

Projeto 1, 2 e PCA

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1. Projeto 1

1.1 Parte A

1.1.1 AUTÔMATOS ANALISADOS

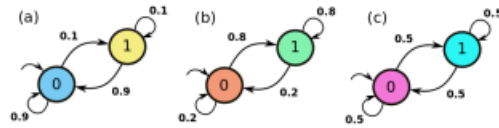


Figura 1: Figura retirada do CDT-22

1.1.2 CÓDIGOS

Parte do programa usado para gerar os padrões:

```

1 class Padroes:
2     def __init__(self,M):
3         self.matriz = np.array(M)
4         self.symbols = []
5         self.padroes_gerados = []
6     def deterministic(self, inter):
7         self.symbols = [0,1]
8         k = 0
9         for i in range(inter):
10            p = 0
11            r = np.random.random_sample()
12            for j in range(len(self.matriz)):
13                p = p + self.matriz[k][j]
14                if r < p:
15                    self.padroes_gerados.append(self.symbols[k])
16                    k = j
17                    break
18                else: continue
19            return self.padroes_gerados
20
21 fig6a = Padroes([[0.9,0.1],[0.9,0.1]])
22 fig6b = Padroes([[0.2,0.8],[0.2,0.8]])
23 fig6c = Padroes([[0.5,0.5],[0.5,0.5]])
24
25 padroes_fig6a = fig6a.deterministic(200)

```

```

26 padroes_fig6b = fig6b.deterministic(200)
27 padroes_fig6c = fig6c.deterministic(200)

```

Parte usada para gerar o stem plot:

```

1 marks = np.arange(len(padroes_fig6a))
2
3 fig, (x1, x2, x3) = plt.subplots(3, sharex = True, sharey = True, figsize=(20,
4                                     5))
5 x1.stem(marks, padroes_fig6a, 'dimgrey', linefmt=None, markerfmt = '. ')
6 x2.stem(marks, padroes_fig6b, 'dimgrey', linefmt=None, markerfmt = '. ')
7 x3.stem(marks, padroes_fig6c, 'dimgrey', linefmt=None, markerfmt = '. ')
8
9 plt.suptitle('Stem Plot')
10 plt.xlim(0, len(padroes_fig6a))
11 plt.xlabel('f')
12
13 plt.show()

```

Parte Usada para gerar o bar plot:

```

1 fig, (x1, x2, x3) = plt.subplots(3, sharex = True, sharey = True, figsize=(20,
2                                     5))
3
4 x = np.arange(len(padroes_fig6a))
5 x1.bar(x, padroes_fig6a, color = 'dimgrey')
6 x2.bar(x, padroes_fig6b, color = 'dimgrey')
7 x3.bar(x, padroes_fig6c, color = 'dimgrey')
8
9 plt.suptitle('Barplot')
10 plt.xlabel('f')
11 pylab.xlim(0, len(padroes_fig6a))
12 pylab.ylim(0,1)
13 plt.show()

```

Parte para gerar o Square Wave plot:

```

1 fig, (x1, x2, x3) = plt.subplots(3, sharex = True, sharey = True, figsize=(20,
2                                     5))
3
4 plt.suptitle("Square wave")
5 x1.step(arange(0, len(padroes_fig6a)), padroes_fig6a, color = 'dimgrey')
6 x2.step(arange(0, len(padroes_fig6b)), padroes_fig6b, color = 'dimgrey')
7 x3.step(arange(0, len(padroes_fig6c)), padroes_fig6c, color = 'dimgrey')
8
9 plt.show()

```

Parte para obter a média e desvio padrão da frequência relativa de símbolos 1's.

```

1 def freq(automato, symbol):
2     automato = np.array(automato)
3     c = 0
4     for i in automato:
5         if i==symbol:
6             c +=1
7     return c/len(automato)
8

```

```

9 freq_fig6a = []
10 freq_fig6b = []
11 freq_fig6c = []
12 k = fig6a.deterministic(2000)
13 p = fig6b.deterministic(2000)
14 t = fig6c.deterministic(2000)
15 for i in range(200,2000,200):
16     freq_fig6a.append(freq(k[:i],1))
17     freq_fig6b.append(freq(p[:i],1))
18     freq_fig6c.append(freq(t[:i],1))
19     sns.distplot(freq_fig6a, hist = False, kde = True, color = 'darkblue')
20     sns.distplot(freq_fig6b, hist = False, kde = True, color = 'lime')
21     sns.distplot(freq_fig6c, hist = False, kde = True, color = 'tomato')
22
23 print('Media fig 6a:', np.mean(freq_fig6a))
24 print('Desvio fig 6a:', np.std(freq_fig6a))
25
26 print('Media fig 6b:', np.mean(freq_fig6b))
27 print('Desvio fig 6b:', np.std(freq_fig6b))
28
29 print('Media fig 6c:', np.mean(freq_fig6c))
30 print('Desvio fig 6c:', np.std(freq_fig6c))
31
32 plt.title('Density Plot')
33 plt.xlabel('f')
34 plt.xlim(0,1)
35 plt.ylabel('Density')
36 plt.show()

```

1.1.3 PLOTS

Para 500 interações

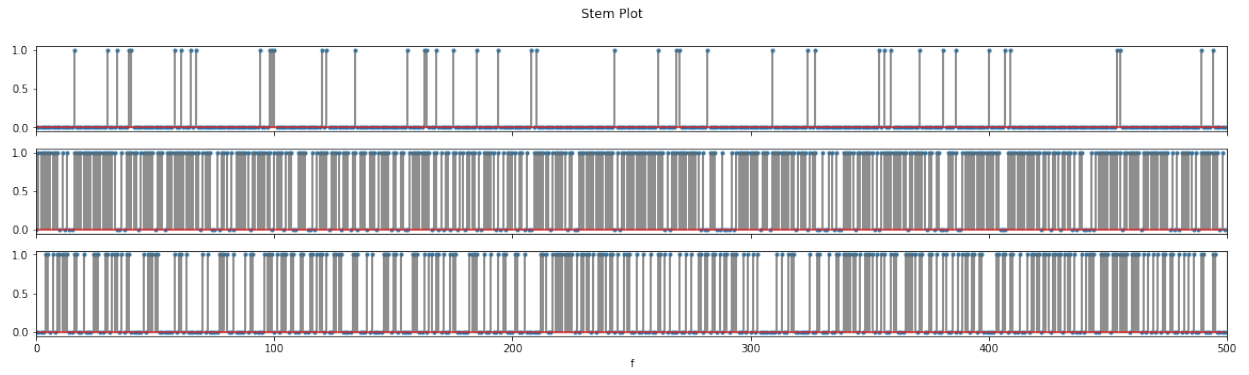


Figura 2: Stem plot, 500 interações; Figura 6a, 6b,6c

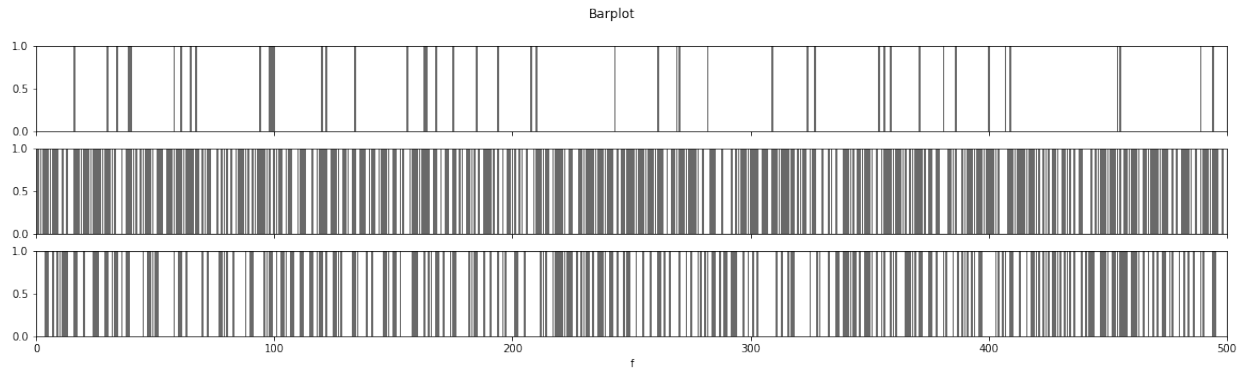


Figura 3: Bar plot, 500 interações; Figura 6a, 6b,6c

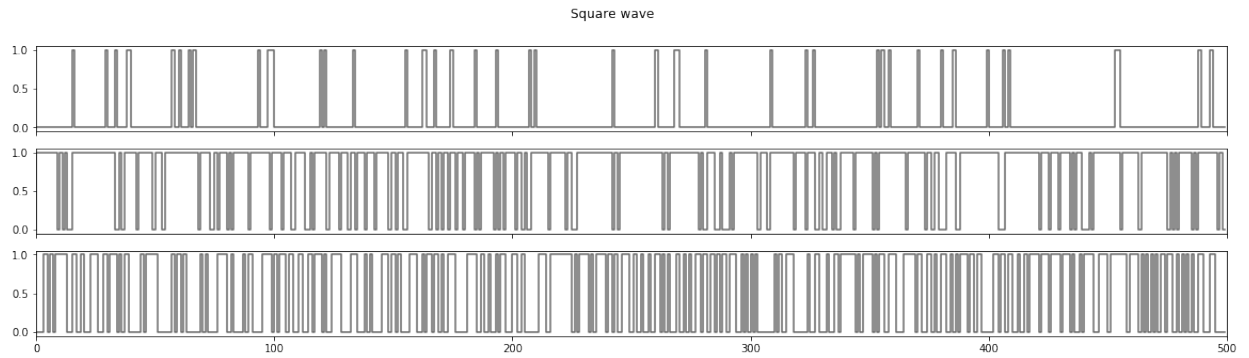


Figura 4: Square Wave, 500 interações; Figura 6a, 6b,6c

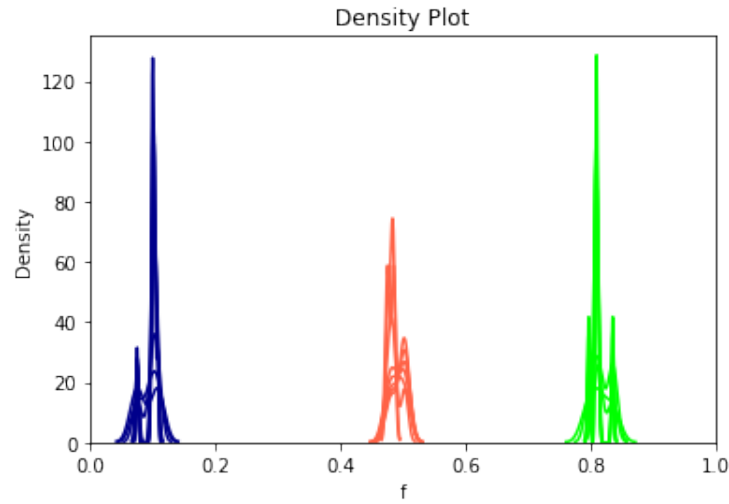


Figura 5: Densidade dos 1's, azul representa o automato da figura 6a, vermelho a figura 6c e o verde a figura 6b

Média fig 6a: 0.09933492063492062
 Desvio fig 6a: 0.004870384740271202
 Média fig 6b: 0.7907349206349207
 Desvio fig 6b: 0.005892153504033305
 Média fig 6c: 0.4985365079365079
 Desvio fig 6c: 0.012015512355410119

Figura 6: Resultados frequência dos 1's

1.2 Parte B

1.2.1 AUTÔMATOS ANALISADOS

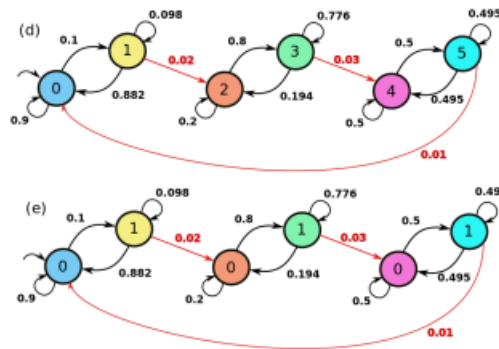


Figura 7: Figura retirada do CDT-22

1.2.2 CÓDIGOS

```

1 class Automatos:
2     def __init__(self, M):
3         self.matriz = np.array(M)
4         self.symb = []
5         self.padroes = []
6     def symbols(self, S):
7         self.symb = S
8     def deterministic(self, inter):
9         k = 0
10        for i in range(inter):
11            r = np.random.random_sample()
12            a = 0
13            for j in range(len(self.matriz)):
14                a = a + self.matriz[k][j]
15                if r < a:
16                    self.padroes.append(self.symb[k])
17                    k = j
18                    break
19        return self.padroes
20 fig_matrix = [
21     [0.9,0.1,0,0,0,0],
22     [0.882,0.098,0.02,0,0,0],
23     [0,0,0.2,0.8,0,0],
24     [0,0,0.194,0.776,0.03,0],
25     [0,0,0,0,0.5,0.5],
26     [0.01,0,0,0,0.495,0.495]
27 ]
28 s_fig1d = [0,1,2,3,4,5]
29 s_fig1e = [0,1,0,1,0,1]
30
31 fig1d = Automatos(fig_matrix)
32 fig1d.symbols(s_fig1d)
33 padroes_fig1d = fig1d.deterministic(500)
34
35 fig1e = Automatos(fig_matrix)
36 fig1e.symbols(s_fig1e)
37 padroes_fig1e = fig1e.deterministic(500)

```

Para gerar o Stem Plot:

```

1 marks_fig1d = np.arange(len(padroes_fig1d))
2 fig, (x1, x2) = plt.subplots(2, sharex = True, figsize=(20, 5))
3 x1.stem(marks_fig1d, padroes_fig1d, 'dimgrey', markerfmt = 'r.')
4 x2.stem(marks_fig1d, padroes_fig1e, 'dimgrey', markerfmt = 'r.')
5 plt.xlim(0, len(padroes_fig1d))
6 plt.suptitle('Stem Plot')
7 plt.xlabel('f')
8 plt.show()

```

Para gerar o Bar Plot:

```

1 fig, (x1, x2) = plt.subplots(2, sharex = True, figsize=(20, 5))
2
3 x = np.arange(len(padroes_fig1d))
4 x1.bar(x, padroes_fig1d, color = 'dimgrey')

```

```

5 x2.bar(x, padroes_fig1e, color = 'dimgrey')
6
7 plt.suptitle('Barplot')
8 plt.xlim(0, len(padroes_fig1d))
9 plt.xlabel('f')
10
11 plt.show()

```

Para gerar o Square Wave plot:

```

1 fig, (x1, x2) = plt.subplots(2, sharex = True, figsize=(20, 5))
2
3 plt.suptitle("Square wave")
4 plt.xlim(0, len(padroes_fig1d))
5 x1.step(arange(0, len(padroes_fig1d)), padroes_fig1d, color = 'dimgrey')
6 x2.step(arange(0, len(padroes_fig1e)), padroes_fig1e, color = 'dimgrey')
7 plt.show()

```

Parte para obter a média e desvio padrão da frequência relativa de símbolos 1's.

```

1 def freq(automato, symbol):
2     automato = np.array(automato)
3     c = 0
4     for i in automato:
5         if i==symbol:
6             c +=1
7     return c/len(automato)
8
9 freq_fig6d = []
10 freq_fig6e = []
11
12 k = fig1d.deterministic(2000)
13 p = fig1e.deterministic(2000)
14
15 for i in range(200,2000,200):
16     freq_fig6d.append(freq(k[:i],1))
17     freq_fig6e.append(freq(p[:i],1))
18     sns.distplot(freq_fig6d, hist = False, kde = True, color = 'darkblue')
19     sns.distplot(freq_fig6e, hist = False, kde = True, color = 'tomato')
20
21 print('Media fig 6d:', np.mean(freq_fig6d))
22 print('Desvio fig 6d:', np.std(freq_fig6d))
23
24 print('Media fig 6e:', np.mean(freq_fig6e))
25 print('Desvio fig 6e:', np.std(freq_fig6e))
26
27
28 plt.xlabel('f')
29 plt.ylabel('Density')
30 plt.title('Density Plot')
31 plt.show()

```

1.2.3 PLOTS

Para 500 interações

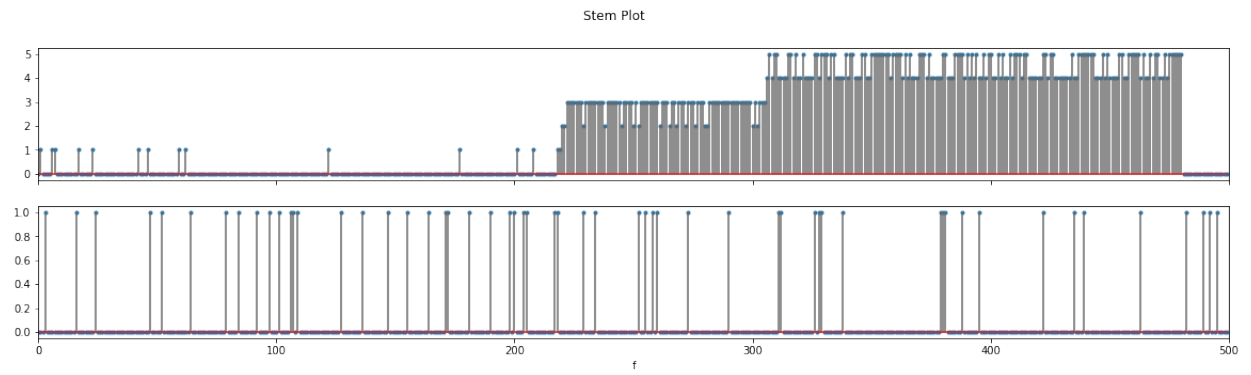


Figura 8: Stem Plot, 500 interações; Figura 6d, 6e

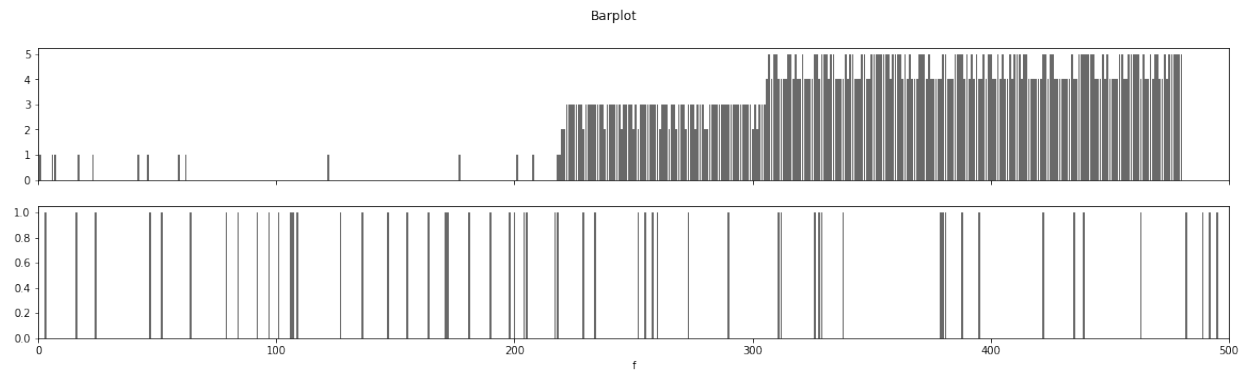


Figura 9: Bar Plot, 500 interações; Figura 6d, 6e

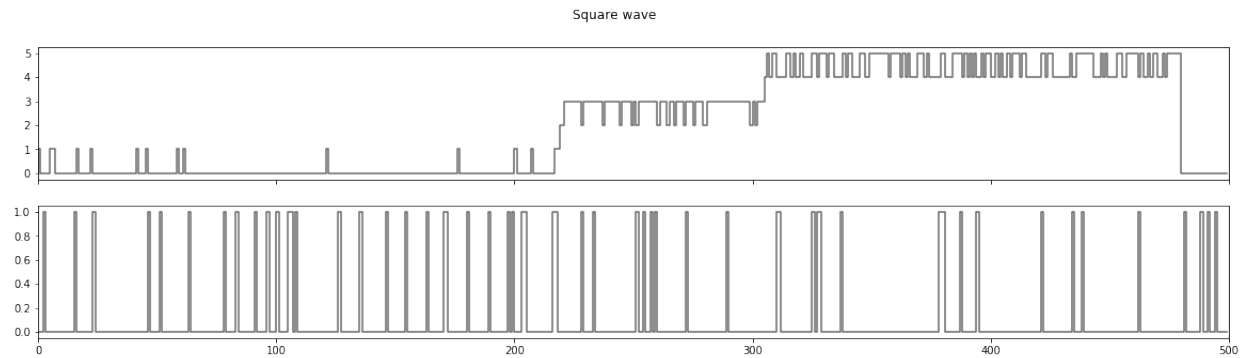


Figura 10: Square Wave, 500 interações; Figura 6d, 6e

Media fig 6d: 0.0711111111111112
Desvio fig 6d: 0.013375223547226103
Media fig 6e: 0.23334259259259257
Desvio fig 6e: 0.06948821336731983

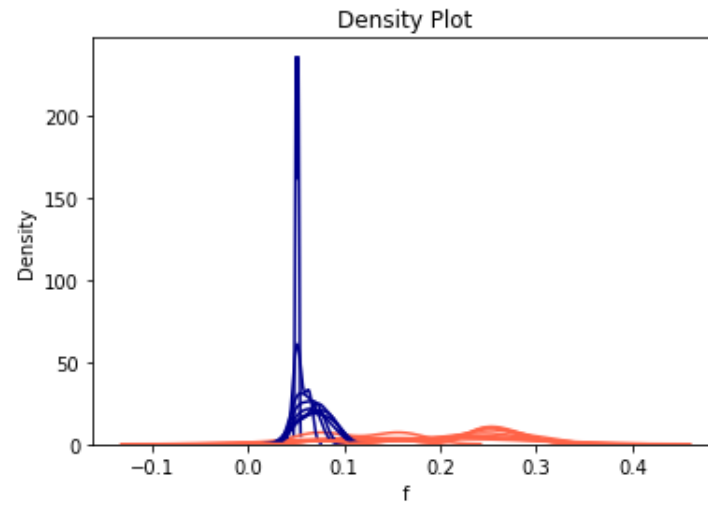


Figura 11: Densidade dos 1's, azul representa o automato da figura 6d e o vermelho a figura 6e, junto com as suas médias e desvios padrão.

2. Projeto 2

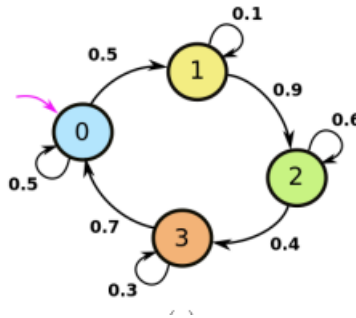


Figura 12: Automato analisado. Figura retirada do CDT-23

Bibliotecas usadas:

```

1 from random import random
2 import numpy as np
3 import matplotlib.pyplot as plt
4 from scipy import signal
5 from scipy import stats
6 import pylab
7 import math

```

Para gerar os padrões da série temporal foi usado o seguinte código:

```

1 class Automatos:
2     def __init__(self, M):
3         self.matriz = np.array(M)
4         self.symb = []
5         self.padroes = []
6     def symbols(self, S):
7         self.symb = S
8     def deterministic(self, inter):
9         k = 0
10        for i in range(inter):
11            r = np.random.random_sample()
12            a = 0
13            for j in range(len(self.matriz)):
14                a = a + self.matriz[k][j]
15                if r < a:
16                    self.padroes.append(self.symb[k])
17                    k = j
18                    break
19        return self.padroes
20
21 M = [[0.5, 0.5, 0, 0],
22      [0, 0.1, 0.9, 0],
23      [0, 0, 0.6, 0.4],
24      [0.7, 0, 0, 0.3]
25      ]
26
27 auto = Automatos(M)

```

```

28 auto.symbols([0,1,2,3])
29 padroes_fig2 = auto.deterministic(200)

```

Parte para gerar os split signals:

```

1 def main():
2     split_fig2 = split_signal()
3     split_fig2.split(padroes_fig2)
4     split_fig2.relative_frequency()
5     split_fig2.plots()
6
7 class split_signal:
8     def __init__(self):
9         self.list_zero = []
10        self.list_one = []
11        self.list_two = []
12        self.list_three = []
13        self.lista_burst = []
14        self.relative_fre = []
15    def split(self, lista):
16        for i in range(len(lista)):
17            if lista[i] == 0:
18                self.list_zero.append(1)
19            else:
20                self.list_zero.append(0)
21        for i in range(len(lista)):
22            if lista[i] == 1:
23                self.list_one.append(1)
24            else:
25                self.list_one.append(0)
26        for i in range(len(lista)):
27            if lista[i] == 2:
28                self.list_two.append(1)
29            else:
30                self.list_two.append(0)
31        for i in range(len(lista)):
32            if lista[i] == 3:
33                self.list_three.append(1)
34            else:
35                self.list_three.append(0)
36    def relative_frequency(self):
37        for i in range(len(self.list_zero)):
38            self.relative_fre.append((self.list_zero[i])/(len(self.list_zero)))
39
40    def plots(self):
41        fig, axs = plt.subplots(4, sharex = True)
42        fig.suptitle('Split signals')
43        x = np.arange(len(padroes_fig2))
44        axs[0].bar(x, self.list_zero, color = 'dimgrey')
45        axs[1].bar(x, self.list_one, color = 'dimgrey')
46        axs[2].bar(x, self.list_two, color = 'dimgrey')
47        axs[3].bar(x, self.list_three, color = 'dimgrey')
48        pylab.xlim(0, len(padroes_fig2))
49        pylab.ylim(0,1)
50

```

```

51 if __name__ == "__main__":
52     main()

```

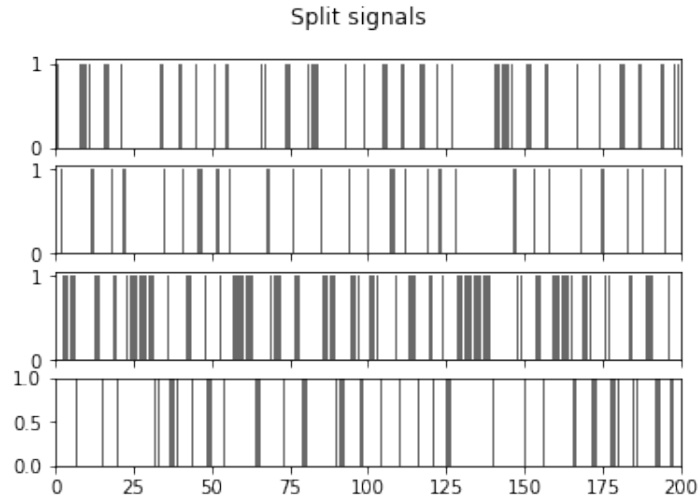


Figura 13: Split Signal

2.1 Parte 1

Número de bursts e média, desvio padrão, entropia e evenness dos tamanhos dos bursts em cada split signal;

2.1.1 CÓDIGOS

```

1 def main():
2     burst_fig2 = burst()
3     burst_fig2.split(padroes_fig2)
4     burst_fig2.relative_frequency()
5     burst_fig2.burst(len(padroes_fig2) - 1)
6     burst_fig2.media_burst()
7     burst_fig2.desvio_burst()
8     burst_fig2.entropia()
9     burst_fig2.plot()
10
11 class burst(split_signal):
12     def __init__(self):
13         split_signal.__init__(self)
14         self.lista_burst = []
15     def burst(self, interacoes):
16         i = 1
17         L = self.list_zero
18         while (i <= interacoes):
19             if (L[i] == 1):
20                 i_0 = 1
21                 while (L[i] == 1) and (i < interacoes):
22                     i += 1

```

```

23         if (i == M) and (L[i] == 1):
24             i = M-1
25             bs = i-i_0
26             if (bs>0):
27                 self.lista_burst.append(bs)
28             i += 1
29     def media_burst(self):
30         media = np.mean(self.lista_burst)
31         print("Media:", media)
32     def desvio_burst(self):
33         desvio = np.std(self.lista_burst)
34         print("Desvio padrao:", desvio)
35     def entropia(self):
36         indices = []
37         for i in range(len(self.relative_fre)):
38             if self.relative_fre[i] != 0:
39                 k = (self.relative_fre[i])*(math.log(self.relative_fre[i], 2))
40                 indices.append(k)
41         epsilon = (-1)*sum(indices)
42         eta = 2*((-1)*sum(indices))
43         print("Entropia:", epsilon)
44         print("Evenness:", eta)
45     def plot(self):
46         plt.title("Burst")
47         contador = []
48         for i in range(len(self.lista_burst)):
49             contador.append(1)
50         plt.stem(self.lista_burst, contador, markerfmt = '. ')
51         plt.show()
52
53
54 if __name__ == '__main__':
55     main()

```

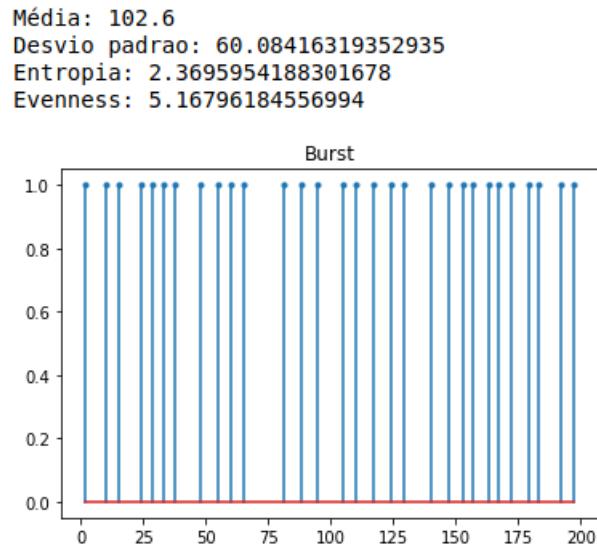


Figura 14: Burst

2.2 Parte 2

Número de distâncias intersímbolos e média, desvio padrão, entropia e evenness das distâncias intersímbolos em cada split signal;

2.2.1 CÓDIGOS

```

1 def main():
2     intersymbol_fig2 = inter_symbol()
3     intersymbol_fig2.split(padroes_fig2)
4     intersymbol_fig2.distance(len(padroes_fig2))
5     intersymbol_fig2.distancias()
6     intersymbol_fig2.number_distance()
7     intersymbol_fig2.media_desvio()
8     intersymbol_fig2.entropia_evenness()
9     intersymbol_fig2.plot()
10
11 class inter_symbol(split_signal):
12     def __init__(self):
13         split_signal.__init__(self)
14         self.list_inter_symbol = []
15         self.list_distancias = []
16     def distance(self, interacoes):
17         L = self.list_zero
18         for i in range(interacoes):
19             if(L[i] == 1):
20                 bs = 0
21                 while(L[i] == 0) & (i < interacoes):
22                     i += 1
23                 if (i == interacoes) & (L[i] == 1):
24                     bs = interacoes

```

```

25         elif (i == interacoes) & (L[i] == 0):
26             bs = 0
27         elif (i < interacoes) & (L[i] == 1):
28             bs = i
29         if (bs > 0):
30             self.list_inter_symbol.append(bs)
31         i += 1
32     def number_distance(self):
33         print("Numero de distancias intersimbolos:", len(self.
list_inter_symbol))
34     def distancias(self):
35         for i in range(0, (len(self.list_inter_symbol)-1)):
36             k = self.list_inter_symbol[i+1] - self.list_inter_symbol[i]
37             self.list_distancias.append(k)
38     def media_desvio(self):
39         media_ = sum(self.list_distancias)/len(self.list_distancias)
40         desvio_ = np.std(self.list_distancias)
41         print("Media:", media_)
42         print("Desvio:", desvio_)
43     def entropia_evenness(self):
44         k = []
45         indices = []
46         for i in range(len(self.list_distancias)):
47             k.append(self.list_distancias[i]/len(self.list_distancias))
48         for i in range(len(k)):
49             if k[i] != 0:
50                 indices_ = ((k[i])*(math.log(k[i], 2)))
51                 indices.append(indices_)
52         entropia = -(sum(indices))
53         evenness = 2**(entropia)
54         print("Entropia:", entropia)
55         print("Evenness:", evenness)
56     def plot(self):
57         plt.title("Inter Symbol distance")
58         contador = []
59         for i in range(len(self.list_inter_symbol)):
60             contador.append(1)
61         plt.stem(self.list_inter_symbol, contador, markerfmt = 'r.')
62         plt.show()
63
64
65 if __name__ == '__main__':
66     main()

```

Média: 3.2666666666666666
 Desvio: 2.688659310676771
 Entropia: 12.20160841294563
 Evenness: 4710.316912269184

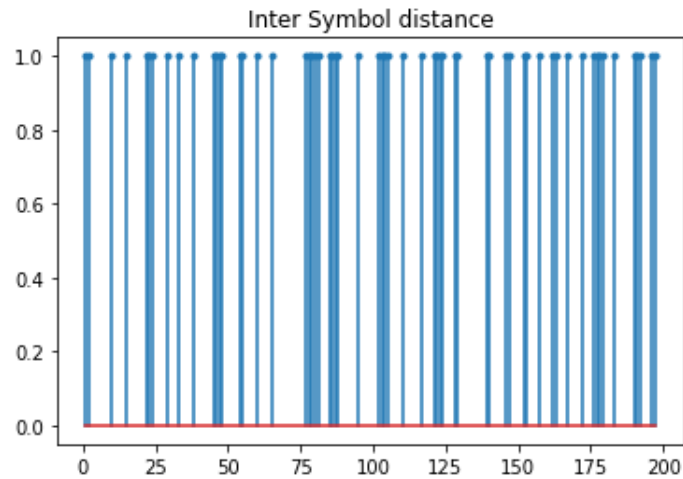


Figura 15: Intersymbol distance

2.3 Parte 3

Média e desvio padrão das magnitudes do espectro de potência da transformada de Fourier discreta de cada split signal (pode usar rotina para FFT);

2.3.1 CÓDIGOS

```

1 def main():
2     fourier_fig2 = fourier_transform()
3     fourier_fig2.split(padroes_fig2)
4     fourier_fig2.transform(len(padroes_fig2))
5     fourier_fig2.media()
6     fourier_fig2.desvio()
7     fourier_fig2.plot()
8
9 class fourier_transform(split_signal):
10     def __init__(self):
11         split_signal.__init__(self)
12         self.transformada = []
13     def transform(self, interacoes):
14         L = self.list_zero
15         self.transformada = np.fft.fft(L, (interacoes-1))
16     def media(self):
17         media = np.mean(self.transformada)
18         print("Media:", media)
19     def desvio(self):
20         desvio_padrao = np.std(self.transformada)
21         print("Desvio padrao:", desvio_padrao)

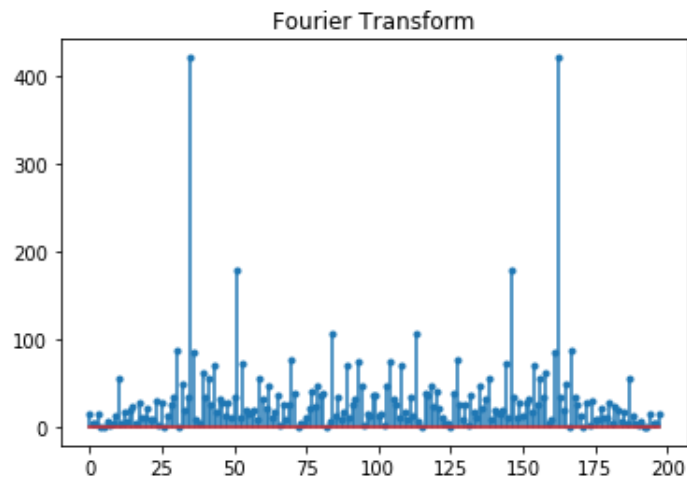
```



```

22     def plot(self):
23         plt.title("Fourier Transform")
24         modulo = []
25         for i in range(len(self.transformada)):
26             modulo.append((np.abs(self.transformada[i]))**2)
27         plt.stem(modulo, markerfmt = 'b.')
28         plt.show()
29
30 if __name__ == "__main__":
31     main()

```



Media: 30.27272727272727
Desvio padrão: 47.65168617784987

Figura 16: Power Spectrum

2.4 Parte 4

Média e desvio do grau e coeficiente de aglomeração de grafos dos sinais originais, obtidos pelo método de visibilidade, que deve ser implementado;

2.4.1 CÓDIGOS

```

1 def main():
2     network_fig2 = network_based()
3     network_fig2.visibility(padroes_fig2)
4     network_fig2.average_degree()
5     network_fig2.clustering_coefficient()
6     #network_fig2.plot()
7     network_fig2.tentativa_plot()
8
9 class network_based(split_signal):
10     def __init__(self):
11         split_signal.__init__(self)
12         self.M = np.zeros((len(padroes_fig2), len(padroes_fig2)), dtype=int)
13     def visibility(self, L):
14         for j in range(1, len(padroes_fig2)):
15             for i in range(0, j-1):
16                 flag = 1
17                 k = i + 1
18                 while (k <= j-1) and (flag == 1):
19                     aux = L[j] + (L[i]-L[j])*(j-k)/(j-i)
20                     if (L[k] >= aux):
21                         flag = 0
22                     k += 1
23                 if (flag == 1):
24                     self.M[i][j] = 1
25                     self.M[j][i] = 1
26     def average_degree(self):
27         edges = 0
28         for i in range(len(padroes_fig2)):
29             for j in range(len(padroes_fig2)):
30                 if self.M[i][j] == 1:
31                     edges += 1
32         average = edges/(len(padroes_fig2))
33         standard_deviation = self.M.std()
34         print("Media:", average)
35         print("Desvio padrao:", standard_deviation)
36     def clustering_coefficient(self):
37         degree_node = []
38         contador = 0
39         for i in range(len(self.M)):
40             degree_node.append(sum(self.M[i]))
41             contador += 1
42         clustering_coefficient = (sum(degree_node)*2)/(contador*(contador-1))
43         print("Coeficiente de aglomeracao:", clustering_coefficient)
44     def tentativa_plot(self):
45         M_array = np.array(self.M)
46         plt.matshow(M_array)
47         plt.title("Visibility procedure")
48         ax = plt.gca()

```

```

49     ax.axes.xaxis.set_visible(False)
50     ax.axes.yaxis.set_visible(False)
51     plt.show()
52     def plot(self):
53         x = []
54         y = []
55         for i in range(len(padroes_fig2)):
56             for j in range(len(padroes_fig2)):
57                 if self.M[i][j] == 1:
58                     x.append(i)
59                     y.append(j)
60         #print(x)
61         plt.title("Visibility procedure")
62         plt.scatter(x,y,s = 1.5)
63         plt.show()
64
65 if __name__ == '__main__':
66     main()

```

Média: 2.77
Desvio padrão: 0.1168682056848654
Coeficiente de aglomeração: 0.0278391959798995

Visibility procedure

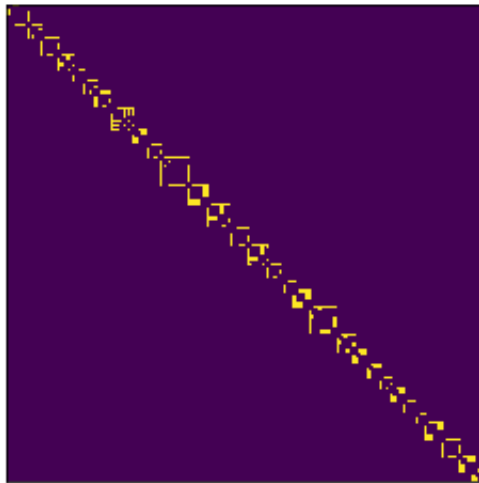


Figura 17: Visibility procedure

3. PCA

CÓDIGOS

```

1 def main():
2     pca_preparo = PCA()
3     pca_preparo.aleatorios(600)
4     pca_preparo.circulo()
5     pca_preparo.alongamento()
6     pca_preparo.rotacao()
7     pca_preparo.covariancia()
8     pca_preparo.subplot()
9     pca_preparo.eigen()
10    pca_preparo.plot()
11
12
13 class PCA:
14     def __init__(self):
15         self.aleatorios_x = []
16         self.aleatorios_y = []
17         self.circulo_x = []
18         self.circulo_y = []
19         self.circulo_y_alongado = []
20         self.rotacao = []
21         self.cov_matrix = []
22         self.eigenvalue = []
23         self.eigenvector = []
24         self.new_circulo_x = []
25         self.new_circulo_y = []
26
27     def aleatorios(self, interacoes):
28         for i in range(interacoes):
29             k = random.uniform(-1,1)
30             h = random.uniform(-1,1)
31             self.aleatorios_x.append(k)
32             self.aleatorios_y.append(h)
33
34     def circulo(self):
35         for i in range(len(self.aleatorios_x)):
36             k = math.sqrt((self.aleatorios_y[i])**2 + (self.aleatorios_x[i])
37             **2)
38             if k <= 1:
39                 self.circulo_x.append(self.aleatorios_x[i])
40                 self.circulo_y.append(self.aleatorios_y[i])
41             else:
42                 continue
43
44     def alongamento(self):
45         for i in range(len(self.circulo_y)):
46             self.circulo_y_alongado.append(0)
47         for i in range(len(self.circulo_y)):
48             self.circulo_y_alongado[i] = 0.2*self.circulo_y[i]
49
50     def rotacao(self):
51         k = np.radians(30)
52         rot = [[np.cos(k), np.sin(k)], [np.sin(k), np.cos(k)]]

```

```

52         self.rotacao_ = np.dot(rot,[ self.circulo_x , self.circulo_y_alongado])
53
54     def covariancia(self):
55         self.cov_matrix = np.cov(self.rotacao_)
56
57     def eigen(self):
58         self.eigenvalue , self.eigenvector = LA.eig(self.cov_matrix)
59         self.eigenvalue.sort()
60         self.eigenvalue = self.eigenvalue[::-1]
61         eta = self.eigenvalue[0]/(sum(self.eigenvalue))
62         print('Autovetor 1:', self.eigenvector[0])
63         print('Lambda 1:', self.eigenvalue[0])
64         print('Autovetor 2:', self.eigenvector[1])
65         print('Lambda 2:', self.eigenvalue[1])
66         print('Eta:', eta)
67
68     def plot(self):
69         origin = [0,0]
70         plt.title("PCA")
71         plt.xlabel("x")
72         plt.ylabel("y")
73         plt.quiver(*origin , *self.eigenvector[:,0], color = 'b' ,
74                   width = 0.004, scale_units='xy', scale=2)
75         plt.quiver(*origin , *self.eigenvector[:,1], color = 'r' ,
76                   width = 0.004, scale_units='xy', scale=2)
77         plt.axis('equal')
78         plt.scatter(self.rotacao_[0], self.rotacao_[1], s = 2, color = 'dimgrey'
79 )
80
81         plt.xlim(-1,1)
82         plt.ylim(-1,1)
83         plt.show()
84
85     def subplot(self):
86         fig , ax = plt.subplots(1,3, figsize=(6*3, 6), sharex = True, sharey =
87 True)
88         plt.xlim(-1,1)
89         plt.ylim(-1,1)
90         fig.suptitle("Distribuicao")
91         ax[0].scatter(self.circulo_x , self.circulo_y , s = 2)
92         ax[0].plot()
93         ax[1].scatter(self.circulo_x , self.circulo_y_alongado , s = 2)
94         ax[2].scatter(self.rotacao_[0], self.rotacao_[1], s = 2)
95         plt.show()
96
97 if __name__ == '__main__':
98     main()

```

3.1 Visualização dos dados, para ver se está parecido com Figura 9, CDT-24

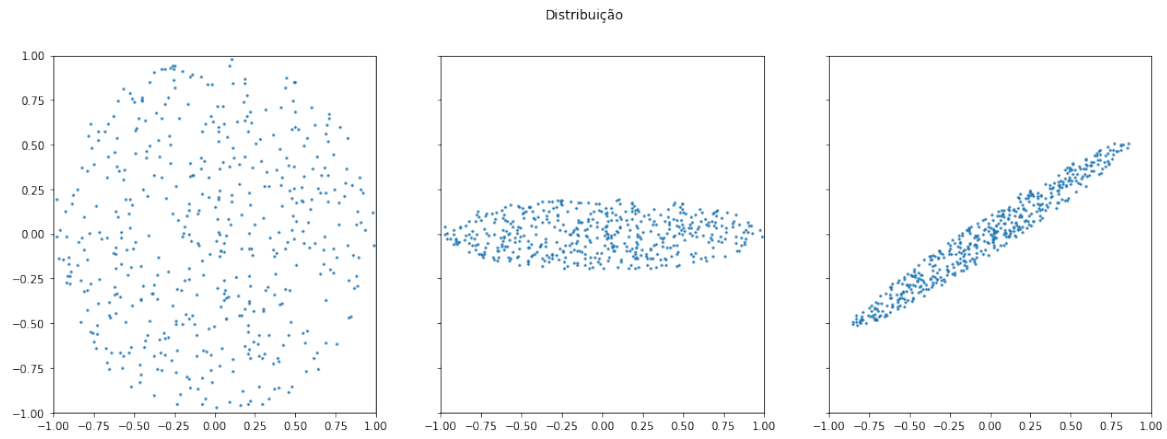


Figura 18: Distribuição

3.2 Resultado final

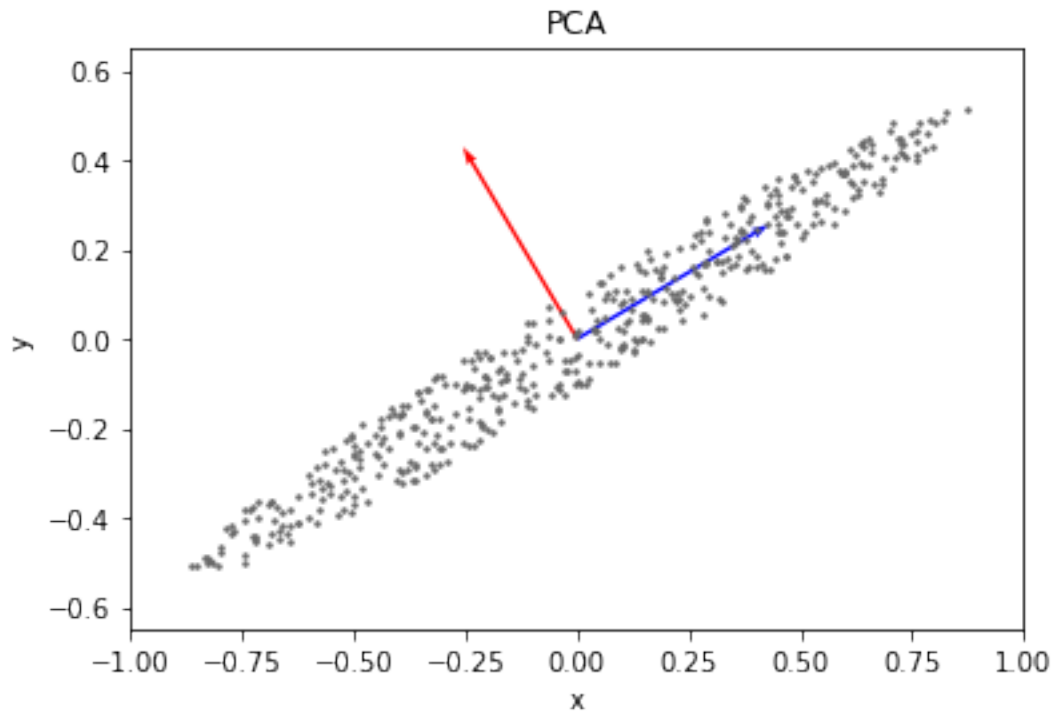


Figura 19: PCA