## BA888 Capstone Project

# Team 1 2/13/2020

```
uber = read_csv("uber_cleandata_with_weather.csv")
```

```
## Parsed with column specification:
## cols(
##
     .default = col_double(),
##
     id = col_character(),
##
     source = col_character(),
##
     destination = col_character(),
##
     car_type = col_character(),
##
     short_summary = col_character(),
##
     icon = col_character()
## )
## See spec(...) for full column specifications.
```

#### glimpse(uber)

```
## Observations: 330,568
## Variables: 39
## $ id
                            <chr> "009e9c53-074d-43cf-aef2-0fbc7a47ed3d"...
## $ hour
                            <dbl> 22, 10, 19, 23, 0, 19, 20, 3, 13, 10, ...
## $ day
                            <dbl> 30, 13, 13, 16, 14, 29, 17, 16, 27, 13...
## $ month
                            <dbl> 11, 12, 12, 12, 12, 11, 12, 12, 11, 12...
## $ source
                            <chr> "North End", "North End", "North End",...
                            <chr> "West End", "West End", "West End", "W...
## $ destination
## $ car_type
                            <chr> "UberXL", "Black", "UberX", "WAV", "Bl...
                            <dbl> 12.0, 16.0, 7.5, 7.5, 26.0, 5.5, 8.5, ...
## $ price
## $ distance
                            <dbl> 1.11, 1.11, 1.11, 1.11, 1.11, 1.11, 2....
## $ surge multiplier
                            <dbl> 42.3647, 42.3647, 42.3647, 42.3647, 42...
## $ latitude
                            <dbl> -71.0542, -71.0542, -71.0542, -71.0542...
## $ longitude
## $ temperature
                            <dbl> 40.13, 20.38, 32.85, 41.29, 31.25, 43....
## $ apparentTemperature
                            <dbl> 38.08, 20.38, 32.85, 36.01, 31.25, 37....
## $ short_summary
                            <chr> "Overcast", "Clear", "Mostly Cloudy", ...
                            <dbl> 0.0000, 0.0000, 0.0000, 0.0567, 0.0000...
## $ precipIntensity
## $ precipProbability
                            <dbl> 0.00, 0.00, 0.00, 0.94, 0.00, 0.00, 0....
## $ humidity
                            <dbl> 0.60, 0.66, 0.56, 0.86, 0.64, 0.52, 0....
                            <dbl> 3.38, 2.94, 2.65, 8.30, 2.62, 12.13, 6...
## $ windSpeed
## $ windGust
                            <dbl> 3.99, 3.22, 3.83, 8.30, 3.54, 19.97, 8...
## $ visibility
                            <dbl> 9.833, 9.831, 9.959, 4.054, 10.000, 9....
## $ temperatureHigh
                            <dbl> 42.52, 33.83, 33.83, 43.83, 33.83, 44....
                            <dbl> 31.71, 27.27, 27.27, 34.25, 27.27, 28....
## $ temperatureLow
## $ apparentTemperatureHigh <dbl> 40.53, 32.85, 32.85, 38.38, 32.85, 38....
## $ apparentTemperatureLow <dbl> 28.06, 24.61, 24.61, 28.30, 24.61, 26....
## $ icon
                            <chr> "cloudy", "clear", "partly-cloudy", "r...
## $ dewPoint
                            <dbl> 27.31, 10.87, 18.66, 37.56, 20.53, 26....
```

```
<dbl> 1017.16, 1031.51, 1033.65, 1012.72, 10...
## $ pressure
## $ windBearing
                           <dbl> 281, 2, 76, 57, 173, 313, 286, 38, 227...
## $ cloudCover
                           <dbl> 1.00, 0.03, 0.64, 1.00, 0.91, 0.53, 0....
                           ## $ uvIndex
## $ visibility.1
                           <dbl> 9.833, 9.831, 9.959, 4.054, 10.000, 9....
## $ ozone
                           <dbl> 281.8, 327.3, 330.8, 325.3, 326.7, 309...
## $ moonPhase
                           <dbl> 0.79, 0.21, 0.21, 0.30, 0.21, 0.75, 0....
                           <dbl> 0.0003, 0.0001, 0.0001, 0.1252, 0.0001...
## $ precipIntensityMax
## $ temperatureMin
                           <dbl> 28.79, 18.29, 18.29, 39.22, 18.29, 35....
## $ temperatureMax
                           <dbl> 42.52, 33.83, 33.83, 43.83, 33.83, 44....
## $ apparentTemperatureMin
                           <dbl> 26.41, 13.79, 13.79, 33.98, 13.79, 31....
                           <dbl> 40.53, 32.85, 32.85, 38.38, 32.85, 38....
## $ apparentTemperatureMax
```

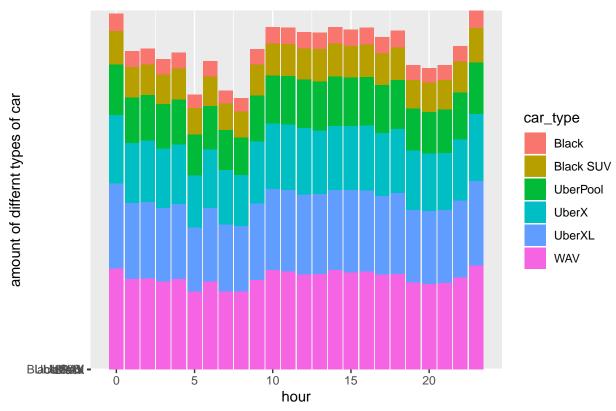
#### head(uber, 10)

```
## # A tibble: 10 x 39
##
             hour
                    day month source destination car_type price distance
##
      <chr> <dbl> <dbl> <dbl> <chr> <chr>
                                                   <chr>>
                                                            <dbl>
                                                                     <dbl>
##
   1 009e~
               22
                     30
                            11 North~ West End
                                                             12
                                                                       1.11
                                                   UberXI.
##
    2 23f1~
               10
                     13
                            12 North~ West End
                                                  Black
                                                             16
                                                                       1.11
##
  3 3575~
               19
                     13
                            12 North~ West End
                                                  UberX
                                                              7.5
                                                                       1.11
##
  4 50ef~
               23
                     16
                            12 North~ West End
                                                   WAV
                                                              7.5
                                                                       1.11
## 5 91c4~
                0
                     14
                            12 North~ West End
                                                   Black S~
                                                             26
                                                                       1.11
##
    6 e219~
               19
                     29
                            11 North~ West End
                                                   UberPool
                                                              5.5
                                                                       1.11
##
   7 1d45~
               20
                     17
                            12 Beaco~ South Stat~ WAV
                                                              8.5
                                                                      2.48
##
    8 1f64~
                3
                            12 Beaco~ South Stat~ UberXL
                                                                       2.48
                     16
                                                             15
##
   9 2ca4~
               13
                     27
                            11 Beaco~ South Stat~ Black
                                                             20.5
                                                                       2.48
## 10 4149~
               10
                     13
                            12 Beaco~ South Stat~ UberX
                                                              8.5
                                                                       2.48
## # ... with 30 more variables: surge_multiplier <dbl>, latitude <dbl>,
       longitude <dbl>, temperature <dbl>, apparentTemperature <dbl>,
       short_summary <chr>, precipIntensity <dbl>, precipProbability <dbl>,
## #
       humidity <dbl>, windSpeed <dbl>, windGust <dbl>, visibility <dbl>,
## #
## #
       temperatureHigh <dbl>, temperatureLow <dbl>,
## #
       apparentTemperatureHigh <dbl>, apparentTemperatureLow <dbl>,
       icon <chr>, dewPoint <dbl>, pressure <dbl>, windBearing <dbl>,
## #
## #
       cloudCover <dbl>, uvIndex <dbl>, visibility.1 <dbl>, ozone <dbl>,
       moonPhase <dbl>, precipIntensityMax <dbl>, temperatureMin <dbl>,
## #
## #
       temperatureMax <dbl>, apparentTemperatureMin <dbl>,
## #
       apparentTemperatureMax <dbl>
```

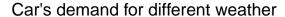
Team 1 discusses about how location and weather refects uber price and try to find how what we found will affect some industy like hotel, which have connections with ordering service. Our data has 39 variables and 330568 rows, which have four different categories data: time, location, price and weather infomation.

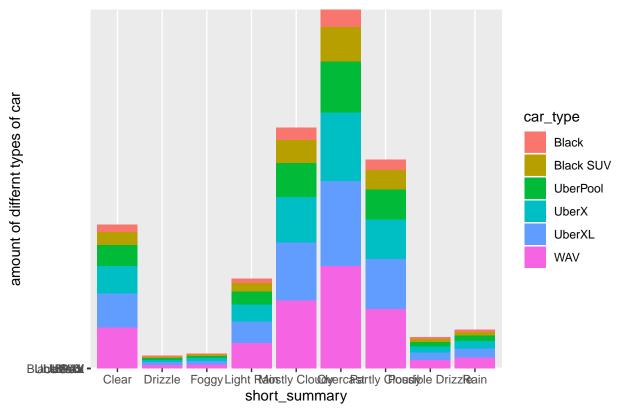
```
ggplot(uber)+
  geom_bar(aes(x=hour,y=car_type,fill=car_type),stat='identity')+
  ylab("amount of differnt types of car")+
  ggtitle("Car's demand for different hour")
```

### Car's demand for different hour



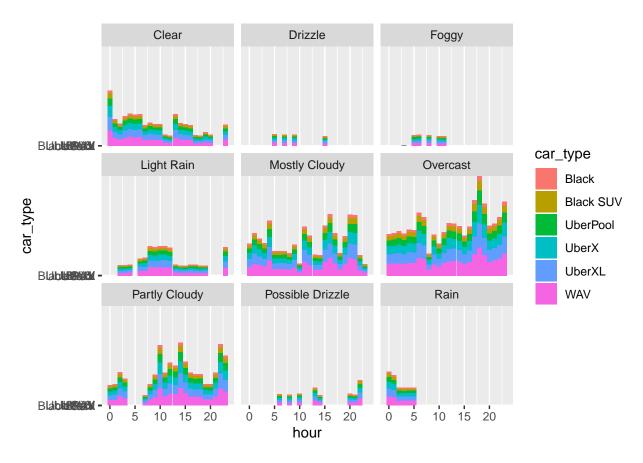
```
ggplot(uber,aes(x=short_summary,y=car_type,fill=car_type))+
  geom_bar(stat='identity')+
  ylab("amount of differnt types of car")+
  ggtitle("Car's demand for different weather")
```





From first chart we can see highest car demand during 11pm and 12am, and the second higher demand for cars is around 10am and 5pm. The second chart we can see the highest car demand on weather on partly cloudy and cloudy, and people have too much demand on good weather of rainy day. We can also see most popular car is WAV, which uber drivers provide affordable rides in wheelchair. Even though most user don't know whether their car provide wheelchair, but most of them order cars with this service. The second most popular cartype is UberXL.

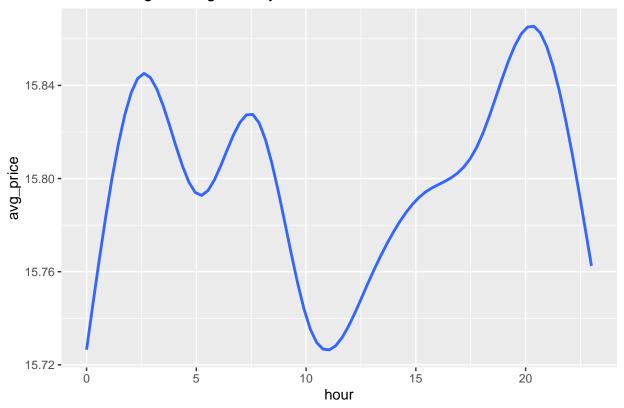
```
ggplot(uber,aes(x=hour,y=car_type,fill=car_type))+
  geom_bar(stat='identity')+
  facet_wrap(~short_summary)
```



```
## price and car amount change during 24h
uber %>% select(price,hour,short_summary) %>% group_by(hour) %>%
  mutate(avg_price=mean(price)) %>%
  ggplot()+
  labs(title = "Price change during the day")+
  geom_smooth(aes(x=hour,y=avg_price),na.rm = FALSE, se = TRUE)
```

##  $geom_smooth()$  using method = gam' and formula  $y \sim s(x, bs = cs')'$ 

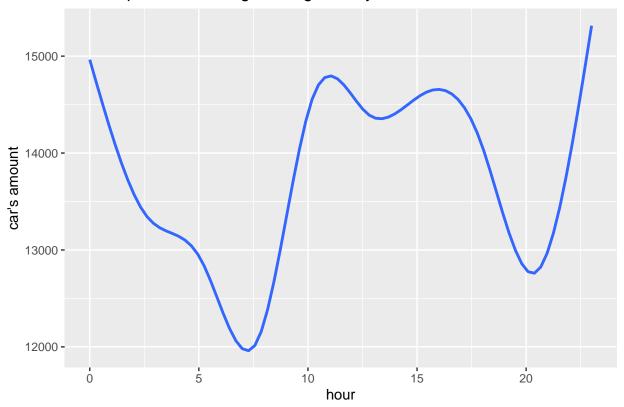
## Price change during the day



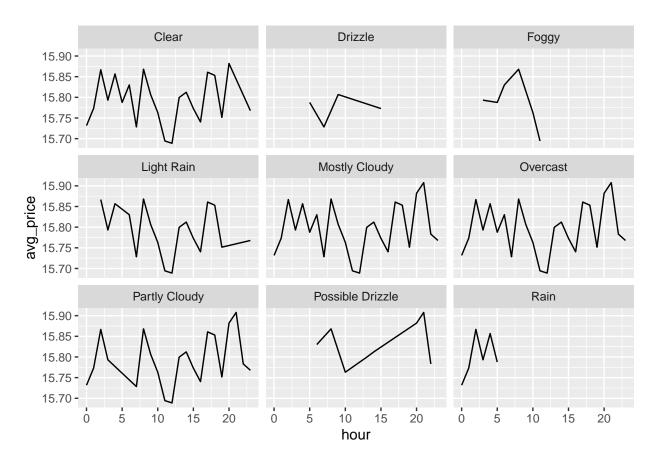
```
uber %>% select(hour,car_type,short_summary) %>% group_by(hour) %>%
  mutate(count=n()) %>%
  ggplot() +
  geom_smooth(aes(x=hour,y=count),na.rm = FALSE, se = TRUE)+
  labs(y="car's amount",title = "car's requirement change during the day")
```

##  $geom_smooth()$  using method = 'gam' and formula 'y ~ s(x, bs = "cs")'

## car's requirement change during the day

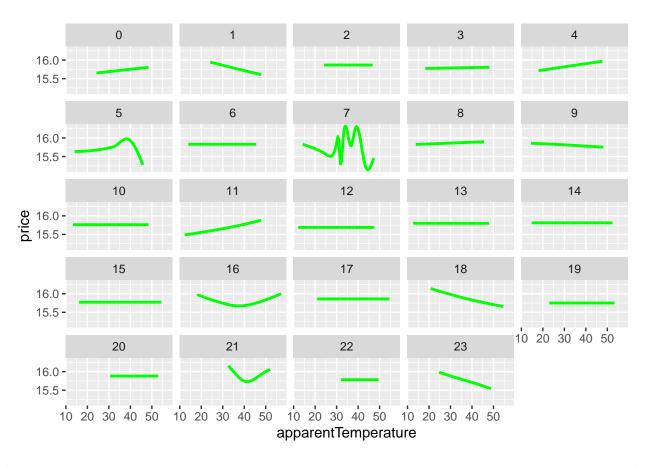


```
uber %>% select(price,hour,short_summary) %>% group_by(hour) %>%
  mutate(avg_price=mean(price)) %>%
  ggplot()+
  geom_line(aes(x=hour,y=avg_price))+
  facet_wrap(~short_summary)
```



```
uber_temperature = uber %>% select(hour,car_type,price,temperature,apparentTemperature,temperatureHigh,
uber_temperature %>%
ggplot() +
geom_smooth(aes(x=apparentTemperature,y=price),se=F,col="green")+
facet_wrap(~hour)
```

## `geom\_smooth()` using method = 'gam' and formula 'y ~ s(x, bs = "cs")'



```
uber_temperature %>%
   ggplot() +
   geom_smooth(aes(x=temperature,y=price),se=F,col="red")+
   facet_wrap(~hour)
```

##  $geom_smooth()$  using method = gam' and formula  $y \sim s(x, bs = cs')'$ 

