

Constrained Optimization (Lecture 5)

March 31st, 2024

Last time → Second order: \rightarrow necessary \checkmark sufficient \checkmark conditions for optimality

→ Case of linear equality constraints.

Using elimination of variables.

$$(1) \begin{cases} \min_{\underline{x}} f(\underline{x}) \\ \text{s.t. } A\underline{x} = b \end{cases}$$

A is $m \times n$,
 $m < n$.

Today

- Example of using elimination of variables for equality constrained problem
- The KKT system for the case when in (1), f is quadratic.
- Using the KKT system for the general case (1).

Example of Equality constrained (solved by elimination of variables).

$$\begin{cases} \min \underline{f(x)} = x_1^2 - 2x_1 + x_2^2 - x_3^2 + 4x_3 \\ \text{s.t. } x_1 - x_2 + 2x_3 = 2 \end{cases}$$

Let's eliminate x_1 . (solve for x_1 in terms of x_2 and from the constraint, and replace it in the obj. function)

$$x_1 = x_2 - 2x_3 + 2.$$

$$\text{So } \underline{f(x_2, x_3)} = \underline{f(x_2, x_3)} =$$

$$= 2x_2^2 + 3x_3^2 + 4x_2x_3 + 2x_2 \rightarrow \text{the reduced objective function}$$

Let's rename the variables:

$$x_2 \rightarrow \alpha_1$$

$$x_3 \rightarrow \alpha_2$$

So, we want to minimize the unconstrained function:

$$\phi(\underline{\alpha}) = \varphi(\alpha_1, \alpha_2) = 2\alpha_1^2 + 3\alpha_2^2 - 4\alpha_1\alpha_2 + 2\alpha_1$$

Set $\nabla \varphi(\underline{\alpha}) = \underline{0}$ and solve for the stationary points.
(first-order necessary cond. for unconstr. optim).

But from last time, we showed that:

$$\nabla \varphi(\underline{\alpha}) = \underline{Z}^\top \nabla \varphi(\hat{\underline{\alpha}} + \underline{Z}\underline{\alpha}) = \underline{Z}^\top \nabla \varphi(\hat{\underline{\alpha}})$$

where $\hat{\underline{\alpha}}$ is a feasible point.
($\hat{\alpha}_1 = \hat{\alpha}_2 = 2\hat{\alpha}_3 \in \mathbb{R}$).

and $\underline{Z} = [\underline{z}_1, \underline{z}_2]$ is an 3×2 matrix
whose columns are vectors in a basis
of $\text{Null}(A)$

Let's find the $\text{null}(A)$: set of vectors in \mathbb{R}^n (\mathbb{R}^3)
that are mapped by A into 0.

$$A = [1 \quad -1 \quad 2], \quad \underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$A\underline{x} = 0$
 \hookrightarrow is a vector, in general, of dimension equal to
the # of equality constraints.

(3)

$$x_1 - x_2 + 2x_3 = 0.$$

so here for x_1 , and let x_2 & x_3 be arbitrary:

$$x_1 = x_2 - 2x_3$$

So the set of vectors mapped by A into zero is of this form:

$$\left\{ \begin{bmatrix} x_2 - 2x_3 \\ x_2 \\ x_3 \end{bmatrix} \right\} = x_2 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}, \quad x_2, x_3 \in \mathbb{R}$$

Write it as a linear combination of two basis vectors.

$$\text{vectors } \underline{z}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \text{ & } \underline{z}_2 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix} \text{ form a basis for Null}(A).$$

$$\boxed{\text{So } Z = \begin{bmatrix} 1 & -2 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}} \quad \boxed{\text{& } Z^T = \begin{bmatrix} 1 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}}$$

Let's solve $D\varphi(\underline{x}) = \underline{0}$.

$$\Rightarrow \varphi(\underline{x}) = \begin{bmatrix} 4x_1 - 4x_2 + 2 \\ 6x_2 - 4x_1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\left. \begin{array}{l} 4x_1 - 4x_2 = -2 \\ -2x_1 + 3x_2 = 0 \end{array} \right| \bullet 2$$

$$\text{add together} \rightarrow 2x_2 = -2 \Rightarrow x_2 = -1$$

$$\left. \begin{array}{l} x_1 = -\frac{3}{2} \\ x_2 = -1 \end{array} \right\} \text{Ansatz}$$

Going back to the notation we get :

$$x_2 = -\frac{3}{2}$$

$$x_3 = -1.$$

$$x_1 = x_2 - 2x_3 + 2 = \frac{5}{2}$$

Use the constraint, get .

$$\text{The only stationary point is: } \underline{x}^* = \begin{bmatrix} \frac{5}{2} \\ -\frac{3}{2} \\ -1 \end{bmatrix}$$

To see if it is a minimizer, use second-order conditions.

($\nabla_{\underline{x}}^2 \Psi(\underline{x})$ is P.D. at \underline{x}^*)

Exercise: Compute $\nabla_{\underline{x}}^2 \Psi(\underline{x})$, and see if it is P.S.D. - or P.D.

Note: The example we tried here was special.

The objective function was quadratic. (original or the transformed one).

$$\Rightarrow \nabla \Psi(\underline{x}) = 0 \text{ was a } \underline{\text{linear system.}}$$

(can be solved with any linear solver)

When obj. func. is not quadratic \Rightarrow $\nabla \Psi(\underline{x}) = 0$ is a]
[non-linear system.]

Q) What if instead of elimination of variables, we use the KKT conditions?

Consider the same problem (Quadratic f, \mathbb{R}^m linear equality constraints)

$$(2) \begin{cases} \min \frac{1}{2} \underline{x} \cdot G \underline{x} + \underline{d} \cdot \underline{x} \\ \text{s.t. } A \underline{x} = b \end{cases}$$

$A \in \mathbb{R}^{m \times n}$ (m equality constraints)

$A \in \mathbb{R}^{m \times n}$

$$\underline{w} \cdot \nabla \mathcal{L}(\underline{x}^*, \underline{\lambda}^*) \leq 0$$

for all $\underline{w} \in C(\underline{x}^*, \underline{\lambda}^*)$

Let \underline{x}^* be
a local min
for (2)

$$\underline{w} \text{ s.t. } \nabla C_i(\underline{x}^*) \cdot \underline{w} = 0, i \in E$$

$$C_i(\underline{x}) = \underline{a}_i \cdot \underline{x} - b_i, i=1, 2, \dots, m.$$

KKT
conditions

(assume

A has full
row rank.

\Rightarrow L.i. rows
 \Rightarrow LICQ holds



there exists $\underline{\lambda}^* \in \mathbb{R}^m$ s.t.

$$\nabla_{\underline{x}} \mathcal{L}(\underline{x}^*, \underline{\lambda}^*) = \underline{0}, \text{ where } \mathcal{L}(\underline{x}, \underline{\lambda}) = \frac{1}{2} \underline{x} \cdot G \underline{x} + \underline{d} \cdot \underline{x} - \underline{\lambda} \cdot (\underline{A} \underline{x} - b)$$

Let's compute $\nabla_{\underline{x}} \mathcal{L}(\underline{x}, \underline{\lambda}) = G \underline{x} + \underline{d} - \underline{\lambda} \cdot \underline{A}$.

$$(\text{indeed: } \nabla_{\underline{x}} (\underline{\lambda} \cdot \underline{A} \underline{x}) = \nabla_{\underline{x}} (\underbrace{\underline{A}^T \underline{\lambda}}_{\underline{A}^T \underline{\lambda}} \cdot \underline{x}) = \underline{A}^T \underline{\lambda})$$

So, KKT condition say that:

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(6)

$$\begin{cases} G\bar{x}^* + \underline{d} - A^T \underline{\lambda}^* = 0 \\ A\bar{x}^* = \underline{b} \end{cases} \Rightarrow \begin{cases} G\bar{x}^* - A^T \underline{\lambda}^* = -\underline{d} \\ A\bar{x}^* = \underline{b} \end{cases}$$

$$m \left[\begin{array}{c} \overbrace{G}^n \\ -A^T \end{array} \right] \left[\begin{array}{c} \bar{x}^* \\ \underline{\lambda}^* \end{array} \right] = \left[\begin{array}{c} -\underline{d} \\ \underline{b} \end{array} \right]_m \quad (3)$$

(written) by $(m+n)$ system of equations.

Metric is not symmetric

Q

How do we use this in practice? (we don't have \bar{x}^*)

A

Start at some arbitrary \underline{x} .

We want to get to \bar{x}^* from \underline{x} .

$$\begin{matrix} \rho & \rightarrow & \bar{x}^* \\ \downarrow & & \downarrow \\ \underline{x} & \xrightarrow{\text{so set } \bar{x}^* = \underline{x} + \frac{1}{\rho}} & \text{known unknown.} \end{matrix}$$

Write system(3) and separate unknowns

$$\begin{matrix} \text{first} \\ \text{block} \\ \text{of } \underline{x} \end{matrix} \quad G(\underline{x} + \rho) - A^T (\underline{\lambda} + \frac{1}{\rho}\underline{\lambda}) \underline{\lambda}^* = -\underline{d} \quad \rightarrow$$

~~$$G(\underline{x} + \rho) - A^T (\underline{\lambda} + \frac{1}{\rho}\underline{\lambda}) \underline{\lambda}^* = -\underline{d} - G\underline{x}$$~~

$$\boxed{G(-\rho) + A^T \underline{\lambda}^* = G\underline{x} + \underline{d}}$$

Second block of m eqs.

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(*)

$$\underline{A}\underline{x} + \underline{A}\underline{P} = \underline{b} \Rightarrow \underline{A}\underline{P} + \underline{O}\underline{x^*} = \underline{b} - \underline{A}\underline{x}$$

$$\underline{A}(-\underline{P}) + \underline{O}\underline{x^*} = \underline{A}\underline{x} - \underline{b}.$$

System becomes:

$$\begin{bmatrix} G & A^\top \\ 0 & I \end{bmatrix} \begin{bmatrix} -\underline{P} \\ \underline{x^*} \end{bmatrix} = \begin{bmatrix} \underline{g} \\ \underline{c} \end{bmatrix},$$

in Chf

$$\begin{aligned} \underline{g} &= G\underline{x} + \underline{d} \\ \underline{c} &= A\underline{x} - \underline{b}. \end{aligned}$$

Remark: This system is symmetric! (more efficient to solve than non-sym. ones)

Note: We will not be able to use Cholesky or C.G.

to solve. Why?

(Matrix is sym. but not P.d.)

Reason or diagonal.

even if G is P.d. \rightarrow (not necessary).

- Can use BEP, or specialized methods.

I does not make use of symmetry -

Sections 16.2 $\xrightarrow{\text{K.3 in textbook}}$ projected c.

↳ Null-space method
↳ Schur-complement method.
or Iterative methods:

Note what if f is nonlinear in general?

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$$(4) \begin{cases} \min f(\underline{x}) \\ \text{s.t. } A\underline{x} = \underline{b} \end{cases}$$

Can start with some $(\underline{x}_0, \underline{\lambda}_0)$ then, at iteration k (we know \underline{x}_k), solve

$$\begin{cases} \min \frac{1}{2} \underline{x} \cdot \nabla^2 f(\underline{x}_k) \underline{x} + \nabla f(\underline{x}_k) \cdot \underline{x} + f(\underline{x}_k) \\ \text{s.t. } A\underline{x} = \underline{b} \end{cases}$$

$\leftarrow KKT$ system is:

$$\begin{bmatrix} \nabla^2 f(\underline{x}_k) & A^T \\ A & 0 \end{bmatrix} \begin{bmatrix} -\underline{\lambda}_k \\ \underline{x} \end{bmatrix} = \begin{bmatrix} G\underline{x}_k + \underline{d} \\ A\underline{x}_k - \underline{b} \end{bmatrix}$$

then $\underline{x}_{k+1} = \underline{x}_k + \underline{\lambda}$

$$\underline{\lambda}_{k+1} = \underline{\lambda}_k + \underline{z}$$

needs to be P.S.D,
not the entire $\nabla^2 f(\underline{x}_k)$.

where $\underline{Z} \rightarrow$ matrix
whose columns

form a basis for $\text{Null}(A)$

$$\underline{Z} = [\underline{z}_1, \underline{z}_2, \dots, \underline{z}_P]$$

$$\underline{w} = \alpha_1 \underline{z}_1 + \dots + \alpha_P \underline{z}_P$$

Remark we now know how to solve

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problem with Linear Equality constraint.

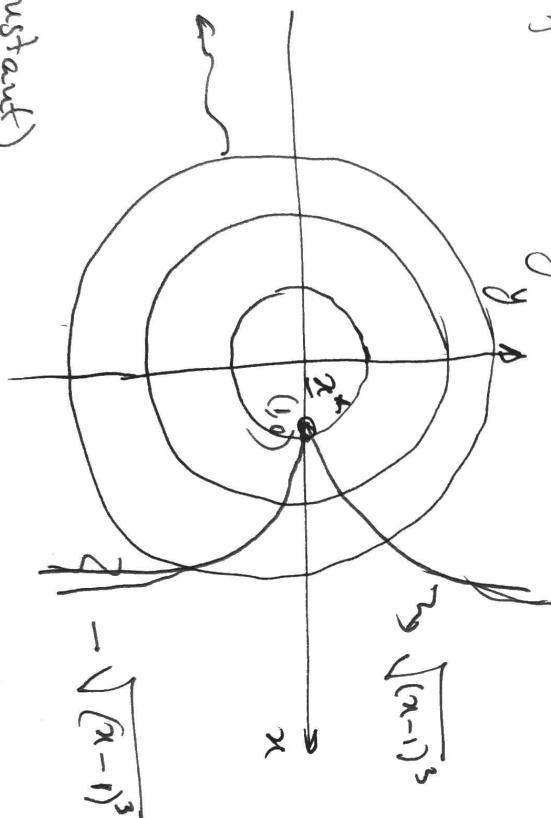
(Q) Can we use elimination of variables method for
Nonlinear Equality constraints?

(A) Not always; It's not safe, so we will not use it.

Example:

$$(5) \begin{cases} \text{min } x^2 + y^2 \\ \text{s.t.} \\ (x-1)^3 = y^2 \end{cases}$$

Graphical method first:
Draw feasible region & level curves:



$$\begin{aligned} y^2 &= (x-1)^3 & x \geq 1 \\ y &= \pm \sqrt{(x-1)^3} \end{aligned}$$

level curves
for $x^2 + y^2$
 $(x^2 + y^2 = \text{constant})$

exact solution: $\underline{x^*} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.

let's try to solve it using elimination of variables:

$$\begin{aligned} \min \quad & x^2 + y^2 = x^2 + (x-1)^3 = x^3 - 3x^2 + 3x - 1 + x^2 = \\ & \uparrow (x-1)^3 \\ & = x^3 - 2x^2 + 3x - 1 \end{aligned}$$

unbounded !!

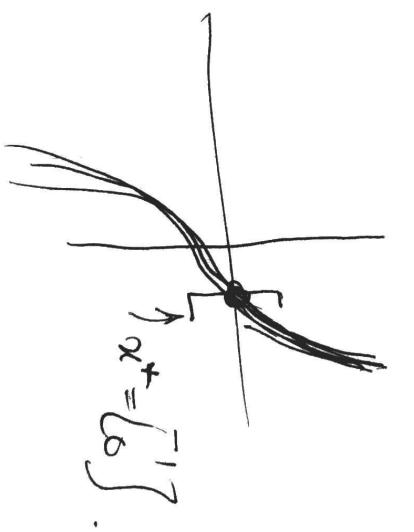
Not solution

What happened? !?

We lost the (implicit) condition
that $\underline{x \geq 1}$ (otherwise the constraint does not
make sense)

So, original problem (5) is equivalent to :

$$\begin{cases} \min & x^2 + (x-1)^3 \\ \text{s.t.} & x - 1 \geq 0 \end{cases}$$



Remark: this type of mistake
is hard to catch if we have many constraints
and n is large; To avoid danger

\rightarrow We will not use elimination of variables
for nonlinear equality constraints.

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Summary :- Showed how to solve
optimization problems
with ~~eg~~ linear equality constraints.

- Next time: Method for solving

$$\begin{cases} \min f(\underline{x}) \\ \text{s.t. } c_i(\underline{x}) \geq 0 \quad i \in I \longrightarrow \text{Linear} \\ c_i(\underline{x}) \leq 0 \quad i \in I \longrightarrow \text{Linear} \end{cases}$$

→ Quadratic

→ a "Quadratic Programming Problem".

added