MATH 151B - Applied Numerical Methods - Homework 6

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Question 1

We begin by finding:

$$F(1,1,1) = \begin{bmatrix} 1^2 + 1 - 37 \\ 1 - 1^2 - 5 \\ 1 + 1 + 1 - 3 \end{bmatrix} = \begin{bmatrix} -35 \\ -5 \\ 0 \end{bmatrix}$$

The Jacobian is:

$$J(x_1, x_2, x_3) = \begin{bmatrix} 2x_1 & 1 & 0 \\ 1 & -2x_2 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

At the initial point:

$$J(1,1,1) = \begin{bmatrix} 2 & 1 & 0 \\ 1 & -2 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

The inverse:

$$J^{-1}(1,1,1) = \begin{bmatrix} .4 & .2 & 0 \\ .2 & -4 & 0 \\ -.6 & .2 & 1 \end{bmatrix}$$

We do matrix multiplication:

$$J^{-1}(1,1,1)F(1,1,1) = \begin{bmatrix} -15 \\ -5 \\ 20 \end{bmatrix}$$

Thus, $x^{(1)}$ is:

$$\begin{bmatrix} 1\\1\\1 \end{bmatrix} - \begin{bmatrix} -15\\-5\\20 \end{bmatrix} = \begin{bmatrix} 16\\6\\-19 \end{bmatrix}$$

From earlier, F(1,1,1) = [-35, -5, 0], we find F(16,6,-19) = [225, -25, 0]. Now, we see that:

$$\sqrt{35^2 + 5^2 + 0^2} = \sqrt{1250} < \sqrt{225^2 + 25^2 + 0^2} = \sqrt{51250}$$

This means that our new approximation is not better than the starting guess because the result of $F(x^{(1)})$ is further than $F(x^{(0)})$ is to [0,0,0].

```
def newton(F_all, J, start, k):
  x = [start]
  e = [np.linalg.norm(F_all(start))]
  for i in range(0, k):
    temp = x[i] - np.dot(inv(J(x[i])),F_all(x[i]))
    x.append(np.around(temp, 10))
    e.append(np.linalg.norm(F_all(temp)))
  return(x, e)
def F_all(x):
  return([[x[0][0]*x[0][0] + x[1][0] - 37],
          [x[0][0] - x[1][0]*x[1][0] - 5],
          [x[0][0] + x[1][0] + x[2][0] - 3]])
def J(x):
  return([[2*x[0][0], 1, 0],
          [1, -2*x[1][0], 0],
          [1, 1, 1]])
x, e = newton(F_all, J, [[1], [1], 4)
pd.DataFrame(data={'[x_1, x_2, x_3]':x, 'error':e})
##
                                        [x_1, x_2, x_3]
                                                              error
## 0
                                        [[1], [1], [1]]
                                                          35.355339
## 1
                               [[16.0], [6.0], [-19.0]]
                                                         226.384628
## 2 [[9.0519480519], [3.3376623377], [-9.3896103896]]
                                                          48.793002
      [[6.4654261642], [1.8883599911], [-5.3537861553]]
                                                           7.012088
## 4
      [[6.0005806085], [1.2091137511], [-4.2096943596]]
                                                           0.509469
x, e = newton(F_all, J, [[1], [1], 8)
pd.DataFrame(data={'[x_1, x_2, x_3]':x, 'error':e})
##
                                        [x_1, x_2, x_3]
                                                                error
## 0
                                        [[1], [1], [1]]
                                                         3.535534e+01
## 1
                               [[16.0], [6.0], [-19.0]] 2.263846e+02
## 2 [[9.0519480519], [3.3376623377], [-9.3896103896]] 4.879300e+01
     [[6.4654261642], [1.8883599911], [-5.3537861553]] 7.012088e+00
## 3
      [[6.0005806085], [1.2091137511], [-4.2096943596]] 5.094688e-01
## 4
## 5 [[5.9985434542], [1.0174805783], [-4.0160240324]] 3.672327e-02
## 6
       [[5.999988146], [1.0001443352], [-4.0001324812]] 3.005526e-04
## 7
        [[5.99999999], [1.0000001], [-4.000000092]]
                                                         2.083024e-08
                                 [[6.0], [1.0], [-4.0]] 0.000000e+00
## 8
```

Question 3

```
def steepest(g, grad_g, max_iter, initial, tol):
 x = [initial]
  i = 0
 while i <= max_iter:</pre>
    g1 = g(x[i])
   z = grad_g(x[i])
    z0 = np.linalg.norm(z)
    if z0 == 0:
     print("here1")
     return(x)
    z = z/z0
    alpha1 = 0
    alpha3 = 1
    g3 = g(x[i] - alpha3*z)
    while(g3 \geq= g1):
      alpha3 = alpha3/2
      g3 = g(x[i] - alpha3*z)
      if alpha3 < tol/2:</pre>
        print("here2")
        return(x)
    alpha2 = alpha3/2
    g2 = g(x[i] - alpha2*z)
    h1 = (g2 - g1)/alpha2
    h2 = (g3 - g2)/(alpha3 - alpha2)
    h3 = (h2 - h1)/alpha3
    alpha0 = .5*(alpha2 - h1)/h3
    g0 = g(x[i] - alpha0*z)
    if g0 <= g3:
      g_real = g0
      alpha = alpha0
      g_real = g3
      alpha = alpha3
    if abs(g_real - g1) < tol:</pre>
      print("here3")
      x.append(x[i] - alpha*z)
      return(x)
    else:
      x.append(x[i] - alpha*z)
    i = i + 1
 return(x)
```

```
def F_all(x):
    x1 = x[0][0]
    x2 = x[1][0]
    x3 = x[2][0]
```

```
return(np.array([[x1*x1*x1 + x1*x1*x2 - x1*x3 + 6],
          [math.exp(x1) + math.exp(x2) - x3],
          [x2*x2 - 2*x1*x3 - 4]]))
def J(x):
  x1 = x[0][0]
  x2 = x[1][0]
  x3 = x[2][0]
  return(np.array([[3*x1*x1 - 2*x1*x2 - x3, x1*x1, -x1],
          [math.exp(x1), math.exp(x2), -1],
          [-2*x3, 2*x2, -2*x1]]).tolist())
def g(x):
  x1 = x[0][0]
  x2 = x[1][0]
  x3 = x[2][0]
  f1 = (x1*x1*x1 + x1*x1*x2 - x1*x3 + 6)*(x1*x1*x1 + x1*x1*x2 - x1*x3 + 6)
  f2 = (math.exp(x1) + math.exp(x2) - x3)*(math.exp(x1) + math.exp(x2) - x3)
  f3 = (x2*x2 - 2*x1*x3 - 4)*(x2*x2 - 2*x1*x3 - 4)
  return(f1 + f2 + f3)
def grad_g(x):
  return(2 * inv(J(x)).dot(F_all(x)))
x_result = steepest(g, grad_g, math.inf, [[1], [1], [1]], .01)
## here3
F_x_result = []
error = []
for i in x result:
  temp = F_{all}([[i[0][0]], [i[1][0]], [i[2][0]])
  F_x_result.append(temp)
  error.append(np.linalg.norm(temp))
x_result = [np.around(x, 3) for x in x_result]
F \times result = [np.around(x, 3) for x in F x result]
error = np.around(error, 5)
pd.DataFrame(data={'x':x_result, 'F(x_result)':F_x_result, 'error':error})
##
                                                         F(x_result)
                                                                        error
## 0
                    [[1], [1], [1]]
                                            [[7.0], [4.437], [-5.0]]
                                                                      9.67900
## 1
        [[0.686], [1.546], [1.777]]
                                        [[5.832], [4.901], [-4.049]] 8.62669
## 2
          [[0.517], [1.8], [2.729]]
                                        [[5.208], [4.997], [-3.584]] 8.05845
## 3
        [[0.427], [1.968], [3.711]]
                                        [[4.852], [4.978], [-3.295]]
                                                                      7.69337
## 4
          [[0.371], [2.1], [4.701]]
                                         [[4.596], [4.914], [-3.08]]
                                                                      7.39960
## 5
        [[0.333], [2.211], [5.694]]
                                        [[4.386], [4.827], [-2.905]]
                                                                      7.13963
## 6
        [[0.305], [2.309], [6.688]]
                                        [[4.202], [4.73], [-2.754]] 6.89978
## 7
       [[0.188], [2.785], [12.102]]
                                        [[3.825], [5.301], [-0.807]] 6.58650
                                        [[3.221], [5.087], [-0.269]] 6.02696
       [[0.161], [3.075], [17.742]]
## 8
```

```
## 9
        [[0.143], [3.336], [24.54]]
                                         [[2.563], [4.711], [0.112]]
                                                                       5.36486
## 10
       [[0.129], [3.575], [32.715]]
                                         [[1.855], [4.118], [0.369]]
                                                                       4.53163
       [[0.117], [3.794], [42.424]]
                                         [[1.092], [3.144], [0.472]]
                                                                       3.36157
       [[0.109], [3.971], [52.719]]
                                         [[0.316], [1.449], [0.307]]
## 12
                                                                       1.51471
## 13
       [[0.107], [4.024], [57.371]]
                                      [[-0.082], [-0.326], [-0.065]]
                                                                       0.34218
       [[0.107], [4.012], [56.371]]
                                                                       0.00337
## 14
                                        [[-0.001], [-0.003], [-0.0]]
                                              [[0.0], [0.002], [0.0]]
       [[0.107], [4.012], [56.355]]
## 15
                                                                       0.00235
x_result = steepest(g, grad_g, math.inf, [[1], [1], [1]], 10**(-5))
```

here3

```
F_x_result = []
error = []
for i in x_result:
    temp = F_all([ [i[0][0]], [i[1][0]], [i[2][0]] ])
    F_x_result.append(temp)
    error.append(np.linalg.norm(temp))

x_result = [np.around(x, 3) for x in x_result]
F_x_result = [np.around(x, 3) for x in F_x_result]
error = np.around(error, 5)

pd.DataFrame(data={'x':x_result, 'F(x_result)':F_x_result, 'error':error})
```

```
##
                                                           F(x_result)
                                                                           error
## 0
                     [[1], [1], [1]]
                                             [[7.0], [4.437], [-5.0]]
                                                                        9.67900
## 1
        [[0.686], [1.546], [1.777]]
                                         [[5.832], [4.901], [-4.049]]
                                                                        8.62669
## 2
          [[0.517], [1.8], [2.729]]
                                         [[5.208], [4.997], [-3.584]]
                                                                        8.05845
        [[0.427], [1.968], [3.711]]
                                         [[4.852], [4.978], [-3.295]]
## 3
                                                                        7.69337
## 4
          [[0.371], [2.1], [4.701]]
                                          [[4.596], [4.914], [-3.08]]
                                                                        7.39960
## 5
        [[0.333], [2.211], [5.694]]
                                         [[4.386], [4.827], [-2.905]]
                                                                        7.13963
## 6
        [[0.305], [2.309], [6.688]]
                                          [[4.202], [4.73], [-2.754]]
                                                                        6.89978
       [[0.188], [2.785], [12.102]]
                                         [[3.825], [5.301], [-0.807]]
## 7
                                                                        6.58650
## 8
       [[0.161], [3.075], [17.742]]
                                         [[3.221], [5.087], [-0.269]]
                                                                        6.02696
## 9
        [[0.143], [3.336], [24.54]]
                                          [[2.563], [4.711], [0.112]]
                                                                        5.36486
## 10
       [[0.129], [3.575], [32.715]]
                                          [[1.855], [4.118], [0.369]]
                                                                        4.53163
       [[0.117], [3.794], [42.424]]
                                          [[1.092], [3.144], [0.472]]
## 11
                                                                        3.36157
## 12
       [[0.109], [3.971], [52.719]]
                                          [[0.316], [1.449], [0.307]]
                                                                        1.51471
       [[0.107], [4.024], [57.371]]
                                       [[-0.082], [-0.326], [-0.065]]
## 13
                                                                        0.34218
       [[0.107], [4.012], [56.371]]
                                         [[-0.001], [-0.003], [-0.0]]
## 14
                                                                        0.00337
## 15
       [[0.107], [4.012], [56.355]]
                                              [[0.0], [0.002], [0.0]]
                                                                        0.00235
```

As you can see above, there does not seem to be an increase as the tolerance gets smaller. The answer converges the same regardless of if the tolerance is .01 or .0001.