In [3]: In [14]:	<pre>import pandas as pd import numpy as np  import pandas as pd  numbers = [33 4 45 6 43 4 34 3 34 7 67 5 56 4]</pre>
Out[14]:	numbers = {23.4,45.6,43.4,34.2,34.7,67.5,56.4}  pop = pd.Series(list(numbers))  pop  0
Out[14].	1 67.5 2 34.7 3 23.4 4 56.4
In [15]:	5 43.4 6 45.6 dtype: float64  pop.name=" population of seven nation"
In [16]: Out[16]:	0 34.2
	1 67.5 2 34.7 3 23.4 4 56.4 5 43.4
	6 45.6 Name: population of seven nation, dtype: float64 # selecting elements
Out[17]: In [18]:	pop[0] 34.2 pop[4]
	56.4  pop.index=['kenya','uganda','zambia','rwanda','Tanzania','somalia','Djibouti']
In [20]:	pop kenya 34.2
	uganda 67.5 zambia 34.7 rwanda 23.4 Tanzania 56.4 somalia 43.4
	Djibouti 45.6 Name: population of seven nation, dtype: float64  #pandas indexing and conditional selection
000[22].	pop['zambia'] 34.7
Out[23]:	pop['kenya'] 34.2
Out[24]:	pop.iloc[0] 34.2
Out[25]:	<pre>pop.iloc[-1] 45.6  pop['kenya':'uganda']</pre>
Out[30]:	kenya 34.2 uganda 67.5 Name: population of seven nation, dtype: float64
<pre>In [31]: Out[31]:</pre>	#conditional selection "boolean arrays" pop>37  kenya False
	uganda True zambia False rwanda False Tanzania True somalia True
	Djibouti True Name: population of seven nation, dtype: bool pop<70
	kenya True uganda True zambia True rwanda True
	Tanzania True somalia True Djibouti True Name: population of seven nation, dtype: bool
	pop[pop>37]  uganda 67.5  Tanzania 56.4 somalia 43.4
	Djibouti 45.6 Name: population of seven nation, dtype: float64  pop.mean()
Out[34]:	43.6000000000001 pop.std()
Out[35]:	14.802589863488977  pop>pop.mean()-pop.std()/2
Out[38]:	kenya False uganda True zambia False rwanda False
	Tanzania True somalia True Djibouti True Name: population of seven nation, dtype: bool
	<pre>pop[pop&gt;pop.mean()-pop.std()/2]  uganda 67.5 Tanzania 56.4 somalia 43.4</pre>
	somalia 43.4 Djibouti 45.6 Name: population of seven nation, dtype: float64  #modifying series
In [41]:	pop['kenya']=67.9 pop
	kenya 67.9 uganda 67.5 zambia 34.7 rwanda 23.4 Tanzania 56.4
	somalia 43.4 Djibouti 45.6 Name: population of seven nation, dtype: float64
Out [43]:	pop[pop>37]=100  pop  kenya 100.0
	uganda 100.0 zambia 34.7 rwanda 23.4 Tanzania 100.0
	somalia 100.0 Djibouti 100.0 Name: population of seven nation, dtype: float64 #Data frames in pandas
" ·^1;	<pre>import pandas import pandas df=pd.DataFrame( {'employees':['mike','cynthia','sunshines','robin','kamau','ruger','otile','maina','Mejja','kiprotich'], 'salary':[75000,53000,34000,56000,34000,45000,45700,34000,23000,34600], 'job description':['manager','assistant','clerk','secretary','gateman','driver','store keeper ','intern','consultant','delivery'],</pre>
	'rating':['good','good','fair','fair','good','excellent','good','fair'] } ,columns=['employees','salary','job description','rating'])
In [41]: Out[41]:	employees salary job description rating
	<ul> <li>mike 75000 manager good</li> <li>cynthia 53000 assistant good</li> <li>sunshines 34000 clerk fair</li> </ul>
	3         robin         56000         secretary         fair           4         kamau         34000         gateman         fair           5         ruger         45000         driver         good
	<ul> <li>otile 45700 store keeper excellent</li> <li>maina 34000 intern good</li> <li>Mejja 23000 consultant good</li> </ul>
In [42]:	9 kiprotich 34600 delivery fair  df.iloc[2]
Out[42]:	employees sunshines salary 34000 job description clerk rating fair
In [43]:	Name: 2, dtype: object  df['employees']  0 mike
Out[43]:	<pre>cynthia sunshines robin kamau</pre>
	5 ruger 6 otile 7 maina 8 Mejja 9 kiprotich
In [44]:	Name: employees, dtype: object  df.info() <class 'pandas.core.frame.dataframe'=""></class>
	RangeIndex: 10 entries, 0 to 9 Data columns (total 4 columns):
	# Column Non-Null Count Dtype
	# Column Non-Null Count Dtype  O employees 10 non-null object  salary 10 non-null int64  job description 10 non-null object  rating 10 non-null object
In [45]:	# Column Non-Null Count Dtype  0 employees 10 non-null object 1 salary 10 non-null int64 2 job description 10 non-null object 3 rating 10 non-null object dtypes: int64(1), object(3) memory usage: 448.0+ bytes  df.shape
<pre>In [45]: Out[45]: In [46]:</pre>	# Column Non-Null Count Dtype
In [45]: Out[45]:	# Column Non-Null Count Dtype
<pre>In [45]: Out[45]: In [46]:</pre>	# Column Non-Null Count Dtype  0 employees 10 non-null object 1 salary 10 non-null object 2 job description 10 non-null object dtypes: int64(1), object(3) memory usage: 448.0+ bytes  df.shape  (10, 4)  count 10.00000
<pre>In [45]: Out[45]: In [46]:</pre>	# Column Non-Null Count btype  0 employees 10 non-null object 1 salary 10 non-null int64 2 job description 10 non-null object 3 rating 10 non-null object dtypes: int64(1), object(3) memory usage: 448.0+ bytes  df.shape  (10, 4)  df.describe()  salary  count 10.00000  mean 43430.00000  std 14966.633259  min 23000.000000
<pre>In [45]: Out[45]: In [46]: Out[46]: In [47]:</pre>	# Column Non-Null Count or Stype    employees
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]:</pre>	# Column Non-Null Count Dyes    Non-Null Count Dyes   Non-Null Cou
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]: Out[48]:</pre>	# Column Non-Wall count Dtype 0 employees 10 non-null object 1 salary 10 non-null object 2 job description 10 non-null object 3 rating 10 non-null object 44.8 ** bytes   ##################################
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]: Out[48]:</pre>	
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]: Out[48]:  Out[49]:</pre>	# Column Non-Null Count Otype
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]: Out[48]:  In [50]: In [51]:</pre>	# Column Non-Wall Count bype
<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]:  Out[48]:  In [50]: In [51]: Out[51]:</pre>	
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<pre>In [45]: Out[45]: In [46]: Out[46]:  In [47]: Out[47]: In [48]:  Out[48]:  In [50]: In [51]: Out[51]:  Out[53]:</pre>	
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