Simulation of the Poisson problem on the 3D Torus

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May 3, 2020

1 Introduction

We consider the torus defined as follow

$$\Gamma = \{(x, y, z) \in \mathbb{R}^3, (\sqrt{x^2 + y^2} - R)^2 + z^2 - r^2 = 0\},\$$

where r > 0 is the minor radius and R > r is the major radius.

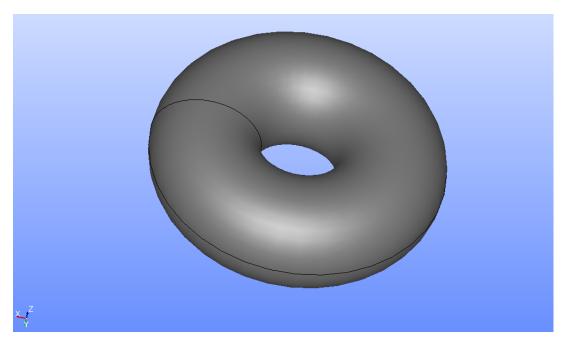


Figure 1: The torus in SALOME CAO module

The torus is a C^{∞} manifold of dimension 2 embedded in \mathbb{R}^3 . The **Laplace-Beltrami operator** \triangle_{Γ} , a generalisation of the euclidean laplacean, can be defined on the torus as the combination of a surface divergence ∇_{Γ} , and of a surface gradient $\vec{\nabla}_{\Gamma}$ (see for instance [3, 6, 12]).

Using the torus coordinates (θ, ϕ) where $\theta, \phi \in [0, 2\pi]$, the Laplace-Beltrami operator takes the following form on the torus

$$\Delta_{\Gamma} f = \frac{1}{r^2 (R + r \cos \theta)} \frac{\partial}{\partial \theta} \left((R + r \cos \theta) \frac{\partial f}{\partial \theta} \right) + \frac{1}{(R + r \cos \theta)^2} \frac{\partial^2 f}{\partial \phi^2}.$$

We consider the following **Poisson problem** on the torus

$$\begin{cases}
-\Delta_{\Gamma} u = f \text{ on } \Gamma \\
\int_{\Gamma} u = 0
\end{cases} , \tag{1}$$

where the right hand side $f \in L^2(\Gamma)$ and the unknown $u \in H^1(\Gamma)$ are **zero** mean functions.

For the following choice of $u(\theta, \phi) = \sin(3\phi)\cos(3\theta + \phi)$ (see [5]), the right hand side is given by

$$f = 9r^{-2}\sin(3\phi)\cos(3\theta + \phi) - (10\sin(3\phi)\cos(3\theta + \phi) + 6\cos(3\phi)\sin(3\theta + \phi))((R + r\cos(\theta))^{-2} - 3\sin(\theta)\sin(3\phi)\sin(3\theta + \phi)(r(R + r\cos(\theta)))^{-1}.$$

Our objective is to solve numerically the Poissson problem (1) using the finite element method described in [1, 4, 2, 3].

2 Finite elements method for 3D Poisson problem

Since Γ is closed (no boundary), we have to impose the global condition $\int_{\Gamma} u = 0$ to guarantee the uniqueness of solution. For this reason we define the following Lebesgue space

$$L^{2}_{\#}(\Gamma) = \{ w \in L^{2}(\Gamma) : \int_{\Gamma} w = 0 \},$$

and the following Sobolev space

$$H^1_{\#}(\Gamma) = \{ w \in H^1(\Gamma) : \int_{\Gamma} w = 0 \}.$$

2.1 Well-posedness of the problem

2.1.1 Variational formulation and Poincaré inequality

In this section we are going to recall two important properties of the surface gradient operator $\vec{\nabla}_{\Gamma}$: the Green-Ostrogradski formula (i.e. integration by part) and Poincaré inequality.

Thanks to the **Green-Ostrograski** theorem, the variational formulation of (1) is:

Find
$$u \in H^1_{\#}(\Gamma)$$
 such that $\forall v \in H^1_{\#}(\Gamma), \int_{\Gamma} \overrightarrow{\nabla}_{\Gamma} u \cdot \overrightarrow{\nabla}_{\Gamma} v = \int_{\Gamma} f v.$ (2)

As for the classical gradient, there is a **Poincaré's inequality** involving the surface gradient (see theorem 2.12 in [3]).

Theorem 1 (Poincaré's inequality).

Assume that Γ is an embedded C^3 hypersurface. There exists a constant c such that, for every function $f \in H^1(\Gamma)$ with $\int_{\Gamma} f = 0$, we have the inequality

$$||f||_{L^2(\Gamma)} \le c||\nabla_{\Gamma} f||_{L^2(\Gamma)}.$$

2.1.2 Existence of a unique weak solution

The bilinear form

$$a(u,v) = \int_{\Gamma} \overrightarrow{\nabla}_{\Gamma} u \cdot \overrightarrow{\nabla}_{\Gamma} v$$

is continuous and coercive thanks to Poincaré inequality.

The linear form

$$b(v) = \int_{\Gamma} fv$$

is continuous.

By application of the **Lax-Milgram theorem**, the variational formulation (2) of problem (1) admits a unique weak solution, which depends continuously on the data f (see Theorem 3.1 in [3]).

Theorem 2 (Well-posedness). Let Γ , be a compact C^2 hypersurface in \mathbb{R}^3 and assume that $f \in L^2(\Gamma)$ and $\int_{\Gamma} f = 0$. Then there exists a unique solution $u \in H^1(\Gamma)$ of (2) with $\int_{\Gamma} u = 0$.

2.1.3 Regularity of the solution

The regularity of the solution requires the regularity of both the right hand side and the manifold. The following theorem is taken from [3] theorem 3.3.

Theorem 3. Let Γ , be a compact C^2 hypersurface in \mathbb{R}^3 and assume that $f \in L^2(\Gamma)$ and $\int_{\Gamma} f = 0$. Then, the unique weak solution of (2) satisfies $u \in H^2_{\#}(\Gamma)$, and there exists a constant C > 0 such that

$$||u||_{H^2_{\#}(\Gamma)} \le C||f||_{L^2_{\#}(\Gamma)}.$$

For more details see [1] for euclidian case and [2, 3] for the case of curved surfaces.

2.2 The P1 finite elements

Following [3], we first approximate the torus Γ by a polyhedral surface Γ_h with triangular faces $(\mathcal{T}_k)_{k\geq 1}$ called elements having their nodes on Γ . We approximate functions $f\in H^1_\#(\Gamma)$ by functions $f_h\in H^1_\#(\Gamma_h)$ via the lift operator (10).

We consider u_h the weak solution of the following Poisson problem on the piecewise linear manifold Γ_h :

$$\begin{cases}
-\Delta_{\Gamma_h} u_h = f_h \text{ on } \Gamma_h \\
u_h \in H^1_\#(\Gamma_h)
\end{cases} , \tag{3}$$

and its variational formulation, analog to (2) is

Find
$$u_h \in H^1_\#(\Gamma_h)$$
 such that $\forall v_h \in H^1_\#(\Gamma_h), \int_{\Gamma_h} \overrightarrow{\nabla}_{\Gamma_h} u_h \cdot \overrightarrow{\nabla}_{\Gamma_h} v_h = \int_{\Gamma_h} f_h v_h.$

(4)

We look for \tilde{u}_h the projection of the solution u_h of (4) on the space $V_0(\Gamma_h)$ of continuous piecewise affine functions with zero mean on Γ_h . The discrete form of the variational formulation (4) is then given by.

Find
$$\tilde{u}_h \in V_0(\Gamma_h)$$
 such that $\forall \tilde{v}_h \in V_0(\Gamma_h), \int_{\Gamma_h} \overrightarrow{\nabla}_{\Gamma_h} \tilde{u}_h \cdot \overrightarrow{\nabla}_{\Gamma_h} \tilde{v}_h = \int_{\Gamma_h} \tilde{f}_h \tilde{v}_h$, (5)

where \tilde{f}_h is the projection of f_h on $V_0(\Gamma_h)$.

2.3 The linear system to be solved

Since $V_0(\Gamma_h)$ is generated by the nodal functions $\phi_i : \Gamma_h \to \mathbb{R}$, i = 1, ..., n such that $\phi_i(x_j) = \delta_{ij}$, (5) takes the following algebraic form

$$A_{\triangle_{\Gamma_h}} X = b_h, \tag{6}$$

where

$$\tilde{u}_h = \sum_{i=1}^n u_i \phi_i,\tag{7}$$

 $A_{\triangle_{\Gamma_h}} = (a_{ij})_{i,j=1,...,n}, X = {}^t(u_1,...,u_n) \text{ and } b_h = {}^t(b_1,...,b_n) \text{ with}$

$$a_{ij} = \int_{\Gamma_h} \overrightarrow{\nabla}_{\Gamma_h} \phi_i \cdot \overrightarrow{\nabla}_{\Gamma_h} \phi_j = \sum_{k=1}^n \int_{\mathcal{T}_k} \overrightarrow{\nabla}_{\Gamma_h} \phi_i \cdot \overrightarrow{\nabla}_{\Gamma_h} \phi_j,$$

$$b_j = \int_{\Gamma_h} f \phi_j = \sum_{k=1}^n \int_{\mathcal{T}_k} f \phi_j.$$

 $A_{\triangle_{\Gamma_h}}$ is symmetric positive and sparse but not invertible since constants are in its kernel, hence the linear system (6) is singular. However it admits a unique solution with zero mean provided the right hand side has zero mean (see [3]).

2.4 Convergence of the numerical method

2.4.1 Fermi coordinates and lift operator

A function u defined on Γ can be extended to a neighborhood of Γ in \mathbb{R}^3 using a lift operator based on the Fermi coordinates around Γ . Following [3], we define the δ -strip around Γ as

$$U_{\delta,\Gamma} = \{ x \in \mathbb{R}^3, dist(x,\Gamma) < \delta \}. \tag{8}$$

For δ small enough it is possible to define the projection $a: U_{\delta} \to \Gamma$ onto Γ and the distance function $d: U_{\delta} \to \mathbb{R}_+$ to Γ . a(x) and d(x) are called the **Fermi coordinates** of x and their existence is given by the following theorem (see Lemma 2.8 in [3] for the proof).

Theorem 4 (Fermi coordinates).

Let Γ be an embedded C^2 hypersurface. There exists $\delta_{Fermi} > 0$ such that for every point $x \in U_{\delta_{Fermi},\Gamma}$, there exists a unique point $a(x) \in \Gamma$, and a function $d \in C^2(U_{\delta_{Fermi},\Gamma})$ such that

$$\forall x \in U_{\delta_{Fermi}, \Gamma}, \quad x = a(x) + d(x) \overrightarrow{n}(x),$$
 (9)

where $\overrightarrow{n}(x)$ is the unit normal vector to Γ at x.

Thanks to the **Fermi coordinates** defined in theorem 4, we can define as in [3] (equation 4.2) a **lift operator** L such that

$$L: C(\Gamma_h) \to C(\Gamma) u_h \to u_h \circ a^{-1} ,$$
 (10)

provided

$$\Gamma_h \subset U_{\delta_{Fermi},\Gamma}.$$
 (11)

2.4.2 Convergence theorems

In order to study the convergence of the finite element approximation, we need to compare $u \in H^1(\Gamma)$ with $\tilde{u}_h \in H^1(\Gamma_h)$ but don't share the same support. Hence we need to use the lift operator (10) which requires the assumption (11) that the triangulated surface Γ_h is close enough to Γ .

As the parameter h goes to zero the distance between Γ_h and Γ converges to zero as expressed in the following theorem taken from [3] Lemma 4.1.

Theorem 5 (Convergence of Γ_h towards Γ). Let $\Gamma \in \mathbb{R}^3$ be an embedded C^2 hypersurface and $\Gamma_h \subset U_{\delta_{Fermi},\Gamma}$ a piecewise linear surface. Let h be the largest diameter of trianges in Γ_h . There exists a constant c such that

$$\forall x \in \Gamma_h, \quad dist(x, \Gamma) \le ch^2.$$

Once proven that Γ_h converges towards Γ , we can prove that \tilde{u}_h converges to u using the lift operator (10). The following convergence theorem is taken from [2] Theorem 8, Lemma 6 and Lemma 7.

Theorem 6 (Convergence of \tilde{u}_h towards u). Let $\Gamma \in \mathbb{R}^3$ be an embedded C^2 hypersurface and $\Gamma_h \subset U_{\delta_{Fermi},\Gamma}$ a piecewise linear surface. Let h be the largest diameter of trianges in Γ_h .

If u is a continuous solution of the Poisson problem (1) and \tilde{u}_h is the discrete solution of (5), then there exists c > 0 such that

$$||u - \tilde{u}_h \circ a^{-1}||_{L^2(\Gamma)} \le ch^2, \quad ||\nabla_{\Gamma}(u - \tilde{u}_h \circ a^{-1})||_{L^2(\Gamma)} \le ch.$$
 (12)

3 Numerical results for Laplace-Beltrami operator on Torus

For the coding the finite element method, we use the Python language and the open-source Linux based library CDMATH [11] which is very simple for the manipulation of large matrices, vectors, meshes and fields. It (CDMATH) can handle finite element and finite volume discretizations, read general 3D geometries and meshes generated by SALOME.

3.1 Meshing of the domain

For the design and meshing of the domain we use GEOMETRY and MESH modules of the software SALOME 9.5 (see [8, 10, 9]).

Below are the meshes used in our convergence analysis.

meshTorus 1	meshTorus 2	meshTorus 3	meshTorus 4
·			
1022 cells	6461 cells	20006 cells	43910 cells

Figure 2: Mesh of domain

3.2 Visualization of the results

For the numerical resolution of our discrete problem, we use an iterative solver because the stiffness matrix A_{\triangle_R} is large and sparse (see [7]).

because the stiffness matrix $A_{\triangle_{\Gamma_h}}$ is large and sparse (see [7]) . For the visualization of the result, we use the PARAVIS module included in SALOME (see [9]).

Below are visualizations of the numerical results obtained on the different meshes of picture 2.

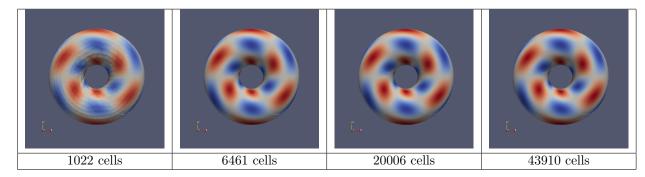


Figure 3: Numerical results of the finite elements on the torus

Below are clipings of the previous numerical results.

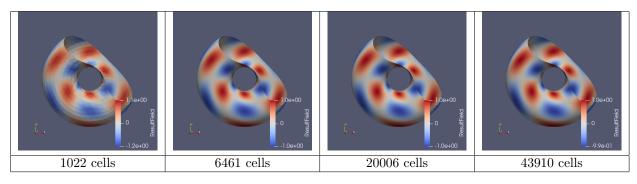


Figure 4: Cliping of the numerical result on the torus

3.3 Numerical convergence of the finite element method

Convergence of finite elements for Laplace operator on 3D torus triangular meshes log(|numerical-exact|) -0.6 least square slope : -1.655 -0.8 -1.0log(error) -1.2 -1.4-1.6-1.8-2.01.8 1.6 1.7 1.9 2.0 2.1 2.2 1.5 2.3 log(sqrt(number of nodes))

Figure 5: Convergence of the finite element method on the torus

The method converges with a numerical order of approximately 1.65.

3.4 Computational time of the finite element method

Computational time of finite elements for Laplace operator on 3D torus triangular meshes 2.0 log(cpu time)) 1.5 1.0 log(cpu time) 0.5 0.0 -0.5 1.7 1.9 2.0 2.1 2.2 1.5 1.6 1.8 2.3 log(sqrt(number of nodes)

Figure 6: Computational time of the finite element method on the torus

3.5 Ploting over slice circle

Here we have drawn a circle on each torus to extract the values. This circle is visible on the torus in Figure 3.

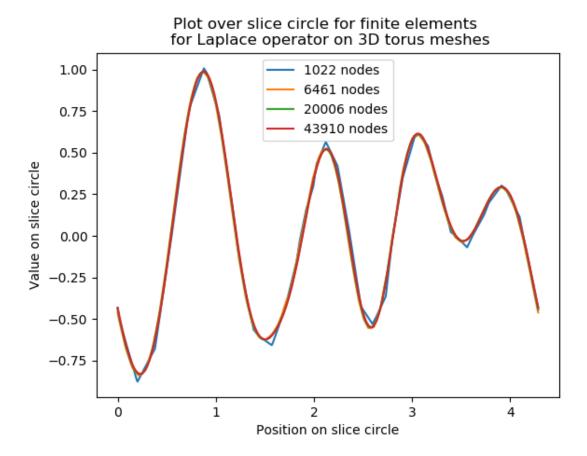


Figure 7: Convergence of the data ploted over a circle drawn on the torus

4 The script

```
Référence : M. A. Olshanskii, A. Reusken, and J.
8 #
       Grande. A finite element method for elliptic equations
                               on surfaces. SIAM J. Num. Anal., 47, p.
9 #
       3355
                   Résolution d'un système linéaire à matrice singuliè
10 #
      re : les vecteurs constants sont dans le noyau
12
13 import cdmath
14 from math import sin, cos, atan2, sqrt
15 import PV_routines
16 import VTK_routines
import paraview.simple as pvs
18
19 #Chargement du maillage triangulaire du tore
20 #
my_mesh = cdmath.Mesh("meshTorus.med")
if(not my_mesh.isTriangular()) :
    raise ValueError("Wrong cell types : mesh is not made of
      triangles")
if (my_mesh.getMeshDimension()!=2) :
   raise ValueError("Wrong mesh dimension : expected a surface of
      dimension 2")
if (my_mesh.getSpaceDimension()!=3) :
   raise ValueError("Wrong space dimension : expected a space of
      dimension 3")
nbNodes = my_mesh.getNumberOfNodes()
nbCells = my_mesh.getNumberOfCells()
print("Mesh building/loading done")
33 print("nb of nodes=", nbNodes)
34 print("nb of cells=", nbCells)
35
_{
m 36} # Torus radii (calculation will fail if the mesh is not correct)
37 R=1 #Grand rayon
38 r=0.6 #Petit rayon
40 #Discrétisation du second membre, de la solution exacte et dé
      termination des noeuds intérieurs
42 my_RHSfield = cdmath.Field("RHS field", cdmath.NODES, my_mesh, 1)
43 exactSolField = cdmath.Field("Exact solution field", cdmath.NODES,
      my_mesh, 1)
44
_{\rm 45} maxNbNeighbours = 0#This is to determine the number of non zero
      coefficients in the sparse finite element rigidity matrix
46
47 #parcours des noeuds pour discrétisation du second membre et
     extraction du nb max voisins d'un noeud
48 for i in range(nbNodes):
   Ni=my_mesh.getNode(i)
49
   x = Ni.x()
50
y = Ni.y()
z = Ni.z()
```

53

```
theta=atan2(z, sqrt(x*x+y*y)-R)
     phi=atan2(y,x)
55
56
     exactSolField[i] = sin(3*phi)*cos(3*theta+ phi) # for the exact
57
       solution we use the funtion given in the article of Olshanskii,
        Reusken 2009, page 19
     my_RHSfield[i] = 9*sin(3*phi)*cos(3*theta+phi)/(r*r) + (10*sin)
       (3*phi)*cos(3*theta+ phi) + 6*cos(3*phi)*sin(3*theta+ phi))/((R
       +r*cos(theta))*(R+r*cos(theta))) - 3*sin(theta)*sin(3*phi)*sin
       (3*theta+ phi)/(r*(R+r*cos(theta))) #for the right hand side we
       use the function given in the article of Olshanskii, Reusken
       2009, page 19
     if my_mesh.isBorderNode(i): # Détection des noeuds frontière
59
      raise ValueError("Mesh should not contain borders")
60
61
       maxNbNeighbours = max(1+Ni.getNumberOfCells(), maxNbNeighbours)
62
       #true only for planar cells, otherwise use function Ni.
       getNumberOfEdges()
63
64 print("Right hand side discretisation done")
print("Max nb of neighbours=", maxNbNeighbours)
print("Integral of the RHS", my_RHSfield.integral(0))
68 # Construction de la matrice de rigidité et du vecteur second
       membre du système linéaire
70 Rigidite = cdmath.SparseMatrixPetsc(nbNodes, nbNodes, maxNbNeighbours)
71 RHS=cdmath. Vector(nbNodes)
73 # Vecteurs gradient de la fonction de forme associée à chaque noeud
        d'un triangle
74 GradShapeFuncO=cdmath.Vector(3)
75 GradShapeFunc1=cdmath.Vector(3)
76 GradShapeFunc2=cdmath.Vector(3)
78 normalFace0=cdmath.Vector(3)
79 normalFace1=cdmath.Vector(3)
80
81 #On parcourt les triangles du domaine
82 for i in range(nbCells):
83
     Ci=my_mesh.getCell(i)
84
85
     #Contribution à la matrice de rigidité
86
     nodeId0=Ci.getNodeId(0)
     nodeId1=Ci.getNodeId(1)
88
     nodeId2=Ci.getNodeId(2)
89
     NO=my_mesh.getNode(nodeIdO)
90
     N1=my_mesh.getNode(nodeId1)
91
92
     N2=my_mesh.getNode(nodeId2)
93
     #Build normal to cell Ci
94
     normalFace0[0] = Ci.getNormalVector(0,0)
95
     normalFace0[1]=Ci.getNormalVector(0,1)
96
     normalFace0[2]=Ci.getNormalVector(0,2)
97
     normalFace1[0] = Ci.getNormalVector(1,0)
98
     normalFace1[1]=Ci.getNormalVector(1,1)
99
100
     normalFace1[2]=Ci.getNormalVector(1,2)
   normalCell = normalFace0.crossProduct(normalFace1)
102
```

```
normalCell = normalCell/normalCell.norm()
103
     cellMat=cdmath.Matrix(4)
105
106
     cellMat[0,0]=N0.x()
     cellMat[0,1]=N0.y()
107
     cellMat[0,2]=N0.z()
108
109
     cellMat[1,0]=N1.x()
     cellMat[1,1]=N1.y()
110
     cellMat[1,2]=N1.z()
111
     cellMat[2,0]=N2.x()
112
     cellMat[2,1]=N2.y()
113
     cellMat[2,2]=N2.z()
114
     cellMat[3,0]=normalCell[0]
     cellMat[3,1]=normalCell[1]
116
117
     cellMat[3,2]=normalCell[2]
     cellMat[0,3]=1
118
     cellMat[1.3]=1
119
     cellMat[2,3]=1
120
     cellMat[3,3]=0
122
     #Formule des gradients voir EF P1 -> calcul déterminants
123
     {\tt GradShapeFunc0[0]=cellMat.partMatrix(0,0).determinant()/2}
124
     GradShapeFunc0[1] = - cellMat.partMatrix(0,1).determinant()/2
125
     GradShapeFunc0[2] = cellMat.partMatrix(0,2).determinant()/2
126
     {\tt GradShapeFunc1[0] = -cellMat.partMatrix(1,0).determinant()/2}
127
128
     GradShapeFunc1[1] = cellMat.partMatrix(1,1).determinant()/2
     GradShapeFunc1[2] = -cellMat.partMatrix(1,2).determinant()/2
129
130
     {\tt GradShapeFunc2} \ [0] = \ {\tt cellMat.partMatrix} \ (2\,,0) \ . \ {\tt determinant} \ () \ / 2
     GradShapeFunc2[1] = -cellMat.partMatrix(2,1).determinant()/2
131
     GradShapeFunc2[2] = cellMat.partMatrix(2,2).determinant()/2
132
133
     #Création d'un tableau (numéro du noeud, gradient de la fonction
134
      de forme
     GradShapeFuncs={nodeId0 : GradShapeFunc0}
135
     GradShapeFuncs [nodeId1] = GradShapeFunc1
136
     GradShapeFuncs[nodeId2] = GradShapeFunc2
137
138
     # Remplissage de la matrice de rigidité et du second membre
139
     for j in [nodeId0, nodeId1, nodeId2] :
140
       #Ajout de la contribution de la cellule triangulaire i au
141
       second membre du noeud j
       RHS[j]=Ci.getMeasure()/3*my_RHSfield[j]+RHS[j] # intégrale dans
        le triangle du produit f x fonction de base
       \#Contribution de la cellule triangulaire i à la ligne j du syst
143
       ème linéaire
       for k in [nodeId0, nodeId1, nodeId2] :
144
         Rigidite.addValue(j,k,GradShapeFuncs[j]*GradShapeFuncs[k]/Ci.
       getMeasure())
147 print("Linear system matrix building done")
148
149 # Résolution du système linéaire
150 #:
LS=cdmath.LinearSolver(Rigidite,RHS,100,1.E-6, "GMRES", "ILU")#
       Remplacer CG par CHOLESKY pour solveur direct
LS.setMatrixIsSingular()#En raison de l'absence de bord
153 SolSyst=LS.solve()
print( "Preconditioner used : ", LS.getNameOfPc() )
print( "Number of iterations used : ", LS.getNumberOfIter() )
print( "Final residual : ", LS.getResidu() )
print("Linear system solved")
158
```

```
159 # Création du champ résultat
my_ResultField = cdmath.Field("Numerical result field", cdmath.
       NODES, my_mesh, 1)
for j in range(nbNodes):
      my_ResultField[j]=SolSyst[j];#remplissage des valeurs pour les
163
       noeuds intérieurs
164 #sauvegarde sur le disque dur du résultat dans un fichier paraview
my_ResultField.writeVTK("FiniteElementsOnTorusPoisson")
167 print("Integral of the numerical solution", my_ResultField.integral
       (0))
168 print("Numerical solution of Poisson equation on a torus using
      finite elements done")
170 #Calcul de l'erreur commise par rapport à la solution exacte
171 # =
172 max_sol_exacte=exactSolField.getNormEuclidean().max()
173 erreur_max=(exactSolField - my_ResultField).getNormEuclidean().max
       ()
174 max_sol_num=my_ResultField.max()
175 min_sol_num=my_ResultField.min()
177 print("Relative error = max(| exact solution - numerical solution
       |)/max(| exact solution |) = ",erreur_max/max_sol_exacte)
178 print("Maximum numerical solution = ", max_sol_num, " Minimum
      numerical solution = ", min_sol_num)
print("Maximum exact solution = ", exactSolField.max(), " Minimum
       exact solution = ", exactSolField.min())
180
181 #Postprocessing:
182 #========
183 # Save 3D picture
PV_routines.Save_PV_data_to_picture_file("
       FiniteElementsOnTorusPoisson"+'_0.vtu',"Numerical result field"
       ,'NODES', "FiniteElementsOnTorusPoisson")
resolution=100
186 VTK_routines.Clip_VTK_data_to_VTK("FiniteElementsOnTorusPoisson"+'
       _O.vtu',"Clip_VTK_data_to_VTK_"+ "FiniteElementsOnTorusPoisson"
       +'_0.vtu',[0.25,0.25,0.25], [-0.5,-0.5,-0.5],resolution)
PV_routines.Save_PV_data_to_picture_file("Clip_VTK_data_to_VTK_"+"
       FiniteElementsOnTorusPoisson"+',_O.vtu', "Numerical result field"
       ,'NODES',"Clip_VTK_data_to_VTK_"+"FiniteElementsOnTorusPoisson"
189 # Plot over slice circle
190 finiteElementsOnTorus_Ovtu = pvs.XMLUnstructuredGridReader(FileName
       =["FiniteElementsOnTorusPoisson"+'_0.vtu'])
slice1 = pvs.Slice(Input=finiteElementsOnTorus_Ovtu)
192 slice1.SliceType.Normal = [0.5, 0.5, 0.5]
renderView1 = pvs.GetActiveViewOrCreate('RenderView')
194 finiteElementsOnTorus_OvtuDisplay = pvs.Show(
      finiteElementsOnTorus_Ovtu, renderView1)
pvs.ColorBy(finiteElementsOnTorus_OvtuDisplay, ('POINTS', '
       Numerical result field'))
slice1Display = pvs.Show(slice1, renderView1)
197 pvs.SaveScreenshot("./FiniteElementsOnTorusPoisson"+"_Slice"+'.png')
       , magnification=1, quality=100, view=renderView1)
plotOnSortedLines1 = pvs.PlotOnSortedLines(Input=slice1)
199 lineChartView2 = pvs.CreateView('XYChartView')
200 plotOnSortedLines1Display = pvs.Show(plotOnSortedLines1,
     lineChartView2)
```

References

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