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Plan


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- 1 Stationary points
  - 2 Second derivative test
  - 3 Global maximum and minimum
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Optimization:  
max/min  $f(x,y)$

Note:

maximum = global maximum  
minimum = global minimum

Defn:

$(x^*, y^*)$  is a maximal pt / maximizer for  $f$   
if  $f(x^*, y^*) \geq f(x,y)$  for all  $(x,y)$  in  $D_f$

$(x^*, y^*)$  is a local max. for  $f$  if  
 $f(x^*, y^*) \geq f(x,y)$  for all  $(x,y)$  close to  $(x^*, y^*)$

$(x^*, y^*)$  is a minimum pt / minimizer for  $f$   
if  $f(x^*, y^*) \leq f(x,y)$  for all  $(x,y)$  in  $D_f$

$(x^*, y^*)$  is a local min. for  $f$  if  
 $f(x^*, y^*) \leq f(x,y)$  for all  $(x,y)$  close to  $(x^*, y^*)$

A stationary pt  $(x^*, y^*)$  of  $f$  that is neither  
local max nor local min is called  
a saddle point

Key result:

If  $(x^*, y^*)$  is a max/min for  $f$ ,  
then we have either

- i)  $(x^*, y^*)$  is a stationary pt for  $f$  ( $f'_x = f'_y = 0$  at  $(x^*, y^*)$ )
- ii) either  $f'_x$  or  $f'_y$  is not defined  
at  $(x^*, y^*)$
- iii)  $(x^*, y^*)$  is a boundary pt for  $D_f$ .

Names: i) or ii) are called critical points

candidate pts: i), ii) or iii)

Note: If  $f$  is "nice" (for example a polynomial), then only pts of type i) exist

Ex.  $f(x,y) = x^3 + 3xy + y^3$

Cand pts = stationary pts

Stationary pts:

$$\left. \begin{aligned} f'_x &= 3x^2 + 3y = 0 \\ f'_y &= 3x + 3y^2 = 0 \end{aligned} \right\} \text{first order conditions (FOC)}$$

$$\begin{aligned} x^2 + y &= 0 \Rightarrow y = -x^2 \\ x + y^2 &= 0 \end{aligned}$$

$$x + (-x^2)^2 = 0$$

$$x + x^4 = 0$$

$$x(1 + x^3) = 0$$

$$\underline{x=0} \quad \text{or} \quad x^3 = -1$$

$$\underline{y=0} \quad x = \sqrt[3]{-1} = -1$$

$$\underline{y=-1}$$

Stationary pts = Cand. pts

$$(x,y) = (0,0), (-1,-1)$$

$$f(0,0) = 0$$

cond. for min

$$f(-1,-1) = 1$$

cond. for max

## ② Second derivative test:

Second derivative test

If  $(x^*, y^*)$  is a stationary pt. of  $f$ , then we compute

$$H(f)(x^*, y^*) = \begin{pmatrix} f''_{xx}(x^*, y^*) & f''_{xy}(x^*, y^*) \\ f''_{xy}(x^*, y^*) & f''_{yy}(x^*, y^*) \end{pmatrix} = \begin{pmatrix} A & B \\ B & C \end{pmatrix}$$

We have that:

(1) If  $\det H(f)(x^*, y^*) > 0$  and  $\text{tr} H(f)(x^*, y^*) > 0$  then  $(x^*, y^*)$  is local min

(2) If  $\det H(f)(x^*, y^*) > 0$  and  $\text{tr} H(f)(x^*, y^*) < 0$  then  $(x^*, y^*)$  is local max

(3) If  $\det H(f)(x^*, y^*) < 0$ , then  $(x^*, y^*)$  is a saddle point.

Note: If  $\det H(f)(x^*, y^*) = 0$ , then the test is inconclusive.

Ex: (continued)  $f(x,y) = x^3 + 3xy + y^3$

$$f'_x = 3x^2 + 3y$$

$$f'_y = 3x + 3y^2$$

Crit. pts:

$$(x,y) = (0,0), (-1,-1)$$

$$f=0 \quad f=1$$

We classify the stationary pts using the second derivative test:

= determine (for each stationary pt) if it is local max, local min, Saddle pt.

$$H(f) = \begin{pmatrix} f''_{xx} & f''_{xy} \\ f''_{xy} & f''_{yy} \end{pmatrix} = \begin{pmatrix} 6x & 3 \\ 3 & 6y \end{pmatrix}$$

For (0,0):

$$H(f)(0,0) = \begin{pmatrix} 0 & 3 \\ 3 & 0 \end{pmatrix} = \begin{pmatrix} A & B \\ B & C \end{pmatrix}$$

$$\det = \det H(f)(0,0) = 0 - 9 = -9 < 0$$

$\Downarrow$   
(0,0) is a saddle pt for  $f$

For (-1,-1):

$$H(f)(-1,-1) = \begin{pmatrix} -6 & 3 \\ 3 & -6 \end{pmatrix}$$

AC-B<sup>2</sup>

$$\det = 36 - 9 = 27 > 0$$

$$\text{tr} = \text{tr} H(f)(-1,-1) = -6 + (-6) = -12 < 0$$

A+C

$\Downarrow$   
(-1,-1) is a local max for  $f$

### ③ Global max/min

Conclusion:

$f(x,y) = x^3 + 3xy + y^3$  has no maximum

— 11 — has local maximum  $f(-1,-1) = 1$

Ad hoc:

$$f(10,10) = 10^3 + 3 \cdot 10 \cdot 10 + 10^3 = 1000 + 300 + 1000 = 2300 > 1$$

$\Rightarrow$  (-1,-1) is no max for  $f$

$\Rightarrow$   $f$  has no maximum

Note: If  $(x^*, y^*)$  is stationary pt with

$$H(f)(x^*, y^*) = \begin{pmatrix} A & B \\ B & C \end{pmatrix}$$

then:

① If  $AC - B^2 > 0$ , then  $AC > B^2 \geq 0$  hence  $AC > 0$

$$\Rightarrow A, C > 0 \Leftrightarrow A + C > 0$$

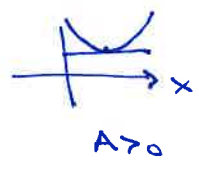
or

$$A, C < 0 \Leftrightarrow A + C < 0$$

In short, the possible cases are  $\begin{cases} \text{i) } A, C > 0 \\ \text{ii) } A, C < 0 \end{cases}$

cuts of the graph  $Z = f(x, y)$

$A, C > 0$ :



$A > 0$



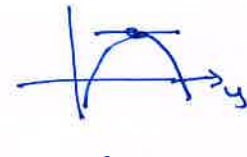
$C > 0$

local min

$A, C < 0$ :



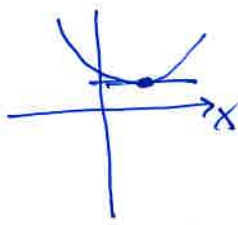
$A < 0$



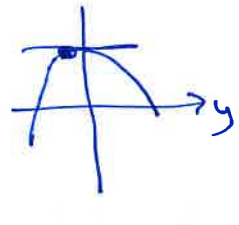
$C < 0$

local max

② If  $AC - B^2 < 0$ , a typical case is  $A > 0, C < 0$



$A > 0$



$C < 0$

Saddle pt

Ex:  $f(x,y) = \sqrt{x^2+y^2}$ ,  $D_f = \mathbb{R}^2 \Rightarrow$  no boundary pts of  $D_f$

$$f'_x = (\sqrt{u})'_x = (u^{1/2})'_x$$

$$u = x^2 + y^2$$

$$= \frac{1}{2} u^{-1/2} \cdot u'_x = \frac{1}{2\sqrt{u}} \cdot 2x$$



Stat. pts:

$$f'_x = \frac{2x}{2\sqrt{x^2+y^2}} = \frac{x}{\sqrt{x^2+y^2}} = 0 \Rightarrow x=0$$

$$f'_y = \frac{1}{2\sqrt{u}} \cdot u'_y = \frac{y}{\sqrt{x^2+y^2}} = 0 \Rightarrow y=0$$

But  $f'_x(0,0)$ ,  $f'_y(0,0)$  gives division by 0  
 $\Rightarrow (x^*, y^*) = (0,0)$  is not stationary pt.  $\Rightarrow$

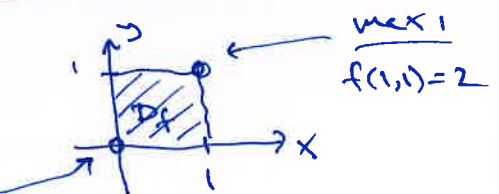
no stat. pts. for  $f$ .

$f'_x, f'_y$  not defined for  $(0,0) \Rightarrow (0,0)$  is a critical pt

Note that  $f(0,0) = 0$  is a minimum pt for  $f$ .

Ex:  $f(x,y) = x+y$ ,  $0 \leq x, y \leq 1$

min:  
 $f(0,0) = 0$



Boundary pts of  $D_f =$   
the sides of the square

Ex:  $f(x,y) = x^2 y^3 + y^2 - 2y$

max/min  $f(x,y) = ?$

Candidate pts = maximum pts:

Foc:  $f'_x = 2xy^3 = 0 \implies x=0$

$f'_y = x^2 \cdot 3y^2 + 2y - 2 = 0$

$\implies x=0$

or  $y=0$

$0 + 2y - 2 = 0$

$2y = 2$

$y = 1$

$0 + 0 - 2 = 0$

impossible

$(x^*, y^*) = (0, 1)$

Stat. pts:  $(x^*, y^*) = (0, 1)$   
 $f = -1$

Classification:

$H(f) = \begin{pmatrix} 2y^3 & 6xy^2 \\ 6xy^2 & 6x^2y + 2 \end{pmatrix}$

For (0,1):  $H(f)(0,1) = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$

$\det = 4 > 0$

$\text{tr} = 4 > 0$

$\Downarrow$

$f(0,1) = -1$  is local min

Ad hoc:  $x=1$

$f(1,y) = y^3 + y^2 - 2y$

$f(1,-10) = -1000 + 100 + 20$

$= -880$

$\Downarrow$

$f$  has no minimum  
(or maximum)

$y \rightarrow \infty \implies f(1,y) \rightarrow \infty$

Note: Example of fn.  $f(x,y)$  with only one stat. pt, which is a local min, but still not global min.

Linear approximation of  $f(x,y)$  at  $(x_0, y_0)$ : tangent plane  
of  $f$  at  
 $(x_0, y_0)$

$$L(x,y) = f(x_0, y_0) + f'_x(x_0, y_0) \cdot (x - x_0) + f'_y(x_0, y_0) \cdot (y - y_0)$$