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DataFrame is a 2-dimensional labeled data structure with columns of potentially different types. You can think of it like a spreadsheet or SQL table, or a dict of Series objects.



```
Import Numpy as np
Import Pandas as pd
#Creating Dataframe Objects
Arr = np.random.randint(0, 10, (5, 3))
Arr
Df = pd.DataFrame(arr)
Df
Df.values
Df.index
Df.columns
Df.values[0]
```



```
Df.index =['R1', 'R2', 'R3', 'R4', 'R5']
Df.columns = [c1, c2, c3]
DF
```

Implicitly obtain the appropriate locks for your application at the point at which they are needed. An operation that reads an object will obtain a read lock; an operation that modifies an object will obtain a write lock.

Explicit Lock: - Lock is explicitly requested for a record or table.

Implicit Lock: - Lock is implied but is not acquired

Df.loc[R3, C2]
Df.iloc[2, 1]
Df.iloc[2:4, 1:3]



Df.loc =['R3': 'R5', 'c2': 'c3']

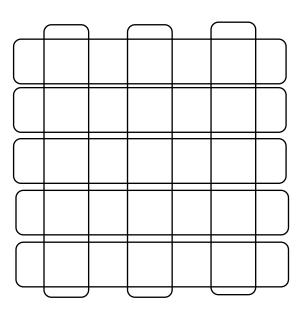
Df.iloc[0]

Type(df.iloc[0]) # you can think of df as a collection of series in both x and y axis

Df.iloc[:, 0]

Df.shape

Df.T # Transpose





Write a function to create data frame with n rows and n columns

```
Def create_df(nrows, ncols, maxrand = 10):
    arr = np.random.randint(0, maxrand, (nrows, ncols))
    df = pd.DataFrame(arr)
    df.index = ['R' + str(x) for x in np.arange(1, nrows + 1)]
    df.coloumns = ['C' + str(x) for x in np.arange(1, ncols + 1)]
    return df
```

```
Create_df(5, 3)
```

```
Create_df(2, 5)
```

```
[27] def create_df(nRows, nCols, maxRand=10):
    arr = np.random.randint(0, maxRand, (nRows, nCols))
    df = pd.DataFrame(arr)
    df.index = ['R' + str(x) for x in np.arange(1, nRows+1)]
    df.columns = ['C' + str(x) for x in np.arange(1, nCols+1)]
    return df
```



Df['earth']

Pandas – Data Frame Objects

Creating a Data Frame using multiple Series

```
mass1 = pd.Series((0.33, 4.07, 5.97, 0.642, 1090, 568, 86.0, 102, 0.0146, 0.000292), index = ['mercury', 'v
enus', 'earth', 'mars', 'jupiter', 'saturn', 'uranus', 'neptune', 'pluto', 'eris'])
#print(mass1)
dia1 = pd.Series((4079, 12104, 12756, 3475, 6792, 142904, 120536, 51110, 49528, 2370, 2326), index = ['
mercury', 'venus', 'earth', 'moon', 'mars', 'jupiter', 'saturn', 'uranus', 'neptune', 'pluto', 'eris'])
#print(dia1)
Df = pd.DataFrame({'mass' : mass1, 'diameter', dia1})
Df
df['mass']
Df['diameter']
```

Creating a Data Frame using multiple Series

Df['mass']['earth']

Df.mass.earth

Creating a new coloumn

Df['pop'] = 0

Df

Df['pop']['earth'] =8000000000

df



Creating a Data Frame using multiple Series

Df['mass'] == df.mass

df['mass'] is df.mass

Df[pop] is df.pop

Df.pop

pop function pops an item from the data frame

Df.loc ['earth']

Df.loc[:, mass]



```
# Creating a new row
Df.loc['mean'] = 0
Df
Df.drop('mean')
Df.drop('pop', axis =1)
df
Df.drop('mean', inplace = True)
Df.drop('pop', axis =1, inplace = True)
df
Np.mean(df['mass'])
Df.loc['mean'] = [np.mean(df['mass]), np.mean(df['diameter])]
df
```



Creating a function to create a new row

```
Def create_mean_row(df):
       df.loc['col_mean'] = [np.mean(df[col1]) for col in df.columns]
       return df
Create_mean_row(df)
Def create_mean_row(df):
       df.loc['col_mean1'] = df.mean()
       return df
Create_mean_row(df)
Df = create_df(5, 3)
Df
Df.mean()
```



```
# Creating a function to create a new row
# what if I want to compute mean for the row:
Df.mean(axis =1)
Df['row_mean'] = df.mean(axis =1)
Df
Df.loc['col_mean'] = df.mean()
Df
Df.median()
Df.min()
Df.max()
Df.quantile(0.25)
Df.drop('row_mean', axis =1)
```



Other ways of computing the details

```
Df.describe()
mass1 = pd.Series((0.33, 4.07, 5.97, 0.642, 1090, 568, 86.0, 102, 0.0146, 0.0002
92), index = ['mercury', 'venus', 'earth', 'mars', 'jupiter', 'saturn', 'uranus', 'nept
une', 'pluto', 'eris'])
#print(mass1)
dia1 = pd.Series((4079, 12104, 12756, 3475, 6792, 142904, 120536, 51110, 495
28, 2370, 2326), index = ['mercury', 'venus', 'earth', 'moon', 'mars', 'jupiter', 'sat
urn', 'uranus', 'neptune', 'pluto', 'eris'])
#print(dia1)
```

Planets = pd.DataFrame({'mass' : mass1, 'diameter', dia1})
Planets.describe()



Seaborn - Planets data set

Import seaborn as sns

Df = sns.load_dataset('planets')

df.info()

Df.head()

Df.tail()

Df.describe()



```
# Task 1: Go through each row of the df and delete (drop) if any of the cols is null
# Regular way
For r in df.index:
        for c in df.columns:
                if pd.isnull(df.loc[r, c]):
                df.drop(r, inplace = True)
                break
Df.describe()
For I, r in df.iterrows():
        print(i)
        print(r)
        break
```



Task 1: Go through each row of the df and delete (drop) if any of the cols is null

```
For I, r in df.iterrows():
        print(pd.isnull(r))
        break
For I, r in df.iterrows():
        print(pd.isnull(r).any()):
        break
For I, r in df. Iterrows():
        if pd.isnull(r).any():
        df.drop(I, inplace = True)
```



```
Df.describe()
```

most efficient way

Df.dropna(inplace = true)
Df.describe()

Task 2: filter rows for planets found in 2010s and methods is 'Radial Velocity' & 'Transit' & distance is > 75%



```
# Method: 1
Df_1 = df.copy()
Per_75 = df.distance.quantile(0.75)
For I, r in df_1.iterrows():
         if r['year'] < 2010:
                  df_1.drop(I, inplace = True)
                  continue
         if r['method'] != 'Radial Velocity' and r['method'] != 'Transit':
                  df_1.drop(I, inplace = True)
                  continue
         if r['distance'] < per_75:</pre>
                  df_1.drop(I, inplace = True)
                  continue
Df_1.describe()
df_1.head()
Per_75
Df_1.tail()
```

Df.describe()

PRIMEINTUIT Pandas – Data Frame Objects

Method: 2 using indexing with conditions

```
Df_2 = df.copy()
Df 3 = df 2
         (df_2['year'] > = 2010) &
         ((df_2['method'] == 'Radial Velocity') | (df_2['method'] == 'Transit')) &
         (df_2['distance'] > per_75)
```

Task 3: Modify the method column to have only the abbreviation of each method



```
# Method: 1
Df_2.method.unique()
S = 'Radial Velocity'
S.split(' ')
[x[0] for x in S.split(' ')]
' '.join([x[0] for x in S.split(' ')])
Short_Names = {}
For s in df.method.unique():
         short_name[s] = ' '.join([x[0] for x in S.split(' ')])
Print(Short_ Name)
```



PRIMEINTUIT Finishing School Pandas - Data Frame Objects

```
# Method: 1 Continued......
For I, r in df.iterrows():
        df.loc[I, 'short_method'] = Short_Names.get(r['method'], r['method'])
        # new coloumn
                                 Get names form short_names dict by using get function
Df.head()
Df.tail()
```

PRIMEINTUIT Pandas – Data Frame Objects

```
# Method: 2
Df5 = sns.load_dataset('planets')
Def Shorten_method(s):
        Short_Names.get(s,s)
df5['Short_Method'] = df['method'].apply(Shorten_method)
#Apply Shorten_method to each coloumn in df5/method column take the result and put it
into Short_Method column.
```

Task 4: count the number of planets discovered for each method type



Task 4: count the number of planets discovered for each method type

Step 1. Split the data frame into smaller chunks (in this case they should have the same method name)

Step 2. Apply some function in each smaller chuck (count function)

Step 3. Aggregate the results from each chunk together.



Method 1



```
# Method 2

Df5.groupby('method').count()
# grouping based on column 'method'

Df5.groupby('method')['method'].count()
# grouping based on column 'method' and count on column 'method'

Print(d)

Df5.groupby('method') ['distance'].mean()
```

Task 5: Find out what fraction of planets have been found in the last decase(2010s) across each method type



Task 5: Find out what fraction of planets have been found in the last decase(2010s) across each method type

```
# Step 1: Filter the data for given condition ( planets found in last decade)
```

- # Step 2: Split based on the method
- # Step 3: Apply count function
- # Step 4: Aggregate the final results

Task 5: Find out what fraction of planets have been found in the last decase(2010s) across each method type

```
Df5[df5.year >= 2010].groupby('method')

Df5[df5.year >= 2010].groupby('method')['method'].count()

S_2010s = Df5[df5.year >= 2010].groupby('method')['method'].count()

S_alltime = Df5.groupby('method')['method'].count()

S_2010s / s_alltime
```