

drawing_conclusions

October 23, 2017

1 Drawing Conclusions

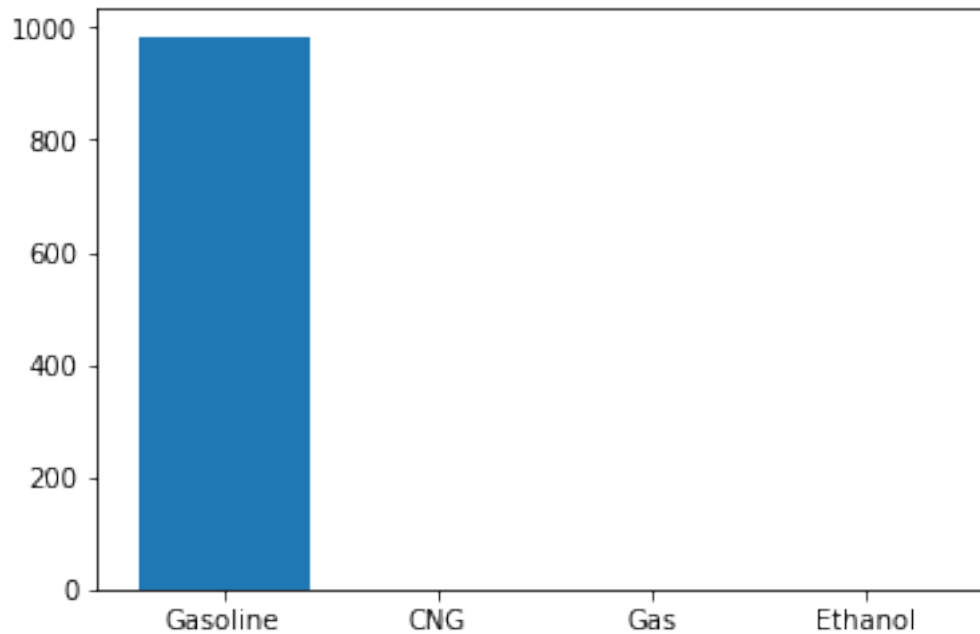
Use the space below to address questions on datasets `clean_08.csv` and `clean_18.csv`

```
In [3]: # load datasets
import pandas as pd
import matplotlib.pyplot as plt
% matplotlib inline
df_08 = pd.read_csv('clean_08.csv')
df_18 = pd.read_csv('clean_18.csv')
```

1.0.1 Q1: Are more unique models using alternative sources of fuel? By how much?

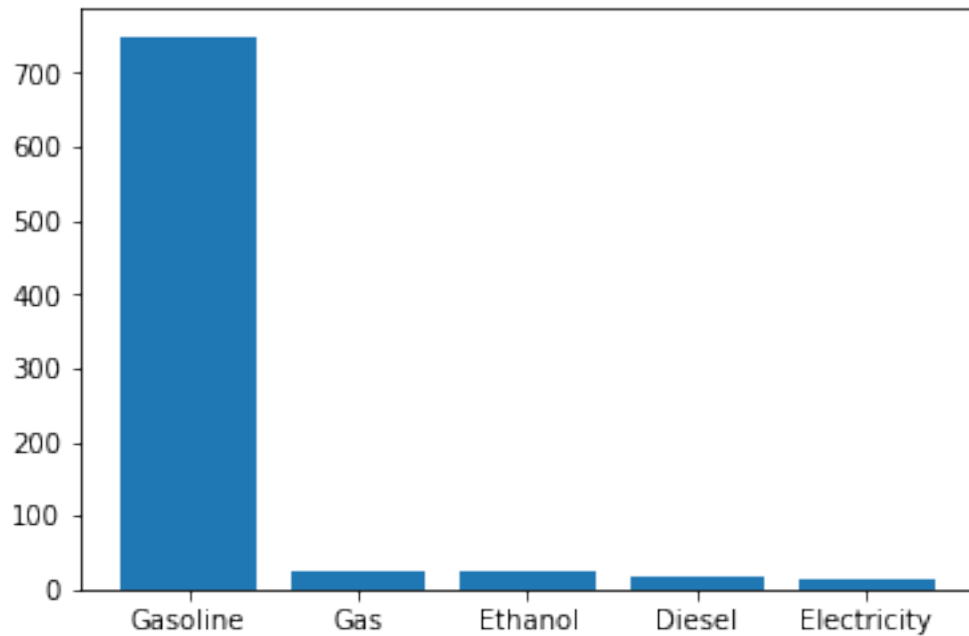
```
In [4]: df_08['fuel'].value_counts()
gasoline = df_08['fuel'].value_counts()[0]
cng = df_08['fuel'].value_counts()[1]
gas = df_08['fuel'].value_counts()[2]
ethanol = df_08['fuel'].value_counts()[3]

x_labels = ['Gasoline', 'CNG', 'Gas', 'Ethanol']
y_values = [gasoline, cng, gas, ethanol]
locations = [1, 2, 3, 4]
plt.bar(locations, y_values, tick_label = x_labels);
```



```
In [5]: df_18['fuel'].value_counts()
gasoline = df_18['fuel'].value_counts()[0]
gas = df_18['fuel'].value_counts()[1]
ethanol = df_18['fuel'].value_counts()[2]
diesel = df_18['fuel'].value_counts()[3]
electricity = df_18['fuel'].value_counts()[4]

x_labels = ['Gasoline', 'Gas', 'Ethanol', 'Diesel', 'Electricity']
y_values = [gasoline, gas, ethanol, diesel, electricity]
locations = [1, 2, 3, 4, 5]
plt.bar(locations, y_values, tick_label = x_labels);
```



1.0.2 Q2: How much have vehicle classes improved in fuel economy?

```
In [6]: #df_08.head()
        #df_08.groupby('veh_class')['cmb_mpg'].describe()
        df_08.groupby('veh_class')['cmb_mpg'].mean()
```

```
Out[6]: veh_class
SUV                18.471429
large car          18.509091
midsize car        21.601449
minivan            19.117647
pickup             16.277108
small car          21.105105
station wagon      22.366667
van                14.952381
Name: cmb_mpg, dtype: float64
```

```
In [7]: #df_18.groupby('veh_class')['cmb_mpg'].describe()
        df_18.groupby('veh_class')['cmb_mpg'].mean()
```

```
Out[7]: veh_class
large car          23.409091
midsize car        27.884058
minivan            20.800000
pickup             18.589744
small SUV          24.074074
```

```

small car          25.421053
special purpose    18.500000
standard SUV       18.197674
station wagon      27.529412
Name: cmb_mpg, dtype: float64

```

```

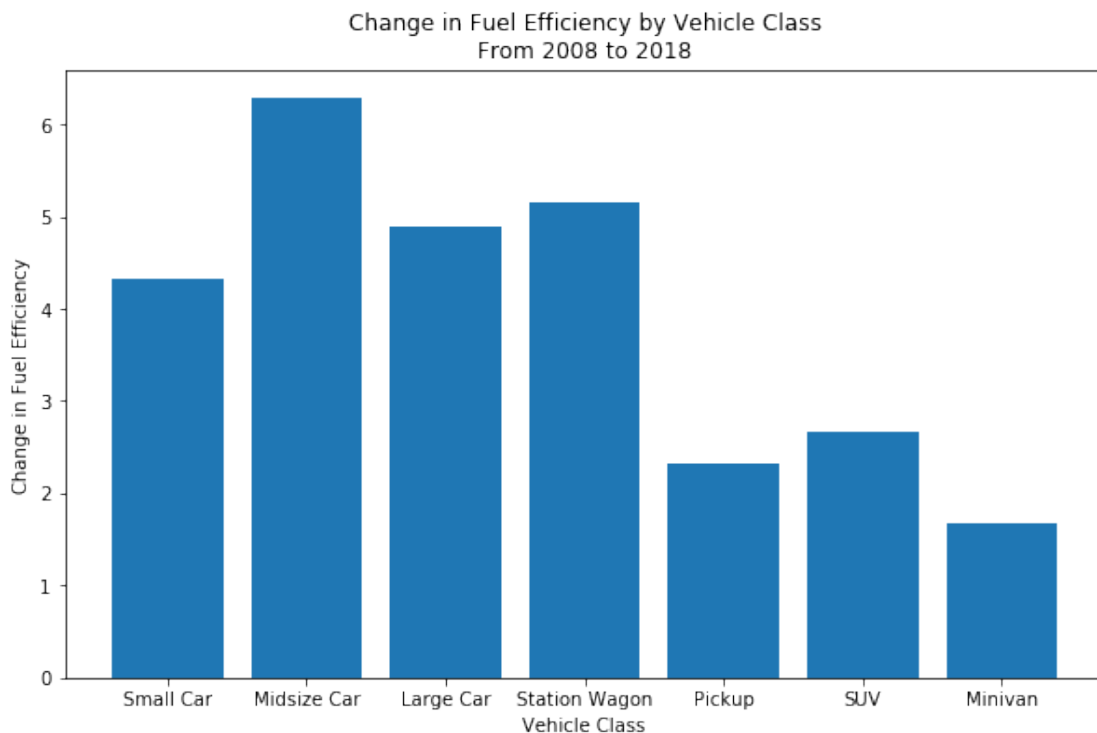
In [8]: station_wagon_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[8] - df_08.groupby('veh_class')['cmb_mpg'].mean()[8]
large_car_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[0] - df_08.groupby('veh_class')['cmb_mpg'].mean()[0]
midsize_car_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[1] - df_08.groupby('veh_class')['cmb_mpg'].mean()[1]
small_car_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[5] - df_08.groupby('veh_class')['cmb_mpg'].mean()[5]
minivan_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[2] - df_08.groupby('veh_class')['cmb_mpg'].mean()[2]
pickup_dx = df_18.groupby('veh_class')['cmb_mpg'].mean()[3] - df_08.groupby('veh_class')['cmb_mpg'].mean()[3]
suv_dx = (df_18.groupby('veh_class')['cmb_mpg'].mean()[4] + df_18.groupby('veh_class')['cmb_mpg'].mean()[6]) - df_08.groupby('veh_class')['cmb_mpg'].mean()[4] + df_08.groupby('veh_class')['cmb_mpg'].mean()[6]

```

```

In [9]: location = [1, 2, 3, 4, 5, 6, 7]
labels = ['Small Car', 'Midsize Car', 'Large Car', 'Station Wagon', 'Pickup', 'SUV', 'Minivan']
values = [small_car_dx, midsize_car_dx, large_car_dx, station_wagon_dx, pickup_dx, suv_dx, minivan_dx]
plt.figure(figsize=(10,6));
plt.bar(location, values, tick_label=labels);
plt.title('Change in Fuel Efficiency by Vehicle Class\nFrom 2008 to 2018');
plt.xlabel('Vehicle Class');
plt.ylabel('Change in Fuel Efficiency');

```



1.0.3 Q3: What are the characteristics of SmartWay vehicles? Have they changed over time?

```
In [10]: df_08.query('smartway == "yes"').describe()
```

```
Out[10]:
```

	displ	cyl	air_pollution_score	city_mpg	hwy_mpg	\
count	380.000000	380.000000	380.000000	380.000000	380.000000	
mean	2.602895	4.826316	7.268421	20.984211	28.413158	
std	0.623436	1.002025	0.970027	3.442672	3.075194	
min	1.300000	4.000000	6.000000	17.000000	22.000000	
25%	2.275000	4.000000	7.000000	19.000000	26.000000	
50%	2.400000	4.000000	7.000000	20.000000	28.000000	
75%	3.000000	6.000000	7.000000	22.000000	30.000000	
max	5.000000	8.000000	9.000000	48.000000	45.000000	

	cmb_mpg	greenhouse_gas_score
count	380.000000	380.000000
mean	23.736842	6.868421
std	3.060379	0.827338
min	20.000000	6.000000
25%	22.000000	6.000000
50%	23.000000	7.000000
75%	25.000000	7.000000
max	46.000000	10.000000

```
In [11]: print("Combined MPG Comparison")
print('2008:', df_08.query('smartway == "yes"')['cmb_mpg'].mean())
print('2018:', df_18.query('smartway == "Yes"')['cmb_mpg'].mean())
```

Combined MPG Comparison

2008: 23.7368421053

2018: 34.4395604396

```
In [12]: print("Greenhouse Gas Score Comparison")
print('2008:', df_08.query('smartway == "yes"')['greenhouse_gas_score'].mean())
print('2018:', df_18.query('smartway == "Yes"')['greenhouse_gas_score'].mean())
```

Greenhouse Gas Score Comparison

2008: 6.86842105263

2018: 7.53846153846

```
In [13]: print("Air Pollution Score")
print('2008:', df_08.query('smartway == "yes"')['air_pollution_score'].mean())
print('2018:', df_18.query('smartway == "Yes"')['air_pollution_score'].mean())
```

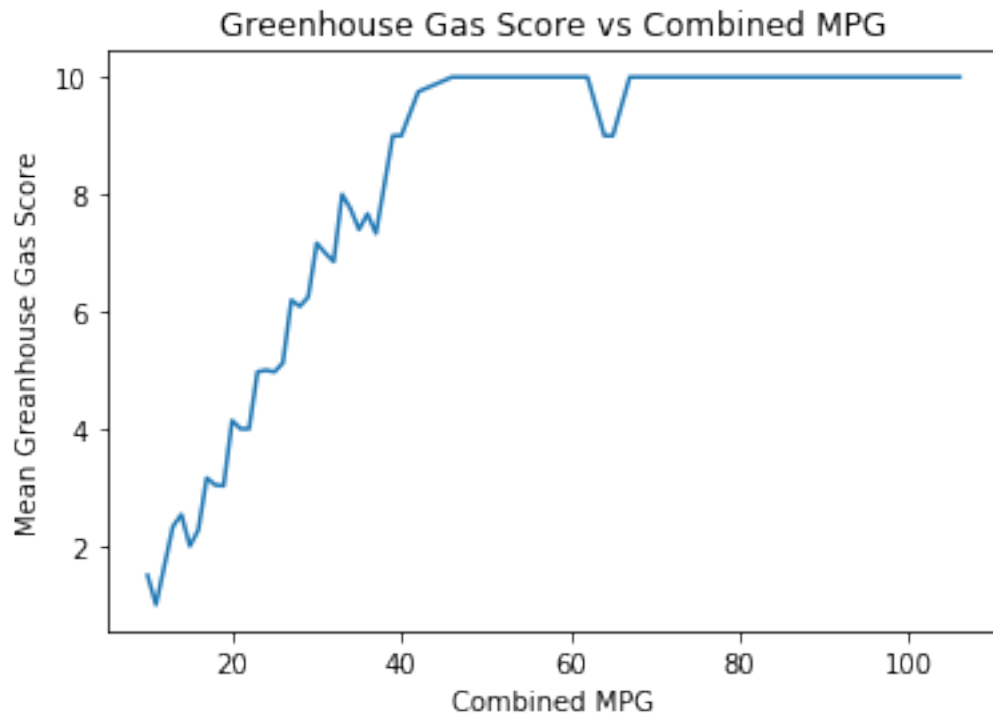
Air Pollution Score

2008: 7.26842105263

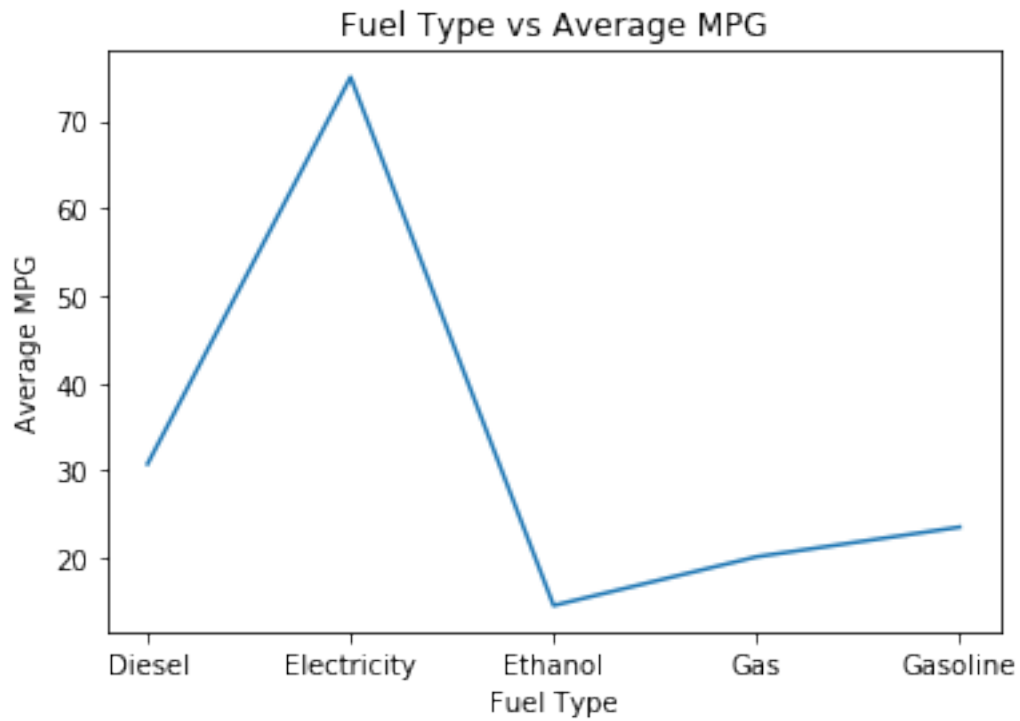
2018: 4.87912087912

1.0.4 Q4: What features are associated with better fuel economy?

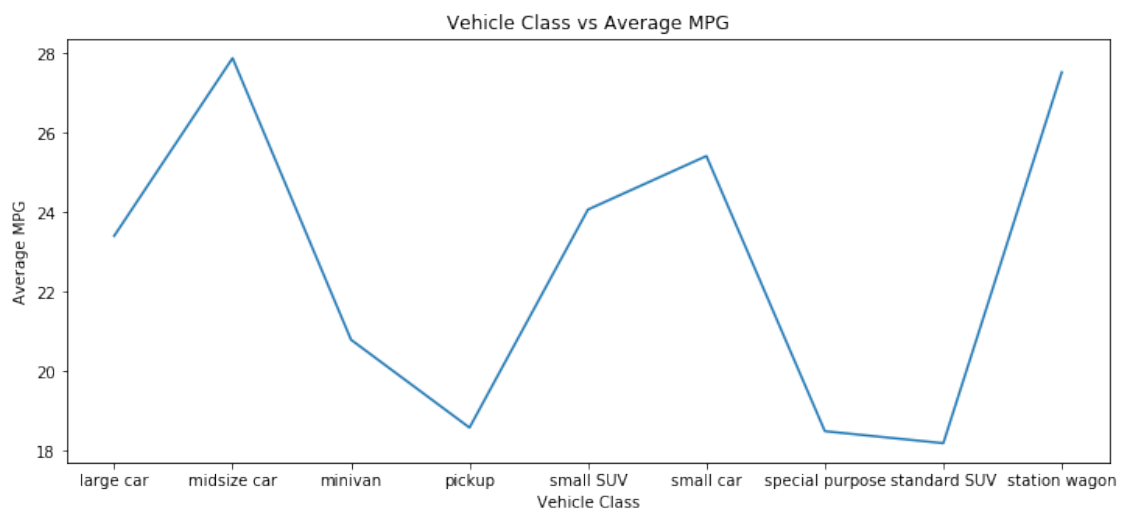
```
In [80]: x = df_18.groupby('cmb_mpg')['cmb_mpg'].mean()
y = df_18.groupby('cmb_mpg')['greenhouse_gas_score'].mean()
plt.plot(x,y);
plt.title('Greenhouse Gas Score vs Combined MPG');
plt.xlabel('Combined MPG');
plt.ylabel('Mean Greenhouse Gas Score');
```



```
In [167]: x = df_18.groupby('fuel')
y = df_18.groupby('fuel')['cmb_mpg'].mean()
plt.plot(y);
plt.title('Fuel Type vs Average MPG');
plt.xlabel('Fuel Type');
plt.ylabel('Average MPG');
```



```
In [115]: y = df_18.groupby('veh_class')['cmb_mpg'].mean()  
plt.figure(figsize=(12,5));  
plt.plot(y);  
plt.title('Vehicle Class vs Average MPG');  
plt.xlabel('Vehicle Class');  
plt.ylabel('Average MPG');
```



```
In [ ]:
```