

Section 4.3

I

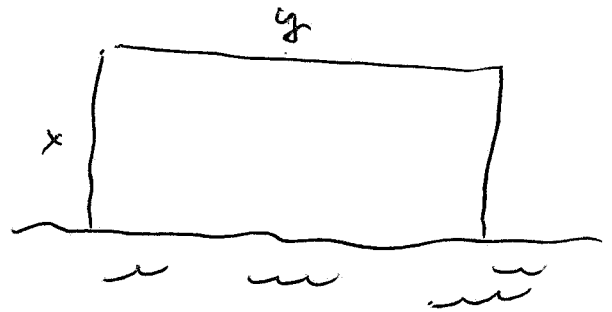
Recall optimization in Calculus I:

ex. Using 1800 linear feet of fence,
construct a ^{rectangle} yard along a straight
river with the largest area possible:

Idea: Maximize $A = xy$

Subject to

$$1800 = 2x + y.$$



Here A is the objective func.

$1800 = 2x + y$ is the constraint

The constraint facilitates calculation by

① allowing us to change the objective
into a func of one variable.

② allow us to use single var. calculus
techniques to help solve.

Here, since $1800 = 2x + y$, $y = 1800 - 2x$, so

$$A = xy = x(1800 - 2x) = 1800x - 2x^2.$$

Note the shape of (the graph of) $A(x)$,
a parabola opening down with a max
at its vertex.

Calculus: Look for critical pts:

$A(x)$ is C^1 , so only at places where $A' = 0$

$$A'(x) = 1800 - 4x = 0$$

when $x = 450$.

Here $A''(450) = -4 < 0$ hence by 2nd der test,

$x = 450$ is a local max.

Solution: Dimension of yard as $x = \text{width} = 450$

and $y = \text{length} = 900$, with area

$$A = xy = 450(900) = \dots$$

Notes ① Here, it is implicit that $x > 0, y > 0$
 so "region" in xy -plane is the
open first quadrant,

~~② A critical pt of $A(x)$~~

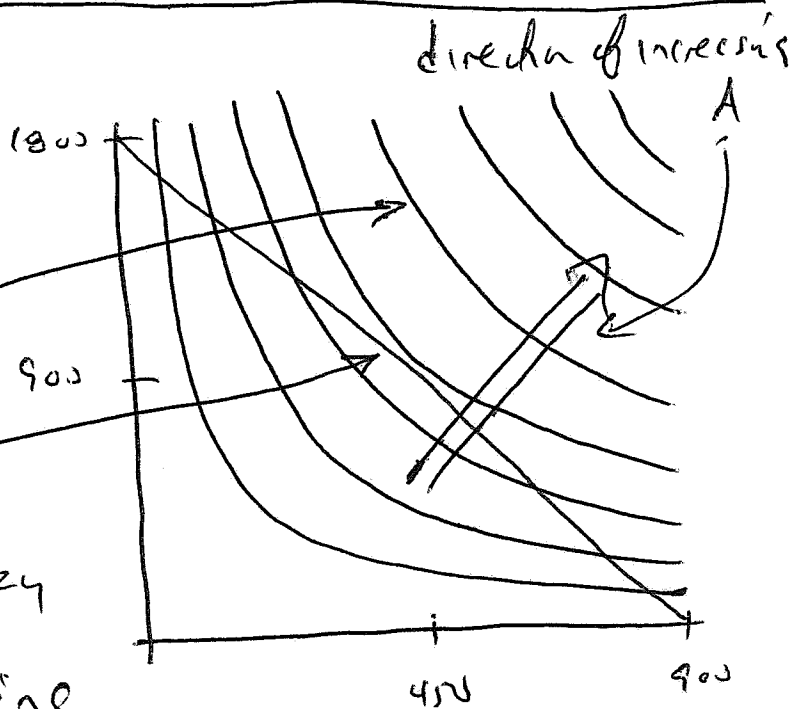
② Also implicit: $x < 1800, y < 1800$.

③ ∇A critical pt of $A(x)$ on $(0, 900)$
 and since "edges" go to 0 and $A(x) \rightarrow \infty$
 on interval, of course it is a max.

Different view point:

$A = xy$ level sets \rightarrow

Constraint \rightarrow



if we are forced to stay
 on the constraint line

and look for the largest value of A along it,
 where will we find it?

Here, we leave both functions as functions of both x and y and look for a geometric ~~reason~~ way to find an extremum of the objective given the constraint.

One can see (maybe look!) that along the constraint line, we are cutting through level sets of A for a while then we go tangent and then we again cut through level sets.

What is happening to the values of A then?

This new view allows us to generalize:

Optimize $f: X \subset \mathbb{R}^n \rightarrow \mathbb{R}$ (X open)

subject to $g: X \rightarrow \mathbb{R}$ where $g(\bar{x}) = c$.

We look for extrema of f while "stuck"
on the c -level set of g .

Note: We could simply solve $g(\bar{x}) = c$ for
one of its variables in terms of the
other, ~~but~~ and then "lower" the number of
variables of f by 1.

But sometimes this is not possible!

ex. Max $f(x, y, z) = x^2 + 3y^2 + y^2 z^4$

subject to $g(x, y, z) = e^{xy} - xyz + \cos\left(\frac{xy}{z}\right) = 2$.

Instead, appeal to the geometry!

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Thm 4.3.1 For $X \subset \mathbb{R}^n$ open, $f, g: X \rightarrow \mathbb{R}$
 C^1 -functions, let

$$S = \{ \vec{x} \in X \mid g(\vec{x}) = c \}$$

be the c -level set of g . Then

16 $f|_S$ has an extremum at $\vec{x}_0 \in S$,

where $\nabla g(\vec{x}_0) \neq \vec{0}$, then $\exists \lambda \in \mathbb{R} \Rightarrow$

$$\nabla f(\vec{x}_0) = \lambda \nabla g(\vec{x}_0)$$

Notes ① Extrema of f will happen at places
where ∇f is a multiple of ∇g , as
vectors.

② $\nabla f(\vec{x}_0) = \lambda \nabla g(\vec{x}_0)$ is a set of
 n -equations (nonlinear) in
 $n+1$ unknowns (all \vec{x} , and λ).
 \Rightarrow Lots of solutions!!!

③ But add in the constraint, and

$$f_{x_1}(\bar{x}) = \lambda g_{x_1}(\bar{x})$$

⋮

$$f_{x_n}(\bar{x}) = \lambda g_{x_n}(\bar{x})$$

$$g(\bar{x}) = c$$

is a set of $(n+1)$ -eqns in $(n+1)$ -unknowns.
(typically isolated solutions).

④ λ is called a Lagrange Multiplier.
Not usually important what its value is ...

ex. Identify all critical pts of

$$f(x, y) = 5x + 2y$$

subject to $g(x, y) = 5x^2 + 2y^2 = 14$.

Soln Here $\nabla f(\bar{x}) = \begin{bmatrix} 5x \\ 2y \end{bmatrix}$ $\nabla g(\bar{x}) = \begin{bmatrix} 10x \\ 4y \end{bmatrix}$

So system is

$$f_x(\bar{x}) = \lambda g_x(\bar{x})$$

$$f_y(\bar{x}) = \lambda g_y(\bar{x})$$

$$g(\bar{x}) = c$$

① $5 = \lambda 10x$
 ② $2 = \lambda 4y$
 ③ $5x^2 + 2y^2 = 14$

Mass with λ : By ①, ② $x = \frac{1}{2\lambda} = y$.

So by ③ $\frac{5}{4\lambda^2} + \frac{2}{4\lambda^2} = 14 \Rightarrow \frac{1}{4\lambda^2} = 2$

or $\lambda^2 = \frac{1}{8} \Rightarrow \lambda = \pm \frac{1}{2\sqrt{2}}$

Hence $(x, y) = (\sqrt{2}, \sqrt{2})$, or

$(x, y) = (-\sqrt{2}, -\sqrt{2})$. See Mellenchic.

Geometrically, one can see whether it is a max or a min, and why the gradient condition is telling us.

How to judge analytically?

So what if we have multiple constraints?

(a) Each constraint tends to reduce the number of independent variables by 1.

(b) Each constraint tends to reduce the dimension of the space we evaluate the objective func on by 1.

(c) In \mathbb{R}^3 , one objective func has level sets which are surfaces.

Each constraint does also: 2 constraints (2 surfaces) typically intersect in a line

We then look for extrema of objective func along a line.

ex. Find extrema of $f(x, y, z) = 2x + y^2 - z^2$

Subject to $g_1(x, y, z) = x - 2z = 0$

$g_2(x, y, z) = x + z = 0$

Thm 4.3.2 Let $X \subset \mathbb{R}^n$ be open and

$f, g_1, \dots, g_k: X \rightarrow \mathbb{R}$ be C^1 -functions, ($k < n$)

Let $S = \{ \bar{x} \in X \mid g_1(\bar{x}) = c_1, \dots, g_k(\bar{x}) = c_k \}$

If $f|_S$ has an extremum at $\bar{x}_0 \in S$, where

$\nabla g_1(\bar{x}_0), \dots, \nabla g_k(\bar{x}_0)$ are linearly indep.

as vectors, then there exist scalars

$\lambda_1, \dots, \lambda_k$ s.t.

$$\nabla f(\bar{x}_0) = \lambda_1 \nabla g_1(\bar{x}_0) + \dots + \lambda_k \nabla g_k(\bar{x}_0)$$

Notes: (1) Recall lin. indep. also means all nonzero!

Analytically $\nabla f(\vec{x}) = \begin{bmatrix} 2 \\ 2y \\ -2z \end{bmatrix}$ $\nabla g_1(\vec{x}) = \begin{bmatrix} 1 \\ -z \\ 0 \end{bmatrix}$ $\nabla g_2(\vec{x}) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$

~~System~~ The constraint vectors are lin. indep. everywhere, so system is:

~~System~~

$$\nabla f(\vec{x}) = \lambda_1 \nabla g_1(\vec{x}) + \lambda_2 \nabla g_2(\vec{x})$$

$$\begin{bmatrix} 2 \\ 2y \\ -2z \end{bmatrix} = \lambda_1 \begin{bmatrix} 1 \\ -z \\ 0 \end{bmatrix} + \lambda_2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

$$x - 2y = 0$$

$$x + z = 0$$

$$\left. \begin{array}{l} 2 = \lambda_1 + \lambda_2 \\ 2y = -2\lambda_1 \\ -2z = \lambda_2 \\ x = 2y \\ x = -z \end{array} \right\} \Rightarrow \begin{array}{l} \textcircled{1} \quad \left. \begin{array}{l} 2 = -y - 2z \\ x = 2y \\ x = -z \end{array} \right\} 2y - z = 0 \\ \textcircled{2} \quad \left(\begin{array}{l} 2 = -y - 2z \\ 0 = 2y - z \end{array} \right) \end{array}$$

$$\textcircled{3} \quad \left. \begin{array}{l} 4 = -2y - 4z \\ 0 = 2y - z \end{array} \right\} \begin{array}{l} 4 = -5z \\ \underline{z = -\frac{4}{5}} \end{array} \dots$$

See McLaurin's