m2-l1-p3

February 3, 2024

1 M2-L1 Problem 3 (5 points)

In this question you will perform regression on 2D data. A linear fit will be demonstrated, and afterward you will extend the code to perform a second-order fit. First, run the Setup cells.

1.1 Setup

1.1.1 Generating data

```
[12]: # Generating data for the problem
      import numpy as np
      import matplotlib.pyplot as plt
      def gaussian2d(A, mx, my, sx, sy):
          F = lambda xy: A*np.exp(-((xy[:,0]-mx)**2/(2*sx*sx))
                                   + (xy[:,1]-my)**2/(2*sy*sy)))
          return F
      def get_data_function():
          f1 = gaussian2d(A=0.7, mx = 0.25, my=0.25, sx=0.25, sy=0.25)
          f2 = gaussian2d(A=0.7, mx = 0.75, my=0.75, sx=0.25, sy=0.45)
          f = lambda xy: f1(xy) + f2(xy)
          return f
      np.random.seed(0)
      x = np.random.rand(60,2)
      f = get_data_function()
      y = f(x)
```

1.1.2 Function for 3D plotting

```
[13]: from mpl_toolkits.mplot3d import Axes3D
from matplotlib import cm

# Writing a 3D Plotting function. Inputs data points and regression function
def plot_data_with_regression(x_data, y_data, regfun=None):
    plt.figure(figsize=(8,8))
    fig = plt.gcf()
```

```
ax = fig.add_subplot(111, projection='3d')
  ax.scatter(x_data[:,0], x_data[:
,1],0*y_data,s=13,c=y_data,zorder=-1,cmap="coolwarm", alpha=1,_
⇔edgecolor="black", linewidth=0.2)
  ax.scatter(x_data[:,0], x_data[:,1],y_data,s=20,c="black",zorder=-1)
  for i in range(len(y data)):
aplot([x_data[i,0],x_data[i,0]],[x_data[i,1],x_data[i,1]],[0,y_data[i]],'k:

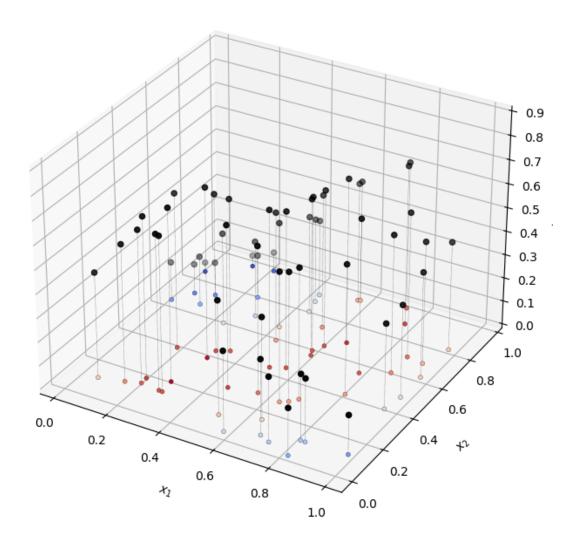
¬',linewidth=0.3)

  ax.set_xlabel('\n' + r"$x_1$")
  ax.set ylabel('\n' + r"$x 2$")
  ax.set_zlabel('\n'+r"$y$")
  ax.set_zlim(0,0.9)
  if regfun is not None:
      vals = np.linspace(0, 1, 100)
      x1grid, x2grid = np.meshgrid(vals, vals)
      y = regfun(np.concatenate((x1grid.reshape(-1,1),x2grid.

¬reshape(-1,1)),1)).reshape(*np.shape(x1grid))
      ax.plot_surface(x1grid, x2grid, y.reshape(x1grid.shape), alpha = 0.8,
⇔cmap = cm.coolwarm)
      plt.show()
```

1.1.3 Data visualized

```
[14]: plot_data_with_regression(x,y)
```



1.2 Demonstration: 2D Linear Regression

First, I generate a design matrix within a function called get_linear_design_matrix()

```
[15]: def get_linear_design_matrix(x):
    x1 = x[:,0].reshape(-1, 1)
    x2 = x[:,1].reshape(-1, 1)
    columns = [x1, x2, np.ones_like(x1)]  # Linear design matrix has a column_u
    of x1, column of x2, and a column of ones
    X = np.concatenate(columns, axis=1)  # Combine each column horizontally to_u
    omake a matrix
    return X
```

```
[16]: X = get_linear_design_matrix(x)
print("First four rows of X:")
print(X[:4,:])
```

```
First four rows of X:

[[0.5488135  0.71518937  1. ]

[0.60276338  0.54488318  1. ]

[0.4236548  0.64589411  1. ]

[0.43758721  0.891773   1. ]]
```

Next, get the coefficients of the regression:

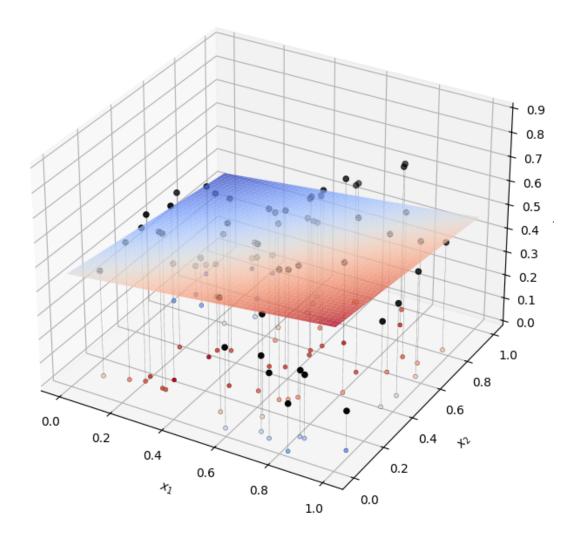
```
[17]: # Get coefficients
w1 = np.linalg.inv(X.T @ X) @ X.T @ y.reshape(-1,1)
print("Linear Coefficients:", w1.flatten())
```

```
Linear Coefficients: [ 0.11233939 -0.10638434  0.45720916]
```

Finally, we plot the result. Here, plot_data_with_regression() takes as input the x input data, y output data, and a function which performs the desired regression. Therefore I first define said regression function, and plug it in as an argument to the plotting function:

```
[18]: def do_2d_linear_regression(x):
    y_fit = get_linear_design_matrix(x) @ w1
    return y_fit

plot_data_with_regression(x, y, do_2d_linear_regression)
```



1.3 Your Turn: 2D Quadratic Regression

The linear regression results are clearly not a great fit. You will see if a 2nd order fit can do any better. Fill in the missing code below to generate a quadratic design matrix and plot the results:

```
[19]: def get_quadratic_design_matrix(x):
    x1 = x[:,0].reshape(-1, 1)
    x2 = x[:,1].reshape(-1, 1)

# YOUR CODE GOES HERE
# 2ND ORDER, 2-D DESIGN MATRIX NEEDS 6 TOTAL COLUMNS
columns = [x1**2, x2**2, x1*x2, x1, x2, np.ones_like(x1)]
X = np.concatenate(columns, axis=1)
```

```
return X
[20]: X = get_quadratic_design_matrix(x)
      print("First four rows of X:")
      print(X[:4,:])
     First four rows of X:
     [[0.30119626 0.51149583 0.39250558 0.5488135 0.71518937 1.
                                                                        ]
      [0.36332369 0.29689768 0.32843563 0.60276338 0.54488318 1.
                                                                        ]
      [0.17948339 0.41717921 0.27363614 0.4236548 0.64589411 1.
                                                                        ]
      [0.19148257 0.79525908 0.39022846 0.43758721 0.891773 1.
                                                                        ]]
[21]: # Get coefficients
      w2 = np.linalg.inv(X.T @ X) @ X.T @ y.reshape(-1,1)
      print("Quadratic Coefficients:", w2.flatten())
     Quadratic Coefficients: [-1.09949493 -0.78655383 1.62592273 0.44193704
     -0.17753776 0.55677679]
[22]: def do_2d_quadratic_regression(x):
         y_fit = get_quadratic_design_matrix(x) @ w2
         return y_fit
      plot_data_with_regression(x, y, do_2d_quadratic_regression)
```

