

LAST TIME: Optimization

Tuesday 10/4/16

↳ To find critical points, one solves

$$\nabla f = 0$$

↳ In two variables, (if $f = f(x, y)$) there was the 2nd derivative test for determining if the critical points are local extrema.

2nd Derivative Test:

Suppose $\nabla f(a, b) = 0$ and define

$$D = \begin{vmatrix} f_{xx}(a, b) & f_{xy}(a, b) \\ f_{yx}(a, b) & f_{yy}(a, b) \end{vmatrix} = f_{xx}(a, b)f_{yy}(a, b) - f_{xy}(a, b)^2$$

↑
Hessian Determinant

★ Remark ★ → $f_{xy} = f_{yx}$ for differentiable functions

Phase 1:

$D > 0 \implies (a, b)$ is local extrema

$D < 0 \implies$ the graph of f at (a, b) has a saddle

$D = 0 \implies$ test is inconclusive

Phase 2: If $D > 0$ we apply the 2nd derivative test in one variable,

$f_{xx}(a, b) > 0 \implies$ concave up (local min) ☺

$f_{xx}(a, b) < 0 \implies$ concave down (local max) ☹

★ Remark ★ → you can use f_{yy} to do this too.

TODAY: Lagrange multipliers & Extreme value Theorem

Lagrange Multipliers

baby example problem: Optimize the function $f(x,y) = x+y$
subject to the constraint, $x^2+y^2=1 \leftarrow g(x,y)$

Terminology:

optimize \rightarrow Find local max or min

Constraint \rightarrow Only look at values of the function where $x^2+y^2=1$

This example problem is a "two variable, Lagrange multiplier problem with one constraint."

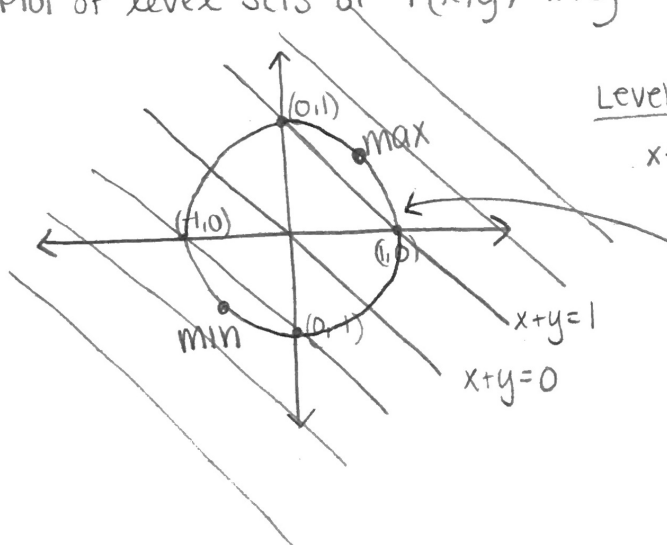
General Lagrange multiplier problem in two vars w/ one constraint:

optimize a given $f(x,y)$ subject to a given constraint

$g(x,y) = k$ where $k \in \mathbb{R}$

Let's look at our "baby example":

Plot of level sets of $f(x,y) = x+y$



Level Sets:

$x+y=c$ for some $c \in \mathbb{R}$

constraint:

$$x^2+y^2=1$$

* Recall

$\rightarrow \nabla f(x,y)$ at a point is the direction of maximal increase

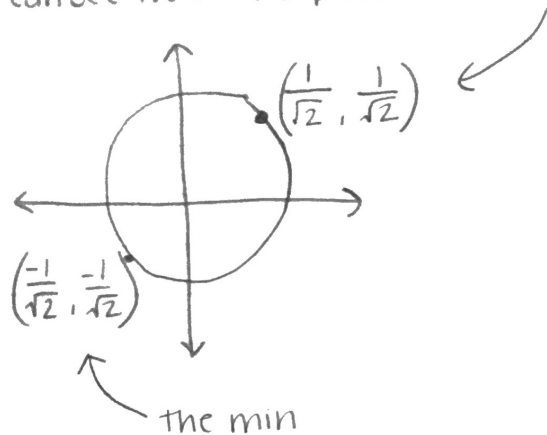
$\rightarrow \nabla g$ is normal to the curve defined by $g(x,y) = k$

or $g(x,y) = 0$

$$g(x,y) = x^2+y^2-1.$$

If $g(x,y) = k$ we can use $G(x,y) = g(x,y) - k$ and look at $G(x,y) = 0$ rather than $g(x,y) = 0$.

We can see from the picture the max is



The extreme values are where the level sets of $f(x,y)$ are tangent to the constraint $g(x,y) = k$.

In math speak:

$\nabla f = \lambda \nabla g$ normal vector of curve
 direction of max increase \leftarrow This is saying they are the same up to a constant multiple

This idea works in general.

Solution of 1 constraint Lagrange multiplier Problem:

Solving the system of equations,

$$\left. \begin{aligned} \nabla f &= \lambda \nabla g \\ g(x,y) &= k \end{aligned} \right\}$$

gives candidates for max/min. Plugging them into $f(x,y)$ and taking the max & min solves the problem

Solution to our baby problem:

$$\left\{ \begin{aligned} f(x,y) &= x+y && \leftarrow \text{function we want to optimize} \\ x^2+y^2 &= 1 && \leftarrow \text{constraint} \end{aligned} \right.$$

OR

$$\left. \begin{aligned} g(x,y) &= x^2+y^2 \\ k &= 1 \end{aligned} \right|$$

$$\left. \begin{aligned} G(x,y) &= x^2+y^2-1 \\ k &= 0 \end{aligned} \right|$$

SETUP OUR SYSTEM:

$$\begin{cases} \nabla f = \lambda \nabla g \\ g(x,y) = 1 \end{cases} \Leftrightarrow \begin{cases} (1,1) = \lambda(2x, 2y) \\ x^2 + y^2 = 1 \end{cases} \Leftrightarrow \begin{cases} 1 = 2\lambda x \\ 1 = 2\lambda y \\ x^2 + y^2 = 1 \end{cases} \star$$

Suppose $\lambda \neq 0$

$$\star \Rightarrow x = \frac{1}{2\lambda}, y = \frac{1}{2\lambda}$$

putting into constraint:

$$(\frac{1}{2\lambda})^2 + (\frac{1}{2\lambda})^2 = 1$$

$$(2\lambda)^2 = 2$$

$$4\lambda^2 = 2$$

$$\lambda^2 = \frac{1}{2}$$

$$\lambda = \pm \sqrt{\frac{1}{2}}$$

$$\lambda = \pm \frac{1}{\sqrt{2}} \rightarrow \lambda = \frac{+1}{\sqrt{2}} \text{ OR } \lambda = \frac{-1}{\sqrt{2}}$$

$$\lambda = \frac{+1}{\sqrt{2}} : x = \frac{1}{2\lambda} = \frac{1}{2(\frac{1}{\sqrt{2}})} = \frac{1}{\sqrt{2}}$$

similarly $y = \frac{1}{\sqrt{2}}$

so we have a candidate point $(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$

$$\lambda = \frac{-1}{\sqrt{2}} : x = \frac{1}{2\lambda} = \frac{1}{2(-\frac{1}{\sqrt{2}})} = -\frac{1}{\sqrt{2}}$$

similarly $y = -\frac{1}{\sqrt{2}}$

so we have a candidate point $(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}})$

IF $\lambda = 0$

$$1 - 2\lambda x \Rightarrow 1 = 0$$

This is nonsense, so $\lambda \neq 0$

Check our values:

$$f\left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right) = \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} = \frac{2}{\sqrt{2}} = \boxed{\sqrt{2}}$$

← max value

$$f\left(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right) = -\frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} = -\frac{2}{\sqrt{2}} = \boxed{-\sqrt{2}}$$

← min value

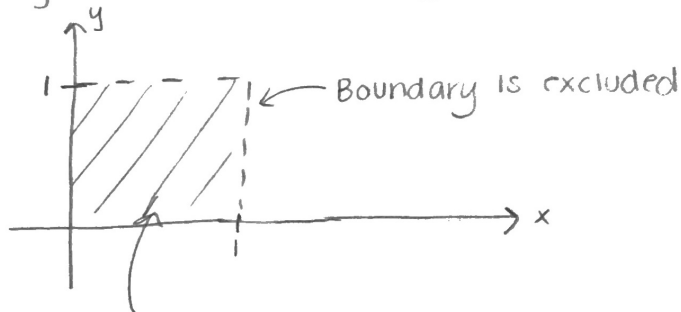
EXTREME VALUE THEOREM

Theorem: Let $R \subseteq \mathbb{R}^2$ which is closed and bounded
 Then any continuous function achieves a max and min on R .

We need to explain what these things mean.

example:

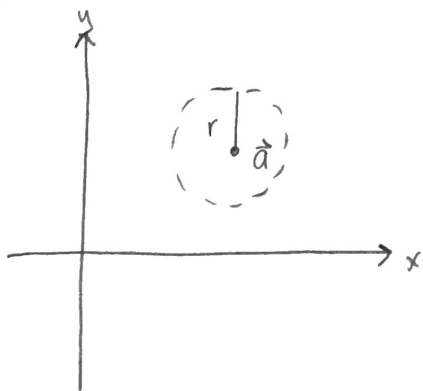
$$\{(x,y) : 0 < x < 1 \text{ and } 0 < y < 1\}$$



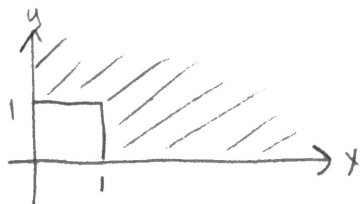
This is an open set

Def: A set $S \subseteq \mathbb{R}^2$ is open if for every $\vec{a} \in S$, there exists an open disc centered at \vec{a} completely confined in S .

$$\begin{aligned} \left(\begin{array}{l} \text{open disc centered at } \vec{a} \\ \text{of radius } r \end{array} \right) &= B_r(\vec{a}) \\ &= \{ \vec{x} \in \mathbb{R}^2 : |\vec{x} - \vec{a}| < r \} \end{aligned}$$



Def: A set $S \subseteq \mathbb{R}^2$ is closed if it is the complement of an open set.



$\mathbb{R}^2 \setminus$ previous example
 complement

$$\{(x,y) : \text{not } (0 < x < 1 \text{ and } 0 < y < 1)\} =$$

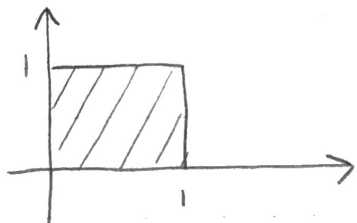
$$\{(x,y) : x \leq 0 \text{ or } x \geq 1 \text{ or } y \leq 0 \text{ or } y \geq 1\}$$

This set is closed but not bounded.

↑ does have parts going off to ∞

Example: A closed & bounded set ("compact" set)

The set $\{(x,y) : 0 \leq x \leq 1 \text{ and } 0 \leq y \leq 1\}$ is closed and bounded



example problem: Find the extreme values of

$$f(x,y) = x^2 + y^2$$

on the square

$$\{(x,y) : 0 \leq x \leq 1 \text{ and } 0 \leq y \leq 1\}$$

How to solve:

Find local extrema on the interior

$\nabla f = 0$ (throw away values outside the domain)

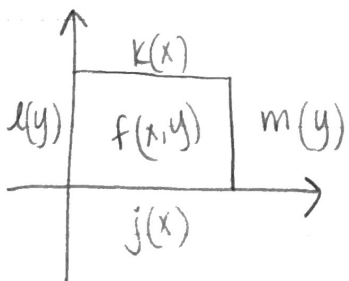
Find extreme values on the boundary

$$j(x) = f(x,0)$$

$$l(y) = f(0,y)$$

$$k(x) = f(x,1)$$

$$m(y) = f(1,y)$$



Taking the max (resp. min) of all the values from the first two parts gives your extrema.