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import numpy as np
import sympy as sp
import matplotlib.pyplot as plt
\#We use symbolic variables x and y for our function
x= sp.symbols('x')
y=sp.symbols('y')
#Here we setup the function we are trying to minimize
\mathsf{f} = ((10^*\mathsf{x}^{**2}) + (22^*\mathsf{x}) + (13^*\mathsf{y}^{**2}) + (8^*\mathsf{y}) + (4^*\mathsf{x}^*\mathsf{y}) + (13))/9
#Creating the gradient
dfx=sp.diff(f,x)
dfy=sp.diff(f,y)
\ensuremath{\text{\#}} Creating the Hessian in this case the hessian is constant
dfxx=sp.diff(dfx,x)
dfxy=sp.diff(dfx,y)
dfyy=sp.diff(dfy,y)
dfyx=sp.diff(dfy,x)
hessian_f=np.array([[dfxx, dfxy],[dfyx,dfyy]])
#Random starting point with
func_truth=np.array([[-15/14],[-1/7]])
point_main=np.array([[10],[20]])
x0=point_main[0,0]
y0=point_main[1,0]
# I am running through 27 iterations to reduce the step size using newtons method. In this method we use the formula
# xk+1=xk- inv(Hessian)*gradient.
# Since the function is a second order function we can estimate it correctly in one step!
newton norm list=[]
for i in range(27):
    z=np.linalg.inv(np.float64(hessian_f))
    j = np.array([[dfx.subs([(x,x0),(y,y0)])],[dfy.subs([(x,x0),(y,y0)])]])
    j=j.astype(float)
    diff_j=point_main-func_truth
    newton_norm_list.append(np.linalg.norm(diff_j))
    k=np.matmul(z,j)
    point main=point main-k
    x0=point_main[0,0]
    y0=point_main[1,0]
for i in range(len(newton_norm_list)):
    if newton_norm_list[i]< 10e-17:</pre>
        newton_norm_list[i]=0
    else:
        continue
print(x0)
print(y0)
x_list=[i for i in range(0,27)]
#Lets use regular gradient descent algorith
#xk+1 = xk - a*gradient where a is a very small step size
point_main_1=np.array([[10],[20]])
x1=point_main_1[0,0]
y1=point_main_1[1,0]
k=0
stp size=0.4
reg_gd=[]
for i in range(27):
    \verb|j1=np.array([[dfx.subs([(x,x1),(y,y1)])],[dfy.subs([(x,x1),(y,y1)])]|)|
    j1=j1.astype(float)
    {\tt diff\_j1=point\_main\_1-func\_truth}
    reg_gd.append(np.linalg.norm(diff_j1))
    point_main_1=point_main_1-(stp_size*j1)
    x1=point_main_1[0,0]
    y1=point_main_1[1,0]
plt.yscale('log')
```

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plt.plot(x_list,newton_norm_list,label="Newton gradient descent")
plt.xlabel('Iterations')
plt.ylabel('f(x)-f(x*)')
plt.title('Log-Linear Graph')
plt.plot(x_list,reg_gd,label="Normal gradient descent")
plt.legend()
print (x1)
print(y1)
plt.show()
```

- -1.0714285714285714
  - -0.14285714285714285
  - -1.0714285714285718
  - -0.1428571428571435

