

# PandaBasics101

September 10, 2018

## 1 Panda Basics 101 : DataFrames

DataFrames (DF) : Creation class `pandas.DataFrame(data=None, index=None, columns=None, dtype=None, copy=False)`

Two-dimensional size-mutable, potentially heterogeneous tabular data structure with labeled axes (rows and columns). Arithmetic operations align on both row and column labels. Can be thought of as a dict-like container for Series objects. The primary pandas data structure.

```
In [1]: #basic imports
import pandas as pd
import numpy as np
```

### 1.1 Create DF From List

```
In [2]: df = pd.DataFrame([1,2,3,4,5])
df
```

```
Out[2]:    0
0    1
1    2
2    3
3    4
4    5
```

### 1.2 Create DF From List of Lists

```
In [3]: df = pd.DataFrame([[1,2,3,4,5],[10,20,30,40,50],[100,120,130,140,150]])
df
```

```
Out[3]:    0    1    2    3    4
0    1    2    3    4    5
1   10   20   30   40   50
2  100  120  130  140  150
```

```
In [4]: # provide column labels- must be enough
df = pd.DataFrame([[1,2,3,4,5],[10,20,30,40,50],[100,120,130,140,150]],
                  columns=list("ABCDE"))
df
```

```
Out [4]:
```

	A	B	C	D	E
0	1	2	3	4	5
1	10	20	30	40	50
2	100	120	130	140	150

### 1.3 Create DF From Dict

```
In [5]: #keys = column names - if val = array, must be same size
df = pd.DataFrame( {"foo": [1,2,3,4], "bar": [9,8,7,6], 'baz': [12,13,14,15] } )
df
```

```
Out [5]:
```

	foo	bar	baz
0	1	9	12
1	2	8	13
2	3	7	14
3	4	6	15

### 1.4 Create DF using random decimals and integers (good for experimenting)

### 1.5 Generate Random Decimals in range -1 to 1 using random.randn(rows,cols)

```
In [6]: df = pd.DataFrame(np.random.randn(5, 4), columns=list('ABCD'))
df
```

```
Out [6]:
```

	A	B	C	D
0	0.875130	-0.567622	-1.184339	0.365110
1	-0.886746	0.225891	-1.463829	1.227129
2	-0.615083	1.105010	-0.284913	1.926585
3	1.421741	-0.020012	0.519023	-0.619867
4	-0.570243	1.274669	-0.128316	-1.354610

### 1.6 Generate Random Integers using: randint(low,hi,size=(rows,cols))

```
In [7]: df = pd.DataFrame(np.random.randint(low=0, high=10, size=(5, 5)),
                           columns=['a', 'b', 'c', 'd', 'e'])
df
```

```
Out [7]:
```

	a	b	c	d	e
0	9	5	0	6	5
1	2	5	8	2	0
2	0	0	4	2	4
3	2	3	6	9	4
4	9	2	0	4	0

### 1.7 Create DF From CSV File

```
In [8]: df = pd.read_csv('worldcup.csv')
df
```

```

Out [8]:   WorldCup  year      location      first      second      third \
0    wc1930  1930      Uruguay      Uruguay      Argentina      USA
1    wc1934  1934        Italy      Italy    Czechoslovakia      Germany
2    wc1938  1938        France      Italy      Hungary      Brazil
3    wc1950  1950        Brazil      Uruguay      Brazil      Sweden
4    wc1954  1954    Switzerland    GermanyFR      Hungary      Austria
5    wc1958  1958        Sweden      Brazil      Sweden      France
6    wc1962  1962        Chile      Brazil    Czechoslovakia      Chile
7    wc1966  1966      England      England      GermanyFR      Portugal
8    wc1970  1970      Mexico      Brazil      Italy      GermanyFR
9    wc1974  1974      Germany      GermanyFR      Netherlands      Poland
10   wc1978  1978    Argentina      Argentina      Netherlands      Brazil
11   wc1982  1982      Spain      Italy      GermanyFR      Poland
12   wc1986  1986      Mexico      Argentina      GermanyFR      France
13   wc1990  1990      Italy      GermanyFR      Argentina      Italy
14   wc1994  1994      USA      Brazil      Italy      Sweden
15   wc1998  1998      France      France      Brazil      Croatia
16   wc2002  2002    Korea_Japan      Brazil      Germany      Turkey
17   wc2006  2006      Germany      Italy      France      Germany
18   wc2010  2010    SouthAfrica      Spain      Netherlands      Germany
19   wc2014  2014      Brazil      Germany      Argentina    Netherlands
20   wc2018  2018      Russia      France      Croatia      Belgium

```

```

      fourth  goalsScored  matchesPlayed  attendance
0    Yugoslavia          70           18      590549
1    Austria           70           17      363000
2    Sweden            84           18      375000
3    Spain             88           22     1045246
4    Uruguay           14           26      768607
5    GermanyFR        126           35      819810
6    Yugoslavia        89           32      893172
7    Soviet_Union      89           32     1563135
8    Uruguay           95           32     1603975
9    Brazil            97           38     1865753
10   Italy            102           38     1545791
11   France           146           52     2109723
12   Belgium          132           52     2394031
13   England          115           52     2516215
14   Bulgaria          141           52     3587538
15   Netherlands       171           64     2785100
16   KoreaRepublic     161           64     2705197
17   Portugal          147           64     3359439
18   Uruguay           145           64     3178856
19   Brazil            171           64     3386810
20   England           169           64     3430000

```

## 2 Creating New Cols/Rows

### 2.1 New Column Creation based on existing column(s)

```
In [9]: df['goalsPerMatch'] = df['goalsScored']/df['matchesPlayed']
df
```

```
Out [9]:
```

	WorldCup	year	location	first	second	third \
0	wc1930	1930	Uruguay	Uruguay	Argentina	USA
1	wc1934	1934	Italy	Italy	Czechoslovakia	Germany
2	wc1938	1938	France	Italy	Hungary	Brazil
3	wc1950	1950	Brazil	Uruguay	Brazil	Sweden
4	wc1954	1954	Switzerland	GermanyFR	Hungary	Austria
5	wc1958	1958	Sweden	Brazil	Sweden	France
6	wc1962	1962	Chile	Brazil	Czechoslovakia	Chile
7	wc1966	1966	England	England	GermanyFR	Portugal
8	wc1970	1970	Mexico	Brazil	Italy	GermanyFR
9	wc1974	1974	Germany	GermanyFR	Netherlands	Poland
10	wc1978	1978	Argentina	Argentina	Netherlands	Brazil
11	wc1982	1982	Spain	Italy	GermanyFR	Poland
12	wc1986	1986	Mexico	Argentina	GermanyFR	France
13	wc1990	1990	Italy	GermanyFR	Argentina	Italy
14	wc1994	1994	USA	Brazil	Italy	Sweden
15	wc1998	1998	France	France	Brazil	Croatia
16	wc2002	2002	Korea_Japan	Brazil	Germany	Turkey
17	wc2006	2006	Germany	Italy	France	Germany
18	wc2010	2010	SouthAfrica	Spain	Netherlands	Germany
19	wc2014	2014	Brazil	Germany	Argentina	Netherlands
20	wc2018	2018	Russia	France	Croatia	Belgium

	fourth	goalsScored	matchesPlayed	attendance	goalsPerMatch
0	Yugoslavia	70	18	590549	3.888889
1	Austria	70	17	363000	4.117647
2	Sweden	84	18	375000	4.666667
3	Spain	88	22	1045246	4.000000
4	Uruguay	14	26	768607	0.538462
5	GermanyFR	126	35	819810	3.600000
6	Yugoslavia	89	32	893172	2.781250
7	Soviet_Union	89	32	1563135	2.781250
8	Uruguay	95	32	1603975	2.968750
9	Brazil	97	38	1865753	2.552632
10	Italy	102	38	1545791	2.684211
11	France	146	52	2109723	2.807692
12	Belgium	132	52	2394031	2.538462
13	England	115	52	2516215	2.211538
14	Bulgaria	141	52	3587538	2.711538
15	Netherlands	171	64	2785100	2.671875
16	KoreaRepublic	161	64	2705197	2.515625
17	Portugal	147	64	3359439	2.296875

18	Uruguay	145	64	3178856	2.265625
19	Brazil	171	64	3386810	2.671875
20	England	169	64	3430000	2.640625

```
In [10]: # sort on this value - creates a copy
df2 = df.sort_values(by='goalsPerMatch', ascending=False)
df2
```

```
Out[10]:
```

	WorldCup	year	location	first	second	third \
2	wc1938	1938	France	Italy	Hungary	Brazil
1	wc1934	1934	Italy	Italy	Czechoslovakia	Germany
3	wc1950	1950	Brazil	Uruguay	Brazil	Sweden
0	wc1930	1930	Uruguay	Uruguay	Argentina	USA
5	wc1958	1958	Sweden	Brazil	Sweden	France
8	wc1970	1970	Mexico	Brazil	Italy	GermanyFR
11	wc1982	1982	Spain	Italy	GermanyFR	Poland
6	wc1962	1962	Chile	Brazil	Czechoslovakia	Chile
7	wc1966	1966	England	England	GermanyFR	Portugal
14	wc1994	1994	USA	Brazil	Italy	Sweden
10	wc1978	1978	Argentina	Argentina	Netherlands	Brazil
15	wc1998	1998	France	France	Brazil	Croatia
19	wc2014	2014	Brazil	Germany	Argentina	Netherlands
20	wc2018	2018	Russia	France	Croatia	Belgium
9	wc1974	1974	Germany	GermanyFR	Netherlands	Poland
12	wc1986	1986	Mexico	Argentina	GermanyFR	France
16	wc2002	2002	Korea_Japan	Brazil	Germany	Turkey
17	wc2006	2006	Germany	Italy	France	Germany
18	wc2010	2010	SouthAfrica	Spain	Netherlands	Germany
13	wc1990	1990	Italy	GermanyFR	Argentina	Italy
4	wc1954	1954	Switzerland	GermanyFR	Hungary	Austria

	fourth	goalsScored	matchesPlayed	attendance	goalsPerMatch
2	Sweden	84	18	375000	4.666667
1	Austria	70	17	363000	4.117647
3	Spain	88	22	1045246	4.000000
0	Yugoslavia	70	18	590549	3.888889
5	GermanyFR	126	35	819810	3.600000
8	Uruguay	95	32	1603975	2.968750
11	France	146	52	2109723	2.807692
6	Yugoslavia	89	32	893172	2.781250
7	Soviet_Union	89	32	1563135	2.781250
14	Bulgaria	141	52	3587538	2.711538
10	Italy	102	38	1545791	2.684211
15	Netherlands	171	64	2785100	2.671875
19	Brazil	171	64	3386810	2.671875
20	England	169	64	3430000	2.640625
9	Brazil	97	38	1865753	2.552632
12	Belgium	132	52	2394031	2.538462

16	KoreaRepublic	161	64	2705197	2.515625
17	Portugal	147	64	3359439	2.296875
18	Uruguay	145	64	3178856	2.265625
13	England	115	52	2516215	2.211538
4	Uruguay	14	26	768607	0.538462

## 2.2 Applying functions to DataFrames for each row or column

When we `df.apply(fn, axis= )` a Python function to a DataFrame, each row (`axis=0`) or col (`axis=1`) is passed as a parameter to the function and a new DataFrame is created.

axis confusion

- `axis=1`: All the COLS in each Row
- `axis=0`: all the ROWS in each COL

```
In [11]: df = pd.DataFrame([[1,1,1,1], [100,200,300,400], [1000,2000,3000,4000]],
                           columns=list('ABCD'))
```

df

```
Out[11]:
```

	A	B	C	D
0	1	1	1	1
1	100	200	300	400
2	1000	2000	3000	4000

```
In [12]: #axis=1: All the COLS in each Row
         #axis=0: all the ROWS in each COL
         def addAll(z):
             #the DF will pass either a row or column (a Series) to this function
             return z.sum()

         df2 = df.apply(addAll, axis=1)
         print (df2)

         #now with axis=0 All ROWS for each COL
         #note: the column names are now the row indexes
         df3 = df.apply(addAll, axis=0)
         print (df3)
```

```
0      4
1    1000
2   10000
dtype: int64
A     1101
B     2201
C     3301
D     4401
dtype: int64
```

## 2.3 Using Boolean DataFrame to select rows from a DataFrame

Creating a boolean index (a single column of true/false) is useful when you want to select subset of rows from your DataFrame. The strategy is: - create new DF based on logical operators over the original DF - use the new boolean DF to select only those rows == True

```
In [13]: df = pd.DataFrame([ [2,4,6,8], [6,12,44,67], [6,6,6,26], [3,4,5,6]],
                           columns=list("abcd"))
df
```

```
Out[13]:
```

	a	b	c	d
0	2	4	6	8
1	6	12	44	67
2	6	6	6	26
3	3	4	5	6

```
In [14]: #simple boolean expression
dbool = df['a'] >= 6
dbool
```

```
Out[14]:
```

0	False
1	True
2	True
3	False

Name: a, dtype: bool

```
In [15]: #compound boolean condition - needs parens around each expression
dbool = (df['a'] >= 6) & (df['d'] > df['a']*2)
dbool
```

```
Out[15]:
```

0	False
1	True
2	True
3	False

dtype: bool

```
In [16]: #apply boolean to original DF
df3 = df[dbool]
df3
```

```
Out[16]:
```

	a	b	c	d
1	6	12	44	67
2	6	6	6	26

## 3 Panda Selection and Slicing from Existing DF

```
In [17]: # set up a dataframe
df = pd.DataFrame(np.random.randint(0,100,size=(5, 5)), columns=list('abcde'))
df
```

```
Out[17]:
```

	a	b	c	d	e
0	64	70	98	63	7
1	46	88	30	62	11
2	3	93	44	89	32
3	38	55	72	54	47
4	84	22	50	84	74

### 3.1 Select Column or Columns from DF

Index refers to entire column (unlike Python Dict) Note: When selecting a single column or row, a Series is returned

#### 3.1.1 Select Single Column (as Series)

```
In [18]: df2 = df['b']
          print (df2)
          print (type(df2))
```

0	70
1	88
2	93
3	55
4	22

Name: b, dtype: int32  
<class 'pandas.core.series.Series'>

#### 3.1.2 Select Multiple Columns (as DataFrame) - pass List

```
In [19]: df3 = df[ ['a','c','e'] ]
          df3
```

```
Out[19]:
```

	a	c	e
0	64	98	7
1	46	30	11
2	3	44	32
3	38	72	47
4	84	50	74

#### 3.1.3 Selecting Single row and single data item using .iloc

```
In [20]: # show our DF
          df
```

```
Out[20]:
```

	a	b	c	d	e
0	64	70	98	63	7
1	46	88	30	62	11
2	3	93	44	89	32
3	38	55	72	54	47
4	84	22	50	84	74



### 3.1.4 df.iloc expects a numeric index - refers to row

Single integer returns the row

```
In [21]: #single index value -> row  
df.iloc[2]
```

```
Out[21]: a      3  
        b     93  
        c     44  
        d     89  
        e     32  
        Name: 2, dtype: int32
```

### 3.1.5 df.iloc with a list of labels -- returns multiple rows

```
In [22]: df.iloc[ [1,3,4]]
```

```
Out[22]:      a  b  c  d  e  
1  46  88  30  62  11  
3  38  55  72  54  47  
4  84  22  50  84  74
```

### 3.1.6 Single Data item with .iloc [row,col] -> yields single value

```
In [23]: df.iloc[1,2]
```

```
Out[23]: 30
```

### 3.1.7 Rows and Columns with df.iloc and numeric slice ranges [1:3,2:4]

```
In [24]: # use : for all rows  
df.iloc[:,2:5]
```

```
Out[24]:      c  d  e  
0  98  63  7  
1  30  62  11  
2  44  89  32  
3  72  54  47  
4  50  84  74
```

```
In [25]: # range over rows and cols with integer indexes  
df.iloc[2:4,2:4]
```

```
Out[25]:      c  d  
2  44  89  
3  72  54
```

### 3.2 Selecting with Labels using df.loc

```
In [26]: # show df
df
```

```
Out[26]:
```

	a	b	c	d	e
0	64	70	98	63	7
1	46	88	30	62	11
2	3	93	44	89	32
3	38	55	72	54	47
4	84	22	50	84	74

```
In [27]: # one parameter gives us the entire row. (note: row label is an integer)
# Sanity check: df[2]      -> column 2 ( if such a label exists)
#              df.loc[2] -> row with 'label 2' - not the second in series
print (df.loc[2])
```

```
a      3
b     93
c     44
d     89
e     32
Name: 2, dtype: int32
```

```
In [28]: # two parameters give single element
df.loc[2, 'b']
```

```
Out[28]: 93
```

```
In [29]: # range of values
df.loc[2:4, 'c':'e']
```

```
Out[29]:
```

	c	d	e
2	44	89	32
3	72	54	47
4	50	84	74

### 3.3 Select based on column value using in or .isin

Pandas isin() method is used to filter data frames. isin() method helps in selecting rows with having a single or multiple values in a particular column.

```
In [30]: df = pd.read_csv('worldcup.csv')
df
```

```
Out[30]:
```

	WorldCup	year	location	first	second	third \
0	wc1930	1930	Uruguay	Uruguay	Argentina	USA
1	wc1934	1934	Italy	Italy	Czechoslovakia	Germany
2	wc1938	1938	France	Italy	Hungary	Brazil

3	wc1950	1950	Brazil	Uruguay	Brazil	Sweden
4	wc1954	1954	Switzerland	GermanyFR	Hungary	Austria
5	wc1958	1958	Sweden	Brazil	Sweden	France
6	wc1962	1962	Chile	Brazil	Czechoslovakia	Chile
7	wc1966	1966	England	England	GermanyFR	Portugal
8	wc1970	1970	Mexico	Brazil	Italy	GermanyFR
9	wc1974	1974	Germany	GermanyFR	Netherlands	Poland
10	wc1978	1978	Argentina	Argentina	Netherlands	Brazil
11	wc1982	1982	Spain	Italy	GermanyFR	Poland
12	wc1986	1986	Mexico	Argentina	GermanyFR	France
13	wc1990	1990	Italy	GermanyFR	Argentina	Italy
14	wc1994	1994	USA	Brazil	Italy	Sweden
15	wc1998	1998	France	France	Brazil	Croatia
16	wc2002	2002	Korea_Japan	Brazil	Germany	Turkey
17	wc2006	2006	Germany	Italy	France	Germany
18	wc2010	2010	SouthAfrica	Spain	Netherlands	Germany
19	wc2014	2014	Brazil	Germany	Argentina	Netherlands
20	wc2018	2018	Russia	France	Croatia	Belgium

	fourth	goalsScored	matchesPlayed	attendance
0	Yugoslavia	70	18	590549
1	Austria	70	17	363000
2	Sweden	84	18	375000
3	Spain	88	22	1045246
4	Uruguay	14	26	768607
5	GermanyFR	126	35	819810
6	Yugoslavia	89	32	893172
7	Soviet_Union	89	32	1563135
8	Uruguay	95	32	1603975
9	Brazil	97	38	1865753
10	Italy	102	38	1545791
11	France	146	52	2109723
12	Belgium	132	52	2394031
13	England	115	52	2516215
14	Bulgaria	141	52	3587538
15	Netherlands	171	64	2785100
16	KoreaRepublic	161	64	2705197
17	Portugal	147	64	3359439
18	Uruguay	145	64	3178856
19	Brazil	171	64	3386810
20	England	169	64	3430000

```
In [31]: #create a bool series
newdf = df['first'].isin(['Brazil'])
#use to filter
df[newdf]
```

```
Out[31]:   WorldCup  year  location  first  second  third \
5    wc1958  1958    Sweden  Brazil    Sweden  France
```

6	wc1962	1962	Chile	Brazil	Czechoslovakia	Chile
8	wc1970	1970	Mexico	Brazil	Italy	GermanyFR
14	wc1994	1994	USA	Brazil	Italy	Sweden
16	wc2002	2002	Korea_Japan	Brazil	Germany	Turkey

	fourth	goalsScored	matchesPlayed	attendance
5	GermanyFR	126	35	819810
6	Yugoslavia	89	32	893172
8	Uruguay	95	32	1603975
14	Bulgaria	141	52	3587538
16	KoreaRepublic	161	64	2705197

```
In [32]: newdf = df['first'].isin(['Brazil', 'France'])
         #use to filter
         df[newdf]
```

```
Out [32]:
```

	WorldCup	year	location	first	second	third \
5	wc1958	1958	Sweden	Brazil	Sweden	France
6	wc1962	1962	Chile	Brazil	Czechoslovakia	Chile
8	wc1970	1970	Mexico	Brazil	Italy	GermanyFR
14	wc1994	1994	USA	Brazil	Italy	Sweden
15	wc1998	1998	France	France	Brazil	Croatia
16	wc2002	2002	Korea_Japan	Brazil	Germany	Turkey
20	wc2018	2018	Russia	France	Croatia	Belgium

	fourth	goalsScored	matchesPlayed	attendance
5	GermanyFR	126	35	819810
6	Yugoslavia	89	32	893172
8	Uruguay	95	32	1603975
14	Bulgaria	141	52	3587538
15	Netherlands	171	64	2785100
16	KoreaRepublic	161	64	2705197
20	England	169	64	3430000

## 4 Get and set single values from DataFrame with .at and .iat

### 4.1 Get/Set with .at (using labels - not indexes)

```
In [33]: # set up a dataframe
         df = pd.DataFrame(np.random.randint(0,100,size=(5, 5)), columns=list('abcde'))
         df
```

```
Out [33]:
```

	a	b	c	d	e
0	51	27	17	3	87
1	88	47	64	66	76
2	88	42	70	54	95
3	32	12	55	20	5
4	77	76	54	8	3

```
In [34]: # get value
df.at[1,'c']
```

```
Out[34]: 64
```

```
In [35]: # set value
df.at[1,'c'] = 1002
df
```

```
Out[35]:
```

	a	b	c	d	e
0	51	27	17	3	87
1	88	47	1002	66	76
2	88	42	70	54	95
3	32	12	55	20	5
4	77	76	54	8	3

## 4.2 Get/Set with .iat (using integer indexes)

```
In [36]: df.iat[3,3]
```

```
Out[36]: 20
```

```
In [37]: df.iat[3,3] = 3333
df
```

```
Out[37]:
```

	a	b	c	d	e
0	51	27	17	3	87
1	88	47	1002	66	76
2	88	42	70	54	95
3	32	12	55	3333	5
4	77	76	54	8	3

## 5 Sorting

```
In [38]: #set up
df = pd.DataFrame({
    'col1' : ['A', 'A', 'B', np.nan, 'D', 'C'],
    'col2' : [2, 1, 9, 8, 7, 4],
    'col3' : [0, 1, 9, 4, 2, 3],
})
df
```

```
Out[38]:
```

	col1	col2	col3
0	A	2	0
1	A	1	1
2	B	9	9
3	NaN	8	4
4	D	7	2
5	C	4	3

```
In [39]: df.sort_values(by=['col1'])
```

```
Out[39]:
```

	col1	col2	col3
0	A	2	0
1	A	1	1
2	B	9	9
5	C	4	3
4	D	7	2
3	NaN	8	4

```
In [40]: #sort multiple columns  
df.sort_values(by=['col1', 'col2'])
```

```
Out[40]:
```

	col1	col2	col3
1	A	1	1
0	A	2	0
2	B	9	9
5	C	4	3
4	D	7	2
3	NaN	8	4

```
In [41]: #sort multiple columns - DESCENDING  
df.sort_values(by=['col1', 'col2'], ascending=False)
```

```
Out[41]:
```

	col1	col2	col3
4	D	7	2
5	C	4	3
2	B	9	9
0	A	2	0
1	A	1	1
3	NaN	8	4

```
In [42]: # Put NAs first  
df.sort_values(by=['col1', 'col2'], ascending=False, na_position='first')
```

```
Out[42]:
```

	col1	col2	col3
3	NaN	8	4
4	D	7	2
5	C	4	3
2	B	9	9
0	A	2	0
1	A	1	1

## 6 Dropping rows and cols with Unknown NaN values

DataFrame.dropna(axis=0, how='any', thresh=None, subset=None, inplace=False)

```
In [43]: #drop rows with NaN  
df.dropna()
```

```
Out[43]:
```

	col1	col2	col3
0	A	2	0
1	A	1	1
2	B	9	9
4	D	7	2
5	C	4	3

```
In [44]: #drop columns
df.dropna(axis=1)
```

```
Out[44]:
```

	col2	col3
0	2	0
1	1	1
2	9	9
3	8	4
4	7	2
5	4	3

## 7 GroupBy

When using groupby you get an 'Object' with methods you can use

```
In [45]: # Create dataframe from dict: default is keys are column headers
raw_data = {'regiment': ['Nighthawks', 'Nighthawks', 'Nighthawks', 'Nighthawks', 'Dragoons', 'Dragoons', 'Dragoons', 'Dragoons', 'Scouts', 'Scouts', 'Scouts', 'Scouts'],
            'company': ['1st', '1st', '2nd', '2nd', '1st', '1st', '2nd', '2nd', '1st', '1st', '2nd', '2nd'],
            'name': ['Miller', 'Jacobson', 'Ali', 'Milner', 'Cooze', 'Jacon', 'Ryaner', 'Sone', 'Sloan', 'Piger', 'Riani', 'Ali'],
            'preTestScore': [4, 24, 31, 2, 3, 4, 24, 31, 2, 3, 2, 3],
            'postTestScore': [22, 94, 57, 62, 70, 25, 94, 57, 62, 70, 62, 70]}
df = pd.DataFrame(raw_data, columns = ['regiment', 'company', 'name', 'preTestScore', 'postTestScore'])
df
```

```
Out[45]:
```

	regiment	company	name	preTestScore	postTestScore
0	Nighthawks	1st	Miller	4	22
1	Nighthawks	1st	Jacobson	24	94
2	Nighthawks	2nd	Ali	31	57
3	Nighthawks	2nd	Milner	2	62
4	Dragoons	1st	Cooze	3	70
5	Dragoons	1st	Jacon	4	25
6	Dragoons	2nd	Ryaner	24	94
7	Dragoons	2nd	Sone	31	57
8	Scouts	1st	Sloan	2	62
9	Scouts	1st	Piger	3	70
10	Scouts	2nd	Riani	2	62
11	Scouts	2nd	Ali	3	70

```
In [46]: # Create a groupby variable that groups our DataFrame by regiment
groupby_regiment = df.groupby(['regiment'])
```

## 7.1 Get size of your Groups

```
In [47]: groupby_regiment.size()
```

```
Out[47]: regiment
         Dragoons      4
         Nighthawks    4
         Scouts        4
         dtype: int64
```

## 7.2 Compute sums and mean of groups : use .sum() and .mean()

```
In [48]: groupby_regiment.mean()
```

```
Out[48]:
```

	preTestScore	postTestScore
regiment		
Dragoons	15.50	61.50
Nighthawks	15.25	58.75
Scouts	2.50	66.00

### 7.2.1 add\_prefix('string') when you sum() or mean() over columns

```
In [49]: groupby_regiment.mean().add_prefix('mean_')
```

```
Out[49]:
```

	mean_preTestScore	mean_postTestScore
regiment		
Dragoons	15.50	61.50
Nighthawks	15.25	58.75
Scouts	2.50	66.00

## 7.3 Iterating over Groups

When iterating you can get the name of the group and a dataframe for the group

```
In [50]: for name, group in groupby_regiment:
         print ("Group name = ", name)
         print (group)
```

```
Group name = Dragoons
```

	regiment	company	name	preTestScore	postTestScore
4	Dragoons	1st	Cooze	3	70
5	Dragoons	1st	Jacon	4	25
6	Dragoons	2nd	Ryaner	24	94
7	Dragoons	2nd	Sone	31	57

```
Group name = Nighthawks
```

	regiment	company	name	preTestScore	postTestScore
0	Nighthawks	1st	Miller	4	22
1	Nighthawks	1st	Jacobson	24	94
2	Nighthawks	2nd	Ali	31	57
3	Nighthawks	2nd	Milner	2	62



```

Group name = Scouts
   regiment company   name preTestScore postTestScore
8   Scouts    1st Sloan           2           62
9   Scouts    1st Piger           3           70
10  Scouts    2nd Riani           2           62
11  Scouts    2nd Ali             3           70

```

## 8 Ploting

Easy to plot one column vs another Include `'%matplotlib inline'` to get Jupyter to plot  
pandas vs matplotlib

Under the hood, pandas plots graphs with the matplotlib library. This is usually pretty convenient since it allows you to just `.plot` your graphs, but since matplotlib is kind of a train wreck pandas inherits that confusion. Which `.plot` do I use?

When you use `.plot` on a dataframe, you sometimes pass things to it and sometimes you don't.

```

.plot plots the index against every column
.plot(x='col1') plots against a single specific column
.plot(x='col1', y='col2') plots one specific column against another specific column

```

```

In [51]: import matplotlib.pyplot as plt
         %matplotlib inline
         #load data
         df = pd.read_csv('worldcup.csv')
         df.head()

```

```

Out[51]:  WorldCup  year    location    first    second    third    fourth \
0   wc1930  1930    Uruguay    Uruguay    Argentina    USA    Yugoslavia
1   wc1934  1934     Italy    Italy    Czechoslovakia    Germany    Austria
2   wc1938  1938     France    Italy    Hungary    Brazil    Sweden
3   wc1950  1950     Brazil    Uruguay    Brazil    Sweden    Spain
4   wc1954  1954  Switzerland    GermanyFR    Hungary    Austria    Uruguay

         goalsScored  matchesPlayed  attendance
0                70             18    590549
1                70             17    363000
2                84             18    375000
3                88             22   1045246
4                14             26    768607

```

```

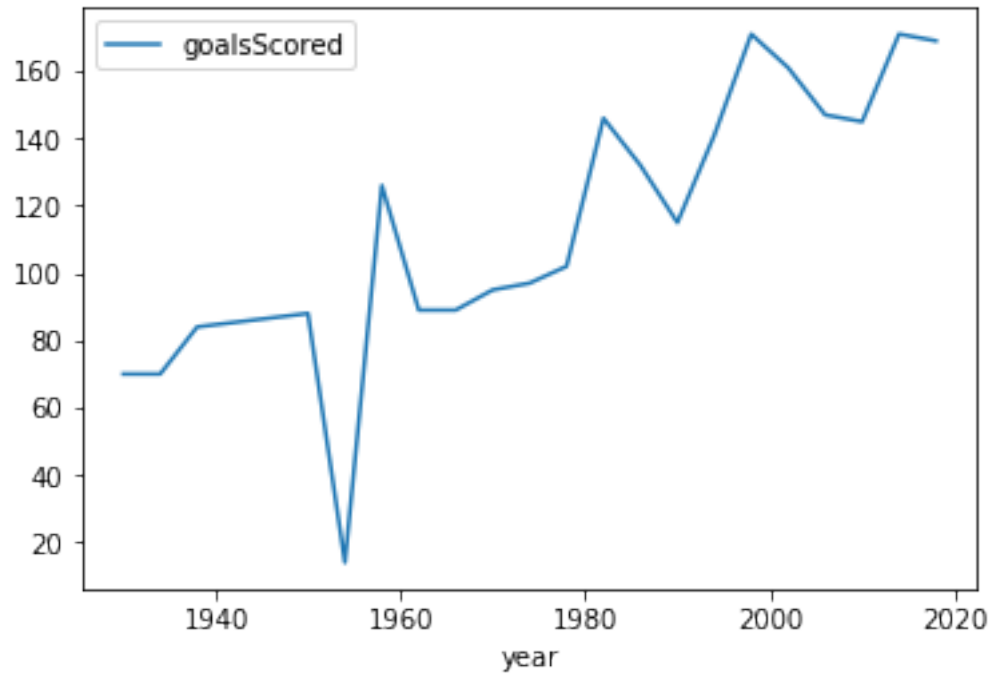
In [52]: df.plot(y='goalsScored', x='year')

```

```

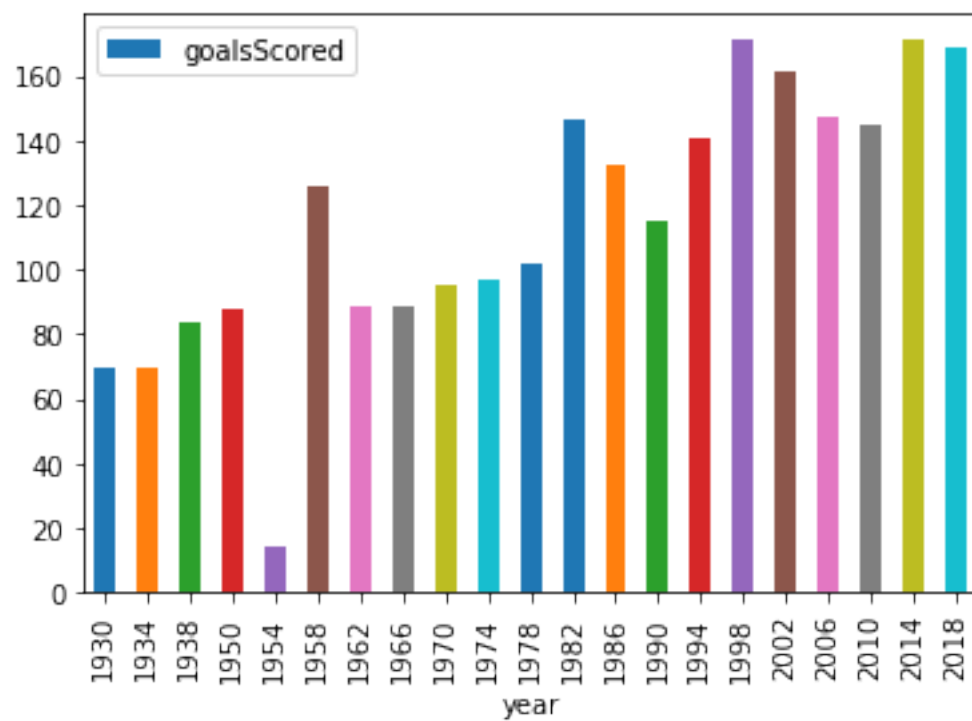
Out[52]: <matplotlib.axes._subplots.AxesSubplot at 0x2123ad05780>

```



In [53]: `df.plot(y='goalsScored', x='year', kind='bar')`

Out[53]: `<matplotlib.axes._subplots.AxesSubplot at 0x2123b015160>`



### 8.0.1 Groupby and plots

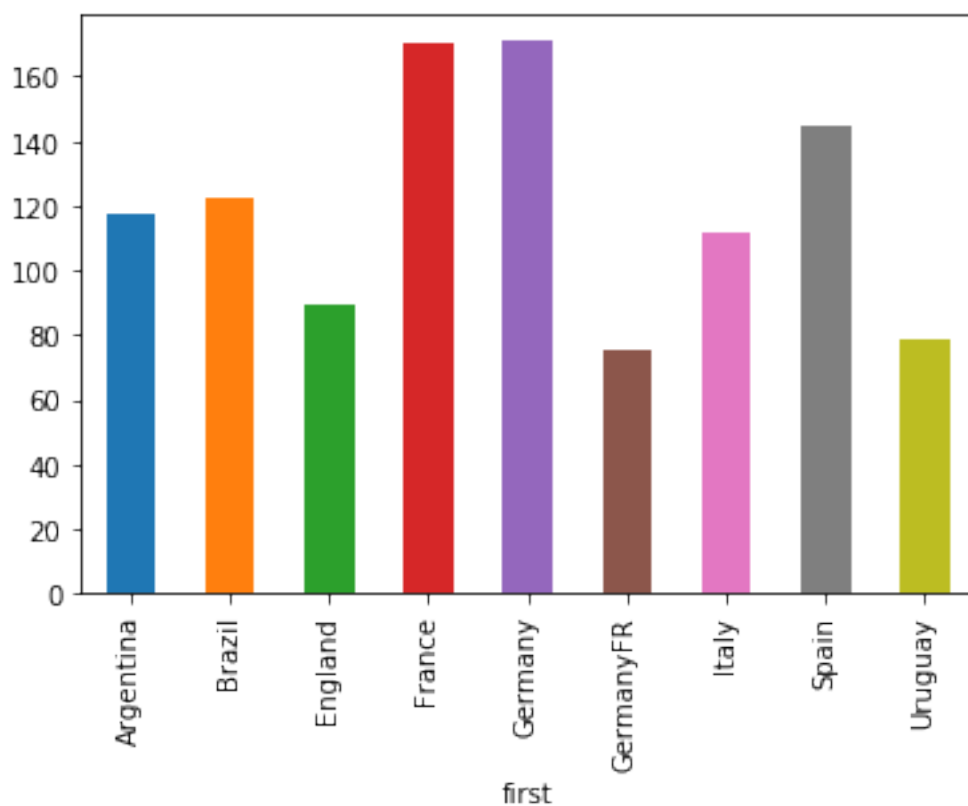
Let's look at worldcup and get a plot of goals scored by winning teams

```
In [54]: df.groupby('first')['goalsScored'].mean()
```

```
Out[54]: first
Argentina    117.000000
Brazil       122.400000
England       89.000000
France       170.000000
Germany      171.000000
GermanyFR     75.333333
Italy        111.750000
Spain        145.000000
Uruguay       79.000000
Name: goalsScored, dtype: float64
```

```
In [55]: # now add plot to that
df.groupby('first')['goalsScored'].mean().plot(kind='bar')
```

```
Out[55]: <matplotlib.axes._subplots.AxesSubplot at 0x2123b0158d0>
```



## 9 pandas.DataFrame.to\_json

`DataFrame.to_json(path_or_buf=None, orient=None, date_format=None, double_precision=10, force_ascii=True, date_unit='ms', default_handler=None, lines=False, compression=None, index=True)`

- Convert the object to a JSON string.
- `orient` defaults to 'columns': for rows use 'index'
- Note NaN's and None will be converted to null and datetime objects will be converted to UNIX timestamps.

```
In [56]: #set up
         df = pd.DataFrame(np.random.randint(low=0, high=20, size=(5, 5)),
                           columns=['a', 'b', 'c', 'd', 'e'], index=['r1', 'r2', 'r3', 'r4', 'r5'])
         print (df)
```

	a	b	c	d	e
r1	12	4	4	5	12
r2	17	6	0	11	0
r3	16	11	2	16	5
r4	14	1	10	2	5
r5	12	10	0	19	17

### 9.1 Create JSON from columns (default)

```
In [57]: s = df.to_json()
         print (s)
```

```
{"a":{"r1":12,"r2":17,"r3":16,"r4":14,"r5":12},"b":{"r1":4,"r2":6,"r3":11,"r4":1,"r5":10},"c":{
```

### 9.2 Create JSON from rows using `orient='index'`

```
In [58]: s = df.to_json(orient='index')
         print (s)
```

```
{"r1":{"a":12,"b":4,"c":4,"d":5,"e":12},"r2":{"a":17,"b":6,"c":0,"d":11,"e":0},"r3":{"a":16,"b
```