

# Longitudinal Data Analysis

Case study of Trenal.XLS using Linear Mixed Effect Model

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# Theory of Linear Mixed Effects Model(LMM)

## Index description

Let us assume that a given input data set  $X$  has a dimension  $N \times p$ , with  $N$  observations and  $p$  predictors. For each subject indexed with  $i, i = 1, \dots, I$ , we can build a linear mixed effect model

$$\mathbf{Y}_i = \mathbf{X}_i\beta + \mathbf{Z}_i\mathbf{b}_i + \epsilon_i$$

## The application of LMM

Linear Mixed Effects Model is used to analyse a data set, where the observations are not fully independent, while the top level clusters are assumed independent. Inside each cluster, the observations are correlated

## Data set Trenal.XLS pre-analysis

### The summary of the data set

#### Import data

```
library(readxl)
trenal <- read_excel("Trenal.XLS") # summary(trenal)
```

#### Data Preprocessing

```
trenal= trenal[,-18] #remove a noninformative column const
# Continuous or discrete variables
trenal$id = as.factor(trenal$id)
trenal$j = as.factor(trenal$j)
trenal$male = as.factor(trenal$male)
trenal$cardio = as.factor(trenal$cardio)
trenal$reject = as.factor(trenal$reject)

trenal.long = trenal[,13:20] # long table form
trenal.wide = trenal[,1:17] # wide table form

library(magrittr) # needs to be run every time you start R and want to use %>%
library(dplyr)

##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

data.long <- trenal.long %>% # reordered long table
  relocate(id) %>%
  relocate(j,.after=id)%>%
  relocate(time,.after = j)%>%
```

```
relocate(respons, .after=time)
#summary(data.long)
sum(!is.na(data.long$respons))
```

```
## [1] 9558
```

```
data.long.noNA <- na.omit(data.long)# reordered long table without NAs
summary(data.long.noNA)
```

```
##          id          j          time          respons          age
## 3      : 12      2      :1159   Min.    : 0.000   Min.    :14.00   Min.    :15.00
## 5      : 12      1      :1158   1st Qu.: 1.000   1st Qu.:34.00   1st Qu.:35.00
## 6      : 12      3      :1158   Median : 3.000   Median :38.00   Median :46.00
## 8      : 12      4      :1072   Mean    : 3.432   Mean    :38.24   Mean    :45.27
## 9      : 12      5      : 954   3rd Qu.: 6.000   3rd Qu.:42.00   3rd Qu.:56.00
## 10     : 12      6      : 845   Max.    :10.000   Max.    :65.00   Max.    :76.00
## (Other):9479   (Other):3205
## male      cardio  reject
## 0:4213    0:7927  0:6314
## 1:5338    1:1624  1:3237
##
##
##
##
##
```

```
data.long.noNA$id[length(data.long.noNA$id)]
```

```
## [1] 1160
```

```
## 1160 Levels: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 ... 1160
```

## Response variable and predictors

**Response variable** From the `summary(data.long.noNA)`, we can read that the response variable is a continuous variable `respons` HC level from (15, 76) with Mean 38.24.

We have totally  $I = 1160$  ids for subjects. Ideally each id would have 12 (start from  $HC_0, HC_{0.5}, HC_1, HC_2, \dots, HC_{10}$ ) HC level measurement, but in really not all subjects have all of the 12 measurements. We have totally  $N = 9558 = \sum_{i=1}^I n_i$  missing values.

**Predictors** The explaining variables are

1.  $X_1 = time$  in year as discrete values, only changes with  $j, j = 1, \dots, n_i$
2.  $X_2 = age$  in year with 12 NAs; will only change with subject id  $i$
3.  $X_3 = male$  0 = female, 1 = male; will only change with subject id  $i$
4.  $X_4 = cardio$  0 = no, 1 = yes; will only change with subject id  $i$
5.  $X_5 = reject$  0 = accept, 1 = reject; will only change with subject id  $i$
6. fixed intercept, continuous

In the data analysis part, we need to try out different variables accounting for fixed effects and random effects.

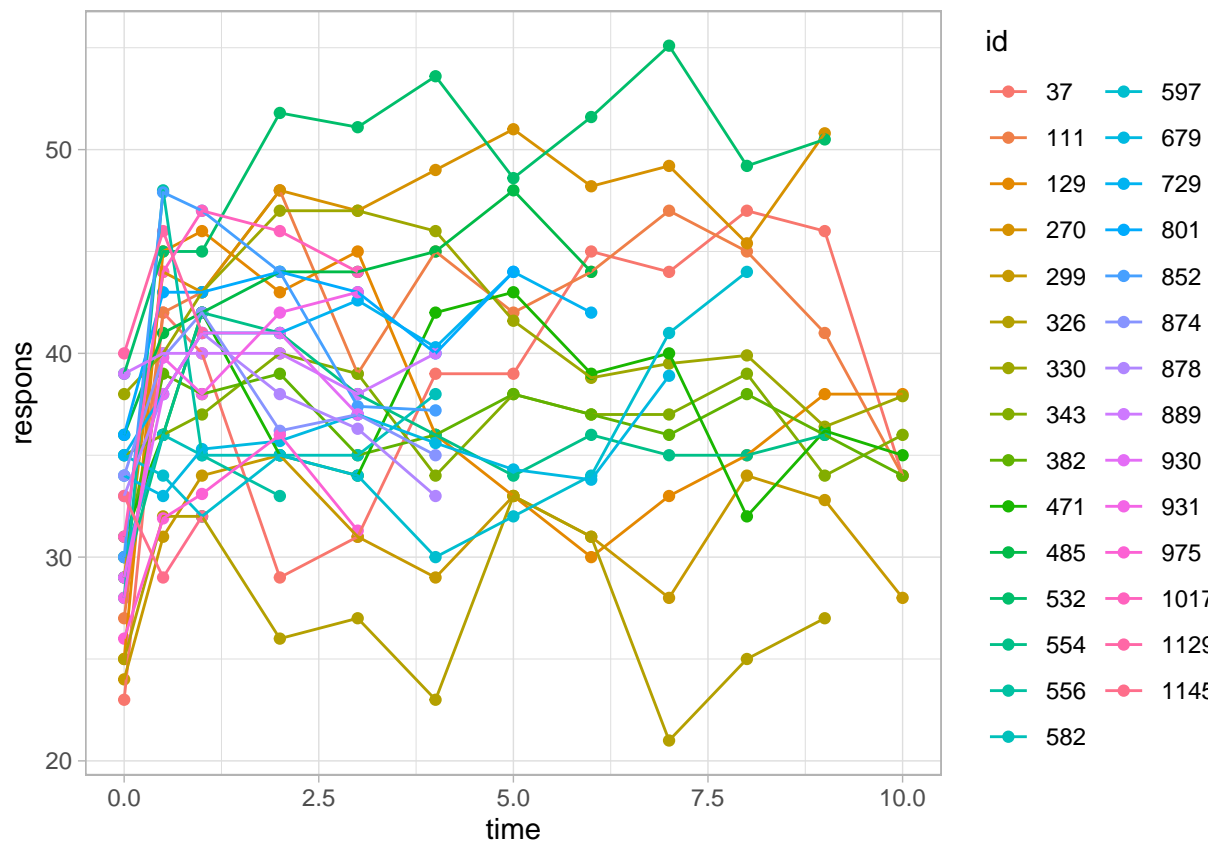
## Data visualization and the information from the data

```
library(ggplot2)
dim(data.long.noNA)
```

```
## [1] 9551      8
```

```
# since the data dimension is large 9551 x 8, we can select random 30 data to have a look
set.seed(1)
selected <- sample(1:length(unique(data.long.noNA$id)),30,replace=T) # random samples and permutations
#selected.vector = as.vector(selected)
data.selected = data.long.noNA[(data.long.noNA$id %in% c(selected)), ]
```

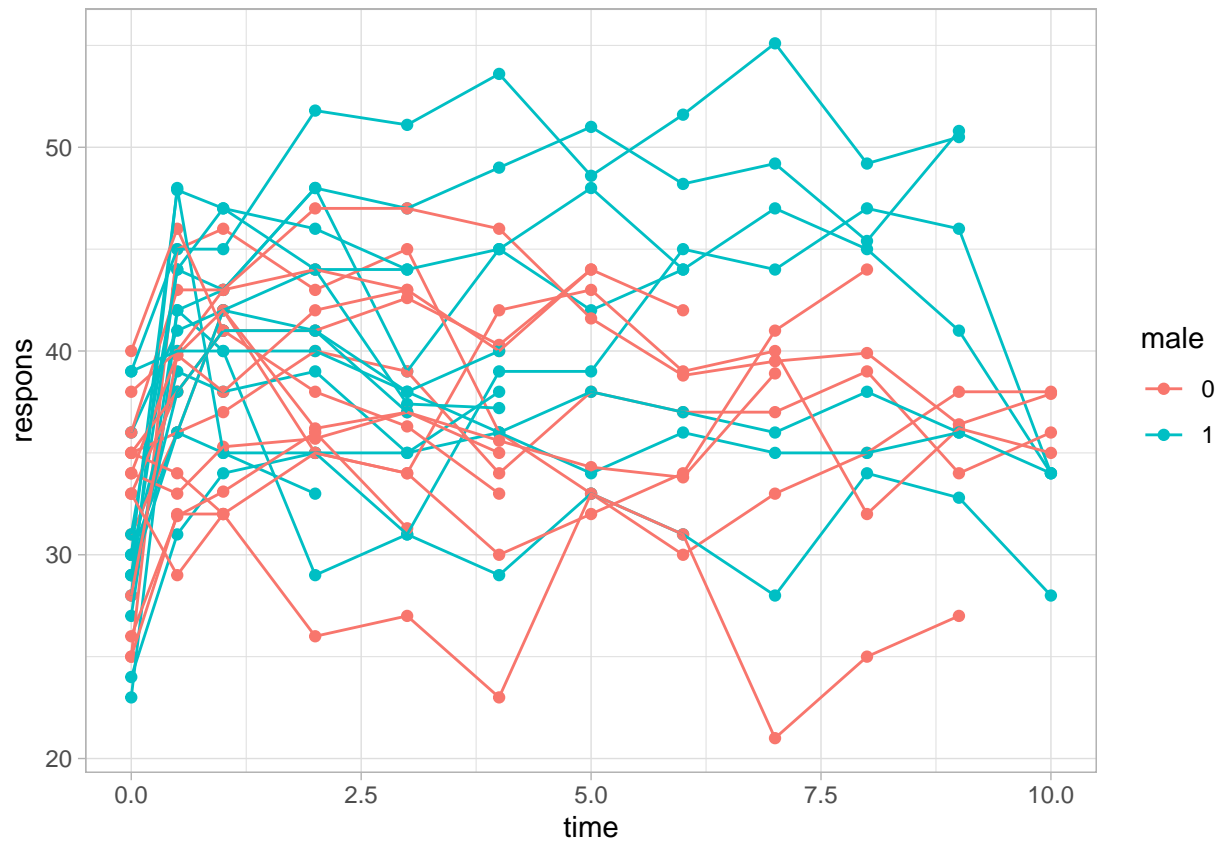
```
ggplot(data.selected,aes(x=time,y=respons,group=id,color=id))+geom_point()+ geom_line()+theme_light()
```



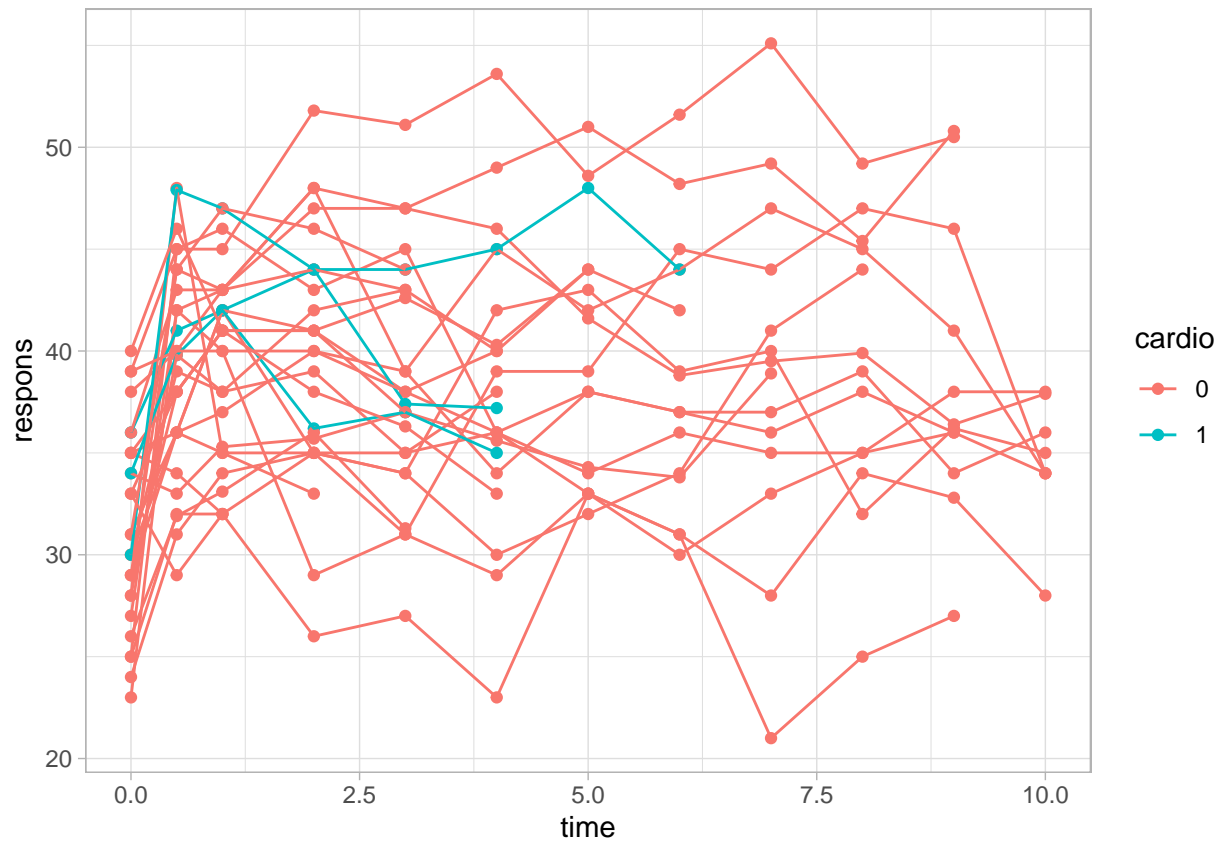
### Spaghettic plot

Plot Info: The intercept may vary according to each individual The slope is not very easy to see

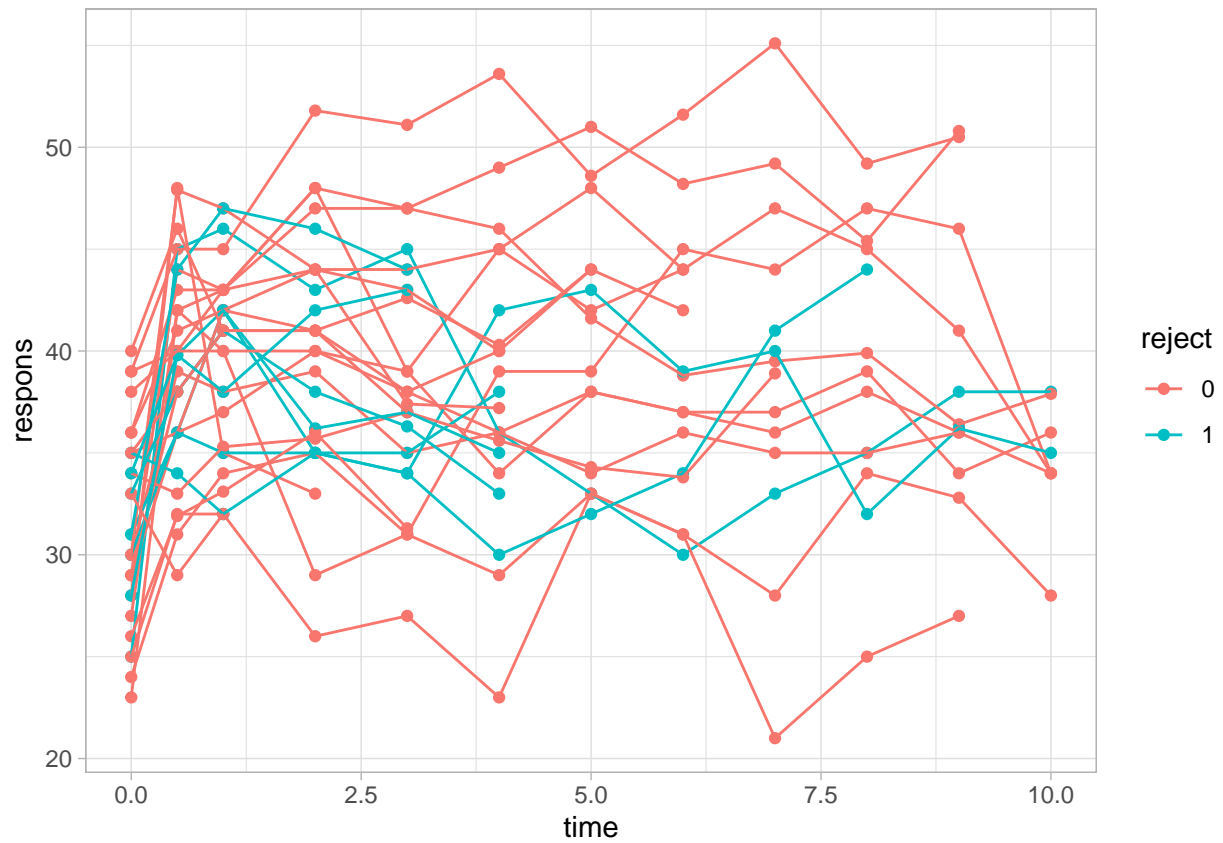
```
ggplot(data.selected,aes(x=time,y=respons,group=id,color=sex))+geom_point()+ geom_line()+theme_light()
```



```
ggplot(data.selected,aes(x=time,y=respons,group=id,color=cardio)) +geom_point()+ geom_line()+theme_ligh
```

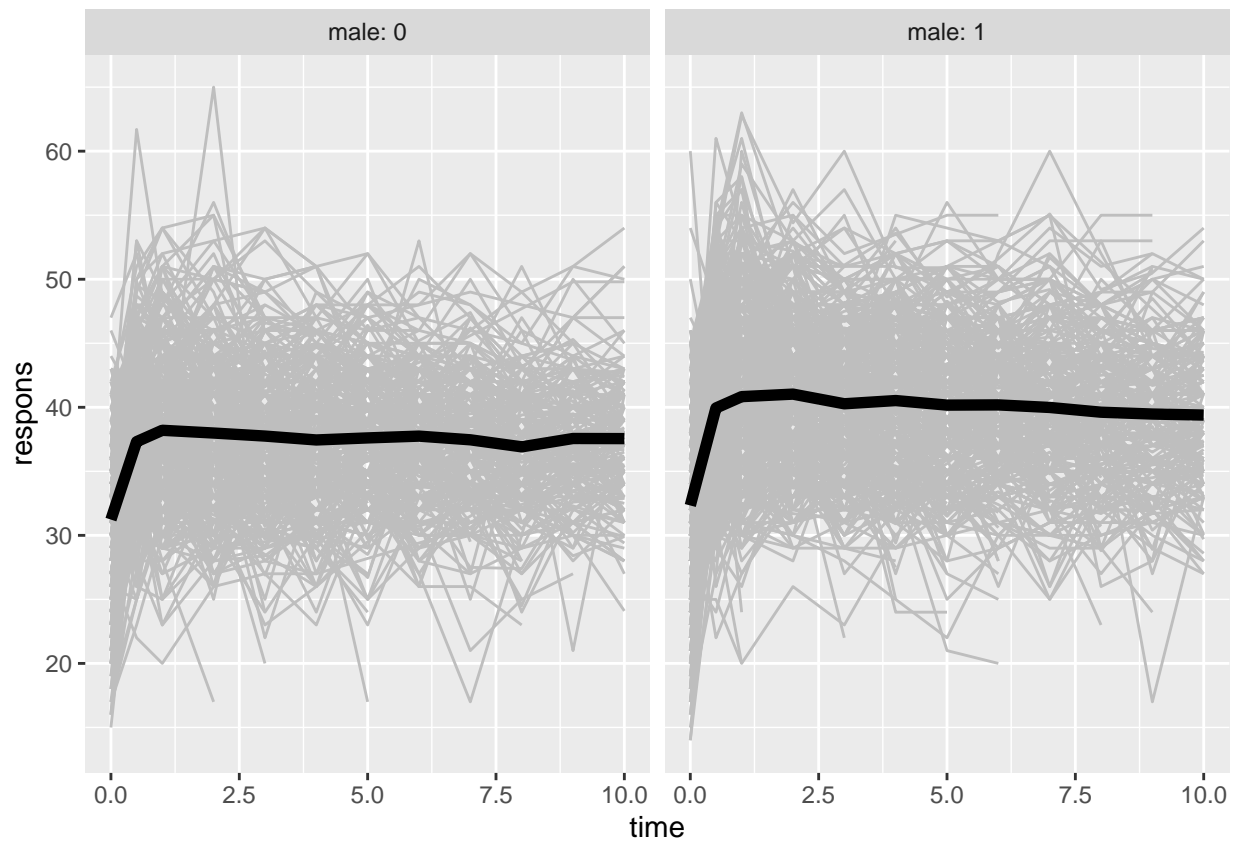


```
ggplot(data.selected,aes(x=time,y=respons,group=id,color=reject)) +geom_point()+ geom_line()+theme_ligh
```

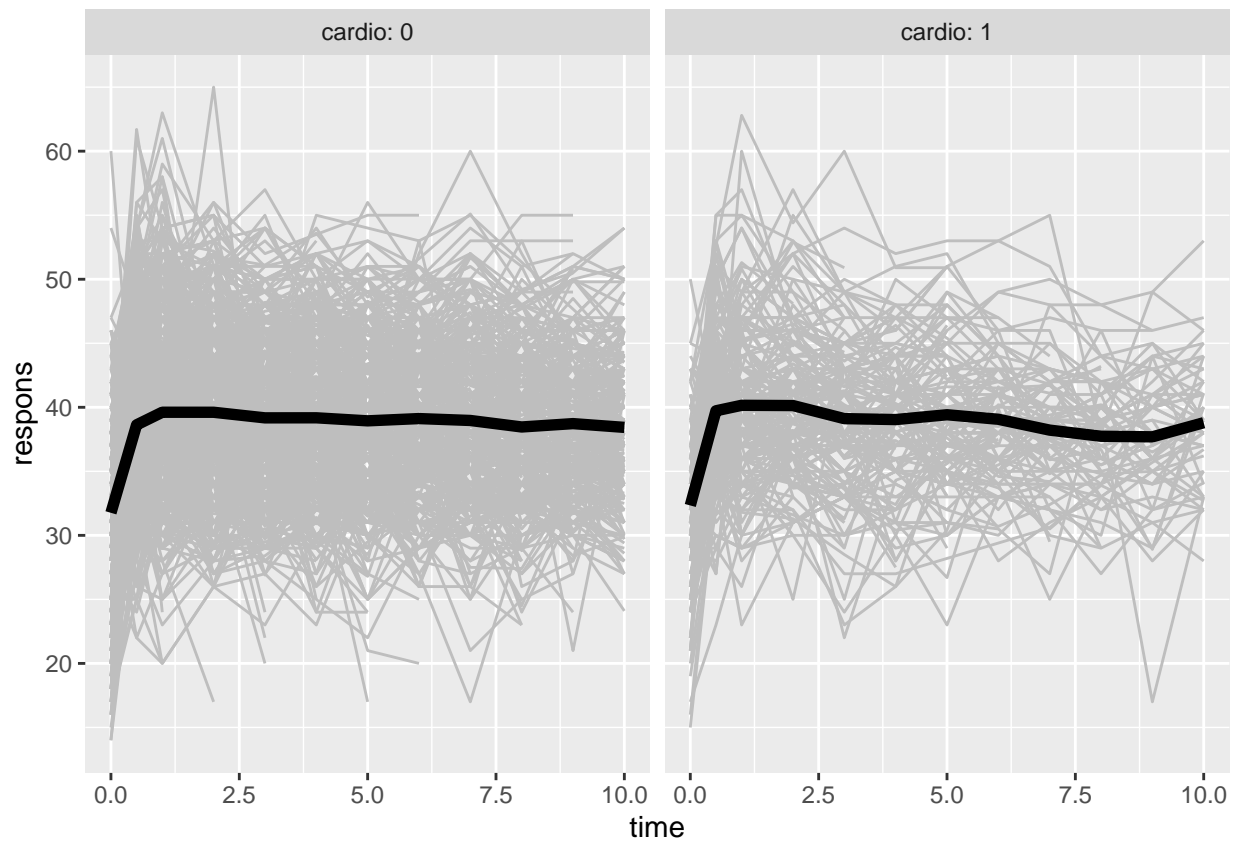


```
# Spaghetti Ggplot separated by male =1
p <- ggplot(data=data.long.noNA,aes(x=time,y=respons,group=id))
p <- p + geom_line(col="grey")+stat_summary(aes(group=1),geom="line",fun=mean,linewidth=2)
p + facet_grid(~male,labeller=label_both)
```

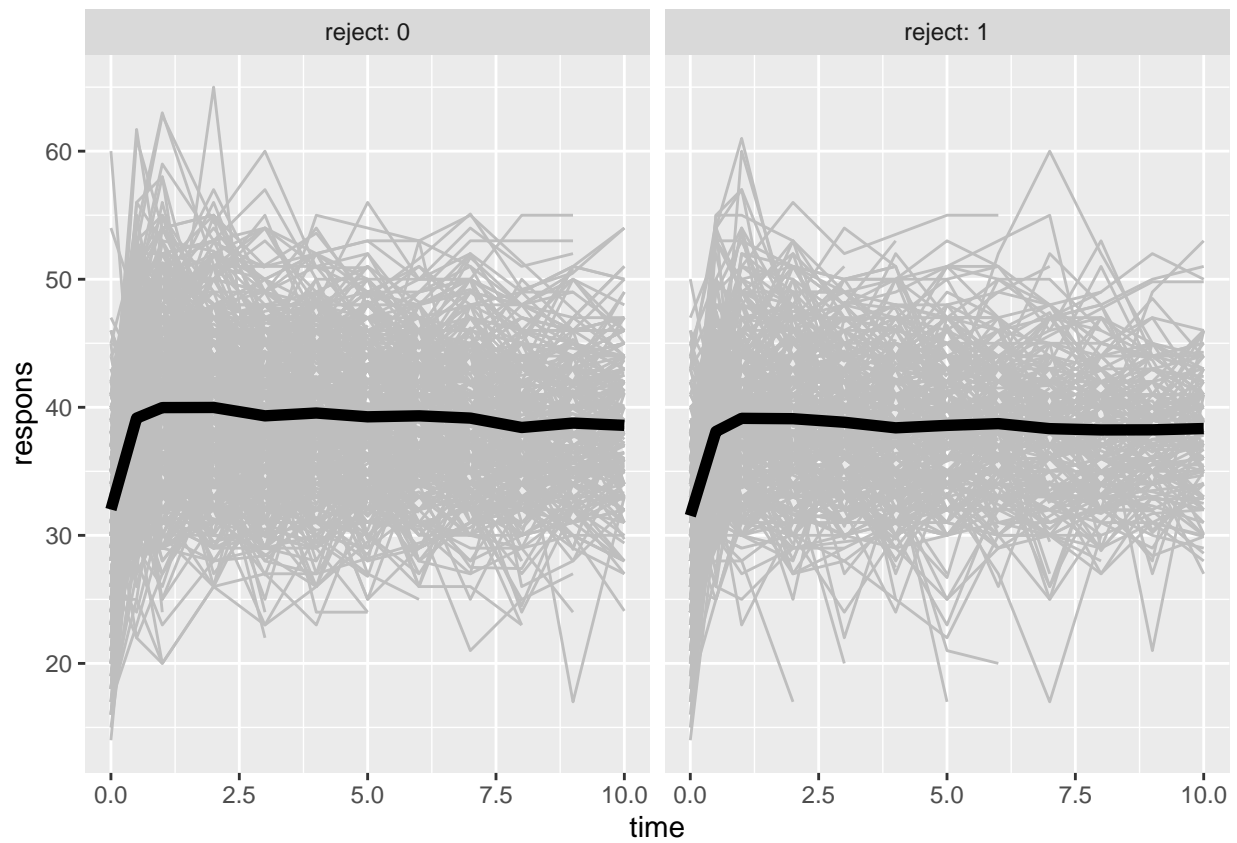




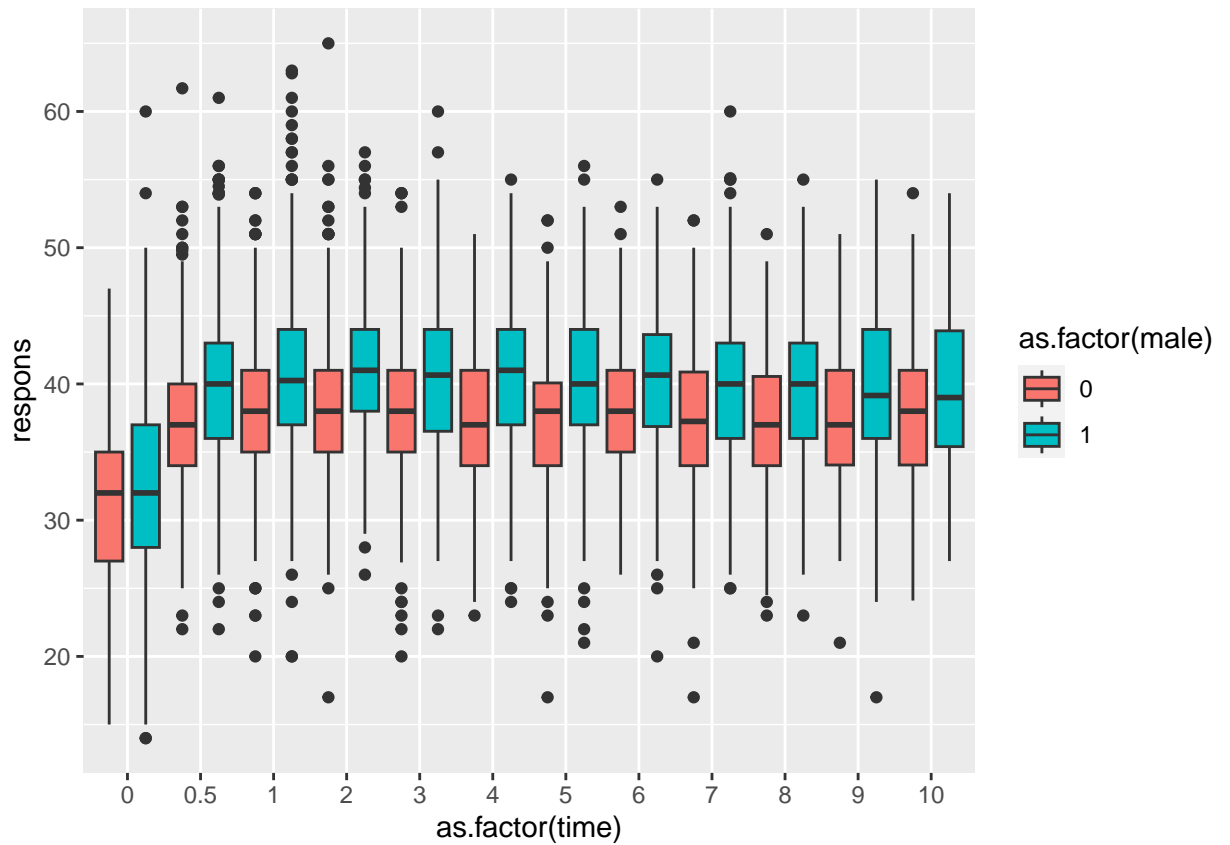
```
# Spaghetti Ggplot separated by cardio
p <- ggplot(data=data.long.noNA,aes(x=time,y=respons,group=id))
p <- p + geom_line(col="grey")+stat_summary(aes(group=1),geom="line",fun=mean,linewidth=2)
cardio.labs <- c("Cardio = 0","Cardio = 1")
p + facet_grid(~cardio,labeller = label_both)
```



```
# Spaghetti Ggplot separated by reject =1
p <- ggplot(data=data.long.noNA,aes(x=time,y=respons,group=id))
p <- p + geom_line(col="grey")+stat_summary(aes(group=1),geom="line",fun=mean,linewidth=2)
p + facet_grid(~reject,labeller=label_both)
```

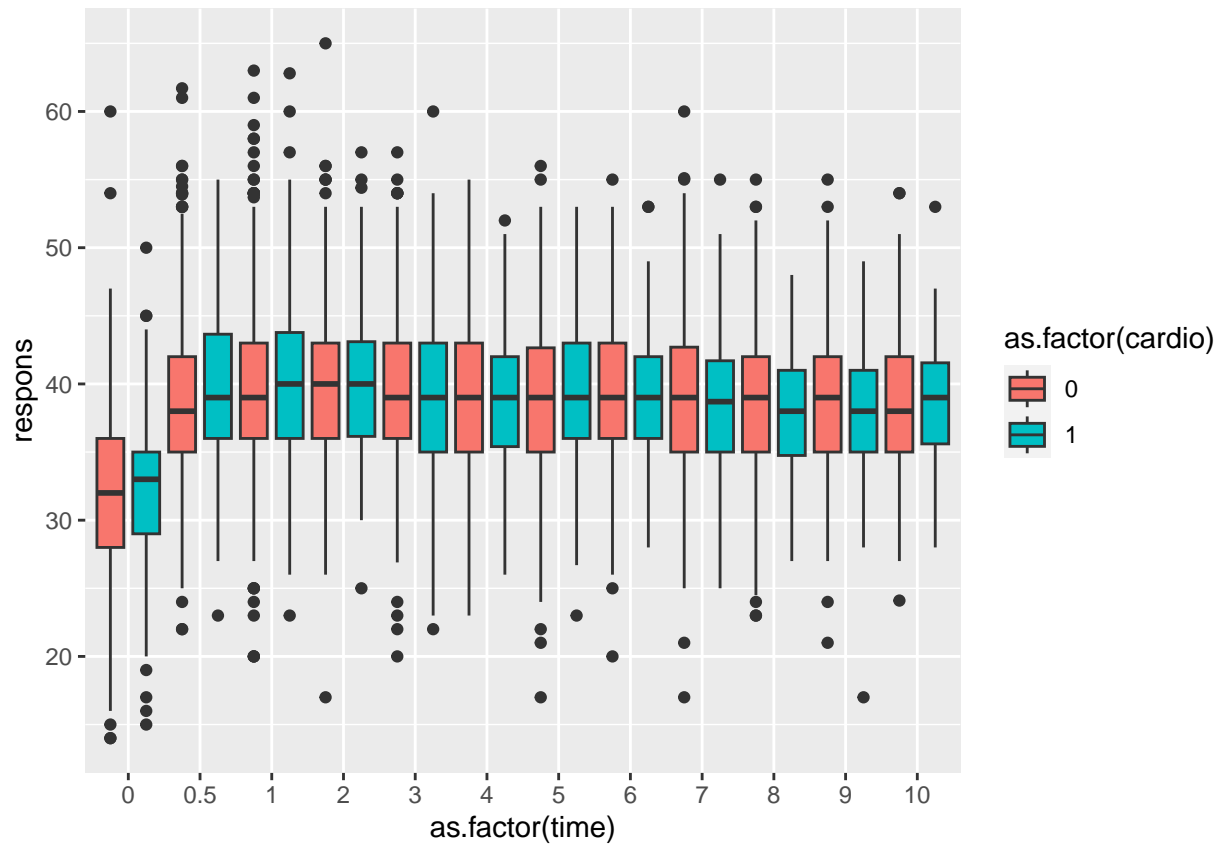


```
# Box plot by sex  
ggplot(data.long.noNA,aes(x=as.factor(time),y=respons,fill=as.factor(male)))+  
  geom_boxplot(position=position_dodge(1))
```

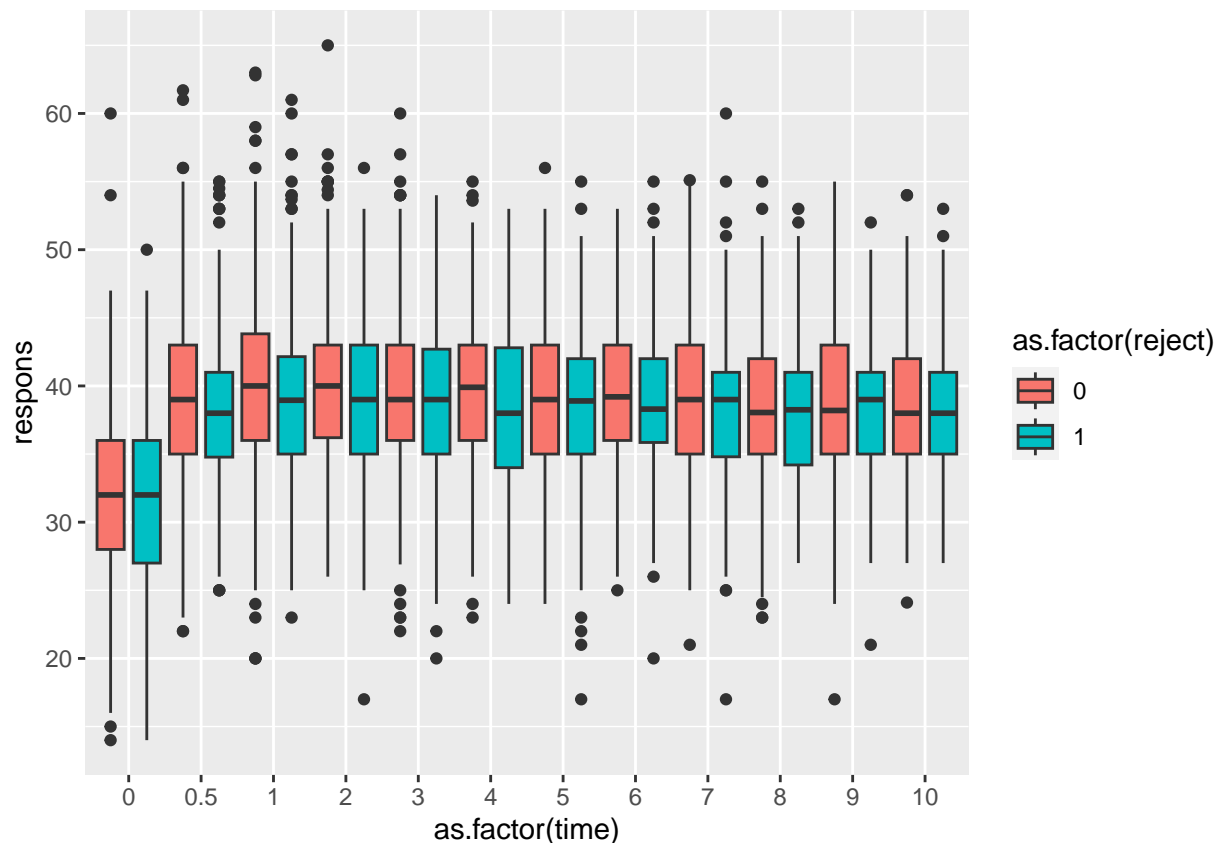


Boxplot

```
# Box plot by cardio
ggplot(data.long.noNA,aes(x=as.factor(time),y=respons,fill=as.factor(cardio)))+
  geom_boxplot(position=position_dodge(1))
```



```
# Box plot by reject
ggplot(data.long.noNA,aes(x=as.factor(time),y=respons,fill=as.factor(reject)))+
  geom_boxplot(position=position_dodge(1))
```



#### Correlation plot of Hc levels in different time

```
HcCorr = trenal.wide[,c(1:12)]
cor(HcCorr,use="complete.obs" ) # also COV for covariance
```

	HC0	HC06	HC1	HC2	HC3	HC4	HC5
## HC0	1.00000000	0.2264123	0.1587116	0.1724777	0.2139805	0.1732267	0.1557624
## HC06	0.22641235	1.0000000	0.7562367	0.6233688	0.5520591	0.5278499	0.5143061
## HC1	0.15871158	0.7562367	1.0000000	0.7315995	0.6656006	0.6119867	0.5873331
## HC2	0.17247771	0.6233688	0.7315995	1.0000000	0.7284046	0.6382434	0.5996189
## HC3	0.21398049	0.5520591	0.6656006	0.7284046	1.0000000	0.7733522	0.7016965
## HC4	0.17322666	0.5278499	0.6119867	0.6382434	0.7733522	1.0000000	0.7888249
## HC5	0.15576243	0.5143061	0.5873331	0.5996189	0.7016965	0.7888249	1.0000000
## HC6	0.13620085	0.4569881	0.5004036	0.4869519	0.5786122	0.6814132	0.7592203
## HC7	0.10156045	0.3936597	0.4541699	0.4724703	0.5402798	0.6466212	0.7067887
## HC8	0.08419757	0.3687935	0.4454882	0.4244221	0.5030428	0.6040136	0.6080051
## HC9	0.08859254	0.3711560	0.4254622	0.3971477	0.4303661	0.5461579	0.5713338
## HC10	0.09718506	0.4210917	0.4301937	0.4647890	0.4972001	0.5629570	0.5800544
	HC6	HC7	HC8	HC9	HC10		
## HC0	0.1362008	0.1015604	0.08419757	0.08859254	0.09718506		
## HC06	0.4569881	0.3936597	0.36879347	0.37115604	0.42109175		
## HC1	0.5004036	0.4541699	0.44548815	0.42546216	0.43019368		
## HC2	0.4869519	0.4724703	0.42442213	0.39714773	0.46478897		
## HC3	0.5786122	0.5402798	0.50304282	0.43036614	0.49720006		
## HC4	0.6814132	0.6466212	0.60401365	0.54615793	0.56295695		
## HC5	0.7592203	0.7067887	0.60800514	0.57133378	0.58005440		
## HC6	1.0000000	0.7414970	0.67347761	0.62938253	0.60329422		

```

## HC7  0.7414970 1.0000000 0.71838142 0.63933448 0.65646214
## HC8  0.6734776 0.7183814 1.00000000 0.70316750 0.68501304
## HC9  0.6293825 0.6393345 0.70316750 1.00000000 0.74259683
## HC10 0.6032942 0.6564621 0.68501304 0.74259683 1.00000000

library("PerformanceAnalytics")

## Warning: package 'PerformanceAnalytics' was built under R version 4.2.3
## Loading required package: xts
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric
##
## ##### WARNING #####
## # We noticed you have dplyr installed. The dplyr lag() function breaks how      #
## # base R's lag() function is supposed to work, which breaks lag(my_xts).      #
## #                                                                              #
## # Calls to lag(my_xts) that you enter or source() into this session won't    #
## # work correctly.                                                            #
## #                                                                              #
## # All package code is unaffected because it is protected by the R namespace  #
## # mechanism.                                                                #
## #                                                                              #
## # Set `options(xts.warn_dplyr_breaks_lag = FALSE)` to suppress this warning.  #
## #                                                                              #
## # You can use stats::lag() to make sure you're not using dplyr::lag(), or you #
## # can add conflictRules('dplyr', exclude = 'lag') to your .Rprofile to stop  #
## # dplyr from breaking base R's lag() function.                            #
## ##### WARNING #####
##
## Attaching package: 'xts'
## The following objects are masked from 'package:dplyr':
##
##      first, last
##
## Attaching package: 'PerformanceAnalytics'
## The following object is masked from 'package:graphics':
##
##      legend
chart.Correlation(HcCorr,historgram=TRUE)

## Warning in par(usr): argument 1 does not name a graphical parameter
## Warning in par(usr): argument 1 does not name a graphical parameter
## Warning in par(usr): argument 1 does not name a graphical parameter

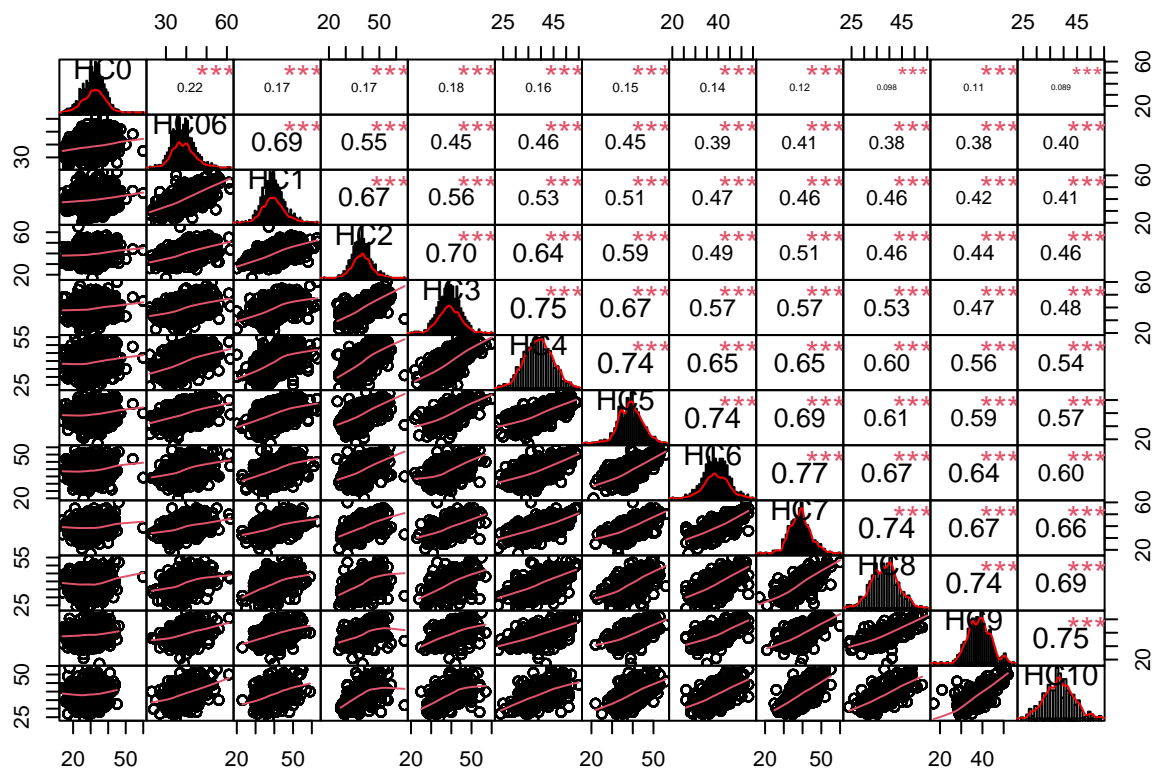
```

[illegible]





```
## Warning in par(usr): argument 1 does not name a graphical parameter
## Warning in par(usr): argument 1 does not name a graphical parameter
## Warning in par(usr): argument 1 does not name a graphical parameter
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## Warning in par(usr): argument 1 does not name a graphical parameter
```



Hypothesis based on the plot 1.Age 2.Male 3.Cardio 4.Reject

## Data set Trenal.XLS analysis with the linear mixed effects model

### The chosen of fixed effects variable

We can choose all the predictors as the fixed effect variables, plus an intercept ## The chosen of random effects variable

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>.

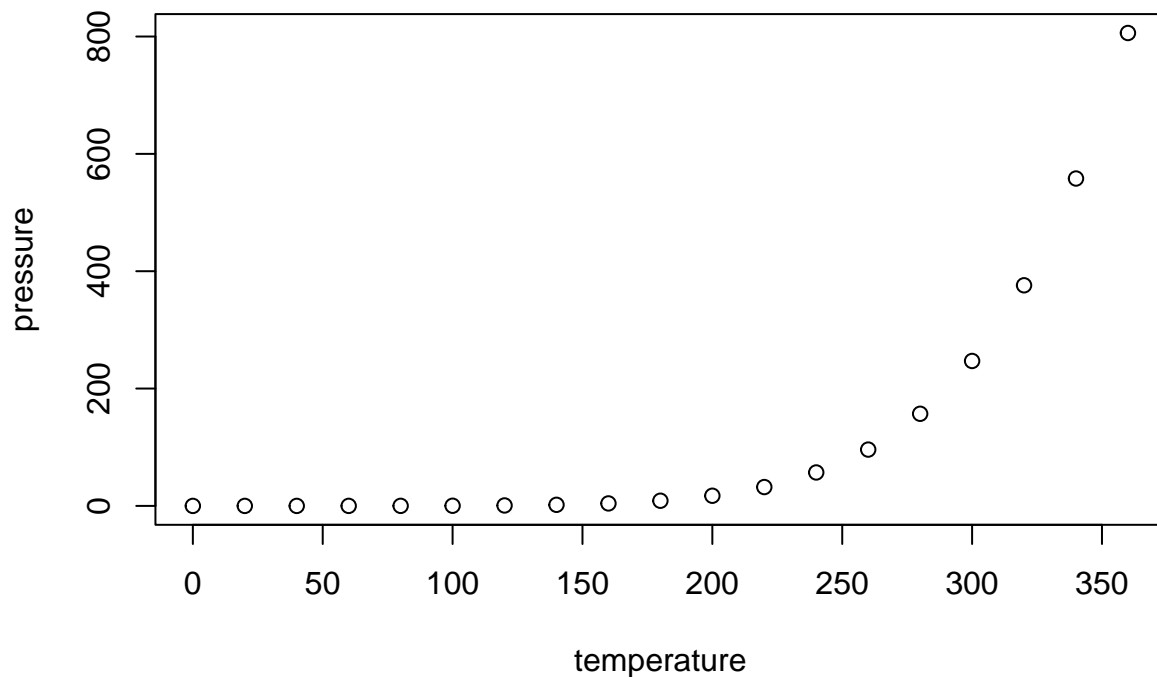
When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
summary(cars)
```

```
##      speed      dist
##  Min.   : 4.0    Min.   :  2.00
## 1st Qu.:12.0    1st Qu.: 26.00
##  Median:15.0    Median : 36.00
##   Mean :15.4    Mean   : 42.98
## 3rd Qu.:19.0    3rd Qu.: 56.00
##   Max. :25.0    Max.   :120.00
```

## Including Plots

You can also embed plots, for example:



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.