

Lecture 24

Extrema under Constraints

So far, we have minimized (maximized) functions $f(x_1, x_2)$ on their domains.

Sometimes, we need to find extrema of $f(x_1, x_2)$ under some additional constraint.

E.g. We want to find the maximum / minimum of

$$f(x_1, x_2) = x_1 x_2 + 2x_1^2 x_2^2$$

under the constraint

$$(x_1 - 1)^2 + 2(x_2 - 2)^2 = 2.$$

In general, we want to find the extrema of the function $f(x_1, x_2)$ under the constraint

$$\boxed{g(x_1, x_2) = 0.}$$

Theorem (Lagrange): Assume that f, g have continuous first-order partial derivatives and $f(x_1, x_2)$ has an extremum at (\bar{x}_1, \bar{x}_2) subject to $\boxed{g(x_1, x_2) = 0.}$

If $\nabla g(\bar{x}_1, \bar{x}_2) \neq 0$, then there exists

...

some s.t.

$$\nabla f(\bar{x}_1, \bar{x}_2) = \lambda \cdot \nabla g(\bar{x}_1, \bar{x}_2).$$

This suggests that when we want to optimize $f(x_1, x_2)$ s.t. the constraint $g(x_1, x_2) = 0$, we have to solve the system

$$(*) \quad \begin{cases} g(x_1, x_2) = 0 \\ \nabla f(x_1, x_2) = \lambda \nabla g(x_1, x_2) \end{cases}$$

The number λ is called the Lagrange multiplier.

Example: Find all extrema of $f(x_1, x_2) = e^{-x_1 x_2}$ subject to the constraint $x_1^2 + 4x_2^2 = 1$.

- The constraint can be written as $g(x_1, x_2) = 0$, where $g(x_1, x_2) = x_1^2 + 4x_2^2 - 1$.
- We now solve the system (*):

$$\nabla f(x_1, x_2) = \begin{pmatrix} -x_2 e^{-x_1 x_2} \\ -x_1 e^{-x_1 x_2} \end{pmatrix}$$

$$\nabla g(x_1, x_2) = \begin{pmatrix} 2x_1 \\ 8x_2 \end{pmatrix}$$

$$\begin{aligned} (*) \Rightarrow \quad & x_1^2 + 4x_2^2 - 1 = 0 \\ & -x_2 e^{-x_1 x_2} = \lambda \cdot 2x_1 \quad | \cdot 4x_2 \\ & -x_1 e^{-x_1 x_2} = \lambda \cdot 8x_2 \quad | \cdot x_1 \end{aligned}$$

$$\Rightarrow \begin{aligned} x_1^2 + 4x_2^2 - 1 &= 0 \\ -4x_2^2 e^{-x_1 x_2} &= 8x_1 x_2 \\ -x_1^2 e^{-x_1 x_2} &= 8x_1 x_2 \end{aligned}$$

$$\Rightarrow \begin{aligned} x_1^2 + 4x_2^2 - 1 &= 0 \\ x_1^2 e^{-x_1 x_2} &= 4x_2^2 e^{-x_1 x_2} \end{aligned}$$

$$\Rightarrow \begin{aligned} x_1^2 + 4x_2^2 &= 1 \\ x_1^2 &= 4x_2^2 \end{aligned} \Rightarrow \begin{aligned} 4x_2^2 + 4x_2^2 &= 1 \\ 8x_2^2 &= 1 \\ \Rightarrow x_2^2 &= \frac{1}{8} \\ \Rightarrow x_2 &= \pm \frac{1}{\sqrt{8}} \end{aligned}$$

$$\Rightarrow x_1^2 = 4 \cdot x_2^2 = 4 \cdot \frac{1}{8} = \frac{1}{2}$$

$$\Rightarrow x_1 = \pm \frac{1}{\sqrt{2}}$$

Therefore, the system (*) has 4 solutions:

$$\left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right), \left(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right), \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{8}}\right),$$

$$\left(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{8}}\right).$$

We have to find the values of f on all these points:

$$\dots \dots \dots (1, 1, -1)$$

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$$\begin{aligned}
 f\left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right) &= f\left(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right) \\
 &= e^{-\frac{1}{\sqrt{2}} \cdot \frac{1}{\sqrt{8}}} = e^{-\frac{1}{\sqrt{2 \cdot 8}}} \\
 &= \underline{\underline{e^{-\frac{1}{4}}}}
 \end{aligned}$$

$$\begin{aligned}
 \underline{f\left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{8}}\right)} &= \underline{f\left(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right)} \\
 &= \underline{\underline{e^{\frac{1}{\sqrt{2}} \cdot \frac{1}{\sqrt{8}}}}} = \underline{\underline{e^{\frac{1}{4}}}}
 \end{aligned}$$

The maximum value of f s.t. the constraint g is $e^{\frac{1}{4}}$.

Therefore, $\left(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right)$ and $\left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{8}}\right)$ are maxima.

The minimum value of f s.t. the constraint g is $e^{-\frac{1}{4}}$.

Therefore, $\left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{8}}\right)$ and $\left(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{8}}\right)$ are minima.

Example: Among two positive numbers with product equal to 1, find those whose sum is minimum.

Of course, we could say that $x_1 \cdot x_2 = 1$
 $\Rightarrow \boxed{x_2 = \frac{1}{x_1}}$, and the sum of the two numbers is

$$x_1 + x_2 = x_1 + \frac{1}{x_1}$$

So, we need to find the minimum of $h(x_1) = x_1 + \frac{1}{x_1}$, $x_1 > 0$.

It is also possible to use Lagrange multipliers:

We want to minimize

$$f(x_1, x_2) = x_1 + x_2, \quad \underline{x_1, x_2 > 0}$$

subject to the constraint $x_1 \cdot x_2 = 1$.

We set $g(x_1, x_2) = x_1 \cdot x_2 - 1$.

Now we solve the system (*) :

$$\cdot \nabla f(x_1, x_2) = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\cdot \nabla g(x_1, x_2) = \begin{pmatrix} x_2 \\ x_1 \end{pmatrix}$$

$$\begin{array}{l} \underline{(*)} \\ \implies \end{array} \left. \begin{array}{l} x_1 x_2 - 1 = 0 \\ 1 = 2 \cdot x_2 \\ 1 = 2 x_1 \end{array} \right\} \begin{array}{l} \boxed{x_1 x_2 = 1} \\ \boxed{2x_1 = 2x_2} \end{array}$$

$$\implies x_1 = x_2 \implies x_1^2 = 1$$

$$\implies \underline{\underline{x_1 = \pm 1}}, \quad \underline{\underline{x_2 = \pm 1}}$$

Because $x_1, x_2 > 0$, we get the solution

$$(\bar{x}_1, \bar{x}_2) = (1, 1).$$

Thus, among all positive numbers x_1, x_2 with $x_1 \cdot x_2 = 1$, the sum $x_1 + x_2$ is minimum when $x_1 = x_2 = 1$.

System of Differential Equations

We have solved certain DE of the form

$$y'(t) = f(t, y)$$

(for example, $y' = y^2 - 2y + 2$).

Such a DE has a "general solution".
E.g., the DE $y' = y$ has the general solution

$$y(t) = c \cdot e^t \quad (c \in \mathbb{R} \text{ is a constant})$$

When we choose some specific value of $c \in \mathbb{R}$, we get a particular solution, e.g., $y = 5 \cdot e^t$.

Also, if we solve the DE with an initial condition, we solve an "initial value problem", and this has a unique solution.

E.g., $y' = y$, $y(0) = 1$ has the solution

$$y(t) = e^t.$$

We now deal with systems of DE, i.e.

$$\begin{pmatrix} y_1'(t) \\ y_2'(t) \\ \vdots \\ y_n'(t) \end{pmatrix} = \begin{pmatrix} f_1(t, y_1, \dots, y_n) \\ f_2(t, y_1, \dots, y_n) \\ \vdots \\ f_n(t, y_1, \dots, y_n) \end{pmatrix}$$

The unknown here is the vector

$$y(t) = \begin{pmatrix} y_1(t) \\ \vdots \\ y_n(t) \end{pmatrix}.$$

Here :

$$\begin{aligned} y_1'(t) &= a_{11} y_1(t) + \dots + a_{1n} y_n(t) \\ &\vdots \\ y_n'(t) &= a_{n1} y_1(t) + \dots + a_{nn} y_n(t) \end{aligned}$$

$$\Leftrightarrow y'(t) = \frac{dy(t)}{dt} = A \cdot y(t)$$

$$\text{with } A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix}.$$

Such a system is called a homogeneous linear first-order system of DEs with constant coefficients.

We will only deal with the case $n=2$,
i.e.,

$$(**) \begin{cases} y_1'(t) = a_{11}y_1(t) + a_{12}y_2(t) \\ y_2'(t) = a_{21}y_1(t) + a_{22}y_2(t) \end{cases}$$

$$\Leftrightarrow y'(t) = A \cdot y(t), \quad A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}.$$

The Direction Field

We consider the system (**).

On the y_1y_2 -plane, at each point (y_1, y_2) , we draw the vector

$$\begin{pmatrix} a_{11}y_1 + a_{12}y_2 \\ a_{21}y_1 + a_{22}y_2 \end{pmatrix}.$$

The collection of all these vectors is called the direction field of the system (**).

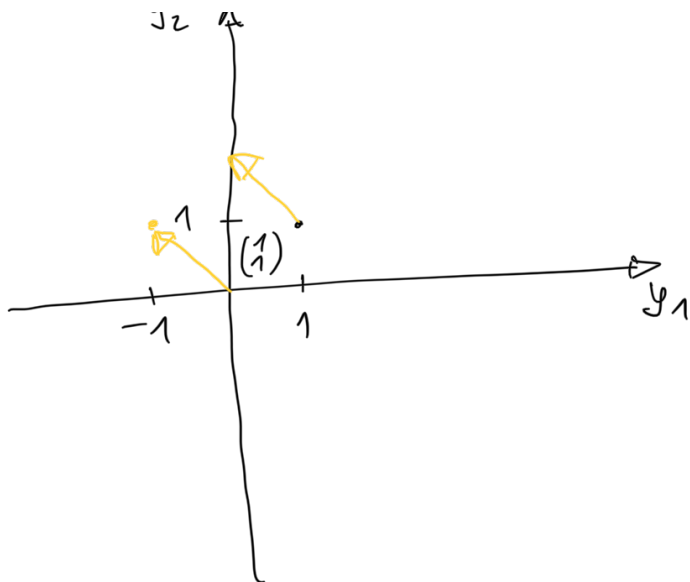
Example: Consider the system

$$\begin{cases} y_1' = y_1 - 2y_2 \\ y_2' = y_2 \end{cases}.$$

$$\text{Here, } A = \begin{pmatrix} 1 & -2 \\ 0 & 1 \end{pmatrix}.$$

u .

$$At (y_1, y_2) = \begin{pmatrix} 1 \\ 1 \end{pmatrix},$$



the vector is

$$\begin{pmatrix} y_1 + (-2)y_2 \\ y_2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 - 2 \\ 1 \end{pmatrix}$$

$$= \underline{\underline{\begin{pmatrix} -1 \\ 1 \end{pmatrix}}}$$

What does the direction field represent?
 Suppose we want to solve the system

$$y' = Ay$$

with initial condition $y(0) = \begin{pmatrix} \bar{y}_1 \\ \bar{y}_2 \end{pmatrix}$.

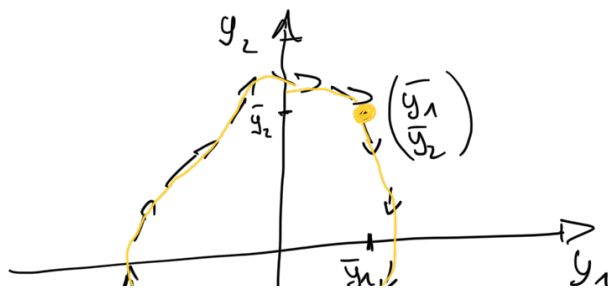
Then the unique solution will have the form

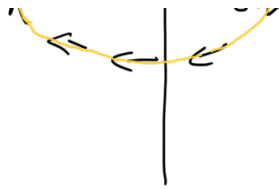
$$y_1 = y_1(t)$$

$$y_2 = y_2(t), \quad t \in \mathbb{R},$$

which represents a curve in the $y_1 y_2$ -plane that goes through the point $\begin{pmatrix} \bar{y}_1 \\ \bar{y}_2 \end{pmatrix}$.

At each point of the solution curve, the direction field is a tangent to the curve:





Solving the system $y'(t) = A \cdot y(t)$

- Assumption: A has two distinct eigenvalues λ_1, λ_2
- First, we find the two eigenvalues λ_1, λ_2 of A
- We find an eigenvector $u = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$ corresponding to λ_1 and an eigenvector $v = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$ corr. to λ_2 .
- The general solution is:

$$y(t) = c_1 e^{\lambda_1 t} u + c_2 e^{\lambda_2 t} v$$

$$\Leftrightarrow \begin{cases} y_1(t) = c_1 e^{\lambda_1 t} u_1 + c_2 e^{\lambda_2 t} v_1 \\ y_2(t) = c_1 e^{\lambda_1 t} u_2 + c_2 e^{\lambda_2 t} v_2 \end{cases}$$

$$(t \in \mathbb{R})$$

Where $c_1, c_2 \in \mathbb{R}$ are constants.

Example: Solve $(**)$ with $A = \begin{pmatrix} 2 & -2 \\ 2 & -3 \end{pmatrix}$
and initial condition $y(0) = \begin{pmatrix} -1 \\ 4 \end{pmatrix}$.

- Find the eigenvalues of A :

$$\det(\lambda \cdot I - A) = \begin{vmatrix} \lambda - 2 & 2 \\ -2 & \lambda + 3 \end{vmatrix}$$

$$\det(A - \lambda I) = \begin{vmatrix} \lambda - 2 & 0 \\ 0 & \lambda + 3 \end{vmatrix}$$

$$= (\lambda - 2)(\lambda + 3) + 4$$

$$= \lambda^2 + 3\lambda - 2\lambda - 6 + 4$$

$$= \lambda^2 + \lambda - 2$$

$$= (\lambda + 2)(\lambda - 1)$$

The eigenvalues of A are

$$\lambda_1 = -2, \quad \lambda_2 = 1.$$

• Find an eigenvector for $\lambda_1 = -2$:

$$A \cdot u = -2 \cdot u \Rightarrow \begin{pmatrix} 2 & -2 \\ 2 & -3 \end{pmatrix} \cdot u = \begin{pmatrix} -2u_1 \\ -2u_2 \end{pmatrix}$$

$$\Rightarrow \begin{cases} 4u_1 - 2u_2 = 0 \\ 2u_1 - u_2 = 0 \end{cases} \quad \left. \vphantom{\begin{cases} 4u_1 - 2u_2 = 0 \\ 2u_1 - u_2 = 0 \end{cases}} \right\} \underline{\underline{2u_1 = u_2}}$$

One eigenvector is $u = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$.

• Find an eigenvector for $\lambda_2 = 1$:

$$A v = 1 \cdot v \Rightarrow \begin{pmatrix} 2 & -2 \\ 2 & -3 \end{pmatrix} \cdot v = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

$$\Rightarrow \begin{cases} v_1 - 2v_2 = 0 \\ 2v_1 - 4v_2 = 0 \end{cases} \quad \left. \vphantom{\begin{cases} v_1 - 2v_2 = 0 \\ 2v_1 - 4v_2 = 0 \end{cases}} \right\} v_1 = 2v_2$$

One eigenvector is $v = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$.

- The general solution of the system is

$$y(t) = C_1 e^{\lambda_1 t} u + C_2 e^{\lambda_2 t} v$$

$$= C_1 e^{-2t} \begin{pmatrix} 1 \\ 2 \end{pmatrix} + C_2 e^t \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$\Leftrightarrow \begin{cases} y_1(t) = C_1 e^{-2t} + 2C_2 e^t \\ y_2(t) = 2C_1 e^{-2t} + C_2 e^t \end{cases}$$

($t \in \mathbb{R}$, $C_1, C_2 \in \mathbb{R}$ constants)

- Specific solution:

$$y(\underbrace{0}_{t=0}) = \begin{pmatrix} -1 \\ 4 \end{pmatrix} = \begin{pmatrix} C_1 \cdot e^{\underline{0}} + 2C_2 \cdot e^{\underline{0}} \\ 2C_1 e^{\underline{0}} + C_2 e^{\underline{0}} \end{pmatrix}$$

$$= \begin{pmatrix} C_1 + 2C_2 \\ 2C_1 + C_2 \end{pmatrix}$$

$$\text{So, } \begin{cases} -1 = C_1 + 2C_2 \\ 4 = 2C_1 + C_2 \end{cases} \Rightarrow \underline{C_1 = -1 - 2C_2}$$

$$\Rightarrow 4 = 2 \cdot (-1 - 2C_2) + C_2$$

$$= -2 - 4C_2 + C_2$$

$$\Rightarrow 6 = -3C_2 \Rightarrow \underline{\underline{C_2 = -2}}$$

$$\Rightarrow C_1 = -1 - 2 \cdot (-2) = 3$$

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The unique solution is

$$y(t) = \begin{pmatrix} 3e^{-2t} - 4e^t \\ 6e^{-2t} - 2e^t \end{pmatrix}, \quad t \in \mathbb{R}$$
