

```
import pandas as pd
```

```
ds = pd.read_csv('iris.csv')
```

```
ds.head()
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

```
# Group by the categorical variable 'Species'
```

```
grouped_stats = ds.groupby('Species').agg({  
    'SepalLengthCm' : ['mean', 'median', 'min', 'max', 'std'],  
    'SepalWidthCm' : ['mean', 'median', 'min', 'max', 'std'],  
    'PetalLengthCm' : ['mean', 'median', 'min', 'max', 'std'],  
    'PetalWidthCm' : ['mean', 'median', 'min', 'max', 'std'],  
})
```

```
grouped_stats
```

```
ds.describe()
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	150.000000	150.000000	150.000000	150.000000	150.000000
mean	75.500000	5.843333	3.054000	3.758667	1.198667
std	43.445368	0.828066	0.433594	1.764420	0.763161
min	1.000000	4.300000	2.000000	1.000000	0.100000
25%	38.250000	5.100000	2.800000	1.600000	0.300000
50%	75.500000	5.800000	3.000000	4.350000	1.300000
75%	112.750000	6.400000	3.300000	5.100000	1.800000
max	150.000000	7.900000	4.400000	6.900000	2.500000

```
# Create a list with numeric values for each Categorical Response

species_numeric_list = {}

for species, group in ds.groupby('Species'):
    species_numeric_list[species] = group['SepalLengthCm'].tolist()

species_numeric_list

{'Iris-setosa': [5.1,
4.9,
4.7,
4.6,
5.0,
5.4,
4.6,
5.0,
4.4,
4.9,
5.4,
4.8,
4.8,
4.3,
5.8,
5.7,
5.4,
5.1,
5.7,
5.1,
5.4,
5.1,
4.6,
5.1,
4.8,
5.0,
5.0,
5.2,
5.2,
4.7,
4.8,
5.4,
5.2,
5.5,
4.9,
5.0,
5.5,
4.9,
4.4,
5.1,
5.0,
```

```
4.5,  
4.4,  
5.0,  
5.1,  
4.8,  
5.1,  
4.6,  
5.3,  
5.0],  
'Iris-versicolor': [7.0,  
6.4,  
6.9,  
5.5,  
6.5,  
5.7,  
6.3,  
4.9,  
6.6,  
5.2,  
5.0,  
5.9,  
6.0,  
6.1,  
5.6,  
6.7,  
5.6,  
5.8,  
6.2,  
5.6,  
5.9,  
6.1,  
6.3,  
6.1,  
6.4,  
6.6,  
6.8,  
6.7,  
6.0,  
5.7,  
5.5,  
5.5,  
5.8,  
6.0,  
5.4,  
6.0,  
6.7,  
6.3,  
5.6,  
5.5,
```

```
5.5,  
6.1,  
5.8,  
5.0,  
5.6,  
5.7,  
5.7,  
6.2,  
5.1,  
5.7],  
'Iris-virginica': [6.3,  
5.8,  
7.1,  
6.3,  
6.5,  
7.6,  
4.9,  
7.3,  
6.7,  
7.2,  
6.5,  
6.4,  
6.8,  
5.7,  
5.8,  
6.4,  
6.5,  
7.7,  
7.7,  
6.0,  
6.9,  
5.6,  
7.7,  
6.3,  
6.7,  
7.2,  
6.2,  
6.1,  
6.4,  
7.2,  
7.4,  
7.9,  
6.4,  
6.3,  
6.1,  
7.7,  
6.3,  
6.4,  
6.0,
```

```
6.9,  
6.7,  
6.9,  
5.8,  
6.8,  
6.7,  
6.7,  
6.3,  
6.5,  
6.2,  
5.9]]}
```

```
#basic statistical details like percentile, mean, sd,
```

```
def basic_statistics(group):  
    return{  
        'Mean' : group.mean(),  
        'Standard Deviation' : group.std(),  
        '25th Percentile' : group.quantile(.25),  
        '50th Percentile' : group.median(),  
        '75th Percentile' : group.quantile(.75),  
        'Minimun' : group.min(),  
        'Maximum' : group.max()  
    }
```

```
#now apply this for all species
```

```
setosa_stats = basic_statistics(df[df['Species'] == 'Iris-setosa']  
['SepalLengthCm'])  
versicolor_stats = basic_statistics(df[df['Species'] == 'Iris-  
versicolor']['SepalLengthCm'])  
virginica_stats = basic_statistics(df[df['Species'] == 'Iris-  
virginica']['SepalLengthCm'])
```

```
# display stats
```

```
print("Statistics for Iris-setosa:\n", setosa_stats)  
print("\nStatistics for Iris-versicolor:\n", versicolor_stats)  
print("\nStatistics for Iris-virginica:\n", virginica_stats)
```

```
Statistics for Iris-setosa:
```

```
{ 'Mean': np.float64(5.006), 'Standard Deviation':  
np.float64(0.35248968721345136), '25th Percentile': np.float64(4.8),  
'50th Percentile': np.float64(5.0), '75th Percentile':  
np.float64(5.2), 'Minimun': np.float64(4.3), 'Maximum':  
np.float64(5.8)}
```

```
Statistics for Iris-versicolor:
```

```
{ 'Mean': np.float64(5.936), 'Standard Deviation':  
np.float64(0.5161711470638634), '25th Percentile': np.float64(5.6),  
'50th Percentile': np.float64(5.9), '75th Percentile':  
np.float64(6.3), 'Minimun': np.float64(4.9), 'Maximum':
```

```
np.float64(7.0)}
```

```
Statistics for Iris-virginica:
```

```
{ 'Mean': np.float64(6.587999999999998), 'Standard Deviation':  
np.float64(0.6358795932744322), '25th Percentile': np.float64(6.225),  
'50th Percentile': np.float64(6.5), '75th Percentile':  
np.float64(6.9), 'Minimun': np.float64(4.9), 'Maximum':  
np.float64(7.9)}
```

```
# for age-income dataset
```

```
data = {  
    'Age' : [22,25,47,52,46,56,55,60,23,34,44,53],  
    'Income': [25000, 48000, 52000, 58000, 60000, 62000, 61000,  
70000, 26000, 40000, 52000, 58000],  
    'AgeGroup': ['Young', 'Young', 'Middle-aged', 'Middle-aged',  
'Middle-aged', 'Senior', 'Senior', 'Senior', 'Young', 'Young',  
'Middle-aged', 'Senior']  
}
```

```
df = pd.DataFrame(data)
```

```
df
```

	Age	Income	AgeGroup
0	22	25000	Young
1	25	48000	Young
2	47	52000	Middle-aged
3	52	58000	Middle-aged
4	46	60000	Middle-aged
5	56	62000	Senior
6	55	61000	Senior
7	60	70000	Senior
8	23	26000	Young
9	34	40000	Young
10	44	52000	Middle-aged
11	53	58000	Senior

```
# Step 1a : grouped summary statistics
```

```
grouped_summary = df.groupby('AgeGroup')  
['Income'].agg(['mean','median','min','max','std'])  
print('\n---- Grouped Summary Statistics (Income grouped by AgeGroup)  
----\n')  
print(grouped_summary)
```

```
---- Grouped Summary Statistics (Income grouped by AgeGroup) ----
```

	mean	median	min	max	std
AgeGroup					
Middle-aged	55500.0	55000.0	52000	60000	4123.105626

Senior	62750.0	61500.0	58000	70000	5123.475383
Young	34750.0	33000.0	25000	48000	11176.612486

*# Step 1b: list of Income values for each agegroup*

```
grouped_list = df.groupby('AgeGroup')['Income'].apply(list).to_dict()
```

```
print("\n--- List of Incomes per AgeGroup ---\n")
for group, values in grouped_list.items():
    print(f"{group} : {values}\n")
```

--- List of Incomes per AgeGroup ---

Middle-aged : [52000, 58000, 60000, 52000]

Senior : [62000, 61000, 70000, 58000]

Young : [25000, 48000, 26000, 40000]

```
df.describe()
```

	Age	Income
count	12.000000	12.000000
mean	43.083333	51000.000000
std	13.667868	14122.837727
min	22.000000	25000.000000
25%	31.750000	46000.000000
50%	46.500000	55000.000000
75%	53.500000	60250.000000
max	60.000000	70000.000000

```
setosa = ds[ds['Species']=='Iris-setosa']
versicolor = ds[ds['Species'] == 'Iris-versicolor']
virginica = ds[ds['Species']=='Iris-virginica']
```

```
print("\n --- Statistical Details for Iris-Setosa ---\n")
print(setosa.describe())
```

--- Statistical Details for Iris-Setosa ---

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.000000	50.000000	50.000000	50.000000	50.000000
mean	25.500000	5.006000	3.418000	1.464000	0.244000
std	14.57738	0.35249	0.381024	0.173511	0.10721

min	1.00000	4.30000	2.300000	1.000000
0.10000				
25%	13.25000	4.80000	3.125000	1.400000
0.20000				
50%	25.50000	5.00000	3.400000	1.500000
0.20000				
75%	37.75000	5.20000	3.675000	1.575000
0.30000				
max	50.00000	5.80000	4.400000	1.900000
0.60000				

```
print("\n--- Statistical Details for Iris-Versicolor ---\n")
print(versicolor.describe())
```

```
print("\n--- Statistical Details for Iris-Virginica ---\n")
print(virginica.describe())
```

```
--- Statistical Details for Iris-Versicolor ---
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.000000	50.000000	50.000000	50.000000
mean	75.50000	5.936000	2.770000	4.260000	1.326000
std	14.57738	0.516171	0.313798	0.469911	0.197753
min	51.00000	4.900000	2.000000	3.000000	1.000000
25%	63.25000	5.600000	2.525000	4.000000	1.200000
50%	75.50000	5.900000	2.800000	4.350000	1.300000
75%	87.75000	6.300000	3.000000	4.600000	1.500000
max	100.00000	7.000000	3.400000	5.100000	1.800000

```
--- Statistical Details for Iris-Virginica ---
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.000000	50.000000	50.000000	50.000000
mean	125.50000	6.58800	2.974000	5.552000	2.02600
std	14.57738	0.63588	0.322497	0.551895	0.27465
min	101.00000	4.90000	2.200000	4.500000	



1.40000				
25%	113.25000	6.22500	2.800000	5.100000
1.80000				
50%	125.50000	6.50000	3.000000	5.550000
2.00000				
75%	137.75000	6.90000	3.175000	5.875000
2.30000				
max	150.00000	7.90000	3.800000	6.900000
2.50000				