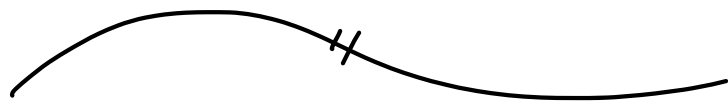


Problem 5 on HW 5 will be optional. HW 6 will replace Exam 3.



Applications:

Principal Axes Theorem (old)

Descartes, Fermat (1600s)

Euler (1700)

High School:

$$f(x, y) = a + bx + cy + dx^2 + exy + fy^2.$$

Equation  $f(x, y) = 0$  can be reduced to standard form by a translation & a rotation.

Standard forms:

parabola :  $y = ax^2$  or  $x = ay^2$

ellipse :  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$

hyperbola :  $\pm \left( \frac{x^2}{a^2} - \frac{y^2}{b^2} \right) = 1.$

Principal Axes Theorem:

$$F(x_1, \dots, x_n) = b + \sum b_i x_i + \sum b_{ij} x_i x_j$$
$$= b + \vec{b}^T \vec{x} + \vec{x}^T B \vec{x}$$

where  $B^T = B$ . If  $B^{-1}$  exists  
then we can find change of  
coords  $\vec{u} = Q \vec{x} + \vec{c}$  where

$$Q^T Q = I \quad (\text{orthogonal})$$

(generalized rotation)

such that

$$F(\vec{u}) = F(u_1, u_2, \dots, u_n)$$
$$= c + \lambda_1 u_1^2 + \lambda_2 u_2^2 + \dots + \lambda_n u_n^2$$

Proof: Since  $B^T = B$ , Spec.

Thm.  $\Rightarrow \exists Q^T Q = I$  so

$$B = Q \Lambda Q^T$$

with  $\Lambda = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$ .

Consider  $\vec{u} = Q\vec{x} + \vec{t}$  for some vector  $\vec{t}$ , yet to be chosen. Then

$$\begin{aligned} f(\vec{u}) &= F(Q\vec{x} + \vec{t}) \\ &= b + \vec{b}^T(Q\vec{x} + \vec{t}) + (Q\vec{x} + \vec{t})^T B (Q\vec{x} + \vec{t}) \\ &= b + \vec{b}^T Q\vec{x} + \vec{b}^T \vec{t} + (\vec{x}^T Q^T + \vec{t}^T) B (Q\vec{x} + \vec{t}) \end{aligned}$$

$$\begin{aligned} &\vec{x}^T Q^T B Q\vec{x} + \vec{x}^T Q^T B \vec{t} \\ &+ \vec{t}^T B Q\vec{x} + \vec{t}^T B \vec{t} \end{aligned}$$

[  $B^T = B$  &  $\vec{x}^T Q^T B \vec{t}$  scalar, so

$$\begin{aligned} \vec{x}^T Q^T B \vec{t} &= (\vec{x}^T Q^T B \vec{t})^T \\ &= \vec{t}^T B^T Q \vec{x} \\ &= \vec{t}^T B Q \vec{x}. \end{aligned}$$

$$= b + \vec{b}^T \vec{t} + \vec{t}^T B \vec{t} + \vec{b}^T Q \vec{x} \\ + 2 \vec{t}^T B Q \vec{x} + \vec{x}^T \cancel{Q^T B Q} \vec{x}.$$



$$= (b + \vec{b}^T \vec{t} + \vec{t}^T B \vec{t}) \\ + (\vec{b}^T Q + 2 \vec{t}^T B Q) \vec{x} + \vec{x}^T \Lambda \vec{x}.$$

$= 0.$

$$\vec{b}^T Q + 2 \vec{t}^T B Q = \vec{0}^T$$

$$2 \vec{t}^T B Q = -\vec{b}^T Q$$

assume  
exists

$$\vec{t}^T B Q = -\frac{1}{2} \vec{b}^T Q$$

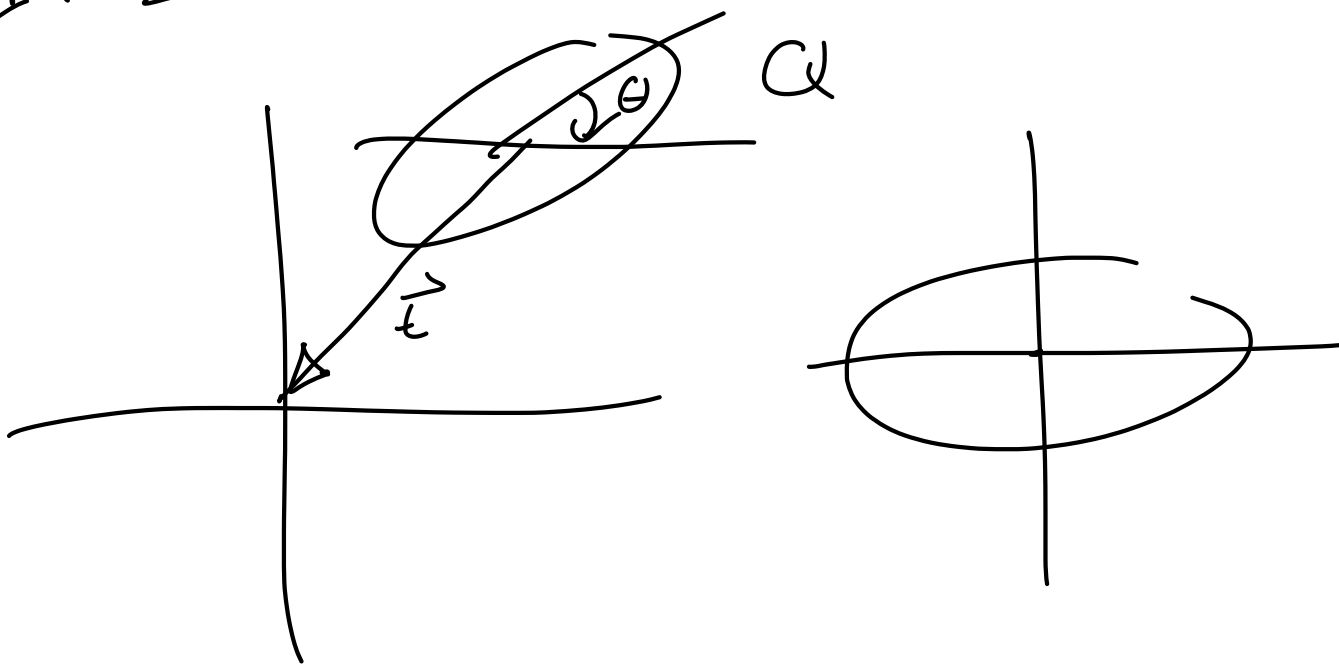
$$\vec{t} = -\frac{1}{2} \vec{b}^T Q Q^{-1} B^{-1}$$

$$\vec{t} = -\frac{1}{2} \vec{b}^T B^{-1}$$

Finally (oops):

$$f(u_1, \dots, u_n) = c + \lambda_1 x_1^2 + \dots + \lambda_n x_n^2$$

In 2D:



Apply to  $F(x, y) = 2x^2 + 2xy + y^2 + x + 1$ .

$$= 1 + (1 \ 0) \begin{pmatrix} x \\ y \end{pmatrix} + (x \ y) \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}.$$

$$R_{\theta}^{-1} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} R_{\theta} = \begin{pmatrix} \lambda & \\ & \mu \end{pmatrix}.$$

"

$$\begin{aligned} c &= \cos \theta \\ s &= \sin \theta \end{aligned}$$

$$\begin{pmatrix} 2c^2 + 2sc + 1s^2 & \dots \\ (1-2)sc + 1(c^2 - s^2) & \dots \end{pmatrix}.$$

$$\text{Want } (1-2)sc + (c^2 - s^2) = 0.$$

$$\Rightarrow \tan \theta = 1/1 = 1$$

$$\theta = \frac{\pi}{4}.$$

Formula: To diagonalize symmetric

$$A = \begin{pmatrix} a & b/2 \\ b/2 & c \end{pmatrix}$$

Rotate by  $\theta$  where

$$\tan \theta = b / (a - c).$$

Translation ... I.O.U.

3D Version (Euler's Rotation Theorem): You can still do it with a rotation.

Theorem: Any  $3 \times 3$   $Q^T Q = I$   
can be written as

$$Q = X \begin{pmatrix} \pm 1 & 0 & 0 \\ 0 & \cos\theta & -\sin\theta \\ 0 & \sin\theta & \cos\theta \end{pmatrix} X^{-1}$$

IF  $\det(Q) = -1$  pick  $-1$

$\det(Q) = 1$  pick  $+1$

Implies any quadratic

$$F(x, y, z)$$

Can be diagonalized by a rotation.



Positive Definite Matrices.

We previously showed,

for  $B^T = B$ :

$$\vec{x}^T B \vec{x} \geq 0 \quad \forall \vec{x} \iff B = A^T A$$

IF  $A$  has ind cols then

$$\vec{x}^T B \vec{x} = 0 \iff \vec{x} = 0.$$

But it's really hard to show  
 $B$  semidef  $\implies B = A^T A$ . It's  
like finding the "square root"  
of a matrix.

Spectral Theorem:

$$B^T = B \implies B = Q \Lambda Q^T$$

for some  $Q^T Q = I$ .

Furthermore  $B$  pos semidef

$\implies$  e.values  $\lambda \geq 0$ .

[We know  $B^T = B \implies$  real e.values.]

I.o.u.  $\vec{x}^T B \vec{x} \geq 0 \forall \vec{x} \implies \geq 0$  e.values]

$$B = Q \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} Q^T$$



TRICK:  $\Lambda$  is diagonal with entries  $\geq 0$  so it makes sense to take the "square root":

$$\sqrt{\Lambda} \stackrel{\text{DEF}}{=} \begin{pmatrix} \sqrt{\lambda_1} & & \\ & \ddots & \\ & & \sqrt{\lambda_n} \end{pmatrix}$$

Check:  $\sqrt{\Lambda} \cdot \sqrt{\Lambda} = \Lambda$

In this special case we can lift this to a "square root" of  $B$ . Take

$$A^T = Q \sqrt{\Lambda}$$

$$A = \sqrt{\Lambda}^T Q^T = \sqrt{\Lambda} Q^T$$

Then

$$A^T A = Q \sqrt{\Lambda} \sqrt{\Lambda} Q$$

$$= Q \Lambda Q^T = B \quad \checkmark$$

Not unique, but this is an algorithm to factor semi-def matrix

$$B \text{ semidef} \implies B = A^T A.$$