SfePy - Simple Finite Elements in Python

Short Introduction ...

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EuroSciPy 2008, July 27, Leipzig, Germany

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   - Notes on Programming Languages
2. Our choice
   - Mixing Languages — Best of Both Worlds
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   - Shape Optimization in Incompressible Flow Problems
   - Finite Element Formulation of Schrödinger Equation
6. Conclusion

Introduction

- SfePy = simple finite elements in Python
  - general finite element analysis software
  - solving systems of PDEs
- BSD open-source license
- available at
  - `http://sfepy.org` (developers)
    - mailing lists, issue (bug) tracking
    - we encourage and support everyone who joins!
  - `http://sfepy.kme.zcu.cz` (project information)
- selected applications:
  - homogenization of porous media (parallel flows in a deformable porous medium)
  - acoustic band gaps (homogenization of a strongly heterogenous elastic structure: phononic materials)
  - shape optimization in incompressible flow problems
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Notes on Programming Languages

Rough Division

- compiled (fortran, C, C++, Java, . . .)

Pros

- speed
- large code base (legacy codes)
- tradition

Cons

- (often) complicated build process, recompile after any change
- low-level ⇒ lots of lines to get basic stuff done
- code size ⇒ maintenance problems
- static!

- interpreted or scripting (sh, tcl, matlab, perl, ruby, python, . . .)

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- no compiling
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Cons

- many are relatively new
- not known very well in many scientific communities
- slow!

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Mixing Languages — Best of Both Worlds

- low level code (C or fortran): element matrix evaluations, costly mesh-related functions, ...
- high level code (Python): logic of the code, particular applications, configuration files, problem description files

www.python.org

SfePy = Python + C (+ fortran)

- notable features:
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  - fast compilation
  - problem description files in pure Python
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Software Dependencies

- to install and use SfePy, several other packages or libraries are needed:
  - NumPy and SciPy: free (BSD license) collection of numerical computing libraries for Python
    - enables Matlab-like array/matrix manipulations and indexing
  - other: UMFPACK, Pyparsing, Matplotlib, Pytables (+ HDF5), swig
  - visualization of results: ParaView, MayaVi2, or any other VTK-capable viewer
- missing:
  - free (BSD license) 3D mesh generation and refinement tool
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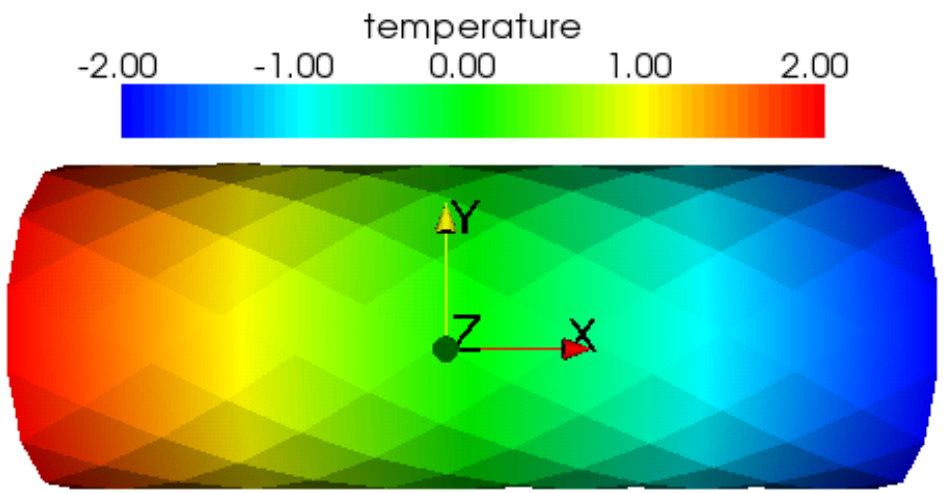
Introduction

- problem description file is a regular Python module, i.e. all Python syntax and power is accessible
- consists of entities defining:
  - fields of various FE approximations, variables
  - equations in the weak form, quadratures
  - boundary conditions (Dirichlet, periodic, "rigid body")
  - FE mesh file name, options, solvers, ...
- simple example: the Laplace equation:

$$c\Delta u = 0 \text{ in } \Omega, \quad u = \bar{u} \text{ on } \Gamma, \text{ weak form: } \int_\Omega c\, \nabla u \cdot \nabla v = 0, \quad \forall v \in V_0$$



Problem Description File

Solving Laplace Equation — FE Approximations

- mesh → define FE approximation to Ω:

```
filename_mesh = 'simple.mesh'
```

- fields → define space V_h:

```
field_1   = {
        'name'     :  'temperature',
        'dim'      :  (1,1),
        'domain'   :  'Omega',
        'bases'    :  'Omega' :  '3_4_P1'
}
```

'3_4_P1' means P1 approximation, in 3D, on 4-node FEs (tetrahedra)

- variables → define u_h, v_h:

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variables   = {
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Problem Description File

Solving Laplace Equation — Boundary Conditions

- regions $\rightarrow$ define domain Ω, regions Γ_{left}, Γ_{right}, $\Gamma = \Gamma_{\text{left}} \cup \Gamma_{\text{right}}$:
  - $_h$ omitted from now on ...

```
regions   = {
        'Omega'         :  ('all', {}),
        'Gamma_Left'    :  ('nodes in (x < 0.0001)', {}),
        'Gamma_Right'   :  ('nodes in (x > 0.0999)', {}),
}
```

- Dirichlet BC $\rightarrow$ define $\bar{u}$ on Γ_{left}, Γ_{right}:

```
ebcs   = {
        't_left'    :  ('Gamma_Left', 'u.0' :  2.0),
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Problem Description File

Solving Laplace Equation — Equations

- materials → define c:

```
material_1   = {
                'name'     :  'm',
                'mode'     :  'here',
                'region'   :  'Omega',
                'c'        :  1.0,
               }
```

- integrals → define numerical quadrature:

```
integral_1   = {
                'name'         :  'i1',
                'kind'         :  'v',
                'quadrature'   :  'gauss_o1_d3',
               }
```

- equations → define what and where should be solved:

```
equations   = {
               'eq'   :  'dw_laplace.i1.Omega( m.c, v, u ) = 0'
              }
```

Problem Description File

Solving Laplace Equation — Equations

- materials → define c:

```
material_1 = {
    'name'   : 'm',
    'mode'   : 'here',
    'region' : 'Omega',
    'c'      : 1.0,
}
```

- integrals → define numerical quadrature:

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integral_1 = {
    'name'      : 'i1',
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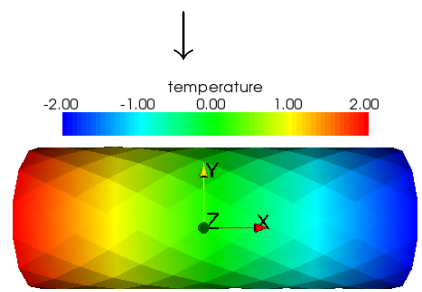
Running SfePy

```
$ ./simple.py input/poisson.py
sfepy: reading mesh...
sfepy: ...done in 0.02 s
sfepy: setting up domain edges...
sfepy: ...done in 0.02 s
sfepy: setting up domain faces...
sfepy: ...done in 0.02 s
sfepy: creating regions...
sfepy:     leaf Gamma_Right region_4
sfepy:     leaf Omega region_1000
sfepy:     leaf Gamma_Left region_03
sfepy: ...done in 0.07 s
sfepy: equation "Temperature":
sfepy: dw_laplace.i1.Omega( coef.val, s, t ) = 0
sfepy: describing geometries...
sfepy: ...done in 0.01 s
sfepy: setting up dof connectivities...
sfepy: ...done in 0.00 s
sfepy: using solvers:
                nls: newton
                 ls: ls
sfepy: matrix shape: (300, 300)
sfepy: assembling matrix graph...
sfepy: ...done in 0.01 s
sfepy: matrix structural nonzeros: 3538 (3.93e-02% fill)
sfepy: updating materials...
sfepy:     coef
sfepy: ...done in 0.00 s
sfepy: nls: iter: 0, residual: 1.176265e-01 (rel: 1.000000e+00)
sfepy:   rezidual:     0.00 [s]
sfepy:      solve:     0.01 [s]
sfepy:     matrix:     0.00 [s]
sfepy: nls: iter: 1, residual: 9.921082e-17 (rel: 8.434391e-16)
```

- top level of SfePy code is a collection of executable scripts tailored for various applications
- `simple.py` is dumb script of brute force, attempting to solve any equations it finds by the Newton method
- . . . exactly what we need here (solver options were omitted in previous slides)



Top-level Scripts

Main scripts / applications:

- `runTests.py` ... run all/selected unit tests
- `simple.py` ... generic problem solver, both for stationary and time-dependent problems
- `eigen.py` ... application: acoustic band gaps in strongly heterogenous media
- `schroedinger.py` ... application: Schrödinger equation solver

Auxiliary:

- `extractor.py` ... extract results stored in a HDF5 file, dump results to VTK
- `findSurf.py` ... extract a mesh surface, mark its components
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Verification of Numerical Results

- to verify numerical results we use method of manufactured solutions: for example, for Poisson's equation $\mathrm{div}(\mathrm{grad}(u)) = f$:
  1. make up a solution, e.g. $u = \sin 3x \cos 4y$
  2. compute corresponding f, here $f = 25 \sin 3x \cos 4y$, and boundary conditions by substituting u into the equation
  3. solve numerically and compare the exact solution of the strong problem with the numerical solution of the weak problem

     $\rightarrow$ allows to assess both the discretization and numerical errors
- manual derivation of f tedious $\rightarrow$ SymPy
  - each term class annotated by a corresponding symbolic expression
  - example: anisotropic diffusion term

```
symbolic = {   'expression':  'div( K * grad( u ) )',
               'map' :  {'u' :  'state', 'K' : 'material'}}
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  - work in progress

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Optimal Flow Problem

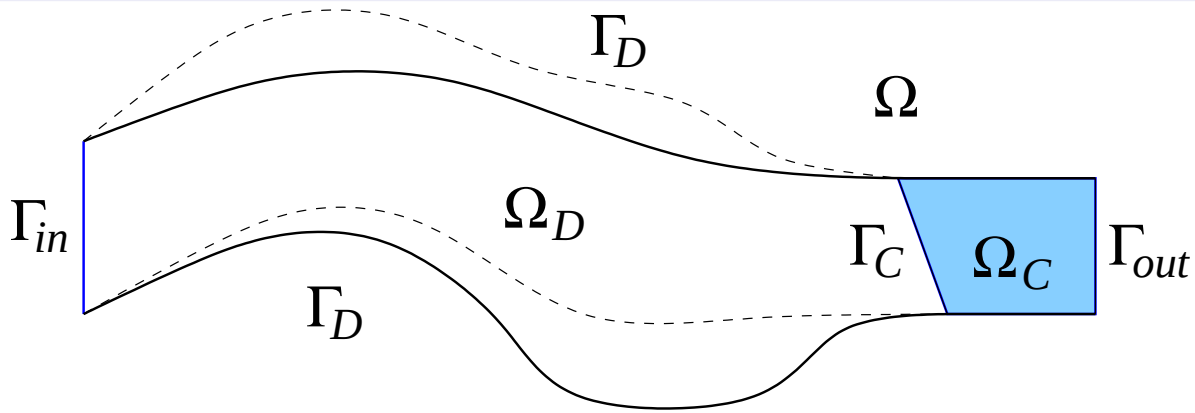
Problem Setting

Objective Function

$$\Psi(u) \equiv \frac{\nu}{2} \int_{\Omega_c} |\nabla u|^2 \longrightarrow \min$$

- minimize gradients of solution (e.g. losses) in $\Omega_c \subset \Omega$
- by moving design boundary $\Gamma \subset \partial\Omega$
- perturbation of Γ by vector field $\mathcal{V}$

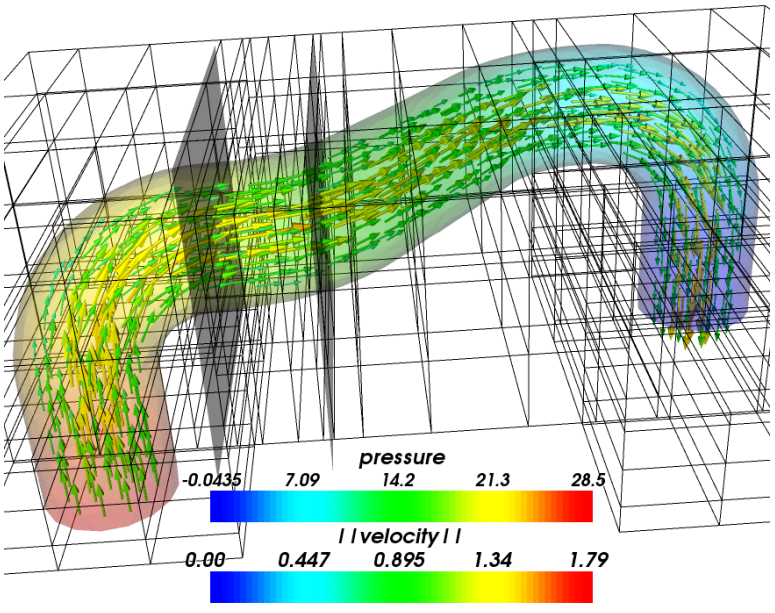
$$\Omega(t) = \Omega + \{t\mathcal{V}(x)\}_{x\in\Omega} \quad \text{where } \mathcal{V} = 0 \text{ in } \bar{\Omega}_c \cup \partial\Omega \setminus \Gamma$$

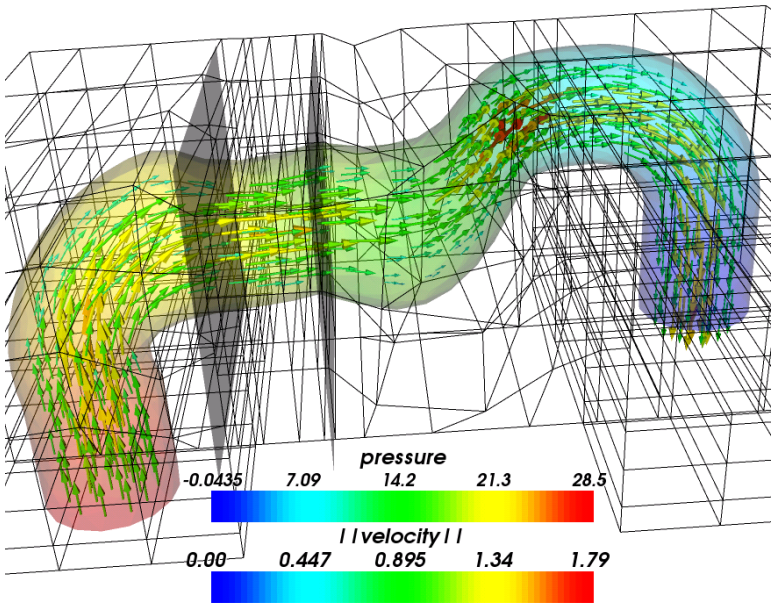


Optimal Flow Problem

Example Results

- flow and domain control boxes, left: initial, right: final





- Ω_C between two grey planes
- work in progress . . .

Direct Problem

. . . paper $\leftrightarrow$ input file

- weak form of Navier-Stokes equations: ? $\mathbf{u} \in \mathbf{V}_0(\Omega)$, $p \in L^2(\Omega)$ such that

$$\begin{aligned} a_\Omega\left(\mathbf{u}, \mathbf{v}\right) + c_\Omega\left(\mathbf{u}, \mathbf{u}, \mathbf{v}\right) - b_\Omega\left(\mathbf{v}, p\right) &= g_{\Gamma_{\text{out}}}\left(\mathbf{v}\right) \quad \forall \mathbf{v} \in \mathbf{V}_0 \ , \\ b_\Omega\left(\mathbf{u}, q\right) &= 0 \quad \forall q \in L^2(\Omega) \ . \end{aligned} \tag{1}$$

- in SfePy syntax:

```
equations = {
'balance'           :  """
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                       + dw_convect.i2.Omega( v, u )
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Adjoint Problem

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- KKT conditions $\delta_{\mathbf{u},p}\mathcal{L} = 0$ yield adjoint state problem for $\mathbf{w}$, r:

$$\delta_{\mathbf{u}}\mathcal{L} \circ \mathbf{v} = 0 = \delta_u \Psi(\mathbf{u}, p) \circ \mathbf{v} + a_\Omega(\mathbf{v}, \mathbf{w}) + c_\Omega(\mathbf{v}, \mathbf{u}, \mathbf{w}) + c_\Omega(\mathbf{u}, \mathbf{v}, \mathbf{w}) + b_\Omega(\mathbf{v}, r) \ ,$$
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Finite Element Formulation of Schrödinger Equation

One particle Schrödinger equation:

$$\left(-\frac{\hbar^2}{2m}\nabla^2 + V\right)\psi = E\psi\,.$$

FEM:

$$(K_{ij} + V_{ij})\,q_j = EM_{ij}q_j + F_i\,,$$

$$\begin{aligned} V_{ij} &= \int \phi_i V \phi_j \,\mathrm{d}V\,, \\ M_{ij} &= \int \phi_i \phi_j \,\mathrm{d}V\,, \\ K_{ij} &= \frac{\hbar^2}{2m}\int \nabla\phi_i \cdot \nabla\phi_j \,\mathrm{d}V\,, \\ F_i &= \frac{\hbar^2}{2m}\oint \frac{\mathrm{d}\psi}{\mathrm{d}n}\phi_i \,\mathrm{d}S\,. \end{aligned}$$

Usually we set $F_i = 0$.

Particle in the Box

$$V(x) = \begin{cases} 0, & \text{inside the box} \quad a \times a \times a \\ \infty, & \text{outside} \end{cases}$$

Analytic solution:

$$E_{n_1 n_2 n_3} = \frac{\pi^2}{2a^2}\left(n_1^2 + n_2^2 + n_3^2\right)$$

where $n_i = 1, 2, 3, \ldots$ are independent quantum numbers. We chose $a = 1$, i.e.: $E_{111} = 14.804$, $E_{211} = E_{121} = E_{112} = 29.608$, $E_{122} = E_{212} = E_{221} = 44.413$, $E_{311} = E_{131} = E_{113} = 54.282$ $E_{222} = 59.217$, $E_{123} = E_{\text{perm.}} = 69.087$.

Numerical solution ($a = 1$, 24702 nodes):

E	1	2-4	5-7	8-10	11	12-
theory	14.804	29.608	44.413	54.282	59.217	69.087
FEM	14.861	29.833	44.919	55.035	60.123	70.305
		29.834	44.920	55.042		70.310
		29.836	44.925	55.047		...

3D Harmonic Oscillator

$$V(r) = \begin{cases} \frac{1}{2}\omega^2 r^2, & \text{inside the box} \quad a \times a \times a \\ \infty, & \text{outside} \end{cases}$$

Analytic solution in the limit $a \to \infty$:

$$E_{nl} = \left(2n + l + \frac{3}{2}\right)\omega$$

where $n, l = 0, 1, 2, \ldots$. Degeneracy is $2l + 1$, so: $E_{00} = \frac{3}{2}$, triple $E_{01} = \frac{5}{2}$, $E_{10} = \frac{7}{2}$, quintuple $E_{02} = \frac{7}{2}$ triple $E_{11} = \frac{9}{2}$, quintuple $E_{12} = \frac{11}{2}$:

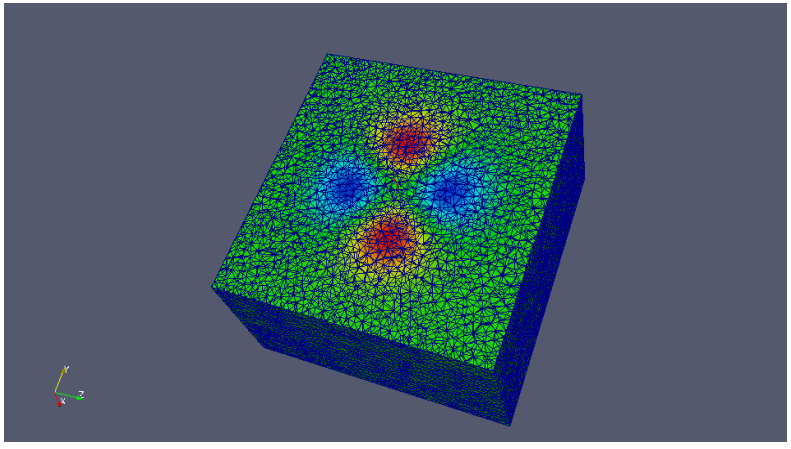
Numerical solution ($a = 15$, $\omega = 1$, 290620 nodes):

E	1	2-4	5-10	11-
theory	1.5	2.5	3.5	4.5
FEM	1.522	2.535	3.554	4.578
		2.536	3.555	4.579
		2.536	3.555	4.579
			3.555	...
			3.556	
			3.556	

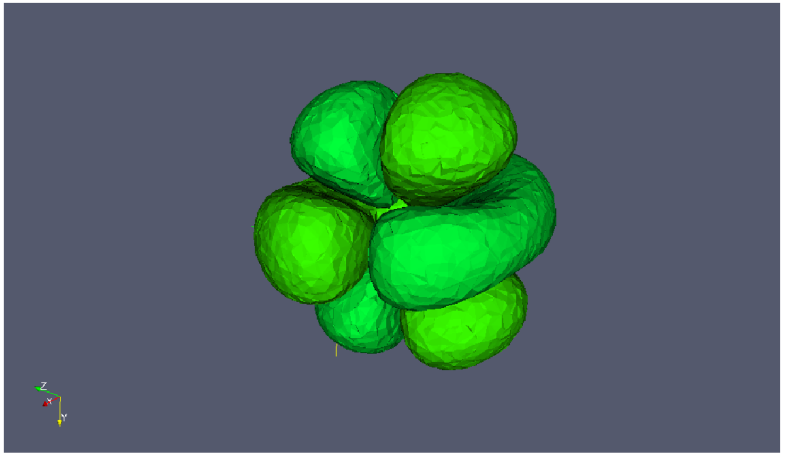
3D Harmonic Oscillator

Eigenvectors:

0th



12th



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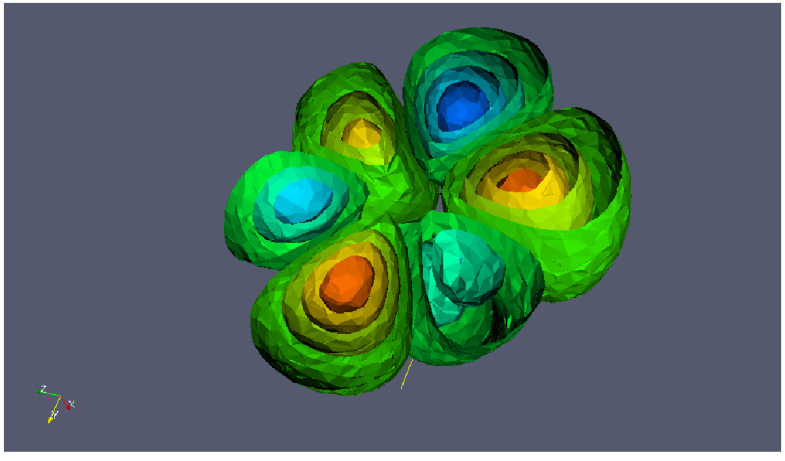
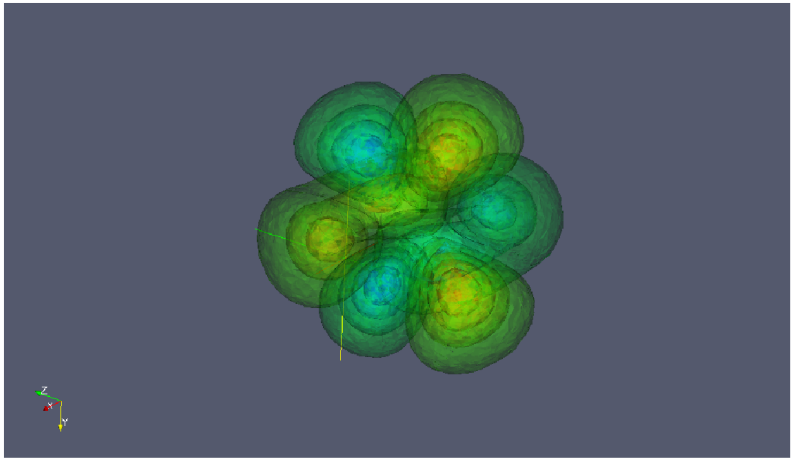


12th



Hydrogen Atom

$$V(r) = \begin{cases} -\frac{1}{r}, & \text{inside the box} \quad a \times a \times a \\ \infty, & \text{outside} \end{cases}$$

Analytic solution in the limit $a \to \infty$:

$$E_n = -\frac{1}{2n^2}$$

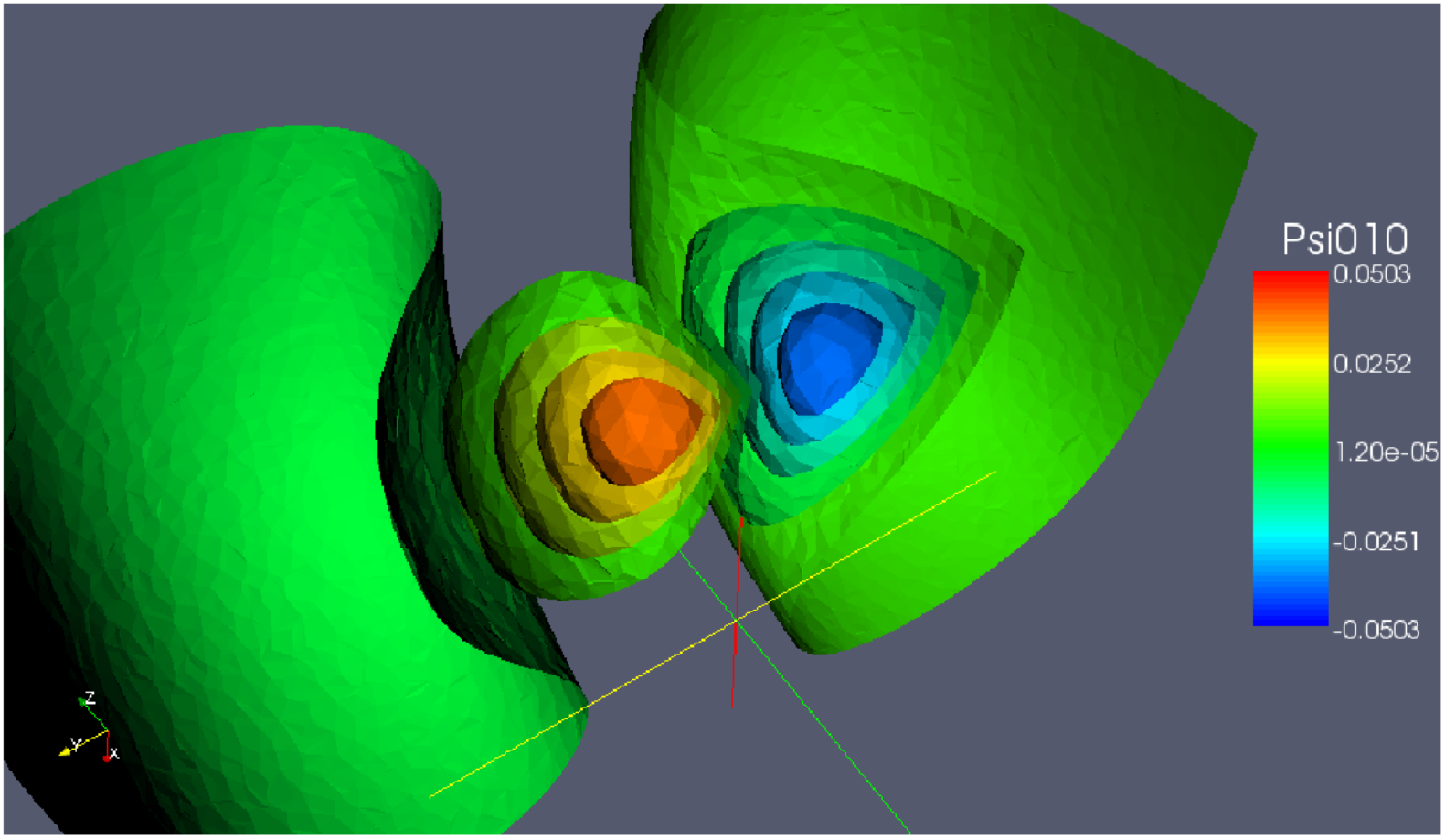
where $n = 1, 2, 3, \ldots$. Degeneracy is n^2, so: $E_1 = -\frac{1}{2} = -0.5$, $E_2 = -\frac{1}{8} = -0.125$, $E_3 = -\frac{1}{18} = -0.055$, $E_4 = -\frac{1}{32} = -0.031$.

Numerical solution ($a = 15$, 160000 nodes):

E	1	2-5	6-14	15-
theory	-0.5	-0.125	-0.055	-0.031
FEM	-0.481	-0.118	-0.006	$\cdots$

Hydrogen Atom

11th eigenvalue (calculated: -0.04398532, exact: -0.056), on the mesh with 976 691 tetrahedrons and 163 666 nodes, for the hydrogen atom (V=-1/r).



Conclusion

What is done

- basic FE element engine:
  - finite-dimensional approximations of continuous fields
  - variables, boundary conditions, FE assembling
  - equations, terms, regions
  - materials, material caches
- various solvers accessed via abstract interface
- unit tests, automatic documentation generation
- mostly linear problems, but multiphysical

What is not done

- general FE engine, possibly with symbolic evaluation (SymPy)
- good documentation
- fast problem-specific solvers (!)
- adaptive mesh refinement (!)
- parallelization (petsc4py)

What will not be done (?)

- GUI
- real symbolic parsing/evaluation of equations

http://sfepy.org

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Yes, the final slide!

Acknowledgements

The work on various parts of SfePy has been supported by the following grants and research projects funded by several agencies in the Czech Republic:

- Robert Cimrman:
  - grant project GAČR 101/07/1471, entitled "Finite element modelling of linear, non-linear and multiscale effects in wave propagation in solids and heterogeneous media"
  - research project MŠMT 1M06031
  - research project MŠMT 4977751303
- Ondřej Čertík:
  - research center project LC06040
  - grant project GAČR IAA100100637

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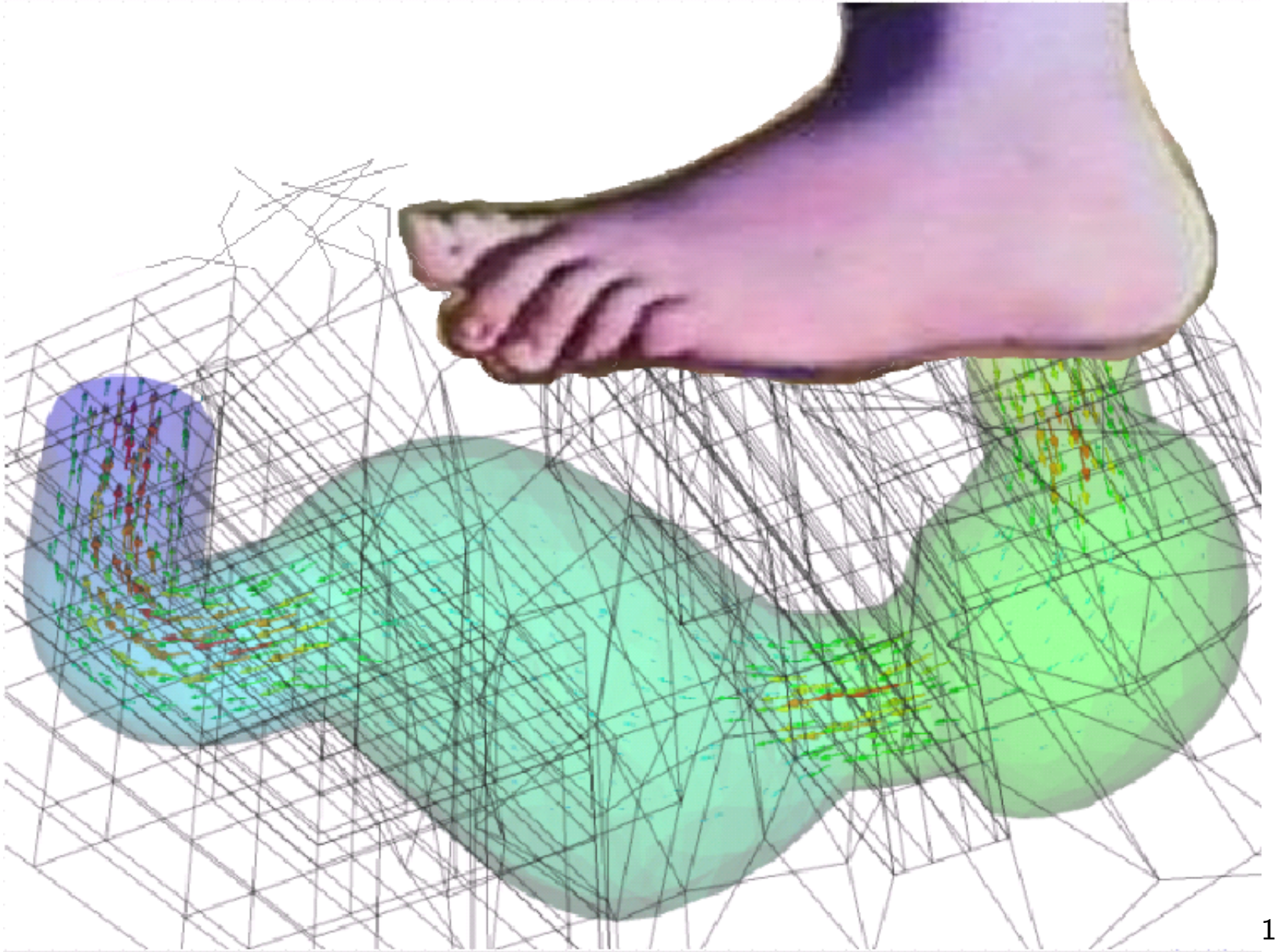

[1]

[1] Do you like Monty Python's Flying Circus? It helps! (Python FAQ 1.1.17)