

LECTURE 5

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GRA 6035

MATHEMATICS

Plan:

- ① Diagonalization and Markov chains
- ② Quadratic forms, definiteness and principal minors.

Reading:

[FES] 6.2 (Ex 3),
23.1 (Ex 23.4),
23.6, 13.1-13.5,
16.1-16.4, 23.8

Review:

A $n \times n$ -matrix is diagonalizable if there is an invertible matrix P and a diagonal matrix D s.t.

$$P^{-1}AP = D$$

A diagonalizable \iff $\left\{ \begin{array}{l} \text{i) there are } n \text{ eigenvalues of } A \\ \lambda_1, \lambda_2, \dots, \lambda_n \text{ (counted with} \\ \text{multiplicity)} \end{array} \right.$ and $\left\{ \begin{array}{l} \text{ii) there are } n \text{ linearly} \\ \text{independent eigenvectors} \\ \underline{v_1}, \underline{v_2}, \dots, \underline{v_n} \end{array} \right.$

$$D = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{pmatrix}$$

$$P = \left(\begin{array}{c|c|c} \underline{v_1} & \underline{v_2} & \dots & \underline{v_n} \end{array} \right)$$

$\underline{v_i}$ eigenvector for λ_i

Fact: If λ is an eigenvalue with multiplicity m , then

$$1 \leq \underbrace{\dim E_\lambda}_{\text{the \# of}} \leq m$$

lin. independent eigenvectors for λ

Condition ii) \iff For each eigenvalue with multiplicity m , we have $\dim E_{\lambda} = m$.

Facts:

i) If A is symmetric, then A is diagonalizable.
($A^T = A$)

ii) If A has n distinct eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ (all multiplicities are 1), then A is diagonalizable.

If not, we must compute all eigenvalues / \sim vectors and check i) - ii).

① Applications: Markov chains

E_t : employment

State vector

$$\underline{x}_t = \begin{pmatrix} e_t \\ u_t \end{pmatrix}$$

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

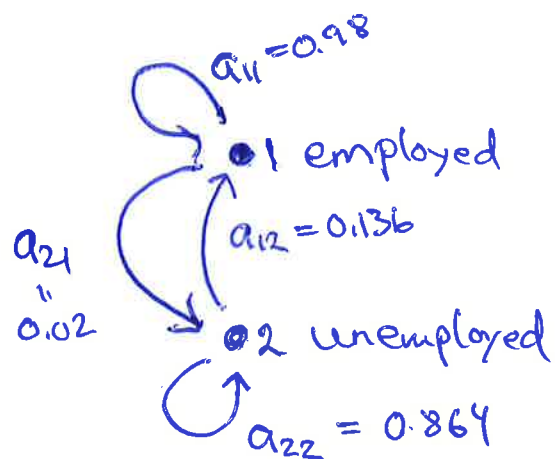
2x2-
matrix

$$\begin{cases} e_t \geq 0, u_t \geq 0 \\ e_t + u_t = 1 \end{cases}$$

$a_{ij} \geq 0$
column sums = 1

transition matrix

$$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = \begin{pmatrix} 0.98 & 0.136 \\ 0.02 & 0.864 \end{pmatrix}$$



a_{ij} : transitions from node j to node i

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

$$\begin{pmatrix} e_{t+1} \\ u_{t+1} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} e_t \\ u_t \end{pmatrix}$$

$$e_{t+1} = a_{11} e_t + a_{12} u_t$$

$$u_{t+1} = a_{21} e_t + a_{22} u_t$$

$$\underline{x}_0 \rightarrow \underline{x}_1 = A \cdot \underline{x}_0 \rightarrow \underline{x}_2 = A \cdot \underline{x}_1 \rightarrow \dots \rightarrow \underline{x}_n = A^n \cdot \underline{x}_0$$

Target: Find $\lim_{n \rightarrow \infty} A^n \cdot x_0$

① Find eigenvalues and eigenvectors of A

$$A = \begin{pmatrix} 0.98 & 0.136 \\ 0.02 & 0.864 \end{pmatrix}$$

$$\begin{vmatrix} 0.98 - \lambda & 0.136 \\ 0.02 & 0.864 - \lambda \end{vmatrix} = 0$$

$$0.98 \cdot 0.864 - 0.02 \cdot 0.136$$

$$\lambda^2 - 1.844\lambda + 0.844 = 0$$

$$\lambda_1 = 1, \lambda_2 = 0.844$$

$\lambda_1 \neq \lambda_2 : A$ is diagonalizable

$$D = \begin{pmatrix} 1 & 0 \\ 0 & 0.844 \end{pmatrix}$$

$$D^n = \begin{pmatrix} 1 & 0 \\ 0 & 0.844^n \end{pmatrix}$$

\downarrow $n \rightarrow \infty$

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$P^{-1}AP = D \Rightarrow A = PDP^{-1}$$

$$A^n = P \cdot D^n \cdot P^{-1}$$

$$\underline{A=1}: \begin{pmatrix} -0.02 & 0.136 \\ 0.02 & -0.136 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$0.02x = 0.136y \rightarrow E_1: \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 6.8y \\ y \end{pmatrix} = y \cdot \begin{pmatrix} 6.8 \\ 1 \end{pmatrix}$$

$$\underline{\lambda = 0.844}: \begin{pmatrix} 0.136 & 0.136 \\ -0.02 & -0.02 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{matrix} x+y=0 \\ x=-y \end{matrix} \quad E_2: \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ y \end{pmatrix} = y \cdot \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$P = \begin{pmatrix} 68 & -1 \\ 10 & 1 \end{pmatrix} \quad P^{-1} = \frac{1}{78} \begin{pmatrix} 1 & 1 \\ -10 & 68 \end{pmatrix}$$

$$A^n = P \cdot D^n \cdot P^{-1} = \begin{pmatrix} 68 & -1 \\ 10 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & 0.844^n \end{pmatrix} \cdot \frac{1}{78} \begin{pmatrix} 1 & 1 \\ -10 & 68 \end{pmatrix}$$

↓ $n \rightarrow \infty$

$$\begin{pmatrix} 68 & -1 \\ 10 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \cdot \frac{1}{78} \begin{pmatrix} 1 & 1 \\ -10 & 68 \end{pmatrix}$$

$$= \frac{1}{78} \begin{pmatrix} 68 & 0 \\ 10 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 \\ -10 & 68 \end{pmatrix}$$

$$= \frac{1}{78} \begin{pmatrix} 68 & 68 \\ 10 & 10 \end{pmatrix} = \begin{pmatrix} 68/78 & 68/78 \\ 10/78 & 10/78 \end{pmatrix}$$

$$\lim_{n \rightarrow \infty} (A^n \cdot \underline{x}_0) = \begin{pmatrix} 68/78 & 68/78 \\ 10/78 & 10/78 \end{pmatrix} \cdot \begin{pmatrix} e_0 \\ u_0 \end{pmatrix} = \underline{\underline{\begin{pmatrix} 68/78 \\ 10/78 \end{pmatrix}}} \approx \begin{pmatrix} 0.872 \\ 0.128 \end{pmatrix}$$

$$e_0 \cdot \begin{pmatrix} 68/78 \\ 10/78 \end{pmatrix} + u_0 \cdot \begin{pmatrix} 68/78 \\ 10/78 \end{pmatrix}$$

Equilibrium state: $\underline{x} = \lim_{n \rightarrow \infty} (A^n \cdot \underline{x}_0)$

employment/
unemployment
rate in the long
run.

$$E_1 \approx y \cdot \begin{pmatrix} 68 \\ 1 \end{pmatrix}$$

State vector:

$$y \cdot 68 + y \cdot 1 = 1$$

$$y \cdot 78 = 1 \Rightarrow y = \frac{1}{78}$$

$$\underline{v} = \begin{pmatrix} 68/78 \\ 1/78 \end{pmatrix} \approx \underline{\underline{\begin{pmatrix} 68/78 \\ 10/78 \end{pmatrix}}}$$

Results for Markov chains

n categories

state vector
(n-vector)

$$\underline{x}_t = \begin{pmatrix} x_t(1) \\ x_t(2) \\ \vdots \\ x_t(n) \end{pmatrix}$$



$$\begin{cases} x_t(i) \geq 0 \\ \sum_i x_t(i) = 1 \end{cases}$$

transition matrix $A = (a_{ij})$
n x n-matrix

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

$$\begin{cases} a_{ij} \geq 0 \text{ for all } i, j \\ \text{col. sums} = 1 \end{cases}$$

The Markov chain is called regular if you can get to any node from any other node.

$$a_{ij} > 0 \text{ for all } i, j \implies A \text{ is regular}$$

Result: Assume that the Markov chain is regular.

Then we have:

i) $\lambda = 1$ is an eigenvalue, and it is dominant
($|\lambda| < 1$ for all other eigenvalues)

ii) there is a unique state vector among the eigenvectors for $\lambda = 1$, \underline{v} .

iii) For any initial state \underline{x}_0 , we have that

$$\lim_{t \rightarrow \infty} \underline{x}_t = \lim_{t \rightarrow \infty} A^t \cdot \underline{x}_0 = \underline{v}$$

Markov process

Ex: Families are classified as U (urban), S (suburban) and R (rural). At time $t=n$ (after n years), the share of families in these groups can be described by the state vector

$$\underline{V}_n = \begin{pmatrix} U_n \\ S_n \\ R_n \end{pmatrix} \begin{cases} U_n \geq 0 \\ S_n \geq 0 \\ R_n \geq 0 \end{cases}, \quad U_n + S_n + R_n = 1$$

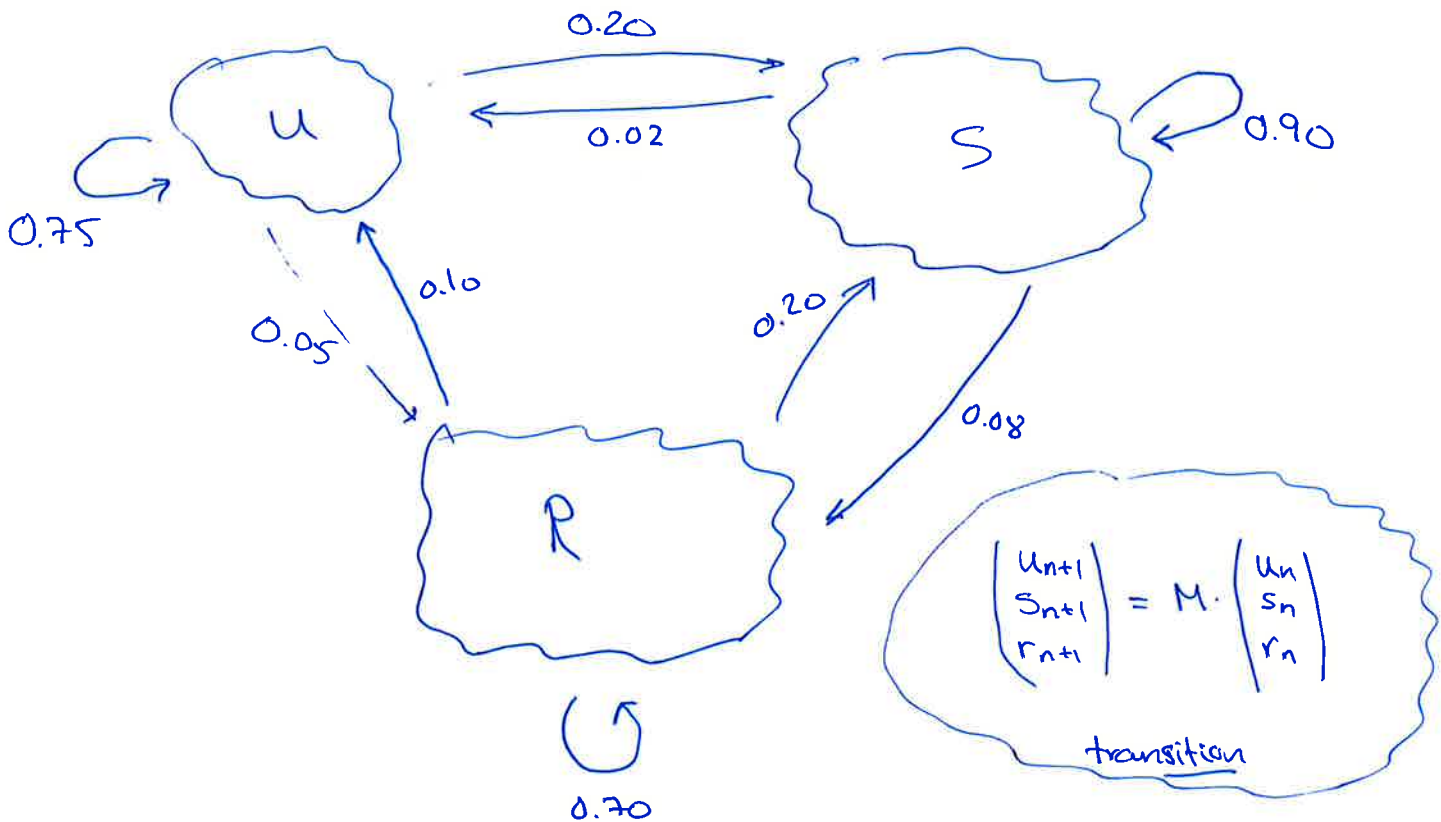
Ex:

$$\underline{V} = \begin{pmatrix} 0.8 \\ 0.1 \\ 0.1 \end{pmatrix}$$

From year n to year $n+1$, the change in the shares are given by a transition matrix or Markov matrix

$$M = \begin{pmatrix} 0.75 & 0.02 & 0.10 \\ 0.20 & 0.90 & 0.20 \\ 0.05 & 0.08 & 0.70 \end{pmatrix} \begin{cases} m_{ij} \geq 0 \\ \text{each column has sum } 1 \end{cases}$$

It can be described graphically as follows:



Markov process:

$$\underline{V}_0 = \begin{pmatrix} U_0 \\ S_0 \\ R_0 \end{pmatrix} \xrightarrow{\text{start}} \underline{V}_1 = M \cdot \underline{V}_0 \xrightarrow{\text{---}} \underline{V}_2 = M \cdot \underline{V}_1 = M^2 \cdot \underline{V}_0 \xrightarrow{\text{---}} \dots \xrightarrow{\text{---}} \underline{V}_n = M^n \cdot \underline{V}_0$$

The Markov process is regular if $m_{ij} > 0$ for all i, j . We assume that this the case. The following holds for all regular Markov processes:

Fact: i) $\lambda = 1$ is an eigenvalue of M , and there is a unique eigenvector \underline{v} with eigenvalue $\lambda = 1$ that is a state vector (i.e. $\underline{v} = (v_i)$ with $v_i \geq 0$, $v_1 + \dots + v_k = 1$)

ii) $\lim_{n \rightarrow \infty} M^n \cdot \underline{v}_0 = \underline{v}$ and $\lim_{n \rightarrow \infty} M^n = \begin{pmatrix} \underline{v} & | & \underline{v}_1 & \dots & | & \underline{v} \end{pmatrix}$

Ex: $M = \begin{pmatrix} 0.75 & 0.02 & 0.10 \\ 0.20 & 0.90 & 0.20 \\ 0.05 & 0.08 & 0.70 \end{pmatrix}$

$D = \begin{pmatrix} 1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{pmatrix}$

$\lambda = 1$: $\begin{pmatrix} -0.25 & 0.02 & 0.10 \\ 0.20 & -0.10 & 0.20 \\ 0.05 & 0.08 & -0.30 \end{pmatrix} \rightarrow \begin{pmatrix} 5 & 8 & -30 \\ -25 & 2 & 10 \\ 20 & -10 & 20 \end{pmatrix} \rightarrow \begin{pmatrix} 5 & 8 & -30 \\ 0 & 42 & -140 \\ 0 & -42 & 140 \end{pmatrix}$

$\rightarrow \begin{pmatrix} 5 & 8 & -30 \\ 0 & 42 & -140 \\ 0 & 0 & 0 \end{pmatrix}$

$5x + 8y - 30z = 0$
 $42y - 140z = 0$
 z free

$y = \frac{140z}{42} = \frac{10}{3}z$

$5x = 30z - 8 \cdot \frac{10}{3}z = \frac{90 - 80}{3}z$

$x = \frac{2}{3}z$

$\frac{2}{3}z + \frac{10}{3}z + z = 1$

$5z = 1$

$z = 1/5$

$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2/3 \cdot z \\ 10/3 \cdot z \\ z \end{pmatrix} = \frac{z}{3} \cdot \begin{pmatrix} 2 \\ 10 \\ 3 \end{pmatrix} \Rightarrow \underline{v} = \begin{pmatrix} 2/15 \\ 10/15 \\ 3/15 \end{pmatrix}$ (with $z = 1/5$)

Conclusion: As $n \rightarrow \infty$ (in the long run) $u = 2/15 \approx 13.3\%$ of families are urban, $s = 10/15 \approx 66.7\%$ are suburban, and $r = 3/15 = 20\%$ are rural.

Check: Compute M^{10} , M^{50} , M^{100} using Wolfram Alpha or other software.

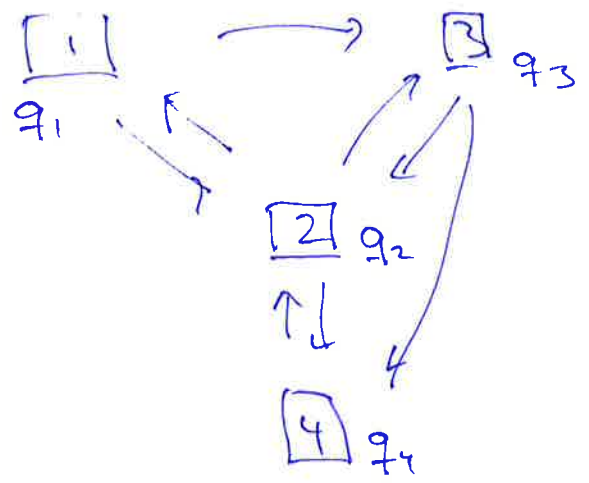
It is also possible to compute M^n as

$M^n = P \cdot \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{pmatrix}^n \cdot P^{-1} \approx P \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \cdot P^{-1}$

↑ since $\lambda_1 = 1, \lambda_2, \lambda_3 < 1$

Ex: Google's PageRank algorithm

Webpages 1-4 with links: r_i is the ranking no.



$$r_1 = K \cdot (r_2)$$

$$r_2 = K \cdot (r_1 + r_3 + r_4)$$

$$r_3 = K \cdot (r_1 + r_2)$$

$$r_4 = K \cdot (r_2 + r_3)$$

$$\underline{r} = \begin{pmatrix} r_1 \\ r_2 \\ r_3 \\ r_4 \end{pmatrix}$$

$$\underline{r} = K \cdot \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix} \underline{r}$$

$$\lambda \cdot \underline{r} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix} \underline{r}$$

Eigenvectors for M with $\lambda=1$,

$$\underline{v} = \begin{pmatrix} 4/29 \\ 12/29 \\ 6/29 \\ 7/29 \end{pmatrix} \quad \begin{matrix} r_1 = 4/29 \\ r_2 = 12/29 \\ r_3 = 6/29 \\ r_4 = 7/29 \end{matrix}$$

} replace with M

$$M = \begin{pmatrix} 0 & 1/3 & 0 & 0 \\ 1/2 & 0 & 1/2 & 1 \\ 1/2 & 1/3 & 0 & 0 \\ 0 & 1/3 & 1/2 & 0 \end{pmatrix}$$

$M^n \rightarrow (\underline{v} | \underline{v}_2 | \dots | \underline{v}_n)$
as $n \rightarrow \infty$

efficient computation

② Quadratic forms

Defn: A quadratic form in n variables x_1, \dots, x_n is a polynomial function where each term has degree two.

Ex:

$n=1$: $f(x) = ax^2$ ($x=x_1$)

$n=2$: $f(x,y) = ax^2 + bxy + cy^2$ ($x=x_1, y=x_2$)

$n=3$: ~~$f(x,y,z) =$~~

$$f(x_1, x_2, x_3) = c_{11}x_1^2 + c_{12}x_1x_2 + c_{13}x_1x_3 + c_{22}x_2^2 + c_{23}x_2x_3 + c_{33}x_3^2$$

general n : $f(x_1, \dots, x_n) =$

$$c_{11}x_1^2 + c_{12}x_1x_2 + \dots + c_{1n}x_1x_n + c_{22}x_2^2 + \dots + c_{2n}x_2x_n + \dots + c_{nn}x_n^2$$

Matrix form: $\underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$

$\underline{x}^T \cdot A \cdot \underline{x}$: Ex: $A = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$ gives:

$$\begin{aligned} & (x_1 \ x_2) \cdot \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \\ &= \begin{pmatrix} x_1 + 2x_2 & 2x_1 + x_2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \\ &= x_1^2 + 2x_2x_1 + 2x_1x_2 + x_2^2 \\ &= x_1^2 + 4x_1x_2 + x_2^2 \end{aligned}$$

$a_{ii} \rightsquigarrow a_{ii}x_i^2$
 $a_{ij}, a_{ji} \rightsquigarrow a_{ij}x_ix_j$ ($i \neq j$)

Note:

Any quadratic form $f(x_1, \dots, x_n)$ can be written in matrix form

$$f(x) = \underline{x}^T A \underline{x}$$

There is a unique matrix A with A symmetric.

Ex:

$$f(x_1, x_2, x_3) = \underline{x_1^2} + 7x_1x_2 - x_1x_3 + \underline{x_2^2} + 4x_2x_3 - \underline{x_3^2}$$

quadratic form

$$A = \begin{pmatrix} 1 & 7/2 & -1/2 \\ 7/2 & 1 & 2 \\ -1/2 & 2 & -1 \end{pmatrix} = \begin{pmatrix} 1 & 7/2 & -1/2 \\ 7/2 & 1 & 2 \\ -1/2 & 2 & -1 \end{pmatrix}$$

Ex:

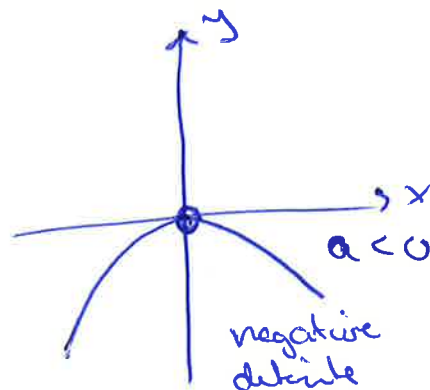
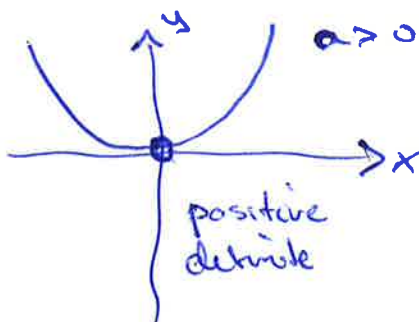
$$A = \begin{pmatrix} 1 & 2 & 0 \\ 2 & -1 & 4 \\ 0 & 4 & 3 \end{pmatrix}$$

Symmetric
matrix

$$f(x_1, x_2, x_3) = x_1^2 + 4x_1x_2 - x_2^2 + 8x_2x_3 + 3x_3^2$$

Ex:

$$n=1: f(x) = ax^2$$



Defn: $f(\underline{x})$ quadratic form in n variables
A symmetric $n \times n$ -matrix

A/f is positive definite if $f(\underline{x}) > 0$ for all $\underline{x} \neq \underline{0}$
 " positive semidefinite " $f(\underline{x}) \geq 0$ for all \underline{x}
 " negative definite " $f(\underline{x}) < 0$ for all $\underline{x} \neq \underline{0}$
 " negative semidefinite " $f(\underline{x}) \leq 0$ for all \underline{x}
 " indefinite " $f(\underline{x})$ takes both positive and negative values

(i.e. none of the above holds)

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

Ex: $f(x,y) = x^2 + 4xy + 3y^2$

$$A = \begin{pmatrix} 1 & 2 \\ 2 & 3 \end{pmatrix}$$

Ex: $f(x,y) = x^2 + 4y^2$
pos. defn.

$f(x,y) > 0$ for all $(x,y) \neq (0,0)$

Result:

$f(\underline{x})$ quadr. form with symm. $n \times n$ -matrix A

If A has eigenvalues $\lambda_1, \dots, \lambda_n$, then:

$\lambda_1, \dots, \lambda_n > 0 \iff$ positive definite

$\lambda_1, \dots, \lambda_n \geq 0 \iff$ positive semidefinite

$\lambda_1, \dots, \lambda_n < 0 \iff$ negative definite

$\lambda_1, \dots, \lambda_n \leq 0 \iff$ negative semidefinite

Both positive and negative eigenvalues \iff indefinite

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

$$\lambda^2 - 4\lambda - 1 = 0$$

$$\lambda = \frac{4 \pm \sqrt{16+4}}{2}$$

$$= 2 \pm \sqrt{5}$$

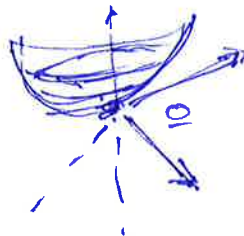
$$\lambda_1 = 2 + \sqrt{5} > 0$$

$$\lambda_2 = 2 - \sqrt{5} < 0$$

indefinite

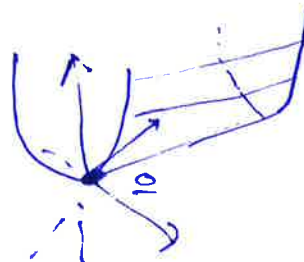
Interpretation: Graph of f

f pos. defn. :



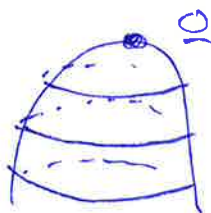
$x=0$ global min.

f pos. semidefn.



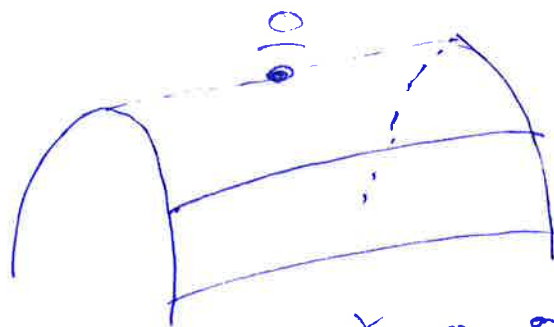
$x=0$ global min
(and some other pts)

f neg. defn.



$x=0$ global max.

f neg. semidefn.



$x=0$ global max
(and some other pts).

f indefinite

$x=0$ Saddle pt.

Ex: $f(x) = 3x_1^2 + 2x_1x_3 - x_2^2 + x_3^2$

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

can compute eigenvalues

Principal minors:

Leading principal minors:

D_i : leading principal minor of order i

$$D_i = M_{1,2,\dots,i} \quad (\text{first } i \text{ rows, } i \text{ cols})$$

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

$$D_1 = 3$$

$$D_2 = 3 \cdot (-1) - 0^2 = -3$$

$$D_3 = |A| = 1 \cdot 1 + 1 \cdot (-3) = -2$$

Result:

$D_1 > 0, D_2 > 0, \dots, D_n > 0 \iff$ positive definite
 $D_1 < 0, D_2 > 0, D_3 < 0, \dots \iff$ negative definite

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

$$D_1 = -1 = \lambda_1$$

$$D_2 = 2 = \lambda_1 \lambda_2$$

$$D_3 = -8 = \lambda_1 \lambda_2 \lambda_3$$

Result:

pos. semidefinite $\leftrightarrow \Delta_1, \Delta_2, \dots, \Delta_n \geq 0$
 neg. semidefinite $\leftrightarrow \Delta_1 \leq 0, \Delta_2 \geq 0, \dots$
 indefinte \leftrightarrow all other cases

Δ_i : principal minor of order i
 = minor of order i where i choose the same rows as columns

Note: Many Δ_i 's for each order i , the first of them is D_i .

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

$D_1 = 1$
 $D_2 = -1 \iff \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix}$
 $D_3 = |A| = 4 \cdot (-14) + 1 \cdot (-9) + 2 \cdot (-1)$
 $= -56 - 9 - 2 = -67$

Choose one row and the same col. $\rightarrow \Delta_1: M_{1,1} = \underline{1} \quad M_{2,2} = \underline{3} \quad M_{3,3} = \underline{2}$

two rows / same two cols. $\rightarrow \Delta_2: M_{2,1,2} = \underline{-1} \quad M_{1,3,1,3} = \begin{vmatrix} 1 & 4 \\ 4 & 2 \end{vmatrix} = \underline{-14}$

$\rightarrow \Delta_3: M_{2,3,2,3} = \begin{vmatrix} 3 & -1 \\ -1 & 2 \end{vmatrix} = \underline{5}$

all three rows and cols $\rightarrow \Delta_3: M_{2,3,2,3} = |A| = -67$

A is indefinte

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

Ex: $f = x^2 + 2xy + y^2 + z^2$

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

$$\left. \begin{aligned} D_1 &= 1 \\ D_2 &= 0 \\ D_3 &= 0 \end{aligned} \right\}$$

can be positive semidefinite,
but must compute all principal
minors to tell (it may also be
indefinite).

$$\begin{array}{l} \Delta_1 = 1, 1, 1 \\ \Delta_2 = 0, 1, 1 \\ \Delta_3 = 0 \end{array}$$

- choose: 1,1 2,2 3,3
- choose: 1,2, 1,3, 1,3, 2,3, 2,3
- choose: 1,2,3, 1,2,3

$$\Delta_1, \Delta_2, \Delta_3 \geq 0 \Rightarrow \text{positive semidefinite}$$