

MATH-H401 Numerical methods

Project 6: Modeling a Capacitor

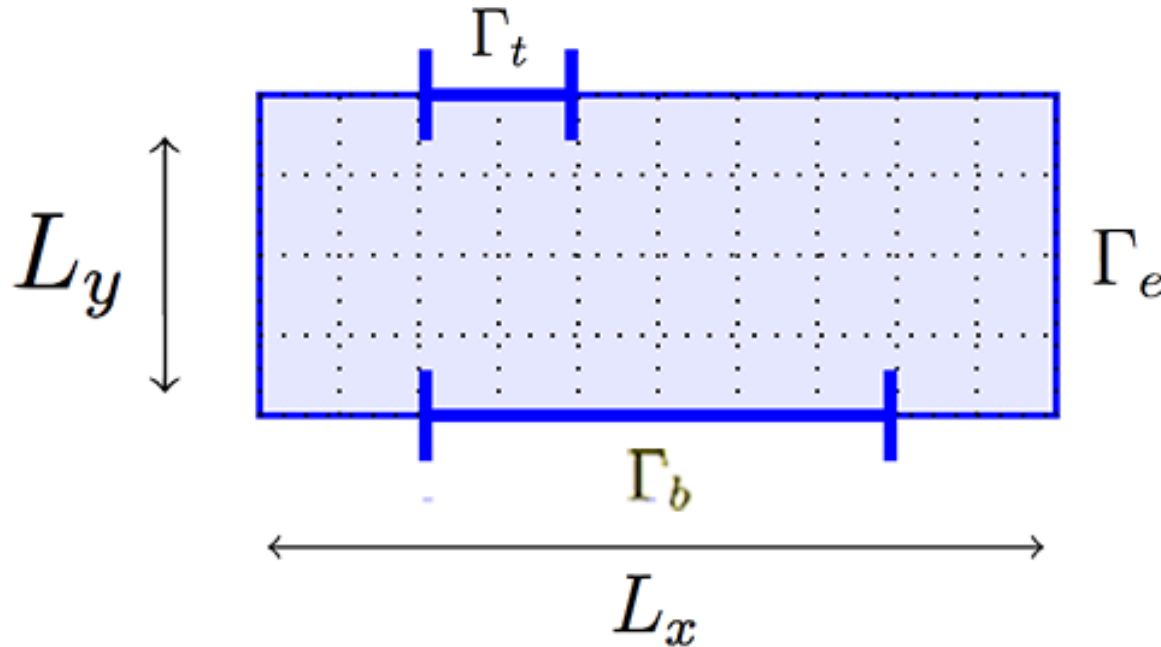
Teachers : Artem Napov
Yvan Notay



Alban Dietrich

1. Introduction

Project 6



Dimensions

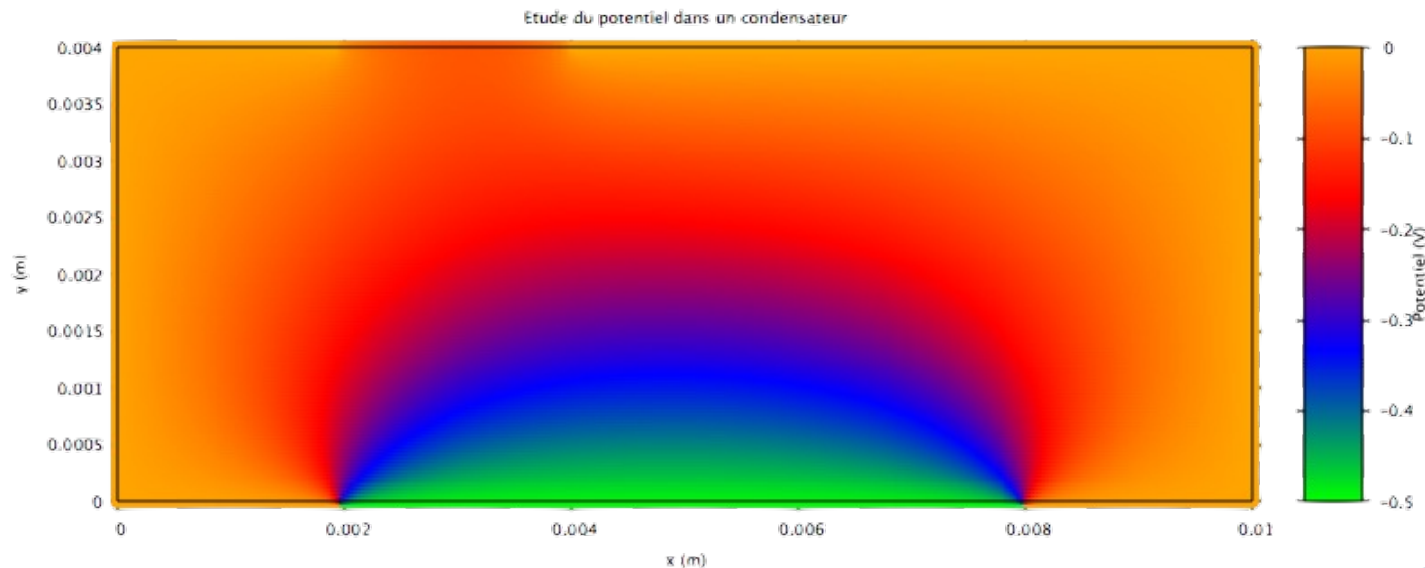
- $L_x = 10 \text{ mm}$
- $L_y = 4 \text{ mm}$

CB

- $\frac{du}{dn} = 0$ on Γ_t
- $u = -0.5$ on Γ_b
- $u = 0$ on Γ_e

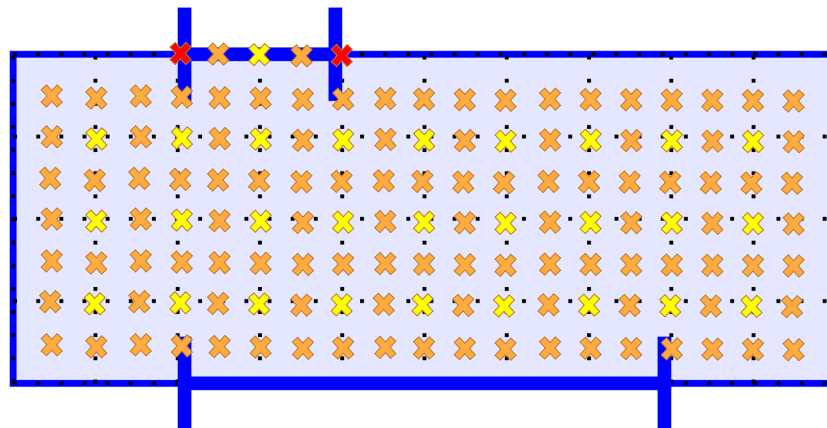
2. Neumann's CB

- Using finite differences
- Implemented in prob.c
- More unknowns

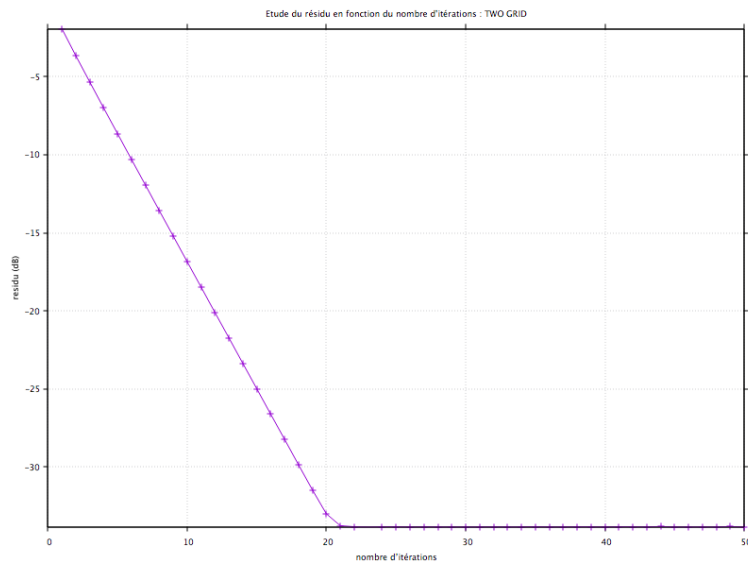


3. Two Grid

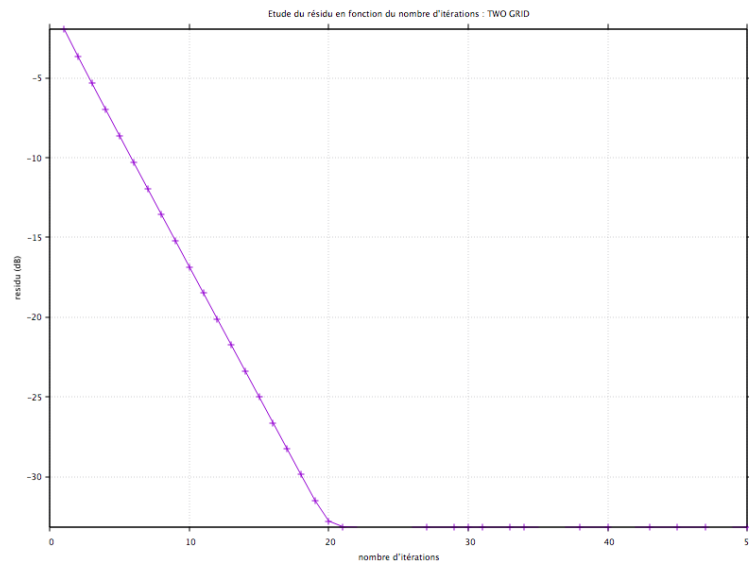
- Creation of symmetrized Gauss-Seidel
- Creation of the restriction
- Solve with UMFPACK $Au_c = r_c$
- Creation of the extension



Result for a step of $h = 3.125 \times 10^{-5} \text{ m}$

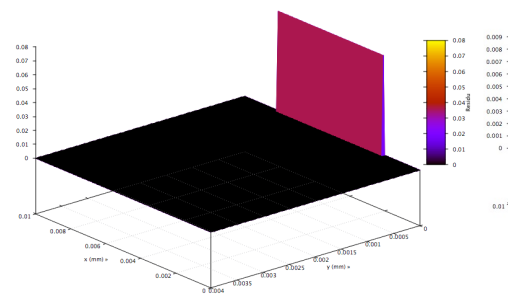


Result for a step of $h = 7.8125 \times 10^{-6} \text{ m}$

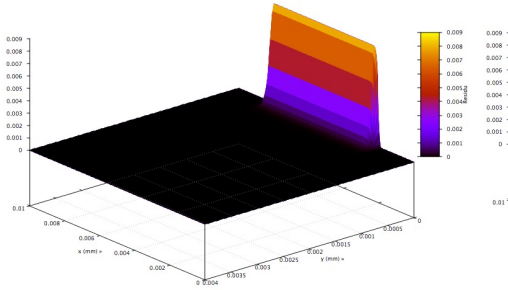


Evolution of the residue

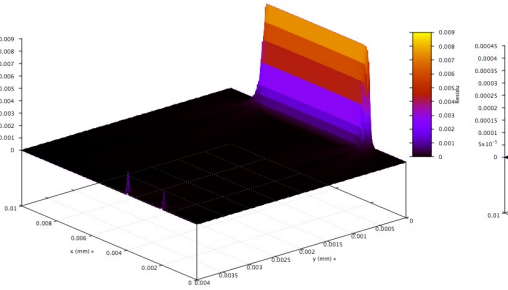
Etude du résidu dans un condensateur



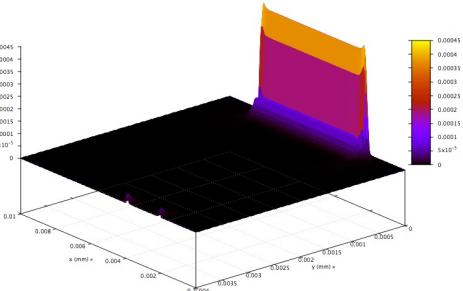
Etude du résidu dans un condensateur



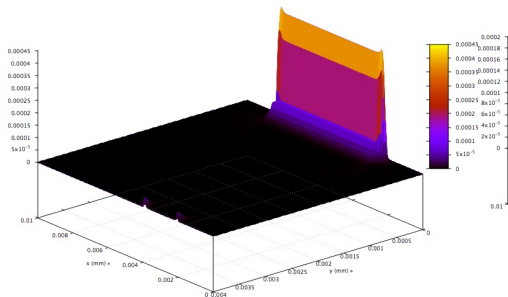
Etude du résidu dans un condensateur



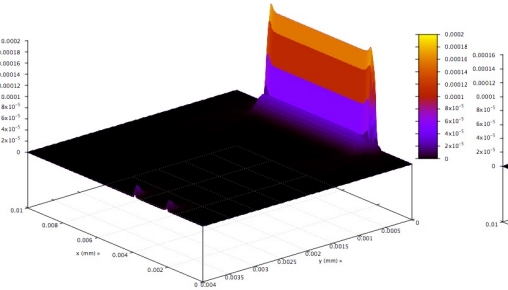
Etude du résidu dans un condensateur



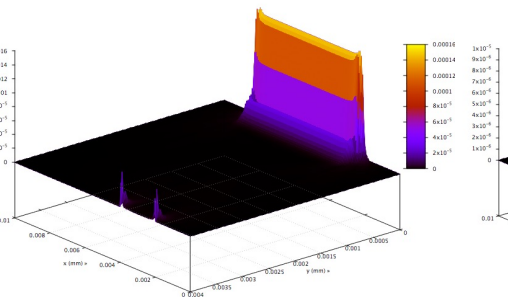
Etude du résidu dans un condensateur



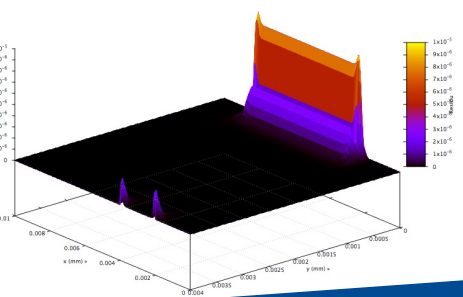
Etude du résidu dans un condensateur



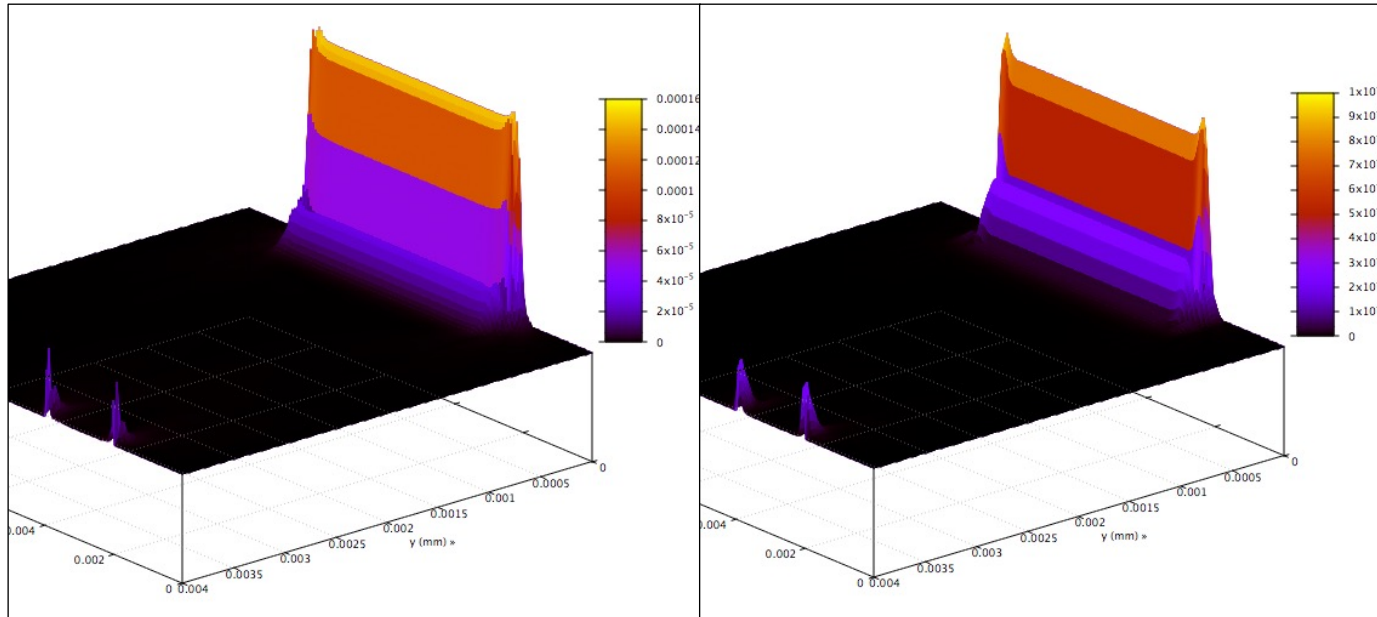
Etude du résidu dans un condensateur



Etude du résidu dans un condensateur

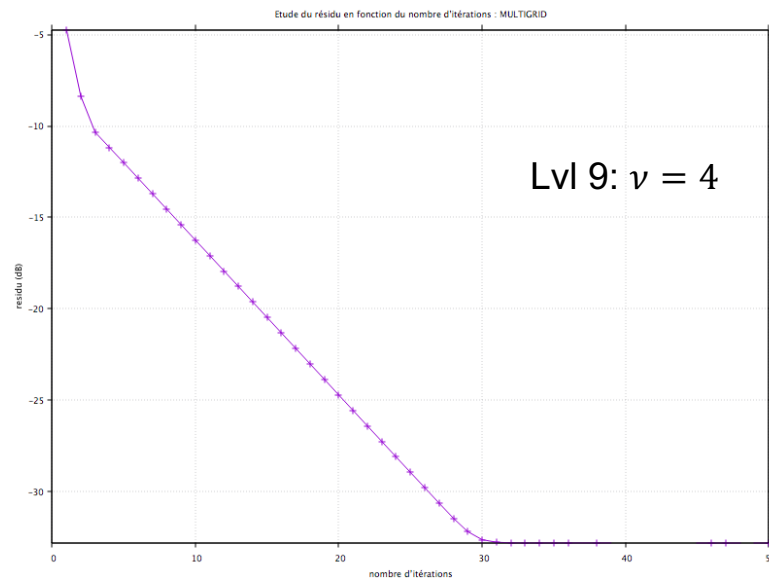
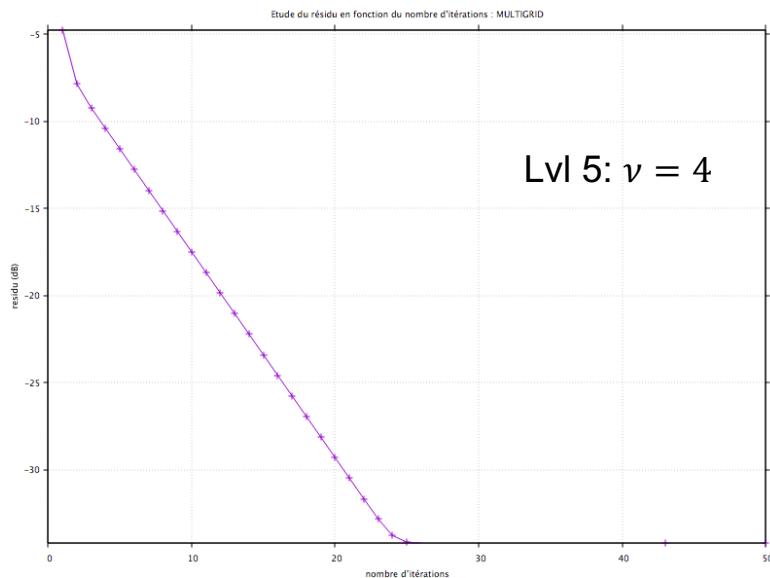


Before and after correction

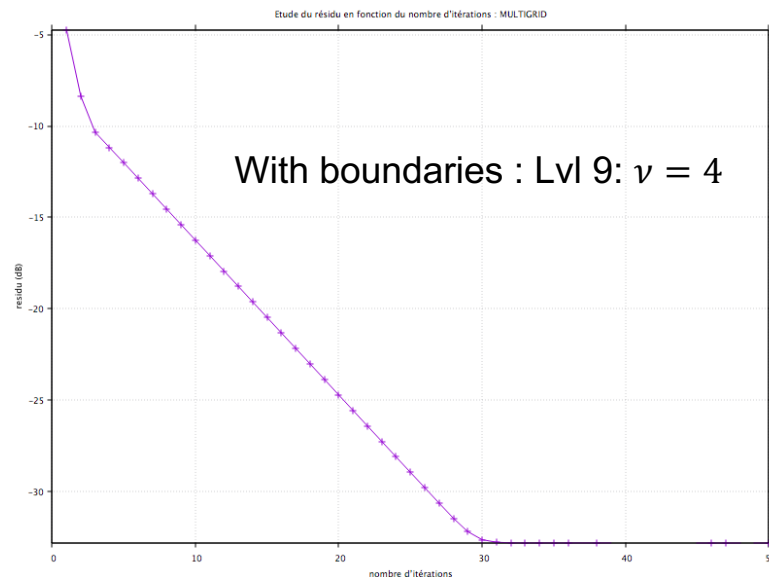
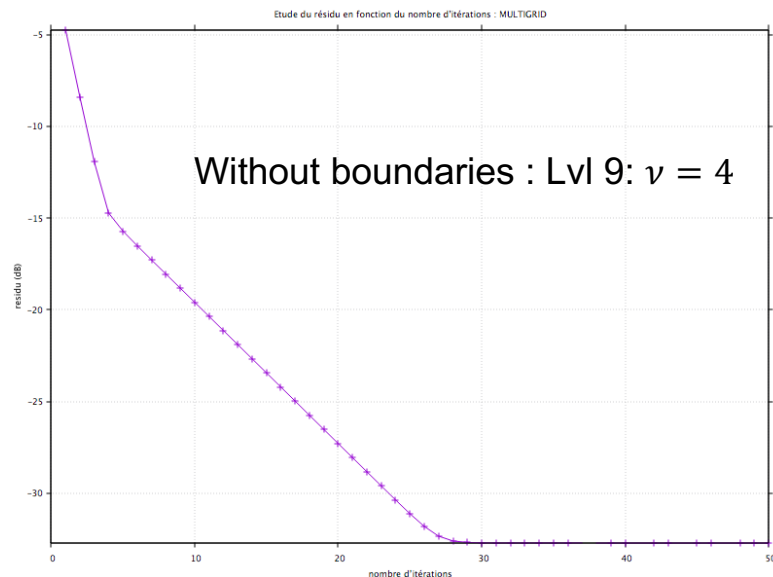


4. Multigrid

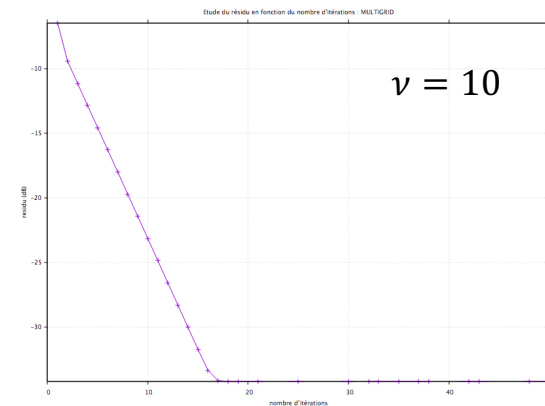
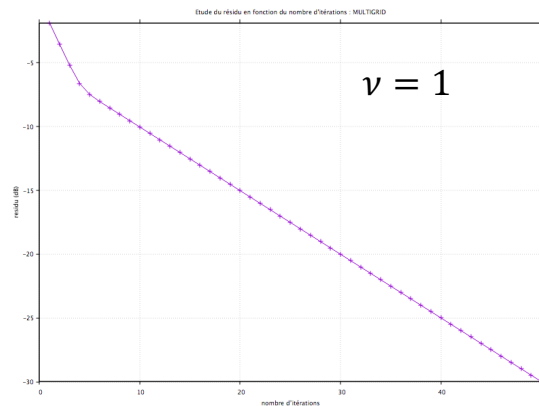
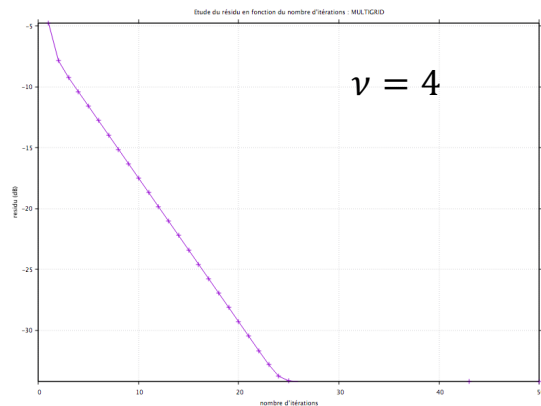
- Problem in the Multigrid : $\nu = 4$ instead of $\nu = 1 \rightarrow$ Visualization of the residue
- With Neumann boundaries \rightarrow good convergence



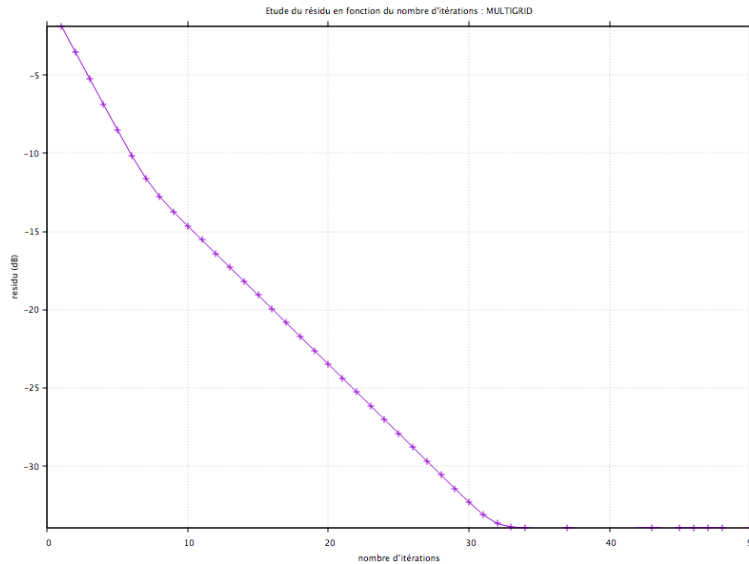
- Comparison WITH or WITHOUT Neumann boundaries
- No big \neq + big break \rightarrow Keeps Neumann boundaries



- For a level 5: passage of $\nu = 4$ instead of $\nu = 1$
- Remove Neumann Boundaries

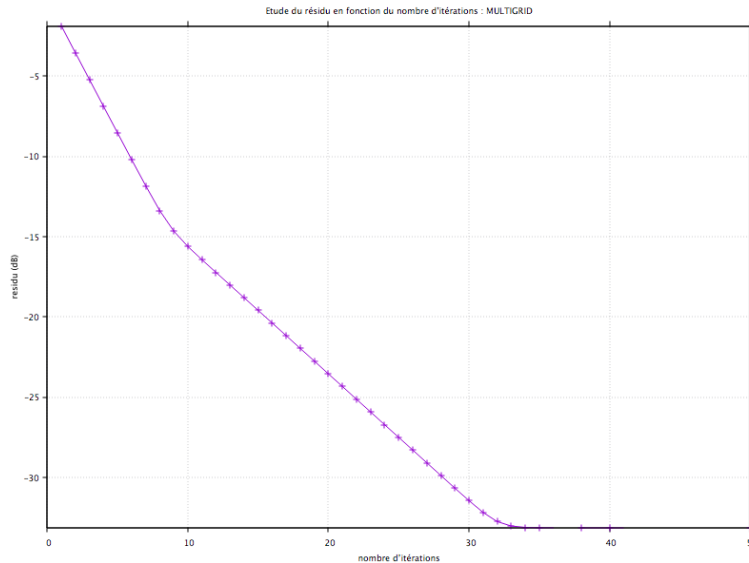


- Faster than Two Grid : for $h = 3.125 \times 10^{-5} \text{ m}$ and 50 iterations
→ TG: 2.586298 s
→ MG: 0.224781s



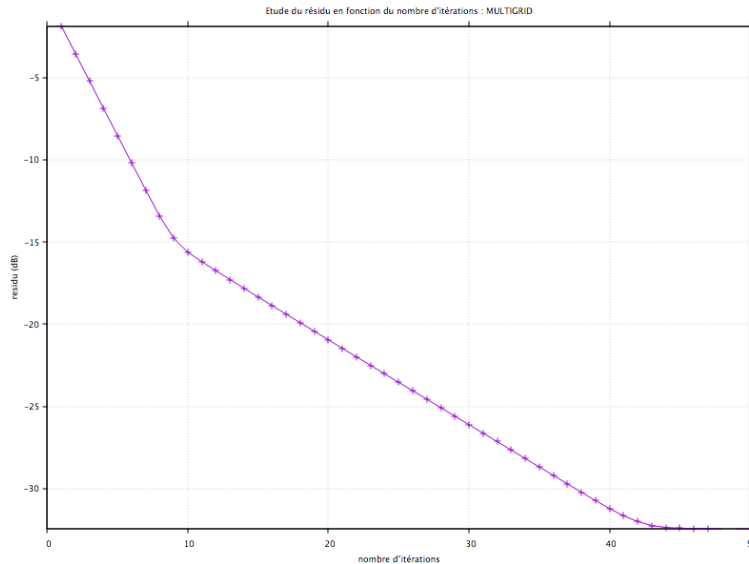
Level 5 : $h = 3.125 \times 10^{-5} \text{ m}$

- Faster than Two Grid : for $h = 3.125 \times 10^{-5} \text{ m}$ and 50 iterations
→ TG: 2.586298 s
→ MG: 0.224781s



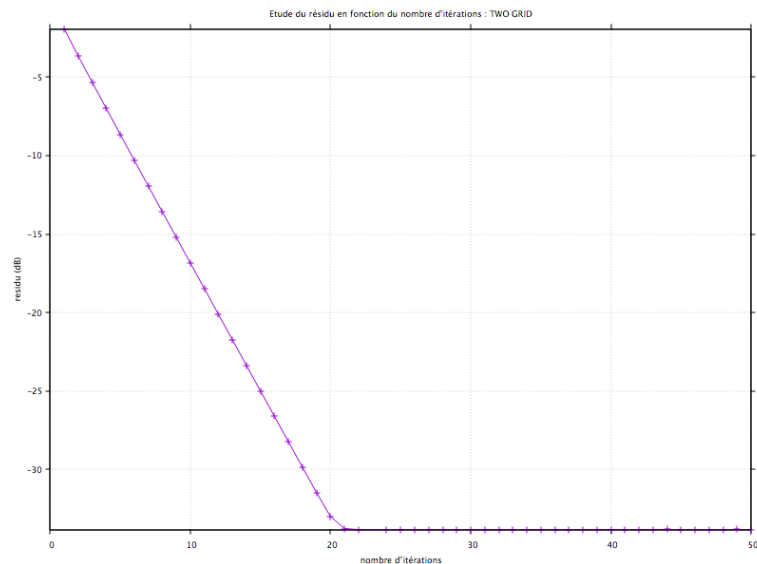
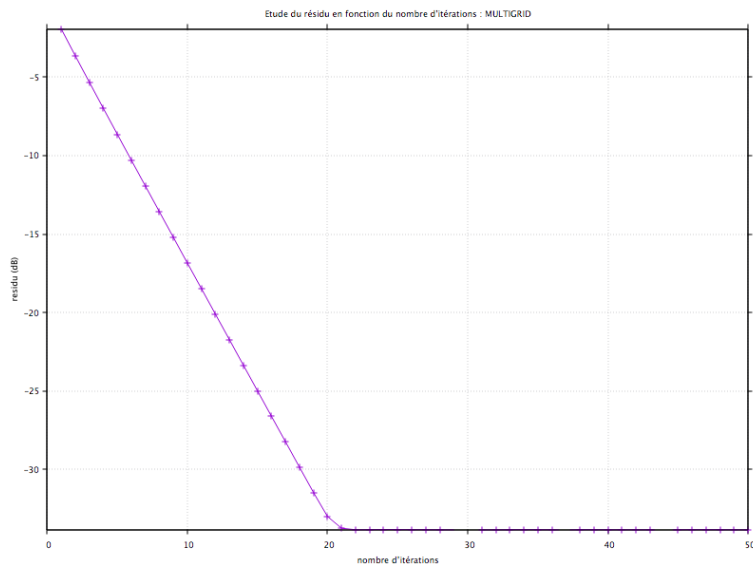
Level 7 : $h = 7.8125 \times 10^{-6} \text{ m}$

- Faster than Two Grid : for $h = 3.125 \times 10^{-5} \text{ m}$ and 50 iterations
 - TG: 2.586298 s
 - MG: 0.224781s

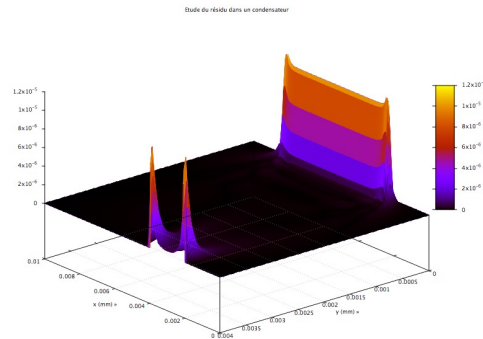
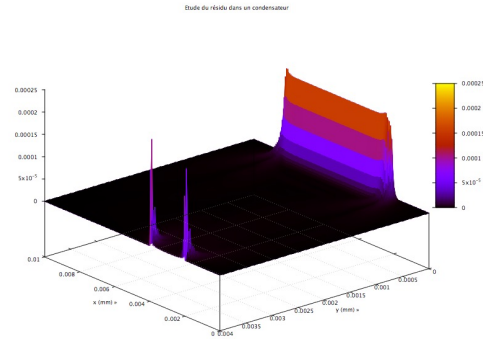


Level 9 : $h = 1.953125 \times 10^{-6} \text{ m}$

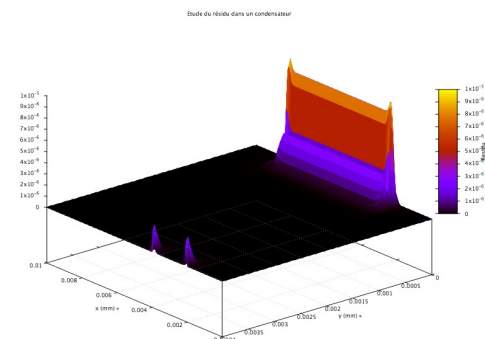
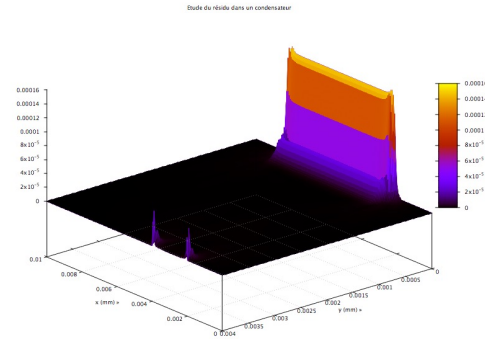
- MG can be used as a Two Grid : for $h = 3.125 \times 10^{-5} \text{ m}$
(resolution at $h = 6.25 \times 10^{-5} \text{ m}$)



Multigrid



Two Grid



Multigrid convergence

Level 9: $h = 1.953125 \times 10^{-6} m$

Stability criterion

$$\|r\| = C v \|A\| \|\tilde{u}\|$$

$$\|r\| \approx 0,1201642$$

$$\|A\| = \lambda_{\max}(A) \approx 2,097152 \cdot 10^{12}$$

$$\|\tilde{u}\| = 639,904812$$

$$\rightarrow C \approx 0,81$$

Calculation of ρ and τ

$$\rho = \frac{r_n}{r_{n-1}} \approx 0,5786$$

$$\lambda_{\max}(B^{-1}A) \approx 1$$

$$\rho(T) = \max(1 - \tau\lambda_{\min}, \tau\lambda_{\max} - 1)$$

$$\rho(T) = 1 - \tau\lambda_{\min} \leftrightarrow \lambda_{\min} = 0.4213$$

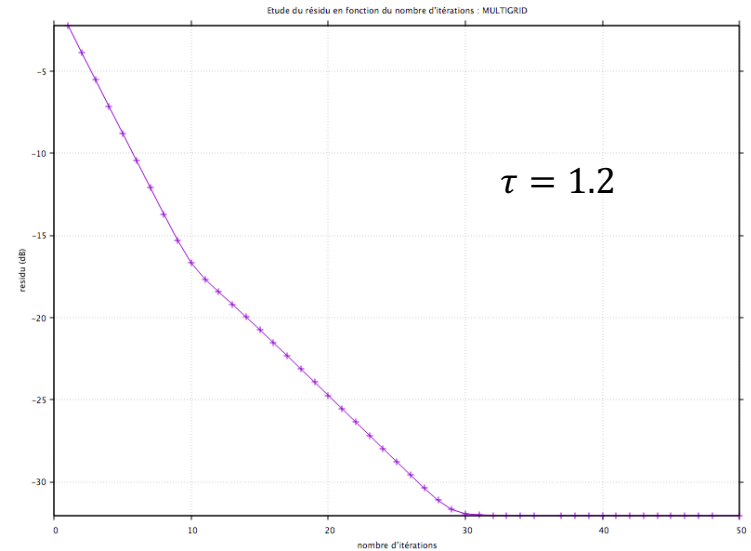
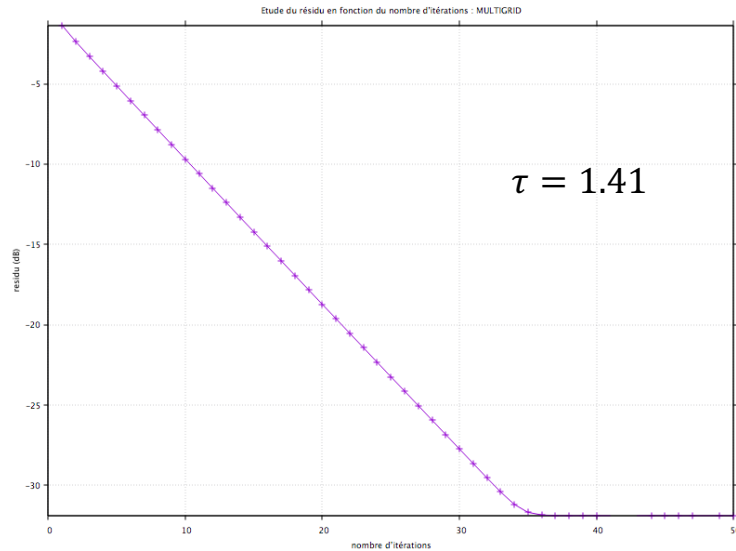
$$\rightarrow \tau_{\text{opti}} = \frac{2}{\lambda_{\max} + \lambda_{\min}} \approx 1,41$$

Conclusion

$$\rho_{\text{exp}} = \frac{r_n}{r_{n-1}} \approx 0,4162$$

$$\rho_{\text{th}} = \frac{\lambda_{\max} - \lambda_{\min}}{\lambda_{\max} + \lambda_{\min}} \approx 0.4072$$

Results



Level 5: $h = 3.125 \times 10^{-5} m$

Stability criterion

$$\|r\| = C \vee \|A\| \|\tilde{u}\|$$

$$\|r\| \approx 0.00003020493$$

$$\|A\| = \lambda_{\max}(A) \approx 8192000000$$

$$\|\tilde{u}\| = 39.820483$$

$$\rightarrow C \approx 0.84$$

Calculation of ρ and τ

$$\rho = \frac{r_n}{r_{n-1}} \approx 0.4497$$

$$\rho(T) = 1 - \tau \lambda_{\min} \leftrightarrow \lambda_{\min} = 0.5503$$

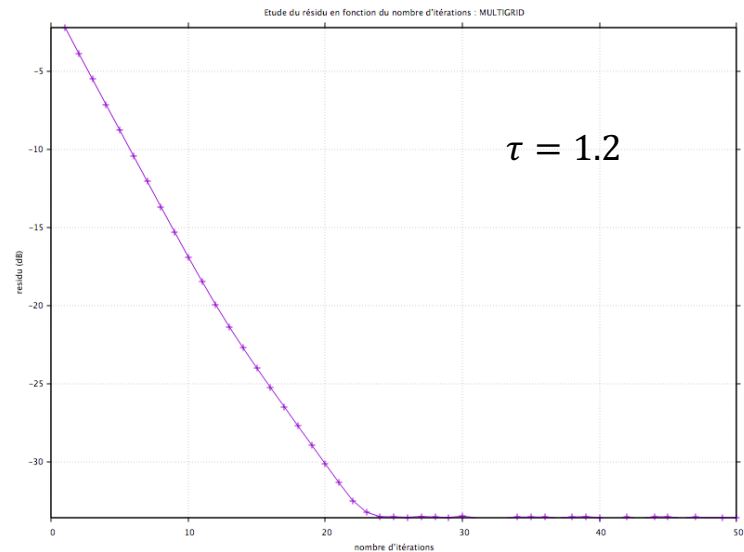
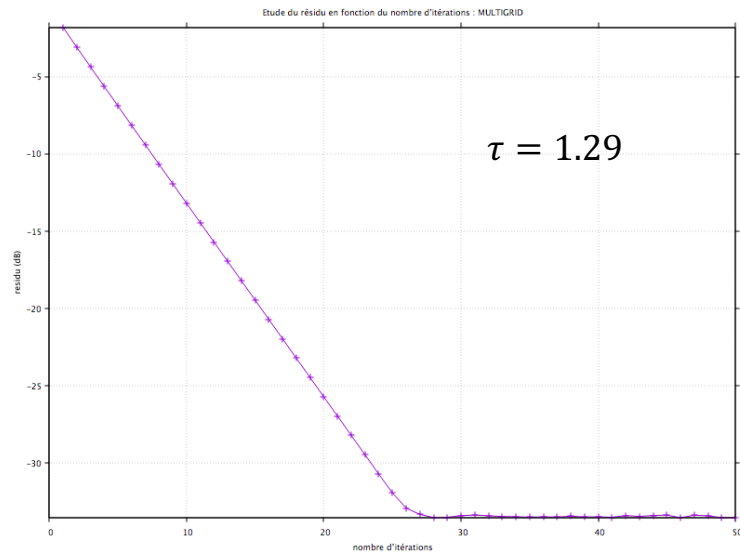
$$\rightarrow \tau_{\text{opti}} = \frac{2}{\lambda_{\max} + \lambda_{\min}} \approx 1.29$$

Conclusion

$$\rho_{\text{exp}} = \frac{r_n}{r_{n-1}} \approx 0.3692$$

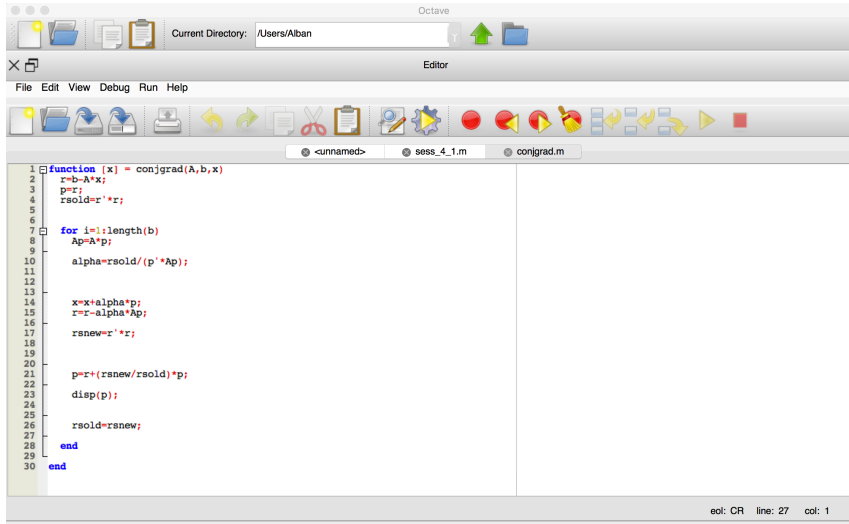
$$\rho_{\text{th}} = \frac{\lambda_{\max} - \lambda_{\min}}{\lambda_{\max} + \lambda_{\min}} \approx 0.29$$

Results



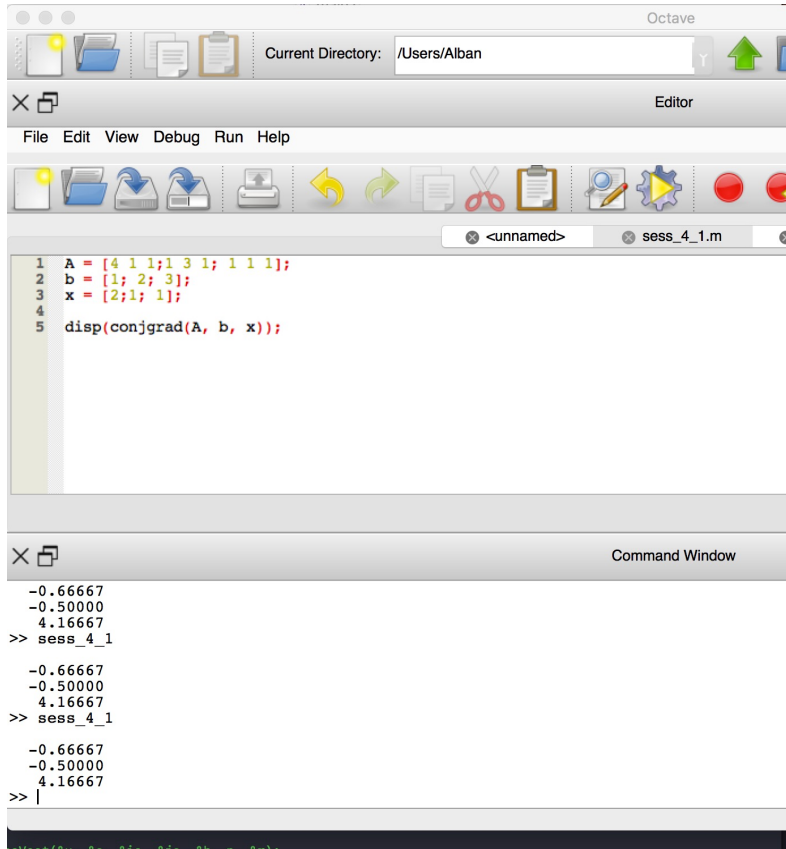
5. Conjugate gradient

- Conjugate gradient code converges to a residual $\propto 10^{-3}$ in 15 iterations
- Works with Octave → Gets GC without preconditioner



```
1 function [x] = conjgrad(A,b,x)
2
3   r=b-A*x;
4   p=r;
5   rsold=r'*r;
6
7   for i=1:length(b)
8     Ap=A*p;
9     alpha=rsold/(p'*Ap);
10
11     x=x+alpha*p;
12     r=r-alpha*Ap;
13     rsnew=r'*r;
14
15     p=p+(rsnew/rsold)*p;
16     disp(p);
17     rsold=rsnew;
18   end
19 end
```

Octave
Current Directory: /Users/Alban
Editor
File Edit View Debug Run Help
eol: CR line: 27 col: 1



Octave

Current Directory: /Users/Alban

Editor

File Edit View Debug Run Help

```

1 A = [4 1 1; 1 3 1; 1 1 1];
2 b = [1; 2; 3];
3 x = [2; 1; 1];
4
5 disp(conjgrad(A, b, x));

```

Command Window

```

-0.66667
-0.50000
4.16667
>> sess_4_1

-0.66667
-0.50000
4.16667
>> sess_4_1

-0.66667
-0.50000
4.16667
>> |

```

```

/* DECLARATION DE VARIABLES */
double beta = 0, rsSolid = 0, rsNew = 0, alpha = 0, numAlpha = 0, denAlpha = 0;
double *Ad;
printf("\nhello\n");
//Choix pour le problème du cours
n = 3;
Ad = malloc(n*sizeof(double));
a = malloc(n*n*sizeof(double));
u = malloc(n*sizeof(double));
r = malloc(n*sizeof(double));
d = malloc(n*sizeof(double));
b = malloc(n*sizeof(double));
ja = malloc(*sizeof(int));
ia = malloc(*sizeof(int));

a[0] = 4.0; a[1] = 1.0; a[2] = 1.0; a[3] = 1.0; a[4] = 3.0; a[5] = 1.0; a[6] = 1.0; a[7] = 1.0; a[8] = 1.0;
ja[0] = 0; ja[1] = 1; ja[2] = 2; ja[3] = 0; ja[4] = 1; ja[5] = 2; ja[6] = 0; ja[7] = 1; ja[8] = 2;
ia[0] = 0; ia[1] = 3; ia[2] = 0; ia[3] = 0;
b[0] = 1; b[1] = 2; b[2] = 3;

CG(&ia, &ja, &a, &b, &u, n);

printf("\nSOL00 %f %f %f\n", u[0], u[1], u[2]);

```

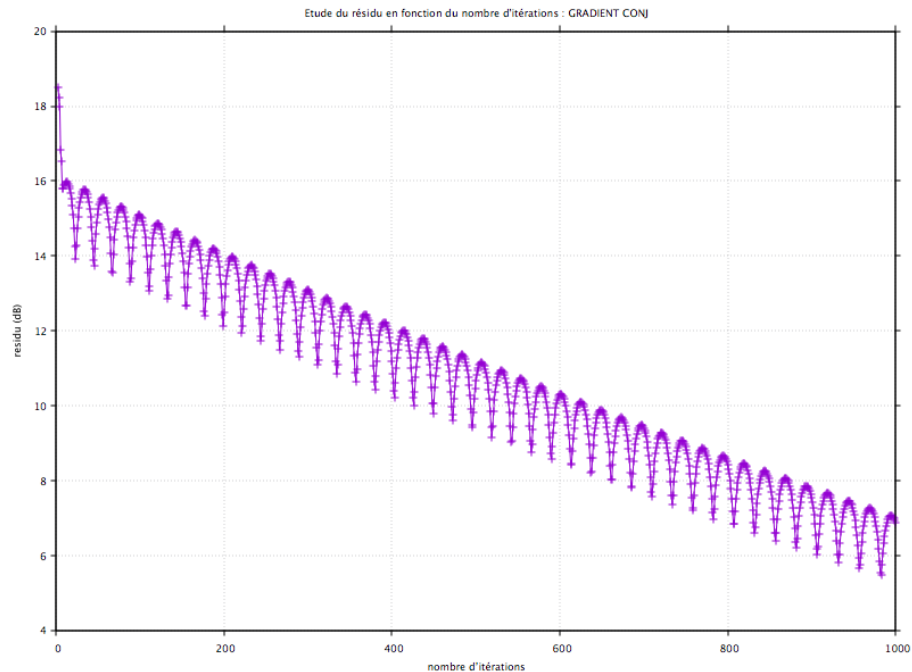
[GENERATION DU PROBLEME]

PROBLEM: mx = 321 my = 129 n = 40576 nnz = 201986

SOLUTION -0.666667 -0.500000 4.166667

MacBook-Pro:CG Sans preconditionner Alban\$

- With preconditioner : bad convergence \rightarrow jumps



INTERLUDE : PROPRIÉTÉS DE LA NORME EUCLIDIENNE

PROPRIÉTÉ 1 : soient A une matrice symétrique et $\lambda_i, i = 1, \dots, n$, ses valeurs propres. Alors

$$\|A\|_2 = \max_i |\lambda_i| = \max_{\mathbf{v}} \frac{|\mathbf{v}^T A \mathbf{v}|}{\mathbf{v}^T \mathbf{v}}.$$

Soit \mathbf{p}_i le vecteur propre normalisé associé à $\lambda_i, i = 1, \dots, n$. L'ensemble de ces vecteurs forment une base orthonormale et tout vecteur \mathbf{v} a une représentation $\mathbf{v} = \sum_{i=1}^n \alpha_i \mathbf{p}_i$ dans cette base. On a alors

$$\|A\|_2^2 = \max_{\mathbf{v}} \frac{\|A\mathbf{v}\|_2^2}{\|\mathbf{v}\|_2^2} = \max_{\mathbf{v}} \frac{\mathbf{v}^T A^2 \mathbf{v}}{\mathbf{v}^T \mathbf{v}} = \max_{\alpha_1, \dots, \alpha_n} \frac{\sum_{i=1}^n \lambda_i^2 \alpha_i^2}{\sum_{i=1}^n \alpha_i^2} = \max_i |\lambda_i|^2.$$

Par analogie avec les deux dernières égalités on a aussi

$$\max_{\mathbf{v}} \frac{|\mathbf{v}^T A \mathbf{v}|}{\mathbf{v}^T \mathbf{v}} = \max_{\alpha_1, \dots, \alpha_n} \frac{|\sum_{i=1}^n \lambda_i \alpha_i^2|}{\sum_{i=1}^n \alpha_i^2} = \max_i |\lambda_i|.$$

PROPRIÉTÉ 2 : pour tout matrice A rectangulaire

$$\|A^T A\|_2 = \|A\|_2^2.$$

$$\|A^T A\|_2 = \max_{\mathbf{v}} \frac{|\mathbf{v}^T A^T A \mathbf{v}|}{\mathbf{v}^T \mathbf{v}} = \max_{\mathbf{v}} \frac{\|A\mathbf{v}\|_2^2}{\|\mathbf{v}\|_2^2}.$$

3.3 Relaxation (5)

ULB

(p. 22)

Using this optimal value of τ , one obtains

$$\rho(T) = \frac{\lambda_{\max}(B^{-1}A) - \lambda_{\min}(B^{-1}A)}{\lambda_{\max}(B^{-1}A) + \lambda_{\min}(B^{-1}A)} = \frac{\kappa - 1}{\kappa + 1},$$

where κ is the condition number of the preconditioned system:

$$\kappa = \kappa(B^{-1}A) = \frac{\lambda_{\max}(B^{-1}A)}{\lambda_{\min}(B^{-1}A)}.$$

- $k_{\varepsilon} \geq \frac{\ln(\frac{1}{\varepsilon})}{\ln(\frac{1}{\rho})}$
- $k_{\varepsilon} \geq 15$ (for $\varepsilon = 10^{-6}$ et $\rho = 0,4$)

The corresponding estimate of the number of iterations is (see above) the smallest integer satisfying

$$k_{\varepsilon} \geq \frac{\ln \frac{1}{\varepsilon}}{\ln \frac{1}{\rho(T)}} = \frac{\ln \frac{1}{\varepsilon}}{\ln \frac{\kappa+1}{\kappa-1}}. \quad (2)$$

where

$$\ln \frac{\kappa+1}{\kappa-1} = \ln \frac{1+\kappa^{-1}}{1-\kappa^{-1}} \geq \frac{2}{\kappa}$$

(with $\mathcal{O}(\kappa^{-3})$ error).