

2 Positive definite matrices

Let $A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$. Then A is positive definite if one of the following (equivalent) conditions hold.

① $\lambda_1 > 0, \lambda_2 > 0$

eigenvalues

② subdets: $a > 0, ac - b^2 > 0$

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

determinants

usually the def'n

③ pivots: $a > 0, \frac{ac - b^2}{a} > 0$

pivots

↳ * ④ $x^T A x > 0$ for $x \neq 0$.

quadratic forms

It should be clear what the analogue is for $n \times n$.

Ex: $\begin{bmatrix} 2 & 6 \\ 6 & 7 \end{bmatrix}$ is positive definite (pos. semi-definite if $d=18$)

$\begin{bmatrix} 2 & 6 \\ 6 & 7 \end{bmatrix}$ not pos. def'n:

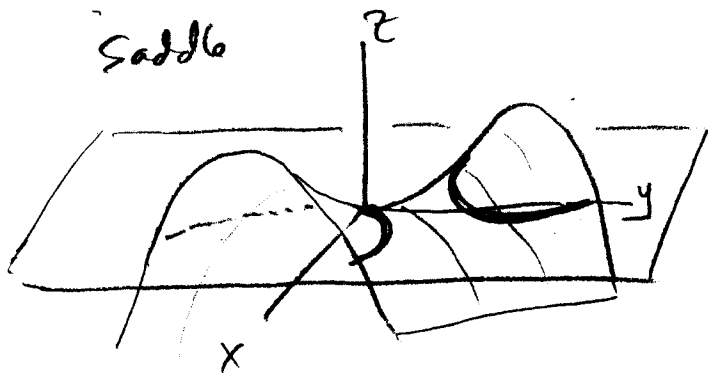
$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 2 & 6 \\ 6 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{matrix} a & 2b & c \\ \downarrow & \downarrow & \downarrow \\ 2x_1^2 & +12x_1x_2 & +7x_2^2 \end{matrix} \neq 0$$

Take $x_1 = 1, x_2 = -1 \rightsquigarrow f(x_1, x_2) = -3$

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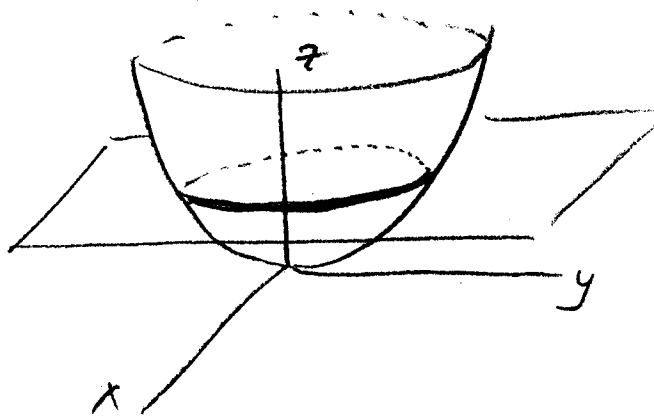
Graph of $f(x, y) = \vec{x}^T A \vec{x} = ax^2 + 2bxy + cy^2$.

Saddle



cross-section: hyperbola

vs.



ellipse

Recall calculus.

min of $f(x)$:

$$f'(x) = 0, \quad f''(x) > 0$$

min of $f(x, y)$

$$Df(x, y) = 0, \quad H(f) = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix} \text{ is}$$

positive definite.

$$\text{e.g., } \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix} = 2 \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

Ex: $f(x, y) = 2x^2 + 12xy + 20y^2$

$$= 2(x+3y)^2 + 2y^2 > 0 \quad \text{for } (x, y) \neq (0, 0)$$

come from elimination.

$$A = \begin{bmatrix} 2 & 6 \\ 6 & 20 \end{bmatrix} = \underset{L}{\begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix}} \underset{u}{\begin{bmatrix} 2 & 6 \\ 0 & 2 \end{bmatrix}}$$

pivot = 2, 2
multiplier = 3

positive pivots

positive squares

positive definite

More properties (think analogues of "positivity" for numbers)

• A pos. def. $\Rightarrow A^{-1}$ pos. def.

Why: Eigenvalues of A^{-1} are $1/\lambda$, where $Av = \lambda v$.

• A, B pos. def. $\Rightarrow A+B$ pos. def.

Why: $x^T(A+B)x = x^T Ax + x^T Bx > 0$ if $x \neq 0$.

• M is $m \times n$, rank n (lin. indep. cols.) $\Rightarrow M^T M$ pos. def.

Why: $x^T(M^T M)x = x^T M^T M x = (Mx)^T (Mx) = \|Mx\|^2 > 0$
if $x \neq 0$.

Note: If rank $M < n$, then $M^T M$ pos. semi-def.

Application: (Conic sections)

let $f(x_1, x_2) = 5x_1 + 8x_1x_2 + 5x_2^2 = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 5 & 4 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

$\lambda_1 = 9 \quad v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$\lambda_2 = 1 \quad v_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

$A = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 9 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$

45° rotation matrix $\rightarrow \begin{matrix} Q & \Lambda & Q^T \end{matrix}$

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The equation $x^T A x = 1$ is a conic section (ellipse).

$$x^T A x = x^T Q \Lambda Q^T x = 1$$

$$(Q^T x)^T \Lambda (Q^T x) = 1$$

$$y^T \Lambda y = 1 \quad \text{where } y = Q^T x \Rightarrow x = Q y.$$

