	3. Analyze Data Data Science Playlist on YouTube
	Data Science Analyze Conce data is read into Python, a first step is to analyze the data with summary statistics. This is especially true if the data set is large.
	Summary statistics include the count, mean, standard deviation, maximum, minimum, and quartile information for the data columns.
	Generate Data Run the next cell to: Generate n linearly spaced values between 0 and n-1 with np.linspace(start,end,count)
In [1]:	 Draw random samples from a uniform distribution between 0 and 1 with np.random.rand(count) Draw random samples from a normal (Gaussian) distribution with np.random.normal(mean,std,count) Combine time, x, and y with a vertical stack np.vstack and transpose .T for column oriented data. Save CSV text file 03-data.csv with header time,x,y. import numpy as np np.random.seed(0) n = 1000 time = np.linspace(0,n-1,n) x = np.random.rand(n) y = np.random.normal(1,1,n) data = np.vstack((time,x,y)).T
	np.savetxt('03-data.csv',data,header='time,x,y',delimiter=',',comments='') -
In [2]:	Display Data Distributions The histogram is a preview of how to create graphics so that data can be evaluated visually. 04. Visualize shows how to create plots to analyze data. import matplotlib.pyplot as plt
	<pre>%matplotlib inline plt.hist(x,10,label='x') plt.hist(y,60,label='y',alpha=0.7) plt.ylabel('Count'); plt.legend() plt.show()</pre>
	80 - tim 60 - 40 - 20 -
	Data Analysis with numpy The np.loadtxt function reads the CSV data file 03-data.csv. Numpy calculates size (dimensions), mean (average), std (standard deviation), and median as summary statistics. If you don't specify the axis then numpy gives a statistic across both the
In [3]:	<pre>rows (axis=0) and columns (axis=1). import numpy as np data = np.loadtxt('03-data.csv',delimiter=',',skiprows=1) print('Dimension (rows,columns):') print(np.size(data,0),np.size(data,1))</pre>
	<pre>print('Average:') print(np.mean(data,axis=0)) print('Standard Deviation:') print(np.std(data,0)) print('Median:') print(np.median(data,0)) Dimension (rows,columns): 1000 3 Average:</pre>
	[4.99500000e+02 4.95921534e-01 1.02904418e+00] Standard Deviation: [288.67499026 0.29059884 0.96616883] Median: [4.99500000e+02 4.81322565e-01 1.03093469e+00]
	Analyze data 1. Calculate the mean, standard deviation, and median of x*y 2. Calculate the skew of x*y with the scipy.stats skew function.
<pre>In [4]: Out[4]:</pre>	<pre>print(np.mean(x*y), np.std(x*y), np.median(x*y)) from scipy.stats import skew skew(x*y) 0.5021305639783985 0.628181267249509 0.35459902459887327 1.1973986839553572</pre>
	Data Analysis with pandas
In [5]:	Pandas simplifies the data analysis with thedescribe() function that is a method of DataFrame that is created with pd.read_csv() . Note that the data file can either be a local file name or a web-address such as url='http://apmonitor.com/pdc/uploads/Main/tclab_data2.txt' data = pd.read_csv(url) data.describe()
Out[5]:	<pre>import pandas as pd data = pd.read_csv('03-data.csv') data.describe() time</pre>
	std 288.819436 0.290744 0.966652 min 0.000000 0.000546 -1.994613 25% 249.750000 0.247613 0.352914 50% 499.500000 0.481323 1.030935 75% 749.250000 0.737260 1.637332 max 999.000000 0.999809 4.170975
	Data Analysis with pandas-profiling Pandas Profiling is a data analysis tool for a more in-depth summary of the data than the descibe() function. Install the package with: pip installuser pandas-profiling[notebook]
In [6]:	<pre>jupyter nbextension enablepy widgetsnbextension You need to restart the Kernel before proceeding. The install only needs to run once. try: import pandas as pd from pandas_profiling import ProfileReport import os except:</pre>
	<pre>!pip installuser pandas-profiling !jupyter nbextension enablepy widgetsnbextension print('Restart the Kernel before proceeding') # import data url='http://apmonitor.com/pdc/uploads/Main/tclab_data2.txt' data = pd.read_csv(url)</pre>
In [7]:	After you install pandas-profiling and enable the widget extension, you can now import and analysis data. Some of the functions take a long time with a large data set. Two methods for dealing with large data sets are to: 1. Sub-sample the data sets such as with data = data[::10] to take every 10th row. 2. Use the minimal option to avoid the correlation and other analysis that is slow with large data sets. profile = ProfileReport(data, explorative=True, minimal=False)
In [8]:	The profile report can be saved as an interactive web-page. The web-page is saved to the current working directory that is displayed with os.getcwd(). profile.to_file('report.html') print('File report.html saved to '+os.getcwd())
	Summarize dataset: 100% 27/27 [00:01<00:00, 16.48it/s, Completed] Generate report structure: 100% 1/1 [00:01<00:00, 1.02s/it] Render HTML: 100% 1/1 [00:00<00:00, 3.28it/s] Export report to file: 100% 1/1 [00:00<00:00, 909.63it/s] File report.html saved to /home/curtis/classes/dynamic_optmization/data_science-master The profile report can also be viewed in the Jupyter Notebook.
	TCLab Activity
	Temperature Control Lab https://apmonitor.com/heat.htm is a significant depth is a signific
	Generate Data Set 1 Generate a file from the TCLab data with seconds (t), heater levels (Q1 and Q2), and temperatures (lab.T1 and lab.T2). Record data every second for 120 seconds and change the heater levels every 30 seconds to a random number between 0 and 80 with np.random.randint(). There is no need to change this program, only run it for 2 minutes to collect the data. If you do not have a TCLab device, read a data file 1 from an online link.
In [10]:	<pre>import tclab, time, csv import pandas as pd import numpy as np try: # connect to TCLab if available n = 120 with open('03-tclab1.csv', mode='w', newline='') as f: cw = csv.writer(f) cw.writerow(['Time','Q1','Q2','T1','T2'])</pre>
	<pre>with tclab.TCLab() as lab: print('t Q1 Q2 T1</pre>
	<pre>print(t,Q1,Q2,lab.T1,lab.T2)</pre>
	TCLab version 0.4.9 Arduino Leonardo connected on port /dev/ttyACM1 at 115200 baud. TCLab Firmware 2.0.1 Arduino Leonardo/Micro. t Q1 Q2 T1
	25 52 75 27.827 27.021 30 53 67 28.923 27.505 35 53 67 29.922 28.536 40 53 67 31.211 29.632 45 53 67 32.5 30.244 50 53 67 33.467 31.082 55 53 67 34.595 31.823 60 44 2 35.723 32.468 65 44 2 36.754 33.145
	70 44 2 37.656 33.563 75 44 2 38.623 33.982 80 44 2 39.332 34.079 85 44 2 40.234 34.401 90 60 66 40.879 34.724 95 60 66 41.523 34.724 100 60 66 42.458 35.11 105 60 66 43.103 35.271 110 60 66 43.779 35.852
In [11]:	TCLab disconnected successfully. Read Data Set 2 Use requests to download a sample TCLab data file for the analysis. It is saved as 03-tclab2.csv.
	<pre>import requests import os url = 'http://apmonitor.com/pdc/uploads/Main/tclab_data2.txt' r = requests.get(url) with open('03-tclab2.csv', 'wb') as f: f.write(r.content) print('File 03-tclab2.csv retrieved to current working directory: ') print(os.getcwd())</pre>
	File 03-tclab2.csv retrieved to current working directory: /home/curtis/classes/dynamic_optmization/data_science-master Data Analysis Read the files 03-tclab1.csv and 03-tclab2.csv and display summary statistics for each with data.describe() . Use the summary statistics to compare the number of samples and differences in average and standard deviation value for T1 and T2 .
<pre>In [31]: Out[31]:</pre>	<pre>data1 = pd.read_csv('03-tclab1.csv') data2 = pd.read_csv('03-tclab2.csv') data1 Time Q1 Q2 T1 T2 0 0 52 75 25.185 25.346</pre>
	1 1 52 75 25.120 25.346 2 2 52 75 25.120 25.056 3 3 52 75 25.088 25.410 4 4 52 75 25.120 25.088 115 60 66 44.714 36.271
	116 116 60 66 44.746 36.432 117 117 60 66 45.068 36.625 118 118 60 66 45.068 36.851 119 119 60 66 45.391 37.044
<pre>In [28]: Out[28]:</pre>	120 rows × 5 columns sum1 = data1.describe() sum2 = data2.describe() sum1 Time Q1 Q2 T1 T2
211	count 120.000000 120.000000 120.000000 120.000000 mean 59.500000 52.250000 52.500000 35.001350 31.313842 std 34.785054 5.697191 29.487214 6.496461 3.761548 min 0.000000 44.000000 25.088000 25.056000 25% 29.750000 50.000000 28.785500 27.762750
In [26]:	50% 59.500000 52.500000 66.500000 35.545500 32.387000 75% 89.250000 54.750000 69.000000 40.782250 34.434000 max 119.000000 60.000000 75.000000 45.391000 37.044000 print(sum1['T1']['count'] - sum2['T1']['count']) print(sum1['T2']['count'] - sum2['T2']['count'])
In [27]:	-481.0 -481.0 print(sum1['T1']['mean'] - sum2['T1']['mean']) print(sum1['T2']['mean'] - sum2['T2']['mean']) -25.96761838602329 -10.481965321686083
In [29]:	-10.481965321686083 print(sum1['T1']['std'] - sum2['T1']['std']) print(sum1['T2']['std'] - sum2['T2']['std'])

-8.452731722923627 -9.677721691051412

In [30]: profile = ProfileReport(data1, explorative=True, minimal=False)
 profile.to_file('report.html')

print('File report.html saved to '+os.getcwd())

T1 .

Use the pandas-profiling package to generate a data analysis report. View the distribution and correlation of the variables Q1 and

Summarize dataset: 100%| 27/27 [00:01<00:00, 21.00it/s, Completed]

Generate report structure: 100%| 1/1 [00:01<00:00, 1.01s/it]

Render HTML: 100%| 1/1 [00:00<00:00, 6.95it/s]

Export report to file: 100%| 1/1 [00:00<00:00, 1091.13it/s]

File report.html saved to /home/curtis/classes/dynamic_optmization/data_science-master