```
# create a table with columns and rows
# column name : students,technology,marks
# add the column teacher
# and delete the technology later
# and try to to practice on all the example methods
# 15 to 20 records
dataset = {
  "students": [
    "Navya", "Vyshnavi", "Keerthi", "Amitha", "Vidya",
    "Raksha", "Abinaya", "Rohit", "Joshua", "Sameer",
    "Akhil", "Manish", "Divya", "Shreya", "Ankit",
    "Priya", "Suresh", "Megha", "Rahul", "Sailesh"
  ],
  "technology": [
    "Python", "Java", "Python", "C++", "Java",
    "Python", "Java", "C++", "Python", "Java",
    "C++", "Python", "Java", "C++", "Python",
    "Java", "Python", "C++", "Java", "Python"
  ],
  "marks": [
    79, 68, 49, 64, 69,
    98, 84, 61, 44, 44,
    59, 93, 67, 96, 85,
    46, 41, 82, 74, 74
}
data frm = pd.DataFrame(dataset)
```

```
print(data frm)
print(data frm.head(3))
print(data frm.tail(3))
print(data frm.shape)
print(data frm.dtypes)
print(data frm.columns)
print(data frm.describe())
print(data_frm['students'])
print(data frm["marks"]>75)
data_frm.drop('technology', axis=1, inplace=True)
data frm["teacher"] = [
  "Mr. Rao", "Ms. Priya", "Mr. Khan", "Ms. Lee", "Mrs. Das",
  "Mr. Joshi", "Ms. Smith", "Mr. Wang", "Mrs. Patel", "Ms. Kim",
  "Mr. Roy", "Ms. Gupta", "Mr. Chen", "Mrs. Brown", "Ms. Singh",
  "Mr. Kumar", "Ms. White", "Mrs. Thomas", "Mr. Clark", "Ms. Lewis"
]
print(data frm)
Output:
students technology marks
0
    Navya
              Python
                       79
  Vyshnavi
                       68
                Java
1
   Keerthi
             Python
                       49
3
    Amitha
                C++
                       64
                      69
4
    Vidya
              Java
```

5

Raksha

Abinaya

Python

Java

98

84

- 7 Rohit C++ 61
- 8 Joshua Python 44
- 9 Sameer Java 44
- 10 Akhil C++ 59
- 11 Manish Python 93
- 12 Divya Java 67
- 13 Shreya C++ 96
- 14 Ankit Python 85
- 15 Priya Java 46
- 16 Suresh Python 41
- 17 Megha C++ 82
- 18 Rahul Java 74
- 19 Sailesh Python 74

students technology marks

- 0 Navya Python 79
- 1 Vyshnavi Java 68
- 2 Keerthi Python 49

students technology marks

- 17 Megha C++ 82
- 18 Rahul Java 74
- 19 Sailesh Python 74

(20, 3)

students object

technology object

marks int64

dtype: object

Index(['students', 'technology', 'marks'], dtype='object')

marks

count 20.000000

mean 68.850000

std 17.921686

min 41.000000

25% 56.500000

50% 68.500000

75% 82.500000

max 98.000000

- 0 Navya
- 1 Vyshnavi
- 2 Keerthi
- 3 Amitha
- 4 Vidya
- 5 Raksha
- 6 Abinaya
- 7 Rohit
- 8 Joshua
- 9 Sameer
- 10 Akhil
- 11 Manish
- 12 Divya
- 13 Shreya
- 14 Ankit
- 15 Priya
- 16 Suresh
- 17 Megha

- 18 Rahul
- 19 Sailesh

Name: students, dtype: object

- 0 True
- 1 False
- 2 False
- 3 False
- 4 False
- 5 True
- 6 True
- 7 False
- 8 False
- 9 False
- 10 False
- 11 True
- 12 False
- 13 True
- 14 True
- 15 False
- 16 False
- 17 True
- 18 False
- 19 False

Name: marks, dtype: bool

students marks teacher

- 0 Navya 79 Mr. Rao
- 1 Vyshnavi 68 Ms. Priya

- 2 Keerthi 49 Mr. Khan
- 3 Amitha 64 Ms. Lee
- 4 Vidya 69 Mrs. Das
- 5 Raksha 98 Mr. Joshi
- 6 Abinaya 84 Ms. Smith
- 7 Rohit 61 Mr. Wang
- 8 Joshua 44 Mrs. Patel
- 9 Sameer 44 Ms. Kim
- 10 Akhil 59 Mr. Roy
- 11 Manish 93 Ms. Gupta
- 12 Divya 67 Mr. Chen
- 13 Shreya 96 Mrs. Brown
- 14 Ankit 85 Ms. Singh
- 15 Priya 46 Mr. Kumar
- 16 Suresh 41 Ms. White
- 17 Megha 82 Mrs. Thomas
- 18 Rahul 74 Mr. Clark
- 19 Sailesh 74 Ms. Lewis

Pandas Assessment

```
import pandas as pd
data = {
  'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva', 'Frank', 'Grace', 'Helen'],
  'Age': [25, 30, 35, 40, 28, 32, 45, 38],
  'Department': ['HR', 'Engineering', 'Finance', 'HR', 'Engineering', 'Finance', 'HR', 'Finance'],
  'Salary': [50000, 60000, 70000, 52000, None, 58000, None, 62000]
}
df = pd.DataFrame(data)
df.to csv('data csv using pandas.csv', index=False)
print("DataFrame with 8 records:")
print(df)
#1. Load data.csv and display the first 5 rows
df = pd.read csv('data csv using pandas.csv')
print("First 5 rows:")
print(df.head())
#2. Check the shape of the DataFrame (rows, columns)
print("\nShape of DataFrame (rows, columns):")
print(df.shape)
#3. Get summary statistics for numeric columns
print("\nSummary statistics (numeric columns):")
print(df.describe())
```

```
# 4. Extract the 'Age' column as a Series
print("\n'Age' column:")
print(df['Age'])
#5. Show rows where 'Salary' > 50000
print("\nRows where Salary > 50000:")
print(df[df]'Salary'] > 50000])
#6. Rows where 'Department' == 'HR' and 'Age' > 30
print("\nRows where Department == 'HR' and Age > 30:")
print(df[(df['Department'] == 'HR') & (df['Age'] > 30)])
#7. Find which columns have NaN values and how many
print("\nMissing values per column:")
print(df.isnull().sum())
#8. Fill NaN values in 'Salary' with 0
df filled = df.copy()
df filled['Salary'] = df filled['Salary'].fillna(0)
print("\n'Salary' column after filling NaNs with 0:")
print(df filled['Salary'])
#9. Drop duplicate rows and reset the index
df no duplicates = df.drop duplicates().reset index(drop=True)
print("\nDataFrame after dropping duplicates and resetting index:")
print(df no duplicates)
```

```
#10. Sort rows by "Age" in descending order
df sorted = df.sort values(by='Age', ascending=False)
print("\nDataFrame sorted by Age (descending):")
print(df sorted)
#11. Group by "Department" and find average "Salary"
print("\nAverage Salary by Department:")
print(df.groupby('Department')['Salary'].mean())
#12. Count unique departments in the "Department" column
print("\nNumber of unique departments:")
print(df['Department'].nunique())
Output:
DataFrame with 8 records:
   Name Age Department Salary
  Alice 25
                  HR 50000.0
    Bob 30 Engineering 60000.0
1
2 Charlie 35
                Finance 70000.0
   David 40
                   HR 52000.0
4
    Eva 28 Engineering
                            NaN
   Frank 32
                Finance 58000.0
5
   Grace 45
                   HR
                         NaN
   Helen 38
                Finance 62000.0
First 5 rows:
```

Name Age Department Salary

HR 50000.0

Alice 25

- 1 Bob 30 Engineering 60000.0
- 2 Charlie 35 Finance 70000.0
- 3 David 40 HR 52000.0
- 4 Eva 28 Engineering NaN

Shape of DataFrame (rows, columns):

(8, 4)

Summary statistics (numeric columns):

Age Salary

count 8.000000 6.000000

mean 34.125000 58666.666667

std 6.664136 7229.568913

min 25.000000 50000.000000

25% 29.500000 53500.000000

50% 33.500000 59000.000000

75% 38.500000 61500.000000

max 45.000000 70000.000000

'Age' column:

- 0 25
- 1 30
- 2 35
- 3 40
- 4 28
- 5 32
- 6 45

Name: Age, dtype: int64

Rows where Salary > 50000:

Name Age Department Salary

- 1 Bob 30 Engineering 60000.0
- 2 Charlie 35 Finance 70000.0
- 3 David 40 HR 52000.0
- 5 Frank 32 Finance 58000.0
- 7 Helen 38 Finance 62000.0

Rows where Department == 'HR' and Age > 30:

Name Age Department Salary

- 3 David 40 HR 52000.0
- 6 Grace 45 HR NaN

Missing values per column:

Name 0

Age 0

Department 0

Salary 2

dtype: int64

'Salary' column after filling NaNs with 0:

- 0 50000.0
- 1 60000.0
- 2 70000.0

- 3 52000.0
- 4 0.0
- 5 58000.0
- 6 0.0
- 7 62000.0

Name: Salary, dtype: float64

DataFrame after dropping duplicates and resetting index:

Name Age Department Salary

- 0 Alice 25 HR 50000.0
- 1 Bob 30 Engineering 60000.0
- 2 Charlie 35 Finance 70000.0
- 3 David 40 HR 52000.0
- 4 Eva 28 Engineering NaN
- 5 Frank 32 Finance 58000.0
- 6 Grace 45 HR NaN
- 7 Helen 38 Finance 62000.0

DataFrame sorted by Age (descending):

Name Age Department Salary

- 6 Grace 45 HR NaN
- 3 David 40 HR 52000.0
- 7 Helen 38 Finance 62000.0
- 2 Charlie 35 Finance 70000.0
- 5 Frank 32 Finance 58000.0
- 1 Bob 30 Engineering 60000.0
- 4 Eva 28 Engineering NaN

0 Alice 25 HR 50000.0

Average Salary by Department:

Department

Engineering 60000.000000

Finance 63333.333333

HR 51000.000000

Name: Salary, dtype: float64

Number of unique departments:

3