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```

Callback

```
from casadi import *
from numpy import *
```

In this example, we will demonstrate callback functionality for Ipopt. Note that you need the fix https://github.com/casadi/casadi/wiki/enableIpoptCallback before this works

We start with constructing the rosenbrock problem

```
x=SX.sym("x")
y=SX.sym("y")

f = (1-x)**2+100*(y-x**2)**2
nlp={'x':vertcat(x,y), 'f':f,'g':x+y}
fcn = Function('f', [x, y], [f])
```

Matplotlib callback

Now let's do some useful visualisations We create a callable python class, i.e. one that has a __call__ member

```

from pylab import figure, subplot, contourf, colorbar, draw, show, plot,
    title

import time
import matplotlib
matplotlib.interactive(True)

class MyCallback(Callback):
  def __init__(self, name, nx, ng, np, opts={}):
    Callback.__init__(self)

    self.nx = nx
    self.ng = ng
    self.np = np

    opts['input_scheme'] = nlpsol_out()
    opts['output_scheme'] = ['ret']

    figure(1)
    subplot(111)

    x_,y_ = mgrid[-1:1.5:0.01,-1:1.5:0.01]
    z_ = DM.zeros(x_.shape)

    for i in range(x_.shape[0]):
      for j in range(x_.shape[1]):
        z_[i,j] = fcn(x_[i,j],y_[i,j])
    contourf(x_,y_,z_)
    colorbar()
    title('Iterations of Rosenbrock')
    draw()

    self.x_sols = []
    self.y_sols = []

    # Initialize internal objects
    self.construct(name, opts)

  def get_n_in(self): return nlpsol_n_out()
  def get_n_out(self): return 1


  def get_sparsity_in(self, i):
    n = nlpsol_out(i)
    if n=='f':
      return Sparsity.scalar()
    elif n in ('x', 'lam_x'):
      return Sparsity.dense(self.nx)
    elif n in ('g', 'lam_g'):
      return Sparsity.dense(self.ng)
    else:
      return Sparsity(0,0)
  def eval(self, arg):
    # Create dictionary
    darg = {}
    for (i,s) in enumerate(nlpsol_out()): darg[s] = arg[i]

    sol = darg['x']
    self.x_sols.append(float(sol[0]))
    self.y_sols.append(float(sol[1]))
    subplot(111)
    if hasattr(self,'lines'):
      self.lines[0].set_xdata(self.x_sols)
      self.lines[0].set_ydata(self.y_sols)
    else:
      self.lines = plot(self.x_sols,self.y_sols,'or-')

    draw()
    time.sleep(0.25)

    return [0]

mycallback = MyCallback('mycallback', 2, 1, 0)
opts = {}
opts['iteration_callback'] = mycallback
opts['ipopt.tol'] = 1e-8
opts['ipopt.max_iter'] = 50
solver = nlpsol('solver', 'ipopt', nlp, opts)
sol = solver(lbx=-10, ubx=10, lbg=-10, ubg=10)
```

```
******************************************************************************
This program contains Ipopt, a library for large-scale nonlinear
    optimization.
 Ipopt is released as open source code under the Eclipse Public License (
     EPL).
         For more information visit http://projects.coin-or.org/Ipopt
******************************************************************************

This is Ipopt version 3.12.3, running with linear solver ma57.

Number of nonzeros in equality constraint Jacobian...:        0
Number of nonzeros in inequality constraint Jacobian.:        2
Number of nonzeros in Lagrangian Hessian.............:        3

Total number of variables............................:        2
                     variables with only lower bounds:        0
                variables with lower and upper bounds:        2
                     variables with only upper bounds:        0
Total number of equality constraints.................:        0
Total number of inequality constraints...............:        1
        inequality constraints with only lower bounds:        0
   inequality constraints with lower and upper bounds:        1
        inequality constraints with only upper bounds:        0

iter    objective    inf_pr   inf_du lg(mu)  ||d||  lg(rg) alpha_du
    alpha_pr  ls
   0  1.0000000e+00 0.00e+00 1.33e+00  -1.0 0.00e+00    -  0.00e+00 0.00e
      +00   0
   1  8.1696108e-01 0.00e+00 8.18e+00  -1.0 8.33e-01    -  9.22e-01 2.50e
      -01f  3
   2  5.0928866e-01 0.00e+00 1.31e+00  -1.0 1.58e-01    -  1.00e+00 1.00e
      +00f  1
   3  3.8213489e-01 0.00e+00 5.32e+00  -1.0 4.89e-01    -  1.00e+00 5.00e
      -01f  2
   4  2.2689357e-01 0.00e+00 1.54e+00  -1.0 1.92e-01    -  1.00e+00 1.00e
      +00f  1
   5  1.9526345e-01 0.00e+00 9.02e+00  -1.0 3.85e-01    -  1.00e+00 1.00e
      +00f  1
   6  6.2791707e-02 0.00e+00 2.66e-01  -1.0 1.22e-01    -  1.00e+00 1.00e
      +00f  1
   7  3.3688142e-02 0.00e+00 3.14e+00  -1.7 4.88e-01    -  1.00e+00 5.00e
      -01f  2
   8  1.1215647e-02 0.00e+00 6.93e-01  -1.7 1.45e-01    -  1.00e+00 1.00e
      +00f  1
   9  3.3356343e-03 0.00e+00 1.69e+00  -1.7 1.91e-01    -  1.00e+00 1.00e
      +00f  1
iter    objective    inf_pr   inf_du lg(mu)  ||d||  lg(rg) alpha_du
    alpha_pr  ls
  10  3.2713823e-04 0.00e+00 8.94e-02  -1.7 5.61e-02    -  1.00e+00 1.00e
     +00f  1
  11  9.2093197e-06 0.00e+00 1.06e-01  -2.5 4.91e-02    -  1.00e+00 1.00e
     +00f  1
  12  1.0134200e-07 0.00e+00 3.06e-04  -2.5 3.33e-03    -  1.00e+00 1.00e
     +00f  1
  13  2.0350914e-10 0.00e+00 3.66e-05  -3.8 9.12e-04    -  1.00e+00 1.00e
     +00f  1
  14  2.9616464e-14 0.00e+00 7.79e-08  -5.7 4.22e-05    -  1.00e+00 1.00e
     +00f  1
  15  5.4181655e-20 0.00e+00 1.17e-11  -8.6 5.16e-07    -  1.00e+00 1.00e
     +00f  1

Number of Iterations....: 15

                                   (scaled)                 (unscaled)
Objective...............:   5.4181654535916011e-20    5.4181654535916011e
    -20
Dual infeasibility......:   1.1724482496621431e-11    1.1724482496621431e
    -11
Constraint violation....:   0.0000000000000000e+00    0.0000000000000000e
    +00
Complementarity.........:   2.5060224083090714e-09    2.5060224083090714e
    -09
Overall NLP error.......:   2.5060224083090714e-09    2.5060224083090714e
    -09

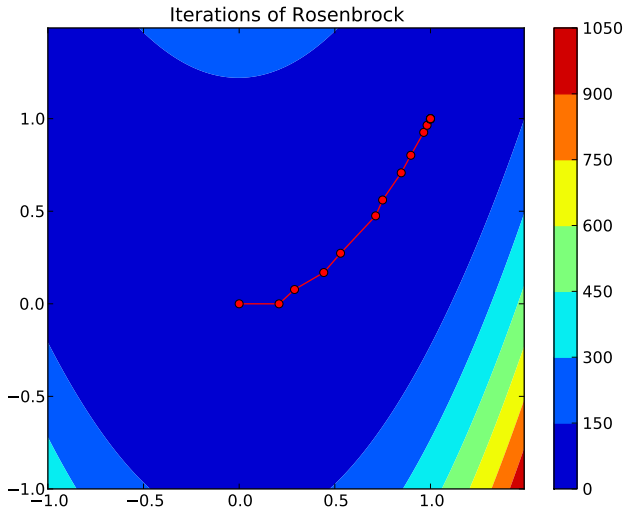

Number of objective function evaluations             = 26
Number of objective gradient evaluations             = 16
Number of equality constraint evaluations            = 0
Number of inequality constraint evaluations          = 26
Number of equality constraint Jacobian evaluations   = 0
Number of inequality constraint Jacobian evaluations = 16
Number of Lagrangian Hessian evaluations             = 15
Total CPU secs in IPOPT (w/o function evaluations)   =      0.305
Total CPU secs in NLP function evaluations           =      0.001

EXIT: Optimal Solution Found.
                   proc           wall      num           mean
                                  mean
                   time           time     evals       proc time
           wall time
       nlp_f     0.000 [s]      0.000 [s]    26       0.00 [ms]
        0.00 [ms]
       nlp_g     0.000 [s]      0.000 [s]    26       0.00 [ms]
        0.01 [ms]
  nlp_grad_f     0.000 [s]      0.000 [s]    17       0.00 [ms]
     0.00 [ms]
   nlp_jac_g     0.000 [s]      0.000 [s]    17       0.00 [ms]
      0.00 [ms]
  nlp_hess_l     0.000 [s]      0.000 [s]    15       0.00 [ms]
     0.02 [ms]
all previous     0.000 [s]      0.001 [s]
callback_fun     0.310 [s]      4.301 [s]    16      19.38 [ms]
   268.84 [ms]
callback_prep    0.000 [s]      0.000 [s]    16       0.00 [ms]
    0.01 [ms]
      solver     0.000 [s]      0.009 [s]
    mainloop     0.310 [s]      4.311 [s]
```

By setting matplotlib interactivity off, we can inspect the figure at ease

```
114 matplotlib.interactive(False)
```

```
115 show()
```


Iterations of Rosenbrock
1.0
0.5
0.0
−0.5
−1.0
−1.0
−0.5
0.0
0.5
1.0
1050
900
750
600
450
300
150
0