

NewtonImplicitSolver

July 23, 2025

This file is part of CasADi.

CasADi -- A symbolic framework for dynamic optimization.
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1 NewtonImplicitSolver

```
[1]: from casadi import *  
     from numpy import *  
     from pylab import *
```

We will investigate the working of rootfinder with the help of the parametrically excited Duffing equation. $\ddot{u} + \dot{u} - \epsilon(2\mu\dot{u} + \alpha u^3 + 2ku \cos(\Omega t))$ with $\Omega = 2 + \epsilon\sigma$. \

The first order solution is $u(t) = a \cos(\frac{1}{2}\Omega t - \frac{1}{2}\gamma)$ with the modulation equations: $\frac{da}{d\epsilon t} = -[\mu a + \frac{1}{2}ka \sin \gamma]$ \ $a \frac{d\gamma}{d\epsilon t} = -[-\sigma a + \frac{3}{4}\alpha a^3 + ka \cos \gamma]$ \

We seek the stationary solution to these modulation equations.

Parameters

```
[2]: eps    = SX.sym("eps")  
     mu     = SX.sym("mu")
```

```
alpha = SX.sym("alpha")
k      = SX.sym("k")
sigma = SX.sym("sigma")
params = [eps,mu,alpha,k,sigma]
```

Variables

```
[3]: a      = SX.sym("a")
     gamma = SX.sym("gamma")
```

Equations

```
[4]: res0 = mu*a+1.0/2*k*a*sin(gamma)
     res1 = -sigma * a + 3.0/4*alpha*a**3+k*a*cos(gamma)
```

Numerical values

```
[5]: sigma_ = 0.1
     alpha_ = 0.1
     k_      = 0.2
     params_ = [0.1,0.1,alpha_,k_,sigma_]
```

We create a NewtonImplicitSolver instance

```
[6]: f=Function("f", [vertcat(a,gamma),vertcat(*params)], [vertcat(res0,res1)])
     opts = {}
     opts["abstol"] = 1e-14
     opts["linear_solver"] = "csparse"
     s=rootfinder("s", "newton", f, opts)
```

Initialize $[a, \gamma]$ with a guess and solve

```
[7]: x_ = s([1,-1], params_)
```

```
[8]: print("Solution = ", x_)
```

Solution = [1.1547, -1.5708]

Compare with the analytic solution:

```
[9]: x = [sqrt(4.0/3*sigma_/alpha_),-0.5*pi]
     print("Reference solution = ", x)
```

Reference solution = [1.1547005383792515, -1.5707963267948966]

We show that the residual is indeed (close to) zero

```
[10]: residual = f(x_, params_)
      print("residual = ", residual)
```

residual = [1.29063e-15, 2.62584e-15]

```
[11]: for i in range(1):  
      assert(abs(x_[i]-x[i])<1e-6)
```

Solver statistics

```
[12]: print(s.stats())
```

```
{'iter_count': 29, 'n_call_g': 0, 'n_call_jac_g_x': 0, 'return_status':  
'success', 'success': True, 't_proc_g': 0.0, 't_proc_jac_g_x': 0.0, 't_wall_g':  
0.0, 't_wall_jac_g_x': 0.0, 'unified_return_status': 'SOLVER_RET_UNKNOWN'}
```