

OUTLINE

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The FEniCS project

The FEniCS Project is a collaborative project set out in 2003 for the development of innovative concepts and tools for automated scientific computing, with a particular focus on automated solution of differential equations by finite element method.

- Automated generation of basis functions
- Automated evaluation of variational forms
- Automated finite element assembly
- Automated adaptive error control

FEniCS is a user-friendly tool for solving partial differential equations (PDEs).

The FEniCS Project is developed by researchers from a number of research institutes from around the world. The research institutes mainly involved are:

- Simula Research Laboratory (Oslo)
- University of Cambridge
- University of Chicago
- Baylor University
- KTH Royal Institute of Technology

The FEniCS project – Useful Information

Website FEniCS project:

http://fenicsproject.org/

Installation:

FEniCS is included as part of Ubuntu GNU/Linux (starting with 10.04/Lucid). Installing procedure from the Ubuntu Software Center.

FEniCS can also be installed by running the following command in a terminal:

sudo apt-get install fenics

Documentation:

Python and C++ demos already installed in the system

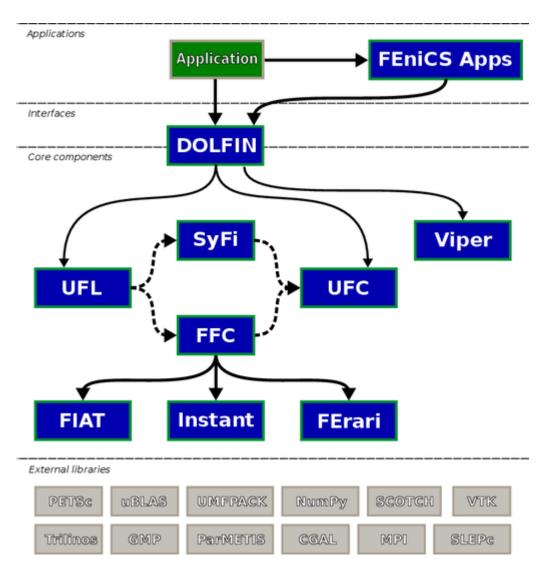
FEniCS book "Automated Solution of Differential Equations by the Finite Element Method - The FEniCS Book"

pydoc dolfin.X to look up documentation of a Python class X from the DOLFIN library

Question and Answers:

(old) https://launchpad.net/fenics

(new) http://fenicsproject.org/qa/



FEniCS is organized as a collection of interoperable components that together form the FEniCS Project.

DOLFIN: the problem-solving environment; it is a C++ library, with a Python interface.

FFC: form compiler.

FIAT: the finite element tabulator.

Instant: the just-in-time compiler.

UFC: the code generation interface.

UFL: the form language; a component implementing the unified form language for specifying finite element forms.

Viper: a component for quick visualization of finite element meshes and solutions.

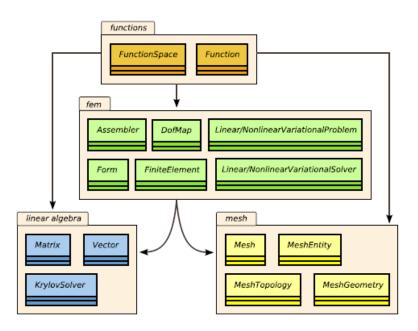
Python Interface

CLASS: a programming construction for creating objects containing a set of variables and functions. Most types of FEniCS objects are defined through the class concept.

INSTANCE: an object of a particular type, where the type is implemented as a class. For instance, mesh = UnitInterval(10) creates an instance of class UnitInterval, which is reached by the name mesh. (Class UnitInterval is actually just an interface to a corresponding C++ class in the DOLFIN C++ library.)

CLASS ATTRIBUTE: a variable in a class, reached by dot notation: instance_name.attribute_name.

Most important components and classes in FEniCS, along with their dependencies indicated by arrows.

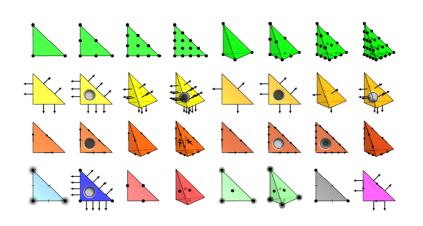


FEATURES

ONLY TRIANGLE AND TETRAHEDRON

FULLY SUPPORTED in BLACK PARTLY SUPPORTED in GREY

Name	Symbol
Argyris	ARG
Arnold–Winther	AW
Brezzi-Douglas-Marini	BDM
Crouzeix-Raviart	CR
Discontinuous Lagrange	DG
Hermite	HER
Lagrange	CG
Mardal-Tai-Winther	MTW
Morley	MOR
Nédélec 1st kind H(curl)	N1curl
Nédélec 2nd kind H(curl)	N2curl
Raviart–Thomas	RT



THIRD-PARTS LIBRARY for LINEAR ALGEBRA BACKENDS

- vectors;
- dense/sparse matrices;
- direct/iterative solvers;
- eigenvalue solver;

Depending on the specific algebra backend

```
list_lu_solver_methods()
list_krylov_solver_methods()
list_krylov_solver_preconditioners()
```

Linear algebra backends PETSc Trilinos/Epetra uBLAS MTL4

```
>>> list_lu_solver_methods()
LU method | Description
default
              default LU solver
umfpack
              UMFPACK (Unsymmetric MultiFrontal sparse LU factorization)
              MUMPS (MUltifrontal Massively Parallel Sparse direct Solver)
mumps
              SPOOLES (SParse Object Oriented Linear Equations Solver)
spooles
              PETSc builtin LU solver
petsc
>>> list_krylov_solver_methods()
Krylov method | Description
default
                  default Krylov method
                  Conjugate gradient method
cg
                  Generalized minimal residual method
gmres
                  Minimal residual method
minres
                  Transpose-free quasi-minimal residual method
tfamr
                  Richardson method
richardson
                  Biconjugate gradient stabilized method
bicgstab
```

FEATURES

Linear solv	er	Preconditioners	
lu	Sparse lu factorization	ilu	Incomplete LU factorization
cholesky	Sparse Cholesky factorization	icc	Incomplete Cholesky factorization
cg	Conjugate gradient method	jacobi	Jacobi iteration
gmres	Generalized minimal residual method	bjacobi	Block Jacobi iteration
bicgstab	Biconjugate gradient stabilized method	sor	Successive over-relaxation
minres	Minimal residual method	amg	Algebraic multigrid (BoomerAMG or ML)
tfgmr	Transpose-free quasi-minimal residual method	additive_schwartz	Additive Schwarz
richardson	Richardson method	hypre_amg	Hypre algebraic multigrid (BoomerAMG)
		hypre_euclid	Hypre parallel incomplete LU factorization
		hypre_parasails	Hypre parallel sparse approximate inverse
		ml_amg	ML algebraic multigrid

The Unified Form Language – UFL (Alnæs and Logg, 2009) – is a domain specific language for the declaration of finite element discretizations of variational forms and functionals. More precisely, the language defines a flexible user interface for defining finite element spaces and expressions for weak forms in a notation close to mathematical notation.

- ☐ A richer form language, especially for expressing nonlinear PDEs.
- ☐ Automatic differentiation of expressions and forms.
- ☐ Improving the performance of the form compiler technology to handle more complicated equations efficiently.

$$F = I + \operatorname{grad} u$$

$$C = F^{T} F$$

$$I_{C} = \operatorname{tr}(C)$$

$$II_{C} = \frac{1}{2} [I_{C}^{2} - \operatorname{tr}(CC)]$$

$$W = c_{1}(I_{C} - 3) + c_{2}(II_{C} - 3)$$

$$S = 2 \frac{\partial W}{\partial C}$$

Find u such that div(FS) = 0 in Ω + boundary conditions

The non linear variational problem reads

$$L(u, \phi) = \int_{\Omega} FS: grad \ \phi dx \ \forall \ \phi \in V$$

```
# Form arguments
phi\theta = TestFunction(element)
phi1 = TrialFunction(element)
u = Coefficient(element)
c1 = Constant(cell)
c2 = Constant(cell)
# Deformation gradient Fij = dXi/dxj
I = Identity(cell.d)
F = I + grad(u)
# Right Cauchy-Green strain tensor C with invariants
C = variable(F.T*F)
I C = tr(C)
II_C = (I_C **2 - tr(C*C))/2
# Mooney-Rivlin constitutive law
W = c1*(I_C-3) + c2*(II_C-3)
# Second Piola-Kirchoff stress tensor
S = 2*diff(W, C)
# Weak forms
L = inner(F*S, grad(phiθ))*dx
```

```
dot(a,b): \mathbf{v} \cdot \mathbf{u} = v_i u_i, \mathbf{A} \cdot \mathbf{u} = A_{ij} u_j e_i, \mathbf{A} \cdot \mathbf{B} = A_{ik} B_{kj} e_i e_j, \mathbf{C} \cdot \mathbf{A} = C_{ijk} A_{kl} e_i e_j e_l
inner(a,b): \mathbf{v} : \mathbf{u} = v_i u_i, \mathbf{A} : \mathbf{B} = A_{ij} B_{ij}, \mathbf{C} : \mathbf{D} = C_{ijkl} D_{ijkl}
outer(a,b): \mathbf{v} \otimes \mathbf{u} = v_i u_j e_i e_j, \mathbf{A} \otimes \mathbf{u} = A_{ij} u_k e_i e_j e_k, \mathbf{A} \otimes \mathbf{B} = A_{ij} B_{kl} e_i e_j e_k e_l
cross(u,v), transpose(A), tr(A), det(A), inv(A), cofac(A), dev(A), skew(A), sym(A), ...
```

Consider Poisson's equation with two different boundary conditions on $\partial \Omega_0$ and $\partial \Omega_1$, the variational problem reads:

$$a(w; u, v) = L(f, g, h; v)$$

$$a(w; u, v) = \int_{\Omega} w \operatorname{grad} u \cdot \operatorname{grad} v \, dx$$

$$L(f, g, h; v) = \int_{\Omega} f v dx + \int_{\partial \Omega_0} g^2 v ds + \int_{\partial \Omega_1} h v ds$$

These forms can be expressed in UFL as

$$a = w * inner(grad(u), grad(v)) * dx$$

 $L = f * v * dx + g * * 2 * v * ds(0) + h * v * ds(1)$

Integrals:

 $\int_{\Omega} (.) dx \to *dx$ defines a cell integral $\int_{\partial\Omega} (.) ds \to *ds$ defines an exterior facet integral

Subdomains:

class Boundary(SubDomain): def inside(self,x,on_boundary): return x[0]>0.5 + DOLFIN+EPS

boundary_parts = MeshFunction("size_t", mesh, 1) boundary_parts.set_all(0) boundary = Boundary() boundary. mark(boundary_parts, 1)

Spatial derivative: $\frac{\partial f}{\partial x_i}$. Derivative of f in the spatial direction x_i

$$df = Dx(f, i)$$

$$df = f. dx(i)$$

The operator **diff** can be applied to expressions to differentiate w.r.t designated variables.

$$g = \sin(cell. x[0])$$

$$v = variable(g)$$

$$f = \exp(v ** 2)$$

$$h = diff(f, v)$$

The form operator **derivative** declares the derivative of a form w.r.t. coefficients of a discrete function. In case of non-trivial equations is useful.

• to linearize your nonlinear residual equation (linear form) automatically for use with the Newton-Raphson method.

```
# Pi is energy; u: function; du: trial function; v: test function F = derivative(Pi, u, v) # Compute Jacobian of F for Newton-Raphson method J = derivative(F, u, du)
```

• if applied multiple times, to derive a linear system from a convex functional, in order to find the function that minimizes the functional.

EXAMPLE – Poisson Equation

For a domain
$$\Omega \subset R^n$$
 with boundary $\partial \Omega = \Gamma_D \cup \Gamma_N$, the Poisson equation is
$$-\nabla^2 u = f \text{ in } \Omega$$
$$u = 0 \text{ in } \Gamma_D$$
$$\nabla u \cdot n = g \text{ in } \Gamma_N$$

where f and g are input and n is the outward directed boundary normal.

The variational form reads

$$a(u,v) = L(v) \quad v \in V$$

$$a(u,v) = \int_{\Omega} \nabla u \cdot \nabla v \, dx, L(v) = \int_{\Omega} fv \, dx + \int_{\Gamma_N} gv ds$$

V is a function space that satisfy the Dirichlet boundary conditions

Data in the example
$$\Omega = [0,1]x[0,1] \text{ (unit square)}$$

$$\Gamma_D = \{(0,y) \cup (1,y) \subset \partial\Omega\} \text{ (Dirichlet boundary)}$$

$$\Gamma_N = \{(x,0) \cup (x,1) \subset \partial\Omega\} \text{ (Neumann boundary)}$$

$$g = \sin 5x$$

$$f = 10e^{\left[-\frac{2(x-0.5)+2(y-0.5)}{0.02}\right]}$$

EXAMPLE – Poisson Equation

```
from dolfin import *
# Create mesh and define function space
mesh = UnitSquareMesh(32, 32)
V = FunctionSpace(mesh, "Lagrange", 1)
# Define Dirichlet boundary (x = 0 \text{ or } x = 1)
def boundary(x):
  return x[0] < DOLFIN EPS or x[0] > 1.0 - DOLFIN EPS
# Define boundary condition
u0 = Constant(0.0)
bc = DirichletBC(V, u0, boundary)
# Define variational problem
u = TrialFunction(V)
v = TestFunction(V)
f = Expression("10*exp(-(pow(x[0] - 0.5, 2) + pow(x[1] - 0.5, 2)) / 0.02)")
g = Expression("sin(5*x[0])")
a = inner(grad(u), grad(v))*dx
L = f^*v^*dx + g^*v^*ds
# Compute solution
u = Function(V)
solve(a == L, u, bc)
# Save solution in VTK format
file = File("poisson.pvd") file << u
# Plot solution
plot(u, interactive=True)
```

from dolfin import *

To use DOLFIN from Python, users need to import functionality from the DOLFIN Python module.

call to required external library. Es. import numpy

parameters["linear_algebra_backend"] = "PETSc"

Create mesh and define function space mesh = UnitSquareMesh(32, 32)

implemented class for simple mesh: Square, Circle, Cube, Cone, Cylinder, Sphere dolfin-convert filename.* filename.msh gmsh, Abaqus...

V = FunctionSpace(mesh, "Lagrange", 1)

FunctionSpace: Class to create a Function Space. It requires the mesh, the element type, the degree

```
# Define Dirichlet boundary (x = 0 \text{ or } x = 1)
def boundary(x):
return x[0] < DOLFIN\_EPS or x[0] > 1.0 - DOLFIN_EPS
```

Or label from gmsh

Define boundary condition u0 = Constant(0.0) bc = **DirichletBC**(V, u0, boundary)

> A Python function, returning a boolean, can be used to define the subdomain for the Dirichlet boundary condition. The function should return True for those points inside the subdomain and False for the points outside.

Define variational problem u = TrialFunction(V) v = TestFunction(V)

EXAMPLE – Poisson Equation

$$f = Expression("10*exp(-(pow(x[0] - 0.5, 2) + pow(x[1] - 0.5, 2)) / 0.02)")$$

$$g = Expression("sin(5*x[0])")$$

strings defining f and g use C++ syntax since, for efficiency, DOLFIN will generate and compile C++ code for these expressions at run-time.

Add parameter

a = inner(grad(u), grad(v))*dx

 $L = f^*v^*dx + g^*v^*ds$

Variational form

Compute solution

u = Function(V)

solve(a == L, u, bc)

u is automatically initialized at null Solving the variational problem; Matrices automatically assembled in the form Ax=b

Save solution in VTK format

file = File("poisson.pvd") file << u

Plot solution

plot(u, interactive=True)

Saving and Visualization. vtk format available

```
solve(A, x, b)
```

This is the simplest approach to solving the linear system Ax = b:

- default methods are considered
- straightforward, but little control
- it discriminates between linear and nonlinear problems

```
solve(A, x, b, ''lu'')
solve(A, x, b, ''gmres'', ''ilu'')
```

Here you provide methods and preconditioners

```
solver = LUSolver(A)
solver.solve(x,b)

solver = KrylovSolver(A)
solver.solve(x,b)
```

More specific choices

```
>>> solver = KrylovSolver()
>>> info(solver.parameters,1)
  rameter set "krylov solver" containing 9 parameter(s) and 2 nested parameter set(s)>
 krylov solver
                      type value range access change
 absolute_tolerance | double 1e-15
                                              0
 divergence limit | double 10000
                                              1
                                                            0
 error_on_nonconvergence | bool true {true, false}
 maximum_iterations | int 10000
 monitor_convergence | bool false {true, false}
                                                     1
 nonzero_initial_guess | bool false {true, false}
                                                     1
                                                            0
 relative_tolerance | double 1e-06
                                                     1
                                              0
                          bool true {true, false}
                                                     1
 report
                                                            0
 use petsc cusp hack
                          bool false {true, false}
                                                     1
                                                            0
 <Parameter set "gmres" containing 1 parameter(s) and 0 nested parameter set(s)>
             type value range access change
   gmres
                    30 []
   restart | int
                             1
 <Parameter set "preconditioner" containing 4 parameter(s) and 2 nested parameter set(s)>
   preconditioner | type value range access change
                         bool false {true, false} 1
   report
                         bool false {true, false} 1
   reuse
                                                           0
   same_nonzero_pattern | bool false {true, false} 1
                                                           0
   shift nonzero
                     | double
                                             1
                                                           0
                                 0
   <Parameter set "ilu" containing 1 parameter(s) and 0 nested parameter set(s)>
```

Boundary value problems for hyperelastic media can be expressed as minimisation problems. For a domain $\Omega \subset \mathbb{R}^n$, the task is to find the displacement field $u: \Omega \to \mathbb{R}^n$ that minimises the total potential energy Π :

$$\min_{v \in V} \Pi$$

where V is a suitable function space that satisfies boundary conditions on u.

The total potential energy is given by

$$\Pi = \int_{\Omega} \Phi(u) dx - \int_{\Omega} B \cdot u dx - \int_{\Gamma} T \cdot u ds$$

where $\Phi(u)$ is the elastic stored energy density, B is a body force (per unit reference volume) and T is a traction force (per unit reference area).

At minimum point, the directional derivative of Π along the test function v with respect to change in function u is equal to zero

$$L(u; v) = D_v \Pi = \frac{d\Pi(u + \epsilon v)}{d\epsilon} \bigg|_{\epsilon = 0} = 0 \qquad \forall v \in V$$

If $\Phi(u)$ is nonlinear, the Jacobian is required for the Newton's method. It is

$$Jac(u; du, v) = D_{du}L = \frac{dL(u + \epsilon du; v)}{d\epsilon} \bigg|_{\epsilon=0}$$

from dolfin import *

```
# Create mesh and define function space
mesh = UnitCubeMesh(24, 16, 16)
V = VectorFunctionSpace(mesh, "Lagrange", 1)
```

A Vector Function Space is required

```
bcl = DirichletBC(V, c, left)
bcr = DirichletBC(V, r, right)
bcs = [bcl, bcr]
```

More than one boundary condition are collected together in a list

```
B = Constant((0.0, -0.5, 0.0)) \# Body force per unit volume

T = Constant((0.1, 0.0, 0.0)) \# Traction force on the boundary
```

In place of Constant, it is also possible to use as_vector, e.g. B = as_vector([0.0, -0.5, 0.0]). The advantage of Constant is that its values can be changed without requiring re-generation and re-compilation of C++ code. On the other hand, using as_vector can eliminate some function calls during assembly.

```
# Kinematics
I = Identity(V.cell().d) # Identity tensor
F = I + grad(u) # Deformation gradient
C = F.T*F # Right Cauchy-Green tensor
# Invariants of deformation tensors
Ic = tr(C) J = det(F)
# Elasticity parameters
E, nu = 10.0, 0.3
mu, lmbda = Constant(E/(2*(1 + nu))), Constant(E*nu/((1 + nu)*(1 - 2*nu)))
# Stored strain energy density (compressible neo-Hookean model)
psi = (mu/2)*(Ic - 3) - mu*ln(J) + (lmbda/2)*(ln(J))**2 # Strain Energy Function
Pi = psi*dx - dot(B, u)*dx - dot(T, u)*ds # Total potential energy
```

Kinematics and Energy are derived with a formalism very similar to the analytical one

```
# Compute first variation of Pi (directional derivative about u in the direction of v)

F = derivative(Pi, u, v)

# Compute Jacobian of F

J = derivative(F, u, du)
```

Function *derivative* performs the directional derivatives automatically.

NB: it is not possible to modify the output of *derivative*

```
# Solve variational problem
solve(F == 0, u, bcs, J=J, form_compiler_parameters=ffc_options)
# Save solution in VTK format
file = File("displacement.pvd"); file << u;
# Plot and hold solution
plot(u, mode = "displacement", interactive = True)
```

Solve with customized options

```
# Periodic Boundary conditions
class PeriodicBoundary(SubDomain):
#Left boundary is "targeted domain" G
def inside(self,x,on boundary):
   return bool(x[0] < 0 + DOLFIN EPS and x[0] > -1 and on boundary)
#map right boundary (H) to left boundary (G)
def map(self,x,y):
   if (x[0] > 20 - DOLFIN EPS and x[0] < 20 + DOLFIN EPS and on boundary):
    y[0] = x[0] - 20
    y[1] = x[1]
pbc = PeriodicBoundary()
# Definition of function spaces
P2 = VectorFunctionSpace(mesh, "CG", user par.fe order u, constrained domain =
PeriodicBoundary()) # Space for displacement
P1 = FunctionSpace(mesh, "CG", user par.fe order p) # Space for pressure
V = MixedFunctionSpace([P1,P2])
                                            Kinematics and Energy are derived with a
                                            formalism very similar to the analytical one
```

```
# Create functions to define the energy and store the results
up = Function(V)
(p,u)=split(up)
                                             Function that extracts one function for each
                                             subspace
# Create test and trial functions for the variational formulation
dup = TrialFunction(V)
vq = TestFunction(V)
(q,v) = TestFunctions(V)
bc1 = DirichletBC(V.sub(0).sub(0), zero scalar, X())
bc1 = DirichletBC(V.sub(0), zero vector, X())
                                              Boundary conditions can be applied
                                              component by component of the subspace
```

```
# Energy (Incompressible neo Hookean)
psi = mu/2 * (Ic - 3) + p * (J - 1)
Pi = psi*dx
```

F=derivative(Pi, up,vq) dF=derivative(F,up,dup)

Lie Derivative is automatically applied to the functions belonging to different subspaces

```
# Setup the variational problem
varproblem = NonlinearVariationalProblem(F, up, bc_u,
J=dF,form_compiler_parameters=ffc_options)
solver = NonlinearVariationalSolver(varproblem)
```

In this case, the NonlinearVariationalSolver is called Info(NonlinearVariationalSolver.default_parameters(),1)

```
# surface energy
gamma = Constant(36.5)
N = FacetNormal(mesh)
NansonOp = transpose(cofac(F))
deformed N = dot(NansonOp, N)
current element of area = (1+u[1]/X[1])*sqrt(dot(deformed_N,deformed_N))
surface energy density = gamma*current element of area
surface energy = 2*math.pi*X[1]*surface energy density*ds(1)
                                          FEniCS, by default, works in a Material
                                          Reference System and with a Cartesian
                                          framework
```

```
# Total potential energy
Pi= Psi + surface_energy
(as before)
```

```
# Implemented Newton Raphson
while ttt <= T:
 fnorm = 1
 eps = 1
 mu.assign(value)
 nIter = 0
 cicle = 0
 while ok == 0:
     while fnorm > 1e-7 or eps > 1e-6 and nIter < nIter max :
    nIter += 1
     bcs du = homogenize(bc u)
     A, b = assemble system(dF, -F, bcs du)
     solve (A, dw inc.vector(), b)
    eps_ok = np.linalg.norm(dw inc.vector().array(), ord = 2)
     up.vector()[:] = up.vector() + omega*dw inc.vector()
```

Kinematics and Energy are derived with a formalism very similar to the analytical one

- Adaptive meshing
- Dofmaps
- Numpy, Scipy
- Fem-Fenics

