

# Dataset\_Z\_Exploration\_Matlab

April 4, 2021

## 0.1 Dataset Prueba 1 - Tesis Javier-Uriel

### 0.1.1 Importamos algunas librerías que nos serán útiles más adelante

```
[1]: import os
import time
import random

import pandas as pd # for dataframe operations.
import numpy as np #for linear algebra operations.
import seaborn as sns # data visualization library
import matplotlib.pyplot as plt # for plotting

from scipy.fftpack import fft, fftfreq

from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
from statsmodels.tsa.stattools import pacf

pd.set_option('display.max_columns', None) #Para mostrar todas las columnas
random.seed(1)
```

### 0.1.2 Leemos el Dataset

```
[2]: #Dataset solo movimientos en Z
#rpm_list = ['RPM0', 'RPM1', 'RPM2', 'RPM3']
#states_list_org = ["vz", "az", "uvz",
#                  "p", "q",
#                  "wp", "wq",
#                  "ap", "aq"]
rpm_list = ['Motor1', 'Motor2', 'Motor3', 'Motor4']
states_list_org = ["Z", "Dz", "Z_r", "Ac_Dz",
                  "Yaw", "Roll",
                  "P", "Q",
                  "Acceleracion P", "Acceleracion Q"]
#states_list_org = ["vz", "az", "uvz"]
states_list_min = ["vz", "az", "uvz"]
dataset_name = "Dataset_Z_Matlab"
directory = "../logs/Datasets/"+dataset_name
```

```
ORDER = 3
dfs = []
states_list=states_list_org.copy()
```

### 0.1.3 Corregir la salida

El estado que entrega Pybullet de RPMs es la salida anterior, en este dataset se tomará RPMs como la salida actual. Si el primer elemento de RPMs es 0, es necesario hacer el shift

```
[3]: for filename in os.listdir(directory):
    if not filename.endswith(".csv"):
        continue
    df = pd.read_csv(os.path.join(directory, filename))
    #if any(df['z']<=1) or any(abs(df['vz'])>=10): #Eliminar si el dron se cae
    # print(filename)
    #else:
    if any(df[rpm_list].loc[0]==0): #Desplazar los estados de RPM si es necesario
        df[rpm_list] = df[rpm_list].shift(periods=-1)
        df = df.dropna()
        df.to_csv(os.path.join(directory, filename), index=False)

    a = []
    ## Desplazamos estados anteriores
    for n in range(1,ORDER+1):
        for column in states_list:
            df[column+str(n)] = df[column].shift(periods=n, fill_value=0)
            a.append(column+str(n))
    dfs.append(df)
states_list+=a

dataset = pd.concat(dfs)
dataset.describe()
```

```
[3]:
```

	Time	X	Y	Z	Yaw \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	4.281013e+01	-3.350768e-04	-3.359317e-04	-1.503027e+00	2.256389e-06
std	2.896393e+01	1.074610e-02	1.136344e-02	2.700165e-01	2.768997e-04
min	0.000000e+00	-3.573908e-01	-5.542104e-01	-3.318702e+00	-1.254526e-02
25%	1.806000e+01	-1.364899e-03	-1.247569e-03	-1.624232e+00	-1.617708e-05
50%	3.521000e+01	-3.312734e-04	-2.898734e-04	-1.499096e+00	1.685380e-07
75%	6.760500e+01	7.783263e-04	6.850931e-04	-1.367094e+00	1.726007e-05
max	1.000000e+02	3.491811e-01	5.142932e-01	-4.311047e-02	1.233279e-02

	Roll	Pitch	Dx	Dy	Dz \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	7.887180e-05	-8.071778e-05	3.052208e-06	2.332470e-06	-1.521121e-02
std	3.506121e-03	5.052975e-03	4.558698e-02	2.659362e-02	4.544270e-01

min	-1.588011e-01	-1.569786e-01	-1.510064e+00	-1.360795e+00	-4.524404e+00
25%	-2.072770e-04	-3.469328e-04	-8.009977e-04	-1.753193e-03	-9.322637e-03
50%	6.043164e-05	-7.090611e-05	-1.754226e-06	0.000000e+00	1.322787e-04
75%	3.423737e-04	1.998152e-04	7.909324e-04	1.652702e-03	8.660311e-03
max	1.636863e-01	1.538717e-01	1.486235e+00	1.353771e+00	3.034264e+00

	P	Q	R	Motor1	Motor2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.573519e-06	1.412977e-06	-3.197492e-07	2.365206e+02	-2.365207e+02
std	9.057596e-03	3.927625e-02	2.225740e-03	3.395921e+01	3.395556e+01
min	-5.037703e-01	-3.575321e+00	-1.820346e-01	1.000000e+01	-4.613083e+02
25%	-1.220251e-03	-1.284665e-03	-9.134351e-04	2.355287e+02	-2.375263e+02
50%	4.287249e-06	-4.722129e-06	-4.495727e-06	2.365282e+02	-2.365222e+02
75%	1.240739e-03	1.283613e-03	9.029967e-04	2.374787e+02	-2.354865e+02
max	4.727377e-01	3.632708e+00	1.794934e-01	4.609845e+02	-1.000000e+01

	Motor3	Motor4	X_r	Y_r	Z_r \
count	1.295895e+06	1.295895e+06	1295895.0	1295895.0	1.295895e+06
mean	2.365206e+02	-2.365208e+02	0.0	0.0	-1.482117e+00
std	3.394866e+01	3.396455e+01	0.0	0.0	2.440522e-01
min	1.000000e+01	-4.609848e+02	0.0	0.0	-2.774623e+00
25%	2.355364e+02	-2.375314e+02	0.0	0.0	-1.620000e+00
50%	2.365288e+02	-2.365235e+02	0.0	0.0	-1.500000e+00
75%	2.374726e+02	-2.354862e+02	0.0	0.0	-1.351574e+00
max	4.611932e+02	-1.000000e+01	0.0	0.0	0.000000e+00

	Yaw_r	Pitch_r	Roll_r	Dx_r	Dy_r	Dz_r \
count	1295895.0	1295895.0	1295895.0	1295895.0	1295895.0	1.295895e+06
mean	0.0	0.0	0.0	0.0	0.0	-2.193838e-02
std	0.0	0.0	0.0	0.0	0.0	8.223455e+00
min	0.0	0.0	0.0	0.0	0.0	-3.460435e+02
25%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
50%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
75%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
max	0.0	0.0	0.0	0.0	0.0	3.807929e+02

	P_r	Q_r	R_r	Flag_Pitch_Roll	Ac_Dx \
count	1295895.0	1295895.0	1295895.0	1295895.0	1.295895e+06
mean	0.0	0.0	0.0	1.0	9.007334e-02
std	0.0	0.0	0.0	0.0	2.089574e-01
min	0.0	0.0	0.0	1.0	-8.013137e-01
25%	0.0	0.0	0.0	1.0	-5.126491e-02
50%	0.0	0.0	0.0	1.0	9.049701e-02
75%	0.0	0.0	0.0	1.0	2.315978e-01
max	0.0	0.0	0.0	1.0	8.437319e-01

Ac_Dy	Ac_Dz	Gyro P	Gyro Q	Gyro R \
-------	-------	--------	--------	----------

count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-5.992936e-02	-9.419827e+00	-9.498428e-03	-7.498578e-03	1.499741e-03
std	1.930278e-01	1.427087e+00	9.045007e-03	3.952919e-02	2.383690e-03
min	-8.834900e-01	-1.963813e+01	-5.125701e-01	-3.605846e+00	-1.798834e-01
25%	-1.905823e-01	-9.646096e+00	-1.071855e-02	-8.792938e-03	4.273835e-04
50%	-5.959882e-02	-9.423019e+00	-9.495718e-03	-7.504752e-03	1.502034e-03
75%	7.084187e-02	-9.203916e+00	-8.260985e-03	-6.208120e-03	2.568620e-03
max	6.365708e-01	8.609541e-01	4.625807e-01	3.648603e+00	1.830248e-01

	Sonar Altitud	Pressure Altitud	Bat_V	Bat_Percentage	\
count	1.295895e+06	1.295895e+06	1295895.0	1295895.0	
mean	1.504551e+00	1.012529e+05	3.5	70.0	
std	2.614817e-01	3.211676e+00	0.0	0.0	
min	4.400000e-01	1.012314e+05	3.5	70.0	
25%	1.367890e+00	1.012514e+05	3.5	70.0	
50%	1.498847e+00	1.012529e+05	3.5	70.0	
75%	1.624728e+00	1.012545e+05	3.5	70.0	
max	3.290363e+00	1.012704e+05	3.5	70.0	

	Acceleracion X	Acceleracion Y	Acceleracion Z	Acceleracion P	\
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	
mean	5.755633e-06	-1.250541e-04	-1.736796e-04	6.004077e-05	
std	5.950564e-01	8.118001e-02	1.435704e+00	4.211311e-01	
min	-8.899236e+01	-4.018375e+00	-1.017578e+01	-8.605778e+00	
25%	-8.635339e-03	-2.117448e-02	-1.253270e-01	-2.800144e-01	
50%	0.000000e+00	0.000000e+00	-1.526624e-03	-3.745025e-03	
75%	8.559134e-03	2.102111e-02	1.203972e-01	2.815473e-01	
max	9.037159e+01	4.432607e+00	1.060796e+01	7.373816e+00	

	Acceleracion Q	Acceleracion R	Z1	Dz1	\
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	
mean	2.189152e-05	-8.116385e-06	-1.502918e+00	-1.521034e-02	
std	1.670874e+00	2.951463e-01	2.703173e-01	4.544223e-01	
min	-1.592394e+02	-8.434417e+00	-3.318702e+00	-4.524404e+00	
25%	-3.060012e-01	-1.960427e-01	-1.624222e+00	-9.322412e-03	
50%	-3.043190e-03	-7.657334e-04	-1.499095e+00	1.304944e-04	
75%	3.080426e-01	1.959933e-01	-1.367004e+00	8.658752e-03	
max	1.689137e+02	7.723484e+00	0.000000e+00	3.034264e+00	

	Z_r1	Ac_Dz1	Yaw1	Roll1	P1	\
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	
mean	-1.482007e+00	-9.419130e+00	2.257514e-06	7.887457e-05	1.273315e-06	
std	2.443724e-01	1.429272e+00	2.768997e-04	3.505640e-03	9.056365e-03	
min	-2.774623e+00	-1.963813e+01	-1.254526e-02	-1.588011e-01	-5.037703e-01	
25%	-1.620000e+00	-9.646093e+00	-1.617320e-05	-2.072770e-04	-1.220251e-03	
50%	-1.500000e+00	-9.423018e+00	1.684218e-07	6.035169e-05	4.136165e-06	
75%	-1.351574e+00	-9.203855e+00	1.726007e-05	3.423065e-04	1.240343e-03	

max	0.000000e+00	8.609541e-01	1.233279e-02	1.636863e-01	4.727377e-01
-----	--------------	--------------	--------------	--------------	--------------

	Q1	Acceleracion P1	Acceleracion Q1	Z2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.303519e-06	1.038755e-05	-3.086813e-05	-1.502809e+00
std	3.927625e-02	4.210900e-01	1.670862e+00	2.706177e-01
min	-3.575321e+00	-8.605778e+00	-1.592394e+02	-3.318702e+00
25%	-1.284665e-03	-2.800144e-01	-3.060012e-01	-1.624207e+00
50%	-4.722129e-06	-3.745025e-03	-3.043190e-03	-1.499094e+00
75%	1.283548e-03	2.814385e-01	3.079379e-01	-1.366925e+00
max	3.632708e+00	7.373816e+00	1.689137e+02	0.000000e+00

	Dz2	Z_r2	Ac_Dz2	Yaw2	Roll2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-1.520958e-02	-1.481896e+00	-9.418387e+00	2.258437e-06	7.887883e-05
std	4.544177e-01	2.446916e-01	1.431431e+00	2.768996e-04	3.505148e-03
min	-4.524404e+00	-2.774623e+00	-1.963813e+01	-1.254526e-02	-1.588011e-01
25%	-9.322241e-03	-1.620000e+00	-9.646035e+00	-1.617313e-05	-2.072770e-04
50%	1.281137e-04	-1.500000e+00	-9.422956e+00	1.683297e-07	6.028301e-05
75%	8.657166e-03	-1.351574e+00	-9.203737e+00	1.726007e-05	3.422434e-04
max	3.034264e+00	0.000000e+00	8.609541e-01	1.233279e-02	1.636863e-01

	P2	Q2	Acceleracion P2	Acceleracion Q2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.221378e-06	1.457860e-06	2.812219e-05	-7.125476e-06
std	9.055216e-03	3.927624e-02	4.210710e-01	1.670856e+00
min	-5.037703e-01	-3.575321e+00	-8.605778e+00	-1.592394e+02
25%	-1.219919e-03	-1.284440e-03	-2.799338e-01	-3.059567e-01
50%	4.130569e-06	-4.424624e-06	-3.700447e-03	-2.996484e-03
75%	1.240317e-03	1.283548e-03	2.814385e-01	3.079379e-01
max	4.727377e-01	3.632708e+00	7.373816e+00	1.689137e+02

	Z3	Dz3	Z_r3	Ac_Dz3	Yaw3 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-1.502699e+00	-1.520887e-02	-1.481784e+00	-9.417694e+00	2.258529e-06
std	2.709177e-01	4.544130e-01	2.450097e-01	1.433600e+00	2.768996e-04
min	-3.318702e+00	-4.524404e+00	-2.774623e+00	-1.963813e+01	-1.254526e-02
25%	-1.624196e+00	-9.322085e-03	-1.620000e+00	-9.646004e+00	-1.617308e-05
50%	-1.499093e+00	1.261767e-04	-1.500000e+00	-9.422915e+00	1.660678e-07
75%	-1.366811e+00	8.655519e-03	-1.351574e+00	-9.203696e+00	1.726007e-05
max	0.000000e+00	3.034264e+00	0.000000e+00	8.609541e-01	1.233279e-02

	Roll3	P3	Q3	Acceleracion P3 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	7.888335e-05	1.080767e-06	1.493488e-06	2.985777e-06
std	3.504645e-03	9.054026e-03	3.927624e-02	4.210606e-01
min	-1.588011e-01	-5.037703e-01	-3.575321e+00	-8.605778e+00

25%	-2.072770e-04	-1.219723e-03	-1.284153e-03	-2.799338e-01
50%	6.020387e-05	3.981350e-06	-4.307566e-06	-3.700447e-03
75%	3.421859e-04	1.240157e-03	1.283347e-03	2.813485e-01
max	1.636863e-01	4.727377e-01	3.632708e+00	7.373816e+00

	Acceleracion Q3
count	1.295895e+06
mean	-3.594663e-05
std	1.670853e+00
min	-1.592394e+02
25%	-3.059567e-01
50%	-2.996484e-03
75%	3.079038e-01
max	1.689137e+02

#### 0.1.4 Estados repetidos

En este caso se eliminan estados repetidos y estados que se encuentren en estado transitorio mientras el dron despega o se estabiliza antes de introducir la señal de control.

```
[4]: shape_b4 = dataset.drop(["Time"], axis=1).shape
shape_drop= dataset.drop(["Time"], axis=1).drop_duplicates().shape
print(f'shape (b4 drop) = {shape_b4}')
print(f'shape = {shape_drop}')
print(f'len (b4 drop) - len = {shape_b4[0]-shape_drop[0]}')
```

```
shape (b4 drop) = (1295895, 75)
shape = (1136656, 75)
len (b4 drop) - len = 159239
```

```
[5]: states = dataset.drop(["Time"], axis=1).drop_duplicates()[states_list]
print(f'columns = {states.columns}')
print(f'shape = {states.shape}')
states.head()
```

```
columns = Index(['Z', 'Dz', 'Z_r', 'Ac_Dz', 'Yaw', 'Roll', 'P', 'Q',
'Acceleracion P',
'Acceleracion Q', 'Z1', 'Dz1', 'Z_r1', 'Ac_Dz1', 'Yaw1', 'Roll1', 'P1',
'Q1', 'Acceleracion P1', 'Acceleracion Q1', 'Z2', 'Dz2', 'Z_r2',
'Ac_Dz2', 'Yaw2', 'Roll2', 'P2', 'Q2', 'Acceleracion P2',
'Acceleracion Q2', 'Z3', 'Dz3', 'Z_r3', 'Ac_Dz3', 'Yaw3', 'Roll3', 'P3',
'Q3', 'Acceleracion P3', 'Acceleracion Q3'],
dtype='object')
shape = (1136656, 40)
```

```
[5]:      Z      Dz  Z_r      Ac_Dz      Yaw      Roll      P  \
0 -0.046000  0.000000  0.0  -9.492839 -5.859605e-07 -1.965233e-06 -0.000393
1 -0.046000  0.047745  0.0 -13.635429 -5.307764e-07  2.704250e-06  0.000934
```

2	-0.045761	0.087995	0.0	-14.080304	-7.171090e-08	-4.059651e-06	-0.001353
3	-0.045321	0.108800	0.0	-13.597730	-1.382231e-06	9.645310e-07	0.001005
4	-0.044777	0.105958	0.0	-13.629545	5.927828e-07	6.776180e-06	0.001162

	Q	Acceleracion P	Acceleracion Q	Z1	Dz1	Z_r1	\
0	-0.000431	-0.078609	-0.086239	0.000000	0.000000	0.0	
1	0.001025	0.265389	0.291147	-0.046000	0.000000	0.0	
2	-0.001527	-0.457335	-0.510345	-0.046000	0.047745	0.0	
3	0.001162	0.471523	0.537868	-0.045761	0.087995	0.0	
4	0.001192	0.031499	0.006026	-0.045321	0.108800	0.0	

	Ac_Dz1	Yaw1	Roll1	P1	Q1	Acceleracion P1	\
0	0.000000	0.000000e+00	0.000000e+00	0.000000	0.000000	0.000000	
1	-9.492839	-5.859605e-07	-1.965233e-06	-0.000393	-0.000431	-0.078609	
2	-13.635429	-5.307764e-07	2.704250e-06	0.000934	0.001025	0.265389	
3	-14.080304	-7.171090e-08	-4.059651e-06	-0.001353	-0.001527	-0.457335	
4	-13.597730	-1.382231e-06	9.645310e-07	0.001005	0.001162	0.471523	

	Acceleracion Q1	Z2	Dz2	Z_r2	Ac_Dz2	Yaw2	\
0	0.000000	0.000000	0.000000	0.0	0.000000	0.000000e+00	
1	-0.086239	0.000000	0.000000	0.0	0.000000	0.000000e+00	
2	0.291147	-0.046000	0.000000	0.0	-9.492839	-5.859605e-07	
3	-0.510345	-0.046000	0.047745	0.0	-13.635429	-5.307764e-07	
4	0.537868	-0.045761	0.087995	0.0	-14.080304	-7.171090e-08	

	Roll2	P2	Q2	Acceleracion P2	Acceleracion Q2	Z3	\
0	0.000000	0.000000	0.000000	0.000000	0.000000	0.000	
1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000	
2	-0.000002	-0.000393	-0.000431	-0.078609	-0.086239	0.000	
3	0.000003	0.000934	0.001025	0.265389	0.291147	-0.046	
4	-0.000004	-0.001353	-0.001527	-0.457335	-0.510345	-0.046	

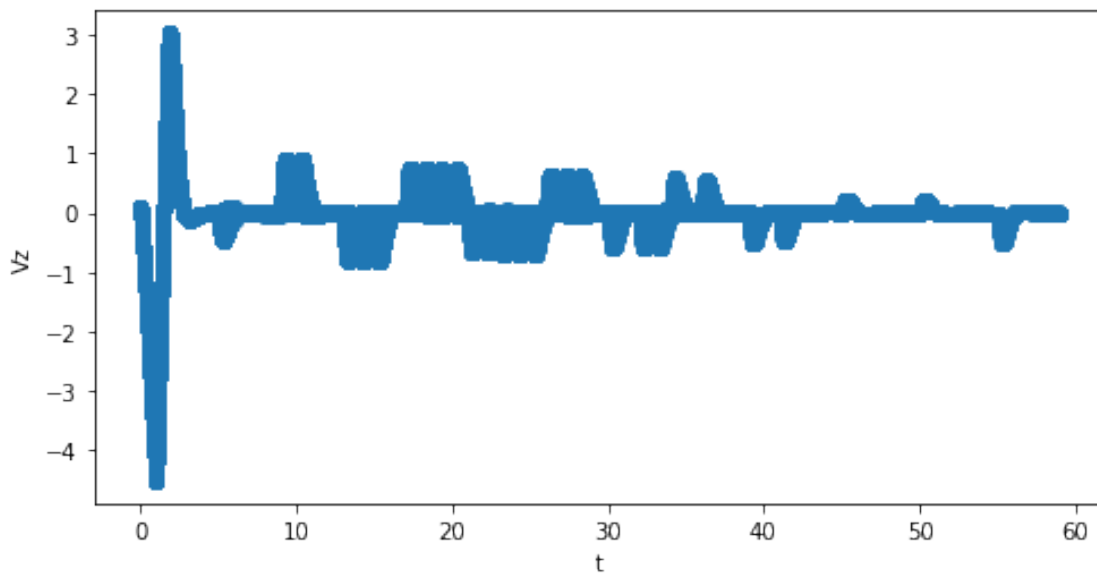
	Dz3	Z_r3	Ac_Dz3	Yaw3	Roll3	P3	Q3	\
0	0.000000	0.0	0.000000	0.000000e+00	0.000000	0.000000	0.000000	
1	0.000000	0.0	0.000000	0.000000e+00	0.000000	0.000000	0.000000	
2	0.000000	0.0	0.000000	0.000000e+00	0.000000	0.000000	0.000000	
3	0.000000	0.0	-9.492839	-5.859605e-07	-0.000002	-0.000393	-0.000431	
4	0.047745	0.0	-13.635429	-5.307764e-07	0.000003	0.000934	0.001025	

	Acceleracion P3	Acceleracion Q3
0	0.000000	0.000000
1	0.000000	0.000000
2	0.000000	0.000000
3	-0.078609	-0.086239
4	0.265389	0.291147

```
[6]: states_duplicates = dataset[dataset.duplicated(keep='last')]
states_duplicates = states_duplicates.dropna()
```

```
[7]: fig = plt.figure(figsize=(8, 4))
t = states_duplicates['Time']
y = states_duplicates['Dz']
#y_ref = states_duplicates['uvz']
plt.scatter(t, y)
#plt.scatter(t, y_ref)
plt.ylabel('Vz')
plt.xlabel('t')
```

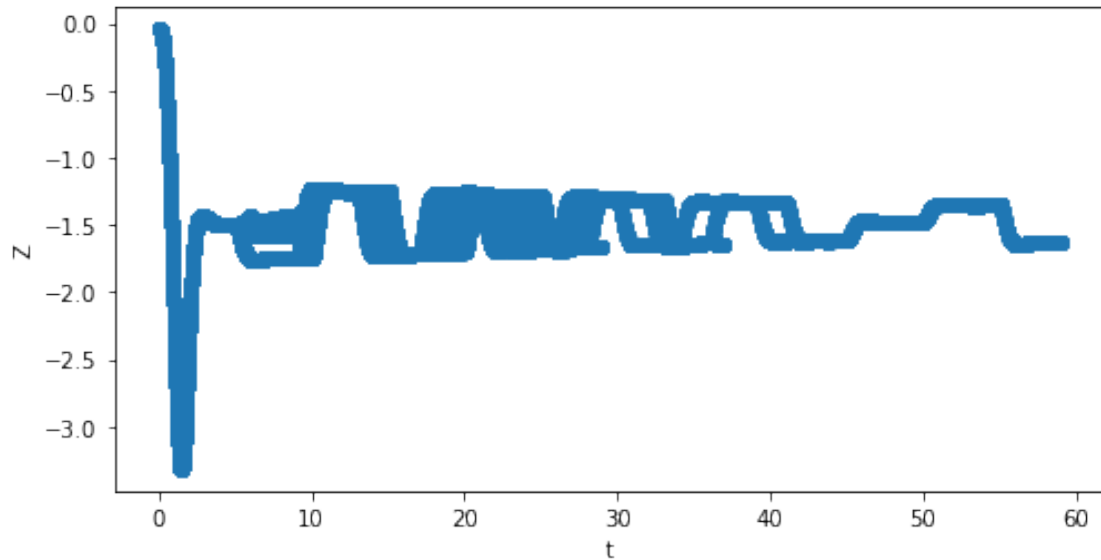
```
[7]: Text(0.5, 0, 't')
```



```
[8]: fig = plt.figure(figsize=(8, 4))
t = states_duplicates['Time']
y = states_duplicates['Z']
#y_ref = states_duplicates['uvz']
plt.scatter(t, y)
#plt.scatter(t, y_ref)
plt.ylabel('Z')
plt.xlabel('t')
```

```
[8]: Text(0.5, 0, 't')
```





### 0.1.5 Se grafican los datos

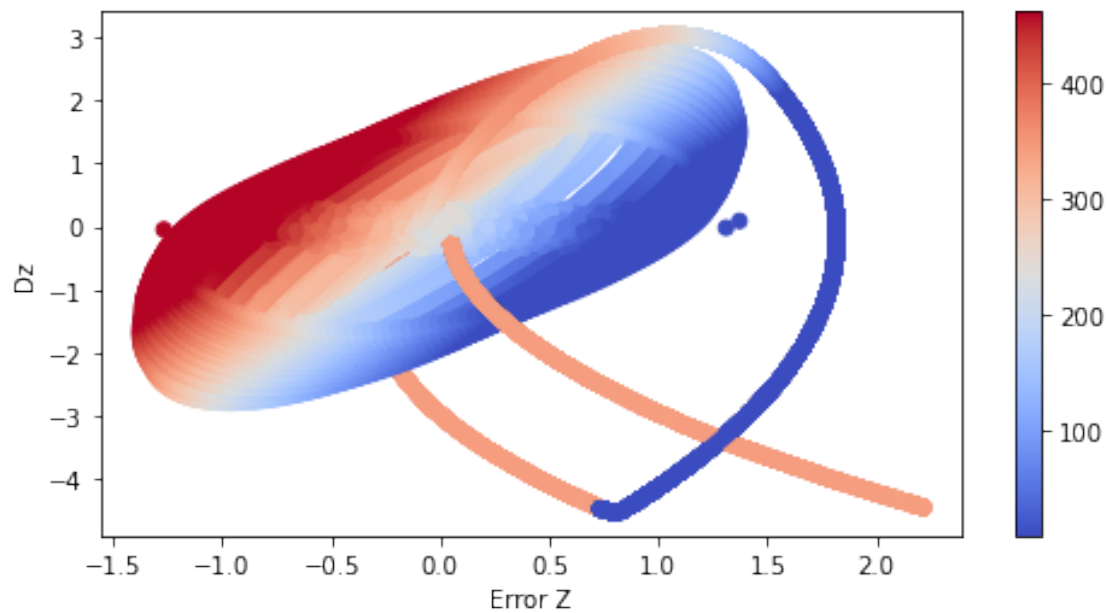
Se grafica un histograma de cada una de las propiedades los datos analizados individualmente por columnas, en el cual se observa que todos tienen distribuciones altamente apuntadas (curosis) y en algunos casos bimodales, pero de cualquier manera, no son uniformes

```
[9]: n_bins = 50
     #_ = dataset.hist(bins=n_bins, figsize=(30,30))
```

### 0.1.6 Análisis de estados

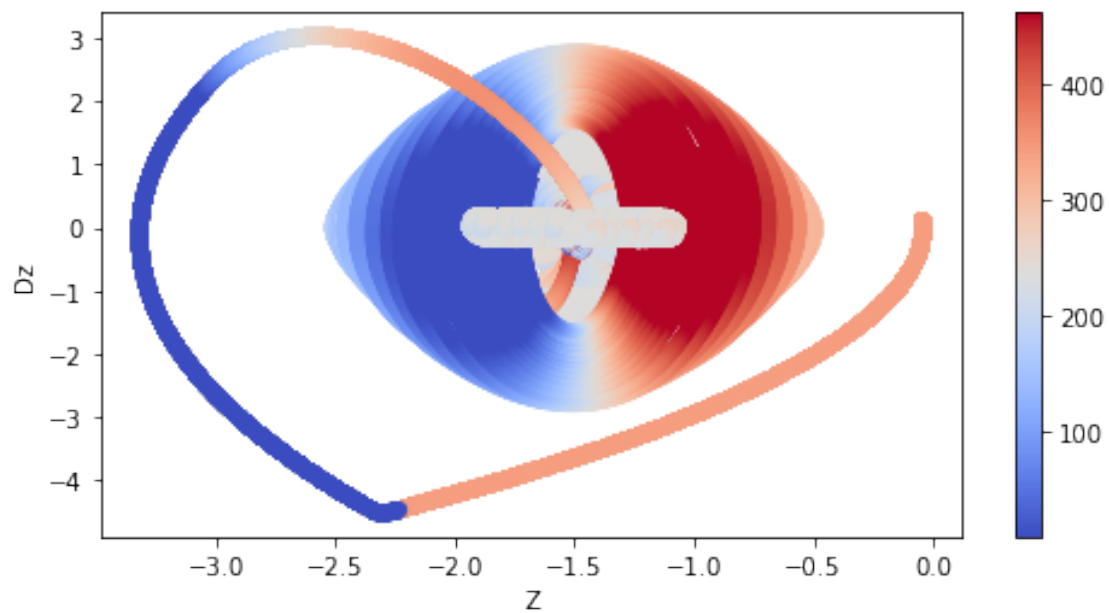
```
[10]: fig = plt.figure(figsize=(8, 4))
      x = dataset['Z_r']-dataset['Z']
      y = dataset['Dz']
      c = dataset['Motor1']
      plt.scatter(x, y, c=c, cmap='coolwarm')
      plt.colorbar()
      plt.ylabel('Dz')
      plt.xlabel('Error Z')
```

```
[10]: Text(0.5, 0, 'Error Z')
```



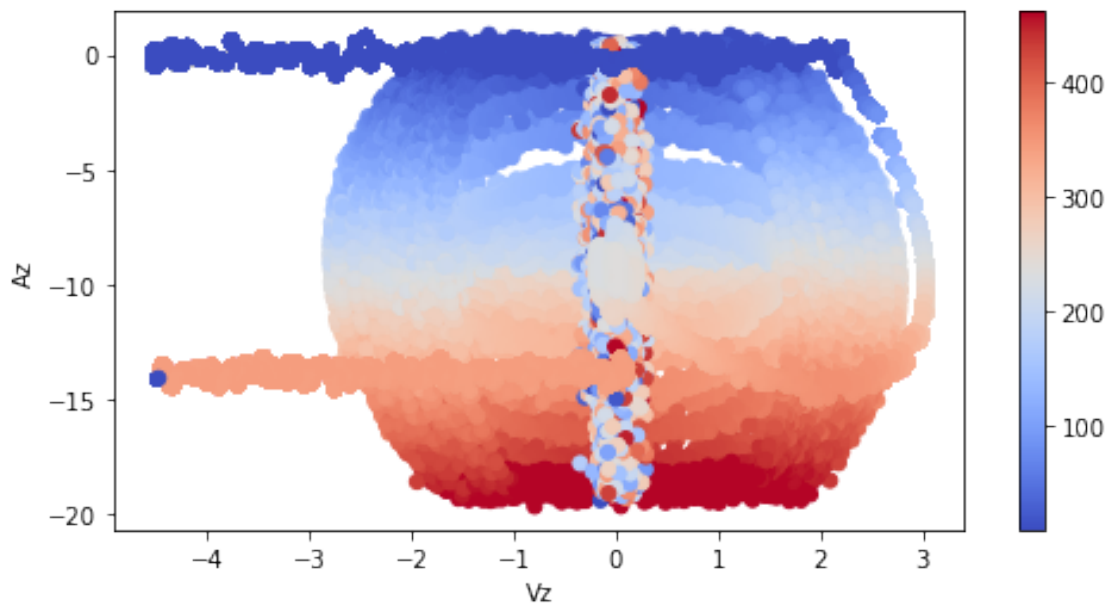
```
[11]: fig = plt.figure(figsize=(8, 4))
      x = dataset['Z']
      plt.scatter(x, y, c=c, cmap='coolwarm')
      plt.colorbar()
      plt.ylabel('Dz')
      plt.xlabel('Z')
```

```
[11]: Text(0.5, 0, 'Z')
```



```
[12]: fig = plt.figure(figsize=(8, 4))
x = dataset['Dz']
y = dataset['Ac_Dz']
plt.scatter(x, y, c=c, cmap='coolwarm')
plt.colorbar()
plt.ylabel('Az')
plt.xlabel('Vz')
```

```
[12]: Text(0.5, 0, 'Vz')
```



### 0.1.7 Análisis de Fourier

#### Gráfica de algunas señales

```
[13]: def plot_fourier(df, states=['Z', 'Dz', 'Z_r']):
    dt = df['Time'][1]-df['Time'][0]
    n = len(df['Time'])
    Y = fft(df[states[0]].to_numpy()) / n # Transformada normalizada
    Y_ref = fft(df[states[1]].to_numpy()) / n
    frq = fftfreq(n, dt)
    fig = plt.figure(figsize=(14, 10))
    ax1 = fig.add_subplot(221)
    ax1.plot(df['Time'], df[states[0]], df['Time'], df[states[2]])
    ax1.set_xlabel('Tiempo (s)')
    ax1.set_ylabel('$y(t)$')
```

```

ax1.set_title(f'Señal en el tiempo ({states[0]} y {states[2]})')
ax2 = fig.add_subplot(223)
ax2.set_title(f'Señal en frecuencia ({states[0]} y {states[2]})')
ax2.vlines(frq[0:int(n/10)], 0, abs(Y[0:int(n/10)]))
ax2.vlines(frq[0:int(n/10)], 0, abs(Y_ref[0:int(n/10)]), color='orange')
plt.xlabel('Frecuencia (Hz)')
plt.ylabel('Abs($Y$)')

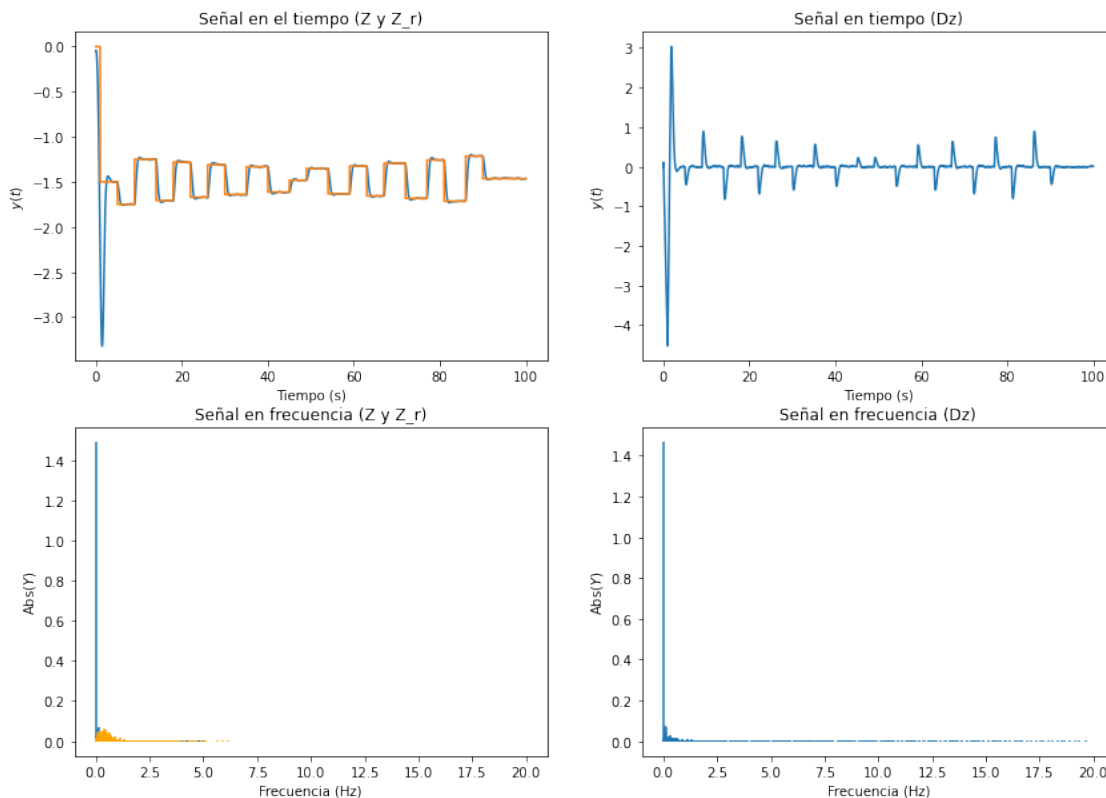
Y = fft(df[states[2]].to_numpy()) / n # Transformada normalizada
ax1 = fig.add_subplot(222)
ax1.plot(df['Time'], df[states[1]])
ax1.set_xlabel('Tiempo (s)')
ax1.set_ylabel('$y(t)$')
ax1.set_title(f'Señal en tiempo ({states[1]})')
ax2 = fig.add_subplot(224)
ax2.set_title(f'Señal en frecuencia ({states[1]})')
ax2.vlines(frq[0:int(n/10)], 0, abs(Y[0:int(n/10)]))
plt.xlabel('Frecuencia (Hz)')
plt.ylabel('Abs($Y$)')

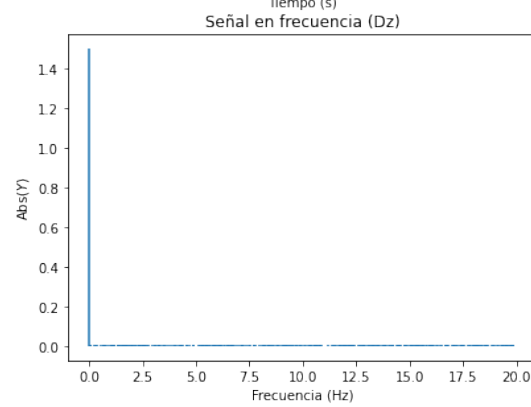
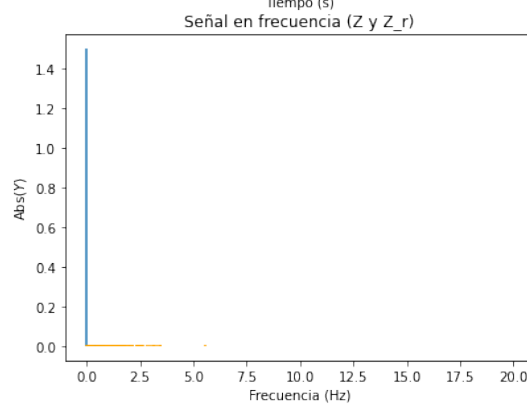
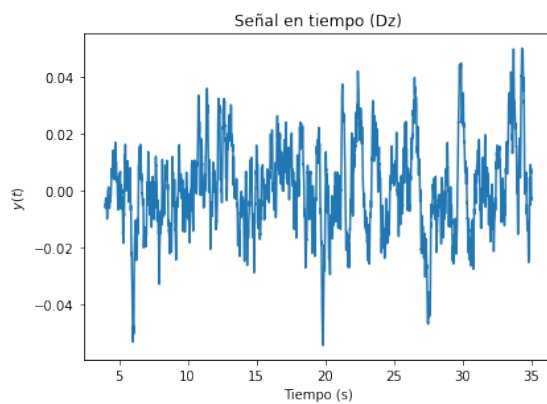
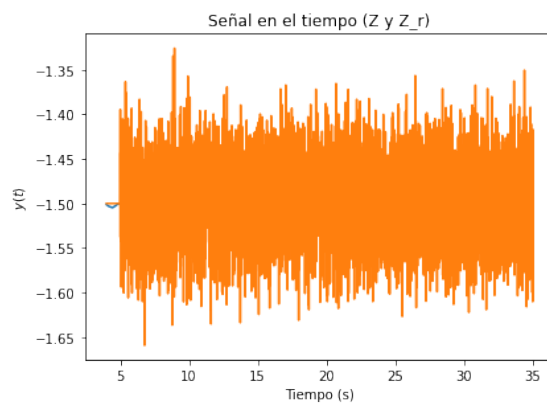
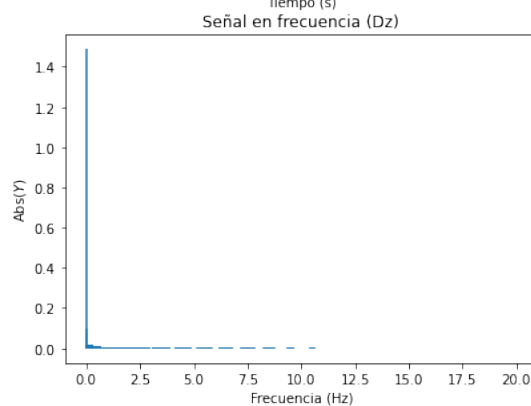
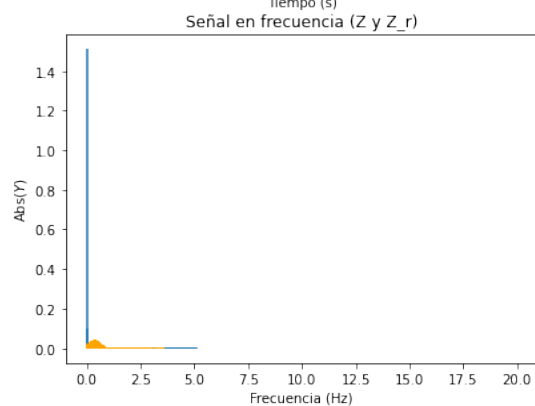
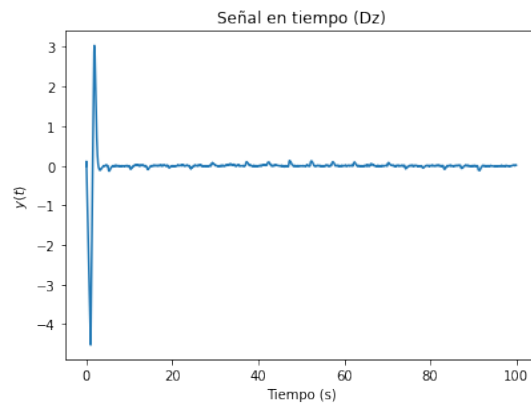
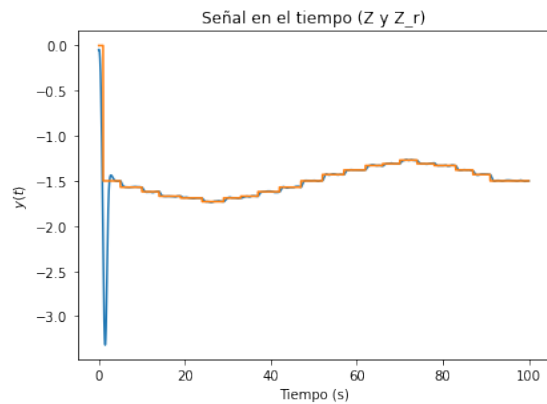
```

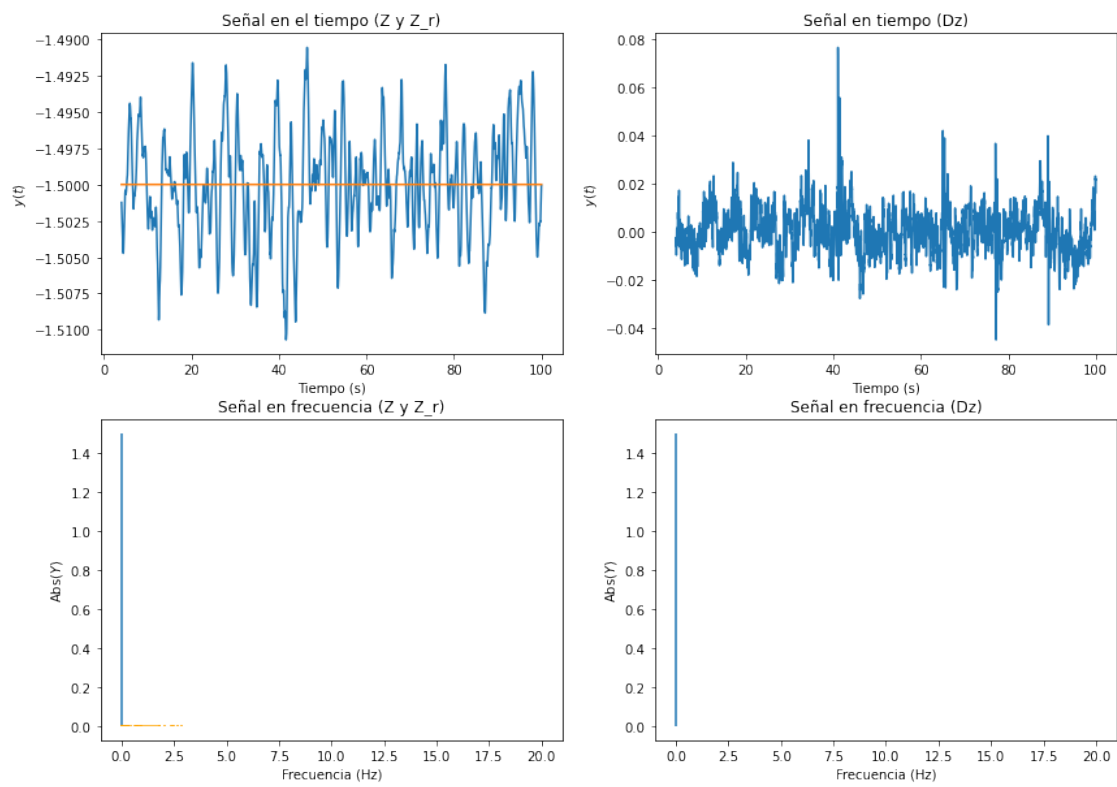
```

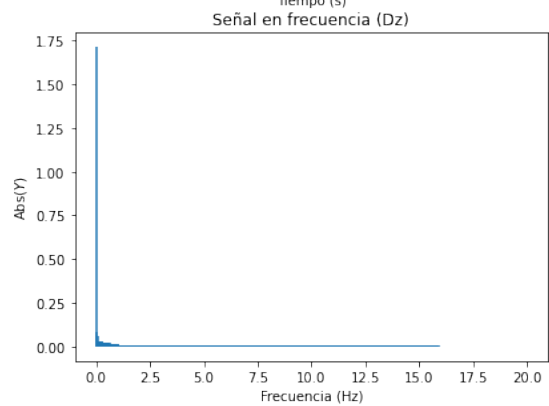
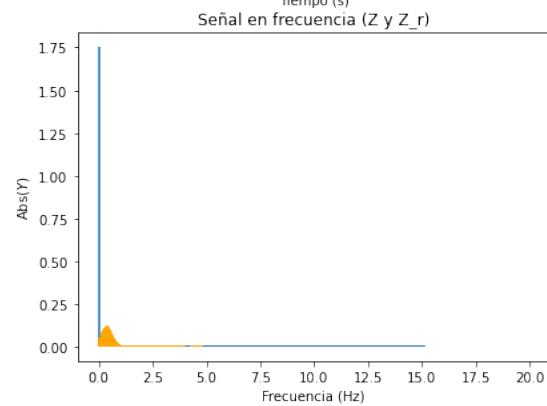
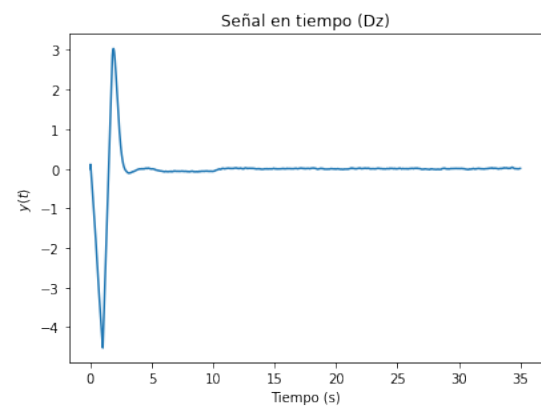
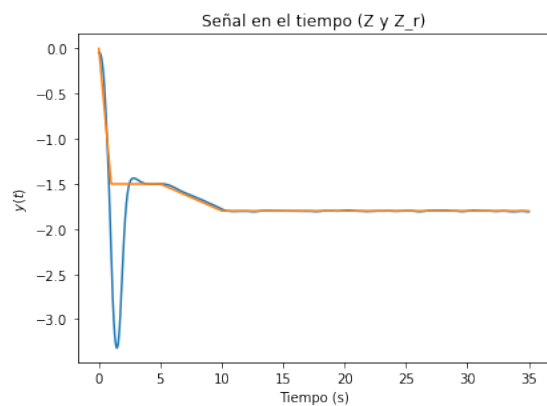
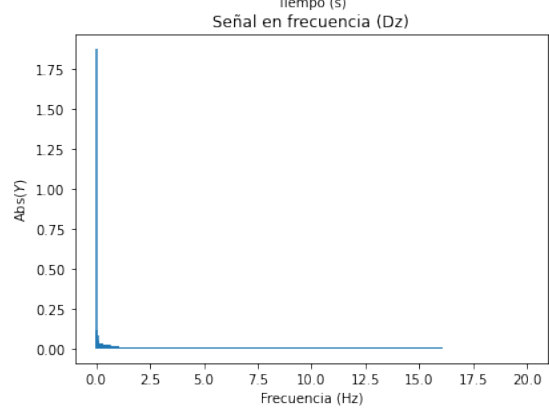
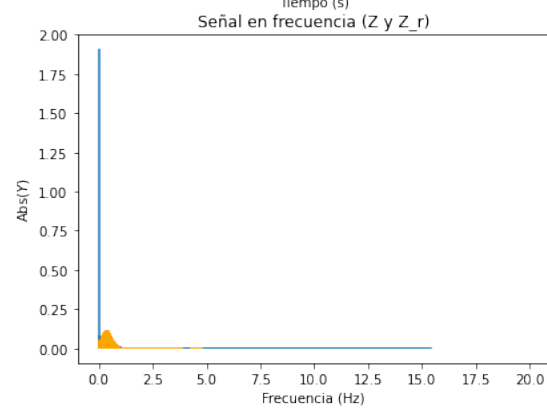
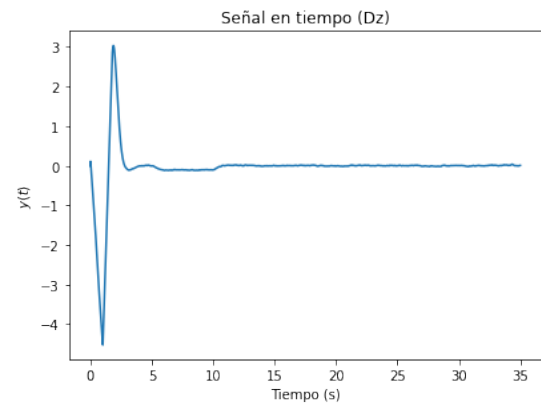
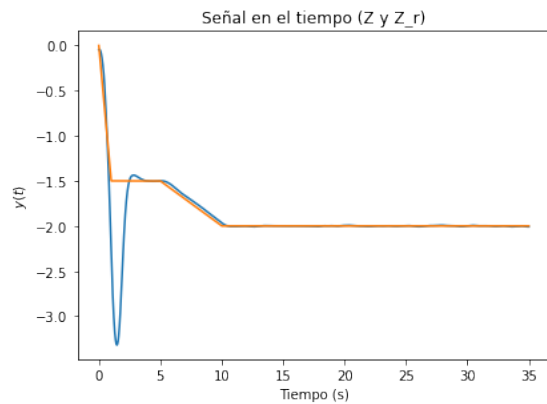
[14]: for df in random.choices(dfs, k = 8):
      plot_fourier(df)

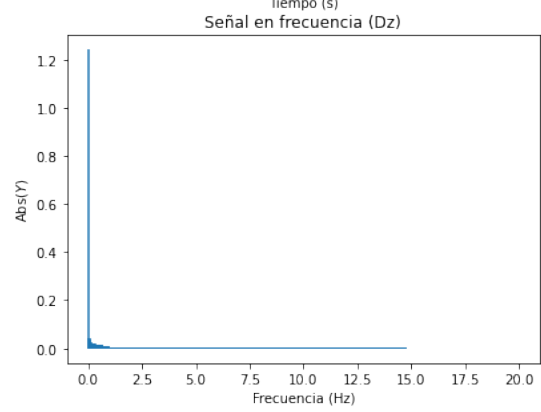
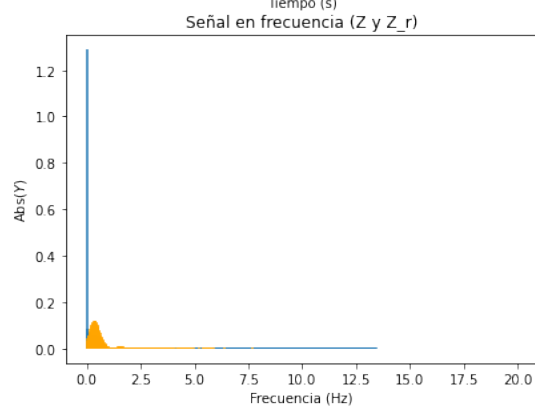
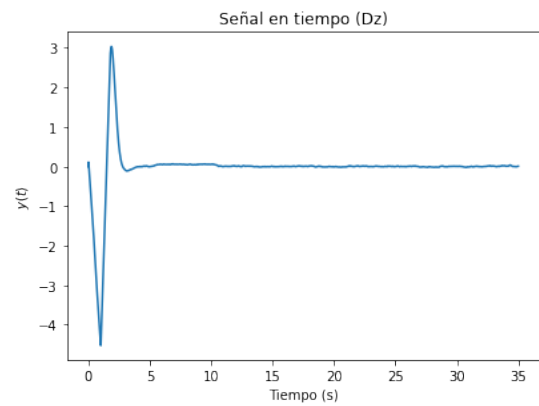
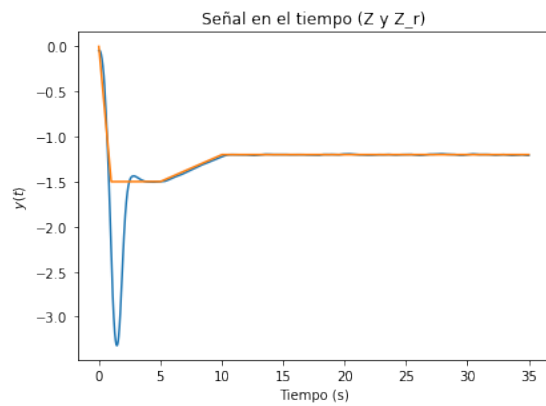
```



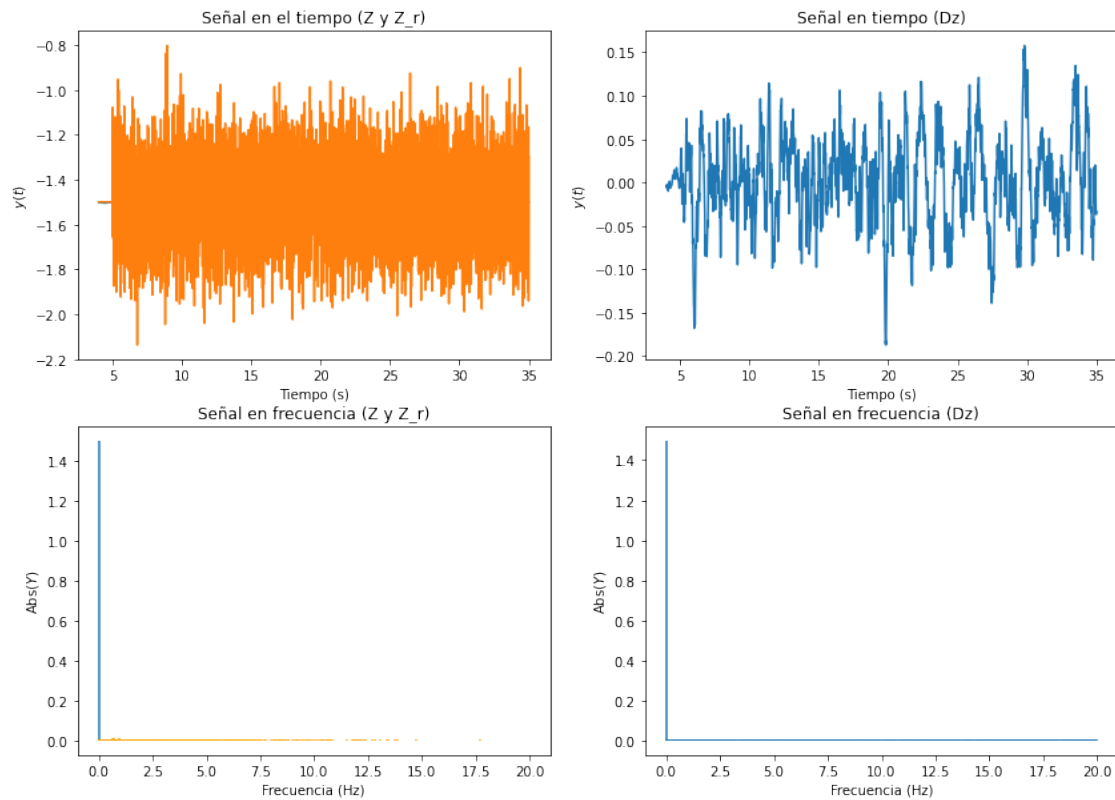




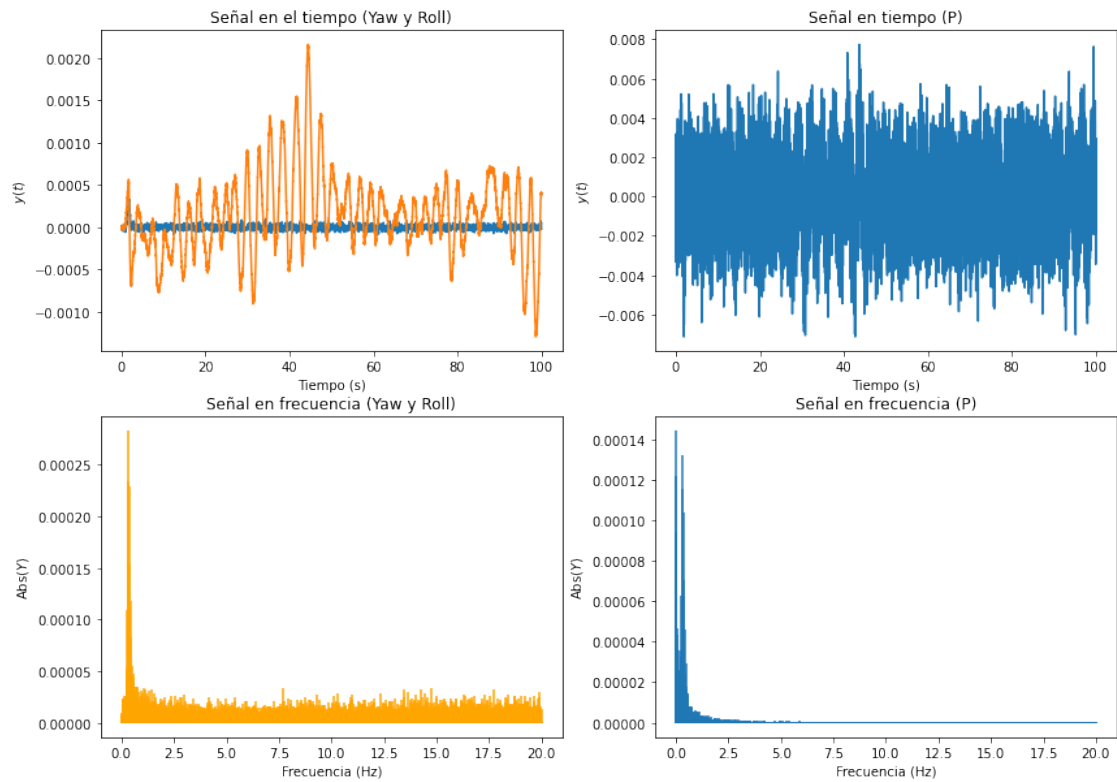
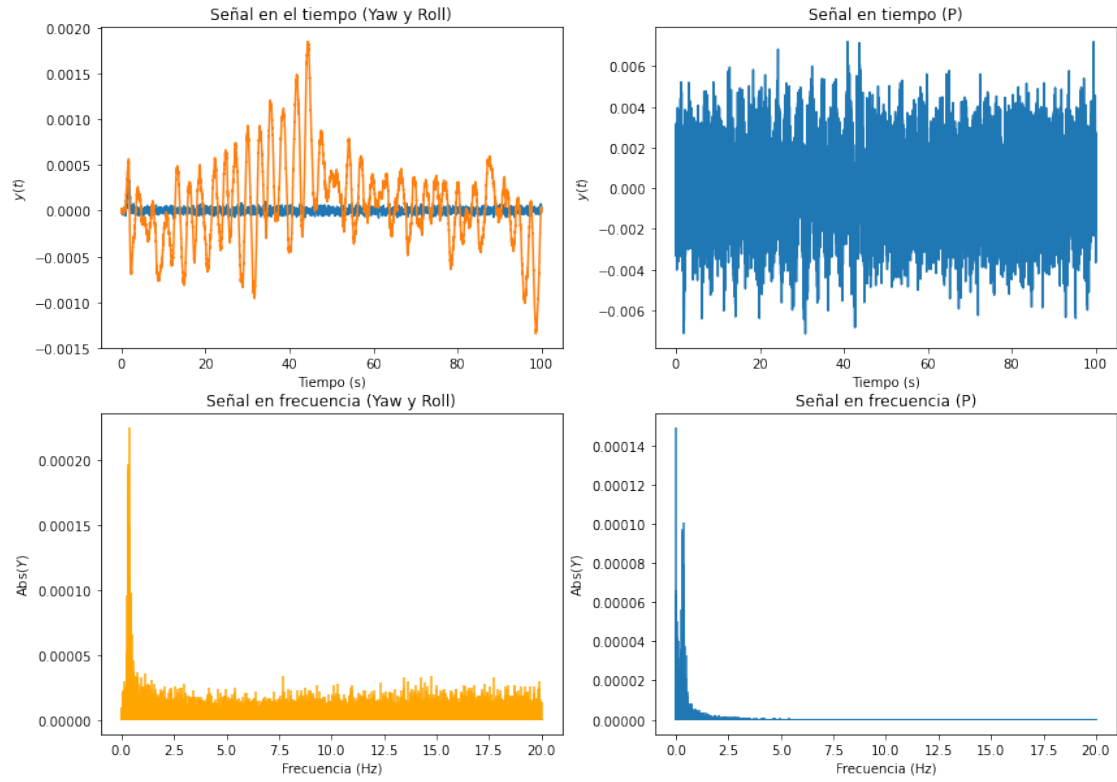


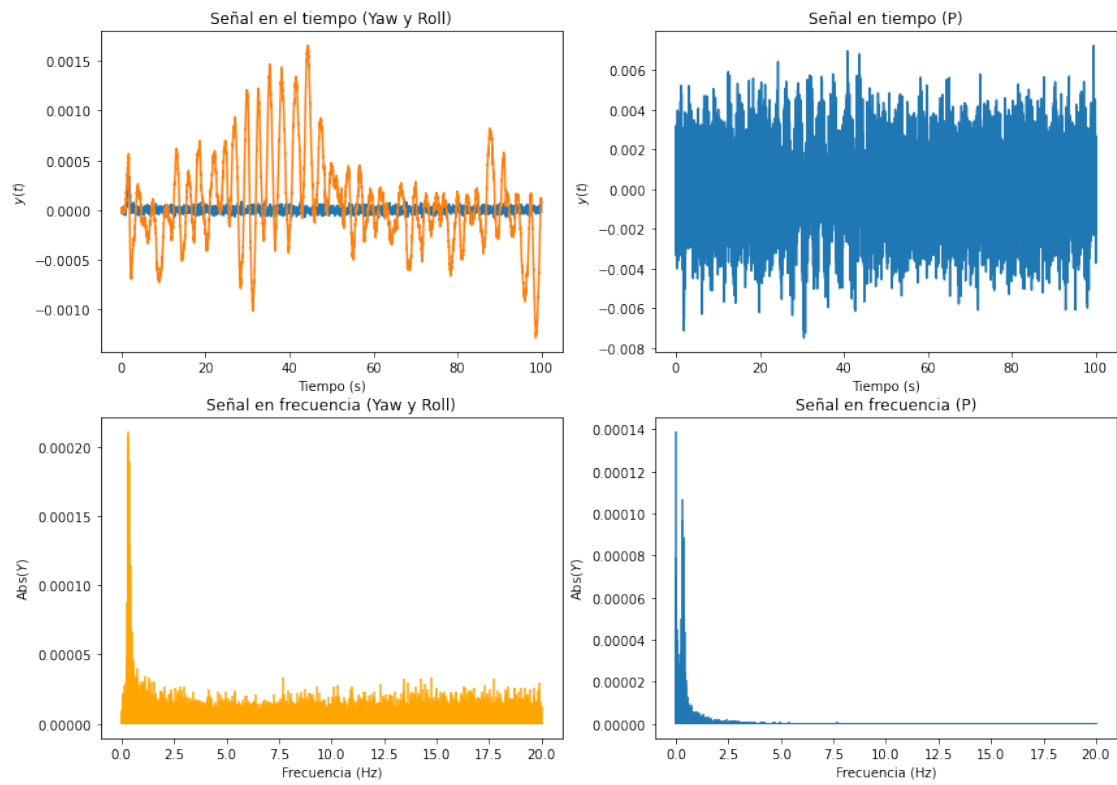


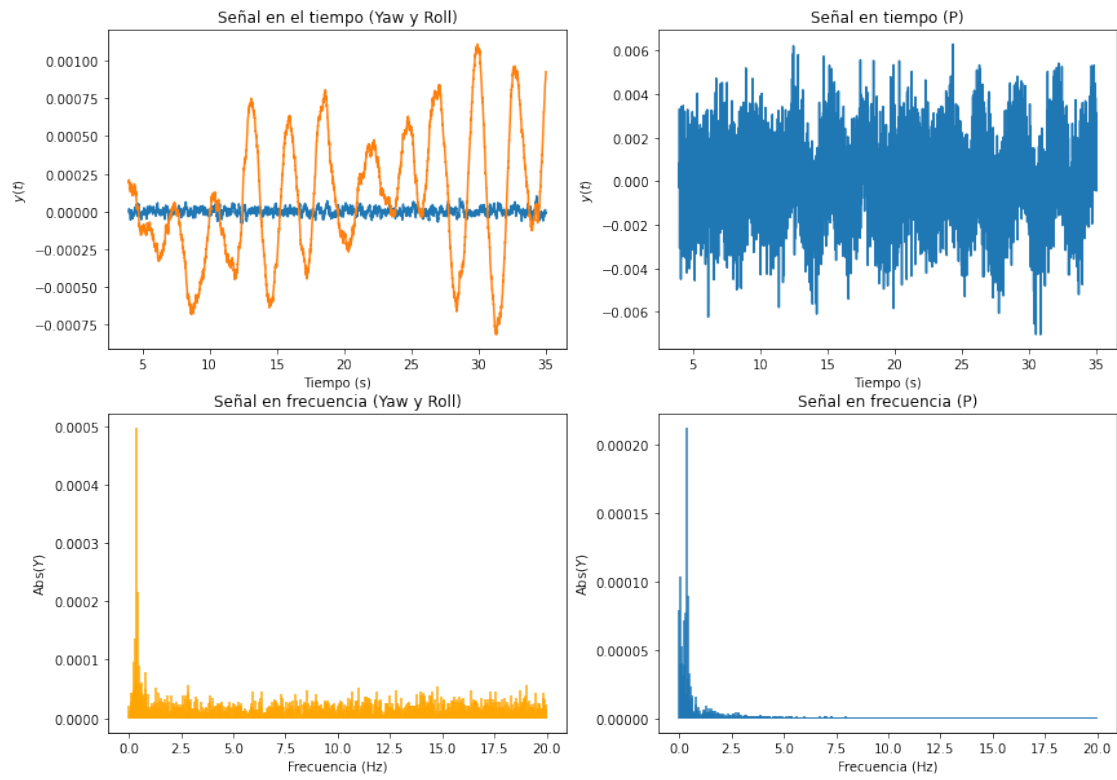
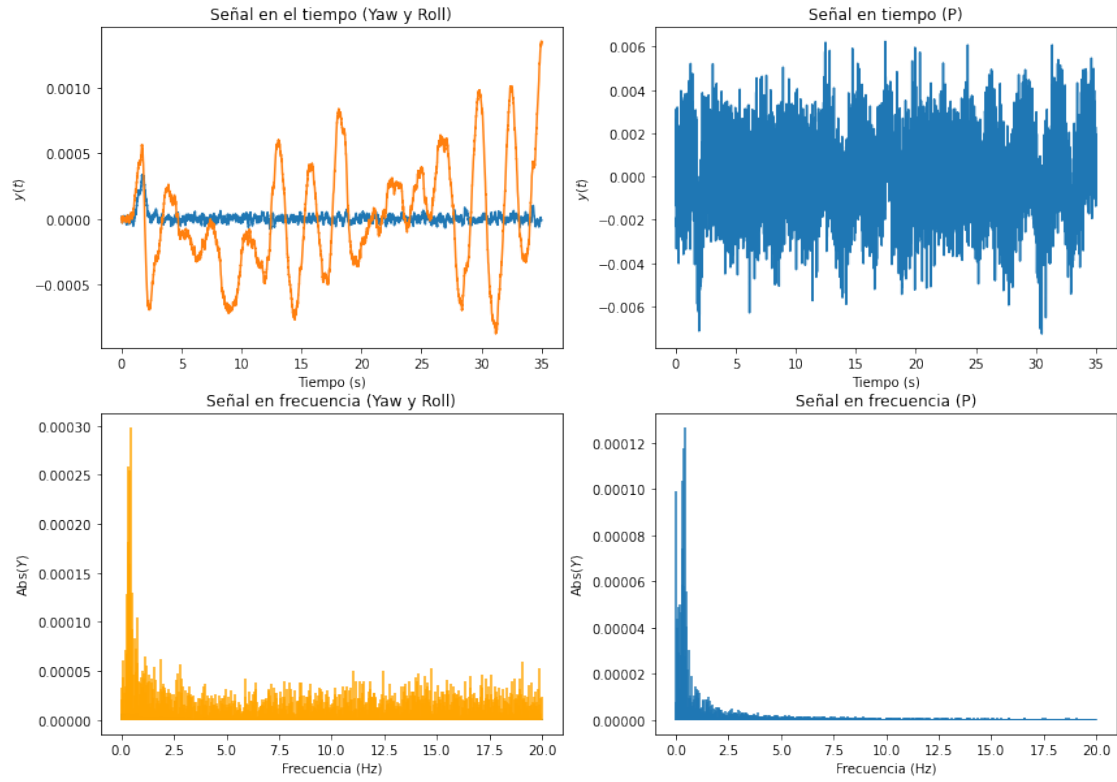


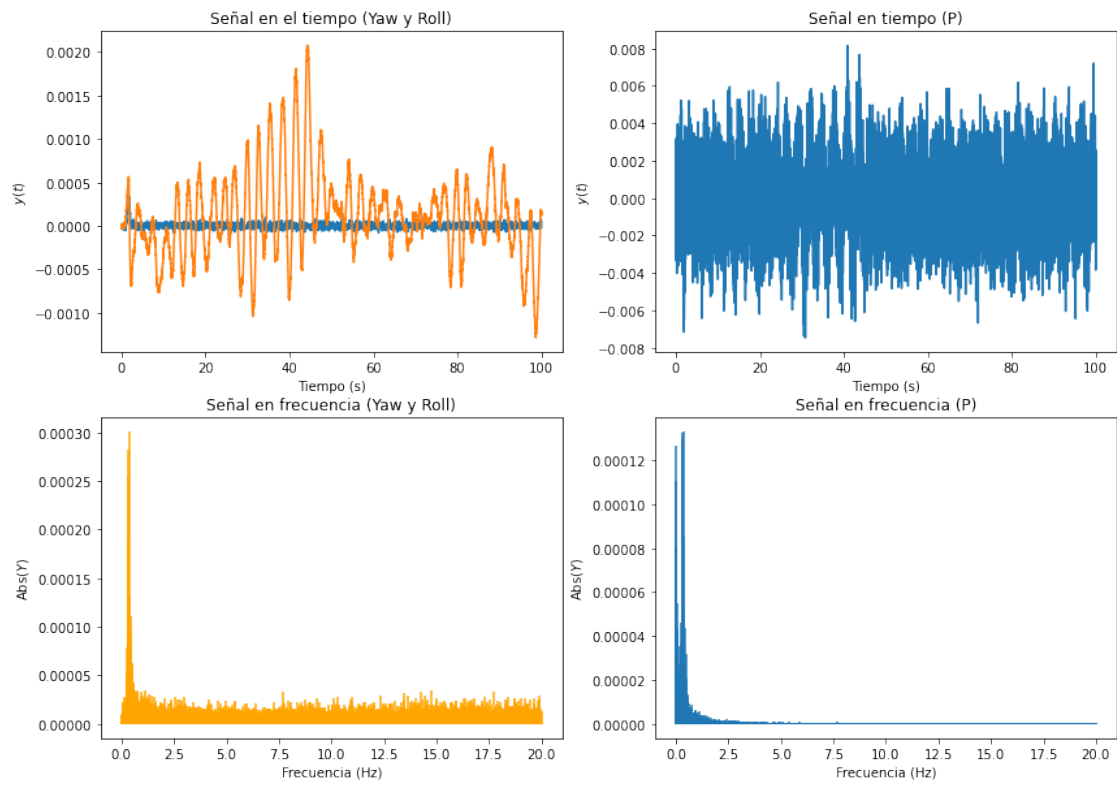


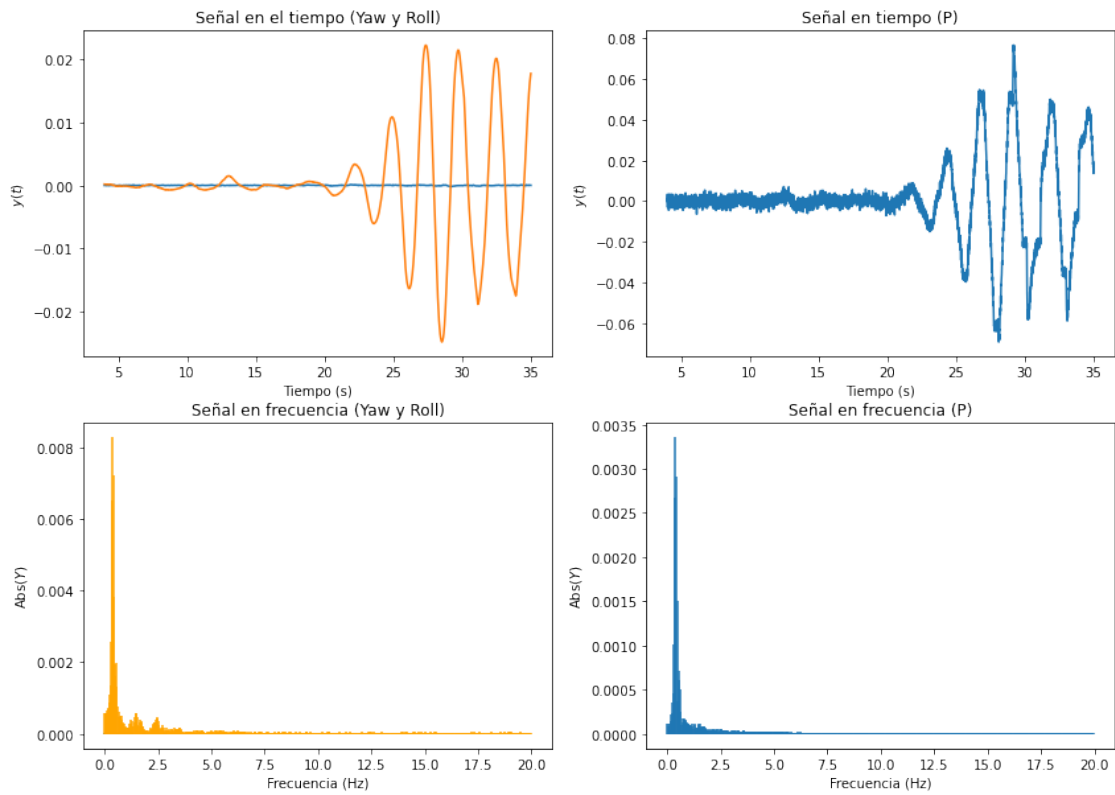
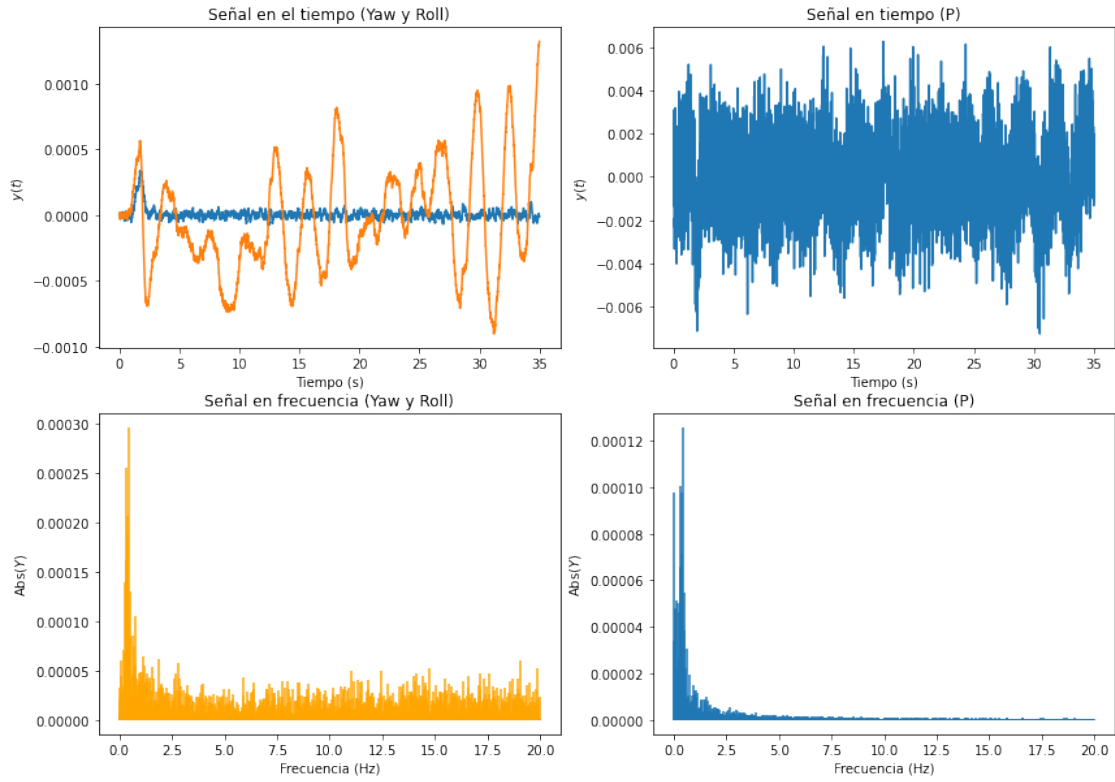
```
[15]: for df in random.choices(dfs, k = 8):
        plot_fourier(df, states=['Yaw', 'P', 'Roll'])
```











## Histograma

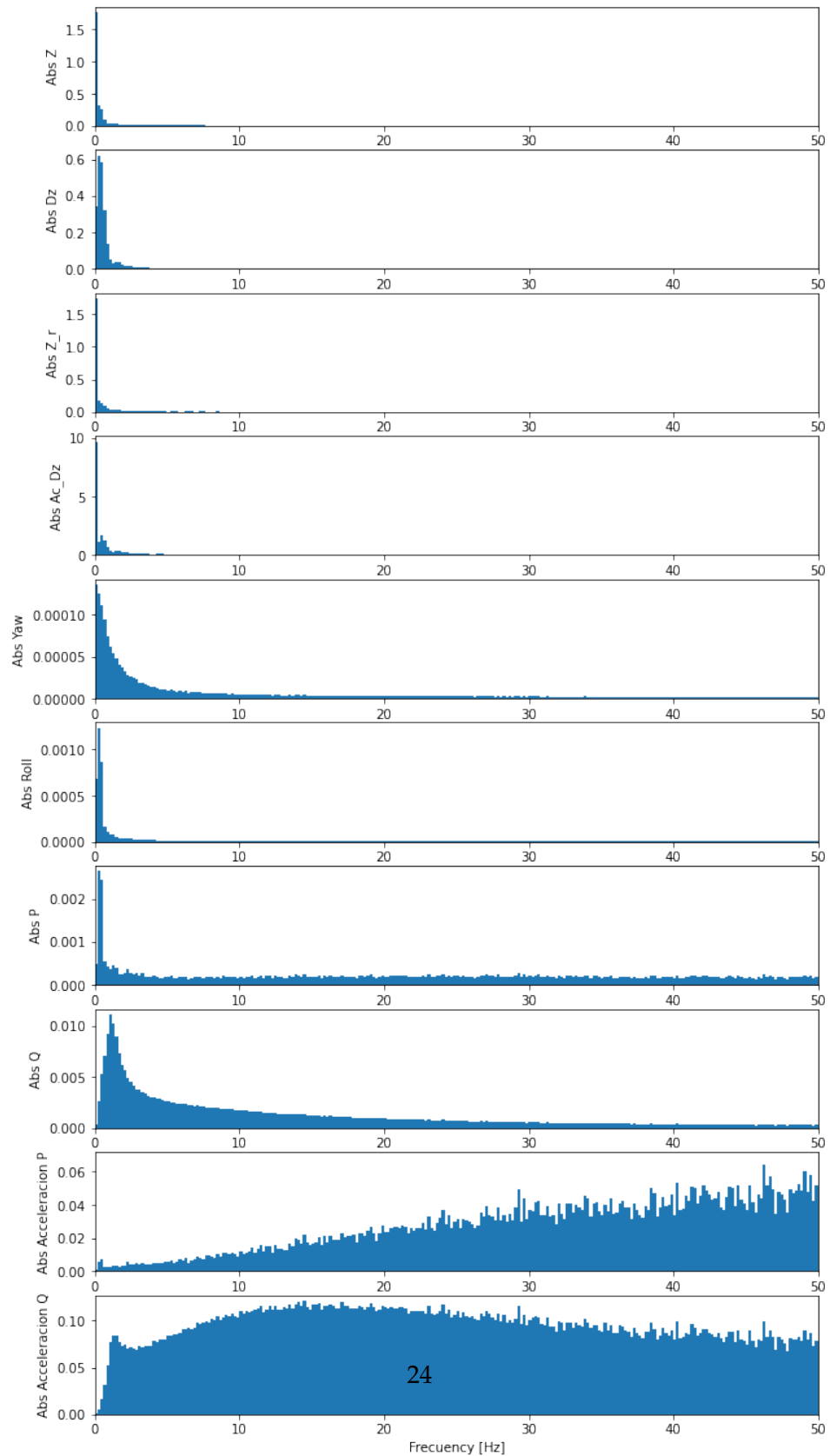
```
[16]: Fourier = []
      for i, df in enumerate(dfs):
          dt = df['Time'][1]-df['Time'][0]
          n = len(df['Time'])
          Fourier.append({})
          for state in states_list_org:
              Fourier[i][state]={}
              Fourier[i][state]['Y'] = abs(fft(df[state].to_numpy())/n)[0:int(n/2)] #
          → Transformada normalizada
              Fourier[i][state]['X'] = fftfreq(n, dt)[0:int(n/2)]

[17]: F = {}
      for state in states_list_org:
          F[state]={}
          F[state]['X'] = []
          F[state]['Y'] = []
          for f in Fourier:
              F[state]['X'] = np.concatenate([F[state]['X'], f[state]['X']])
              F[state]['Y'] = np.concatenate([F[state]['Y'], f[state]['Y']])

[22]: fig, axs = plt.subplots(len(states_list_org), 1, figsize=(10, 20))
      fig.suptitle('Fourier Transform Histogram per State')
      for i, state in enumerate(states_list_org):
          axs[i].hist(F[state]['X'], bins=10*n_bins, weights=((F[state]['Y']+1e-7)/
          → len(Fourier)))
          axs[i].set_ylabel(f'Abs {state}')
          axs[i].set_xlim(0, 50)
          axs[i].set_xlabel('Frecuency [Hz]')

[22]: Text(0.5, 0, 'Frecuency [Hz]')
```

Fourier Transform Histogram per State





## 0.1.8 Análisis de Características - Método Estático

```
[19]: dataset.describe()
```

```
[19]:
```

	Time	X	Y	Z	Yaw \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	4.281013e+01	-3.350768e-04	-3.359317e-04	-1.503027e+00	2.256389e-06
std	2.896393e+01	1.074610e-02	1.136344e-02	2.700165e-01	2.768997e-04
min	0.000000e+00	-3.573908e-01	-5.542104e-01	-3.318702e+00	-1.254526e-02
25%	1.806000e+01	-1.364899e-03	-1.247569e-03	-1.624232e+00	-1.617708e-05
50%	3.521000e+01	-3.312734e-04	-2.898734e-04	-1.499096e+00	1.685380e-07
75%	6.760500e+01	7.783263e-04	6.850931e-04	-1.367094e+00	1.726007e-05
max	1.000000e+02	3.491811e-01	5.142932e-01	-4.311047e-02	1.233279e-02

	Roll	Pitch	Dx	Dy	Dz \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	7.887180e-05	-8.071778e-05	3.052208e-06	2.332470e-06	-1.521121e-02
std	3.506121e-03	5.052975e-03	4.558698e-02	2.659362e-02	4.544270e-01
min	-1.588011e-01	-1.569786e-01	-1.510064e+00	-1.360795e+00	-4.524404e+00
25%	-2.072770e-04	-3.469328e-04	-8.009977e-04	-1.753193e-03	-9.322637e-03
50%	6.043164e-05	-7.090611e-05	-1.754226e-06	0.000000e+00	1.322787e-04
75%	3.423737e-04	1.998152e-04	7.909324e-04	1.652702e-03	8.660311e-03
max	1.636863e-01	1.538717e-01	1.486235e+00	1.353771e+00	3.034264e+00

	P	Q	R	Motor1	Motor2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.573519e-06	1.412977e-06	-3.197492e-07	2.365206e+02	-2.365207e+02
std	9.057596e-03	3.927625e-02	2.225740e-03	3.395921e+01	3.395556e+01
min	-5.037703e-01	-3.575321e+00	-1.820346e-01	1.000000e+01	-4.613083e+02
25%	-1.220251e-03	-1.284665e-03	-9.134351e-04	2.355287e+02	-2.375263e+02
50%	4.287249e-06	-4.722129e-06	-4.495727e-06	2.365282e+02	-2.365222e+02
75%	1.240739e-03	1.283613e-03	9.029967e-04	2.374787e+02	-2.354865e+02
max	4.727377e-01	3.632708e+00	1.794934e-01	4.609845e+02	-1.000000e+01

	Motor3	Motor4	X_r	Y_r	Z_r \
count	1.295895e+06	1.295895e+06	1295895.0	1295895.0	1.295895e+06
mean	2.365206e+02	-2.365208e+02	0.0	0.0	-1.482117e+00
std	3.394866e+01	3.396455e+01	0.0	0.0	2.440522e-01
min	1.000000e+01	-4.609848e+02	0.0	0.0	-2.774623e+00
25%	2.355364e+02	-2.375314e+02	0.0	0.0	-1.620000e+00
50%	2.365288e+02	-2.365235e+02	0.0	0.0	-1.500000e+00
75%	2.374726e+02	-2.354862e+02	0.0	0.0	-1.351574e+00
max	4.611932e+02	-1.000000e+01	0.0	0.0	0.000000e+00

	Yaw_r	Pitch_r	Roll_r	Dx_r	Dy_r	Dz_r \
count	1295895.0	1295895.0	1295895.0	1295895.0	1295895.0	1.295895e+06
mean	0.0	0.0	0.0	0.0	0.0	-2.193838e-02
std	0.0	0.0	0.0	0.0	0.0	8.223455e+00
min	0.0	0.0	0.0	0.0	0.0	-3.460435e+02
25%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
50%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
75%	0.0	0.0	0.0	0.0	0.0	0.000000e+00
max	0.0	0.0	0.0	0.0	0.0	3.807929e+02

	P_r	Q_r	R_r	Flag_Pitch_Roll	Ac_Dx \
count	1295895.0	1295895.0	1295895.0	1295895.0	1.295895e+06
mean	0.0	0.0	0.0	1.0	9.007334e-02
std	0.0	0.0	0.0	0.0	2.089574e-01
min	0.0	0.0	0.0	1.0	-8.013137e-01
25%	0.0	0.0	0.0	1.0	-5.126491e-02
50%	0.0	0.0	0.0	1.0	9.049701e-02
75%	0.0	0.0	0.0	1.0	2.315978e-01
max	0.0	0.0	0.0	1.0	8.437319e-01

	Ac_Dy	Ac_Dz	Gyro P	Gyro Q	Gyro R \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-5.992936e-02	-9.419827e+00	-9.498428e-03	-7.498578e-03	1.499741e-03
std	1.930278e-01	1.427087e+00	9.045007e-03	3.952919e-02	2.383690e-03
min	-8.834900e-01	-1.963813e+01	-5.125701e-01	-3.605846e+00	-1.798834e-01
25%	-1.905823e-01	-9.646096e+00	-1.071855e-02	-8.792938e-03	4.273835e-04
50%	-5.959882e-02	-9.423019e+00	-9.495718e-03	-7.504752e-03	1.502034e-03
75%	7.084187e-02	-9.203916e+00	-8.260985e-03	-6.208120e-03	2.568620e-03
max	6.365708e-01	8.609541e-01	4.625807e-01	3.648603e+00	1.830248e-01

	Sonar Altitud	Pressure Altitud	Bat_V	Bat_Percentage \
count	1.295895e+06	1.295895e+06	1295895.0	1295895.0
mean	1.504551e+00	1.012529e+05	3.5	70.0
std	2.614817e-01	3.211676e+00	0.0	0.0
min	4.400000e-01	1.012314e+05	3.5	70.0
25%	1.367890e+00	1.012514e+05	3.5	70.0
50%	1.498847e+00	1.012529e+05	3.5	70.0
75%	1.624728e+00	1.012545e+05	3.5	70.0
max	3.290363e+00	1.012704e+05	3.5	70.0

	Acceleracion X	Acceleracion Y	Acceleracion Z	Acceleracion P \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	5.755633e-06	-1.250541e-04	-1.736796e-04	6.004077e-05
std	5.950564e-01	8.118001e-02	1.435704e+00	4.211311e-01
min	-8.899236e+01	-4.018375e+00	-1.017578e+01	-8.605778e+00
25%	-8.635339e-03	-2.117448e-02	-1.253270e-01	-2.800144e-01
50%	0.000000e+00	0.000000e+00	-1.526624e-03	-3.745025e-03

75%	8.559134e-03	2.102111e-02	1.203972e-01	2.815473e-01
max	9.037159e+01	4.432607e+00	1.060796e+01	7.373816e+00

	Acceleracion Q	Acceleracion R	Z1	Dz1 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	2.189152e-05	-8.116385e-06	-1.502918e+00	-1.521034e-02
std	1.670874e+00	2.951463e-01	2.703173e-01	4.544223e-01
min	-1.592394e+02	-8.434417e+00	-3.318702e+00	-4.524404e+00
25%	-3.060012e-01	-1.960427e-01	-1.624222e+00	-9.322412e-03
50%	-3.043190e-03	-7.657334e-04	-1.499095e+00	1.304944e-04
75%	3.080426e-01	1.959933e-01	-1.367004e+00	8.658752e-03
max	1.689137e+02	7.723484e+00	0.000000e+00	3.034264e+00

	Z_r1	Ac_Dz1	Yaw1	Roll1	P1 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-1.482007e+00	-9.419130e+00	2.257514e-06	7.887457e-05	1.273315e-06
std	2.443724e-01	1.429272e+00	2.768997e-04	3.505640e-03	9.056365e-03
min	-2.774623e+00	-1.963813e+01	-1.254526e-02	-1.588011e-01	-5.037703e-01
25%	-1.620000e+00	-9.646093e+00	-1.617320e-05	-2.072770e-04	-1.220251e-03
50%	-1.500000e+00	-9.423018e+00	1.684218e-07	6.035169e-05	4.136165e-06
75%	-1.351574e+00	-9.203855e+00	1.726007e-05	3.423065e-04	1.240343e-03
max	0.000000e+00	8.609541e-01	1.233279e-02	1.636863e-01	4.727377e-01

	Q1	Acceleracion P1	Acceleracion Q1	Z2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.303519e-06	1.038755e-05	-3.086813e-05	-1.502809e+00
std	3.927625e-02	4.210900e-01	1.670862e+00	2.706177e-01
min	-3.575321e+00	-8.605778e+00	-1.592394e+02	-3.318702e+00
25%	-1.284665e-03	-2.800144e-01	-3.060012e-01	-1.624207e+00
50%	-4.722129e-06	-3.745025e-03	-3.043190e-03	-1.499094e+00
75%	1.283548e-03	2.814385e-01	3.079379e-01	-1.366925e+00
max	3.632708e+00	7.373816e+00	1.689137e+02	0.000000e+00

	Dz2	Z_r2	Ac_Dz2	Yaw2	Roll2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-1.520958e-02	-1.481896e+00	-9.418387e+00	2.258437e-06	7.887883e-05
std	4.544177e-01	2.446916e-01	1.431431e+00	2.768996e-04	3.505148e-03
min	-4.524404e+00	-2.774623e+00	-1.963813e+01	-1.254526e-02	-1.588011e-01
25%	-9.322241e-03	-1.620000e+00	-9.646035e+00	-1.617313e-05	-2.072770e-04
50%	1.281137e-04	-1.500000e+00	-9.422956e+00	1.683297e-07	6.028301e-05
75%	8.657166e-03	-1.351574e+00	-9.203737e+00	1.726007e-05	3.422434e-04
max	3.034264e+00	0.000000e+00	8.609541e-01	1.233279e-02	1.636863e-01

	P2	Q2	Acceleracion P2	Acceleracion Q2 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	1.221378e-06	1.457860e-06	2.812219e-05	-7.125476e-06
std	9.055216e-03	3.927624e-02	4.210710e-01	1.670856e+00

min	-5.037703e-01	-3.575321e+00	-8.605778e+00	-1.592394e+02
25%	-1.219919e-03	-1.284440e-03	-2.799338e-01	-3.059567e-01
50%	4.130569e-06	-4.424624e-06	-3.700447e-03	-2.996484e-03
75%	1.240317e-03	1.283548e-03	2.814385e-01	3.079379e-01
max	4.727377e-01	3.632708e+00	7.373816e+00	1.689137e+02

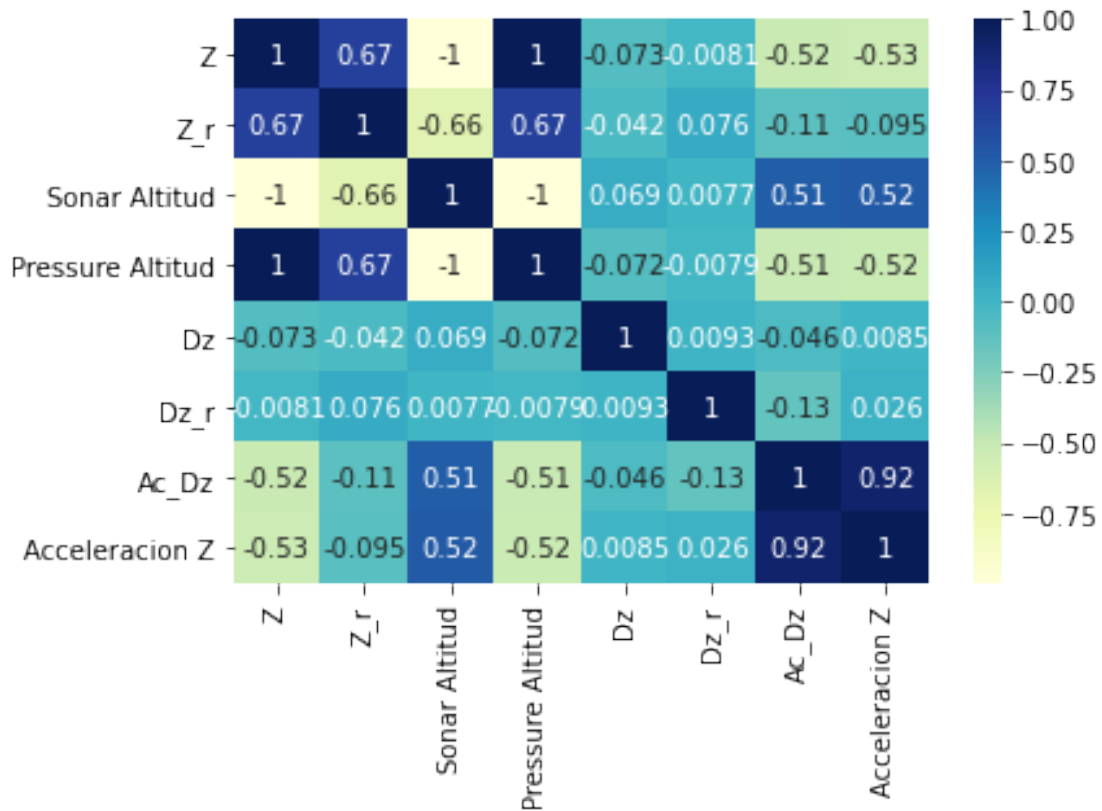
	Z3	Dz3	Z_r3	Ac_Dz3	Yaw3 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	-1.502699e+00	-1.520887e-02	-1.481784e+00	-9.417694e+00	2.258529e-06
std	2.709177e-01	4.544130e-01	2.450097e-01	1.433600e+00	2.768996e-04
min	-3.318702e+00	-4.524404e+00	-2.774623e+00	-1.963813e+01	-1.254526e-02
25%	-1.624196e+00	-9.322085e-03	-1.620000e+00	-9.646004e+00	-1.617308e-05
50%	-1.499093e+00	1.261767e-04	-1.500000e+00	-9.422915e+00	1.660678e-07
75%	-1.366811e+00	8.655519e-03	-1.351574e+00	-9.203696e+00	1.726007e-05
max	0.000000e+00	3.034264e+00	0.000000e+00	8.609541e-01	1.233279e-02

	Roll3	P3	Q3	Acceleracion P3 \
count	1.295895e+06	1.295895e+06	1.295895e+06	1.295895e+06
mean	7.888335e-05	1.080767e-06	1.493488e-06	2.985777e-06
std	3.504645e-03	9.054026e-03	3.927624e-02	4.210606e-01
min	-1.588011e-01	-5.037703e-01	-3.575321e+00	-8.605778e+00
25%	-2.072770e-04	-1.219723e-03	-1.284153e-03	-2.799338e-01
50%	6.020387e-05	3.981350e-06	-4.307566e-06	-3.700447e-03
75%	3.421859e-04	1.240157e-03	1.283347e-03	2.813485e-01
max	1.636863e-01	4.727377e-01	3.632708e+00	7.373816e+00

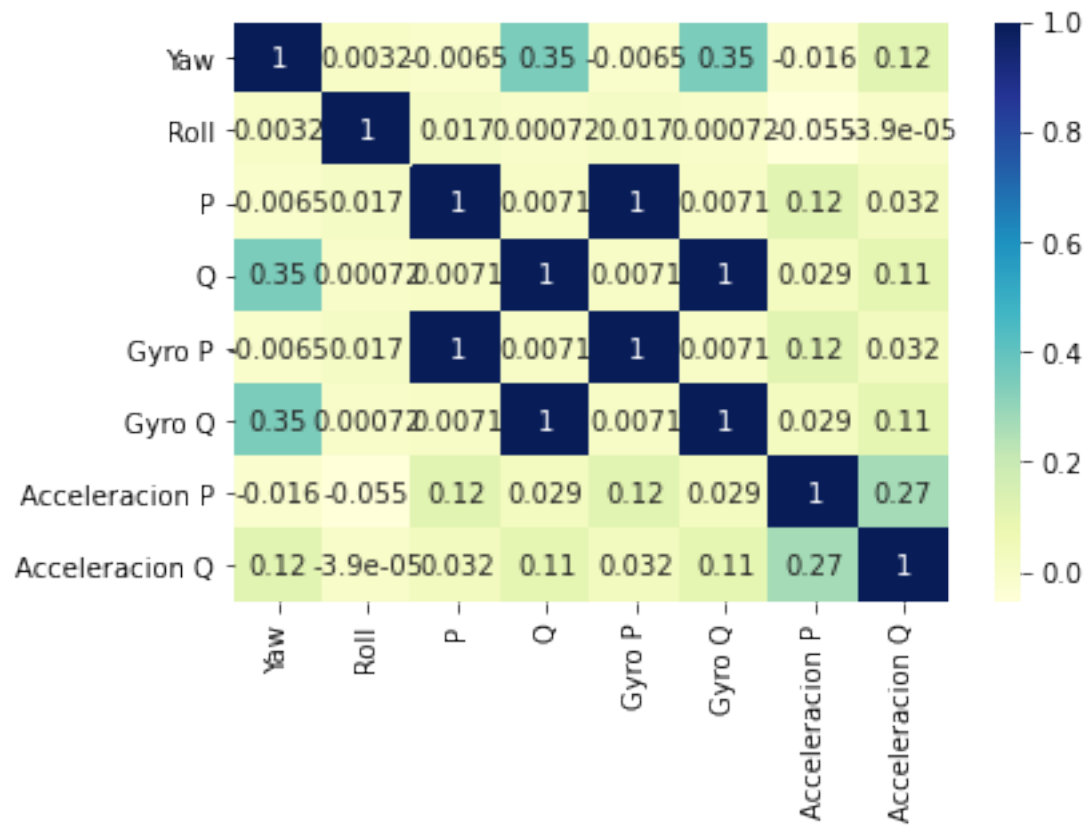
	Acceleracion Q3
count	1.295895e+06
mean	-3.594663e-05
std	1.670853e+00
min	-1.592394e+02
25%	-3.059567e-01
50%	-2.996484e-03
75%	3.079038e-01
max	1.689137e+02

## Mapa de Correlación

```
[27]: corr_states = ["Z", "Z_r", "Sonar Altitud", "Pressure Altitud",
                    "Dz", "Dz_r", "Ac_Dz", "Acceleracion Z"]
correlation = dataset[corr_states].corr() #corr() method of pandas library
→calculates correlation between columns of dataframe
sns.heatmap(correlation, cmap="YlGnBu", annot=True)
plt.show()
```



```
[29]: corr_states = ["Yaw", "Roll", "P", "Q",
                    "Gyro P", "Gyro Q",
                    "Acceleracion P", "Acceleracion Q"]
correlation = dataset[corr_states].corr() #corr() method of pandas library
→calculates correlation between columns of dataframe
sns.heatmap(correlation,cmap="YlGnBu",annot=True)
plt.show()
```



[ ]: