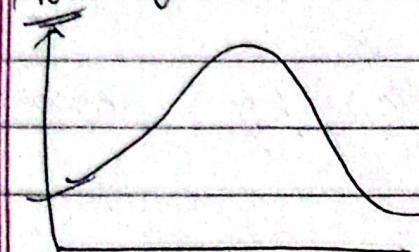


Function transformers.

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Probability



↳
normal distribution

- we won't get the data normally distributed
- The main objective of the function transformer is to distribute data normally.
- It makes problem solving more easier.
- Statistics algorithms like linear regression, logistic regression needs to be normally distributed data.
However, DecisionTree, RandomForest etc they doesn't need normally distributed data.

Sklearn

Function
Transformers

→ log transform

→ Reciprocal

→ square √, sqrt

↳ custom

Power
Transformer

→ Box-Cox

→ Yeo-Johnson.

Quantile
Transformer.

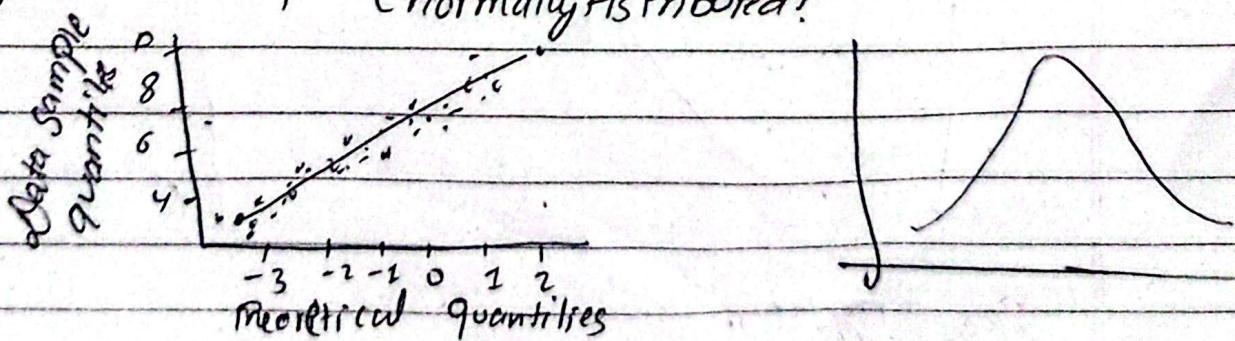
② How to find if data is normal?

way 1: Seaborn disPlot:

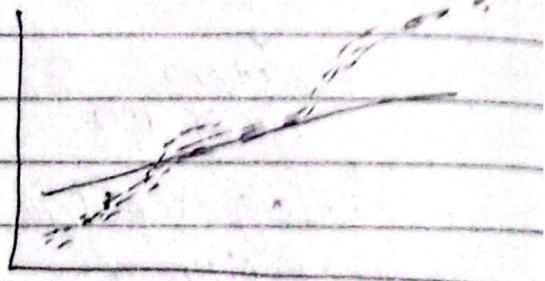
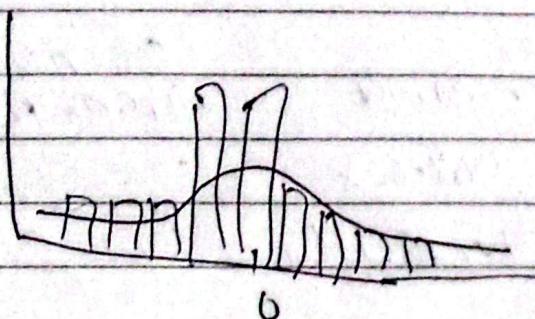
way 2: Pandas .skew()

it should be 0.

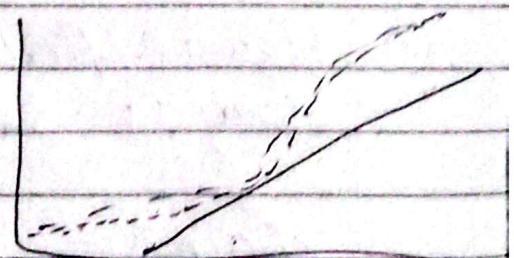
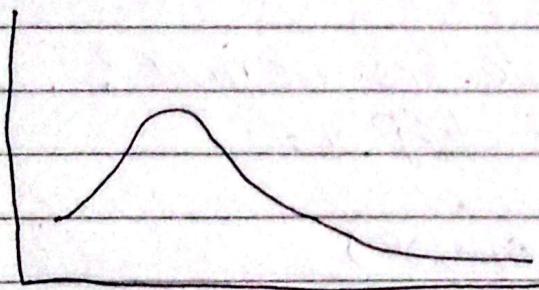
way 3: QQ plot: [most reliable]
normally distributed?



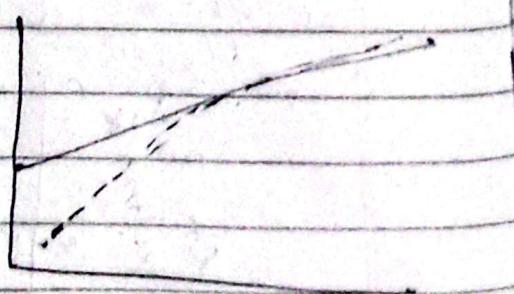
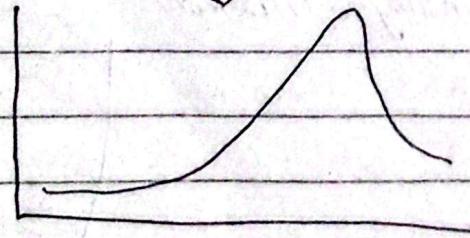
too peak in middle
long peak:



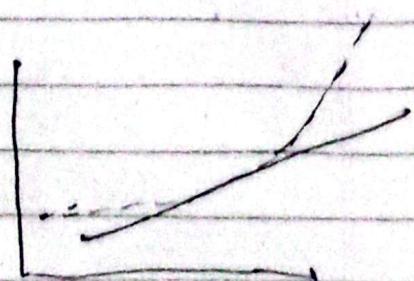
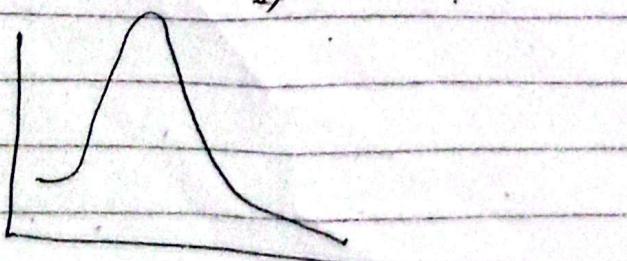
• Skewed data:



(Skew left)

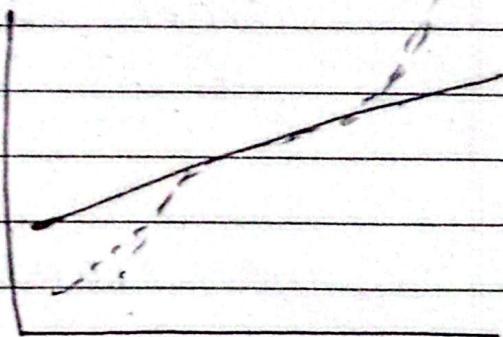
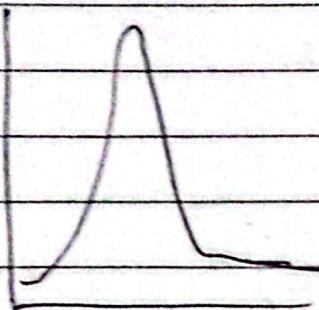


(Skew right)

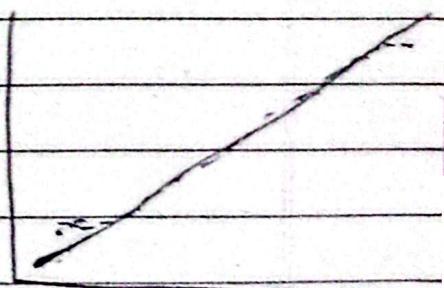
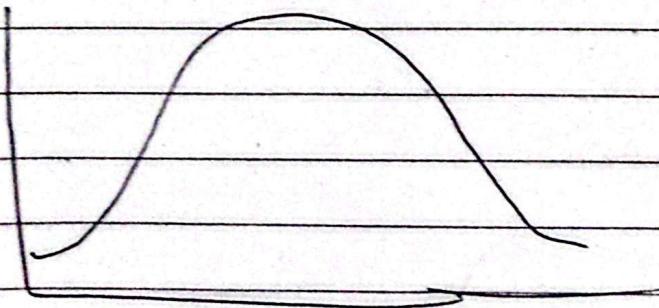


Fat-tails

Q-Q plot.



Thin-tails



Log Transform.

- Take log of that value
- Then, data will be more normally distributed, ~~than~~ than that of right now.

• Better to use in right skewed data.
Don't do in -ve values

*

Reciprocal Transforms: ($\frac{1}{n}$)
larger values changes to smaller
values and vice versa.

#

Square transforms (n^2)
→ use for left skewed data,

Square root transforms (\sqrt{n})

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

      import scipy.stats as stats

      from sklearn.model_selection import train_test_split
      from sklearn.metrics import accuracy_score
      from sklearn.model_selection import cross_val_score
      from sklearn.linear_model import LogisticRegression
      from sklearn.tree import DecisionTreeClassifier
      from sklearn.compose import ColumnTransformer
      from sklearn.preprocessing import FunctionTransformer

[2]: df = pd.read_csv('titanic_dataset.csv', usecols=['Age', 'Fare', 'Survived'])
      df.sample(5)
```

```
[2]:   Survived  Age      Fare
    267        0  NaN     7.5500
    284        1  2.0    20.2125
      8        1  18.0    7.2292
    110        0  41.0   15.0458
    239        1  48.0   106.4250
```



Feb 28 23:01

[9]: X_train.shape

[9]: (334, 2)

[10]: X_test.shape

[10]: (84, 2)

```
[11]: print(X_train['Fare'])
print(X_train['Fare'].dtype)
print(X_train['Fare'].head())
```

```
155    24.00000
287    24.00000
300    32.00000
394    29.00000
261    21.00000
```

```
...
211    30.27259
67     47.00000
25     50.00000
196    6.00000
175    15.00000
```

```
Name: Fare, Length: 334, dtype: float64
float64
```

```
155    24.0
287    24.0
300    32.0
394    29.0
261    21.0
```

```
Name: Fare, dtype: float64
```

+ X C Code

Name: Fare, dtype: float64

```
[14]: stats.probplot(X_train['Fare'], dist = 'norm', plot=plt)
plt.show()
```

