

questao03

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1 Questão 3 - Prova 2 de Inteligência Artificial

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2 1. Pré-processamento de dados

```
In [63]: # Importando as bibliotecas
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import warnings
import matplotlib.cbook
warnings.filterwarnings("ignore",category=matplotlib.cbook.mplDeprecation)

# Importando os datasets
dataset_treino = pd.read_csv('base_treinamento.csv')
X = dataset_treino.iloc[:, :-2].values
y = dataset_treino.iloc[:, [5,6]].values

dataset_teste = pd.read_csv('base_teste.csv')
Xt = dataset_teste.iloc[:, :-2].values
yt = dataset_teste.iloc[:, [5,6]].values
```

```
In [2]: dataset_treino.describe()
```

```
Out[2]:
```

	AP1	AP2	AP3	AP4	AP5 \
count	262.00000	262.00000	262.00000	262.00000	262.00000
mean	-67.90458	-62.129771	-58.122137	-71.454198	-65.183206
std	10.29165	8.530672	8.323924	9.232823	7.128596
min	-90.00000	-81.00000	-75.00000	-86.00000	-84.00000
25%	-75.75000	-68.00000	-65.00000	-78.00000	-69.00000
50%	-69.00000	-63.00000	-58.00000	-74.00000	-65.00000
75%	-61.00000	-57.00000	-53.00000	-66.25000	-60.00000
max	-38.00000	-35.00000	-33.00000	-38.00000	-51.00000

Coordenada X Coordenada Y

count	262.000000	262.000000
mean	12.026985	12.612634
std	5.079505	5.139619
min	1.390000	1.570000
25%	8.442500	9.412500
50%	12.400000	13.630000
75%	16.357500	16.710000
max	20.450000	19.630000

In [3]: dataset_teste.describe()

Out [3]:

	AP1	AP2	AP3	AP4	AP5	Coordenada X \
count	65.000000	65.000000	65.000000	65.000000	65.000000	65.000000
mean	-56.953846	-63.815385	-62.030769	-69.261538	-65.815385	12.094923
std	8.088245	7.717456	8.154694	9.361285	10.037262	5.005830
min	-71.000000	-82.000000	-76.000000	-82.000000	-86.000000	2.410000
25%	-63.000000	-69.000000	-67.000000	-76.000000	-74.000000	8.370000
50%	-58.000000	-64.000000	-62.000000	-73.000000	-68.000000	12.950000
75%	-52.000000	-58.000000	-58.000000	-64.000000	-59.000000	15.480000
max	-39.000000	-51.000000	-43.000000	-43.000000	-42.000000	20.460000

	Coordenada Y
count	65.000000
mean	12.153231
std	4.951980
min	2.110000
25%	9.090000
50%	12.580000
75%	17.490000
max	19.120000

2.0.1 Melhorando a visualização dos datasets...

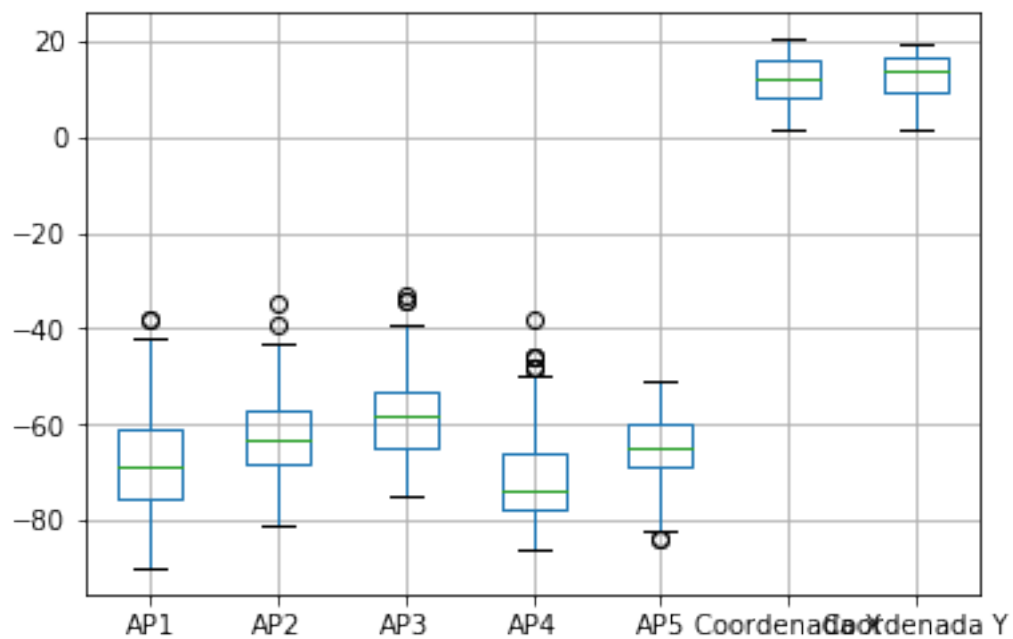
In [4]: `print(pd.options.display.max_columns, pd.options.display.max_rows)`
`pd.options.display.max_columns = 7`
`pd.options.display.max_rows = 262`

20 60

2.0.2 Visualizando os datasets em formato de boxplot...

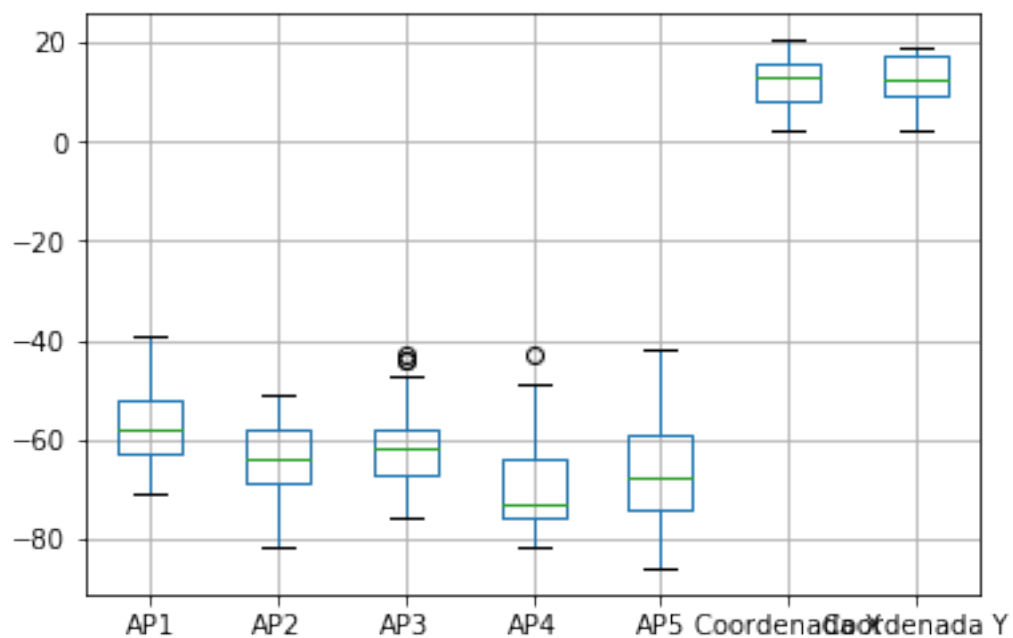
In [5]: dataset_treino.boxplot()

Out [5]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa294297fd0>



In [6]: dataset_teste.boxplot()

Out[6]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa2906a2470>



2.0.3 Checando se existe algum dado incompleto nos datasets...

```
In [7]: dataset_treino.isnull().values.any()
```

```
Out[7]: False
```

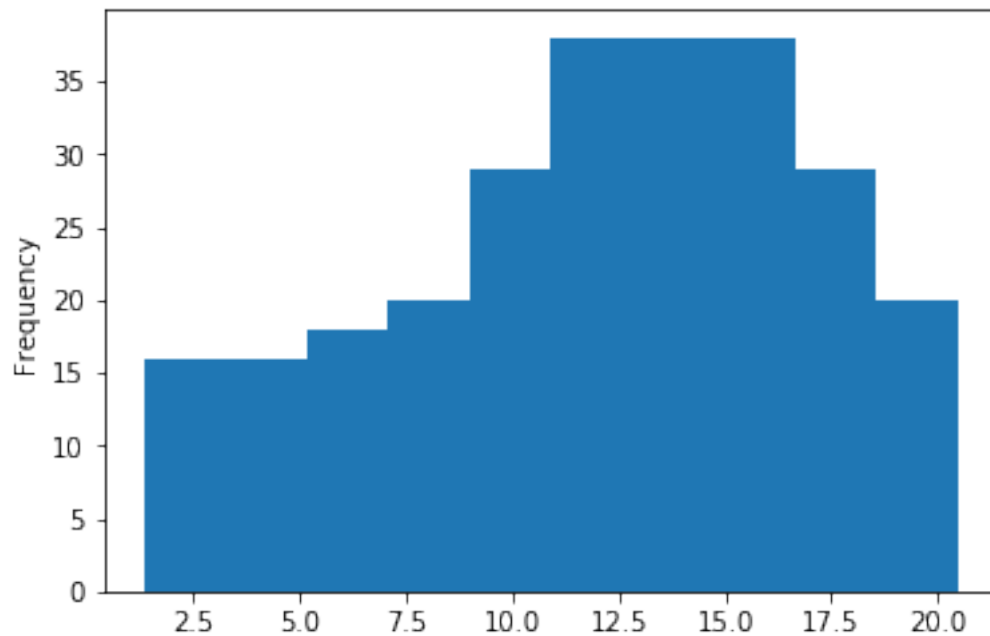
```
In [8]: dataset_teste.isnull().values.any()
```

```
Out[8]: False
```

2.0.4 Analisando os histogramas

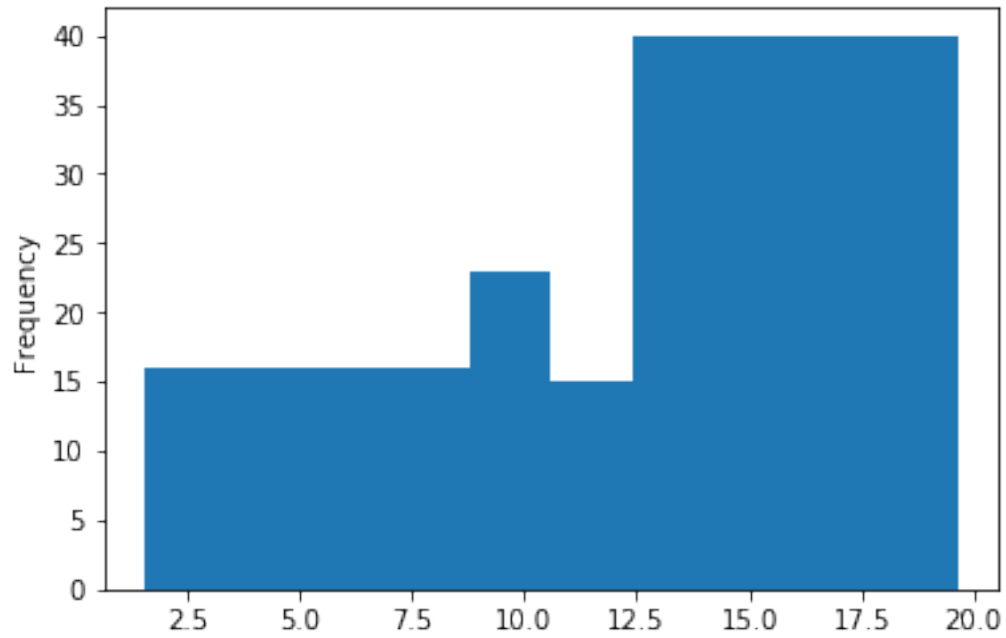
```
In [9]: dataset_treino['Coordenada X'].plot(kind='hist')
```

```
Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa290581748>
```



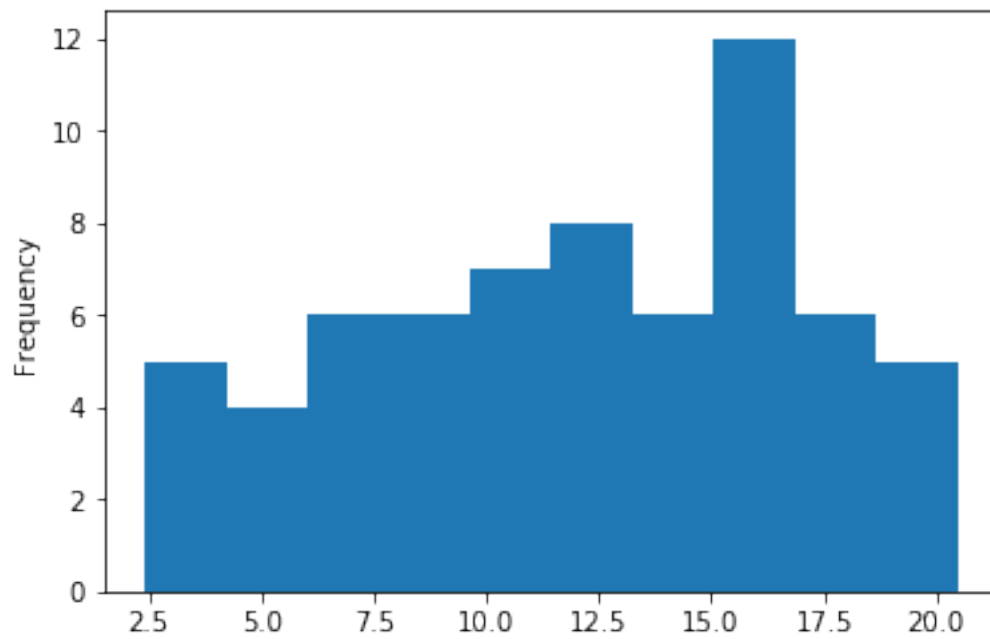
```
In [10]: dataset_treino['Coordenada Y'].plot(kind='hist')
```

```
Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa2905087f0>
```



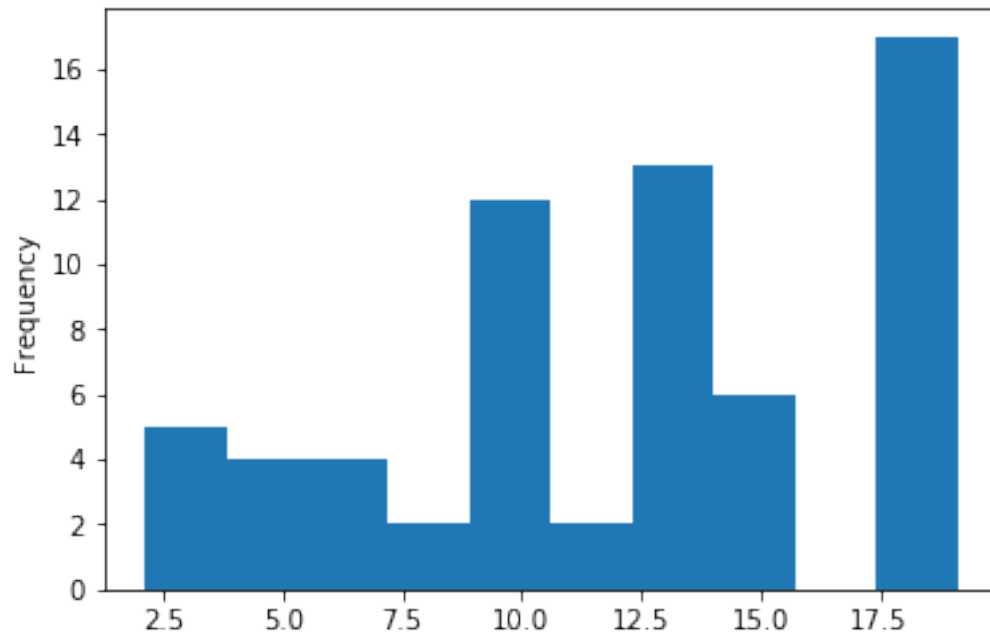
```
In [11]: dataset_teste['Coordenada X'].plot(kind='hist')
```

```
Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa290484390>
```



```
In [12]: dataset_teste['Coordenada Y'].plot(kind='hist')
```

```
Out[12]: <matplotlib.axes._subplots.AxesSubplot at 0x7fa290471978>
```



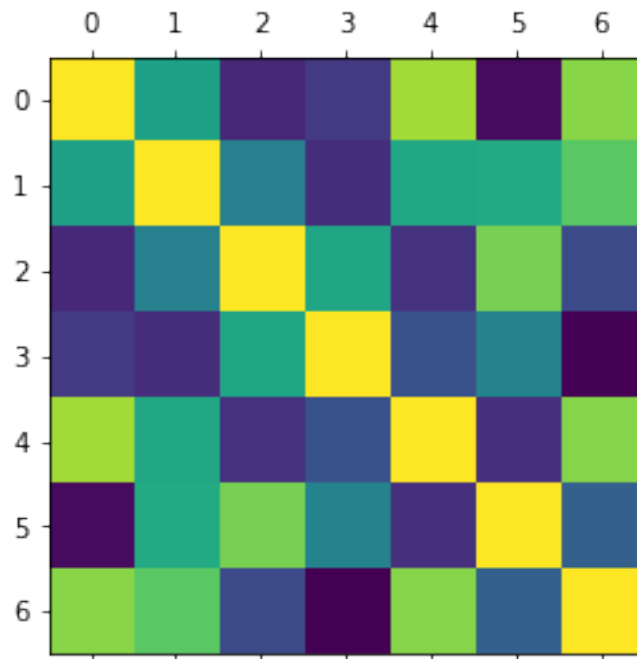
2.0.5 Preparando as variáveis para treino e teste dos algoritmos de aprendizado de máquina...

```
In [30]: X_train = X  
         X_test = Xt  
         y_train = y  
         y_test = yt
```

2.0.6 Plotando os gráficos de correlação dos datasets...

```
In [14]: plt.matshow(dataset_treino.corr())
```

```
Out[14]: <matplotlib.image.AxesImage at 0x7fa29039c1d0>
```



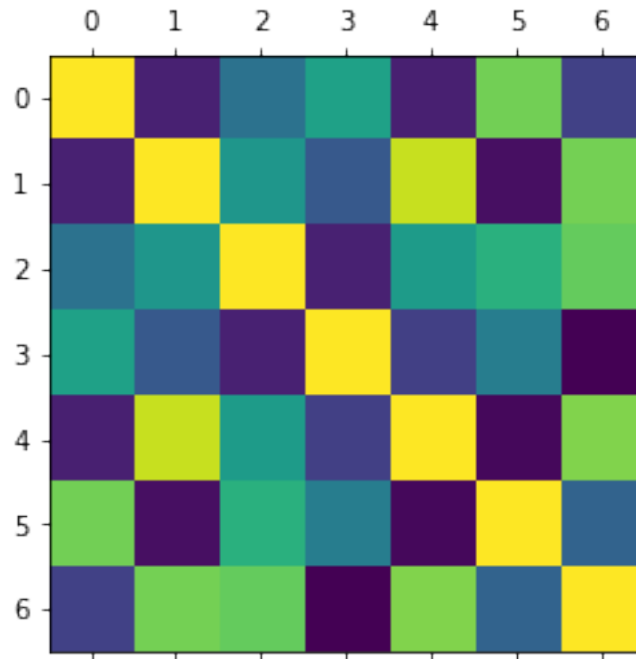
```
In [15]: print(dataset_treino.corr())
```

	AP1	AP2	AP3	AP4	AP5	Coordenada X \
AP1	1.000000	0.239991	-0.579181	-0.479896	0.749445	-0.732574
AP2	0.239991	1.000000	-0.009559	-0.552974	0.286467	0.307589
AP3	-0.579181	-0.009559	1.000000	0.262054	-0.519194	0.635752
AP4	-0.479896	-0.552974	0.262054	1.000000	-0.336926	0.011004
AP5	0.749445	0.286467	-0.519194	-0.336926	1.000000	-0.543906
Coordenada X	-0.732574	0.307589	0.635752	0.011004	-0.543906	1.000000
Coordenada Y	0.683116	0.544512	-0.382358	-0.783691	0.676932	-0.236476

	Coordenada Y
AP1	0.683116
AP2	0.544512
AP3	-0.382358
AP4	-0.783691
AP5	0.676932
Coordenada X	-0.236476
Coordenada Y	1.000000

```
In [16]: plt.matshow(dataset_teste.corr())
```

```
Out[16]: <matplotlib.image.AxesImage at 0x7fa2902fcda0>
```



```
In [17]: print(dataset_teste.corr())
```

	AP1	AP2	AP3	AP4	AP5	Coordenada X \
AP1	1.000000	-0.624181	-0.116768	0.236859	-0.632930	0.618729
AP2	-0.624181	1.000000	0.151045	-0.290863	0.849767	-0.720831
AP3	-0.116768	0.151045	1.000000	-0.625816	0.189058	0.347374
AP4	0.236859	-0.290863	-0.625816	1.000000	-0.453619	-0.030745
AP5	-0.632930	0.849767	0.189058	-0.453619	1.000000	-0.753619
Coordenada X	0.618729	-0.720831	0.347374	-0.030745	-0.753619	1.000000
Coordenada Y	-0.446523	0.626146	0.578855	-0.791972	0.657350	-0.220059

	Coordenada Y
AP1	-0.446523
AP2	0.626146
AP3	0.578855
AP4	-0.791972
AP5	0.657350
Coordenada X	-0.220059
Coordenada Y	1.000000

2.0.7 Alterando a escala dos dados para que nenhum deles seja menos relevante do que o outro na fase de treinamento do aprendizado de máquina.

```
In [18]: from sklearn.preprocessing import StandardScaler
sc_X = StandardScaler()
```



```
X_train = sc_X.fit_transform(X_train)
X_test = sc_X.transform(X_test)
```

```
/home/nath/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py:475: DataConversionWarning:
  warnings.warn(msg, DataConversionWarning)
```

3 2. Regressão Linear Múltipla

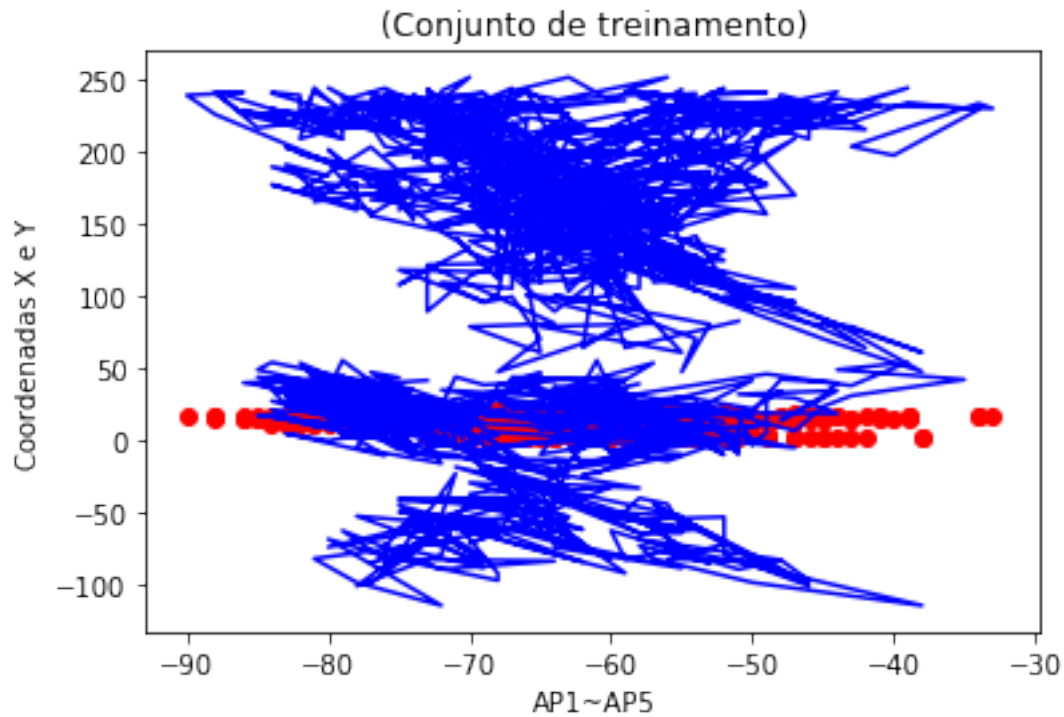
```
In [19]: # Realizando o fitting da regressão linear múltipla para a base de treinamento
from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)
```

```
Out[19]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=1, normalize=False)
```

```
In [20]: # Prevendo os resultados da base de testes
y_pred = regressor.predict(X_test)
```

```
In [87]: # Visualizando os resultados da base de treinamento
```

```
plt.scatter(X_train[:, 0:1], y_train[:, 0:1], color = 'red')
plt.scatter(X_train[:, 2:3], y_train[:, 0:1], color = 'red')
plt.scatter(X_train[:, [4]], y_train[:, 0:1], color = 'red')
plt.plot(X_train, regressor.predict(X_train), color = 'blue')
plt.title('(Conjunto de treinamento)')
plt.xlabel('AP1~AP5')
plt.ylabel('Coordenadas X e Y')
plt.show()
```



```
In [91]: # Visualizando os resultados da base de testes
```

```
plt.scatter(X_test[:, 0:1], y_test[:, 0:1], color = 'red')
plt.scatter(X_test[:, 2:3], y_test[:, 0:1], color = 'red')
plt.scatter(X_test[:, [4]], y_test[:, 0:1], color = 'red')
plt.plot(X_test, regressor.predict(X_test), color = 'blue')
plt.title('(Conjunto de testes)')
plt.xlabel('AP1~AP5')
plt.ylabel('Coordenadas X e Y')
plt.show()
```