

# Untitled

May 9, 2022

## 0.1 InPTA Test eccentricity data and plotting

### 0.1.1 adding the libraries for analysis

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

## 0.2 importing the file using pandas

```
[156]: exoplanet = pd.read_csv('exoplanet.eu_catalog_1000.csv', index_col=False)
↳#exoplanet data
exoplanet.shape
exoplanet.head()
```

```
[156]:
```

	#	name	planet_status	mass	mass_error_min	mass_error_max	mass_sini	\
0	11	Com b	Confirmed	NaN	NaN	NaN	16.1284	
1	11	Oph b	Confirmed	21.0	3.0	3.0	NaN	
2	11	UMi b	Confirmed	NaN	NaN	NaN	11.0873	
3	14	And b	Confirmed	NaN	NaN	NaN	4.6840	
4	14	Her b	Confirmed	9.1	NaN	NaN	5.2150	

  

	mass_sini_error_min	mass_sini_error_max	radius	radius_error_min	...	\
0	1.50	1.50	NaN	NaN	...	
1	NaN	NaN	NaN	NaN	...	
2	1.10	1.10	NaN	NaN	...	
3	0.23	0.23	NaN	NaN	...	
4	NaN	NaN	NaN	NaN	...	

  

	star_sp_type	star_age	star_age_error_min	star_age_error_max	star_teff	\
0	G8 III	NaN	NaN	NaN	4742.0	
1	M9	0.011	0.002	0.002	2375.0	
2	K4III	1.560	0.540	0.540	4340.0	
3	K0III	NaN	NaN	NaN	4813.0	
4	K0 V	5.100	NaN	NaN	5311.0	

  

	star_teff_error_min	star_teff_error_max	star_detected_disc	\
0	100.0	100.0	NaN	

1	175.0	175.0	NaN
2	70.0	70.0	NaN
3	20.0	20.0	NaN
4	87.0	87.0	NaN

	star_magnetic_field	star_alternate_names
0	NaN	NaN
1	NaN	Oph 1622-2405, Oph 11A
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

[5 rows x 98 columns]

### 0.3 extracting eccentricity column from the data

```
[157]: e = exoplanet['eccentricity']
e.head()
```

```
[157]: 0    0.231
1     NaN
2    0.080
3    0.000
4    0.369
Name: eccentricity, dtype: float64
```

### 0.4 removing the NaN values from the list using dropna() function

```
[159]: e_non_nan = e.dropna()
e_non_nan.head()
```

```
[159]: 0    0.231
2    0.080
3    0.000
4    0.369
5    0.640
Name: eccentricity, dtype: float64
```

### 0.5 plotting the histogram of the eccentricity data

```
[160]: import plotly.express as px
fig = px.histogram(e_non_nan, nbins=100, x="eccentricity")
fig.show()
```



0.6 replotting the after Gaussianization of the distribution using Box-transformation either using `scipy.stats.boxcox` or from first principles

```
[161]: from scipy.stats import boxcox
       e_non_nan
```

```
[161]: 0      0.23100
      2      0.08000
      3      0.00000
      4      0.36900
      5      0.64000
      ...
     5023    0.01186
     5024    0.24450
     5025    0.31600
     5026    0.00536
     5027    0.32000
      Name: eccentricity, Length: 2027, dtype: float64
```

0.6.1 eccentricity values contains a 0 eccentric orbit and boxcox needs purely positive values to function. need to do something with 0

```
[163]: e = e_non_nan.values
       idx = np.where(e==0)

       #perturbing the 0 eccentric orbits with an epsilon quantity for boxcox to work
       for i in idx:
           epsilon = 1e-20 #very small perturbation
           e[i] = epsilon
```

```
[142]: e_gauss = boxcox(e)[0] #gaussianized eccentricities with pertubation for the 0
       ↪eccentricity
```

```
[164]: plt.hist(e_gauss, bins=20, range = [-3, 0])  
plt.xlabel('bins')  
plt.ylabel('dN')  
plt.show()
```

