hw5

October 5, 2023

1 Homework 5

1.1 Question 1 (5 points)

- Read the `mpg` dataset in a `pandas` data frame.
- Execute the `describe` method of the data frame you created.
- What function in R corresponds to `describe`?

[3]:
```
#!pandoc --version
```

[77]:
```
import pandas as pd
mpg=pd.read_csv('mpg.csv')
mpg.describe()
```

[77]:

	displ	year	cyl	cty	hwy
count	234.000000	234.000000	234.000000	234.000000	234.000000
mean	3.471795	2003.500000	5.888889	16.858974	23.440171
std	1.291959	4.509646	1.611534	4.255946	5.954643
min	1.600000	1999.000000	4.000000	9.000000	12.000000
25%	2.400000	1999.000000	4.000000	14.000000	18.000000
50%	3.300000	2003.500000	6.000000	17.000000	24.000000
75%	4.600000	2008.000000	8.000000	19.000000	27.000000
max	7.000000	2008.000000	8.000000	35.000000	44.000000

1.1.1 What function in R corresponds to describe?

Summary built in function

1.2 Question 2 (25 points)

This will make you miss the `ggplot2` library! :D

Modify the following plot

```
_ = sns.lmplot(
    data = mpg,
    x = "displ",
    y = "hwy",
    ci = None,
    line_kws = {"color": "black"},
```

```
    scatter_kws = {"s": 10},
    lowess = True,
    x_jitter = 0.5,
    y_jitter = 0.5
).set(
    title = "Fuel consumption by engine displacement",
    xlabel = "Engine displacement",
    ylabel = "Fuel consumption"
).tight_layout()
```

in such a way that:

1. the points in the plot are colored by the variable `class` in the `mpg` dataset
2. there is a single trend line applied that applies to all points in the plot instead of one trend line for each group defined by the `class` variable in the `mpg` dataset.

Hint: - First, create a `lmplot` and set in that plot `hue = "class"` (to color the points by the "class" variable) and `fit_reg = False` (to suppress the trend line); assign the output of the `lmplot` function to a variable named `lm_plot` (you'll need this in a second). - Next, in the same cell, create a `regplot` and set `scatter = False` (to suppress plotting the points, which are already produced by the previous plot) as well as `ax = lm_plot.ax` (to tell `seaborn` to overlay the new plot to the previous one, rather than creating a new plot). - You should still pass relevant parameters to each of the two components of the plot (`lmplot` and `regplot`) in a sensible way so to get the desired final plot as the result.

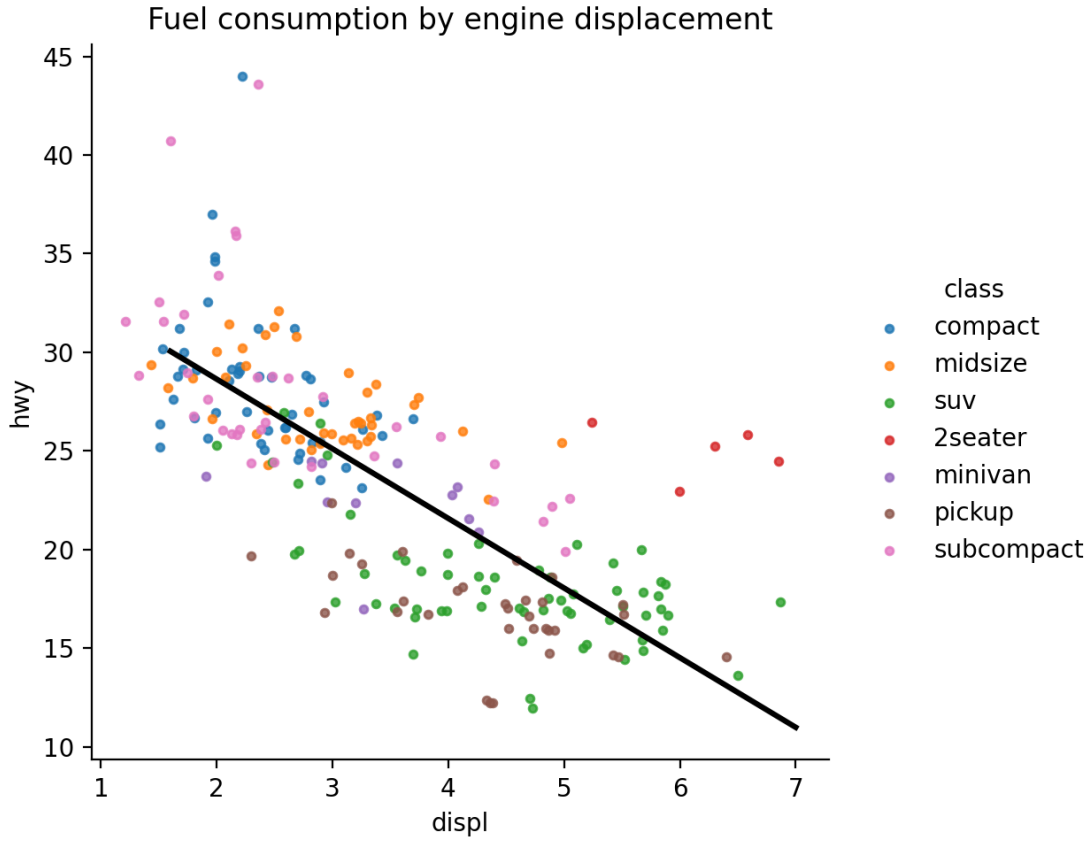
Here's a "skeleton" of the code that you'll need. It's up to you to fill it out!

```
lm_plot = sns.lmplot(
    ... <- configure the lmplot appropriately
).set(
    title = "Fuel consumption by engine displacement",
    xlabel = "Engine displacement",
    ylabel = "Fuel consumption"
).tight_layout()

_ = sns.regplot(
    ... <- configure the regplot appropriately
)
```

[78]:
```
import seaborn as sns
import pandas as pd
import matplotlib.pyplot as plt
lm_plot = sns.lmplot(
    data = mpg,
    x = "displ",
    y = "hwy",
    ci = None,
    line_kws = {"color": "black"},
    scatter_kws = {"s": 10},
```

```
    hue="class",
    fit_reg=False,
    lowess = True,
    x_jitter = 0.5,
    y_jitter = 0.5
).set(
    title = "Fuel consumption by engine displacement",
    xlabel = "Engine displacement",
    ylabel = "Fuel consumption"
).tight_layout()
s=sns.regplot(
    data = mpg,
    x = "displ",
    y = "hwy",
    scatter=False,
    line_kws = {"color": "black"},
    ax=lm_plot.ax,ci=None
)
plt.show()
```



[ ]:
```
#!pip install seaborn==0.12.2
```

1.3 Question 3 (20 points)

Use the `plotnine` library to recreate this `ggplot2` plot (from lab 1) in Python:

```
ggplot(data = diamonds, mapping = aes(x = color, fill = clarity)) +
  geom_bar(position = "dodge") +
  scale_fill_brewer(palette = "Set2") +
  labs(
    x = "Color",
    y = element_blank(),
    fill = "Diamond clarity",
    title = "Clarity distribution by color"
  )
```

Hint: this stackoverflow discussion might come in handy to set the palette colors. Also, note that the variable `clarity` is actually an ordered categorical variable (to see this, read the docs in R for the `diamonds` dataset with `? diamonds`). We can create a `pandas` equivalent of an R factor by using the `pandas.Categorical` class (see the docs here). To get the same ordering that we got in R for the dodged bars, you will need to transform the `clarity` variable in the `diamonds` dataset that you read with `pandas` to a `pandas.Categorical` variable.

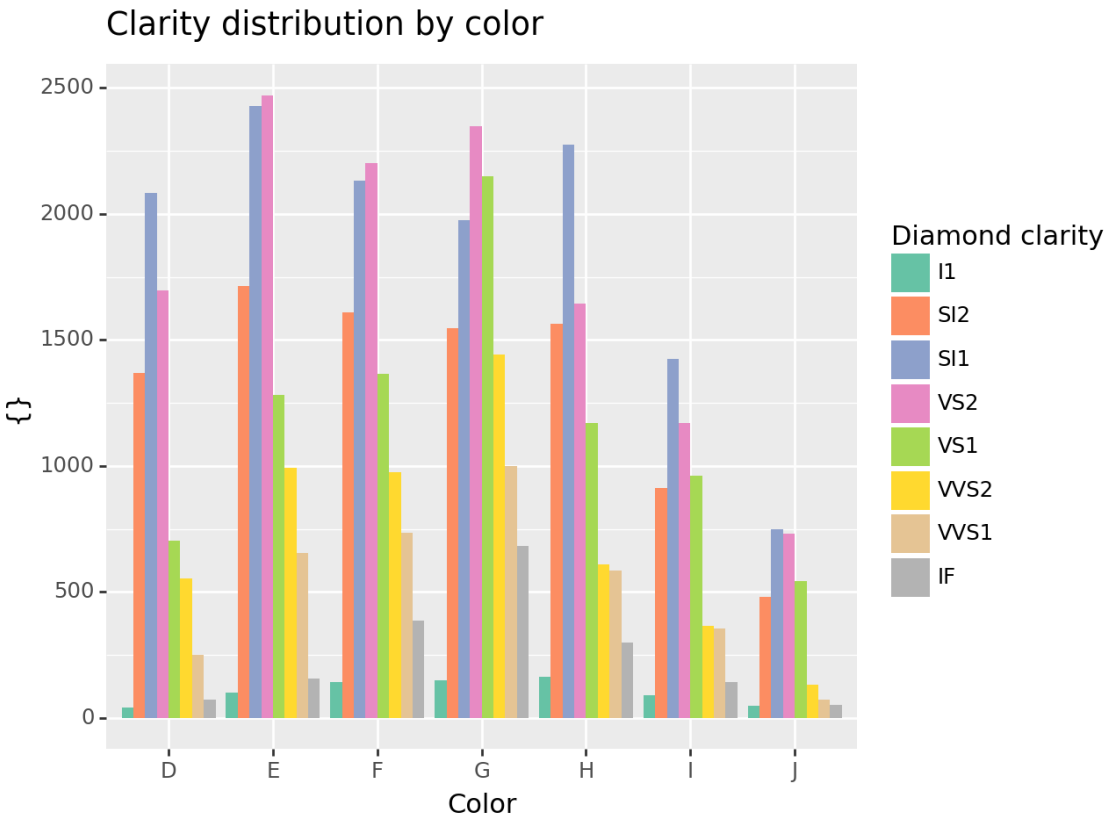
[4]:
```
#!pip install plotnine
```

[6]:
```
from plotnine import *
import pandas as pd
```

[7]:
```
diamonds=pd.read_csv('diamonds.csv')
```

[24]:
```
# Define the desired order for clarity levels
clarity_order = ['I1', 'SI2', 'SI1', 'VS2', 'VS1', 'VVS2', 'VVS1', 'IF']
diamonds['clarity'] = pd.Categorical(diamonds['clarity'],␣
 ↪categories=clarity_order, ordered=True)

(ggplot(data = diamonds, mapping = aes(x = 'color', fill = 'clarity')) +
  geom_bar(position = "dodge") +
  scale_fill_brewer(type="qual", palette = "Set2") +
  labs(
    x = "Color",
    y = element_blank(),
    fill = "Diamond clarity",
    title = "Clarity distribution by color"
  ))
```



[24]:
```
<Figure Size: (640 x 480)>
```

[6]:
```
#!pip install plotnine
```

1.4 Question 4 (25 points)

Solve again exercises 1 and 4 of Homework 2, but this time use `pandas`!

You are encouraged to use the published solutions of Homework 2 and "translate" them from R's `tidyverse`/`dplyr` to Python's `pandas`.

You might find this "translation guide" helpful!

[40]:
```
# Load the flights and airlines datasets (assuming you have them as CSV files)
flights = pd.read_csv('flights.csv')  # Replace with your dataset filename
airlines = pd.read_csv('airlines.csv')  # Replace with your dataset filename

# Filter flights that occurred between July and November
flights = flights[(flights['month'] >= 7) & (flights['month'] <= 11)]

# Calculate the gain
flights['gain'] = flights['arr_delay'] - flights['dep_delay']
```

```
# Group by airline and compute average gain (ignoring NAs)
#Note: missing values are automatically skipped when calculating the mean. You␣
 ↪do not need to specify a parameter to skip them.
average_gain = flights.groupby('carrier')['gain'].mean().reset_index()

# Sort by average_gain in decreasing order
average_gain = average_gain.sort_values(by='gain', ascending=False)

# Join with the airlines dataset
average_gain = average_gain.merge(airlines, left_on='carrier',␣
 ↪right_on='carrier', how='left')

result_tibble = average_gain[['gain', 'name']]
# Rename the columns to match the R solution
result_tibble.columns = ['average_gain', 'name']

# Display the resulting tibble
print(result_tibble)
```

```
    average_gain                         name
0       1.285235  AirTran Airways Corporation
1       0.545455       Frontier Airlines Inc.
2       0.063590                    Envoy Air
3      -0.909774       Hawaiian Airlines Inc.
4      -2.837182              US Airways Inc.
5      -2.846154        SkyWest Airlines Inc.
6      -3.017921           Mesa Airlines Inc.
7      -4.905070              JetBlue Airways
8      -5.787007     ExpressJet Airlines Inc.
9      -7.960461         Delta Air Lines Inc.
10     -8.571100       Southwest Airlines Co.
11     -8.669899       American Airlines Inc.
12     -9.621162        United Air Lines Inc.
13     -9.875375               Virgin America
14    -10.288170           Endeavor Air Inc.
15    -22.211409         Alaska Airlines Inc.
```

[44]:
```
# Read the transactions.csv dataset
transactions = pd.read_csv("transactions.csv")

# Reshape the data to be tidy
transactions = pd.melt(transactions, id_vars=['person', 'month'],␣
 ↪var_name='transaction_type', value_name='amount')

# Filter the data for Jenna or John
transactions = transactions[transactions['person'].isin(['jenna', 'john'])]
```

```
# Get the unique months in which either Jenna or John had a transaction
unique_months = transactions['month'].unique()

# Create a pandas DataFrame with the unique months
result_tibble = pd.DataFrame({'month': unique_months})

# Display the resulting tibble
print(result_tibble)
```

```
   month
0      1
1      3
```

1.5 Question 5 (25 points)

Solve again exercises 1 and 2 of Homework 3, but this time use `pandas`!

You are encouraged to use the published solutions of Homework 3 and "translate" them from R's `tidyverse`/`dplyr` to Python's `pandas`.

Again, you might find this "translation guide" helpful!

[53]:
```
# Read the datasets
dinners = pd.read_csv("dinners.csv")
drinks = pd.read_csv("drinks.csv")
food = pd.read_csv("food.csv")
persons = pd.read_csv("persons.csv")

# Merge the datasets to create the dinners_explicit tibble
dinners_explicit = pd.merge(dinners, drinks, left_on="drink_id",␣
 ↪right_on="item_id", how="left")
dinners_explicit = dinners_explicit.rename(columns={"price": "drink_price",␣
 ↪"item_name": "drink"})

dinners_explicit = pd.merge(dinners_explicit, food, left_on="food_id",␣
 ↪right_on="food_id", how="left")
dinners_explicit = dinners_explicit.rename(columns={"price": "food_price",␣
 ↪"name": "food"})

dinners_explicit = pd.merge(dinners_explicit, persons, left_on="person_id",␣
 ↪right_on="id", how="left")

# Select the specified columns
dinners_explicit = dinners_explicit[["drink", "drink_price", "food",␣
 ↪"food_price", "first_name", "last_name", "age"]]

# Display the resulting tibble
```

```
dinners_explicit
```

[53]:

	drink	drink_price	food	food_price	first_name	last_name	\
0	NaN	NaN	pasta	$8.50	Valter	Evangelista	
1	NaN	NaN	ice cream	$4.50	Polly	Verity	
2	water	$1.00	NaN	NaN	NaN	NaN	
3	beer	$5.00	cake	$4.50	Aysha	Freitas	
4	NaN	NaN	pizza	$12	Rayno	Van Kann	
5	water	$1.00	fish	$15.00	Valter	Evangelista	
6	NaN	NaN	pizza	$12	Rayno	Van Kann	
7	sparkling water	$2.00	ice cream	$4.50	Ksenya	Dunai	
8	soda	$2.50	pop corn	$1.50	Polly	Verity	
9	water	$1.00	salad	$5.00	Aysha	Freitas	
10	wine	$9.00	NaN	NaN	NaN	NaN	
11	soda	$2.50	salad	$5.00	NaN	NaN	
12	beer	$5.00	cake	$4.50	Aysha	Freitas	
13	NaN	NaN	steak	$12.00	NaN	NaN	
14	wine	$9.00	NaN	NaN	Valter	Evangelista	
15	wine	$9.00	pop corn	$1.50	NaN	NaN	
16	NaN	NaN	fish	$15.00	NaN	NaN	
17	water	$1.00	fries	$3.00	Polly	Verity	
18	sparkling water	$2.00	burger	$5.00	Polly	Verity	
19	soda	$2.50	steak	$12.00	Polly	Verity	
20	NaN	NaN	fries	$3.00	Valter	Evangelista	
21	NaN	NaN	NaN	NaN	NaN	NaN	
22	NaN	NaN	pasta	$8.50	NaN	NaN	
23	soda	$2.50	pasta	$8.50	NaN	NaN	
24	sparkling water	$2.00	salad	$5.00	NaN	NaN	
25	sparkling water	$2.00	pizza	$12	Rayno	Van Kann	
26	sparkling water	$2.00	ice cream	$4.50	NaN	NaN	
27	sparkling water	$2.00	fish	$15.00	NaN	NaN	
28	NaN	NaN	fries	$3.00	Polly	Verity	
29	beer	$5.00	pasta	$8.50	Valter	Evangelista	
30	water	$1.00	burger	$5.00	NaN	NaN	
31	NaN	NaN	pasta	$8.50	Rayno	Van Kann	
32	NaN	NaN	salad	$5.00	Aysha	Freitas	
33	NaN	NaN	fries	$3.00	NaN	NaN	
34	soda	$2.50	cake	$4.50	NaN	NaN	
35	sparkling water	$2.00	fish	$15.00	NaN	NaN	
36	soda	$2.50	NaN	NaN	NaN	NaN	
37	beer	$5.00	pop corn	$1.50	NaN	NaN	
38	wine	$9.00	fish	$15.00	NaN	NaN	
39	NaN	NaN	salad	$5.00	NaN	NaN	
40	soda	$2.50	NaN	NaN	NaN	NaN	
41	NaN	NaN	fries	$3.00	NaN	NaN	
42	wine	$9.00	pizza	$12	Polly	Verity	
43	soda	$2.50	NaN	NaN	Aysha	Freitas	

```
44            water      $1.00       pizza       $12         NaN          NaN
45            water      $1.00       fries     $3.00       Polly       Verity
46              NaN         NaN   pop corn     $1.50         NaN          NaN
47             soda      $2.50       pizza       $12       Rayno     Van Kann
48  sparkling water      $2.00       pizza       $12       Aysha      Freitas
49              NaN         NaN      steak    $12.00       Polly       Verity
```

```
     age
0   34.0
1   61.0
2    NaN
3   55.0
4   29.0
5   34.0
6   29.0
7   31.0
8   61.0
9   55.0
10   NaN
11   NaN
12  55.0
13   NaN
14  34.0
15   NaN
16   NaN
17  61.0
18  61.0
19  61.0
20  34.0
21   NaN
22   NaN
23   NaN
24   NaN
25  29.0
26   NaN
27   NaN
28  61.0
29  34.0
30   NaN
31  29.0
32  55.0
33   NaN
34   NaN
35   NaN
36   NaN
37   NaN
38   NaN
```

```
39   NaN
40   NaN
41   NaN
42  61.0
43  55.0
44   NaN
45  61.0
46   NaN
47  29.0
48  55.0
49  61.0
```

[81]:
```
# Assuming you have the 'dinners_explicit' DataFrame from the previous steps

# Create a summary tibble for the two most popular drinks
top2_drinks = dinners_explicit['drink'].value_counts().reset_index()
top2_drinks.columns = ['drink', 'n']
top2_drinks = top2_drinks.head(2)

# Create a summary tibble for the two most popular foods
top2_foods = dinners_explicit['food'].value_counts().reset_index()
top2_foods.columns = ['food', 'n']
top2_foods = top2_foods.head(2)

# Display the summary tibbles
print("Top 2 Drinks:")
top2_drinks
```

```
Top 2 Drinks:
```

[81]:

	drink	n
0	soda	9
1	sparkling water	8

[82]:
```
print("\nTop 2 Foods:")
top2_foods
```

```
Top 2 Foods:
```

[82]:

	food	n
0	pizza	7
1	fries	6

[5]:
```
#!pip install nbconvert
```