

Lecture 32: Optimization

In the last lecture, we were mainly concerned with local extrema at interior points of a set, showing that — for a differentiable function — there must be stationary points (where the gradient = $\vec{0}$). What about global extrema?

Theorem 1: (i) If $S \subseteq \mathbb{R}^n$ is a closed, bounded set, and f is continuous on S , then it attains (absolute/global) max + min on S .

(ii) These extrema can occur at

- $\partial S =$ boundary of S
- singular points of $f(x)$ (i.e. f is not differentiable on a neighborhood of the point)
- stationary points of $f(x)$ (in $\text{int}(S)$ by definition).

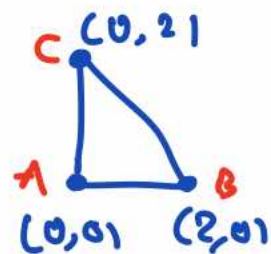
I defer the proof of (i) for now; (ii) is immediate from yesterday's result: let $S_0 \subseteq S$ be the interior of the subset on which f is differentiable, & let \vec{a} be an extremum of f .

If $\vec{a} \in S_0$, we know $\vec{\nabla} f(\vec{a}) = \vec{0}$. Otherwise, either $\vec{a} \in \partial S$ or $\vec{a} \in \text{int}(S) \setminus S_0 =$ singular points.

Upshot: We get an algorithm for locating f 's extrema, known as optimizing f :

- Step 1** minimize/maximize f on $\partial\mathcal{D}$
- Step 2** evaluate f at stationary points (if any obvious bad pts.)
- Step 3** pick the largest/smallest from Steps 1 & 2.

Ex 1 / Optimize $f(x,y) = 8xy - x - y$ on the closed triangular region shown:



• \overline{AC} : $y=0$. $f(x,0) = -x \rightarrow$ max at A: 0
 \rightarrow min at C: -2

• \overline{AB} : $x=0$. $f(0,y) = -y \rightarrow$ max at A: 0
 \rightarrow min at B: -2

• \overline{BC} : $f(2-t, t) = 16t - 8t^2 - 2$

$\Rightarrow 0 = \frac{d}{dt} f(2-t, t) = 16 - 16t$ gives $t=1$, and $f(1,1) = 6$

• Interior: $\vec{0} = \vec{\nabla} f = (8y-1, 8x-1)$

$\Rightarrow (x,y) = (\frac{1}{8}, \frac{1}{8})$, at which $f = -\frac{1}{8}$ (it's a saddle point, but we don't care)

Conclude that overall \swarrow max = 6, at (1,1)

\searrow min = -2, at B & C. //

Ex 2 / Optimize $f(x,y) = 2x^2 + y^2 - 4x - 2y + 5$ on the closed set $\mathcal{D} = \{(x,y) \mid x^2 + \frac{y^2}{2} \leq 1\}$.

• $\partial\mathcal{D}$ = ellipse, parametrized by $(x,y) = (\cos t, \sqrt{2} \sin t)$

Write $F(t) = f(x(t), y(t))$, so that Chain Rule \Rightarrow

$$\begin{aligned}
 F'(t) &= f_x(x(t), y(t)) x'(t) + f_y(x(t), y(t)) y'(t) \\
 &= (4x(t) - 4)(-\sin t) + (2y(t) - 2)(\sqrt{2} \cos t) \\
 &= (\cancel{4 \cos t} - 4)(-\sin t) + (\cancel{2\sqrt{2} \sin t} - 2)(\sqrt{2} \cos t) \\
 &= 4 \sin t - 2\sqrt{2} \cos t.
 \end{aligned}$$



Setting $0 = F'(t)$ gives $\tan(t) = \frac{\sqrt{2}}{2}$

$$\Rightarrow x(t) = \cos(\arctan(\frac{\sqrt{2}}{2}) + \pi) = -\frac{2}{\sqrt{6}}$$

$$y(t) = \sqrt{2} \sin(\arctan(\frac{\sqrt{2}}{2}) + \pi) = -\frac{2}{\sqrt{6}}$$

$$f(\frac{2}{\sqrt{6}}, \frac{2}{\sqrt{6}}) = 7 - 2\sqrt{6}, \quad f(-\frac{2}{\sqrt{6}}, -\frac{2}{\sqrt{6}}) = 7 + 2\sqrt{6}$$

• interior: $\vec{0} = \nabla f = (4x - 4, 2y - 2) \Rightarrow (x, y) = (1, 1)$

... and $f(1, 1) = 2 (< 7 - 2\sqrt{6})$ is the minimum
 NOGO! $(1, 1)$ is not in \mathcal{D} !

So $7 \pm 2\sqrt{6}$ are the max/min values. //

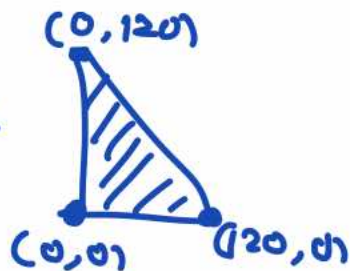
Ex 3 / Find 3 positive numbers that add up to 120

& such that their product is a maximum.

• What are the set \mathcal{D} & function f ?

want to maximize xyz on $\begin{cases} x+y+z = 120 \\ x, y, z \geq 0 \end{cases}$

i.e. $f(x, y) = xy(120 - x - y)$ on $\mathcal{D} =$



• f is 0 on $\partial\mathcal{D}$

• so consider $\vec{0} = \nabla f = \nabla(120xy - x^2y - xy^2)$

$$\Rightarrow (0,0) = (120y - 2xy - y^2, 120x - 2xy - x^2)$$

$$\Rightarrow y(120 - 2x - y) = 0 = x(120 - 2y - x)$$

$$\Rightarrow 120 - 2x - y = 0 = 120 - 2y - x$$

don't want x or $y = 0$

$$\Rightarrow y = x$$

$$\Rightarrow 120 - 3x = 0 \Rightarrow x = 40 \Rightarrow y = 40 \Rightarrow z = 40$$

$$\Rightarrow \text{max} = 40^3 = 64,000.$$

//

Ex 4 / Find the minimum distance between the origin and the surface $z^2 = x^2y + 4$:

• let $P = (x, y, z)$ be any point on the surface.

$(\text{dist}(P, 0))^2 = x^2 + y^2 + z^2$ is easier to work with and optimizing it is equivalent to optimizing $\text{dist}(P, 0)$

So consider $f(x, y) = x^2 + y^2 + (x^2y + 4)$ on $\mathcal{D}_0 = \{(x, y) \mid x^2y + 4 \geq 0\}$.
(why?)

• start by finding the critical points:

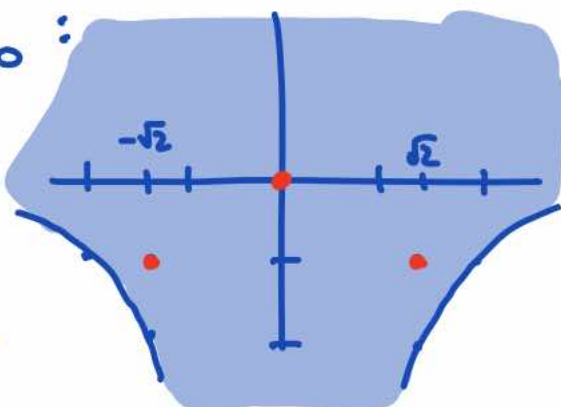
$$\vec{0} = \nabla f = (2x + 2xy, 2y + x^2) \Rightarrow y = -\frac{x^2}{2}$$

$$\Rightarrow 0 = 2x + 2xy = 2x - x^3 \Rightarrow x = 0, \pm\sqrt{2}$$

$$\Rightarrow (x, y) = (\underline{0, 0}), (\underline{\sqrt{2}, -1}), (\underline{-\sqrt{2}, -1}) \Rightarrow f = 4, 5, 5$$

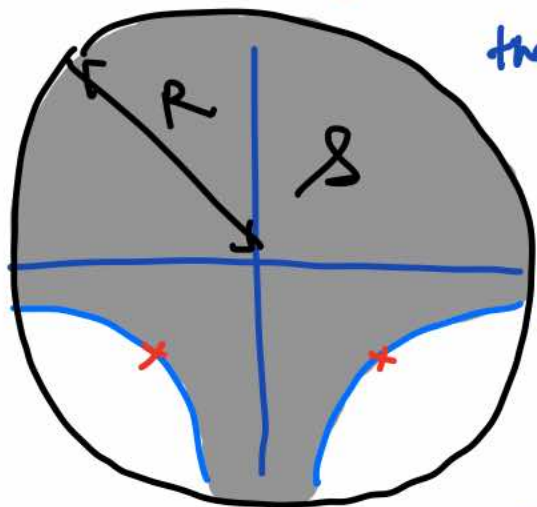
• now look at \mathcal{D}_0 :

This isn't bounded, so let \mathcal{D} be its intersection with



It is defined by $y \geq -\frac{1}{2}x^2$

a big disk of radius R : on the part of the boundary



that's circular, $f \geq R^2$ which we can take arbitrarily large. On the

curve $t \mapsto (t, -4t^2)$ ($t \neq 0$),

we have $f = x^2 + y^2 = t^2 + \frac{16}{t^4}$

which is minimized when $0 = f' = 2t - \frac{64}{t^5}$

$$\Rightarrow t^6 = 32 \Rightarrow t = 2^{5/6} \Rightarrow f = 2^{5/3} + 2^{2/3} = 3 \cdot 2^{2/3}$$

- Since $3 \cdot 2^{2/3} > 4$, the closest points are at $(0, 0, \pm 2)$ with distance $\sqrt{4} = 2$. //

The last 2 examples are constrained extremum problems, which in general can be quite difficult: in particular, sometimes you can't use the constraint to solve for one variable in terms of the others.

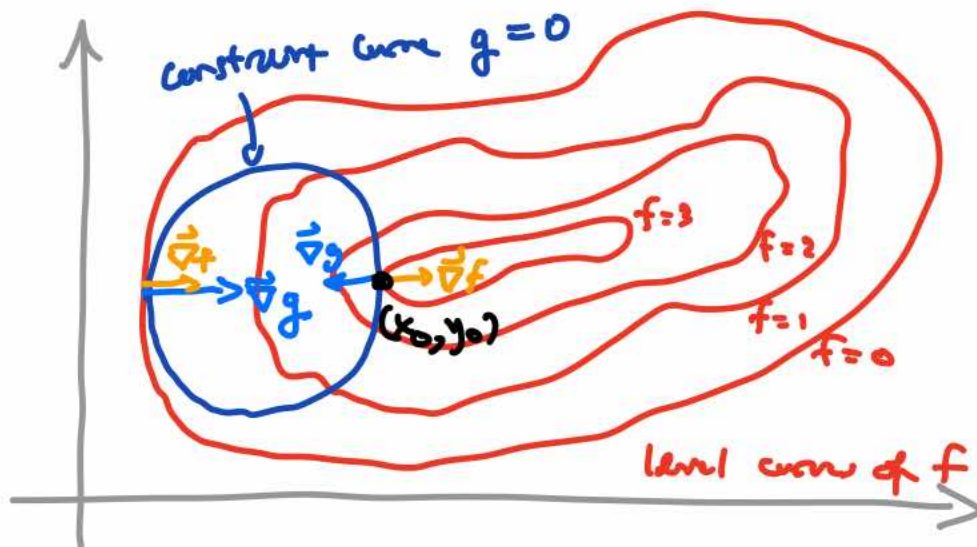
Fortunately, there is a better approach: Consider first

the case where we

want to optimize

$f(x, y)$ subject to

$g(x, y) = 0$:



We shall take it to be geometrically evident that the level curve $f = k$ with greatest possible k intersecting the constraint curve is tangent to it at the intersection point(s). Hence, their normal vectors are parallel, and

$$\vec{\nabla} f = \lambda \vec{\nabla} g$$

at (x_0, y_0) on $g = 0$ where f is maximized.

A little less heuristically, if $\vec{r}(t) = (x(t), y(t))$ parametrizes $g(x, y) = 0$, then the function f is maximized on $g = 0$ where

$$0 = \frac{d}{dt} f(\vec{r}(t)) = \vec{\nabla} f(\vec{r}(t)) \cdot \vec{r}'(t)$$

for $t = \text{some } t_0$; and then $\vec{\nabla} f(\vec{r}(t_0)) \perp \vec{r}'(t_0)$.

But $\vec{\nabla} g(\vec{r}(t_0)) \perp \vec{r}'(t_0)$ as well (since gradient is normal to $g = 0$, and $\vec{r}'(t_0)$ is tangent to it), so $\vec{\nabla} f(\vec{r}(t_0))$ is parallel to $\vec{\nabla} g(\vec{r}(t_0))$.

This approach to solving constrained extremum problems generalizes to 3, 4, etc. variables. For (say) 3 variables, here's the step-by-step:

Step 1 Make sure the constraint is in the form $g(x, y, z) = 0$. [e.g. $x^2 + y^2 = z^2$ is not]

Step 2 Set $\vec{\nabla} f = \lambda \vec{\nabla} g$

Step 3 Solve the resulting set of equations

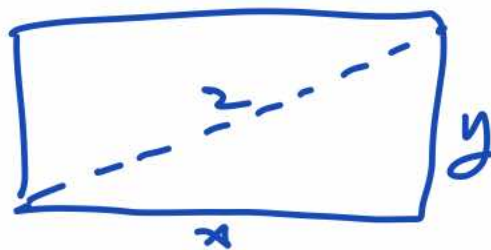
(3 from step 2, 1 from Step 1: $g=0$) to find (x, y, z) .

Step 4 Evaluate f at each of the points found, to see when it is really largest/smallest.

Ex 5 / Find the greatest area a rectangle can have with diagonal of length 2.

1) $f(x, y) = xy$, $g(x, y) = x^2 + y^2 - 4$

2) $\vec{\nabla} f = \lambda \vec{\nabla} g \rightarrow (y, x) = \lambda(2x, 2y)$



3) $y = 2\lambda x$, $x = 2\lambda y$, $x^2 + y^2 = 4$

$\Rightarrow xy = 2\lambda x^2$, $xy = 2\lambda y^2 \Rightarrow \lambda = 0$ or $x^2 = y^2$
impossible: would give $(x, y) = (0, 0) \rightarrow g \neq 0$

$\Rightarrow \begin{cases} 2x^2 = 4 \\ x = y \end{cases} \Rightarrow x = \sqrt{2} = y$

4) $f(\sqrt{2}, \sqrt{2}) = 2 = A_{\max}$. //

The method we've just been describing is that of Lagrange multipliers. We will continue with this on Wednesday.