

Ironhack Student Portal

 preview.my.ironhack.com/lms/courses/course-

Lesson Goals

In this lesson we will learn:

- Grouping data in Pandas
- Using the aggregation functions to summarize grouped data

Introduction

Aggregating and summarizing are essential tools in data analysis. They allow us to perform computations on our data or look at descriptive statistics for subsets of the data. These calculations can help us make meaningful inference regarding our data.

We will use the `vehicles.csv` data set you used in Module 1. In case you don't have the data set handy, download it again from [here](#). Extract the content of the downloaded file to your machine. `vehicles.csv` is contained in the extracted folder.

Grouping

We have looked at the group by clause in SQL in previous lessons. Pandas has a similar function that enables us to perform aggregations - the `groupby` function.

Applying the `groupby` function to a DataFrame will return a DataFrameGroupBy object. We then specify the columns that we intend to group on.

Recall the vehicles dataset from previous lessons:

```
import numpy as np
import pandas as pd

vehicles = pd.read_csv('vehicles.csv')
vehicles.groupby(['Transmission'])
<pandas.core.groupby.groupby.DataFrameGroupBy object at 0x00000177FBAB0F98>
```

This object contains information that can be "unleashed" when an aggregation is applied to this object.

Aggregations

We can apply different aggregation functions to our grouped data. We can use some standard functions or define our own functions and then apply them to the aggregated data using the `agg` function.

Some standard aggregation functions are: `mean` , `sum` , `count` , `median` , `min` , `max` , `std` .

We can also use the `agg` function to apply multiple aggregations at once to all columns specified.

After aggregating, we can subset the data to only apply the aggregation to the columns that we choose.

Here are some examples of standard aggregation functions:

```
vehicles.groupby(['Transmission'])['Highway MPG', 'City MPG', 'Combined MPG'].mean()

      Highway MPG      City MPG      Combined MPG
Transmission
Auto (AV)      40.000000      35.000000      37.000000
Auto (AV-S6)    25.000000      22.000000      23.000000
Auto (AV-S8)    22.000000      20.000000      21.000000
Auto(A1)        37.000000      41.000000      39.000000
Auto(AM-S6)     32.978261      24.315217      27.554348
...           ...           ...
Manual 5 spd    14.000000      14.000000      14.000000
Manual 5-spd    25.664312      19.242327      21.634391
Manual 6-spd    26.202229      18.306232      21.153941
Manual 7-spd    26.205882      18.220588      21.117647
Manual(M7)      22.333333      14.000000      17.000000
45 rows Γ 3 columns
```

```
vehicles.groupby(['Fuel Type', 'Cylinders'])['CO2 Emission Grams/Mile'].median()

Fuel Type      Cylinders
CNG            4.0      253.197321
              6.0      417.030882
              8.0      568.070913
Diesel         4.0      308.484848
              5.0      391.538462
              ...
Regular        8.0      634.785714
              10.0     776.500000
              12.0     683.615385
Regular Gas and Electricity 4.0      129.000000
Regular Gas or Electricity  4.0      51.000000
Name: CO2 Emission Grams/Mile, Length: 48, dtype: float64
```

```
vehicles.groupby(['Fuel Type'])['Combined MPG'].agg(['mean', 'median', 'std'])

      mean      median      std
Fuel Type
CNG      18.133333      14.5      7.436663
Diesel    23.488474      21.0      7.054702
Gasoline or E85 17.572385      17.0      3.822538
Gasoline or natural gas 15.350000      12.0      5.343712
Gasoline or propane 13.500000      13.5      1.603567
...           ...           ...
Premium and Electricity 26.300000      25.5      5.141165
Premium or E85          20.090909      20.0      3.676502
Regular                 20.144698      20.0      5.317500
Regular Gas and Electricity 41.937500      38.5      5.246824
Regular Gas or Electricity 42.000000      42.0      0.000000
13 rows Γ 3 columns
```

Custom Aggregation Functions

We do not have to be limited by the range of standard aggregation functions. If the need arises, we can write our own aggregation function.

For example, in our vehicle dataset, we might want to find out for each level of transmission, what is the most common vehicle class. In other words, we would like to find the mode.

We can write our own implementation of the mode function, but it would be more efficient to use the scipy implementation of this function. Scipy is a Python package for scientific computing.

Let us first define our custom function using the scipy mode function. We create a custom function since the mode function returns a tuple with the mode and the frequency of the mode. We are only interested in the first part of the tuple.

```
from scipy import stats

def agg_mode(x):
    return(stats.mode(x)[0])
```

Now we can use our custom aggregation function using the `agg` function:

```
vehicles.groupby("Transmission")["Vehicle Class"].agg(agg_mode)
Transmission
Auto (AV)          Compact Cars
Auto (AV-S6)       Compact Cars
Auto (AV-S8)       Midsize Cars
Auto(A1)           Subcompact Cars
Auto(AM-S6)        Compact Cars
...
Manual 5 spd              Vans
Manual 5-spd             Compact Cars
Manual 6-spd             Compact Cars
Manual 7-spd      Minicompact Cars
Manual(M7)              Two Seaters
Name: Vehicle Class, Length: 45, dtype: object
```

Summary

In this lesson we learned how to summarization and aggregation with DataFrames. We learned to use the standard aggregation functions and how to make custom aggregation functions.