

Data Wrangling

December 6, 2021

```
[1]: #Data Wrangling
# It is the process of transforming raw data to a clean and organised format
    ↳ ready to use

import pandas as pd

url = "https://osf.io/aupb4/download"

dataframe = pd.read_csv(url)
dataframe.head(5)
```

```
[1]:
```

	Name	PClass	Age	Sex	\
0	Allen, Miss Elisabeth Walton	1st	29.00	female	
1	Allison, Miss Helen Loraine	1st	2.00	female	
2	Allison, Mr Hudson Joshua Creighton	1st	30.00	male	
3	Allison, Mrs Hudson JC (Bessie Waldo Daniels)	1st	25.00	female	
4	Allison, Master Hudson Trevor	1st	0.92	male	

	Survived
0	1
1	0
2	0
3	0
4	1

```
[2]: #Creating Data Frame

dataframe = pd.DataFrame()

#Adding columns
dataframe["Name"]=["A","B","C"]
dataframe["Age"]=[23,12,8]
dataframe["Gender"]=["M","F","F"]
```

```
[3]: #Create row

new_person = pd.Series(["M",22,"M"],index=["Name","Age","Gender"])
```

```
#Append row to dataframe
dataframe = dataframe.append(new_person,ignore_index=True)
print(dataframe)
```

	Name	Age	Gender
0	A	23	M
1	B	12	F
2	C	8	F
3	M	22	M

```
[4]: #Describing the Data
url = "https://osf.io/aupb4/download"

dataframe = pd.read_csv(url)
print(dataframe.head(4))
```

	Name	PClass	Age	Sex	\
0	Allen, Miss Elisabeth Walton	1st	29.0	female	
1	Allison, Miss Helen Loraine	1st	2.0	female	
2	Allison, Mr Hudson Joshua Creighton	1st	30.0	male	
3	Allison, Mrs Hudson JC (Bessie Waldo Daniels)	1st	25.0	female	

	Survived
0	1
1	0
2	0
3	0

```
[5]: #Shape
print(dataframe.shape)

#Show statistics
print(dataframe.describe())
```

```
(1313, 5)
```

	Age	Survived
count	756.000000	1313.000000
mean	30.397989	0.342727
std	14.259049	0.474802
min	0.170000	0.000000
25%	21.000000	0.000000
50%	28.000000	0.000000
75%	39.000000	1.000000
max	71.000000	1.000000

```
[6]: #Navigating DataFrames
```

```

#Selecting rows
dataframe.iloc[0] #Selects first row

dataframe.iloc[1:4] #Selecting 3 rows
dataframe.iloc[:4] #Selecting rows upto 4th row

#Set index to any other feature i.e Name
dataframe = dataframe.set_index(dataframe["Name"])

dataframe.loc["Allen, Miss Elisabeth Walton"]

```

```

[6]: Name      Allen, Miss Elisabeth Walton
     PClass      1st
     Age      29.0
     Sex      female
     Survived      1
     Name: Allen, Miss Elisabeth Walton, dtype: object

```

```

[7]: #loc is used when the index of the DataFrame is a label
     #iloc works by looking for the position in the DataFrame.

```

```

[8]: # Selecting Rows Based on Conditionals

#selecting only the female data
dataframe[dataframe["Sex"]=="female"].head(5)

#Females with age more than 65
dataframe[(dataframe["Sex"]=="female") & (dataframe["Age"]>=65)]

#Males who survived
dataframe[(dataframe["Sex"]=="male") & (dataframe["Survived"]==1)].head(5)

```

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[8]:

```

	Name	PClass	Age	\
Name				
	Allison, Master Hudson Trevor	Allison, Master Hudson Trevor	1st	0.92
	Anderson, Mr Harry	Anderson, Mr Harry	1st	47.00
	Barkworth, Mr Algernon H	Barkworth, Mr Algernon H	1st	NaN
	Beckwith, Mr Richard Leord	Beckwith, Mr Richard Leord	1st	37.00
	Behr, Mr Karl Howell	Behr, Mr Karl Howell	1st	26.00

	Sex	Survived
Name		
	Allison, Master Hudson Trevor	male 1
	Anderson, Mr Harry	male 1
	Barkworth, Mr Algernon H	male 1
	Beckwith, Mr Richard Leord	male 1
	Behr, Mr Karl Howell	male 1

```
[9]: # Replacing Values in a DataFrame
#DataFrame.replace(a,b) -- replace a with b

dataframe["Survived"].replace(1,"alive").head(5)

dataframe["Sex"].replace("female","Woman").head(5)

dataframe["Sex"].replace(["female","male"],["F","M"]).head(5)

dataframe.replace(1,"one").head(6) #It replaces in whole data
```

```
[9]:      Name \
Name
Allen, Miss Elisabeth Walton      Allen, Miss
Elisabeth Walton
Allison, Miss Helen Loraine      Allison, Miss
Helen Loraine
Allison, Mr Hudson Joshua Creighton      Allison, Mr Hudson
Joshua Creighton
Allison, Mrs Hudson JC (Bessie Waldo Daniels) Allison, Mrs Hudson JC (Bessie
Waldo Daniels)
Allison, Master Hudson Trevor      Allison, Master
Hudson Trevor
Anderson, Mr Harry
Anderson, Mr Harry
```

	PClass	Age	Sex	Survived
Name				
Allen, Miss Elisabeth Walton	1st	29.0	female	one
Allison, Miss Helen Loraine	1st	2.0	female	0
Allison, Mr Hudson Joshua Creighton	1st	30.0	male	0
Allison, Mrs Hudson JC (Bessie Waldo Daniels)	1st	25.0	female	0
Allison, Master Hudson Trevor	1st	0.92	male	one
Anderson, Mr Harry	1st	47.0	male	one

```
[10]: # Renaming Columns

dataframe.rename(columns={"PClass":"Passenger Class","Sex":"Gender"}).head(5)
#It takes a dictionary to replace names of one column or multiple columns
```

```
[10]:      Name \
Name
Allen, Miss Elisabeth Walton      Allen, Miss
Elisabeth Walton
Allison, Miss Helen Loraine      Allison, Miss
Helen Loraine
Allison, Mr Hudson Joshua Creighton      Allison, Mr Hudson
```

Joshua Creighton
 Allison, Mrs Hudson JC (Bessie Waldo Daniels) Allison, Mrs Hudson JC (Bessie Waldo Daniels)
 Allison, Master Hudson Trevor Allison, Master Hudson Trevor

Name	Passenger Class	Age	Gender	\
Allen, Miss Elisabeth Walton	1st	29.00	female	
Allison, Miss Helen Loraine	1st	2.00	female	
Allison, Mr Hudson Joshua Creighton	1st	30.00	male	
Allison, Mrs Hudson JC (Bessie Waldo Daniels)	1st	25.00	female	
Allison, Master Hudson Trevor	1st	0.92	male	

Name	Survived
Allen, Miss Elisabeth Walton	1
Allison, Miss Helen Loraine	0
Allison, Mr Hudson Joshua Creighton	0
Allison, Mrs Hudson JC (Bessie Waldo Daniels)	0
Allison, Master Hudson Trevor	1

```
[11]: import collections

column_names = collections.defaultdict(str)

for name in dataframe.columns:
    column_names[name]

print(column_names)
```

```
defaultdict(<class 'str'>, {'Name': '', 'PClass': '', 'Age': '', 'Sex': '',
'Survived': ''})
```

```
[12]: # Maximum, Minimum, Sum, Average and Count

print("Max: ",dataframe["Age"].max())
print("Min: ",dataframe["Age"].min())
print("Avg: ",dataframe["Age"].mean())
print("Sum: ",dataframe["Age"].sum())
print("Count: ",dataframe["Age"].count())
```

```
Max: 71.0
Min: 0.17
Avg: 30.397989417989418
Sum: 22980.88
Count: 756
```

```
[13]: dataframe.count()
```

```
[13]: Name          1313
      PClass       1312
      Age          756
      Sex          1313
      Survived     1313
      dtype: int64
```

```
[14]: # Unique Values
      dataframe["PClass"].unique()

      dataframe["PClass"].value_counts() #Freq table of unique values

      dataframe["Sex"].value_counts()

      dataframe["PClass"].nunique() #Give the count of unique values
```

```
[14]: 3
```

```
[15]: # Missing values
      dataframe[dataframe["Age"].isnull()].head(2)
```

```
[15]:
```

	Name	PClass	Age	\
Name				
Aubert, Mrs Leontine Pauline	Aubert, Mrs Leontine Pauline	1st	NaN	
Barkworth, Mr Algernon H	Barkworth, Mr Algernon H	1st	NaN	

	Sex	Survived
Name		
Aubert, Mrs Leontine Pauline	female	1
Barkworth, Mr Algernon H	male	1

```
[16]: # Replacing values with NaN
      import numpy as np
      dataframe["Sex"].replace("male",np.nan)
```

```
[16]:
```

Name	
Allen, Miss Elisabeth Walton	female
Allison, Miss Helen Loraine	female
Allison, Mr Hudson Joshua Creighton	NaN
Allison, Mrs Hudson JC (Bessie Waldo Daniels)	female
Allison, Master Hudson Trevor	NaN
...	
Zakarian, Mr Artun	NaN
Zakarian, Mr Maprieder	NaN
Zenni, Mr Philip	NaN

```
Lievens, Mr Rene                NaN
Zimmerman, Leo                  NaN
Name: Sex, Length: 1313, dtype: object
```

```
[17]: #Load Data, Set missing values
dataframe = pd.read_csv(url,na_values=[np.nan,"NONE",-999])
```

```
[18]: # Deleting Columns

dataframe.drop("Age",axis=1).head(5)

dataframe.drop(["Age","Sex"],axis=1).head(5)

dataframe.drop(dataframe.columns[1],axis=1).head(5) #using index if names not
↳available
```

```
[18]:
```

	Name	Age	Sex	Survived
0	Allen, Miss Elisabeth Walton	29.00	female	1
1	Allison, Miss Helen Loraine	2.00	female	0
2	Allison, Mr Hudson Joshua Creighton	30.00	male	0
3	Allison, Mrs Hudson JC (Bessie Waldo Daniels)	25.00	female	0
4	Allison, Master Hudson Trevor	0.92	male	1

```
[19]: # Deleting Rows

dataframe[dataframe["Sex"] != "female"].head(4)

dataframe[dataframe["Age"] >=18 ].head(4)

dataframe[dataframe.index != 0].head(3) #Using index
```

```
[19]:
```

	Name	PClass	Age	Sex	\
1	Allison, Miss Helen Loraine	1st	2.0	female	
2	Allison, Mr Hudson Joshua Creighton	1st	30.0	male	
3	Allison, Mrs Hudson JC (Bessie Waldo Daniels)	1st	25.0	female	

	Survived
1	0
2	0
3	0

```
[20]: # Dropping Duplicate rows
dataframe.drop_duplicates().head(4) #It will drop only when the both rows have
↳exact row values

dataframe.drop_duplicates(subset=["Sex"]) # only unique value will be obtained
```

```
dataframe.drop_duplicates(subset=["Sex"],keep="last") #keeps last unique
```

```
[20]:
```

	Name	PClass	Age	Sex	Survived
1307	Zabour, Miss Tamini	3rd	NaN	female	0
1312	Zimmerman, Leo	3rd	29.0	male	0

```
[21]: # Grouping rows by values
```

```
dataframe.groupby("Sex").mean()

dataframe.groupby("Survived")["Name"].count()

dataframe.groupby("Survived").count()

dataframe.groupby(["Sex", "Survived"])["Age"].count()
```

```
[21]: Sex      Survived
female 0          71
       1         217
male   0         372
       1          96
Name: Age, dtype: int64
```

```
[22]: # Group Rows by Time
#Create date range
time_index = pd.date_range("14/12/2012",periods=100000,freq="30S") # "30S"-30_
↳secs frequency

#Create DataFrame
dataframe = pd.DataFrame(index=time_index)

#Add Columns
dataframe["Sale_Amount"] = np.random.randint(1,10,100000)

#Group rows by week, calculate sum per week
dataframe["Sale_Amount"].resample("W").sum()

dataframe["Sale_Amount"].resample("2W").mean() # 2 weeks mean

dataframe["Sale_Amount"].resample("M").count() #monthly count

#resample returns the label of the right "edge" (the last label) of the time_
↳group.
#We can control this behavior using the label parameter:

dataframe["Sale_Amount"].resample("M",label="left").count()
```



```
[22]: 2012-11-30    51840
      2012-12-31    48160
      Freq: M, Name: Sale_Amount, dtype: int64
```

```
[23]: # Looping over a column

url = "https://osf.io/aupb4/download"

dataframe = pd.read_csv(url)

for name in dataframe["Name"][0:5]:
    print(name.upper())

#List comprehension can be used as
print([name.upper() for name in dataframe["Name"][0:5]])
```

```
ALLEN, MISS ELISABETH WALTON
ALLISON, MISS HELEN LORAINÉ
ALLISON, MR HUDSON JOSHUA CREIGHTON
ALLISON, MRS HUDSON JC (BESSIE WALDO DANIELS)
ALLISON, MASTER HUDSON TREVOR
['ALLEN, MISS ELISABETH WALTON', 'ALLISON, MISS HELEN LORAINÉ', 'ALLISON, MR
HUDSON JOSHUA CREIGHTON', 'ALLISON, MRS HUDSON JC (BESSIE WALDO DANIELS)',
'ALLISON, MASTER HUDSON TREVOR']
```

```
[24]: # Applying a Function over all elements in a column

#Creating a function
def uppercase(x):
    return x.upper()

#apply function to column
dataframe["Name"].apply(uppercase)[0:5]
```

```
[24]: 0          ALLEN, MISS ELISABETH WALTON
      1          ALLISON, MISS HELEN LORAINÉ
      2          ALLISON, MR HUDSON JOSHUA CREIGHTON
      3  ALLISON, MRS HUDSON JC (BESSIE WALDO DANIELS)
      4          ALLISON, MASTER HUDSON TREVOR
      Name: Name, dtype: object
```

```
[25]: # Applying a Function to groups

'''By combining groupby and apply we can calculate cus-
tom statistics or apply any function to each group separately.'''

dataframe.groupby("Sex").apply(lambda x:x.count())
```

```
[25]:
```

	Name	PClass	Age	Sex	Survived
Sex					
female	462	462	288	462	462
male	851	850	468	851	851

```
[26]: # Concatenating DataFrames

data1 = {"id":["1","2","3"],
         "name":["sd","ds","ss"]}
dataframe1 = pd.DataFrame(data1,columns=["id","name"])

data2 = {"id":["4","5","6"],
         "name":["fd","df","ff"]}
dataframe2 = pd.DataFrame(data2,columns=["id","name"])

pd.concat([dataframe1,dataframe2],axis=0)

pd.concat([dataframe1,dataframe2],axis=1)

# Add Series to Dataframe
row = pd.Series(["5","huihedf"],index=["id","name"])

dataframe1.append(row,ignore_index=True)
```

```
[26]:
```

	id	name
0	1	sd
1	2	ds
2	3	ss
3	5	huihedf

```
[27]: # Merging DataFrames
import pandas as pd
employees = {"employee_id":["1','2','3','4'],
            'name':["A","B","C","D"]}
dataframe_employees = pd.DataFrame(employees,columns=["employee_id","name"])

sales = {"employee_id":["3","4","5","6"],
         "total_sales": [1323,3254,2343,424]}
dataframe_sales = pd.DataFrame(sales,columns=["employee_id","total_sales"])

#Merge DataFrames
pd.merge(dataframe_employees,dataframe_sales,on="employee_id")
# Defaults to inner join
```

```
[27]:
```

	employee_id	name	total_sales
0	3	C	1323
1	4	D	3254

```
[28]: # Outer join
pd.merge(dataframe_employees,dataframe_sales,on="employee_id",how="outer")
```

```
[28]:  employee_id name  total_sales
0           1    A           NaN
1           2    B           NaN
2           3    C        1323.0
3           4    D        3254.0
4           5  NaN        2343.0
5           6  NaN         424.0
```

```
[29]: # Left join
pd.merge(dataframe_employees,dataframe_sales,on="employee_id",how="left")
```

```
[29]:  employee_id name  total_sales
0           1    A           NaN
1           2    B           NaN
2           3    C        1323.0
3           4    D        3254.0
```

```
[30]: # Right join
pd.merge(dataframe_employees,dataframe_sales,on="employee_id",how="right")
```

```
[30]:  employee_id name  total_sales
0           3    C        1323
1           4    D        3254
2           5  NaN        2343
3           6  NaN         424
```

```
[31]: '''If instead of merging on two columns we want to merge on the indexes of each
↳Data-
Frame, we can replace the left_on and right_on parameters with right_index=True
and left_index=True .'''
```

```
pd.merge(dataframe_employees,
dataframe_sales,
left_on='employee_id',
right_on='employee_id')
```

```
[31]:  employee_id name  total_sales
0           3    C        1323
1           4    D        3254
```

```
[32]: #Joins
'''Inner
Return only the rows that match in both DataFrames (e.g., return any row with
an employee_id value appearing in both dataframe_employees and data
```

```

frame_sales ).
Outer
Return all rows in both DataFrames. If a row exists in one DataFrame but not in
the other DataFrame, fill NaN values for the missing values (e.g., return all
↳rows
in both employee_id and dataframe_sales ).
Left
Return all rows from the left DataFrame but only rows from the right DataFrame
that matched with the left DataFrame. Fill NaN values for the missing values (e.
↳g.,
return all rows from dataframe_employees but only rows from data
frame_sales that have a value for employee_id that appears in data
frame_employees ).
Right
Return all rows from the right DataFrame but only rows from the left DataFrame
that matched with the right DataFrame. Fill NaN values for the missing values
(e.g., return all rows from dataframe_sales but only rows from data
↳frame_employees that have a value for employee_id that appears in data
frame_sales ).'''

```

[32]: 'Inner\nReturn only the rows that match in both DataFrames (e.g., return any row with\nnan employee_id value appearing in both dataframe_employees and data\nframe_sales).\nOuter\nReturn all rows in both DataFrames. If a row exists in one DataFrame but not in\nthe other DataFrame, fill NaN values for the missing values (e.g., return all rows\nin both employee_id and dataframe_sales).\nLeft\nReturn all rows from the left DataFrame but only rows from the right DataFrame\nthat matched with the left DataFrame. Fill NaN values for the missing values (e.g.,\nreturn all rows from dataframe_employees but only rows from data\nframe_sales that have a value for employee_id that appears in data\nframe_employees).\nRight\nReturn all rows from the right DataFrame but only rows from the left DataFrame\nthat matched with the right DataFrame. Fill NaN values for the missing values\n(e.g., return all rows from dataframe_sales but only rows from data frame_employees that have a value for employee_id that appears in data\nframe_sales).'