You are currently looking at **version 1.0** of this notebook. To download notebooks and datafiles, as well as get help on Jupyter notebooks in the Coursera platform, visit the <u>Jupyter Notebook FAQ (https://www.coursera.org/learn/python-data-analysis/resources/0dhYG)</u> course resource.

## **Distributions in Pandas**

```
In [1]: import pandas as pd
         import numpy as np
In [2]: '''
         Suppose we want to simulate the probability of flipping a fair coin 20 times
         , and getting a number greater than or equal to 15. Use np.random.binomial(n, p,
         size)
         to do 10000 simulations of flipping a fair coin 20 times, then see what proporti
         on of the simulations are 15 or greater.
         x = np.random.binomial(20, .5, 10000)
         print((x>=15).mean())
         0.0234
In [3]: np.random.binomial(1, 0.5)
Out[3]: 1
In [4]: np.random.binomial(5000, 0.35)/5000
Out[4]: 0.3548
In [5]: x = np.random.ranf()
         z = np.random.binomial(1000, x, 102)/1000
         print(x)
         print(z)
         print(z.mean())
         0.27826569985021343
         [ 0.285 \ 0.27 \ 0.269 \ 0.252 \ 0.274 \ 0.277 \ 0.276 \ 0.296 \ 0.28 ]
                                                                                 0.287
           0.272 0.274 0.298 0.303
                                         0.284 0.288 0.253 0.297 0.279 0.28
           0.269 \quad 0.267 \quad 0.279 \quad 0.278 \quad 0.272 \quad 0.255 \quad 0.277 \quad 0.271 \quad 0.286 \quad 0.273
           0.274 \quad 0.279 \quad 0.277 \quad 0.307 \quad 0.276 \quad 0.296 \quad 0.284 \quad 0.29 \quad 0.305 \quad 0.284
           0.277 0.274 0.272 0.302 0.298 0.265 0.281 0.287 0.259 0.258
           0.268 \quad 0.277 \quad 0.267 \quad 0.289 \quad 0.269 \quad 0.247 \quad 0.269 \quad 0.293 \quad 0.262 \quad 0.286
           0.276 0.285 0.285 0.27 0.273 0.274 0.277 0.29 0.269 0.265
           0.265 0.28 0.297 0.279 0.321 0.275 0.261 0.277 0.282 0.273
           0.295 \quad 0.309 \quad 0.3 \qquad 0.278 \quad 0.276 \quad 0.286 \quad 0.281 \quad 0.286 \quad 0.31 \quad 0.267
                  0.259 0.252 0.294 0.272 0.266 0.275 0.282 0.278 0.264
           0.29
           0.266 0.286]
         0.27881372549
```

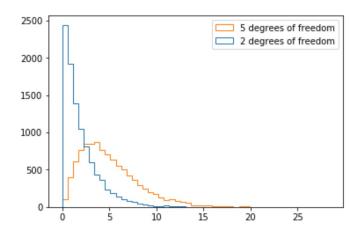
```
In [6]: # unevenly weighted
       chance of tornado = 0.01/100
       np.random.binomial(100000, chance_of_tornado)
Out[6]: 15
In [7]: chance of tornado = 0.01
       sum = 0
       x = 1000000/365 \# number of years
       tornado events = np.random.binomial(1, chance of tornado, int(365*x))
       print(tornado events[0:100])
       print(tornado_events.mean())
       two_days_in_a_row = 0
       for j in range(1,len(tornado events)-1):
          if tornado events[j]==1 and tornado events[j-1]==1:
             two_days_in_a_row+=1
       for i in range(1, len(tornado events)-1):
          if tornado_events[i] == 1:
             sum += 1
       #print('{} tornadoes back to back in {} years'.format(two_days_in_a_row, 1000000
       print(f'{two days in a row} tornadoes back to back in {x} years')
      print(f'{sum} tornadoe(s) in {x}years')
       0.009911
      119 tornadoes back to back in 2739.72602739726 years
      9911 tornadoe(s) in 2739.72602739726years
In [8]: np.random.uniform(0, 1)
Out[8]: 0.46235787224468294
In [9]: | np.random.normal(0.75)
Out[9]: 0.8462291338979055
```

## Formula for standard deviation

$$\sqrt{\frac{1}{N}\sum_{i=1}^N(x_i-\overline{x})^2}$$

```
In [12]: x = (distribution-np.mean(distribution))**2
         print(x[0:10])
         y = (np.mean(distribution)-distribution)**2
         print(y[0:10])
         print(np.mean(distribution))
         \begin{bmatrix} 5.26945543e-01 & 3.22328928e+00 & 2.13017999e-01 & 4.10300405e-02 \end{bmatrix}
            4.89640312e-01 3.91815864e+00 8.04591668e-01 2.69779143e+00
            1.53787345e-04 5.08229271e-01]
         [ 5.26945543e-01 3.22328928e+00 2.13017999e-01 4.10300405e-02
            4.89640312e-01 3.91815864e+00 8.04591668e-01 2.69779143e+00
            1.53787345e-04 5.08229271e-01]
         0.751197815752
In [13]: np.random.normal?
In [14]: | np.std(distribution)
Out[14]: 0.99374325582731005
In [15]: import scipy.stats as stats
         stats.kurtosis(distribution)
Out[15]: 0.057261953745207705
In [16]: stats.skew(distribution)
Out[16]: 0.08946591005782702
In [17]: print(stats.skew([1, 2, 3, 4, 5, 6]))
         print(stats.skew([1, 4, 6, 8, 10, 12]))
         print(stats.skew([1, 2**3, 3**3, 4**3, 5**3, 6**3]))
         print(stats.skew([1, 5, 10, 15, 20]))
         0.0
         -0.17039575849885127
         0.8568494780241179
         0.08448539032773617
In [18]: chi squared df2 = np.random.chisquare(2, size=10000)
         stats.skew(chi_squared_df2)
Out[18]: 1.984520065939756
In [19]: chi_squared_df5 = np.random.chisquare(5, size=10000)
         stats.skew(chi_squared_df5)
Out[19]: 1.2152057290912823
```

Out[20]: <matplotlib.legend.Legend at 0x7f734a4efdd8>



## **Hypothesis Testing**

```
In [21]: df = pd.read_csv('grades.csv')
```

In [22]: df.head()

Out[22]:

|   | student_id                                   | assignment1_grade | assignment1_submission           | assignment2_grade | а |
|---|--|-------------------|----------------------------------|-------------------|---|
| 0 | B73F2C11-70F0-<br>E37D-8B10-1D20AFED50B1     | 92.733946         | 2015-11-02<br>06:55:34.282000000 | 83.030552         | 2 |
| 1 | 98A0FAE0-<br>A19A-13D2-4BB5-<br>CFBFD94031D1 | 86.790821         | 2015-11-29<br>14:57:44.429000000 | 86.290821         | 2 |
| 2 | D0F62040-CEB0-904C-<br>F563-2F8620916C4E     | 85.512541         | 2016-01-09<br>05:36:02.389000000 | 85.512541         | 2 |
| 3 | FFDF2B2C-F514-<br>EF7F-6538-A6A53518E9DC     | 86.030665         | 2016-04-30<br>06:50:39.801000000 | 68.824532         | 2 |
| 4 | 5ECBEEB6-F1CE-80AE-<br>3164-E45E99473FB4     | 64.813800         | 2015-12-13<br>17:06:10.750000000 | 51.491040         | 2 |

```
In [23]: len(df)
Out[23]: 2315
```

```
In [24]: early = df[df['assignment1_submission'] <= '2015-12-31']
late = df[df['assignment1_submission'] > '2015-12-31']
```

```
In [25]: early.mean()
Out[25]: assignment1_grade 74.972741
assignment2_grade 67.252190
assignment3_grade 61.129050
assignment4_grade 54.157620
          assignment5_grade 48.634643
          assignment6_grade 43.838980
          dtype: float64
In [26]: late.mean()
Out[26]: assignment1_grade 74.017429
         assignment2_grade 66.370822
         assignment3_grade 60.023244
         assignment4_grade 54.058138
          assignment5_grade 48.599402
          assignment6_grade 43.844384
          dtype: float64
In [27]: from scipy import stats
          stats.ttest_ind?
In [28]: stats.ttest_ind(early['assignment1_grade'], late['assignment1_grade'])
Out[28]: Ttest indResult(statistic=1.400549944897566, pvalue=0.16148283016060577)
In [29]: stats.ttest_ind(early['assignment2_grade'], late['assignment2_grade'])
Out[29]: Ttest indResult(statistic=1.3239868220912567, pvalue=0.18563824610067967)
In [30]: stats.ttest ind(early['assignment3 grade'], late['assignment3 grade'])
Out[30]: Ttest indResult(statistic=1.7116160037010733, pvalue=0.087101516341556676)
```