

# MATH693a\_HW1

September 21, 2018

```
In [2]: import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
from scipy import linalg as LA
import time
%matplotlib notebook
%matplotlib inline
```

**MATH693a**

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**Homework 1**

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The Rosenbrock Function is given by

$$f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2 \quad (1)$$

We can easily define this function in Python.

```
In [3]: def rb(x):
rb = 100*(x[1]-x[0]**2)**2 + (1-x[0])**2
return rb
```

Its gradient can easily be found by computing the first derivative with respect to  $x_1$  and  $x_2$ . For  $f'(x_1)$  we have

$$f'(x_1) = 100(4x_1^3 - 4x_2x_1) + 2x_1 - 2 \quad (2)$$

$$= 100(4x_1^3 - 4x_2x_1) + 2(x_1 - 1) \quad (3)$$

$$= 400x_1(x_1^2 - x_2) + 2(x_1 - 1); \quad (4)$$

and for  $f'(x_2)$

$$f'(x_2) = 100(2x_2 - 2x_1^2) \quad (5)$$

$$= 200(x_2 - x_1^2). \quad (6)$$

Therefore, the gradient of the Rosenbrock Function is given by

$$\nabla f(x) = \langle 400x_1(x_1^2 - x_2) + 2(x_1 - 1), 200(x_2 - x_1^2) \rangle. \quad (7)$$

In order to find the minimum, we need to set the gradient equal to zero

$$\nabla f(x) = 0 \quad (8)$$

for which it holds

$$(x_1, x_2) = (1, 1) \quad (9)$$

As for the Hessian, we have

$$f_{x_1x_1} = 1200x_1^2 - 400x_2 + 2; \quad (10)$$

$$f_{x_1x_2} = -400x_1 = f_{x_2x_1}; \quad (11)$$

$$f_{x_2x_2} = 200 \quad (12)$$

$$(13)$$

Thus the Hessian is

$$\begin{bmatrix} 1200x_1^2 - 400x_2 + 2 & -400x_1 \\ -400x_1 & 200 \end{bmatrix}$$

We can easily define the gradient and the Hessian in Python and then use them to apply the steepest descent and the Newton method for the backline search algorithm

```
In [4]: def rb_grad(x):
        df1 = 400*x[0]*(x[0]**2-x[1])+2*(x[0]-1)
        df2 = 200*(x[1]-x[0]**2)
        grad = np.array([df1,df2])
        return grad
```

```
In [5]: def hess(x):
        h11 = 1200*x[0]**2-400*x[1]+2
        h12 = -400*x[0]
        h21 = -400*x[0]
        h22 = 200
        hess = np.array([[h11,h12],[h21,h22]])
        return hess
```

```
In [6]: def invhess(x):
        invhess = LA.solve(hess(x),rb_grad(x))
        return invhess
```

```
In [7]: def rb_back_line_search(x,method):
        alpha_bar = 1.
        iiter = 1
        max_iter = 100000
```

```

tol = 1e-8
c1 = 1e-4
rho = .5
alphavals = []
pvals = []
start = time.time()
while LA.norm(rb_grad(x)) > tol and iiter < max_iter:
    iiter = iiter + 1
    alpha = alpha_bar
    if method=='Newton':
        p = -invhess(x) #search direction for Newton method
    else:
        p = -rb_grad(x)/(LA.norm(rb_grad(x))) #search direction for steepest descent
    pt = np.transpose(p)
    while rb(x+alpha*p) > rb(x)+c1*alpha*(pt).dot(rb_grad(x)):
        alpha = rho*alpha
    xnew = x + alpha*p
    x = xnew
    alpha_k = alpha
    alphavals.append(alpha_k)
    alpha10 = alphavals[0:10]
end = time.time()
print("Elapsed time for " + method + " method is " + np.str(end-start))
print("Minimum found at " + np.str(x))
print("Value of Rosenbrock Function at minimum is " + np.str(rb(x)))
print("Search direction is p = " + np.str(p))
print("First 10 alpha values found to be " + np.str(alpha10))
print("Number of iterations necessary for convergence is " + np.str(iiter))

```

In [8]: rb\_back\_line\_search([1.2,1.2], 'steepest descent')

Elapsed time for steepest descent method is 9.69500017166

Minimum found at [1. 1.00000001]

Value of Rosenbrock Function at minimum is 7.244209187686237e-18

Search direction is p = [ 0.89012922 -0.45570822]

First 10 alpha values found to be [0.125, 0.03125, 0.001953125, 0.00048828125, 0.00048828125, 0.00048828125, 0.00048828125, 0.00048828125, 0.00048828125, 0.00048828125]

Number of iterations necessary for convergence is 17067

In [13]: rb\_back\_line\_search([-1.2,1.], 'steepest descent')

Elapsed time for steepest descent method is 9.42100000381

Minimum found at [1. 0.99999999]

Value of Rosenbrock Function at minimum is 7.033102791657379e-18

Search direction is p = [ 0.89927774 -0.43737804]

First 10 alpha values found to be [0.25, 0.125, 0.0625, 0.0078125, 0.0078125, 0.0078125, 0.0078125, 0.0078125, 0.0078125, 0.0078125]

Number of iterations necessary for convergence is 17895

```
In [11]: rb_back_line_search([1.2,1.2], 'Newton')
```

Elapsed time for Newton method is 0.00599980354309

Minimum found at [1. 1.]

Value of Rosenbrock Function at minimum is 1.088287359901554e-25

Search direction is p = [-1.77859130e-07 -3.53202652e-07]

First 10 alpha values found to be [1.0, 0.5, 1.0, 1.0, 1.0, 1.0, 1.0, 1.0]

Number of iterations necessary for convergence is 9

```
In [12]: rb_back_line_search([-1.2,1.], 'Newton')
```

Elapsed time for Newton method is 0.00499987602234

Minimum found at [1. 1.]

Value of Rosenbrock Function at minimum is 3.743975643139474e-21

Search direction is p = [1.11032194e-06 2.49053263e-06]

First 10 alpha values found to be [1.0, 0.125, 1.0, 1.0, 1.0, 0.25, 1.0, 1.0, 1.0, 0.5]

Number of iterations necessary for convergence is 22