Ellipses and Positive Definite Matrices

Let's start with the p.d. matrix

$$A = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}.$$

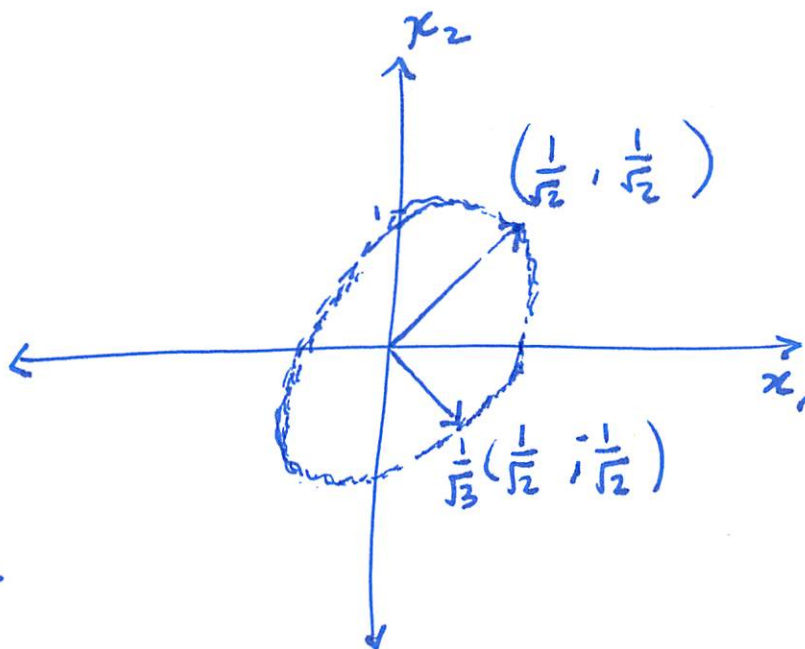
Equation for Ellipse

$$x^T A x = c^2$$

$$(x_1, x_2) \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = c^2$$

$$(x_1, x_2) \begin{pmatrix} 2x_1 - x_2 \\ 2x_2 - x_1 \end{pmatrix} = c^2$$

$$2x_1^2 - 2x_1x_2 + 2x_2^2 = c^2$$



When $c = 1$,
the ellipse looks like this
and expands and shrinks according
to the distance c .

How ought one to know the geometry
of the ellipse?

- know the major and minor axis
direction and magnitude (a vector).

The major and minor axis of the ellipse
is described by the eigenvectors of A .

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Since A was ~~stated~~ p.d. with

$$\lambda_1 = 1, \lambda_2 = 3 \quad \text{and} \quad e_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix} \quad e_2 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$x^T A x = c^2$ can be re-expressed by the spectral decomposition
 $x^T P \Lambda P^T x = c^2$

$$x^T \left(\sum_{j=1}^p \lambda_j e_j e_j^T \right) x = \sum_{j=1}^p \lambda_j x^T e_j e_j^T x = \sum_{j=1}^p \lambda_j (x^T e_j)^2$$

Now in the $p=2$ case, the ellipse is seen to be

$$c^2 = \lambda_1 \tilde{x}_1^2 + \lambda_2 \tilde{x}_2^2 \quad \swarrow \text{notice no cross-product term}$$

where $\tilde{x}_j = x^T e_j$ and \tilde{x}_1 and \tilde{x}_2 are the ~~major and minor~~ ^{principle} axes of the ellipse. In the example, we have

$$c^2 = 1 * \left(\frac{x_1 + x_2}{\sqrt{2}} \right)^2 + 3 * \left(\frac{x_1 - x_2}{\sqrt{2}} \right)^2$$

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In summary,

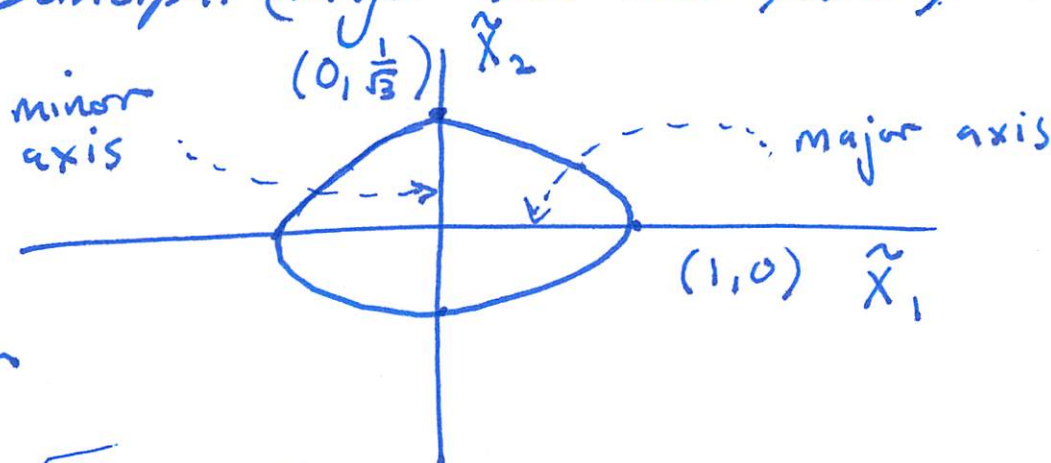
The eigenvectors dictate the major and minor axes of the ellipse direction and the eigenvalues scale the magnitude.

If $\lambda_1 \neq \lambda_2$, then the

~~major~~ minor axis has length $\frac{C}{\sqrt{\lambda_2}}$

~~minor~~ major axis has length $\frac{C}{\sqrt{\lambda_1}}$

In our example, then, we see the ellipse ~~about~~ rotated about on ~~the~~ to the principal (major and minor) axes ~~as~~



Expression
for
Rotation
onto
 \tilde{x}_1, \tilde{x}_2
axes.

$$\left[\text{for } C^2 = 1 = \begin{bmatrix} x'e_1 & x'e_2 \end{bmatrix} \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{bmatrix} x'e_1 \\ x'e_2 \end{bmatrix} \right. \\ \left. = (\tilde{x}_1 \ \tilde{x}_2) \Lambda \begin{pmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{pmatrix} = \tilde{X}^T \Lambda \tilde{X} \right]$$