## Introduction to Data Visualization with Python 3). Statistical plots with Seaborn

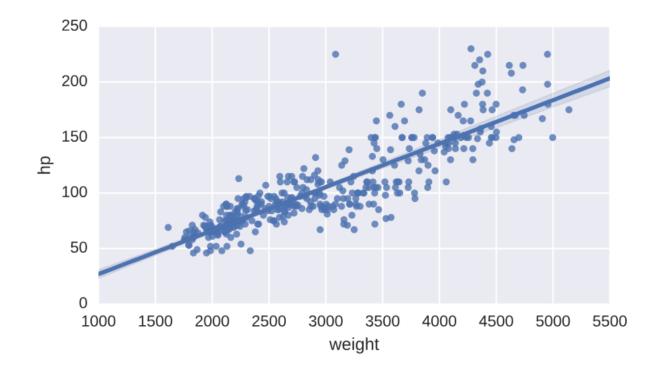
a).

# Import plotting modules

import matplotlib.pyplot as plt

import seaborn as sns

# Plot a linear regression between 'weight' and 'hp' sns.lmplot(x='weight', y='hp', data=auto)

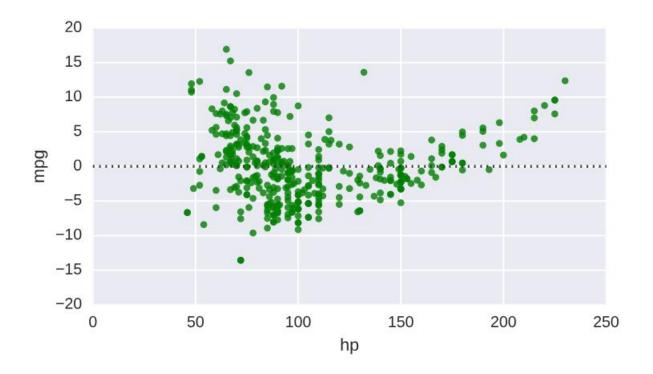


Chapter 3

b).

# Import plotting modules import matplotlib.pyplot as plt import seaborn as sns

# Generate a green residual plot of the regression between 'hp' and 'mpg' sns.residplot(x='hp', y='mpg', data=auto, color='green')



```
c).
```

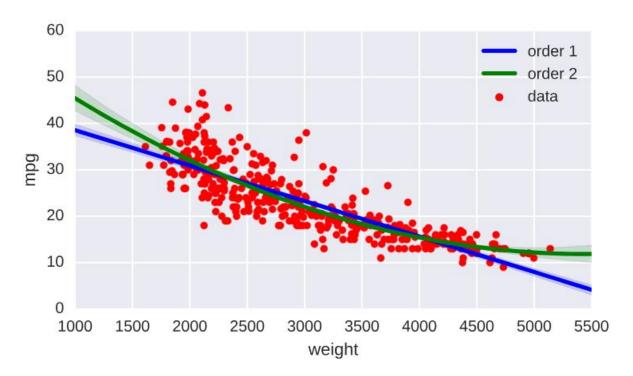
# Generate a scatter plot of 'weight' and 'mpg' using red circles

plt.scatter(auto['weight'], auto['mpg'], label='data', color='red', marker='o')

# Plot in blue a linear regression of order 1 between 'weight' and 'mpg'
sns.regplot(x='weight', y='mpg', data=auto, scatter=None, color='blue', label='order 1')

# Plot in green a linear regression of order 2 between 'weight' and 'mpg'
sns.regplot(x='weight', y='mpg', data=auto, scatter=None, color='green', order=2,label='order 2')

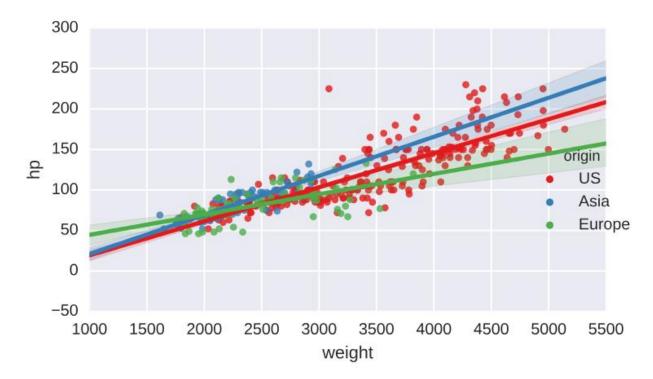
# Add a legend and display the plot plt.legend(loc='upper right') plt.show()



d).

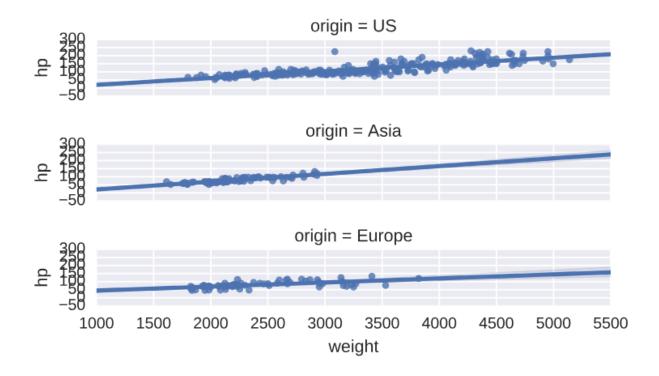
# Plot a linear regression between 'weight' and 'hp', with a hue of 'origin' and palette of 'Set1' sns.lmplot(data=auto,x='weight',y='hp',hue='origin',palette='Set1')

# Display the plot



Chapter 3

- e). Grouping linear regressions by row or column
- # Plot linear regressions between 'weight' and 'hp' grouped row-wise by 'origin' sns.lmplot(data=auto,x='weight',y='hp',row='origin')



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f). constructing strip plots

# Make a strip plot of 'hp' grouped by 'cyl'

plt.subplot(2,1,1)

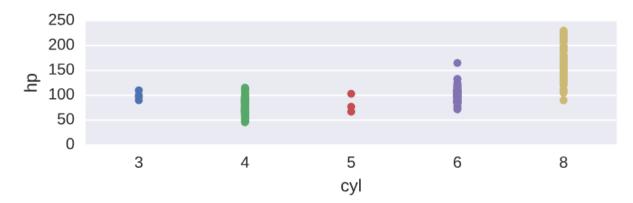
sns.stripplot(x='cyl', y='hp', data=auto)

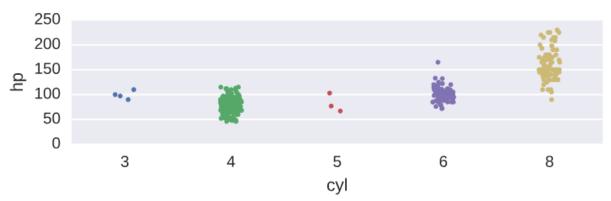
# Make the strip plot again using jitter and a smaller point size

plt.subplot(2,1,2)

sns.stripplot(x='cyl', y='hp', data=auto, jitter=True, size=3)

# Display the plot





Chapter 3

g). Constructing swarm plots

# Generate a swarm plot of 'hp' grouped horizontally by 'cyl'

plt.subplot(2,1,1)

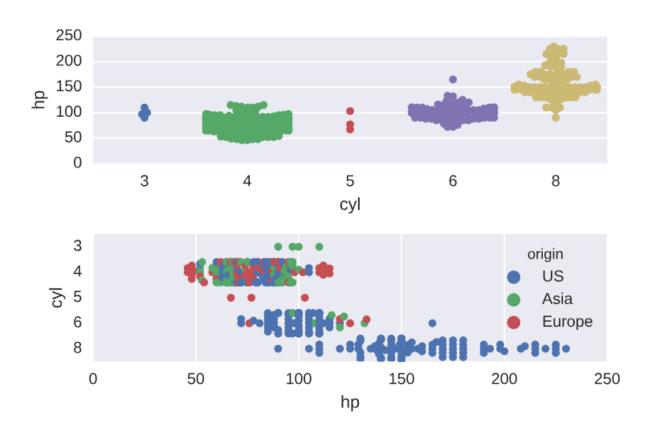
sns.swarmplot(x='cyl',y='hp',data=auto)

# Generate a swarm plot of 'hp' grouped vertically by 'cyl' with a hue of 'origin'

plt.subplot(2,1,2)

sns.swarmplot(x='hp',y='cyl',data=auto,hue='origin',orient='h')

# Display the plot



Chapter 3

h). Constructing Violon plots and overlaying plots

# Generate a violin plot of 'hp' grouped horizontally by 'cyl'

plt.subplot(2,1,1)

sns.violinplot(x='cyl', y='hp', data=auto)

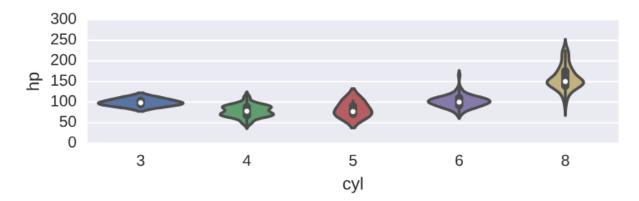
# Generate the same violin plot again with a color of 'lightgray' and without inner annotations plt.subplot(2,1,2)

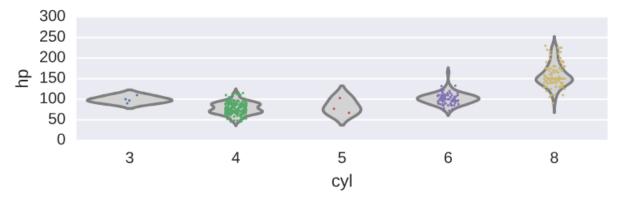
sns.violinplot(x='cyl', y='hp', data=auto,inner=None,color='lightgray')

# Overlay a strip plot on the violin plot

sns.stripplot(x='cyl',y='hp',data=auto,size=1.5,jitter=True)

# Display the plot



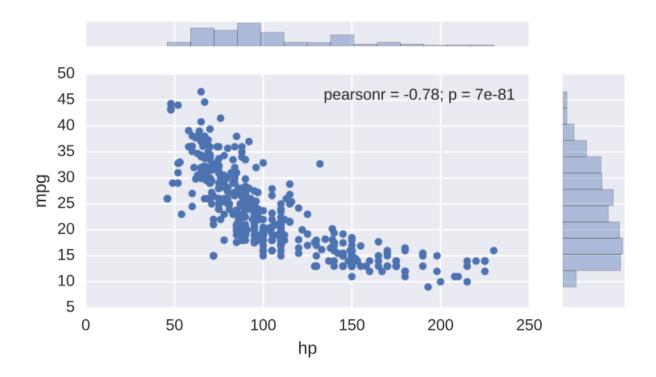


Chapter 3

i). Plotting joint distributions (1)

# Generate a joint plot of 'hp' and 'mpg'

sns.jointplot(x='hp',y='mpg', data=auto)



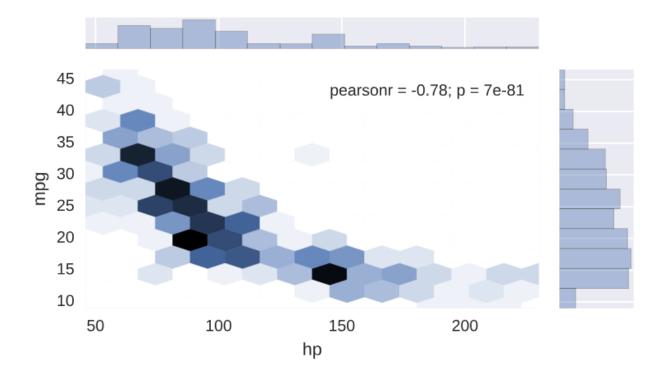
## j). Plotting joint distributions (2)

- kind='scatter' uses a scatter plot of the data points
- kind='reg' uses a regression plot (default order 1)
- kind='resid' uses a residual plot
- kind='kde' uses a kernel density estimate of the joint distribution
- kind='hex' uses a hexbin plot of the joint distribution

# Generate a joint plot of 'hp' and 'mpg' using a hexbin plot

sns.jointplot(x='hp',y='mpg',data=auto,kind='hex')

# Display the plot



Chapter 3

k). Plotting distributions pairwise (1)

# Print the first 5 rows of the DataFrame

print(auto.head())

# Plot the pairwise joint distributions from the DataFrame

sns.pairplot(auto)

# Display the plot

plt.show()

mpg hp origin

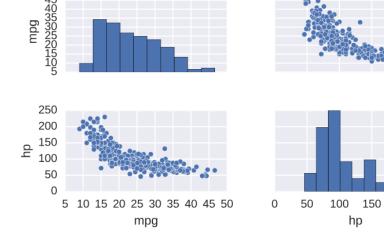
0 18.0 88 US

1 9.0 193 US

2 36.1 60 Asia

3 18.5 98 US

4 34.3 78 Europe



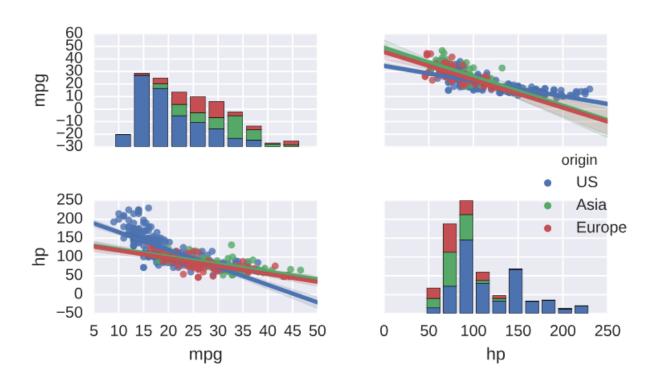
200

250

Chapter 3

- I). Plotting distributions pairwise (2)
- # Print the first 5 rows of the DataFrame
  print(auto.head())

# Plot the pairwise joint distributions grouped by 'origin' along with regression lines sns.pairplot(auto,kind='reg',hue='origin')



Chapter 3

m). Visualizing correlations with heatmap

# Print the covariance matrix
print(cov\_matrix)

# Visualize the covariance matrix using a heatmap sns.heatmap(cov\_matrix)

# Display the heatmap
plt.show()

mpg hp weight accel displ
mpg 1.000000 -0.778427 -0.832244 0.423329 -0.805127
hp -0.778427 1.000000 0.864538 -0.689196 0.897257
weight -0.832244 0.864538 1.000000 -0.416839 0.932994
accel 0.423329 -0.689196 -0.416839 1.000000 -0.543800
displ -0.805127 0.897257 0.932994 -0.543800 1.000000

