Numpy

Sahil Danayak

India in Digits

A Data visualization workshop by Coding Club



Numpy

- Written in C language
- Numpy arrays provide more efficient data storage and operations compared list
- Allows to manage vectors, matrices and higher dimensional arrays
- Used in Scientific computing, Deep learning and financial analysis

Installation

pip install numpy

Numpy Arrays



import numpy as np

```
a = np.array([1, 2, 3, 4, 5])
b = np.array((1, 2, 3, 4, 5))
c = np.array([[1, 2, 3], [4, 5, 6]])
```

Shape of Array

Returns a tuple with each index having the number of corresponding elements

print(a.shape)

Rank of Array

Returns a integer representing dimension of array

print('number of dimensions :', a.ndim)

Array slicing

```
arr = np.array([1, 2, 3, 4, 5, 6, 7])
print(arr[1:5]) #Slice elements from index 1 to index 5
print(arr[4:]) #Slice elements from index 4 to the end of the array
print(arr[:4]) #Slice elements from the beginning to index 4 (not included)
print(art)[1:5:2]) #Return every other element from index 1 to index 5
print(arr[::2]) #From the second element, slice elements from index 1
                #to index 4 (not included)
```

Random, Randint

The array is completely filled with elements between 0 and 100.

Reshaping Array

The outermost dimension will have 4 arrays, each with 3 elements.

Other Numpy Operations

Creates array with all elements as 0.11

Creates array with all elements as

1

Creates array with all elements as 0

Dot Product

```
#Traditional method
def multiply_matrix(A,B):
    C=[[] for i in range (len(A))]
        for i in range(len(A)):
            for j in range(len(B[0])):
                C[i].append(A[i][j]*B[j][i])
    return C
#Using Numpy
np.dot(A,B)
```

Pandas

India in Digits

A Data visualization workshop by Coding Club

Sahil Danayak



Pandas

- Pandas is a Python library used for working with data sets.
- It has functions for analyzing, cleaning, exploring, and manipulating data.
- The name "Pandas" has a reference to both "Panel Data", and "Python Data"

Analysis" and was created by Wes McKinney in 2008.

Installation

```
pip install pandas
```

Database Creation

```
import pandas as pd

df = pd.read_csv('data.csv')
  df =pd.read_table('user.tsv')
  df =pd.read_table('http://bit.ly/music-csv')
  df =pd.read_table('http://bit.ly/movieusers',sep='|')
```

Printing Values from Dataframe



df =pd.read_csv('http://bit.ly/uforeports')
df.head()

	City	Colors Reported	Shape Reported	State	Time
0	Ithaca	NaN	TRIANGLE	NY	6/1/1930 22:00
1	Willingboro	NaN	OTHER	NJ	6/30/1930 20:00
2	Holyoke	NaN	OVAL	со	2/15/1931 14:00
3	Abilene	NaN	DISK	KS	6/1/1931 13:00
4	New York Worlds Fair	NaN	LIGHT	NY	4/18/1933 19:00

df['City']

	7.74
0	Ithaca
1	Willingboro
2	Holyoke
3	Abilene
4	New York Worlds Fair
5	Valley City
6	Crater Lake
7	Alma
8	Eklutna
9	Hubbard

Creating New Fields



df['Location']=df['City']+','+df['State']

	City	Colors Reported	Shape Reported	State	Time	Location
0	Ithaca	NaN	TRIANGLE	NY	6/1/1930 22:00	Ithaca, NY
1	Willingboro	NaN	OTHER	NJ	6/30/1930 20:00	Willingboro, NJ
2	Holyoke	NaN	OVAL	со	2/15/1931 14:00	Holyoke, CO
3	Abilene	NaN	DISK	KS	6/1/1931 13:00	Abilene, KS
4	New York Worlds Fair	NaN	LIGHT	NY	4/18/1933 19:00	New York Worlds Fair, NY

```
#Filters out the rows which contains state as New York
df2=df[df["State"]=='NY']
#Deletes the Colors column
df.drop("Colors Reported",axis=1,inplace=True)
#Gives the shape of dataframe
df.shape
#Replacing all space by hyphen
df.columns=df.columns.str.replace(' ','-')
```



df = pd.read_csv("nba.csv")

0 Avery Bradley Boston Celtics 0.0 PG 25.0 6-2 180.0 Texas 7730337. 1 Jae Crowder Boston Celtics 99.0 SF 25.0 6-6 235.0 Marquette 6796117. 2 John Holland Boston Celtics 30.0 SG 27.0 6-5 205.0 Boston University Nal 3 R.J. Hunter Boston Celtics 28.0 SG 22.0 6-5 185.0 Georgia State 1148640. 4 Jonas Jerebko Boston Celtics 8.0 PF 29.0 6-10 231.0 NaN 5000000. 5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga <										
1 Jae Crowder Boston Celtics 99.0 SF 25.0 6-6 235.0 Marquette 6796117. 2 John Holland Boston Celtics 30.0 SG 27.0 6-5 205.0 Boston University Nal 3 R.J. Hunter Boston Celtics 28.0 SG 22.0 6-5 185.0 Georgia State 1148640. 4 Jonas Jerebko Boston Celtics 8.0 PF 29.0 6-10 231.0 NaN 5000000. 5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville		Name	Team	Number	Position	Age	Height	Weight	College	Salary
2 John Holland Boston Celtics 30.0 SG 27.0 6-5 205.0 Boston University Nal 3 R.J. Hunter Boston Celtics 28.0 SG 22.0 6-5 185.0 Georgia State 1148640. 4 Jonas Jerebko Boston Celtics 8.0 PF 29.0 6-10 231.0 NaN 5000000. 5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State <th>0</th> <th>Avery Bradley</th> <th>Boston Celtics</th> <th>0.0</th> <th>PG</th> <th>25.0</th> <th>6-2</th> <th>180.0</th> <th>Texas</th> <th>7730337.0</th>	0	Avery Bradley	Boston Celtics	0.0	PG	25.0	6-2	180.0	Texas	7730337.0
3 R.J. Hunter Boston Celtics 28.0 SG 22.0 6-5 185.0 Georgia State 1148640. 4 Jonas Jerebko Boston Celtics 8.0 PF 29.0 6-10 231.0 NaN 5000000. 5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	1	Jae Crowder	Boston Celtics	99.0	SF	25.0	6-6	235.0	Marquette	6796117.0
4 Jonas Jerebko Boston Celtics 8.0 PF 29.0 6-10 231.0 NaN 5000000. 5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	2	John Holland	Boston Celtics	30.0	SG	27.0	6-5	205.0	Boston University	NaN
5 Amir Johnson Boston Celtics 90.0 PF 29.0 6-9 240.0 NaN 12000000. 6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	3	R.J. Hunter	Boston Celtics	28.0	SG	22.0	6-5	185.0	Georgia State	1148640.0
6 Jordan Mickey Boston Celtics 55.0 PF 21.0 6-8 235.0 LSU 1170960. 7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	4	Jonas Jerebko	Boston Celtics	8.0	PF	29.0	6-10	231.0	NaN	5000000.0
7 Kelly Olynyk Boston Celtics 41.0 C 25.0 7-0 238.0 Gonzaga 2165160. 8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	5	Amir Johnson	Boston Celtics	90.0	PF	29.0	6-9	240.0	NaN	12000000.0
8 Terry Rozier Boston Celtics 12.0 PG 22.0 6-2 190.0 Louisville 1824360. 9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	6	Jordan Mickey	Boston Celtics	55.0	PF	21.0	6-8	235.0	LSU	1170960.0
9 Marcus Smart Boston Celtics 36.0 PG 22.0 6-4 220.0 Oklahoma State 3431040.	7	Kelly Olynyk	Boston Celtics	41.0	С	25.0	7-0	238.0	Gonzaga	2165160.0
	8	Terry Rozier	Boston Celtics	12.0	PG	22.0	6-2	190.0	Louisville	1824360.0
10 Jared Sullinger Boston Celtics 7.0 C 24.0 6-9 260.0 Ohio State 2569260.	9	Marcus Smart	Boston Celtics	36.0	PG	22.0	6-4	220.0	Oklahoma State	3431040.0
	10	Jared Sullinger	Boston Celtics	7.0	С	24.0	6-9	260.0	Ohio State	2569260.0

applying groupby() function to group the data on team value.

gk = df.groupby('Team')

Let's print the first entries in all the groups formed

gk.first()

	Name	Number	Position	Age	Height	Weight	College	Salary
Team								
Atlanta Hawks	Kent Bazemore	24.0	SF	26.0	6-5	201.0	Old Dominion	2000000.0
Boston Celtics	Avery Bradley	0.0	PG	25.0	6-2	180.0	Texas	7730337.0
Brooklyn Nets	Bojan Bogdanovic	44.0	SG	27.0	6-8	216.0	Oklahoma State	3425510.0
Charlotte Hornets	Nicolas Batum	5.0	SG	27.0	6-8	200.0	Virginia Commonwealth	13125306.0
Chicago Bulls	Cameron Bairstow	41.0	PF	25.0	6-9	250.0	New Mexico	845059.0
Cleveland Cavaliers	Matthew Dellavedova	8.0	PG	25.0	6-4	198.0	Saint Mary's	1147276.0
Dallas Mavericks	Justin Anderson	1.0	SG	22.0	6-6	228.0	Virginia	1449000.0
Denver Nuggets	Darrell Arthur	0.0	PF	28.0	6-9	235.0	Kansas	2814000.0



Scan this QR Code to download nba.csv

```
import pandas as pd
df = pd.read_csv("nba.csv")
df.info()
 <class 'pandas.core.frame.DataFrame'>
 RangeIndex: 458 entries, 0 to 457
 Data columns (total 9 columns):
           457 non-null object
 Name
           457 non-null object
 Team
           457 non-null float64
 Number
         457 non-null object
 Position
          457 non-null float64
 Age
          457 non-null object
 Height
          457 non-null float64
 Weight
           373 non-null object
 College
           446 non-null float64
 Salary
 dtypes: float64(4), object(5)
 memory usage: 32.3+ KB
df.sort_values(by=['Weight'])
```

```
import pandas as pd
df = pd.DataFrame([[1, 2], [4, 5], [7, 8]],
     index=['cobra', 'viper', 'sidewinder'],
     columns=['max speed', 'shield'])
           max_speed shield
cobra
viper
sidewinder
df.loc['viper']
max_speed
shield
```

```
#CONCATENATING DATAFRAMES
df1 = pd.DataFrame(
        "A": ["A0", "A1", "A2", "A3"],
        "B": ["B0", "B1", "B2", "B3"],
        "C": ["C0", "C1", "C2", "C3"],
        "D": ["D0", "D1", "D2", "D3"],
    index=[0, 1, 2, 3],
```

```
df2 = pd.DataFrame(
        "A": ["A4", "A5", "A6", "A7"],
        "B": ["B4", "B5", "B6", "B7"],
        "C": ["C4", "C5", "C6", "C7"],
        "D": ["D4", "D5", "D6", "D7"],
    },
    index=[4, 5, 6, 7],
frames = [df1, df2]
result = pd.concat(frames)
```

```
df = pd.DataFrame({'year': [2015, 2016],
                   'month': [2, 3],
                   'day': [4, 5]})
pd.to_datetime(df)
  2015-02-04
1 2016-03-05
dtype: datetime64[ns]
index = pd.Index([3, 1, 2, 3, 4])
index.value_counts()
dtype: int64
```

```
born
                                 toy
  age
                      name
0 5.0
           NaT
                      Alfred
                                 None
           1939-05-27 Batman
                                 Batmobile
  NaN
           1940-04-25
                                 Joker
df.isna()
                        toy
    age
          born name
         True False
0 False
                      True
  False False False
   True False False False
pd.unique([("a", "b"), ("b", "a"), ("a", "c"), ("b", "a")])
array([('a', 'b'), ('b', 'a'), ('a', 'c')], dtype=object)
```

Congrats on learning "Pandas & Numpy".

Coming up next "Matplotlib & Seaborn"

The
End

