

# PYBER DATA ANALYSIS

\*Despite lower fare (6.8%) and driver (2.6%) percentages in rural cities, rural areas did yield the highest average fare amounts (\$30-45). Of note, the bulk of rural city fare averages were equal to their suburban counterparts.

\*Although urban cities boast the majority of rides per city (68.4%), their average fare amounts were the lowest of the three groups.

\*The majority of Pyber drivers work in urban cities (80.9%).

```
In [94]: %matplotlib inline
# Dependencies and Setup
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np

# File to Load (Remember to change these)
city_data_to_load = "data/city_data.csv"
ride_data_to_load = "data/ride_data.csv"

# Read the City and Ride Data

city_data_df = pd.read_csv(city_data_to_load)
ride_data_df = pd.read_csv(ride_data_to_load)

# Combine the data into a single dataset

combined_data_df = pd.merge(ride_data_df, city_data_df, how = 'outer',
                             on='city')

# Display the data table for preview
combined_data_df.head()
```

Out[94]:

	city	date	fare	ride_id	driver_count	type
0	Lake Jonathanshire	2018-01-14 10:14:22	13.83	5739410935873	5	Urban
1	Lake Jonathanshire	2018-04-07 20:51:11	31.25	4441251834598	5	Urban
2	Lake Jonathanshire	2018-03-09 23:45:55	19.89	2389495660448	5	Urban
3	Lake Jonathanshire	2018-04-07 18:09:21	24.28	7796805191168	5	Urban
4	Lake Jonathanshire	2018-01-02 14:14:50	13.89	424254840012	5	Urban

### Urban Only Data

```
In [95]: # create urban only DF

only_urban = combined_data_df.loc[combined_data_df["type"] == "Urban", :]
only_urban_grouped = only_urban.groupby(['city'])

# urban only calculations

urban_rides_per_city = only_urban_grouped['ride_id'].count()
urban_avgfare_per_city = round(only_urban_grouped['fare'].mean(), 2)
driver_count_urban = only_urban_grouped['driver_count'].count()
```

## Suburban Only Data

```
In [96]: # create suburban only DF
only_suburban = combined_data_df.loc[combined_data_df["type"] == "Suburb"]
only_sub_grouped = only_suburban.groupby(['city'])

# suburban only calculations
sub_rides_per_city = only_sub_grouped['ride_id'].count()
sub_avgfare_per_city = round(only_sub_grouped['fare'].mean(), 2)
driver_count_sub = only_sub_grouped['driver_count'].count()
```

## Rural Data Only

```
In [97]: # create rural only DF
only_rural = combined_data_df.loc[combined_data_df["type"] == "Rural", :]
only_rural_grouped = only_rural.groupby(['city'])

# rural only calculations

rural_rides_per_city = only_rural_grouped['ride_id'].count()
rural_avgfare_per_city = round(only_rural_grouped['fare'].mean(), 2)
driver_count_rural = only_rural_grouped['driver_count'].count()
```

# Bubble Plot of Ride Sharing Data

```

In [98]: # Build the scatter plots for each city types
Urban = plt.scatter(urban_rides_per_city, urban_avgfare_per_city, marker=
    edgecolors = "black", facecolors = "coral", alpha=0.50, label = 'Urban')
Suburban = plt.scatter(sub_rides_per_city, sub_avgfare_per_city, marker=
    edgecolors = "black", facecolors = "skyblue", alpha=0.50, label = 'Suburban')
Rural = plt.scatter(rural_rides_per_city, rural_avgfare_per_city, marker=
    edgecolors = "black", facecolors = "gold", alpha=0.50, label = 'Rural')

# Incorporate the other graph properties

plt.title("Pyber Ride Sharing Data (2016)")
plt.xlabel("Total Number of Rides (Per City)")
plt.ylabel("Average Fare ($)")
plt.grid()

# Create a legend

plt.legend(title="City Types")
plt.legend

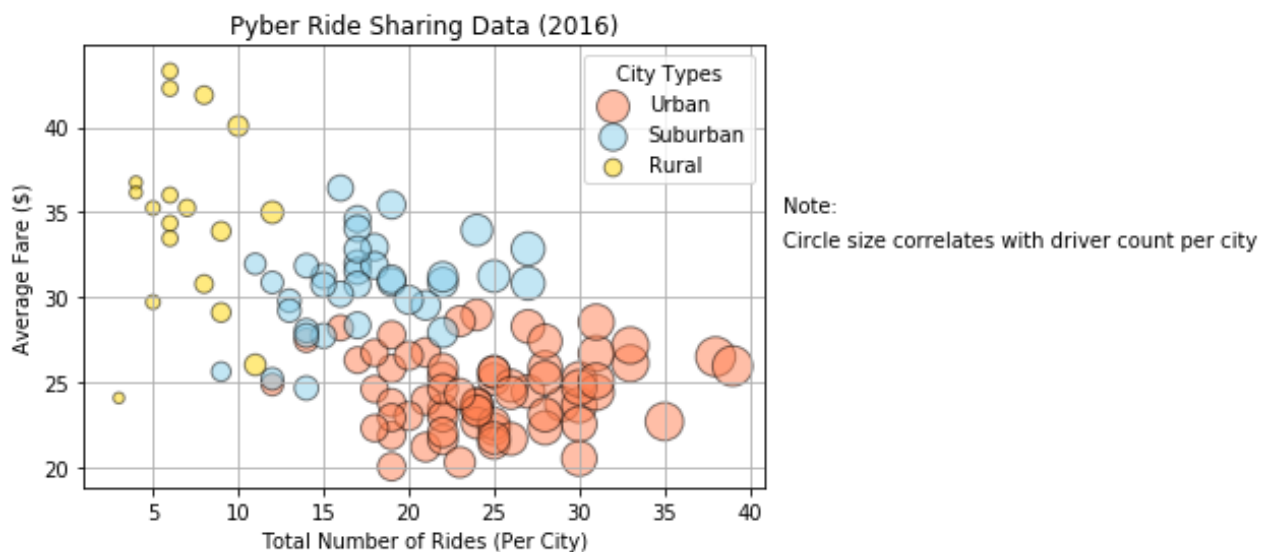
# Incorporate a text label regarding circle size

plt.text(42,35,'Note:')
plt.text(42,33,'Circle size correlates with driver count per city')

# Save Figure
plt.savefig("Images/BubblePlot.png")

# Show Figure
plt.show()

```



## Total Fares by City Type

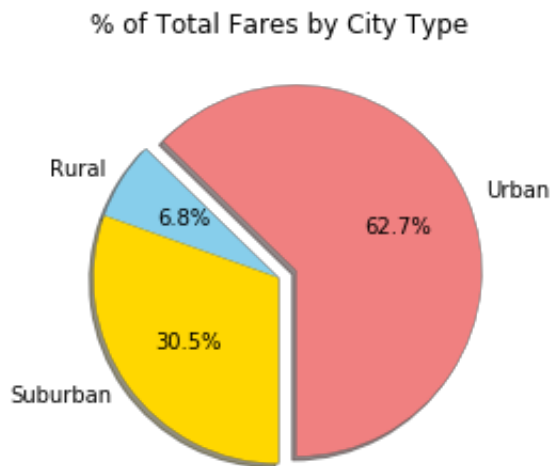
```
In [99]: # create fare DF
fare_group = combined_data_df[['city','type','fare']]
grouped_fare = fare_group.groupby('type')

# get fare calculations
fare_totals = grouped_fare['fare'].sum()
total_fare = fare_group['fare'].sum()
fare_percent = fare_totals/total_fare * 100

# Build Pie Chart
labels = ["Urban", "Rural", "Suburban"]
sizes = [62.6, 6.8, 30.4,]
colors = ["lightcoral", "skyblue", "gold"]
explode = (0.1, 0, 0)
plt.pie(sizes, explode=explode, labels=labels, colors=colors,
        autopct="%1.1f%%", shadow=True, startangle=270)
plt.title("% of Total Fares by City Type")

# Save Figure
plt.savefig("Images/TotalFaresByCity.png")

# Show Figure
plt.show()
```



## Total Rides by City Type

```
In [100]: # Calculate Ride Percents

type_count = combined_data_df['type'].value_counts()
ride_percentage = combined_data_df['type'].value_counts()/combined_data_

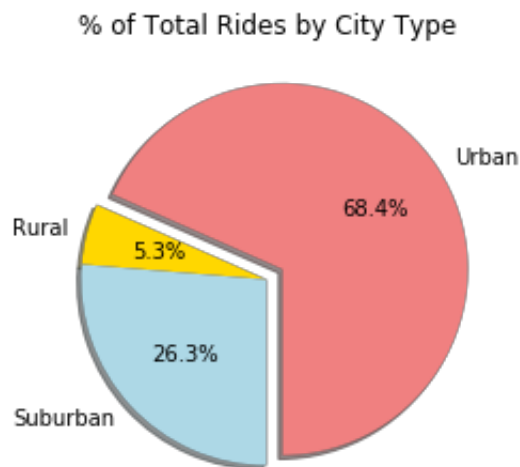
# Build Pie Chart
labels = ["Urban", "Rural", "Suburban"]
sizes = [68.4, 5.3, 26.3]
colors = ["lightcoral", "gold", "lightblue"]
explode = (0.1, 0, 0)
plt.pie(sizes, explode=explode, labels=labels, colors=colors,
        autopct="%1.1f%%", shadow=True, startangle=270)

plt.title("% of Total Rides by City Type")

# Save Figure
plt.savefig("Images/TotalRidesByCity.png")

# Show Figure

plt.show()
```



## Total Drivers by City Type

```
In [101]: # create driver DF
driver_group = combined_data_df[['city', 'type', 'driver_count']]
unique_driver_group = driver_group.drop_duplicates()
grouped_driver = unique_driver_group.groupby('type')

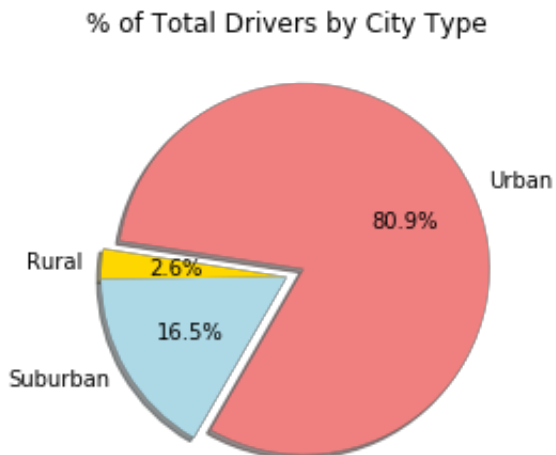
# Calculate Driver Percents
group_totals = grouped_driver['driver_count'].sum()
driver_total = unique_driver_group['driver_count'].sum()
driver_percent = group_totals/driver_total * 100

# Build Pie Charts

labels = ["Urban", "Rural", "Suburban"]
sizes = [80.9, 2.6, 16.5]
colors = ["lightcoral", "gold", "lightblue"]
explode = (0.1, 0, 0)
plt.pie(sizes, explode=explode, labels=labels, colors=colors,
        autopct="%1.1f%%", shadow=True, startangle=240)

plt.title("% of Total Drivers by City Type")

# # Save Figure
plt.savefig("Images/TotalDriversByCity.png")
plt.show()
```



In [ ]: