Homework 4

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This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

1. Watch the video presentation by Tamara Munzner: Keynote on Visualization Principles, https://vimeo.com/26205288 and slides - http://www.cs.ubc.ca/~tmm/talks/vizbi11/vizbi11.pdf Summarize the key takehome messages from her presentation.

I believe the major aim of the presentation is to point out that the option to use visualization should depend on the data available to you. When we have different types of data it is easier to visualize relationships and patterns between them by manipulating with the color, shape of the display value etc in order to bring actionable insights to the surface. There is a lot of power in using spatial position to analyze data as in almost every case it works out fine. An important point she also made was concerning "data transformation". Transforming the data into a suitable abstraction is needed in data visualization in order to represent data correctly.

In respect to 3D data visualization, one has to be cautious in this approach because important things could be easily missed (whereas avoidable anyways) and a good example of this is the fact that 3D does distort perspective. Taking a look at text legibility shows the extent to which 3D might be dangerous however 3D visualization is quite good (and mostly adviceable) for true 3D spatial data. Usage of 3D could result to important values from the data to be missing due to an overriding of the value by other values. In majority of the situation, it is easier to draw conclusions from 2-dimensional data visualization as opposed to 3-dimensional.

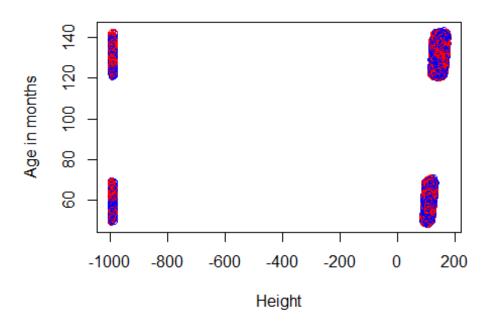
2. Fetch the UK child height/weight measurements data from here. (File).

Study a sufficiently large subset of measurements - plot dotplots comparing age, height, weight, BMI;

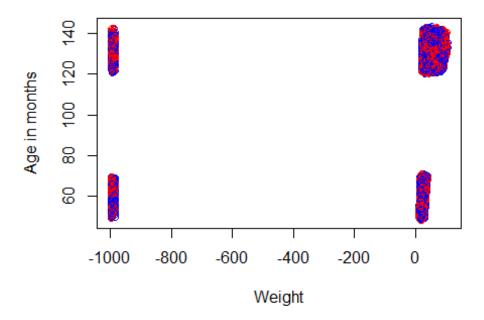
Experiment with color - highlighting gender, BMI clinical score, or age, by different markers, color or color scales.

(hint: you may want to develop ideas first with smaller subsets of data).

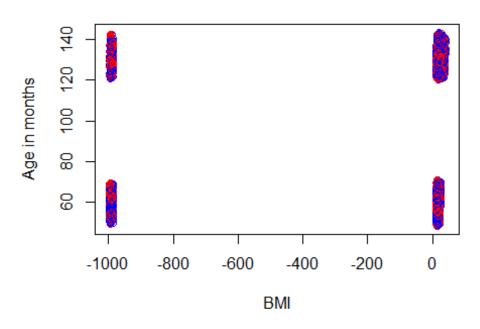
Age vs Height



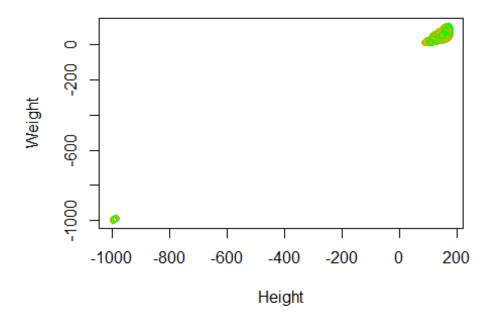
Age vs Weight



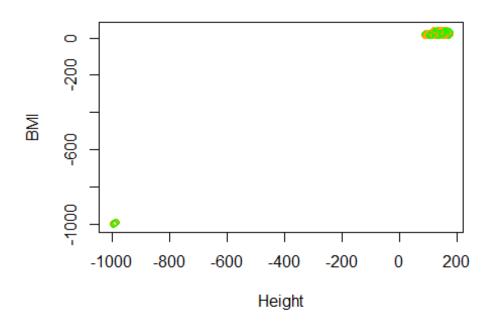
Age vs BMI



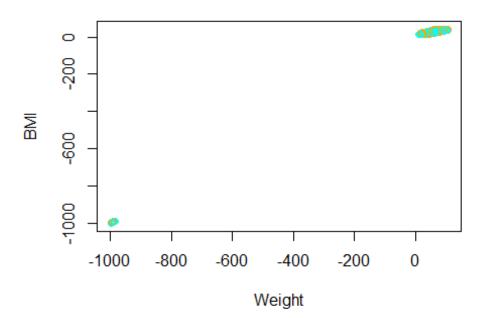
Hieght vs weight



Hieght vs BMI



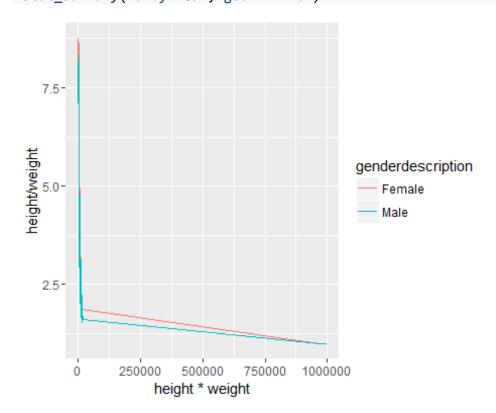
Wieght vs BMI



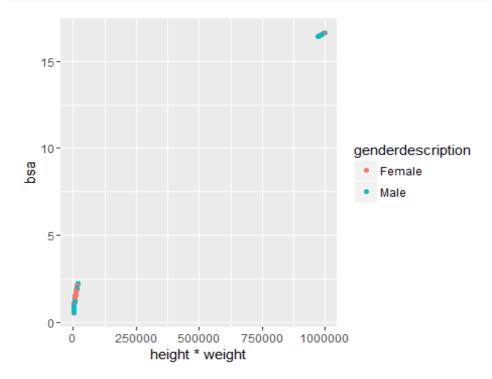
3. Derive new features to plot: height*weight; height/weight; Body Surface Area; Plot them against each other; and against BMI, height, or weight. Try to identify interesting meaningful trends or examples, provide some interpretation.

```
#calculate the Body surface Area using Mosteller formular
childData$bsa <-sqrt((childData$weight*childData$height)/3600)
childData$genderdescription <-factor(childData$genderdescription)
library(ggplot2)

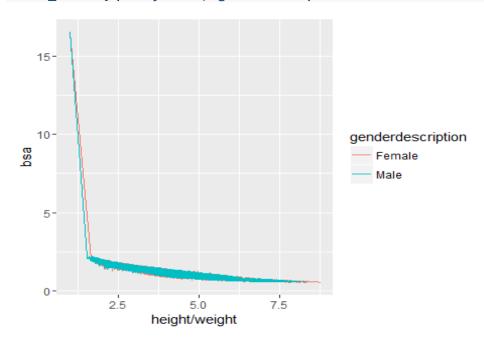
ggplot(childData, aes(height*weight, height/weight, color=genderdescription))
+stat_summary(fun.y=mean, geom="line")</pre>
```



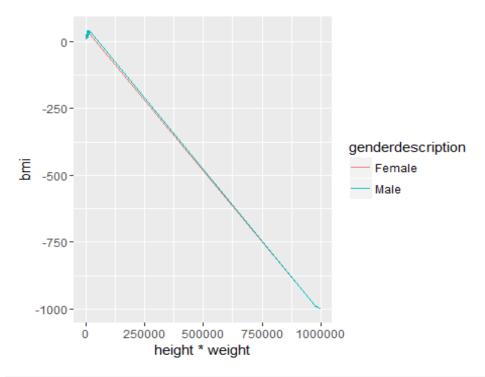
ggplot(childData, aes(height*weight, bsa, color=genderdescription)) +geom_point()



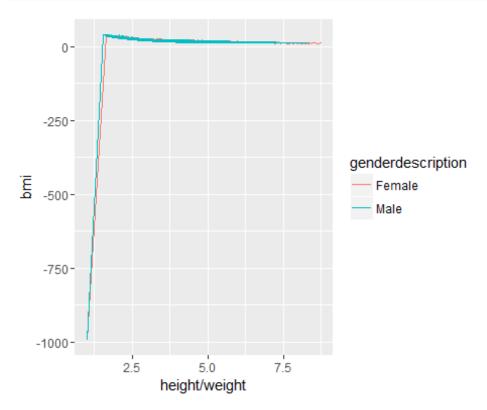
ggplot(childData, aes(height/weight, bsa, color=genderdescription))
+stat_summary(fun.y=mean, geom="line")



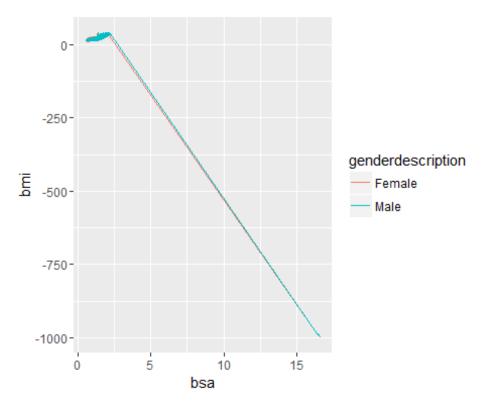
ggplot(childData, aes(height*weight, bmi, color=genderdescription))
+stat_summary(fun.y=mean, geom="line")



ggplot(childData, aes(height/weight, bmi, color=genderdescription))
+stat_summary(fun.y=mean, geom="line")



ggplot(childData, aes(bsa, bmi, color=genderdescription))
+stat_summary(fun.y=mean, geom="line")



I subsetted a small

chunk of the data that included the first 10,000 rows and exported it to excel where I calculated the height*weight, height/weight and Body surface Area and then imported the data back to try our some plots. For the calculation of the body surface area I took the Mosteller formular which calculates the BSA by multiplying the height and weight, dividing it by 3600 and finding the square root of the outcome.

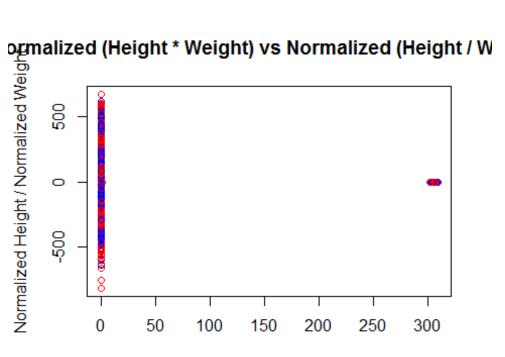
My observation is that the female children seem to grow a bit bigger and faster than the male hence there's a tendency that obesity will be more common in the female children than the male children.

4. Normalise height and weight based on the gender and age, repeat some of the plots from above.

```
library('recommenderlab')
childData$normHeight<-scale(childData$height)
childData$normWeight<-scale(childData$weight)

childData$nHeightMulnWeight <-childData$normHeight*childData$normWeight
childData$nHeightDivnWeight <-childData$normHeight/childData$normWeight

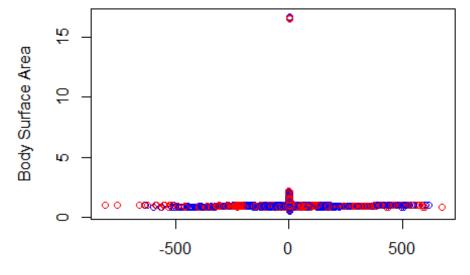
plot(x=childData$nHeightMulnWeight, y=childData$nHeightDivnWeight, xlab
="Normalized Height * Normalized Weight" , ylab ="Normalized Height /
Normalized Weight" , main ="Normalized (Height * Weight) vs Normalized
(Height / Weight)" , col=c("red", "blue"))</pre>
```



Normalized Height * Normalized Weight

```
plot(x=childData$nHeightDivnWeight, y=childData$bsa, xlab ="Normalized Height
* Normalized Weight", ylab = "Body Surface Area", main = "Normalized (Height
* Weight) vs Body surface area" , col=c("red","blue"))
```

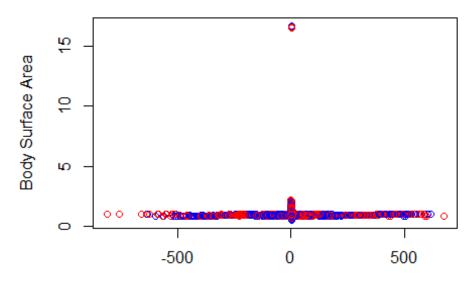
Normalized (Height * Weight) vs Body surface are



Normalized Height * Normalized Weight

```
plot(x=childData$nHeightDivnWeight, y=childData$bsa, xlab ="Normalized Height
/ Normalized Weight" , ylab ="Body Surface Area" , main ="Normalized (Height
/ Weight) vs Body surface area" , col=c("red","blue"))
```

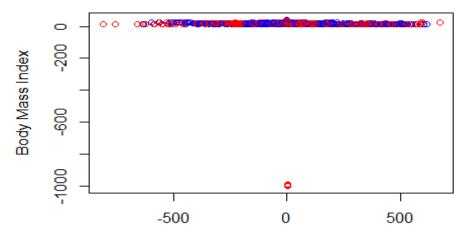
Normalized (Height / Weight) vs Body surface are



Normalized Height / Normalized Weight

plot(x=childData\$nHeightDivnWeight, y=childData\$bmi, xlab ="Normalized Height
/ Normalized Weight" , ylab ="Body Mass Index" , main ="Normalized (Height /
Weight) vs Body Mass Index" , col=c("red","blue"))

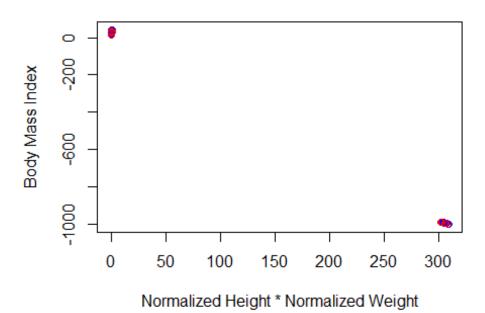
Normalized (Height / Weight) vs Body Mass Index



Normalized Height / Normalized Weight

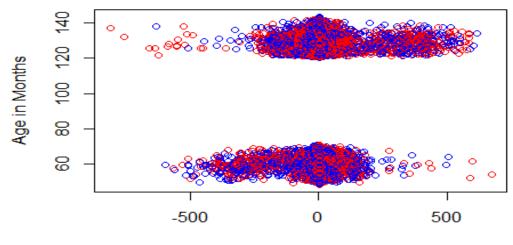
```
plot(x=childData$nHeightMulnWeight, y=childData$bmi, xlab ="Normalized Height
* Normalized Weight" , ylab ="Body Mass Index" , main ="Normalized (Height *
Weight) vs Body Mass Index" , col=c("red","blue"))
```

Normalized (Height * Weight) vs Body Mass Index



plot(x=childData\$nHeightDivnWeight, y=childData\$ageinmonths, xlab ="Normalized Height / Normalized Weight", ylab ="Age in Months", main ="Normalized (Height / Weight) vs Age in Months", col=c("red","blue"))

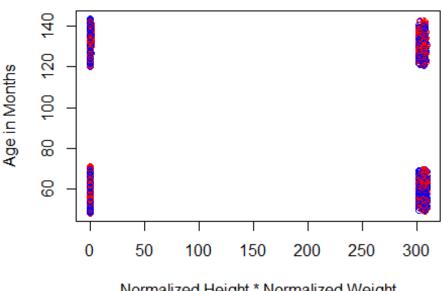
Normalized (Height / Weight) vs Age in Months



Normalized Height / Normalized Weight

```
plot(x=childData$nHeightMulnWeight, y=childData$ageinmonths, xlab
="Normalized Height * Normalized Weight", ylab ="Age in Months", main
="Normalized (Height * Weight) vs Age in Months", col=c("red","blue"))
```

Normalized (Height * Weight) vs Age in Months

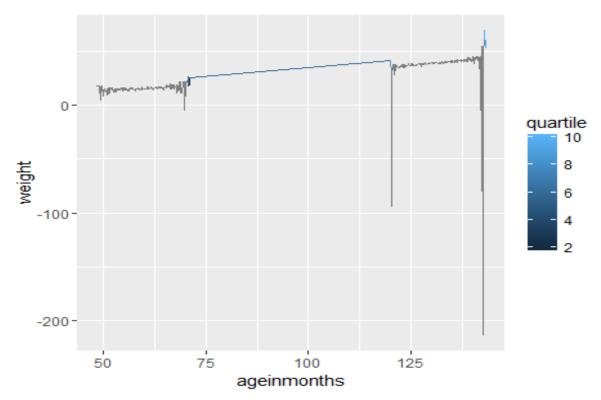


Normalized Height * Normalized Weight

5. Draw approximate growth curves over age. Calculate and plot growth curves of the different deciles (0%-10%, 10%-20%, 20%-30%, ...90%-100%) of obesity categories both by BMI and by weight, for both genders.

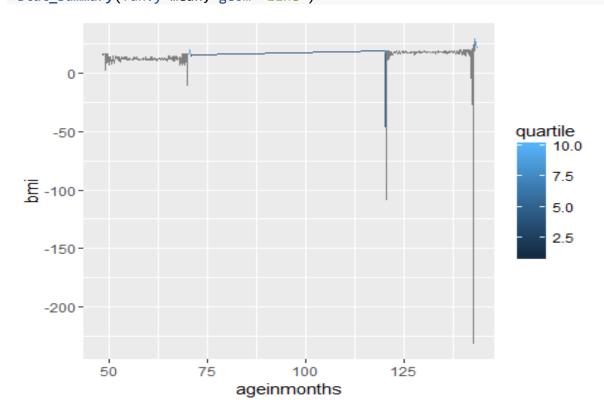
```
childData$genderdescription <-factor(childData$genderdescription)</pre>
library(nlme)
growth<-lme(height ~ageinmonths, data=childData, random= ~ageinmonths</pre>
|genderdescription, method="ML")
print(summary(growth))
## Linear mixed-effects model fit by maximum likelihood
##
    Data: childData
##
          AIC
                   BIC
                         logLik
     12643623 12643694 -6321805
##
##
## Random effects:
    Formula: ~ageinmonths | genderdescription
    Structure: General positive-definite, Log-Cholesky parametrization
##
##
               StdDev
                             Corr
## (Intercept) 1.344267109 (Intr)
## ageinmonths 0.007189537 -0.999
```

```
## Residual
               61.403857258
##
## Fixed effects: height ~ ageinmonths
                  Value Std.Error
                                        DF
                                            t-value p-value
## (Intercept) 71.87310 0.9638840 1141856 74.56613
                                                           0
## ageinmonths 0.54823 0.0053335 1141856 102.79057
                                                           0
   Correlation:
##
               (Intr)
## ageinmonths -0.986
##
## Standardized Within-Group Residuals:
             Min
                                                         Q3
##
                            Q1
                                          Med
                                                                      Max
## -18.663167642
                 -0.007958668
                                  0.058420722
                                                0.120625439
                                                              0.431812942
##
## Number of Observations: 1141859
## Number of Groups: 2
library(dplyr)
obesityByweight<-childData[,c(1,2,4)]
obesityByweight$quartile<-ntile(obesityByweight$weight, 10)</pre>
ggplot(obesityByweight, aes(ageinmonths, weight, color=quartile))
+stat_summary(fun.y=mean, geom="line")
```



```
obesityBybmi <-childData[,c(1,2,5)]
obesityBybmi$quartile <-ntile(obesityBybmi$bmi, 10)</pre>
```

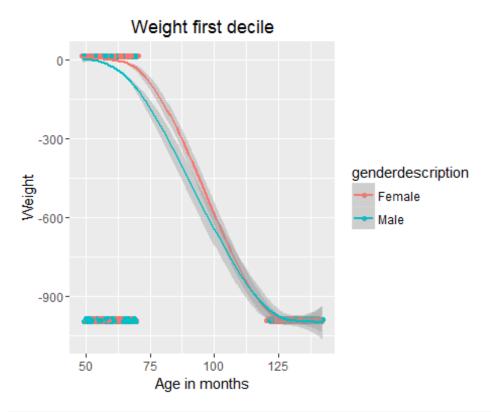
```
ggplot(obesityBybmi, aes(ageinmonths, bmi, color=quartile))
+stat summary(fun.y=mean, geom="line")
```



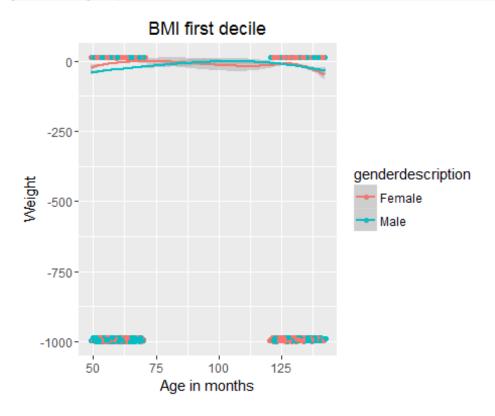
```
#Making subsets to plot based on Male or Female for 1st decile (0% - 10%)
firstdecile_weight <-subset(obesityByweight, quartile ==1)
firstdecile_weight <-na.omit(firstdecile_weight)
firstdecile_bmi <-subset(obesityBybmi, quartile ==1)
firstdecile_bmi <-na.omit(firstdecile_bmi)

# Separate regressions of Ageinmonths on weight for each gender

qplot(ageinmonths, weight, data=firstdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight first decile", xlab="Age in months", ylab="Weight")</pre>
```



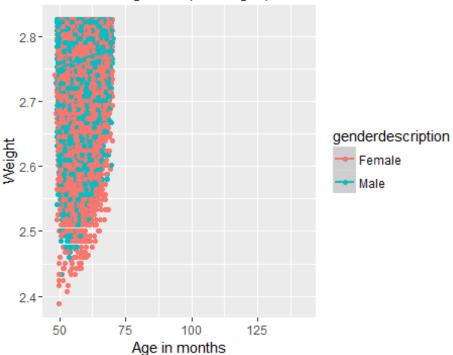
qplot(ageinmonths, bmi, data=firstdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI first decile", xlab="Age in months",
ylab="Weight")



```
#Using the log of variables weight and bmi to plot against time (Age in month) in order to get the growth curve
```

qplot(ageinmonths, log(weight), data=firstdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="First decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

First decile growth(in weight) curve



```
qplot(ageinmonths, log(bmi), data=firstdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="First decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")

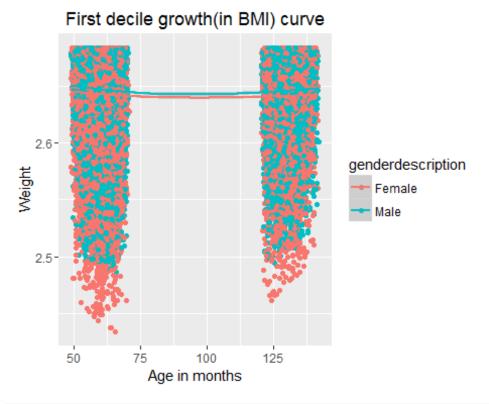
## Warning in log(bmi): NaNs produced

## Warning in log(bmi): NaNs produced

## Warning in log(bmi): NaNs produced

## Warning: Removed 3472 rows containing non-finite values (stat_smooth).

## Warning: Removed 3472 rows containing missing values (geom point).
```

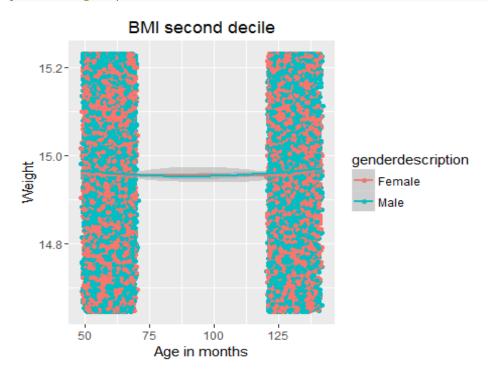


```
#Using Linear modelling function to first perform Regression and get the
intercept dor the growth curve
#print(lm(log(firstdecile_weight[2.5:2.8, "weight"]) ~
firstdecile_weight[2.5:2.8, "ageinmonths"]))
print(lm(log(firstdecile bmi[2.5:2.8,"bmi"])
~firstdecile bmi[2.5:3, "ageinmonths"]))
##
## Call:
## lm(formula = log(firstdecile_bmi[2.5:2.8, "bmi"]) ~ firstdecile_bmi[2.5:3,
       "ageinmonths"])
##
##
## Coefficients:
                              (Intercept)
##
##
                                    2.648
## firstdecile_bmi[2.5:3, "ageinmonths"]
#Making subsets to plot based on Male or Female for 2nd decile (10% - 20%)
seconddecile weight <-subset(obesityByweight, quartile ==2)</pre>
seconddecile_weight <-na.omit(seconddecile_weight)</pre>
seconddecile bmi <-subset(obesityBybmi, quartile ==2)</pre>
seconddecile_bmi <-na.omit(seconddecile_bmi)</pre>
# Separate regressions of Ageinmonths on weight for each gender
```

qplot(ageinmonths, weight, data=seconddecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight second decile", xlab="Age in
months", ylab="Weight")



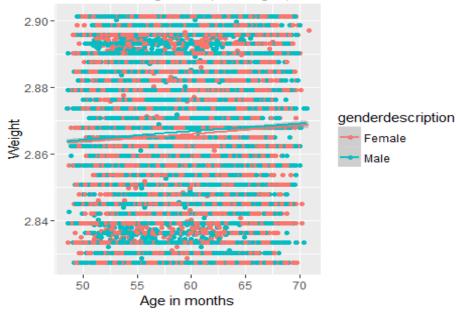
qplot(ageinmonths, bmi, data=seconddecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI second decile", xlab="Age in months",
ylab="Weight")



#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve

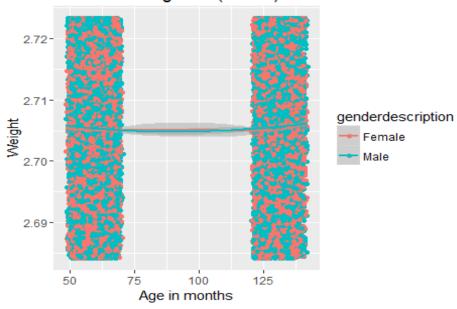
qplot(ageinmonths, log(weight), data=seconddecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Second decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

Second decile growth(in weight) curve



qplot(ageinmonths, log(bmi), data=seconddecile_bmi, geom=c("point",
"smooth"), color=genderdescription, main="Second decile growth(in BMI)
curve", xlab="Age in months", ylab="Weight")

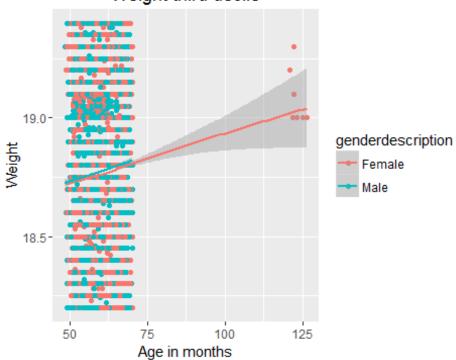
Second decile growth(in BMI) curve



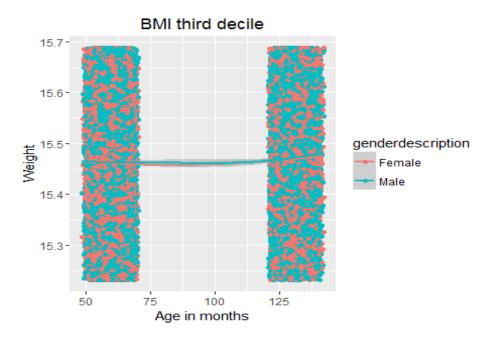
```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(seconddecile_weight[2.862:2.87,"weight"])
~seconddecile_weight[2.862:2.87, "ageinmonths"]))
Call:
lm(formula = log(seconddecile_weight[2.862:2.87, "weight"]) ~
seconddecile_weight[2.862:2.87, "ageinmonths"])
Coefficients:
                                    (Intercept)
                                           2.896
 seconddecile_weight[2.862:2.87, "ageinmonths"]
print(lm(log(seconddecile_bmi[2.685 :2.723,"bmi"]) ~seconddecile_bmi[2.685
:2.723, "ageinmonths"]))
##
## Call:
## lm(formula = log(seconddecile_bmi[2.685:2.723, "bmi"]) ~
seconddecile bmi[2.685:2.723,
       "ageinmonths"])
##
##
## Coefficients:
##
                                     (Intercept)
                                            2.693
## seconddecile bmi[2.685:2.723, "ageinmonths"]
##
                                               NA
#Making subsets to plot based on Male or Female for 3rd decile (20% - 30%)
thirddecile_weight <-subset(obesityByweight, quartile ==3)</pre>
thirddecile_weight <-na.omit(thirddecile_weight)</pre>
thirddecile_bmi <-subset(obesityBybmi, quartile ==3)</pre>
thirddecile bmi <-na.omit(thirddecile bmi)</pre>
```

Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=thirddecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight third decile", xlab="Age in
months", ylab="Weight")

Weight third decile



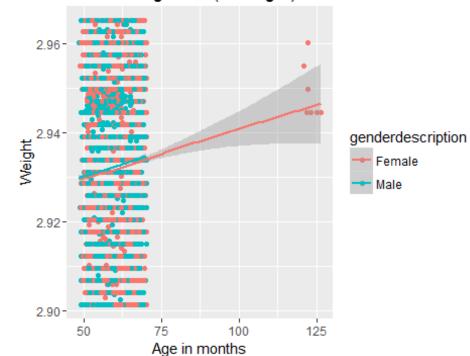
qplot(ageinmonths, bmi, data=thirddecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI third decile", xlab="Age in months",
ylab="Weight")



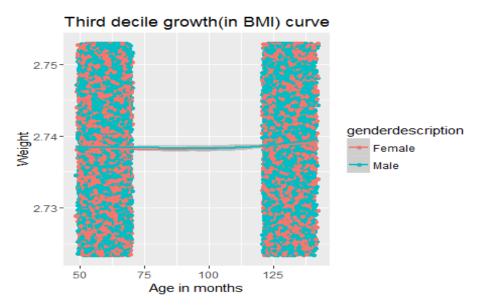
#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve

qplot(ageinmonths, log(weight), data=thirddecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Third decile growth (in weight)
curve", xlab="Age in months", ylab="Weight")

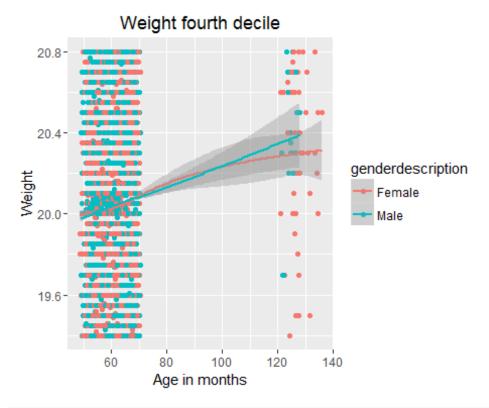
Third decile growth (in weight) curve



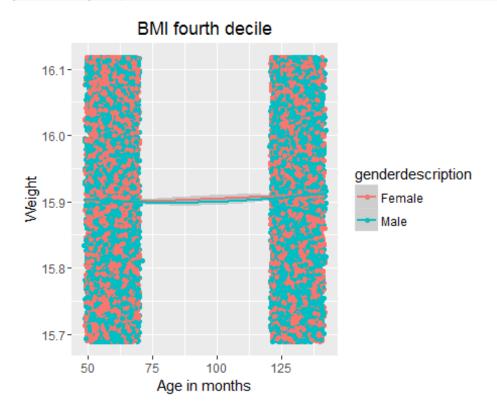
qplot(ageinmonths, log(bmi), data=thirddecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="Third decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")



```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(thirddecile_weight[2.93:2.95,"weight"])
~thirddecile weight[2.93:2.95, "ageinmonths"]))
##
## Call:
## lm(formula = log(thirddecile_weight[2.93:2.95, "weight"]) ~
thirddecile weight[2.93:2.95,
##
       "ageinmonths"])
##
## Coefficients:
##
                                     (Intercept)
##
                                           2.904
## thirddecile weight[2.93:2.95, "ageinmonths"]
print(lm(log(thirddecile_bmi[2.723 :2.753,"bmi"]) ~thirddecile_bmi[2.723
:2.753, "ageinmonths"]))
##
## Call:
## lm(formula = log(thirddecile bmi[2.723:2.753, "bmi"]) ~
thirddecile_bmi[2.723:2.753,
##
       "ageinmonths"])
##
## Coefficients:
##
                                    (Intercept)
                                          2.731
##
## thirddecile_bmi[2.723:2.753, "ageinmonths"]
#Making subsets to plot based on Male or Female for 4th decile (30% - 40%)
fourthdecile weight <-subset(obesityByweight, quartile ==4)</pre>
fourthdecile_weight <-na.omit(fourthdecile_weight)</pre>
fourthdecile_bmi <-subset(obesityBybmi, quartile ==4)</pre>
fourthdecile bmi <-na.omit(fourthdecile bmi)</pre>
# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=fourthdecile weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight fourth decile", xlab="Age in
months", ylab="Weight")
```



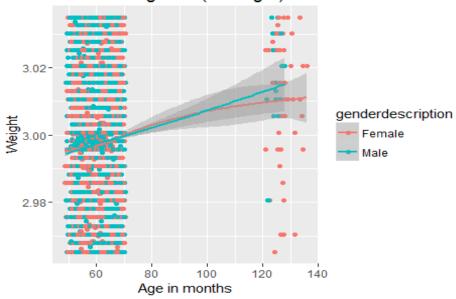
qplot(ageinmonths, bmi, data=fourthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI fourth decile", xlab="Age in months",
ylab="Weight")



#Using the log of variables weight and bmi to plot against time (Age in month) in order to get the growth curve

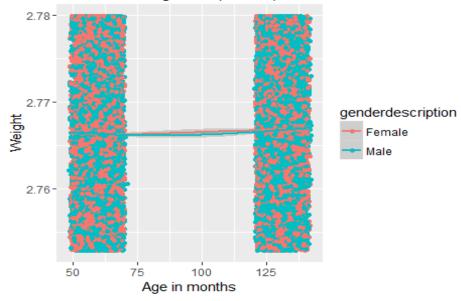
qplot(ageinmonths, log(weight), data=fourthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Fourth decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

Fourth decile growth(in weight) curve

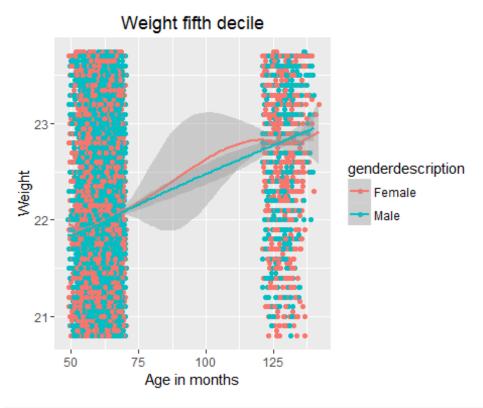


qplot(ageinmonths, log(bmi), data=fourthdecile_bmi, geom=c("point",
"smooth"), color=genderdescription, main="Fourth decile growth(in BMI)
curve", xlab="Age in months", ylab="Weight")

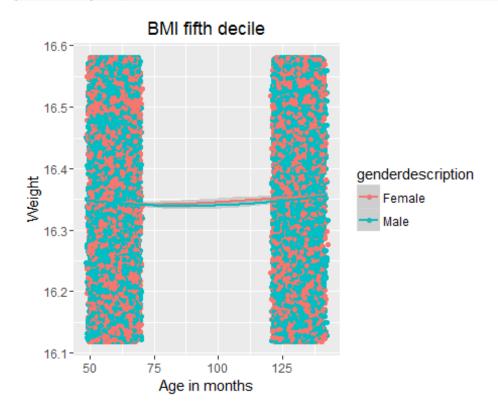
Fourth decile growth(in BMI) curve



```
#Using Linear modelling function to first perform Regression and get the
intercept dor the growth curve
print(lm(log(fourthdecile_weight[2.5:2.8,"weight"])
~fourthdecile_weight[2.5:2.8, "ageinmonths"]))
##
## Call:
## lm(formula = log(fourthdecile_weight[2.5:2.8, "weight"]) ~
fourthdecile weight[2.5:2.8,
##
       "ageinmonths"])
##
## Coefficients:
##
                                    (Intercept)
                                          2.996
##
## fourthdecile weight[2.5:2.8, "ageinmonths"]
print(lm(log(fourthdecile bmi[2.5:2.8,"bmi"])
~fourthdecile_bmi[2.5:3, "ageinmonths"]))
##
## Call:
## lm(formula = log(fourthdecile bmi[2.5:2.8, "bmi"]) ~
fourthdecile_bmi[2.5:3,
##
       "ageinmonths"])
##
## Coefficients:
##
                               (Intercept)
##
                                     2.755
## fourthdecile_bmi[2.5:3, "ageinmonths"]
#Making subsets to plot based on Male or Female for 5th decile (40% - 50%)
fifthdecile weight <-subset(obesityByweight, quartile ==5)</pre>
fifthdecile_weight <-na.omit(fifthdecile_weight)</pre>
fifthdecile_bmi <-subset(obesityBybmi, quartile ==5)</pre>
fifthdecile bmi <-na.omit(fifthdecile bmi)</pre>
# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=fifthdecile weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight fifth decile", xlab="Age in
months", ylab="Weight")
```



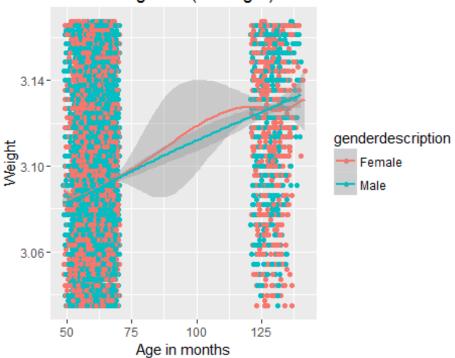
qplot(ageinmonths, bmi, data=fifthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI fifth decile", xlab="Age in months",
ylab="Weight")



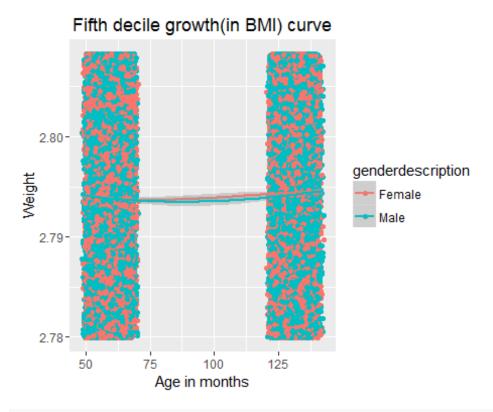
#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve

qplot(ageinmonths, log(weight), data=fifthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Fifth decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

Fifth decile growth(in weight) curve



qplot(ageinmonths, log(bmi), data=fifthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="Fifth decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")



```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(fifthdecile_weight[2.862:2.87, "weight"])
~fifthdecile_weight[2.862:2.87, "ageinmonths"]))
##
## Call:
## lm(formula = log(fifthdecile_weight[2.862:2.87, "weight"]) ~
       fifthdecile_weight[2.862:2.87, "ageinmonths"])
##
## Coefficients:
##
                                      (Intercept)
                                            3.054
##
## fifthdecile weight[2.862:2.87, "ageinmonths"]
##
print(lm(log(fifthdecile_bmi[2.685 :2.723,"bmi"]) ~fifthdecile_bmi[2.685
:2.723, "ageinmonths"]))
##
## Call:
## lm(formula = log(fifthdecile_bmi[2.685:2.723, "bmi"]) ~
fifthdecile_bmi[2.685:2.723,
##
       "ageinmonths"])
##
## Coefficients:
                                    (Intercept)
##
```

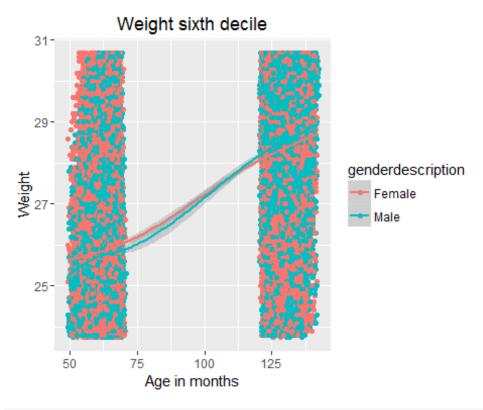
```
## fifthdecile_bmi[2.685:2.723, "ageinmonths"]
## NA

#Making subsets to plot based on Male or Female for 6th decile (50% - 60%)

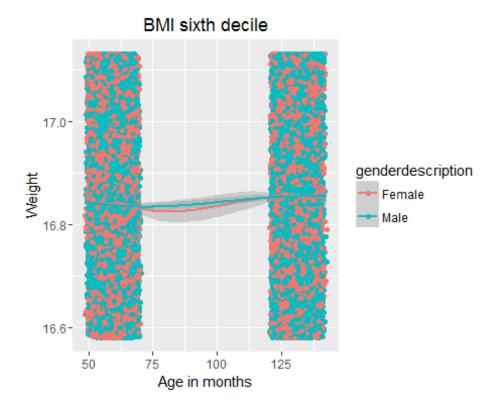
sixthdecile_weight <-subset(obesityByweight, quartile ==6)
sixthdecile_weight <-na.omit(sixthdecile_weight)
sixthdecile_bmi <-subset(obesityBybmi, quartile ==6)
sixthdecile_bmi <-na.omit(sixthdecile_bmi)

# Separate regressions of Ageinmonths on weight for each gender

qplot(ageinmonths, weight, data=sixthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight sixth decile", xlab="Age in months", ylab="Weight")</pre>
```



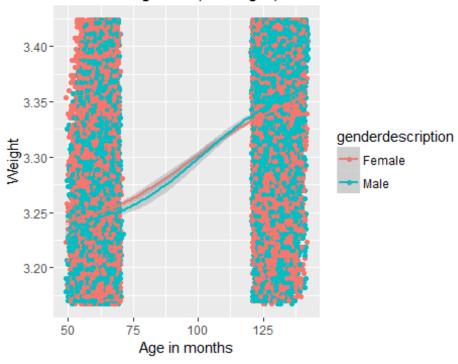
qplot(ageinmonths, bmi, data=sixthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI sixth decile", xlab="Age in months",
ylab="Weight")



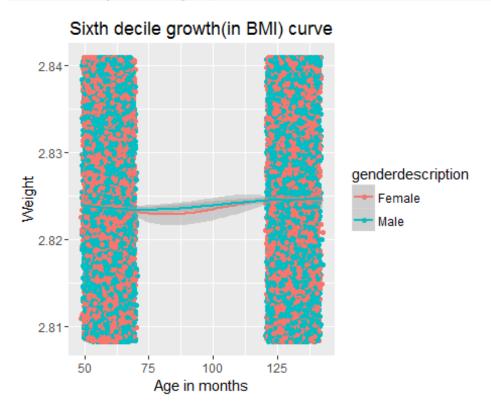
#Using the log of variables weight and bmi to plot against time (Age in month) in order to get the growth curve

qplot(ageinmonths, log(weight), data=sixthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Sixth decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

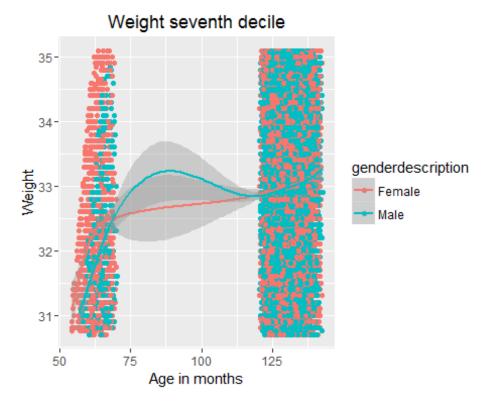
Sixth decile growth(in weight) curve



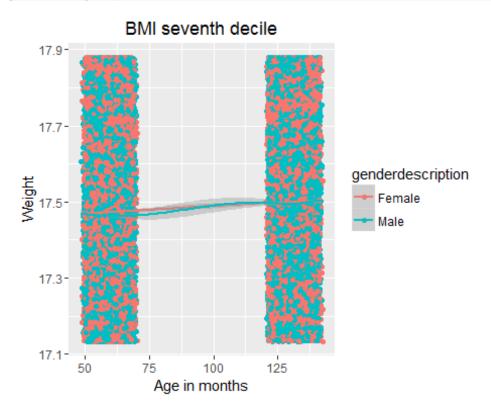
qplot(ageinmonths, log(bmi), data=sixthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="Sixth decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")



```
#Using Linear modelling function to first perform Regression and get the
intercept dor the growth curve
print(lm(log(sixthdecile_weight[2.5:2.8,"weight"])
~sixthdecile_weight[2.5:2.8, "ageinmonths"]))
##
## Call:
## lm(formula = log(sixthdecile_weight[2.5:2.8, "weight"]) ~
sixthdecile weight[2.5:2.8,
##
       "ageinmonths"])
##
## Coefficients:
##
                                   (Intercept)
##
                                         3.391
## sixthdecile weight[2.5:2.8, "ageinmonths"]
print(lm(log(sixthdecile bmi[2.5:2.8,"bmi"])
~sixthdecile_bmi[2.5:3, "ageinmonths"]))
##
## Call:
## lm(formula = log(sixthdecile_bmi[2.5:2.8, "bmi"]) ~ sixthdecile_bmi[2.5:3,
       "ageinmonths"])
##
##
## Coefficients:
##
                              (Intercept)
                                    2.833
##
## sixthdecile_bmi[2.5:3, "ageinmonths"]
#Making subsets to plot based on Male or Female for 7th decile (60% - 70%)
seventhdecile_weight <-subset(obesityByweight, quartile ==7)</pre>
seventhdecile weight <-na.omit(seventhdecile weight)</pre>
seventhdecile_bmi <-subset(obesityBybmi, quartile ==7)</pre>
seventhdecile_bmi <-na.omit(seventhdecile_bmi)</pre>
# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=seventhdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight seventh decile", xlab="Age
in months", ylab="Weight")
```

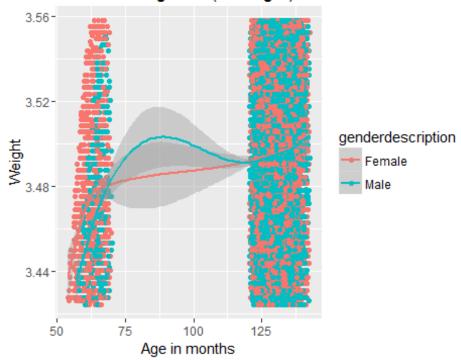


qplot(ageinmonths, bmi, data=seventhdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI seventh decile", xlab="Age in months",
ylab="Weight")



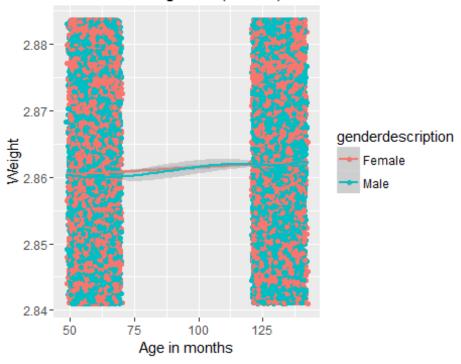
#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve
qplot(ageinmonths, log(weight), data=seventhdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Seventh decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

Seventh decile growth(in weight) curve



qplot(ageinmonths, log(bmi), data=seventhdecile_bmi, geom=c("point",
"smooth"), color=genderdescription, main="Seventh decile growth(in BMI)
curve", xlab="Age in months", ylab="Weight")

Seventh decile growth(in BMI) curve



```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(seventhdecile_weight[2.862:2.87,"weight"])
~seventhdecile_weight[2.862:2.87, "ageinmonths"]))
##
## Call:
## lm(formula = log(seventhdecile_weight[2.862:2.87, "weight"]) ~
       seventhdecile_weight[2.862:2.87, "ageinmonths"])
##
## Coefficients:
##
                                        (Intercept)
                                              3.487
##
## seventhdecile_weight[2.862:2.87, "ageinmonths"]
##
print(lm(log(seventhdecile_bmi[2.685 :2.723,"bmi"]) ~seventhdecile_bmi[2.685
:2.723, "ageinmonths"]))
##
## Call:
## lm(formula = log(seventhdecile_bmi[2.685:2.723, "bmi"]) ~
seventhdecile_bmi[2.685:2.723,
##
       "ageinmonths"])
##
## Coefficients:
                                      (Intercept)
##
```

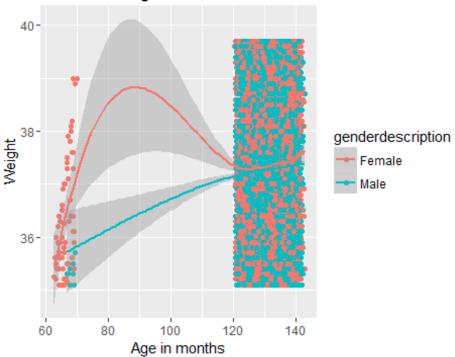
```
## 2.877
## seventhdecile_bmi[2.685:2.723, "ageinmonths"]
## NA

#Making subsets to plot based on Male or Female for 8th decile (70% - 80%)

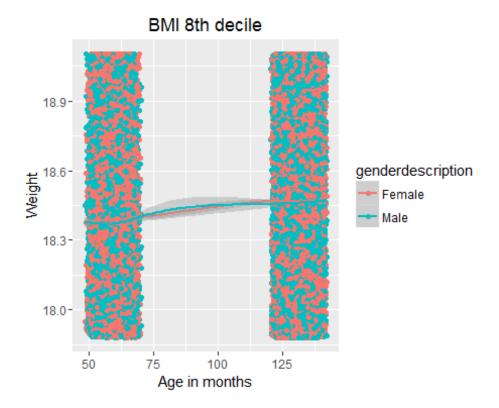
eightdecile_weight <-subset(obesityByweight, quartile ==8)
eightdecile_weight <-na.omit(eightdecile_weight)
eightdecile_bmi <-subset(obesityBybmi, quartile ==8)
eightdecile_bmi <-na.omit(eightdecile_bmi)

# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=eightdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight 8th decile", xlab="Age in months", ylab="Weight")</pre>
```

Weight 8th decile

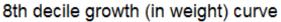


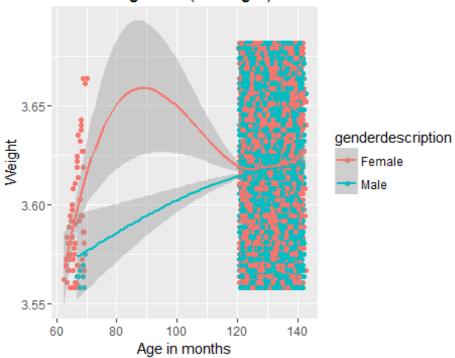
qplot(ageinmonths, bmi, data=eightdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI 8th decile", xlab="Age in months",
ylab="Weight")



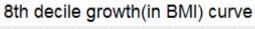
#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve

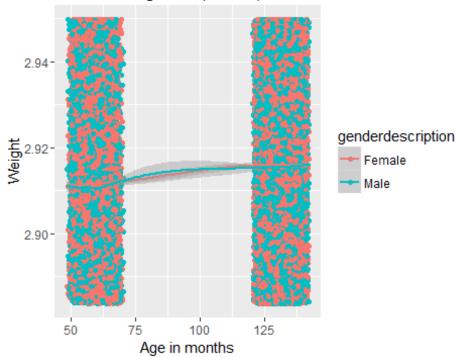
qplot(ageinmonths, log(weight), data=eightdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="8th decile growth (in weight)
curve", xlab="Age in months", ylab="Weight")



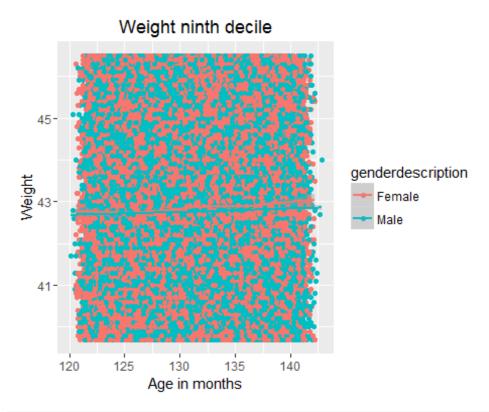


qplot(ageinmonths, log(bmi), data=eightdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="8th decile growth(in BMI) curve", xlab="Age in
months", ylab="Weight")

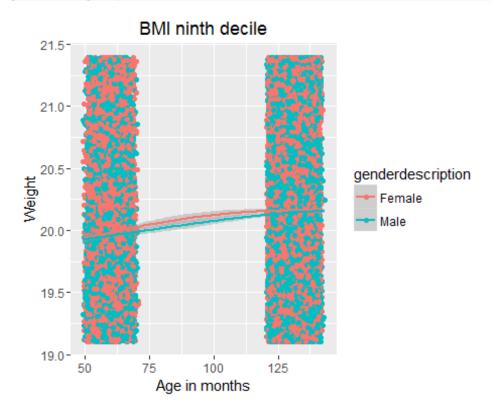




```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(eightdecile_weight[2.93:2.95,"weight"])
~eightdecile_weight[2.93:2.95, "ageinmonths"]))
##
## Call:
## lm(formula = log(eightdecile_weight[2.93:2.95, "weight"]) ~
eightdecile weight[2.93:2.95,
##
       "ageinmonths"])
##
## Coefficients:
##
                                     (Intercept)
##
                                           3.578
## eightdecile weight[2.93:2.95, "ageinmonths"]
print(lm(log(eightdecile_bmi[2.723 :2.753,"bmi"]) ~eightdecile_bmi[2.723
:2.753, "ageinmonths"]))
##
## Call:
## lm(formula = log(eightdecile bmi[2.723:2.753, "bmi"]) ~
eightdecile_bmi[2.723:2.753,
##
       "ageinmonths"])
##
## Coefficients:
##
                                    (Intercept)
##
                                          2.922
## eightdecile_bmi[2.723:2.753, "ageinmonths"]
#Making subsets to plot based on Male or Female for 9th decile (80% - 90%)
ninthdecile weight <-subset(obesityByweight, quartile ==9)</pre>
ninthdecile_weight <-na.omit(ninthdecile_weight)</pre>
ninthdecile_bmi <-subset(obesityBybmi, quartile ==9)</pre>
ninthdecile bmi <-na.omit(ninthdecile bmi)</pre>
# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=ninthdecile weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight ninth decile", xlab="Age in
months", ylab="Weight")
```



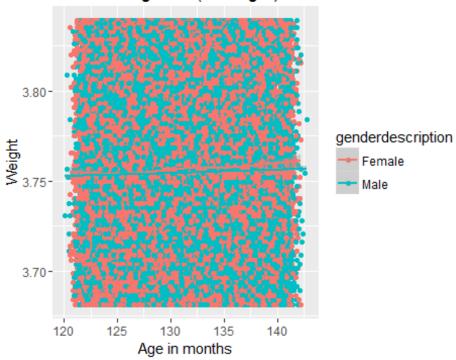
qplot(ageinmonths, bmi, data=ninthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI ninth decile", xlab="Age in months",
ylab="Weight")



#Using the log of variables weight and bmi to plot against time (Age in month) in order to get the growth curve

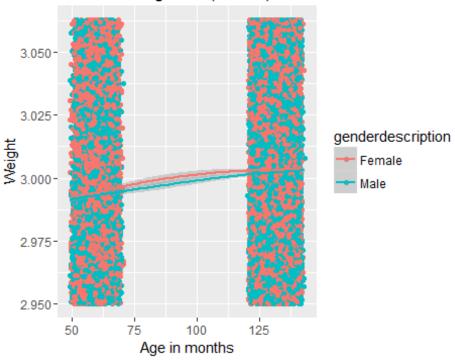
qplot(ageinmonths, log(weight), data=ninthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Ninth decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")

Ninth decile growth(in weight) curve



qplot(ageinmonths, log(bmi), data=ninthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="Ninth decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")

Ninth decile growth(in BMI) curve



```
#Using Linear modelling function to first perform Regression and get the
intercept dor the growth curve
print(lm(log(ninthdecile_weight[2.5:2.8,"weight"])
~ninthdecile_weight[2.5:2.8, "ageinmonths"]))
##
## Call:
## lm(formula = log(ninthdecile_weight[2.5:2.8, "weight"]) ~
ninthdecile_weight[2.5:2.8,
       "ageinmonths"])
##
##
## Coefficients:
##
                                   (Intercept)
##
                                         3.752
## ninthdecile_weight[2.5:2.8, "ageinmonths"]
##
print(lm(log(ninthdecile_bmi[2.5:2.8,"bmi"])
~ninthdecile_bmi[2.5:3, "ageinmonths"]))
##
## Call:
## lm(formula = log(ninthdecile_bmi[2.5:2.8, "bmi"]) ~ ninthdecile_bmi[2.5:3,
##
       "ageinmonths"])
##
## Coefficients:
                              (Intercept)
##
```

```
## 2.984
## ninthdecile_bmi[2.5:3, "ageinmonths"]
## NA

#Making subsets to plot based on Male or Female for 10th decile (90% - 100%)

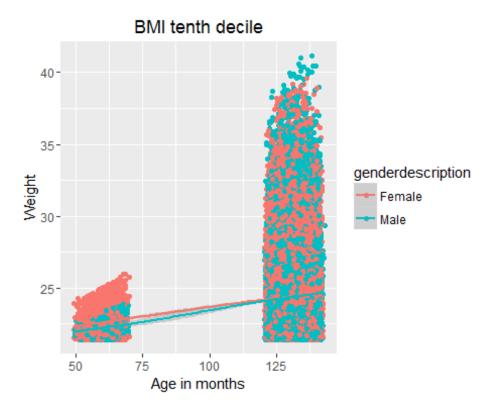
tenthdecile_weight <-subset(obesityByweight, quartile ==10)
tenthdecile_weight <-na.omit(tenthdecile_weight)
tenthdecile_bmi <-subset(obesityBybmi, quartile ==10)
tenthdecile_bmi <-na.omit(tenthdecile_bmi)

# Separate regressions of Ageinmonths on weight for each gender
qplot(ageinmonths, weight, data=tenthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Weight tenth decile", xlab="Age in months", ylab="Weight")</pre>
```

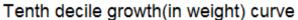
Weight tenth decile

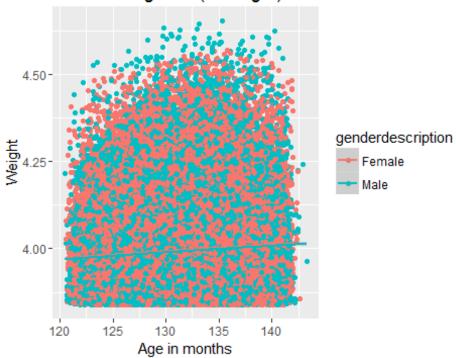


qplot(ageinmonths, bmi, data=tenthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="BMI tenth decile", xlab="Age in months",
ylab="Weight")

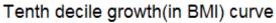


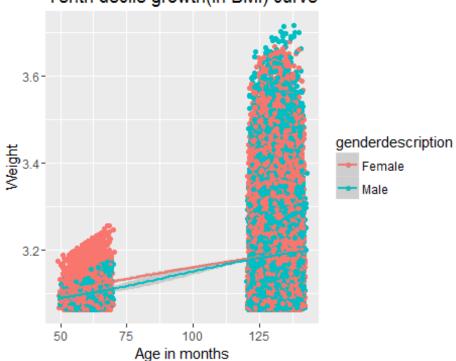
#Using the log of variables weight and bmi to plot against time (Age in
month) in order to get the growth curve
qplot(ageinmonths, log(weight), data=tenthdecile_weight, geom=c("point",
"smooth"), color=genderdescription, main="Tenth decile growth(in weight)
curve", xlab="Age in months", ylab="Weight")





qplot(ageinmonths, log(bmi), data=tenthdecile_bmi, geom=c("point", "smooth"),
color=genderdescription, main="Tenth decile growth(in BMI) curve", xlab="Age
in months", ylab="Weight")





```
#Using Linear modelling function to first perform Regression and get the
intercept for the growth curve
print(lm(log(tenthdecile_weight[2.862:2.87,"weight"])
~tenthdecile_weight[2.862:2.87, "ageinmonths"]))
##
## Call:
## lm(formula = log(tenthdecile_weight[2.862:2.87, "weight"]) ~
       tenthdecile weight[2.862:2.87, "ageinmonths"])
##
## Coefficients:
                                      (Intercept)
##
##
                                            3.875
## tenthdecile weight[2.862:2.87, "ageinmonths"]
##
print(lm(log(tenthdecile bmi[2.685 :2.723,"bmi"]) ~tenthdecile bmi[2.685
:2.723, "ageinmonths"]))
##
## Call:
## lm(formula = log(tenthdecile bmi[2.685:2.723, "bmi"]) ~
tenthdecile bmi[2.685:2.723,
       "ageinmonths"])
##
##
## Coefficients:
                                    (Intercept)
##
## tenthdecile_bmi[2.685:2.723, "ageinmonths"]
```

6. Attend the Tartu city open data event in the Garage 48 hub on Saturday (March 5th, 2pm). Participate in some working group. Write a report about the ideas, problem statement, envision the goals of the project and main steps that need to be executed in order to achieve the stated goals.

Kindergartens, Schools, planning...

Group: JaakVilo, OmisakinOluwatobi Samuel, Valdur Kana, MykhailoDorokhov, Kenigbolo Meya Stephen, Darwin Sivalingapandi, JevgeniSavostkin, FortunatMutunda

What is the need for DATA?

- **Population Registry** - all people by precise **age** (of children), **address**, and language (?) *It can be aggregated for privacy reasons...* (do many siblings affect the need - siblings should go to same (kindergarten/schools)

- **Kindergarten sizes and age distribution of kids currently in -** age groups/cohorts how many admitted and will be leaving at any given time point (know the number of available places upfront)
- Waiting list actual status of the kindergarten (who has signed up what address, when wants place, how many have already signed up - how many may be missing)
- **Where do kids actually go to kindergarten** now to understand the real "waste" travel (should siblings go to same school?)
- **New building permits** how many new families going to move in in short time from now... (match the future demand)
- Maybe some data about **playgrounds**, **transport**, ...
- **Public aggregated data snapshots made available for planning needs**... for businesses to build more efficient services (based on better planning)

WHY

I Planning needs: Predict and Match - demand and supply. Match a demand and supply 2-3 years ahead of actual time ... All the time real time changes.

Observe the dynamic changes - observe or predict who moves in, where, how does that change. (Time series, make predictions on trends)

New building permits - how many flats, what type of flats being built, what is GOING TO BE a new demand.

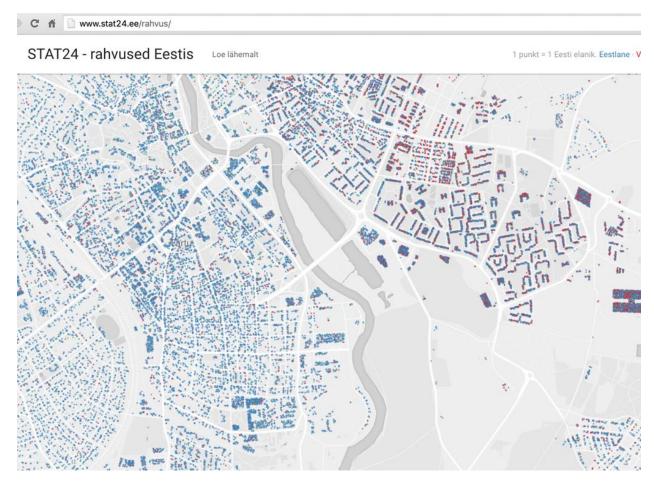
Goal: aid the city and urban planning for Tartu. (new kindergartens, closing, places volume tuning, new teachers to be hired, etc.).

II Teleport like scenario: like companies could take this to the global level examples. - Where and how to move in, what are the costs and options for new people...

Family looks for a place to live, they should know number of places in kindergartens/schools available and some qualitative criteria: Estonian schools vs schools

when it is possible to learn Estonian not being a native speaker. (Probably more valuable for international case)
III SWAPPING service: Are there parents/families that might want to switch the places? Find and propose those switches. E.g. after family moved to a new location their travel could have been made too long And some other may be there available
Travel distances and SWAPPING: service School catchment areas to calculate actual travel distances to kindergartens, schools.
Proximity to school/kindergarten. What are other factors - e.g. workplaces affecting choice of kindergarten ? Local demand can also be higher if many people work nearby
Research issues - how to ensure the privacy, differential privacy queries, etc. Play the different scenarios with models and predictions.
Can it be hacked together at Hackathon? Yes, it can:)

Example of visualisations of Estonians and Russians based purely on open data - census counts per postal code and open street map...



Public access - $\frac{\text{http://eid.ee/3z7}}{\text{public}}$

My personal conclusions

The report above is the final document that summarizes our discussions (barring some editing I made to the doc) however I will like to summarize and highlight some important factors about the topic.

Considering the intricate nature of the data we require, there definitely will be some issues surrounding policy and privacy. In as much as our intentions are good, having such sensitive data publicly could play into the hands of a wrong set of people (for example a pedophile interested in finding out where a large concentration of kids stay or a criminal getting to find out where the newest apartment buildings in the city are located) however as with any data made publicly available the possibility of such questions arising is quite high.

My suggestion will be the use of snapshots that will cover a specific time period. These snapshots could be monthly, quarterly or yearly and will help in comparing the trends in several situations.

Also this data can and most probably will be useful (if used and implemented via statistics) for the city council and government in budgeting as they will be able to properly plan for (if necessary) opening a new Kindergarten or school (the amount of kids to attend to etc.). It could also help marketers to know exactly where to position a business (for example, it makes sense to have an Ice cream shop close to a school or an amusement park close to where there are a large number of children).