Homework_4

February 18, 2019

1 Data sets for Homework 4

There are three data sets in this notebook. Each data set appears in two different ways: as a list of data points and with the independent and dependent variables given separately in two lists.

Note: You may add cells using the menu Insert -> Insert Cell Above/Below. You are welcome to copy commands that we have used in other notebooks into this one.

1.0.1 **Problem 1**

The following data set gives the number of people in Bangladesh living without electricity by year: elec gives a list of data points, elecx gives a list of years, and elecy gives a list of the number of people without electricity.

```
In [1]: elec = [[1990, 97115487],
                 [1991,93190282],
                 [1992,96268982],
                 [1993,95627988],
                 [1994,95501823],
                 [1995,94060109],
                 [1996,93157160],
                 [1997,96111201],
                 [1998,91079158],
                 [1999,89888449],
                 [2000,89475245],
                 [2001,87171255],
                 [2002,85636155],
                 [2003,83949183],
                 [2004,83936648],
                 [2005,79991525],
                 [2006,71920671],
                 [2007,78719467],
                 [2008,72543169],
                 [2009,69761553],
                 [2010,68071508],
                 [2011,62180414],
                 [2012,60923202],
                 [2013,60664947],
                 [2014,59936385],
```

```
[2016,39238736]]
        elecx, elecy = map(vector, zip(*elec))
        def ones(m):
            return vector([1] * m)
        def projection(b, basis):
            return sum([b.dot_product(v)/v.dot_product(v)*v for v in basis])
        def unit(v):
            return v/v.norm()
        def vectors2matrix(vectors):
            return matrix(vectors).transpose()
        def gs(basis):
            onbasis = []
            for b in basis:
                if len(onbasis) == 0: onbasis.append(b)
                else: onbasis.append(b-projection(b, onbasis))
            return map(unit, onbasis)
        def QR(A):
            Q = vectors2matrix(gs(A.columns()))
            return Q, Q.T*A
        def poly_regression(data, k):
            ind, dep = zip(*data)
            A = np.array([ [v**j for j in range(k+1)] for v in ind])
            A = matrix(A)
            B = A.T*A
            b = A.T*vector(dep)
            coefficients = B \ b
            return coefficients * vector([x^i for i in range(k+1)])
        def plot_regression(data, k, color='blue'):
            x, y = zip(*data)
            f = poly_regression(data, k)
            return list_plot(data, color=color, size=20) + plot(f, min(x), max(x), color=color
In [2]: A = vectors2matrix([ones(len(elecx)), elecx])
        b = elecy
In [3]: Q, R = QR(A)
In [4]: y_{int}, b = n(R.inverse() * Q.T * b)
        y_int, b
```

[2015,51254335],

```
Out[4]: (3.88369074321062e9, -1.89923032173382e6)
In [5]: y_int/-(b)
Out[5]: 2044.87612627476
```

By the end of the year 2044 there will be 0 people in bangladesh without electricity

1.0.2 **Problem 2**

The following data set shows the relative risk of having an accident at a particular level of blood alcohol content. The list risk gives the data points, while riskx gives a list of blood alcohol levels and risky gives a list of relative risks.

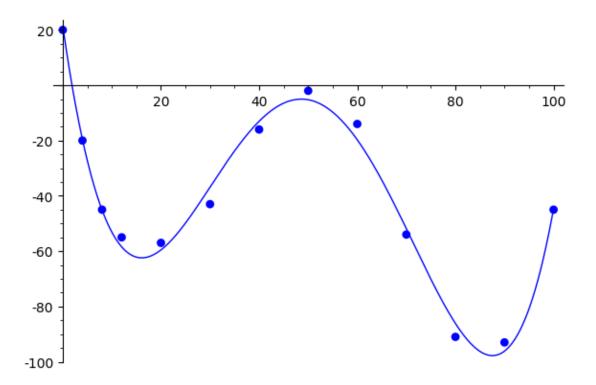
Note: If you have a list my_list, you can create a new list whose entries are the natural log of the entries in my_list by saying vector(np.log(my_list)).

```
In [6]: import numpy as np
        risk = [[0, 1],
                 [.01, 1.03],
                 [.03, 1.06],
                 [.05, 1.38],
                 [.07, 2.09],
                 [.09, 3.54],
                 [.11, 6.41],
                 [.13, 12.6],
                 [.15, 22.1],
                 [.17, 39.5],
                 [.19, 65.32],
                 [.21, 99.78]]
        riskx, risky = map(vector, zip(*risk))
In [7]: A = vectors2matrix([ones(len(riskx)), vector(riskx)])
        b = vector(np.log(risky))
In [8]: Q, R = QR(A)
        y_{int}, b = n(R.inverse() * Q.T * b)
        y_int, b
Out[8]: (-0.539980939336522, 23.8319617111994)
   Part 2a. Our equation is now y = 23.83x - 0.54 where \beta_0 = -0.54 and \beta_1 = 23.83
In [9]: C = e ** y_int
        a = b
        C, a
Out[9]: (0.582759360048180, 23.8319617111994)
   Part 2b. These are our logarithmic parameters where C = e^b = 0.5828 and a = m (in y = mx + b
form).
In [10]: np.log(1.50/C)/a
Out[10]: 0.039671348036800155
   Part 2c. Our BAC for a risk of 1.50 would be 0.0396
```

1.0.3 **Problem 3**

The following data set gives the temperature in the earth's atmosphere at various altitudes. The list temp gives a list of data points while tempx gives a list of altitudes and tempy gives a list of temperatures.

```
In [11]: def power_of(vector, power):
             new_vec = []
             for val in vector:
                 new_vec.append(val ** power)
             return new_vec
In [12]: temp = [[0, 20],
                 [4, -20],
                 [8, -45],
                 [12, -55],
                 [20, -57],
                 [30, -43],
                 [40, -16],
                 [50, -2],
                 [60, -14],
                 [70, -54],
                 [80, -91],
                 [90, -93],
                 [100, -45]]
         tempx, tempy = map(vector, zip(*temp))
In [13]: tempx_2 = power_of(tempx, 2)
         tempx_3 = power_of(tempx, 3)
         tempx_4 = power_of(tempx, 4)
         A = vectors2matrix([ones(len(tempx)), tempx, vector(tempx_2), vector(tempx_3), vector
         b = tempy
         Q, R = QR(A)
         x_hat = R.inverse() * Q.T * b
         v0, v1, v2, v3, v4 = n(x_hat)
In [14]: var('t')
         list_plot(temp, color='blue', size=40) + plot((21.310 - 12.41*t + 0.585*t^2 - 0.0092*
         list_plot(temp, color='blue', size=40) + plot((v0 + v1*t + v2*t**2 + v3*t**3 + v4*t**
Out [14]:
```



Part 3a. because of the almose sine-wave look of this data, we opted to a quartic polynomial to allow for enough peaks and valleys to adaquately fit all of the data

Part 3b. We can see here that the predicted temperature at an altitude of 55km is -9.73

In [0]: