

Seaborn

Visualization Distribution

Library that uses Matplotlib to plot graphs.
Used to visualize random distribution.

Displot

stands for distribution plots.

Takes array as input.

plots a curve corresponding to the distribution points.

Note Matplotlib is used to do so.

```
import matplotlib.pyplot as plt
import seaborn as sns
```

```
sns.displot([0, 1, 2, 3, 4])
plt.show()
```

if we do not want histogram then we can use parameter
'kind = "kde"'

```
sns.displot([0, 1, 2, 3, 4], kind = 'kde')
```

Normal Distribution

Most important distribution

Also called Gaussian distribution.

Fits probability of many events eg → IQ score, heartbeat.

we use `random.normal()` method.



Three parameters.

- ~~me~~ loc \rightarrow (Mean) where peak of bell exists.
- scale \rightarrow (standard deviation) how flat the graph should
- size \rightarrow shape of returned array

Note loc and scale are not mandatory.

```
import numpy as np
from numpy import random as np
```

```
x = np.random.normal(loc=1, scale=2, size=(2, 3))
print(x)
```

\Rightarrow creates a 2×3 distribution.

Note curve of normal distribution also known as bell curve.

Binomial distribution

It is a Discrete distribution.

describe outcome of binary scenarios i.e. toss of coin etc.

Three parameters

n - number of trials

p - probability of each trial

size - shape of array.

Discrete distribution

separate set of events eg toss of coin

while height of people is continuous.

```
from numpy import random
x = random.binomial(n=10, p=0.5, size=10)
print(x)
```

=> [4 7 3 2 8 6 3 4 7 3]

Here, For each value (10 values) the success is counted out of 10 trials.

For first value '4', it denotes there were 4 successful trials out of 10.

Note if there are enough data points it will be quite similar to normal distribution with certain loc and scale.

iii) Poisson Distribution

Discrete distribution.

Estimates how many times an event can happen in specified time.

Two parameters

lam \rightarrow known no. of occurrence

size \rightarrow shape of array.

```
x = random.poisson(lam=2, size=10)
print(x)
```




Normal vs Poisson

- Normal is continuous while Poisson is discrete.
- For large enough Poisson value it will become similar to normal.

Binomial vs Poisson

Binomial can have only two possible outcomes while Poisson can have number of outcomes.

For very large n and near-zero p , it is near identical to Poisson such that $n \cdot p$ is nearly equal to λ .

`binomial = random.binomial (n=1000, p=0.01, size=1000)`

`Poisson = random.poisson (lam=10, size=1000)`

iv Uniform distribution

describe probability where every event has equal chance of occurring. Eg generation of random numbers.

Three parameters

`low` - lower bound, 0.0 default

`high` - higher bound, 1.0 default.

`size` - shape of array.

i continuous uniform distribution

eg \rightarrow generation of real number.

`x = random.uniform (low=2, high=5, size=5)`

`print(x)`

=> generate 5 floating numbers between 2 and 5.

2 Discrete uniform distribution

have given limited values. eg rolling a dice.
Here numbers 1-6 have equal probability in all.

$x = \text{random.randnint}(low=2, high=5, size=5)$

✓ logistic Distribution

used to describe growth.
used extensively in ML in logistic regression etc.

Three parameters:

loc \rightarrow mean, Default 0

scale \rightarrow standard deviation, default 1

size \rightarrow shape of array.

logistic vs Normal

- Near identical but logistic has more area under tails.
- represents more probability of occurrence of an event further away from mean.

Note:- higher value of scale, normal is identical apart from peak.



vii Multinomial distribution

- generalization of binomial distribution.
- describe probabilities of multinomial scenarios. eg blood type

Three parameters

n - no of trials

$pvals$ - list of probability outcomes.

$size$ - shape.

Note Multinomial samples will not produce a single value. They will produce one value for each $pval$.

$x = \text{randem} \cdot \text{multinomial}(n=6, pvals[1/6]^*6)]$

viii Exponential Distribution

used in describing time till next event eg failure/success

Two parameters

scale - inverse of rate. default is 1.0.

size - shape.

Relation Between Poisson & Exponential

poisson deals with number of occurrences of an event in a time period.

exponential distribution deals with time between these events.

viii) chi square Distribution

used as a basis to verify the hypothesis

Two parameters

df - degree of freedom

size - shape

`x = random.chisquare(df=2, size=(2,3))`

`sns.displot(x)`

`plt.show()`

ix) Rayleigh Distribution

used for signal processing. It has two parameters:-

scale - decide how flat distribution will be default 1.0

size - shape of array.

Note At unit std and 2 deg of freedom rayleigh and chi square represent the same distribution

x) Pareto Distribution

Pareto's law is 80-20 distribution (20% factor cause 80%)

Two parameter

a - shape parameter

size - shape