

$\phi(x_1, x_2) \rightarrow$  fm of 2 variables.

Searching for minima for convenience const term = 0  
At a critical pt  $\phi(0, 0) = 0$ . — (1)

$$\hookrightarrow \frac{\partial \phi}{\partial x_1} = 0 = \frac{\partial \phi}{\partial x_2} \quad \text{--- (2)}$$

$$\Rightarrow \phi(x_1, x_2) = \frac{1}{2} \left[ \frac{\partial^2 \phi}{\partial x_1^2} x_1^2 + 2 \frac{\partial^2 \phi}{\partial x_1 \partial x_2} x_1 x_2 + \frac{\partial^2 \phi}{\partial x_2^2} x_2^2 \right]$$

$$\rightarrow = \frac{1}{2} \left[ a x_1^2 + 2b x_1 x_2 + c x_2^2 \right]$$

Is  $(0, 0)$  a minima?

called a quadratic form.

Must always be  $> 0$ , i.e.  $\phi(x_1, x_2) > 0 \quad \forall x$

then  $\phi(x_1, x_2)$  is called POSITIVE DEFINITE.

$$\phi(x_1, x_2) > 0 \quad \rightarrow \quad \text{POS DEF}$$

$\geq 0 \rightarrow$  P&S SEMI DEF

$< 0 \rightarrow$  NEG DEF

$\leq \rightarrow$  NEG SEMI DEF

$> 0, \neq 0 \rightarrow$  INDEFINITE. (SADDLE)

$$ax_1^2 + 2bx_1x_2 + cx_2^2 = a\left(x_1 + \frac{b}{a}x_2\right)^2 + \left(c - \frac{b^2}{a}\right)x_2^2$$

In order for  $\downarrow > 0$

$$\Rightarrow a > 0, c > b^2/a \rightarrow ac > b^2$$

$$\left[ \begin{array}{l} \frac{\partial^2 \phi}{\partial x_1^2} > 0 \\ \frac{\partial^2 \phi}{\partial x_2^2} \frac{\partial^2 \phi}{\partial x_1^2} > \left( \frac{\partial^2 \phi}{\partial x_1 \partial x_2} \right)^2 \end{array} \right]$$

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\|y\|^2 = y^T y$$
$$\|y - Ax\|_2^2$$

$A$  always a symmetric matrix.

Defn: A sym matrix  $A$  is POS DEF IF  $x^T A x > 0 \quad \forall x \neq (0,0)$

$A$  pure quadratic form

↪ Different tests for PD matrices.

$$A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

$$a > 0, \quad ac > b^2$$

1)

$$|A| = ac - b^2 > 0$$

$$\lambda_1, \lambda_2 > 0$$

$$\text{tr}(A) = \sum \lambda_i$$

$$a + c = \lambda_1 + \lambda_2$$

$$\therefore a > 0, \Rightarrow c > 0$$

$$\begin{aligned} &= (y - Ax)^T (y - Ax) \\ &= \underbrace{y^T y} - y^T A x - x^T A^T y + \underbrace{x^T A^T A x} \end{aligned}$$

$$\therefore \lambda_1, \lambda_2 > 0$$

$$2) \begin{bmatrix} a & b \\ b & c \end{bmatrix} \rightarrow \begin{bmatrix} a & b \\ 0 & c - b^2/a \end{bmatrix}$$

Pivots are +ve!

$$R_2 - \frac{b}{a} \times R_1$$

→ Eigenvalues.  $Ax = \lambda x$ .  $x^T$  l. mult

$$x^T Ax = \lambda x^T x = \lambda \|x\|^2$$

• if  $A$  is PD  $\Rightarrow x^T Ax > 0 \Rightarrow \lambda > 0$

If  $\lambda$ 's  $> 0 \stackrel{?}{\Rightarrow} A$  is PD.

Spectral thm  $\Rightarrow A$  must have  $n$  orthonormal eig vectors.

Take any  $x \in \mathbb{R}^n$   $\{x_i\}$

$$\Rightarrow \underline{x} = \sum_{i=1}^n \underline{c}_i \underline{x}_i \quad (\text{defn of basis})$$

$$\underline{x}^T \curvearrowright \underline{Ax} = \sum_{i=1}^n c_i \underline{\lambda}_i \underline{x}_i \curvearrowright \underline{x}^T$$

$$\underline{x}^T \underline{Ax} = \left( \sum_{i=1}^n c_i \underline{x}_i^T \right) \left( \sum_{j=1}^n c_j \lambda_j \underline{x}_j \right)$$

$$= \sum_{i=1}^n c_i^2 \lambda_i \underbrace{\|\underline{x}_i\|^2}_{=1} > 0 \text{ when } \lambda_i > 0$$

$\Rightarrow$  If  $\lambda$ 's are +ve,  $A$  is PD.

A real sym matrix  $A$  is PD iff eigvals are +ve.

$\hookrightarrow$  Pivot test. (Recall  $A$  is sym).

$$A = LU = LDU = LDL^T$$

have 1s on diag.

$$\begin{aligned} x^T A x &= \underbrace{x^T}_y L D \underbrace{L^T}_y x = y^T D y \\ &= [y_1 \dots y_n]^T \begin{bmatrix} D_{11} & & 0 \\ & \ddots & \\ 0 & & D_{nn} \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \\ &= \sum y_i^2 D_{ii} \end{aligned}$$

If  $A$  is PD  $\Rightarrow x^T A x > 0 \Rightarrow D_{ii} > 0$

If  $D_{ii} > 0 \Rightarrow x^T A x > 0 \therefore$  PD

A real sym matrix  $A$  is PD iff the pivots are +ve.

$\hookrightarrow$  Test 3  $\left[ \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \\ \vdots \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \vdots \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \vdots \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{array} \right] A_k$  All upper left sub matrices have  $|A_k| > 0$

Test 4 →. A sym matrix  $A$  is PD iff there is a matrix  $R$  with indepn cols s.t.  $A = R^T R$ .

1) Assume  $A = R^T R$ .  $\stackrel{?}{\Rightarrow}$   $A$  is PD ✓

$$x^T A x = x^T R^T R x = \|R x\|_2^2 \geq 0$$

But  $R$  has indepn cols

$$\therefore \|R x\|_2^2 > 0$$

2) Assume  $A$  is PD  $\stackrel{?}{\Rightarrow}$   $A = R^T R$

$$A = L D L^T \quad \text{diag}(D) = D_{ii} > 0 \quad (\text{Test 2})$$

$$\begin{matrix} \downarrow \\ \sqrt{D} \end{matrix} \sqrt{D} = \underbrace{L \sqrt{D}}_{R^T} \underbrace{\sqrt{D} L^T}_R$$

$$R = \sqrt{D} L^{-1} = \begin{bmatrix} \sqrt{d_{11}} & & & \\ & \sqrt{d_{22}} & & \\ & & \ddots & \\ & & & \sqrt{d_m} \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 0 & 1 & \\ & & & \ddots \\ & & & & 1 \\ & & & & & \ddots \\ & & & & & & 1 \end{bmatrix}$$

$$= \begin{bmatrix} \sqrt{d_{11}} & \sqrt{d_{11}} e & \sqrt{d_{11}} f \\ 0 & \sqrt{d_{22}} c & \sqrt{d_{22}} d \\ 0 & 0 & \sqrt{d_{33}} \\ 0 & 0 & 0 \end{bmatrix}$$

$\Rightarrow R$  has lin indep col.

PSD,

1) Eigenvalues  $\lambda_i \geq 0$

2) Pivots  $D_{ii} \geq 0$

3) Submatrices  $|A_k| \geq 0$

4)  $A = Q \Lambda Q^T$   $\forall$   $Q$   $\perp$   $\|Q\| = 1$



4)  $A$  is  $K \cdot K$  with possibly depn cols.

—  $x$  —

Simplest case  $\begin{bmatrix} a & b \\ b & c \end{bmatrix}$  set  $b = 0$

$$x^T A x = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a & 0 \\ 0 & c \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 1$$

$$a x_1^2 + c x_2^2 = 1.$$

$\therefore$  Ellipse in general if  $A$  is PD