STOCHASTIC PROCESSES

HW04

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Stochastic processes course

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1 Abstract

In this exercise, we want to plot $D^{(1)}$, $D^{(2)}$ and $D^{(4)}$ in terms of x once with binning and with error bars, and once with interaction matrix.

2 Results

2.1 Binning:

2.1.1 $D^{(1)}$

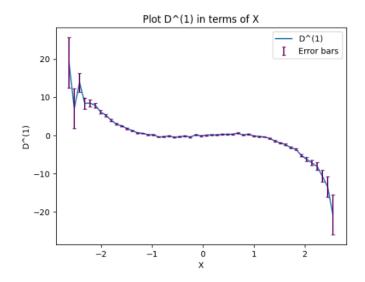


Figure 1: $D^{(1)}$ in terms of X

2.1.2 $D^{(2)}$

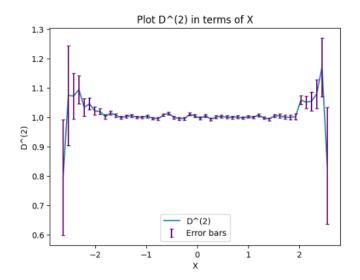


Figure 2: $D^{(2)}$ in terms of X

2.1.3 $D^{(4)}$

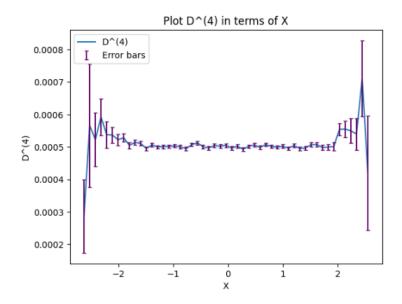


Figure 3: $D^{(4)}$ in terms of X

2.2 Interaction Matrix:

2.2.1 $D^{(1)}$

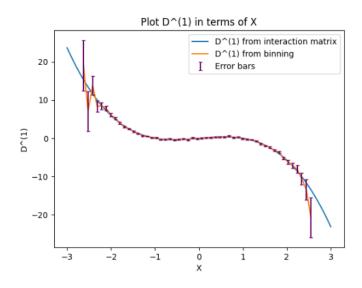


Figure 4: $D^{(1)}$ in terms of X

2.2.2 $D^{(2)}$

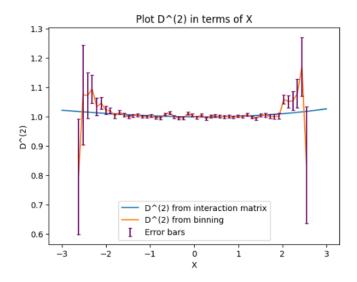


Figure 5: $D^{(2)}$ in terms of X

2.2.3 $D^{(4)}$

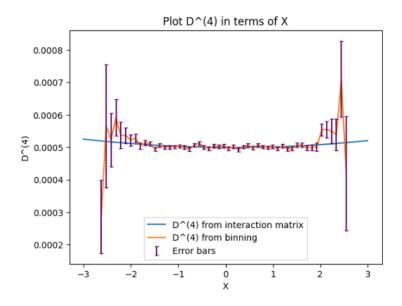


Figure 6: $D^{(4)}$ in terms of X

3 Conclusion

 $D^{(1)}$ represents our trend. Plot $D^{(1)}$ shows us that the data tends to go up in very negative X's and go down in very positive X's, but in small and medium X's, the probability of increasing and decreasing X is the same, and the data between There has no difference between going up and down.

 $D^{(2)}$ shows the intensity of these differences and movements. As if it shows the slope of the data. As we can see in the $D^{(2)}$ plot, in very large and very small X's, the value of $D^{(2)}$ is lower, which indicates that the intensity of the differences is less, but in the middle X's, we see the highest value of $D^{(2)}$ and in the value of 2, which indicates that Jumps are more intense.

Since our $D^{(4)}$ is not zero, we have jump in the data and it is not continuous.

As we can see, the results we obtained from the interaction matrix part are very consistent and similar to the results we obtained by binning, and we can conclude from.

4 Code

```
# Import needed libraries
2 import pandas as pd
3 import numpy as np
4 import matplotlib.pyplot as plt
6 # Mount drive to Read data and plot it
7 # Load get data
8 from google.colab import drive
9 drive.mount('/content/drive')
# Read data and plot it
12 # Load data
df = np.loadtxt('/content/drive/MyDrive/data/0-001.txt')
15 #Plot the data
plt.plot(df)
17 \text{ bins} = 51
18
# Calculating needed values
20 hist_values, bin_edges = np.histogram(df, bins=bins)
_{21} max = np.max(df)
22 min = np.min(df)
24 # Calculation diffrences
25 \text{ tool} = len(df)
26 diffrences = []
27 binmid_list = []
28 for i in range(bins):
   binmid = (bin_edges[i] + bin_edges[i+1]) /2
   binmid_list.append(binmid)
   diff = []
    for j in range(tool):
      if df[j]>= bin_edges[i] and df[j] < bin_edges[i+1] and j != tool-1:</pre>
33
        ekhtelaf = df[j+1] - df[j]
34
        diff.append(ekhtelaf)
    diffrences.append(diff)
36
38 # Calculating D^(1)
39 D1_dt = []
40 errors_dt =[]
41 for j in range(bins):
   d1 = (sum(diffrences[j]))/len(diffrences[j])
   res = pd.Series(diffrences[j]).var()
    sem = (res/len(diffrences[j]))**(1/2)
    errors_dt.append(sem)
    D1_dt.append(d1)
48 D1 = [i * 1000 for i in D1_dt]
49 errors = [i * 1000 for i in errors_dt]
50 plt.plot(binmid_list,D1, label='D^(1)')
51 plt.errorbar(binmid_list, D1, yerr=errors, fmt='none', color='#660066',
      capsize=2, label='Error bars')
52 plt.legend()
plt.xlabel("X")
```

```
54 plt.ylabel("D^(1)")
plt.title('Plot D^(1) in terms of X')
56 plt.show()
58 # Calculating D^(2)
59 D2_dt = []
60 errors_dt =[]
61 for j in range(bins):
    double_list = np.power(diffrences[j], 2)
     d1 = (sum(double_list))/len(diffrences[j])
    res = pd.Series(double_list).var()
     sem = (res/len(double_list))**(1/2)
     errors_dt.append(sem)
     D2_dt.append(d1)
D2 = [i * (1000/2) \text{ for } i \text{ in } D2_dt]
70 errors = [i * (1000/2) for i in errors_dt]
71 plt.plot(binmid_list,D2, label='D^(2)')
72 plt.errorbar(binmid_list, D2, yerr=errors, fmt='none', color='#660066',
       capsize=2, label='Error bars')
73 plt.legend()
74 plt.xlabel("X")
75 plt.ylabel("D^(2)")
76 plt.title('Plot D^(2) in terms of X')
77 plt.show()
79 # Calculating D^(4)
80 D4_dt = []
81 errors_dt =[]
82 for j in range(bins):
   power4_list = np.power(diffrences[j], 4)
    d1 = (sum(power4_list))/len(diffrences[j])
    res = pd.Series(power4_list).var()
     sem = (res/len(power4_list))**(1/2)
     errors_dt.append(sem)
     D4_dt.append(d1)
90 D4 = [i * (1000/24) \text{ for } i \text{ in } D4_dt]
91 errors = [i * (1000/24) for i in errors_dt]
92 plt.plot(binmid_list,D4, label='D^(4)')
93 plt.errorbar(binmid_list, D4, yerr=errors, fmt='none', color='#660066',
       capsize=2, label='Error bars')
94 plt.legend()
95 plt.xlabel("X")
96 plt.ylabel("D^(4)")
97 plt.title('Plot D^(4) in terms of X')
98 plt.show()
99
100 \text{ data} = \text{df}
101 diffr = []
for i in range(len(data)-1):
     diff = data[i+1]-data[i]
     diffr.append(diff)
105
106
108 y1 = [map(lambda x, y: x * y**i, diffr, data) for i in range(4)]
```

```
109
110 tau=1
y = data[tau:] - data[:-tau]
112 data1 = np.ones(len(data))
113 ys = np.zeros(4)
114 for i in range(4):
    ys[i] = np.mean(y)
     y *= data[:-1]
116
117
119 \text{ mmnts} = \text{np.ones}(7)
120 xk = np.ones(len(data))
for i in range(7):
    mmnts[i] = xk.mean()
    xk *= data
123
125 A = np.zeros((4,4))
126 for i in range(4):
   A[i] = np.roll(mmnts, -i)[:4]
129 A_inv = np.linalg.inv(A)
130 phis = A_inv@ys
131 phis = phis/0.001
x = np.linspace(-3,3, 10001)
134 plt.plot(x,np.array([phis[i]*x**i for i in range(4)]).sum(0), label='D^(1)
       from interaction matrix')
plt.plot(binmid_list,D1, label='D^(1) from binning')
136 plt.errorbar(binmid_list, D1, yerr=errors, fmt='none', color='#660066',
       capsize=2, label='Error bars')
plt.legend()
138 plt.xlabel("X")
139 plt.ylabel("D^(1)")
plt.title('Plot D^(1) in terms of X')
plt.show()
143 tau=1
y = data[tau:] - data[:-tau]
_{145} y = y**2
146 \text{ ys2} = \text{np.zeros}(4)
147 for i in range(4):
   ys2[i] = np.mean(y)
    y *= data[:-1]
150
152 phis2 = A_inv@ys2
_{153} phis2 = phis2/0.002
x = np.linspace(-3,3, 10001)
plt.plot(x,np.array([phis2[i]*x**i for i in range(4)]).sum(0), label='D^(2)
       from interaction matrix')
plt.plot(binmid_list,D2, label='D^(2) from binning')
plt.errorbar(binmid_list, D2, yerr=errors, fmt='none', color='#660066',
       capsize=2, label='Error bars')
plt.legend()
plt.xlabel("X")
161 plt.ylabel("D^(2)")
```

```
plt.title('Plot D^(2) in terms of X')
163 plt.show()
165 tau=1
166 y = data[tau:] - data[:-tau]
y = y**4
168 ys4 = np.zeros(4)
for i in range(4):
ys4[i] = np.mean(y)
   y *= data[:-1]
172
174 phis4 = A_inv@ys4
175 phis4 = phis4/0.024
x = np.linspace(-3,3, 10001)
178 plt.plot(x,np.array([phis4[i]*x**i for i in range(4)]).sum(0), label='D^(4)
      from interaction matrix')
plt.plot(binmid_list,D4, label='D^(4) from binning')
180 plt.errorbar(binmid_list, D4, yerr=errors, fmt='none', color='#660066',
      capsize=2, label='Error bars')
plt.legend()
182 plt.xlabel("X")
183 plt.ylabel("D^(4)")
plt.title('Plot D^(4) in terms of X')
185 plt.show()
```