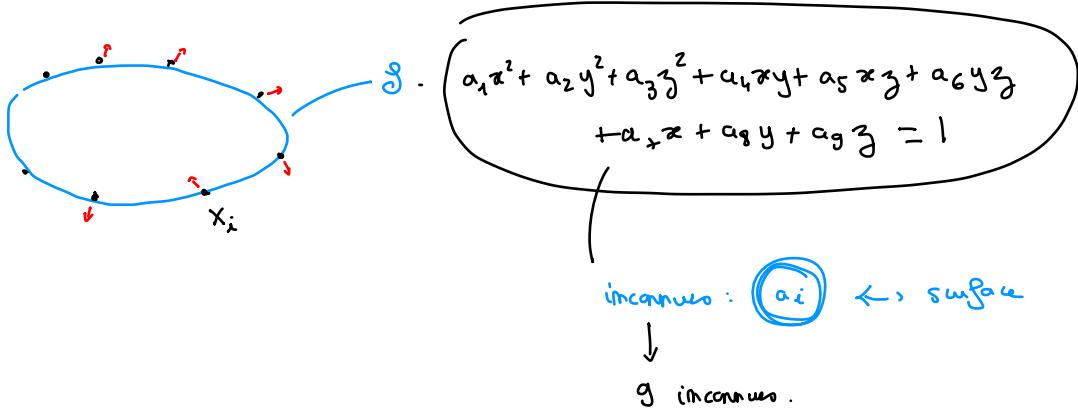


TD / TP 3

Ex. 3



Syst - g inconnus

$$\downarrow \\ g \text{ eq} \leftrightarrow \boxed{g \text{ pts}} \quad x_i$$

Syst:

$$x x_i \rightarrow \left\{ \begin{array}{l} \vdots \\ a_1 x_i^2 + a_2 y_i^2 + a_3 z_i^2 + a_4 x_i y_i + a_5 x_i z_i + a_6 y_i z_i + a_7 x_i + a_8 y_i + a_9 z_i = 1 \leftarrow i \\ \vdots \end{array} \right.$$

$g \text{ eq}$ $g \text{ inconnus}$

$$\downarrow \\ w = \begin{pmatrix} a_1 \\ \vdots \\ a_9 \end{pmatrix}$$

syst. primaire

$$\downarrow$$

$$Xw = y$$

$$Xw = y$$

$$X = \begin{pmatrix} \vdots & \vdots \\ x_i^2 & y_i^2 & z_i^2 & x_i y_i & x_i z_i & y_i z_i & x_i & y_i & z_i \\ \vdots & \vdots \end{pmatrix} \quad g \times g$$

$$y = \begin{pmatrix} \vdots \\ 1 \\ \vdots \\ 1 \end{pmatrix} \quad g \times 1$$

Résoudre en

$$\text{éch.}: \quad w = X \setminus y$$

pts ~ matrice \mathcal{M} - par col

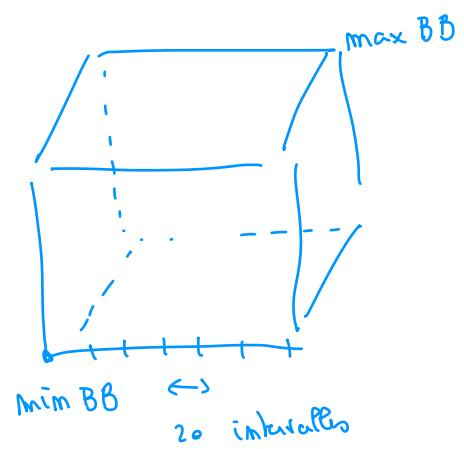
$$g \text{ pts} \rightarrow \mathcal{M} : \boxed{\begin{matrix} & x_i \\ & \vdots \\ 1 & | & 1 & | & 1 & | & 1 \\ & \vdots & & i & & & & g \end{matrix}} \quad \leftarrow \mathcal{M} = \text{rand}(3, g)$$

`plotImplicit3D (f , min BB , max BB , stcf BB)`
 ↓
 prenant
en arg
 X, Y, Z
 (vects du
coords
et calculant
 f en //
 $(... * ...)$

`pt_3D`
 } limites
de la
boîte
englobante
de représentation

`pt_enclu_3D`
 ↘ mème de
seg. selon x, y, z

\rightarrow `plot Implicit3D (f, [-1, -1, -1], [1, 1, 1], [20, 20, 20])`



Script

```
clear
clc
clf
hold on

addpath('MarchingCubes')

N = 9 ;
M = rand(3,N) ;
eps = 10/100. ;

f = compute_quad(M) ;

% Représentation graphique
window = 1.5 ;

plotImplicit3D(f, [-window, -window, -window], [window, window, window], [20, 20, 20]) ;
plot3(M(1,:), M(2,:), M(3,:), 'g', 'MarkerSize',20) ;

% Perturbation des points

delta = (rand(size(M))-0.5)*2*eps ;
M = M+delta ;
f2 = compute_quad(M) ;

% Représentation graphique

plotImplicit3D(f2, [-window, -window, -window], [window, window, window], [20, 20, 20]) ;
plot3(M(1,:), M(2,:), M(3,:), 'r', 'MarkerSize',20) ;
```

Function compute_quad

```
function [f] = compute_quad(M)
    N = size(M,2) ;
    X = [(M(1,:)).^2, (M(2,:)).^2, (M(3,:)).^2, (M(1,:)).*(M(2,:)), (M(1,:)).*(M(3,:)), ...
        (M(2,:)).*(M(3,:)), M(1,:)', M(2,:)', M(3,:)'];
    y = ones(N, 1) ;
    disp(cond(X))

    % Résolution
    w = X\y ; % a1, a2 .....

    % Reconstruction de la fonction f
    f = @(x,y,z) w(1)*x.^2 + w(2)*y.^2 + w(3)*z.^2 + w(4)*x.*y + w(5)*x.*z + ...
        w(6)*y.*z + w(7)*x + w(8)*y + w(9)*z - 1;
end
```

TD \rightarrow ex 1, 2, 5

GSL \rightarrow GNU scientific library \leftrightarrow Matlab C/C++

Linear algebra - Eigen \rightarrow BLAS
Matrices C/C++ / JAVA / Python

Ex. 1

1) We get

$$A = \begin{pmatrix} d_1 & & & \\ & \ddots & & 0 \\ & & \ddots & \\ 0 & & & d_m \end{pmatrix}$$

$$\text{cond}(A) ? = \frac{10^5}{10^{-3}} = 10^8$$

Gauss / LU

2) $[Q, R] = \text{qr}(A)$

Cholesky / $R \cdot R^T$
Householder / QR



$$A = Q \cdot R$$

upper

orthogonal

$$\left| \begin{array}{l} M = \text{rand}(n, n) \\ [Q, R] = \text{qr}(M) \end{array} \right.$$

\hookrightarrow random orthogonal matrix ...

$A \sim$



$Q_n \times \dots \times Q_1$
orthogonal

/symmetries

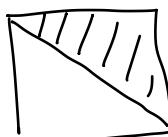
3) $B = Q \times A$ $/10^8 \leftarrow$ can be chosen through
 $\text{cond}(B) = \text{cond}(A)$ the sol. on the diag. of A

solve syst.

$B \rightarrow$

Gauss \rightarrow LU

Householder \rightarrow QR



$\text{cond} ?$



$\text{cond}(B)$

4) Cholesky \rightarrow ~~symmetric~~ definite positive matrices

$$\mathcal{B} = \begin{matrix} Q \times A \\ | \\ \text{diag} \\ \text{ortho} \end{matrix}$$

$$\begin{matrix} \mathcal{B} ? \\ " \\ \mathcal{B}^t = A^t \times Q^t \\ | \\ \text{A} \end{matrix}$$

$$Q^t = Q^{-1} \neq Q$$

$$5) A - \text{diag} - \text{cond}(A) = \underbrace{\frac{c}{\downarrow}}_{\text{ortho}} \rightarrow A^t = A$$

$$\mathcal{B} = Q \times A \times Q^t \rightarrow \text{symmetric} \quad \checkmark$$

$$\text{Definite positive} \xrightarrow{\text{def}} X^t M X > 0 \quad \forall X \neq 0$$

$$\text{eigenvalues of } X^t M X > 0$$

$$\begin{matrix} R \\ | \\ \text{diag} \\ | \\ R \end{matrix}$$

$$\det(\text{sub } R \times R) > 0$$

$$\text{Let } X \in \mathbb{R}^m \quad (*) \quad X^t \cdot (Q \cdot \boxed{A} \cdot Q^t) \cdot X \quad \text{diag} \oplus$$

$$\mu^t \times v \quad \begin{matrix} u^t \\ | \\ \text{---} \\ x \\ | \\ v \end{matrix}$$

$$A = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} = \underbrace{\begin{pmatrix} \sqrt{\lambda_1} & & \\ & \ddots & \\ & & \sqrt{\lambda_n} \end{pmatrix}}_R^t \times \begin{pmatrix} \sqrt{\lambda_1} & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} \quad \frac{\langle u, v \rangle}{\|u\|^2}$$

$$\rightarrow \mu^t \times u = \langle u, u \rangle = \|u\|^2$$

$$(*) = \boxed{X^t \cdot Q \cdot R^t} \underbrace{\boxed{R \cdot Q^t \cdot X}}_{u} = u^t \cdot u = \|u\|^2 \geq 0$$

$= ?$

$$\begin{array}{c} u = 0 \\ | \\ \cancel{R \cdot Q^t \cdot X = 0} \\ | \\ \left(\begin{array}{c} \text{diag} \\ \sqrt{d_1} & \cdots & \sqrt{d_n} \end{array} \right) \\ | \\ \downarrow \\ R^{-1} \checkmark \end{array} \quad \begin{array}{l} Q \text{ is orthogonal} \\ \downarrow \\ Q^t = Q^{-1} \\ \text{invertible} \end{array}$$

$$u = 0 \iff X = 0$$

The matrix we have built is definite positive.

$$B = Q \times A \times Q^t \quad \begin{array}{l} \text{random} \\ \text{sym. def positive} \\ \text{matrix cond. } \subseteq \mathbb{N} \end{array}$$

Ex. 2

$$\begin{array}{c} A \rightsquigarrow \text{eigenvalues } |\lambda_1| > |\lambda_2| > \dots > |\lambda_d| \\ | \\ \text{diagonalizable} \\ | \\ \xrightarrow{\text{---}} \end{array} \quad \begin{array}{c} \downarrow \\ \{ \\ \underbrace{\mu_1}_{\text{---}} \dots \underbrace{\mu_d}_{\text{---}} \\ \text{basis} \end{array}$$

$$x_0 := \sum_i \lambda_i \cdot \vec{u}_i$$

$$x_m ? = A \cdot x_{m-1}$$

$$\begin{aligned} x_1 &= A \cdot x_0 \\ &= A \left(\sum_i \lambda_i \vec{u}_i \right) \\ &= \sum_i \lambda_i \cdot \underbrace{A \vec{u}_i}_{\lambda_i \cdot \vec{u}_i} \\ &= \sum_i \lambda_i \cdot \lambda_i \cdot u_i \end{aligned}$$

$$x_2 = A x_1 = A \left(\sum_i \frac{\lambda_i}{\alpha_i} u_i \right) = \sum_i \alpha_i \cdot \lambda_i \cdot \underbrace{A u_i}_{\alpha_i u_i} = \sum_i \alpha_i \cdot \lambda_i^2 u_i$$

⋮

$$x_n = \sum_i \alpha_i \cdot \lambda_i^n u_i$$

$$\lambda_1 > \lambda_2 \dots > \lambda_d$$

Then

$$\frac{x_n}{\lambda_1^n} = \sum_i \alpha_i \left(\frac{\lambda_i}{\lambda_1} \right)^n \cdot u_i$$

$$\left(\frac{\lambda_i}{\lambda_1} \right)^n \xrightarrow[n \rightarrow \infty]{} \begin{cases} 1 & i=1 \\ 0 & i>1 \end{cases}$$

$$\lambda_i < \lambda_1$$

$$\frac{x_n}{\lambda_1^n} \xrightarrow[n \rightarrow \infty]{} \alpha_1 u_1$$

first eigen vector.

Random diagonalizable matrix

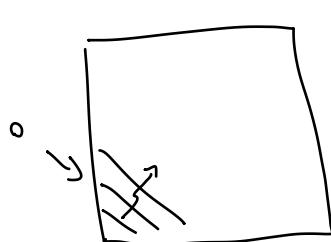
→ rand diag matrix A

→ rand base change

↑
invert. matrix P

change basis
↔

$$P^{-1} \times A \times P$$



↔

