
 Plan

- 1 Revision: Vectors and matrices (Lecture 1-5)
 - 2 Revision: Optimization problems (Lecture 6-9)
 - 3 Revision: Differential and difference equations (Lecture 10-12)
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 ① Vectors and matrices

Lecture 1-5

Topics:

- linear systems, Gaussian elimination
- vectors, span, linear independence, dimension, base, nullspace, column space
- matrices, determinants / inverse matrix, minors, rank, matrix multiplication, powers
- eigenvalues and eigenvectors, diagonalization, Markov chains, computing A^n
- quadratic forms, definiteness, principal minors, reduced rank criterion, symmetric matrix

Important to know:

- (a) Definitions
- (b) How to solve problems
- (c) Connections / results

Ex: Know defn. of eigenvectors / eigenvalues: $A \cdot \underline{v} = \lambda \cdot \underline{v}$
 λ eigenvalue $\Leftrightarrow (A - \lambda I) \underline{v} = \underline{0}$ has non-trivial soln.
 $\Leftrightarrow |A - \lambda I| = 0$

i) Rank: A
 $n \times n$
 matrix

$rk(A) = \#$ pivot positions
 in A
 = max. order of a
 non-zero minor in A

A $n \times n$: $rk A = n \iff |A| \neq 0$

ii) Span and linear independence:

$\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n\}$ vectors in \mathbb{R}^n

a) $Span(\underline{v}_1, \dots, \underline{v}_n) =$ all linear combinations

$c_1 \underline{v}_1 + c_2 \underline{v}_2 + \dots$

$\dots + c_n \underline{v}_n$

b) the vectors $\{\underline{v}_1, \dots, \underline{v}_n\}$ are lin. independent if none of the vectors are a lin. comb. of the others

$A = (\underline{v}_1 | \underline{v}_2 | \dots | \underline{v}_n)$
 $m \times n$

$Col(A) = span(\underline{v}_1, \dots, \underline{v}_n)$

$Null(A) =$ all soln of $A \cdot \underline{x} = \underline{0}$

$dim Col(A) = rk A$

$dim Null(A) = n - rk A = \#$ free vars

iii) Eigenvalues and eigenvectors:

A
 $n \times n$
 matrix

$A \cdot \underline{u} = \lambda \cdot \underline{u}$

$(A - \lambda I) \underline{u} = \underline{0}$

Solutions:

$\underline{0} \neq \underline{u}$ eigenvector
 λ eigenvalue

a) λ eigenvalue $\iff |A - \lambda I| = 0$
 ch. eqn.

b) solve the linear system $(A - \lambda I) \underline{u} = \underline{0}$

A diagonalizable if

$P^{-1} A P = D$ for a invertible matrix

it exists if:

- a) there are n eigenval.
- b) there are n lin. indep. eigenvectors

P and a diagonal matrix D

A symmetric \Rightarrow A diagonalizable

iv) Definiteness: $A \leftrightarrow f(x) = x^T \cdot A \cdot x$
 Symm. $n \times n$ matrix quadratic form

RRC (reduced rank criteria);

- i) A has $\text{rk } A = r < n$
 - +
 - ii) $D_1, D_2, \dots, D_r > 0$
 (leading principal minors of order up to the rank)
- \Downarrow
- A pos. semidefinite

$f(x) \geq 0$ for all $x \Leftrightarrow f$ is pos. semidef.
 \Uparrow
 $\lambda_1, \lambda_2, \dots, \lambda_n \geq 0$ (eigenval. of A)
 \Uparrow
 $\Delta_1, \Delta_2, \dots, \Delta_n \geq 0$ all principal minors of A

Ex: $A = \begin{pmatrix} 4 & 7 & -1 \\ -2 & 0 & 3 \\ 2 & 7 & 2 \end{pmatrix}$

i) $|A| = -(-2) \cdot \begin{vmatrix} 7 & -1 \\ 7 & 2 \end{vmatrix} + (-1) \cdot 3 \cdot \begin{vmatrix} 4 & 7 \\ 2 & 7 \end{vmatrix} = 2(14-7) - 3(28-14) = 2 \cdot 7 - 3 \cdot 14 = 14 - 42 = -28$

ii) Null(A): $A \cdot x = 0$
 $\begin{pmatrix} 4 & 7 & -1 & | & 0 \\ -2 & 0 & 3 & | & 0 \\ 2 & 7 & 2 & | & 0 \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix} -2 & 0 & 3 & | & 0 \\ 4 & 7 & -1 & | & 0 \\ 2 & 7 & 2 & | & 0 \end{pmatrix} \xrightarrow{R_2 + 2R_1, R_3 + R_1} \begin{pmatrix} -2 & 0 & 3 & | & 0 \\ 0 & 7 & -5 & | & 0 \\ 0 & 7 & 5 & | & 0 \end{pmatrix} \xrightarrow{R_3 - R_2} \begin{pmatrix} -2 & 0 & 3 & | & 0 \\ 0 & 7 & -5 & | & 0 \\ 0 & 0 & 10 & | & 0 \end{pmatrix}$

$\text{rk } A = 2 \rightarrow$
 $\# \text{ free var} = 3 - 2 = 1$
 $n - \text{rk}(A)$

$\begin{pmatrix} 2x + 7y - 2z = 0 \\ 7y - 5z = 0 \\ z \text{ free} \end{pmatrix}$
 echelon form

$7y = 5z \Rightarrow y = \frac{5}{7}z$

Solutions: $\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 3z/2 \\ -5z/2 \\ z \end{pmatrix} = z \cdot \begin{pmatrix} 3/2 \\ -5/2 \\ 1 \end{pmatrix} = \frac{z}{2} \cdot \begin{pmatrix} 3 \\ -5 \\ 2 \end{pmatrix}$

$2x + 7y - 2z = 0$
 $2x = -7 \left(\frac{5}{7}z \right) - 2z = -5z - 2z = -7z$
 $x = \frac{-7z}{2} = -\frac{7}{2}z$

Base of Null(A): $\{w_1\}, w_1 = \begin{pmatrix} 3 \\ -5 \\ 2 \end{pmatrix}$
 $\dim \text{Null}(A) = 1$ or $\{w_2\}, w_2 = \begin{pmatrix} 3/2 \\ -5/2 \\ 1 \end{pmatrix}$

iii) Eigenvalues (eigenvectors) of A:

$$A = \begin{pmatrix} 4 & 7 & -1 \\ -2 & 0 & 3 \\ 2 & 7 & 2 \end{pmatrix}$$

Eigenvalues:

$$|A - \lambda I| = 0$$

$$\begin{vmatrix} 4-\lambda & 7 & -1 \\ -2 & -\lambda & 3 \\ 2 & 7 & 2-\lambda \end{vmatrix} = 0$$

$$(4-\lambda) \cdot (-\lambda(2-\lambda) - 21) - 7(-2(2-\lambda) - 6) - 1(-14 + 2\lambda) = 0$$

$$(4-\lambda)(\lambda^2 - 2\lambda - 21) - 7(2\lambda - 10) - 2(\lambda - 7) = 0$$

$$-\lambda^3 + 4\lambda^2 + 2\lambda^2 - 8\lambda + 21\lambda - 84 - 14\lambda + 70 - 2\lambda + 14 = 0$$

$$-\lambda^3 + 6\lambda^2 - 3\lambda = 0$$

$$-\lambda(\lambda^2 - 6\lambda + 3) = 0$$

$$\lambda = 0 \text{ or } \lambda^2 - 6\lambda + 3 = 0$$

$$\lambda = \frac{6 \pm \sqrt{36 - 4 \cdot 3}}{2} = 3 \pm \frac{\sqrt{24}}{2}$$

$$= 3 \pm \frac{\sqrt{4} \cdot \sqrt{6}}{2} = 3 \pm \sqrt{6}$$

$$\lambda_1 = 0 \quad \lambda_2 = 3 + \sqrt{6} \quad \lambda_3 = 3 - \sqrt{6}$$

$\lambda = 0$ is an eigenvalue since $|A| = 0$

Eigenvectors:

$E_0: \lambda = 0$

$$\text{Null}(A - 0 \cdot I) = \text{Null}(A) = c \cdot \begin{pmatrix} 3 \\ -5 \\ 2 \end{pmatrix}$$

$$\text{Basis of } E_0: \left\{ \begin{pmatrix} 3 \\ -5 \\ 2 \end{pmatrix} \right\}$$

Ex:

$$A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & -1 \\ 0 & -1 & 1 \end{pmatrix}$$

Symm 3×3 matrix

$$D_1 = 1 > 0$$

$$D_2 = 2 - 1 = 1 > 0$$

$$|A| = D_3 = -(-1) \cdot (-1 - 0) + 1 \cdot 1 = -1 + 1 = 0$$

pos. semidefn. or indefn.

$$f = x^2 + 2xy + 2y^2 - 2yz + z^2$$

Alt I:

$$\Delta_1 = 1, 2, 1 > 0$$

$$\Delta_2 = 1, 1, 1 > 0$$

$$\Delta_3 = 0$$

all $\Delta_i \geq 0$

\Leftrightarrow A pos. semidefn.

Alt 2:

RRC

$$rk A = 2 < 3$$

$$r = 2, n = 3$$

$$D_1, D_2 > 0$$

\Leftrightarrow A pos. semid.

② Optimization problems

Lecture 6-9

- partial derivatives, stationary pts, Hessian, convex/concave fn., second derivative test
- compact sets, extreme value theorem, Lagrange / Kuhn-Tucker problems, FOC+C / FOC+C+CSC, NOCC, SOC
- envelope theorems

* unconstrained optimization:
max/min $f(x_1, \dots, x_n)$

Env. thm: max/min $f(x; a)$
 $\frac{df^*(a)}{da} = f'_a(x^*(a))$

FOC: $f'_{x_1} = \dots = f'_{x_n} = 0 \rightarrow$ stationary pts on the sol'n.

f convex $\Leftrightarrow H(f)(x)$ pos. semidef. for all x
 f concave \Leftrightarrow - - - - - neg. semidef. - - - - -

f concave \Rightarrow any stationary pt. is global max

classification of all stationary pts

local min, local max, saddle pt.

Second derivative test

x^* stat. pt and $H(f)(x^*)$ pos. def., $\Rightarrow x^*$ local min

* Constrained optimization:

Lagrange: max/min $f(x)$ wh $\left\{ \begin{array}{l} g_1(x) = a_1 \\ \vdots \\ g_m(x) = a_m \end{array} \right.$

Kuhn-Tucker: max $f(x)$ wh $\left\{ \begin{array}{l} g_1(x) \leq a_1 \\ \vdots \\ g_m(x) \leq a_m \end{array} \right.$
(std. form)

Lagrange: $L = f(x) - \lambda_1 \cdot g_1(x) - \lambda_2 \cdot g_2(x) - \dots - \lambda_m \cdot g_m(x)$

FOC: $\left\{ \begin{array}{l} h'_{x_1} = 0 \\ \vdots \\ h'_{x_n} = 0 \end{array} \right.$ \leq $\left\{ \begin{array}{l} g_1(x) = a_1 \\ \vdots \\ g_m(x) = a_m \end{array} \right.$

FOC+C = Lagrange condition, solution = candidate for max/min

NDCQ: If J is maximal when $J = \begin{pmatrix} \partial g_1 / \partial x_1 & \dots & \partial g_1 / \partial x_n \\ \vdots & & \vdots \\ \partial g_m / \partial x_1 & \dots & \partial g_m / \partial x_n \end{pmatrix}$
adn pts that fails NDCQ \Rightarrow exceptional candidates for max/min

i) D \Rightarrow there is a maximum away from candidate pts.
 set of adn pts bounded \Rightarrow EVT

(x^*, λ^*) cand. pt. that satisfy FOC+C:

ii) SOC: $h(x) = L(x; \lambda^*)$ concave/conv. $\Rightarrow x^*$ is max

Envelope thm: max/min $f(x; a)$ when $\begin{cases} g_1(x; a) = 0 \\ \vdots \\ g_m(x; a) = 0 \end{cases}$

$$\frac{df^*(a)}{da} = L'_\lambda(x^*(a); \lambda^*(a))$$

Kuhn-Tucker case: max $f(x)$ when $\begin{cases} g_1(x) \leq a_1 \\ \vdots \\ g_m(x) \leq a_m \end{cases}$

Candidate pts for max: $h = f(x) - \lambda_1 g_1(x) - \dots - \lambda_m g_m(x)$

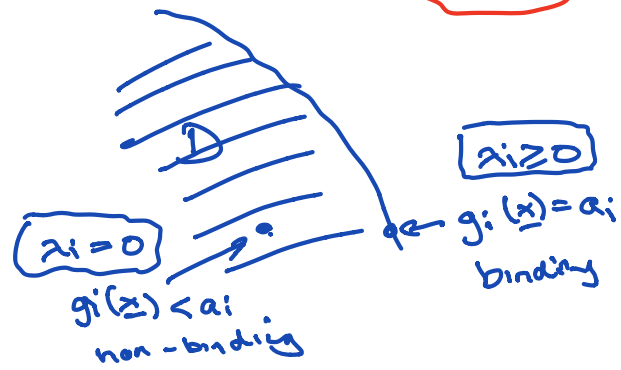
FOC: $\begin{cases} L'_{x_1} = 0 \\ \vdots \\ L'_{x_n} = 0 \end{cases}$

C: $\begin{cases} g_1(x) \leq a_1 \\ \vdots \\ g_m(x) \leq a_m \end{cases}$

CSC: $\begin{cases} \lambda_1, \dots, \lambda_m \geq 0 \\ \text{and} \\ \lambda_i \cdot (g_i(x) - a_i) = 0 \\ \vdots \\ \lambda_m \cdot (g_m(x) - a_m) = 0 \end{cases}$

FOC + C + CSC = Kuhn-Tucker conditions

Complementary slackness condition



Ex: $\max f = 2x^2 - 4y^2 - 2z^2$ (where $\underbrace{x^4 + y^4 + z^4}_{g(x,y,z)} \leq \underbrace{16}_a$)

a) Unconstrained pb: $\max f$

$$\left. \begin{aligned} f'_x = 4x &= 0 \\ f'_y = -8y &= 0 \\ f'_z = -4z &= 0 \end{aligned} \right\} \text{FOC} \quad \begin{aligned} x &= 0 \\ y &= 0 \\ z &= 0 \end{aligned}$$

Stab. pts: $(x,y,z) = (0,0,0)$

$$H(f) = \begin{pmatrix} 4 & 0 & 0 \\ 0 & -8 & 0 \\ 0 & 0 & -4 \end{pmatrix}$$

matrix

$$D_1 = 4 \\ D_2 = -32$$

$\Rightarrow (0,0,0)$ saddle pt
no max for f

b) Kuhn-Tucker: $L = 2x^2 - 4y^2 - 2z^2 - \lambda(x^4 + y^4 + z^4)$
(std form)

FOC: $\begin{aligned} L'_x &= 4x - \lambda \cdot 4x^3 = 0 \\ L'_y &= -8y - \lambda \cdot 4y^3 = 0 \\ L'_z &= -4z - \lambda \cdot 4z^3 = 0 \end{aligned}$

c: $x^4 + y^4 + z^4 \leq 16$

CSC: $\lambda \geq 0$ and ($\lambda = 0$ if $x^4 + y^4 + z^4 < 16$)

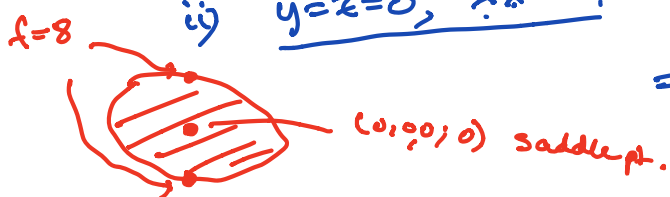
FOC: $\begin{aligned} 4x(1 - \lambda x^2) &= 0 \\ 4y(-2 - \lambda y^2) &= 0 \\ 4z(-1 - \lambda z^2) &= 0 \end{aligned}$

$\begin{aligned} x=0 & \text{ or } \lambda x^2=1 \\ y=0 & \text{ or } \lambda y^2=-2 \\ z=0 & \text{ or } \lambda z^2=-1 \end{aligned}$

$\hat{=} y=z=0$ and ($x=0$ or $\lambda x^2=1$)

i) $x=y=z=0$: $c: 0 < 16$ ok nonbinding $\Rightarrow \lambda=0$
 $\Rightarrow (0,0,0;0) \quad f=0$

ii) $y=z=0, \lambda x^2=1$: $x^2 = 1/\lambda \quad (\lambda > 0)$ $\lambda > 0 \Rightarrow x^4 + y^4 + z^4 = 16$
 $\Rightarrow (\pm 2, 0, 0; 1/4) \quad f=8$
 $x^4 + 0 + 0 = 16$
 $x^4 = 16$
 $x^2 = 4 \Rightarrow \lambda = 1/4$
 $x = \pm 2$



Conclusion:

$$x^4 + y^4 + z^4 \leq 16$$

bounded set

 \Rightarrow
EUT

there is a max

$$\begin{cases} -2 \leq x \leq 2 \\ -2 \leq y \leq 2 \\ -2 \leq z \leq 2 \end{cases}$$

$$f_{\max} = 8$$

since NPGC holds at the boundary.

$$J = (4x^3 \ 4y^3 \ 4z^3)$$

$$\text{rk } J = 0 \iff x=y=z=0$$

not on the boundary

③ Differential and difference equations Lecture 10-12

i) First order differential equations:

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

- separable $y' = f(t) \cdot g(y)$
- linear $y' + a(t)y = b(t)$
- exact $P + Q \cdot y' = 0$
where
 $P = h(t)$
 $Q = h'(y)$

Autonomous case: $y' = F(y)$

- equilibrium states, stability

ii) Second order differential equations:

$$y'' + ay' + by = f(t)$$

- superposition: $y = y_h + y_p$

$$- y_h: \boxed{r^2 + ar + b = 0}$$

$$r_1 \neq r_2: y_h = C_1 e^{r_1 t} + C_2 e^{r_2 t}$$

$$r_1 = r_2: y_h = C_1 e^{r_1 t} + C_2 t \cdot e^{r_1 t}$$

iii) Systems of linear d.f. eqn.

$$\begin{pmatrix} y_1' \\ \vdots \\ y_n' \end{pmatrix} = A \cdot \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

A diagonalizable

$$y = C_1 \underline{v}_1 e^{\lambda_1 t} + \dots + C_n \underline{v}_n e^{\lambda_n t}$$

iv) Difference equations:

$$y_{t+1} + a y_t = f_t$$

$$y_{t+2} + a y_{t+1} + b y_t = f_t$$

$$\underline{y}_{t+1} = A \cdot \underline{y}_t$$

} first/second order linear
diff. eqn.
⇒ superposition

system of linear difference
eqn.

|| A diagonalizable

$$\underline{y}_t = c_1 \cdot \underline{u}_1 \cdot \lambda_1^t + \dots$$

$$\dots + c_n \cdot \underline{u}_n \cdot \lambda_n^t$$