

# Assignment1

## Data:

The data set is from Kaggle containing the McDonalds menu with nutrition information and some notes are available using R and python with plotly. Some demos in (<https://www.kaggle.com/limitpointinfo/mc-d-nutrition-eda>) and just keep a note for the talk [https://github.com/jrauser/writing/blob/master/how\\_humans\\_see\\_data/hhsd\\_notes.Rmd](https://github.com/jrauser/writing/blob/master/how_humans_see_data/hhsd_notes.Rmd)

I eat a lot of fast food and McDonalds is one of my most usually visit restaurant after midnight. So I am very interested in how I gain so much weights, so this data set is fairly interesting to me.

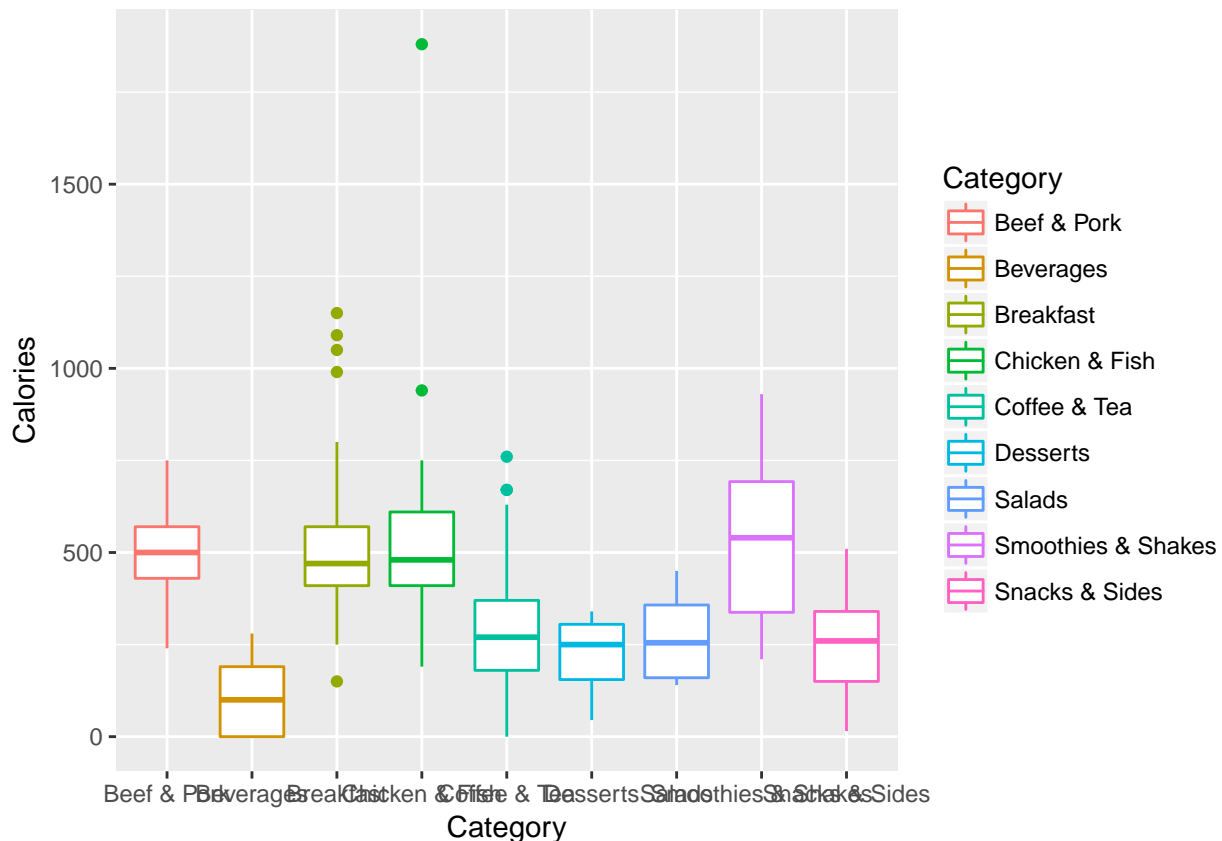
## Graphs:

```
library(tidyverse)
library(ggplot2)
aa<-read.csv('menu.csv')
bb<-as_data_frame(aa)
```

Here, I have a question, why is that if I use read\_csv('menu.csv'), this considers some categorigal data as char, so I have to use function and change into dataframe. Can I have a way fixing it?

Graph 1:

```
gg1<-ggplot(bb, aes(x=Category ,y=Calories, colour = Category)) +geom_boxplot()
gg1
```



This is very interesting for me that beverage has the least calories, I always thought the pop makes me fat.

```
result <- bb %>%
  select(Category, Item, Serving.Size, Calories) %>% group_by(Category) %>% top_n(3) # %>% summarise(Category)
```

## Selecting by Calories

```
data.frame(result[,c('Item', 'Calories')])
```

	Item	Calories
## 1	Big Breakfast with Hotcakes (Regular Biscuit)	1090
## 2	Big Breakfast with Hotcakes (Large Biscuit)	1150
## 3	Big Breakfast with Hotcakes and Egg Whites (Large Biscuit)	1050
## 4	Quarter Pounder with Bacon Habanero Ranch	610
## 5	Double Quarter Pounder with Cheese	750
## 6	Bacon Clubhouse Burger	720
## 7	Bacon Clubhouse Crispy Chicken Sandwich	750
## 8	Chicken McNuggets (20 piece)	940
## 9	Chicken McNuggets (40 piece)	1880
## 10	Premium Bacon Ranch Salad with Crispy Chicken	380
## 11	Premium Southwest Salad with Crispy Chicken	450
## 12	Premium Southwest Salad with Grilled Chicken	290
## 13	Chipotle BBQ Snack Wrap (Crispy Chicken)	340
## 14	Ranch Snack Wrap (Crispy Chicken)	360
## 15	Medium French Fries	340
## 16	Large French Fries	510
## 17	Hot Fudge Sundae	330
## 18	Hot Caramel Sundae	340
## 19	Strawberry Sundae	280
## 20	Coca-Cola Classic (Large)	280
## 21	Sprite (Large)	280
## 22	Minute Maid Orange Juice (Large)	280
## 23	Frappé Mocha (Large)	670
## 24	Frappé Caramel (Large)	670
## 25	Frappé Chocolate Chip (Large)	760
## 26	Strawberry Shake (Large)	850
## 27	Chocolate Shake (Large)	850
## 28	McFlurry with M&M's Candies (Medium)	930

We can see from the select top 3 highest calories food from each Category, which is related with the size.

Graph 2:

```
(gg2<-ggplot(bb,aes(x=Calories,y=Saturated.Fat,colour=Category))
+geom_point()+facet_wrap(~Category)+ geom_smooth(method='loess')+theme_minimal())
```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric =
## parametric, : pseudoinverse used at 138.45
```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric =
## parametric, : neighborhood radius 151.55
```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric =
## parametric, : reciprocal condition number 0
```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric =
## parametric, : There are other near singularities as well. 6400
```

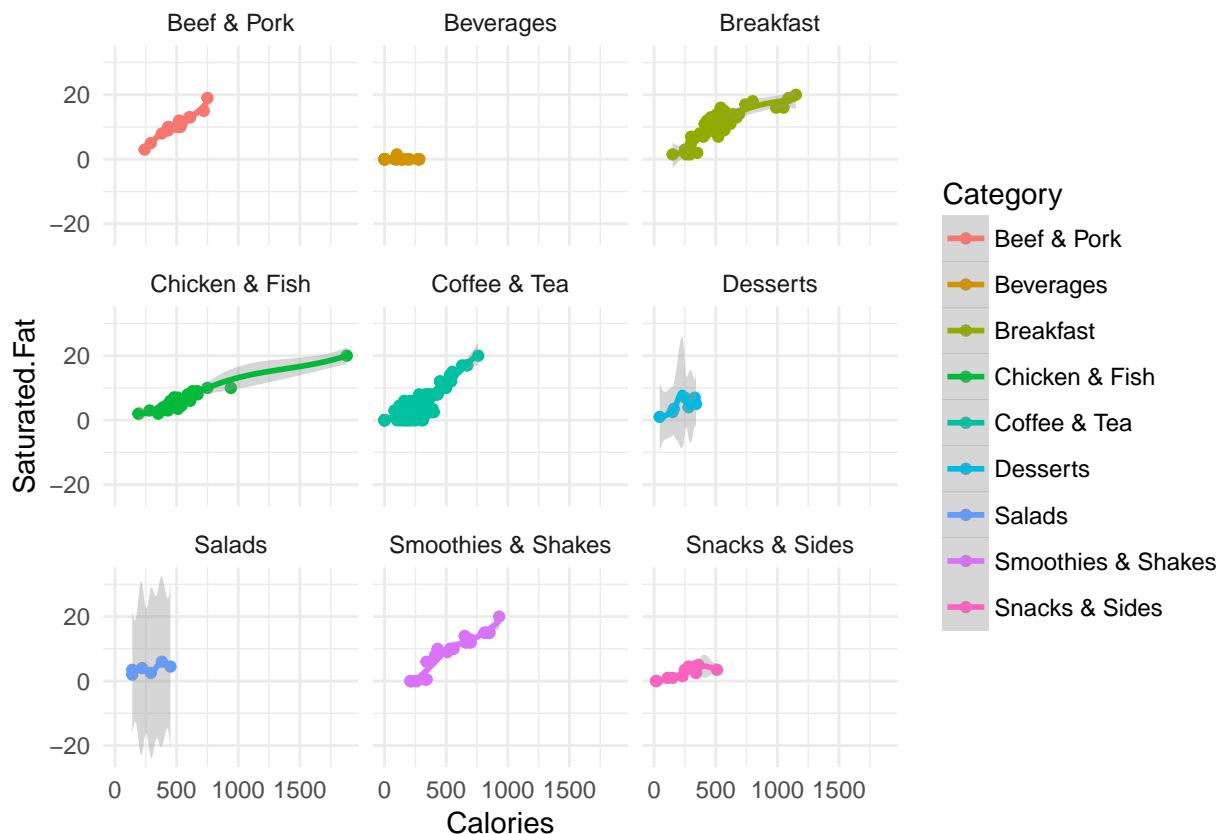
```
## Warning in predLoess(object$y, object$x, newx = if
```

```
## (is.null(newdata)) object$x else if (is.data.frame(newdata))
## as.matrix(model.frame(delete.response(terms(object))), : pseudoinverse used
## at 138.45

## Warning in predLoess(object$y, object$x, newx = if
## (is.null(newdata)) object$x else if (is.data.frame(newdata))
## as.matrix(model.frame(delete.response(terms(object))), : neighborhood radius
## 151.55

## Warning in predLoess(object$y, object$x, newx = if
## (is.null(newdata)) object$x else if (is.data.frame(newdata))
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## condition number 0

## Warning in predLoess(object$y, object$x, newx = if
## (is.null(newdata)) object$x else if (is.data.frame(newdata))
## as.matrix(model.frame(delete.response(terms(object))), : There are other
## near singularities as well. 6400
```



```
#scale_colour_hue(h = c(270, 360))+
```

From the second graph, we can tell that bad fat has positive linear correlation with calories, for beverages, salads and snacks & Sides, the slopes are rather soft, which means that even though they have high calories, they do not contribute much to my fatness.

Therefore, sometimes high calories does not necessary means that it makes you fat, it depends on what kind of food you are taking.

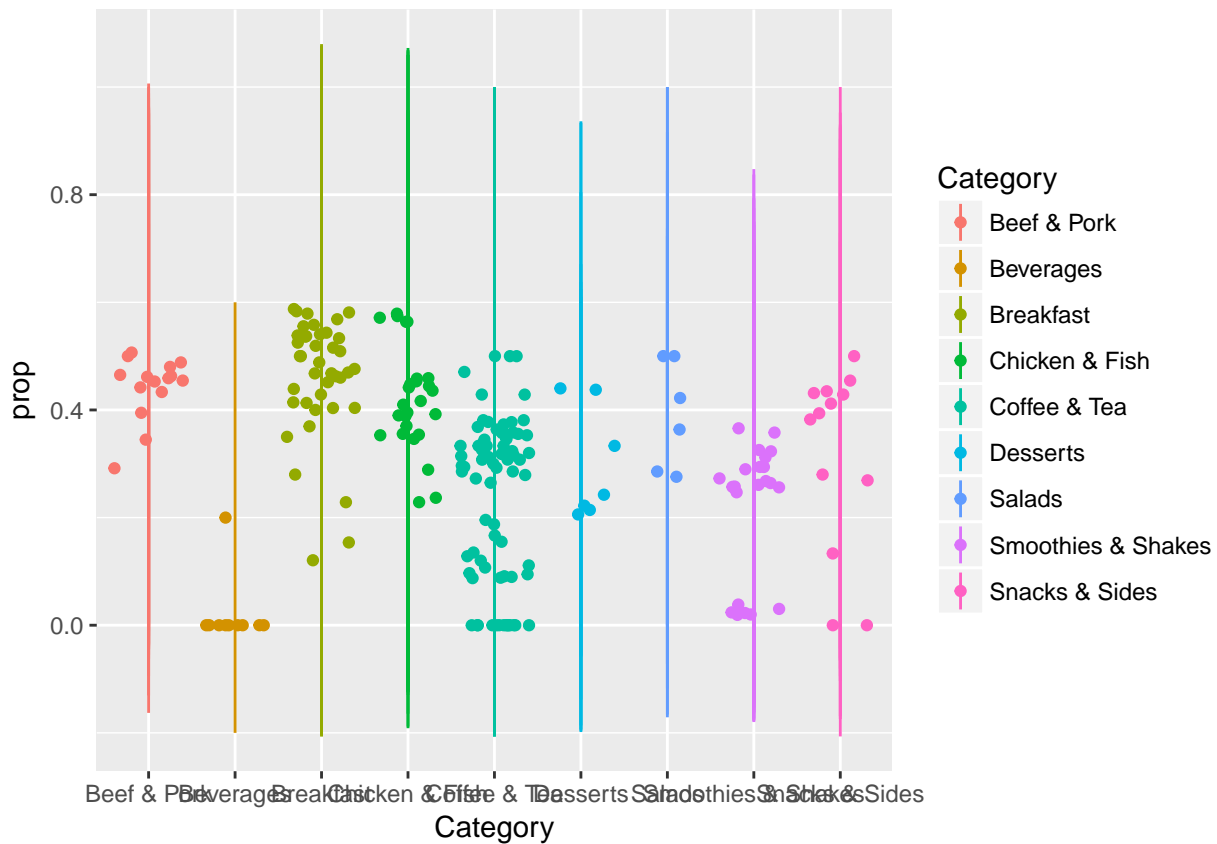
Graph 3:

```
(bb
  %>% group_by(Category, Calories.from.Fat, Calories)
  %>% summarise(prop=mean(Calories.from.Fat/Calories),
                 n=n(),
                 se=sqrt(prop*(1-prop)/n))
)->cc
```

```
(ggplot(cc, aes(Category, prop, colour=Category))+geom_jitter()
  +geom_linerange(aes(ymin=prop-se, ymax=prop+se)))
```

## Warning: Removed 2 rows containing missing values (geom\_point).

## Warning: Removed 2 rows containing missing values (geom\_linerange).



This graph shows the proportion of Calories from Fat for different kinds of food, we can tell that for different groups, we have barely the same proportion except for beverage, which is the lowest among all.