Pyber Ridesharing Company Analysis by Alex Koynoff

#### Observations:

Observation 1: Based on the source data, as shown in the scatter plot, we can conclude that average fare is lowest in the Urban cities, while Rural cities have the highest average fares, with Suburban cities in the middle. However, Urban cities have the highest volume of rides. An explanation of that could be that in Urban cities, there are alot more customers, thus the higher volume of rides, however, the distances in the Urban cities are alot shorter compared to Rural and Suburban, thus the lower average fare price. Even though distance is not part of the dataset provided, an assumption could be made that distances within an Urban city are alot shorter compared to rural areas. Another assumption that could explain the the lower average fare in the Urban cities is that there are more transportation options such as public transit, regular taxis, and other ride sharing companies. That could force Pyber to lower their fair prices in Urban cities to provide an incentive for customers to use their service.

Observation 2: When comparing the Total Fares by City Type vs Total Drivers by City Type, it shows that even though Suburban drivers make up 16.5% of all drivers, they bring in 30.5% of total fares, based on the datasets provided.

Observation 3: While the analysis shows that the Urban cities bring in the most revenue for the Pybar company based on the total Fares by City Type, it might not be very lucrative for the drivers. Urban cities have the highest number of drivers at 80.9%, however, they only bring in 62.7% of total fares. That could mean that on average, Urban drivers make less salary compared to Suburban drivers for example, while also needing to do more rides, thus potentially decreasing the lifecycle of their vehicle. This could cause more repairs and maintenance needed, which can mean that drivers in Urban areas would have higher expenses than Suburban and Rural drivers.

```
In [103]: %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_df = pd.read_csv(city_data_to_load)
    ride_df = pd.read_csv(ride_data_to_load)
```

In [104]: # Display the data table for preview city\_df.head()

## Out[104]:

	city	driver_count	type
0	Richardfort	38	Urban
1	Williamsstad	59	Urban
2	Port Angela	67	Urban
3	Rodneyfort	34	Urban
4	West Robert	39	Urban

In [105]: # Display the data table for preview ride\_df.head()

### Out[105]:

	city	date	fare	ride_id
0	Lake Jonathanshire	2018-01-14 10:14:22	13.83	5739410935873
1	South Michelleport	2018-03-04 18:24:09	30.24	2343912425577
2	Port Samanthamouth	2018-02-24 04:29:00	33.44	2005065760003
3	Rodneyfort	2018-02-10 23:22:03	23.44	5149245426178
4	South Jack	2018-03-06 04:28:35	34.58	3908451377344

```
In [106]: # Combine the data into a single dataset using left join
    merge_data = pd.merge(ride_df,city_df, on="city",how="left")

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

### Out[106]:

	city	date	fare	ride_id	driver_count	type
0	Lake Jonathanshire	2018-01-14 10:14:22	13.83	5739410935873	5	Urban
1	South Michelleport	2018-03-04 18:24:09	30.24	2343912425577	72	Urban
2	Port Samanthamouth	2018-02-24 04:29:00	33.44	2005065760003	57	Urban
3	Rodneyfort	2018-02-10 23:22:03	23.44	5149245426178	34	Urban
4	South Jack	2018-03-06 04:28:35	34.58	3908451377344	46	Urban

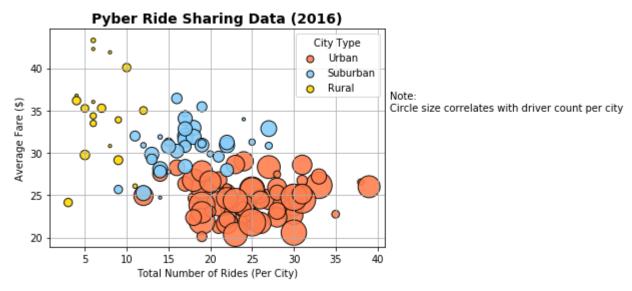
# **Bubble Plot of Ride Sharing Data**

```
In [107]: #create data frames for each city type then group by city
          urban = merge data[merge data["type"] =="Urban"].groupby([merge data["city"]])
          suburban = merge data[merge data["type"] =="Suburban"].groupby([merge data["city"]])
          rural = merge data[merge data["type"] =="Rural"].groupby([merge data["city"]])
          #calculate the values for the x \& y axis. also calculate a value to use for the size of the bubbles
          urban x = urban["ride id"].count()
          urban y = urban["fare"].mean()
          urban size = urban["driver count"].mean()
          suburban x = suburban["ride id"].count()
          suburban y = suburban["fare"].mean()
          suburban size = suburban["driver count"].mean()
          rural x = rural["ride id"].count()
          rural y = rural["fare"].mean()
          rural size = rural["driver count"].mean()
          #create scatter plot
          plt.scatter(urban x,urban y, s=urban size * 10, marker="o", facecolor="coral", edgecolors="black", alpha = 0.
          90, label = "Urban")
          plt.scatter(suburban x,suburban y,s=suburban size * 10, marker="o", facecolor="lightskyblue", edgecolors="bla
          ck", alpha = 0.90, label = "Suburban")
          plt.scatter(rural x,rural y,s=rural size * 10, marker="o", facecolor="gold", edgecolors="black", alpha = 0.90
          , label = "Rural")
          plt.grid()
          #create title and lables
          plt.title("Pyber Ride Sharing Data (2016)", fontsize=14, weight="bold")
          plt.xlabel("Total Number of Rides (Per City)", fontsize=10)
          plt.ylabel("Average Fare ($)", fontsize=10)
          #create Legend
          legend = plt.legend(title="City Type", fontsize = 10, loc="upper right")
          #change markers size for legend one by one using _sizes property found here: https://stackoverflow.com/questi
          ons/24706125/setting-a-fixed-size-for-points-in-legend
          legend.legendHandles[0]. sizes = [50]
          legend.legendHandles[1]. sizes = [50]
          legend.legendHandles[2]. sizes = [50]
          #text outside of plot found here: https://stackoverflow.com/questions/42435446/how-to-put-text-outside-python
```

```
-plots
plt.text(41.5,35, "Note: \nCircle size correlates with driver count per city", fontsize=10)

#save the figure. using bbox in order to fix the note's text so it shows
plt.savefig("data/PyberRideSharingData2016.png",bbox_inches="tight")

#shows the scatter plot
plt.show()
```

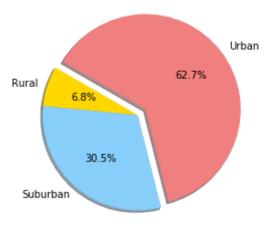


# **Total Fares by City Type**

Out[108]: type

Rural 4327.93 Suburban 19356.33 Urban 39854.38 Name: fare, dtype: float64

#### % of Total Fares by City Type



## **Total Rides by City Type**

## Out[110]: type

Rural 125 Suburban 625 Urban 1625

Name: ride\_id, dtype: int64

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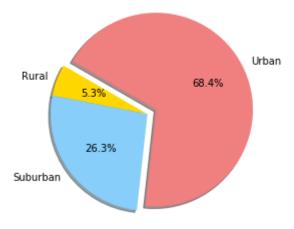
```
In [111]: #set up the characteristics for the pie graph
labels = ["Rural", "Suburban", "Urban"]
colors = ["gold", "lightskyblue", "lightcoral"]
explode = [0,0,0.1]

#set up the pie graph
plt.pie(total_rides_count, labels=labels, colors=colors, explode=explode, shadow=True, startangle=150, autopc
t="%1.1f%%")
plt.title("% of Total Rides by City Type", fontsize=10, weight="bold")
plt.axis("equal")

#save the figure
plt.savefig("data/TotalRidesbyCityType.png",bbox_inches="tight")

#show the pie graph
plt.show()
```





## **Total Drivers by City Type**

```
In [112]: #create a data frame grouped by "type" using the city_df, then calculate the sum of driver_count by "type"
    city_type = city_df.groupby(["type"])
    total_drivers_sum = city_type["driver_count"].sum()

#show the sum values by type for reference
    total_drivers_sum.head()
```

## Out[112]: type

 Rural
 78

 Suburban
 490

 Urban
 2405

Name: driver\_count, dtype: int64

```
In [113]: #set up the characteristics for the pie graph
labels = ["Rural", "Suburban", "Urban"]
colors = ["gold", "lightskyblue", "lightcoral"]
explode = [0,0,0.1]

#set up the pie graph
plt.pie(total_drivers_sum, labels=labels, colors=colors, explode=explode, shadow=True, startangle=150, autopc
t="%1.1f%")
plt.title("% of Total Drivers by City Type", fontsize=10, weight="bold")
plt.axis("equal")

#save the figure
plt.savefig("data/TotalDriversbyCityType.png",bbox_inches="tight")

#show the pie graph
plt.show()
```

#### % of Total Drivers by City Type

