### **Problem Set 3**

```
In [2]: library(ggplot2)
    library(NHANES)
    df = NHANES
    df = df[,c("BPSysAve","Age","Weight","Height","Gender")]
```

In [3]: head(df)

BPSysAve	Age	Weight	Height	Gender
113	34	87.4	164.7	male
113	34	87.4	164.7	male
113	34	87.4	164.7	male
NA	4	17.0	105.4	male
112	49	86.7	168.4	female
86	9	29.8	133.1	male

Here we are dropping those rows which does not have the BPSysAve as replacing them with mean gives us spurious results

```
In [4]: colSums(is.na(df))
df = df[complete.cases(df, df$BPSysAve),]

BPSysAve 1449
          Age 0
          Weight 78
```

In [36]: colSums(is.na(df))

BPSysAve 0
Age 0
Weight 0
Height 0
Gender 0

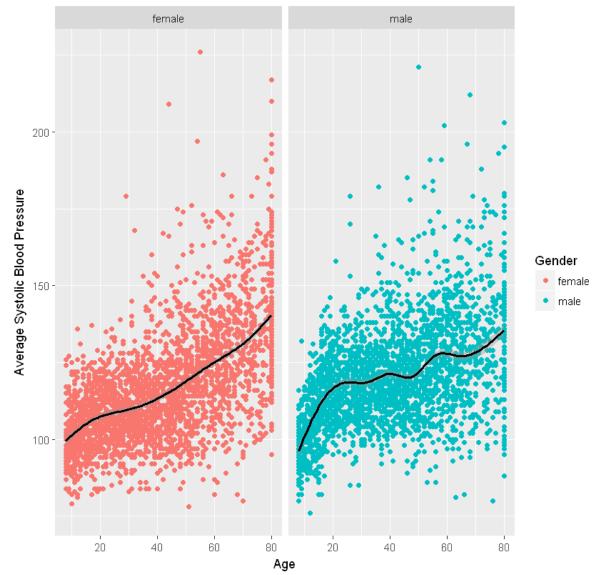
Height

Gender

353

## # Section1: Relationship between Average Systolic Blood Pressure and Age

#### Relationship between Average Systolic Blood Pressure and Age by Gender



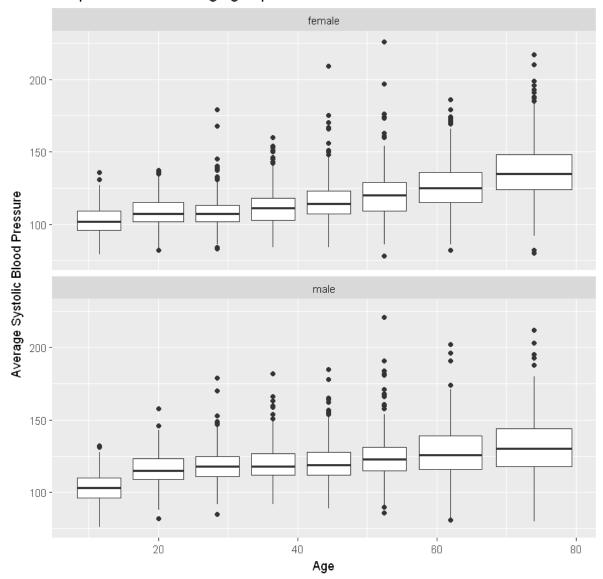
<sup>`</sup>geom\_smooth()` using method = 'gam'

It looks like there is a weak Positive Correlation between Average Systolic Blood Pressure and Age.

The average systolic Blood Pressure increases gradually with age for both genders.

A definitive linear trend is not observed.

#### Boxplots for different age groups

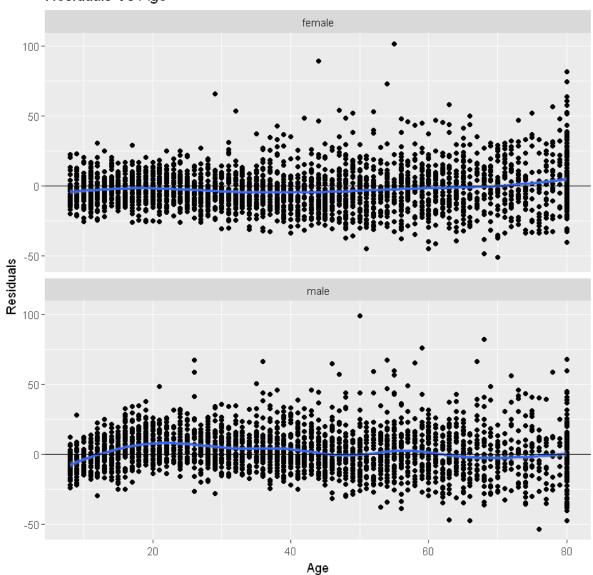


For each of the 8 age groups, it can be observed that the spread of average systolic blood pressure is similar with age-group for both genders.

```
In [113]: library(broom)
    df.lm= lm(BPSysAve ~ Age, data=df)
    df.lm.au = augment(df.lm)
    df.lm.au$Gender = df$Gender
```

`geom\_smooth()` using method = 'gam'

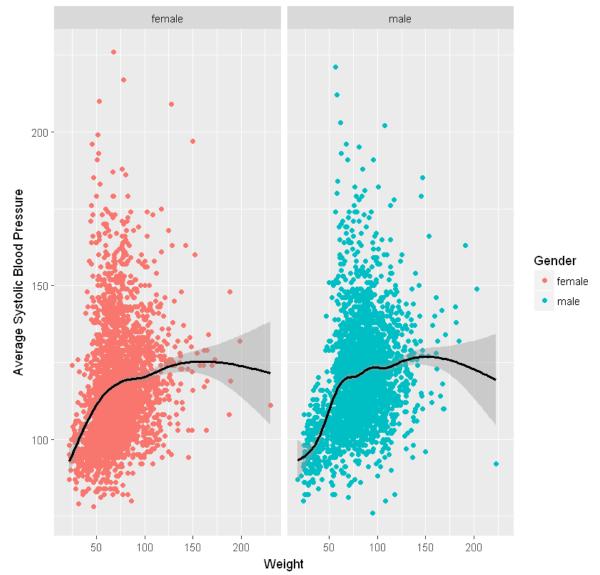
#### Residuals Vs Age



It looks like the model can be fitted best with a linear model. But the confidence level does not contain all the points, it means there is a less correlation between Average Systolic Blood Pressure and Age.

# **Section2: Relationship between Average Systolic Blood Pressure and Weight**

#### Relationship between Average Systolic Blood Pressure and Weight by Gender



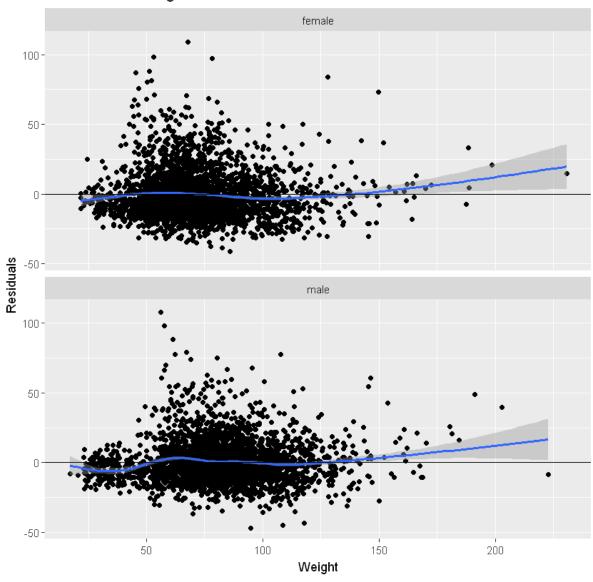
<sup>`</sup>geom\_smooth()` using method = 'gam'

Eyeballing at the above plot, it can be said that Linear approximation model is not appropriate for this data for both the genders.

For weights about 150 and above, the Blood Pressure values deviate from the existing trend.

```
In [100]: library(broom)
    df.lm.wt = lm(BPSysAve ~ Weight + I(Weight^2), data=df)
    df.lm.wt.au = augment(df.lm.wt)
    df.lm.wt.au$Gender = df$Gender
```

#### Residuals Vs Weight

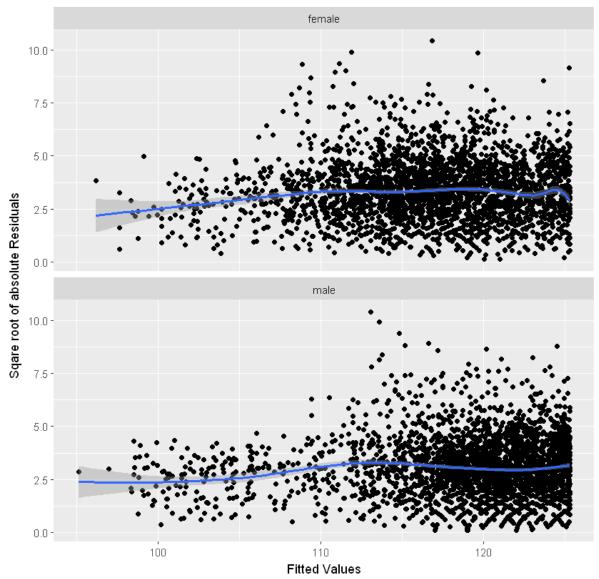


<sup>`</sup>geom\_smooth()` using method = 'gam'

The curve moves just around zero. Thus the Quadratic model seems apt for this data.

```
In [116]: ggplot(df.lm.wt.au, aes(x=.fitted, y= sqrt(abs(.resid)))) + geom_point() + geo
    m_smooth() +
        facet_wrap(~Gender,ncol=1) + xlab("Fitted Values") + ylab("Sqare root of abs
        olute Residuals") +
        ggtitle("Transformed Residuals Vs Fitted Values")
```

#### Transformed Residuals Vs Fitted Values

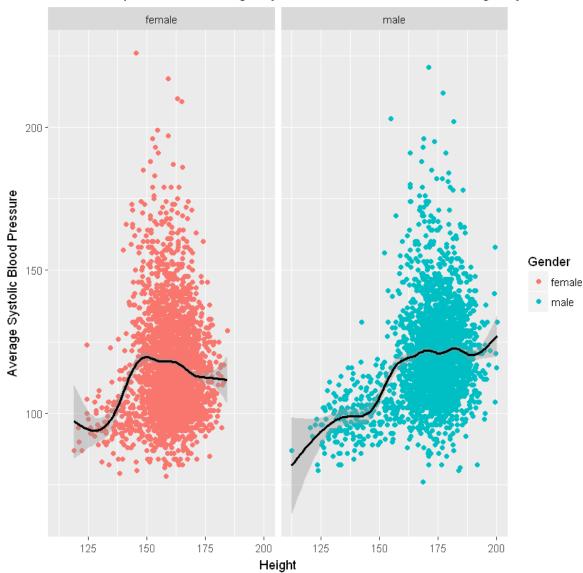


The curve clearly appears to be a horizontal line which proves Homoscedasticity in the data. But the confidence band does not cover majority of data points.

<sup>`</sup>geom\_smooth()` using method = 'gam'

### Section3: Relationship between Average Systolic Blood Pressure and Height

Relationship between Average Systolic Blood Pressure and Height by Gender



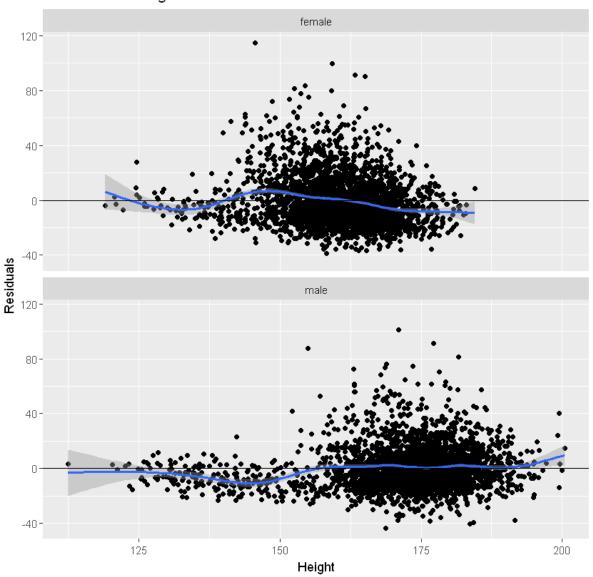
By looking at the above plot we can say that Linear model is not a best fit for the given data

<sup>`</sup>geom\_smooth()` using method = 'gam'

```
In [120]: library(broom)
    df.lm.ht = lm(BPSysAve ~ Height + I(Height^2), data=df)
    df.lm.ht.au = augment(df.lm.ht)
    df.lm.ht.au$Gender = df$Gender
```

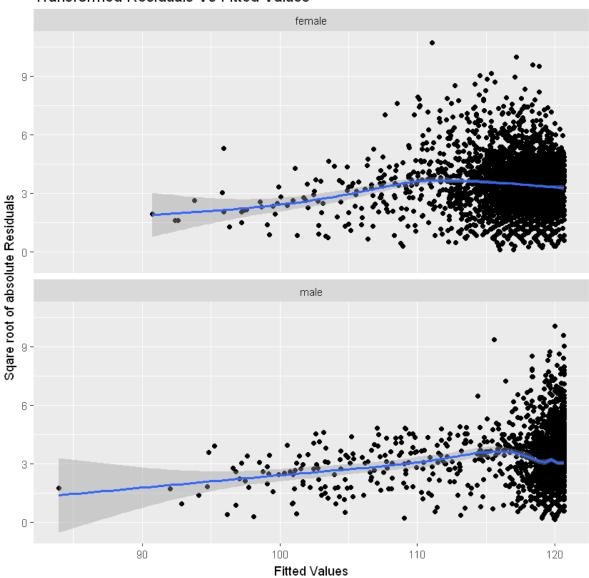
`geom\_smooth()` using method = 'gam'

#### Residuals vs Height



The residuals move just around zero. It implies that the quadratic model is appropriately fitted to the data

#### Transformed Residuals Vs Fitted Values



It can clearly be observed that the confidence band does not include majority of the data points. The line is nearly horizontal for most portion of the curve.

<sup>`</sup>geom\_smooth()` using method = 'gam'