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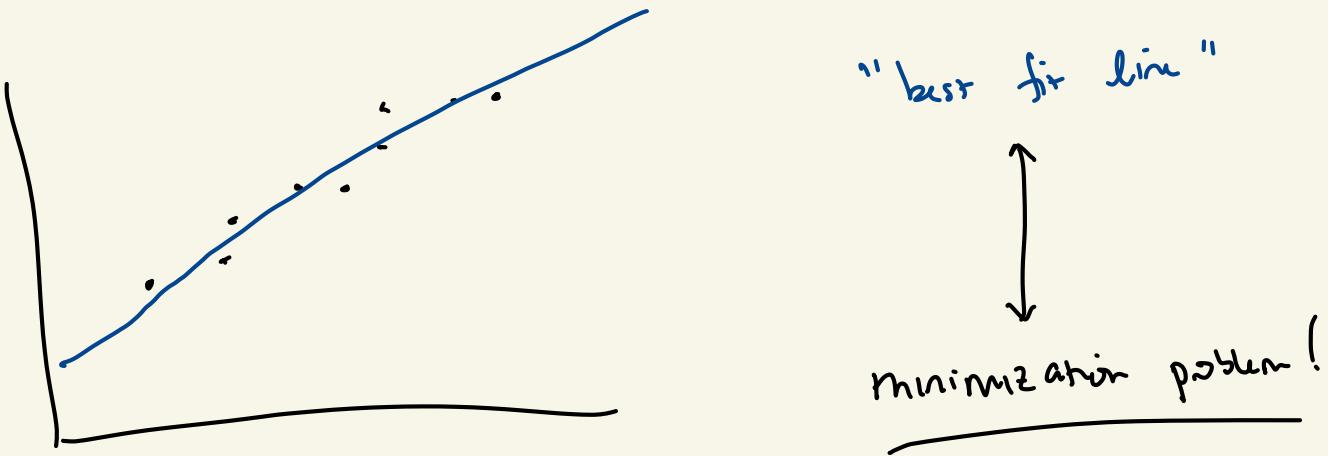
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HW 8 available

Ch 5 : Linear algebra approach to linear regression

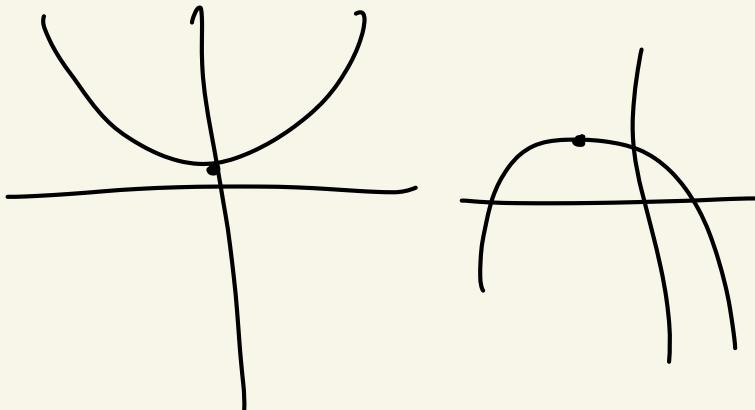


## Section 5.2

Minimization of quadratics (degree 2 polynomial in many variables)

1 variable

$$p(x) = ax^2 + bx + c$$

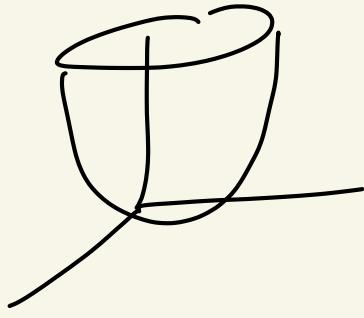


$a > 0$  min

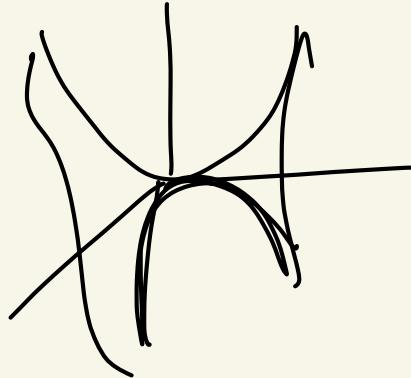
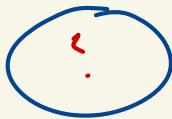
$a < 0$  max

$$p(x,y) = \underbrace{ax^2 + bx + cy^2}_{\text{deg 2}} + \underbrace{dx + ey}_{\text{deg 1}} + f$$

constant

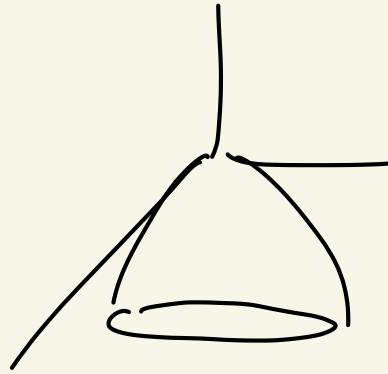


facing up



pringle shape  
"saddle"

? no min



facing down

?  
no min

$$p(x,y) = \underbrace{ax^2 + bxy + cy^2}_{\text{quadratic form}} + dx + ey + f$$

quadratic form

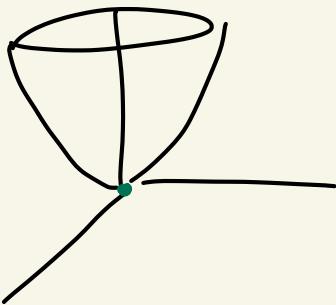
$$x^T K x + x^T \vec{f} + c \quad (\text{constant})$$

$$= (x \ y) \begin{pmatrix} a & \frac{b}{2} \\ \frac{b}{2} & c \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} + (x \ y) \begin{pmatrix} d \\ e \end{pmatrix} + c$$

let  $K = \begin{pmatrix} a & \frac{b}{2} \\ \frac{b}{2} & c \end{pmatrix}$ .

If  $K$  is pos def then  $p(x, y)$  faces up

and  $p$  has a minimum value!



So if  $K > 0$  have a minimization problem! We'll stick to  $K$  being positive definite.

Problem: Given a quadratic w/ n variables

$$p(x_1, x_2, \dots, x_n) = \sum_{i,j=1}^n a_{ij} x_i x_j + \sum_{i=1}^n b_i x_i + c.$$

$$x_i x_j = x_j x_i$$

so I'm not adding this term twice

When does this have a min value? If so  
for  $\vec{x}$  does it occur? what's the actual minimum  
value?

$$\text{First: } p(\vec{x}) = \sum a_{ij}x_i x_j + \sum b_i x_i + c \quad \vec{x} = (x_1, \dots, x_n)$$

$\underbrace{x^T K x}_{\text{}}_{\text{}}$ 
 $x^T b$ 
+ c

$- 2 x^T f$ 
+ c

$$K = \begin{pmatrix} a_{11} & & a_{1j} \\ & a_{22} & \vdots \\ \frac{a_{ij}}{2} & \ddots & a_{nn} \end{pmatrix}$$

$$f = \begin{pmatrix} -\frac{b_1}{2} \\ -\frac{b_2}{2} \\ \vdots \\ -\frac{b_n}{2} \end{pmatrix}$$

$$p(\vec{x}) = (x_1, \dots, x_n) \begin{pmatrix} a_{11} & & a_{1j} \\ \frac{a_{ij}}{2} & \ddots & \\ & & a_{nn} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} - 2 (x_1, \dots, x_n) \begin{pmatrix} -\frac{b_1}{2} \\ \vdots \\ -\frac{b_n}{2} \end{pmatrix} + c$$

$$p(\bar{x}) = \underline{\underline{x^T K x - 2x^T f + c}}$$

the form of  
the quadratic  
that we'll use.

Claim:  $p(\bar{x})$  has a min only when  $K$  is pos def.

The min occurs at  $\underline{\underline{x^* = K^{-1}f}}$ , and the min value is  $\underline{\underline{p(x^*) = c - f^T x^*}}$ .

Pf: Assume  $K > 0$ . If  $\underline{\underline{x^* = K^{-1}f}} \Rightarrow \underline{\underline{f = Kx^*}}$ .

$$\begin{aligned} p(\bar{x}) &= \cancel{x^T K x} - 2x^T (\cancel{Kx^*}) + c \\ &= (\bar{x} - x^*)^T K (\bar{x} - x^*) + (c - x^{*T} K x^*) \end{aligned}$$

Complete the square

$$\begin{aligned}
 &= \left( x^T K x - 2x^T (Kx^*) + x^{*T} K x^* \right) - \left( x^{*T} K x^* + c \right) \\
 &= (x - x^*)^T K (x - x^*) + (c - f^T x^*)
 \end{aligned}$$

$p(\vec{x}) = \underbrace{(x - x^*)^T K (x - x^*)}_{K > 0 \text{ so } g(\vec{y}) = \vec{y}^T K \vec{y}} + (c - f^T x^*) \xrightarrow{\text{constant!}}$

$g(\vec{y}) = \vec{y}^T K \vec{y}$  has a min at  $\vec{y} = 0$   
 i.e.  $g(\vec{y}) = 0$  iff  $\vec{y} = 0$ !

$p(\vec{x})$  has a minimum only at  $x - x^* = 0$   
 $\Rightarrow x^*$  is the minimizer.

When  $x = x^*$ , the  $*$  = 0  $\Rightarrow p(x^*) = c - f^T x^*$ . c - f^T x^\* □

Ex!

$$p(x, y, z) = \underbrace{1x^2 + 2xy + 1xz}_{+ 0x} + \underbrace{-2y^2 + 1yz + 2z^2}_{+ 0x} + \underbrace{6y - 7z + 5}_{+ 0x}$$

$$\textcircled{1} \quad p(x, y, z) = x^T K x - 2x^T f + c. \quad K = ? \quad f = ?$$

$$K = \begin{pmatrix} x & y & z \\ x & 1 & \frac{1}{2} \\ y & 1 & 2 \\ z & \frac{1}{2} & 2 \end{pmatrix} \quad f = \begin{pmatrix} 0 \\ -3 \\ 7/2 \end{pmatrix} \quad b = \begin{pmatrix} 0 \\ 6 \\ -7 \end{pmatrix}$$

$K$  is symmetric by construction

$$K^T = K$$

$$K^{-1} ??$$

$$p(\vec{x}) = (x \ y \ z) \begin{pmatrix} 1 & 1 & \frac{1}{2} \\ 1 & 2 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 2 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} - 2(x \ y \ z) \begin{pmatrix} 0 \\ -3 \\ \frac{7}{2} \end{pmatrix} + 5$$

$K$  is pos def! So  $\text{wt}$  has a min

$$x^* = K^{-1}f$$

$$= \begin{pmatrix} 1 & 1 & \frac{1}{2} \\ 1 & 2 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 2 \end{pmatrix}^{-1} \begin{pmatrix} 0 \\ -3 \\ \frac{7}{2} \end{pmatrix} = \begin{pmatrix} 2 \\ -3 \\ 2 \end{pmatrix}.$$

↖ row reduction

Note: All problems on HW involve pos definite matrices.

$p(x, y, z)$  has the lowest value at  $(2, -3, 2)$ .

$$p(2, -3, 2) = 5 - f^T x^* = 5 - (0 \ -3 \ \frac{7}{2}) \begin{pmatrix} 2 \\ -3 \\ 2 \end{pmatrix}$$

$$= -11$$

$$\underline{\text{Ex}} \quad p(x,y) = 4x^2 - 2xy + 3y^2 + \boxed{3x} \boxed{-2y} + 1$$

$$= (x \ y) \begin{pmatrix} 4 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} - 2(x \ y) \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix} + 1$$

The min occurs at  $x^* = K^{-1}f = \begin{pmatrix} 4 & -1 \\ -1 & 3 \end{pmatrix}^{-1} \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix}$

$$= \frac{1}{11} \begin{pmatrix} 3 & 1 \\ 1 & 4 \end{pmatrix} \begin{pmatrix} -\frac{3}{2} \\ 1 \end{pmatrix} = \begin{pmatrix} -\frac{2}{11} \\ \frac{5}{11} \end{pmatrix}$$

In Calc III, multi:

$$\begin{pmatrix} \frac{\partial P}{\partial x} \\ \frac{\partial P}{\partial y} \end{pmatrix} = \begin{pmatrix} 8x - 2y + 3 \\ 6y - 2x - 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$8x - 2y = -3 \quad \Rightarrow \quad \begin{pmatrix} 8 & -2 \\ -2 & 6 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -3 \\ 2 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 4 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 4 & -1 \\ -1 & 3 \end{pmatrix}^{-1} \begin{pmatrix} -3 \\ 1 \end{pmatrix} \quad !!$$

$$\frac{\partial}{\partial x} \left( 4x^2 - 2xy + \underline{3y^2} + \underline{3x} \cancel{- 2y} + x \right)$$

$$= 8x - 2y + 3$$

Do it this way!