### EE475 Homework 1

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#### Problem 3.1: Find the stationary points and plot the functions.

a:

$$g(w) = w \log(w) + (1 - w) \log(1 - w)$$
$$g'(w) = \log(w) + w \frac{1}{w \ln(10)} - \log(1 - w) - (1 - w) \frac{1}{(1 - w) \ln(10)}$$
$$g'(w) = \log(w) - \log(1 - w)$$

Stationary point @ g' = 0:

$$0 = \log(w) - \log(1 - w)$$
$$w = \frac{1}{2}$$

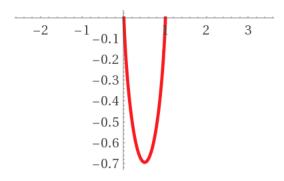


Figure 1: Minimum at w = 0.5

b:

$$g(w) = \log(1 + e^w)$$
$$g'(w) = \frac{1}{(1 + e^w)\ln(10)}e^w\log(e)$$
$$g'(w) = \frac{e^w}{(1 + e^w)}$$

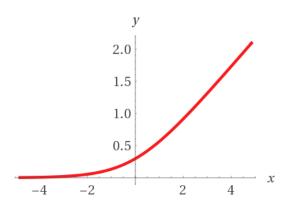


Figure 2: Saddle Point at  $w = -\infty$ 

Stationary point @ g' = 0:

$$0 = \frac{e^w}{(1 + e^w)}$$
$$0 = e^w$$
$$w = \ln(0) = -\infty$$

c:

$$g(w) = w \tanh(w)$$
$$g'(w) = \tanh(w) + w \frac{1}{\cosh^2(w)}$$

Stationary point @ g' = 0:

$$0 = \tanh(w) + w \frac{1}{\cosh^{2}(w)}$$
$$-w = \sinh(w) \cosh(w)$$
$$w = 0$$

d:

$$g(w) = \frac{1}{2}w^T C w + b^T w$$
$$g'(w) = b^T + w^T C$$

Stationary point @ g' = 0:

$$\begin{bmatrix} 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \end{bmatrix} + \begin{bmatrix} w_1 & w_2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$$
$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} -0.4 \\ -0.2 \end{bmatrix}$$

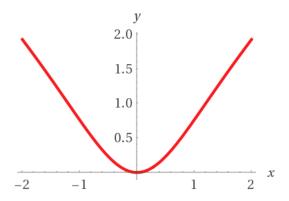


Figure 3: Minimum at w=0

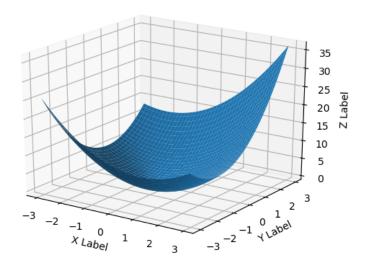


Figure 4: Minimum at x=-0.4 and y=-0.2

## Problem 3.3

$$g(w) = \frac{w^T C w}{w^T w}$$

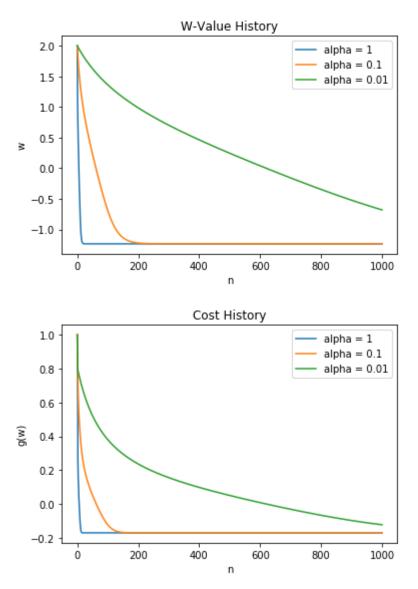
$$Dg(w) = \frac{2w^T C w^T w - w^T C w (2w^T)}{(w^T w)^2}$$

Stationary point @ Dg(w) = 0:

$$\mathbf{0} = 2w^T C w^T w - w^T C w (2w^T)$$

# Problem 3.5

```
In [3]: import numpy as np
import matplotlib.pyplot as plt
def f(w):
    return (1./50.)*(w**4 + w**2 + 10*w)
def df(w):
    return (1./50.)*(4.0*w**3 + 2*w + 10)
w init = 2
n = 1000
alpha = [1, 0.1, 0.01]
w = np.ones([len(alpha), n+1])
output w = np.copy(w)
for i, a in enumerate(alpha):
    w[i][0] = w init
    for j in range(n):
        w[i][j+1] = w[i][j] - a * df(w[i][j])
        output_w[i][j+1] = f(w[i][j])
x = np.arange(n+1)
plt.figure()
plt.plot(x,w[0])
plt.plot(x,w[1])
plt.plot(x,w[2])
plt.legend(['alpha = 1', 'alpha = 0.1', 'alpha = 0.01'])
plt.xlabel("n")
plt.ylabel("w")
plt.title("W-Value History")
plt.figure()
plt.plot(x,output w[0])
plt.plot(x,output w[1])
plt.plot(x,output_w[2])
plt.title("Cost History")
plt.legend(['alpha = 1', 'alpha = 0.1', 'alpha = 0.01'])
plt.xlabel("n")
plt.ylabel("g(w)")
plt.show()
```



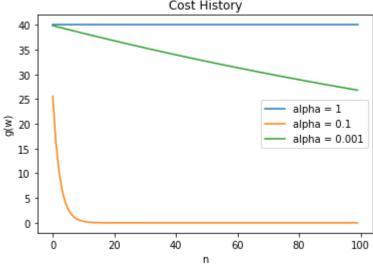
For this particular function and initial condition,  $\alpha=1$  converged the quickest.

# Problem 3.6

```
In [5]: def g(w):
     return abs(w)
 def dg(w):
     if(w > 0):
          return 1
     elif(w < 0):
          return -1
 w_init = 1.75
 n = 20
 w = np.ones([2, n+1])*w_init
 for i in range(n):
     try:
          w[0][i+1] = w[0][i] - 0.5 * dg(w[0][i])
     except:
          w[0][i+1] = w[0][i]
     try:
          w[1][i+1] = w[1][i] - 1/(i+1) * dg(w[1][i])
     except:
          print("error")
 x = np.arange(n+1)
 plt.plot(x,w[0],'co-')
 plt.plot(x,w[1],'ko-')
 plt.legend(['alpha = 0.5', 'alpha = 1/k'])
 plt.xlabel("k")
 plt.ylabel("w^k")
 plt.show()
     1.75
                                            alpha = 0.5
                                            alpha = 1/k
     1.50
     1.25
     1.00
     0.75
     0.50
     0.25
     0.00
    -0.25
                             10.0
                                  12.5
                                       15.0
                                            17.5
          0.0
               2.5
                    5.0
                         7.5
                               k
```

### Problem 3.8

```
In [6]: def h(w):
    w = np.reshape(w, [10,1])
     return np.dot(np.transpose(w), w)
def dh(w):
     return (2 * np.transpose(w))
w init = 2
n = 100
alpha = [1, 0.1, 0.001]
h_out = np.zeros([len(alpha),n])
for j, a in enumerate(alpha):
    w = np.ones([10,n+1])*w_init
    for i in range(n):
         w[:,i+1] = w[:,i] - a * dh(w[:,i])
         h_{out[j][i]} = h(w[:,i+1])
x = np.arange(n)
plt.plot(x,h_out[0])
plt.plot(x,h_out[1])
plt.plot(x,h out[2])
plt.title("Cost History")
plt.legend(['alpha = 1', 'alpha = 0.1', 'alpha = 0.001'])
plt.xlabel("n")
plt.ylabel("g(w)")
plt.show()
                      Cost History
```



 $\alpha=0.1$  performs the best