

Data Exploration of blood pressure in Rabbits

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Description:

The dataset is the study of five rabbits on two different occasions, after treatment with saline (control) and after treatment with the 5-HT₃ antagonist MDL 72222. Doses of phenylbiguanide were injected intravenously at 10 minute intervals in ascending order for each control and MDL. The change in blood pressure of the 5 rabbits were measured. Variables are described from the Source as

- **BPchange:** change in blood pressure.
- **Dose:** dose of Phenylbiguanide in micrograms.
- **Run:** label of run (“C1” to “C5”, then “M1” to “M5”).
- **Treatment:** placebo (control) or the 5-HT₃ antagonist MDL 72222 (MDL).
- **Animal:** label of animal used (“R1” to “R5”).

Goal:

The aim of this work is to know whether change in blood pressure (cardiogenic chemoreflex) triggered by phenylbiguanide doses is related to treatment with MDL. In other words, we want to know if BPchange of MDL is significantly different/higher than that of control among all animals(rabbits) for each of the 6 doses.

Data frame exploration

Setting up:

First, to load the Rabbit R package, run `install.packages("MASS")` in the console first.

```
require(plyr)
```

```
## Loading required package: plyr
```

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 3.3.2
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:plyr':
```

```
##
```

```
##      arrange, count, desc, failwith, id, mutate, rename, summarise,  
##      summarize
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.3.2
```

```
library(MASS)
```

```
## Warning: package 'MASS' was built under R version 3.3.2
```

```
##
```

```
## Attaching package: 'MASS'
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      select
```

Exploration of Rabbit data frames

Let's explore Rabbit with functions like `head`, `ncol`, `str`, `summary`.

```
head(Rabbit) #shows the first 6 rows of the data
```

```
##   BPchange   Dose Run Treatment Animal
## 1      0.5   6.25 C1   Control     R1
## 2      4.5  12.50 C1   Control     R1
## 3     10.0  25.00 C1   Control     R1
## 4     26.0  50.00 C1   Control     R1
## 5     37.0 100.00 C1   Control     R1
## 6     32.0 200.00 C1   Control     R1
```

```
str(Rabbit) # shows the dataset in a string form
```

```
## 'data.frame':   60 obs. of  5 variables:
## $ BPchange : num  0.5 4.5 10 26 37 32 1 1.25 4 12 ...
## $ Dose : num  6.25 12.5 25 50 100 200 6.25 12.5 25 50 ...
## $ Run : Factor w/ 10 levels "C1","C2","C3",...: 1 1 1 1 1 1 2 2 2 2 ...
## $ Treatment: Factor w/ 2 levels "Control","MDL": 1 1 1 1 1 1 1 1 1 1 ...
## $ Animal : Factor w/ 5 levels "R1","R2","R3",...: 1 1 1 1 1 1 2 2 2 2 ...
```

```
ncol(Rabbit) #shows the number of column of variables
```

```
## [1] 5
```

```
tail(Rabbit) #shows the last 6 rows of the data
```

```
##   BPchange   Dose Run Treatment Animal
## 55      2.4   6.25 M5         MDL     R5
## 56      2.5  12.50 M5         MDL     R5
## 57      1.5  25.00 M5         MDL     R5
## 58      2.0  50.00 M5         MDL     R5
## 59      9.0 100.00 M5         MDL     R5
## 60     19.0 200.00 M5         MDL     R5
```

```
dim(Rabbit) #shows the number of variables in the column and the total number of observations in the row
```

```
## [1] 60  5
```

```
names(Rabbit) #shows the name of the all variables
```

```
## [1] "BPchange" "Dose"      "Run"       "Treatment" "Animal"
```

```
summary(Rabbit) #summary statistics of the dataset
```

```
##      BPchange      Dose      Run      Treatment  Animal
##  Min.   : 0.50   Min.   : 6.25   C1      : 6   Control:30   R1:12
##  1st Qu.: 1.65   1st Qu.: 12.50   C2      : 6   MDL      :30   R2:12
##  Median : 4.75   Median : 37.50   C3      : 6                      R3:12
##  Mean   :11.22   Mean   : 65.62   C4      : 6                      R4:12
##  3rd Qu.:20.50   3rd Qu.:100.00   C5      : 6                      R5:12
##  Max.   :37.00   Max.   :200.00   M1      : 6
##                                     (Other):24
```

Extracting columns/“variables”

Let's extract a column with `$`. This first column contains the change in blood pressure of all animals

```
Rabbit$BPchange
```

```
## [1] 0.50 4.50 10.00 26.00 37.00 32.00 1.00 1.25 4.00 12.00 27.00
## [12] 29.00 0.75 3.00 3.00 14.00 22.00 24.00 1.25 1.50 6.00 19.00
## [23] 33.00 33.00 1.50 1.50 5.00 16.00 20.00 18.00 1.25 0.75 4.00
## [34] 9.00 25.00 37.00 1.40 1.70 1.00 2.00 15.00 28.00 0.75 2.30
## [45] 3.00 5.00 26.00 25.00 2.60 1.20 2.00 3.00 11.00 22.00 2.40
## [56] 2.50 1.50 2.00 9.00 19.00
```

```
min(Rabbit$BPchange)
```

```
## [1] 0.5
```

```
max(Rabbit$BPchange)
```

```
## [1] 37
```

```
Rabbit$Dose
```

```
## [1] 6.25 12.50 25.00 50.00 100.00 200.00 6.25 12.50 25.00 50.00
## [11] 100.00 200.00 6.25 12.50 25.00 50.00 100.00 200.00 6.25 12.50
## [21] 25.00 50.00 100.00 200.00 6.25 12.50 25.00 50.00 100.00 200.00
## [31] 6.25 12.50 25.00 50.00 100.00 200.00 6.25 12.50 25.00 50.00
## [41] 100.00 200.00 6.25 12.50 25.00 50.00 100.00 200.00 6.25 12.50
## [51] 25.00 50.00 100.00 200.00 6.25 12.50 25.00 50.00 100.00 200.00
```

```
min(Rabbit$Dose)
```

```
## [1] 6.25
```

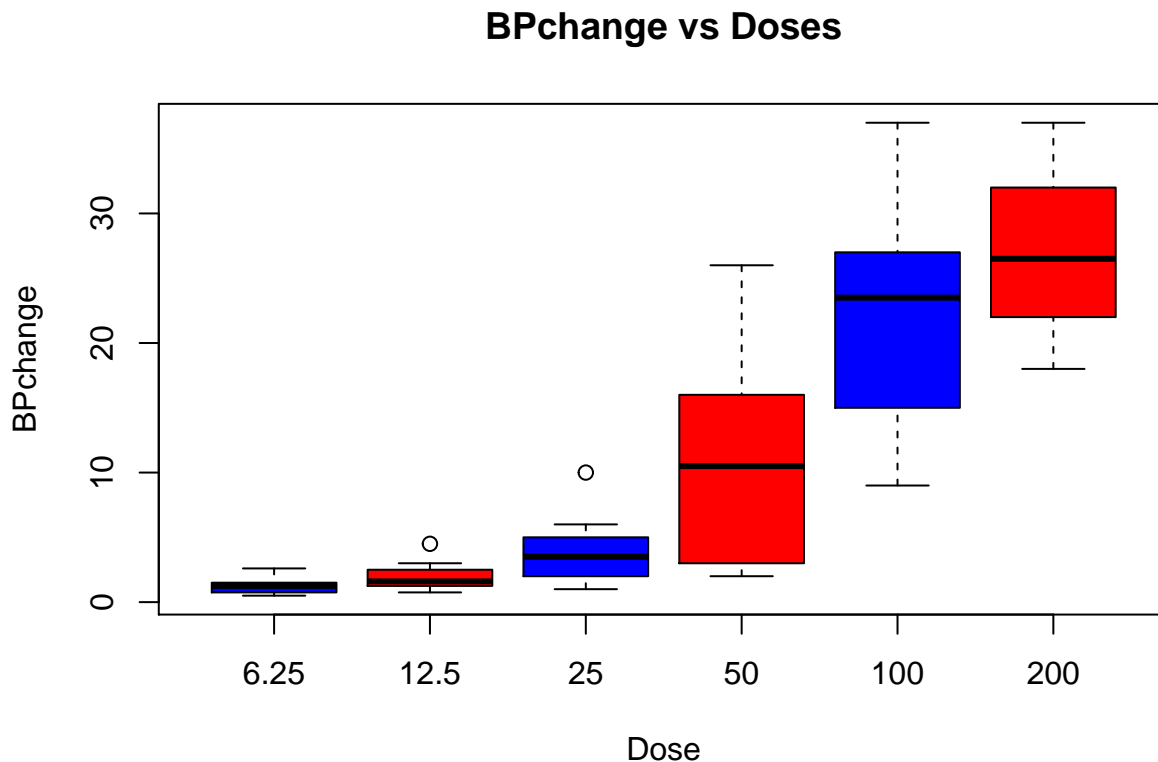
```
max(Rabbit$Dose)
```

```
## [1] 200
```

Change in blood pressure with the dose use

There does seem to be a relationship between change in blood pressure and dose based on the side-by-side boxplot, i.e increasing the doses cause the change in blood pressure to increase. The higher the dose, the more the BPchange.

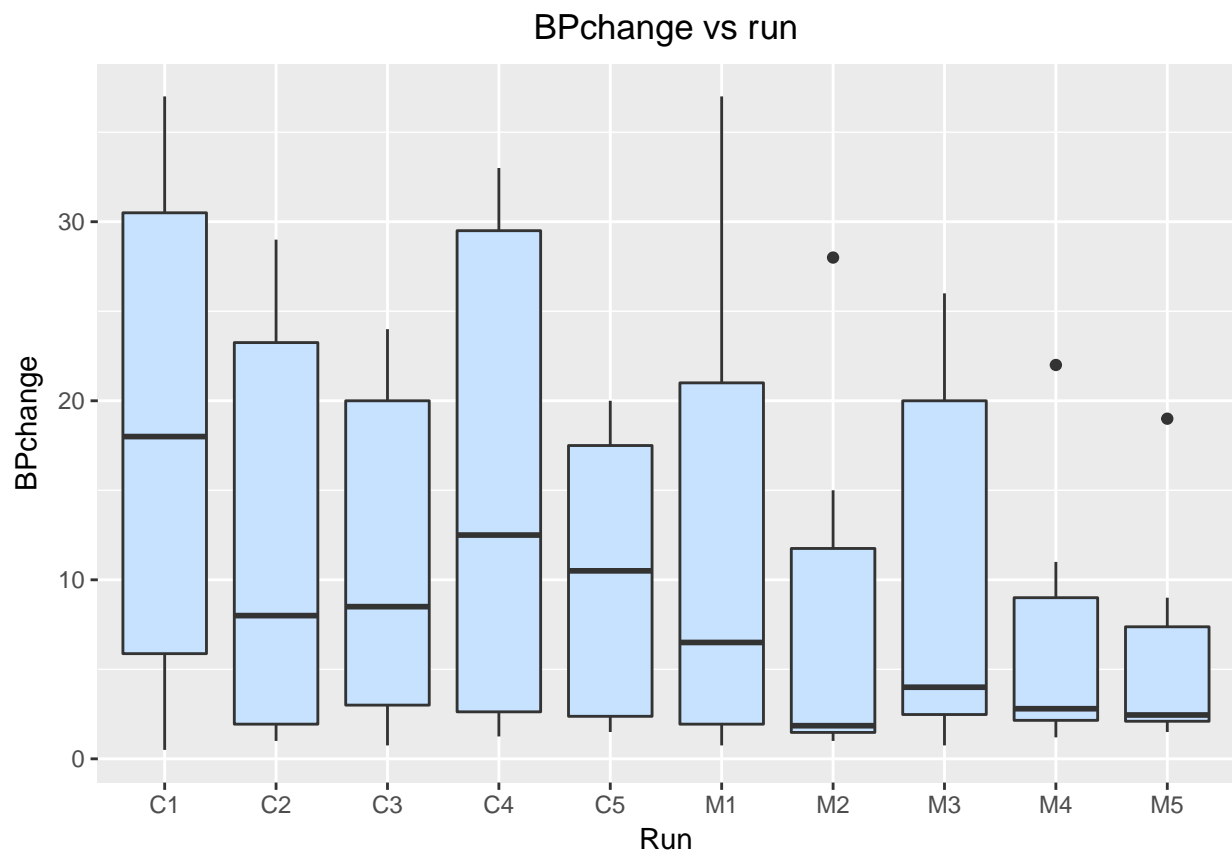
```
attach(Rabbit)
par(mfrow=c(1,1))
boxplot(BPchange~Dose, data=Rabbit, col=(c("blue","red")),main="BPchange vs Doses",ylab="BPchange", xlab="Dose")
```



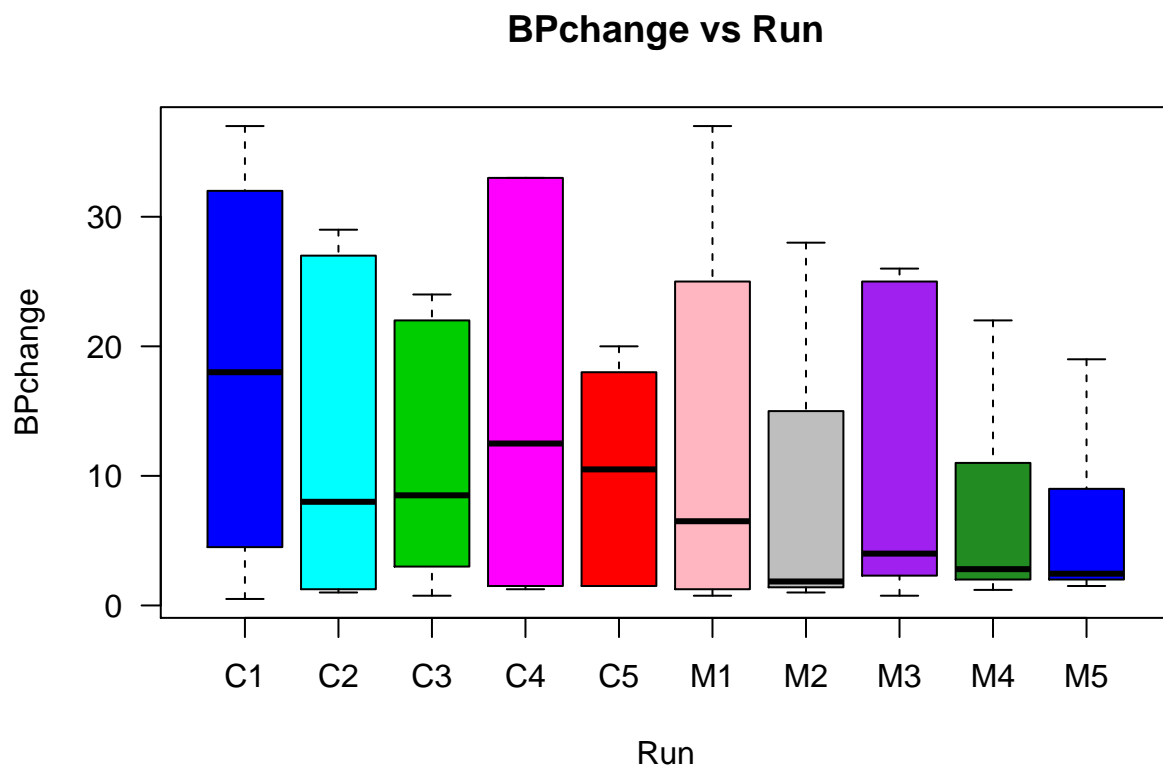
Change in BPchange versus Run

There seem to be a relationship between change in blood pressure and label of run.

```
ggplot(aes(x = Run, y = BPchange), data =Rabbit) +
geom_boxplot(fill = "slategray1") +
ggtitle("BPchange vs run")+theme(plot.title = element_text(hjust = 0.5))
```



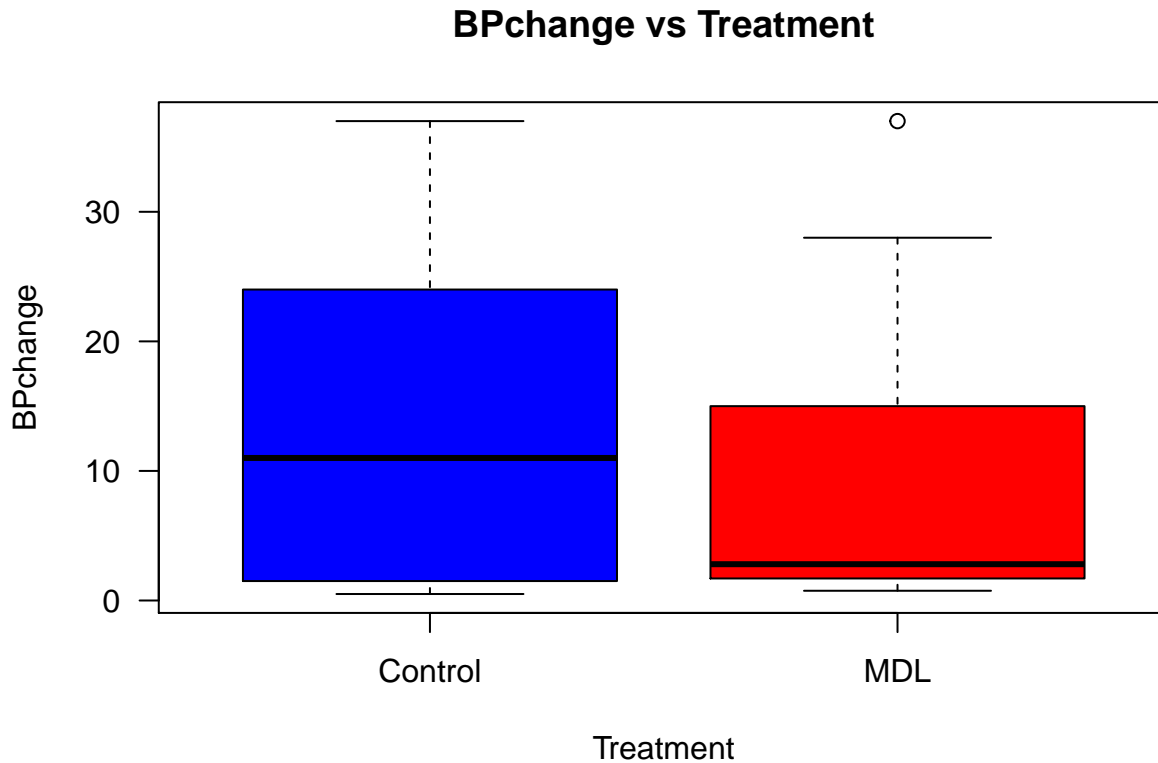
```
boxplot(BPchange~Run, data=Rabbit, ylab="BPchange", main="BPchange vs Run", las=1, col=c(4,5,3,6,2,"lightblue"), las=1)
```



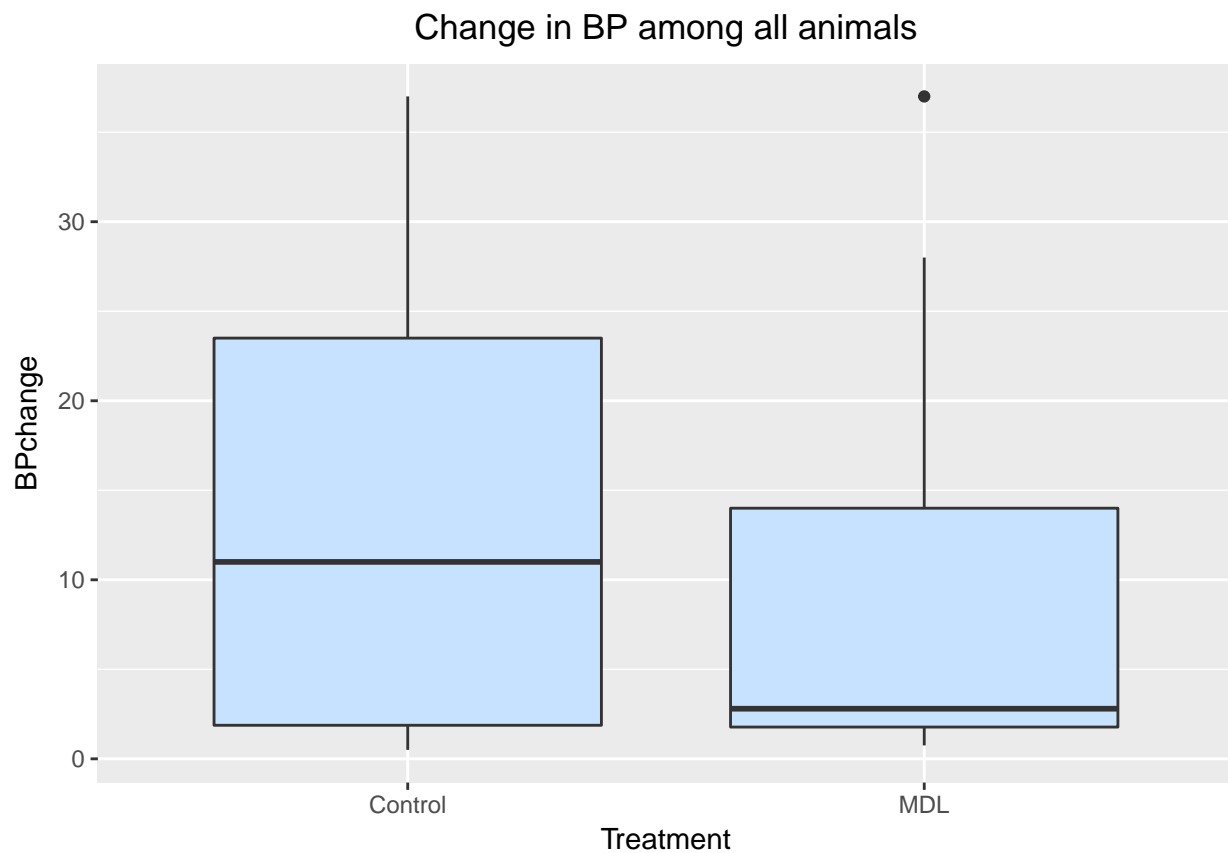
Change in BPchange versus Treatment

Control group seems to have a higher change in blood pressure and we need to know the baseline values of their respective blood pressures to know if MDL increases BP.

```
boxplot(BPchange~Treatment, data=Rabbit, ylab="BPchange", names=c("Control","MDL"), main="BPchange vs T
```

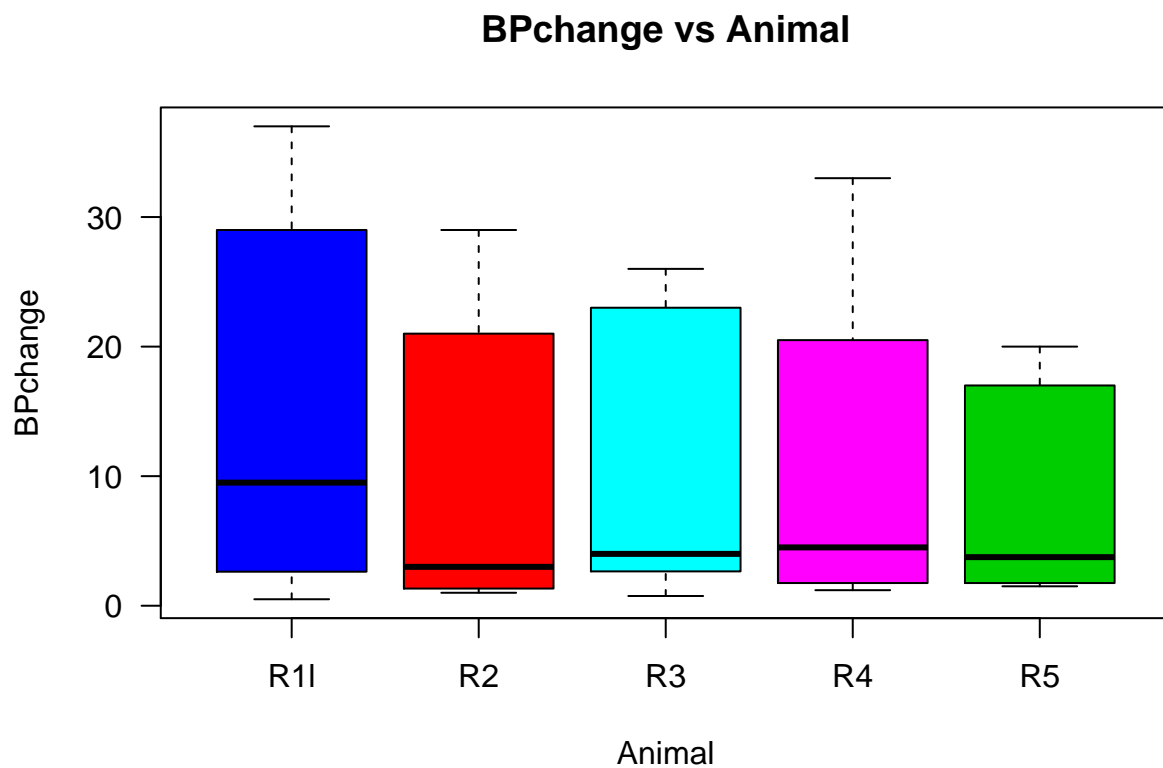


```
ggplot(aes(x = Treatment, y = BPchange), data =Rabbit) +  
geom_boxplot(fill = "slategray1") +  
ggtitle("Change in BP among all animals")+theme(plot.title = element_text(hjust = 0.5))
```

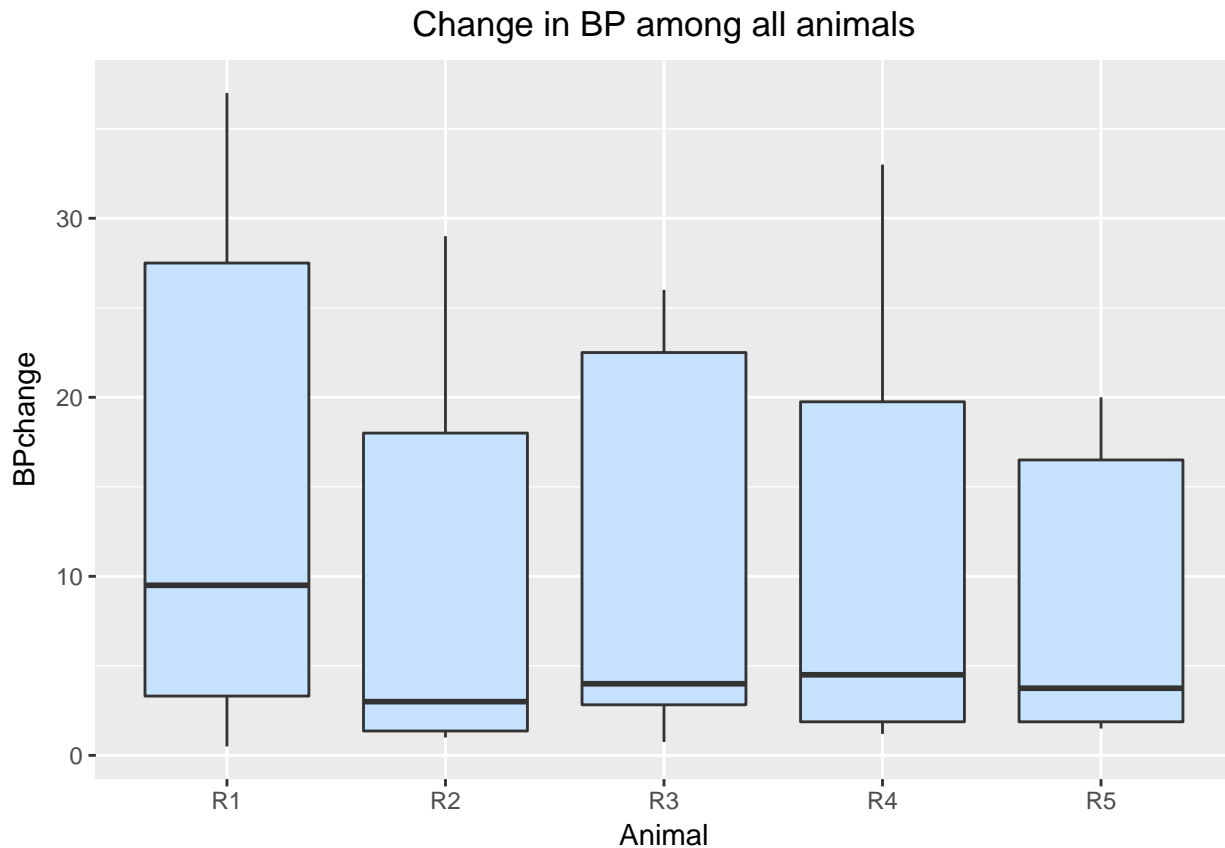


Change in BPchange versus Animal In general, R_1 have the highest BPchange with R_5 being the lowest.

```
boxplot(BPchange~Animal, data=Rabbit, ylab="BPchange", names=c("R1l","R2", "R3","R4", "R5"), main="BPch
```




```
ggplot(aes(x = Animal, y = BPchange), data =Rabbit) +
  geom_boxplot(fill = "slategray1") +
  ggtitle("Change in BP among all animals")+theme(plot.title = element_text(hjust = 0.5))
```



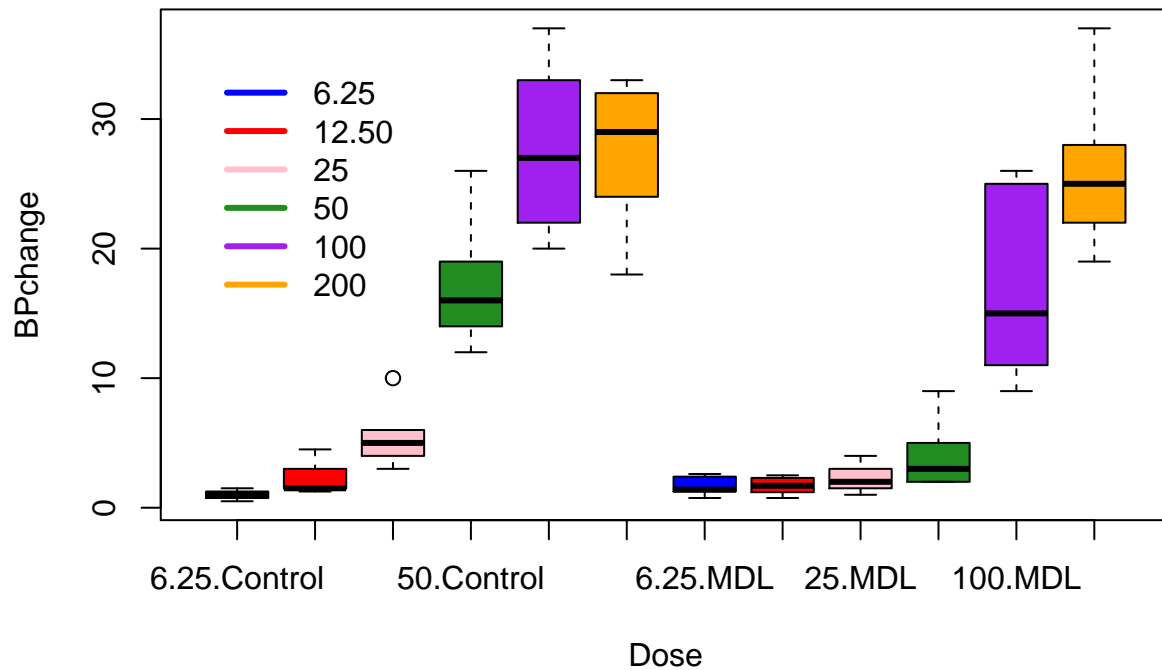
Change in blood pressure versus dose use, and stratified by treatment type (Control/MDL)

It is worth noting that stratifying the dose according to different treatment may give us a better interpretation of the relationship. Comparing the control group of 6 different doses on the left side and MDL group on the right side, we have;

- Increase in change in blood pressure as a result of ascending increase in doses.
- The control group may generally have a lower BPchange at baseline compared to the MDL group, which may mean different things:
 1. MDL increases blood pressure and say that the drug does not work.
 2. Animal on MDL may generally have a high blood pressure.
 3. There may be report bias.

```
boxplot(BPchange~Dose*Treatment, data=Rabbit, col=c("blue","red","pink","forestgreen","purple","orange"),
  legend(x=0.5, y=35, legend=c("6.25", "12.50", "25", "50", "100", "200"), col=c("blue","red","pink","fo
```

Change in blood pressure vs Dose and by treatment

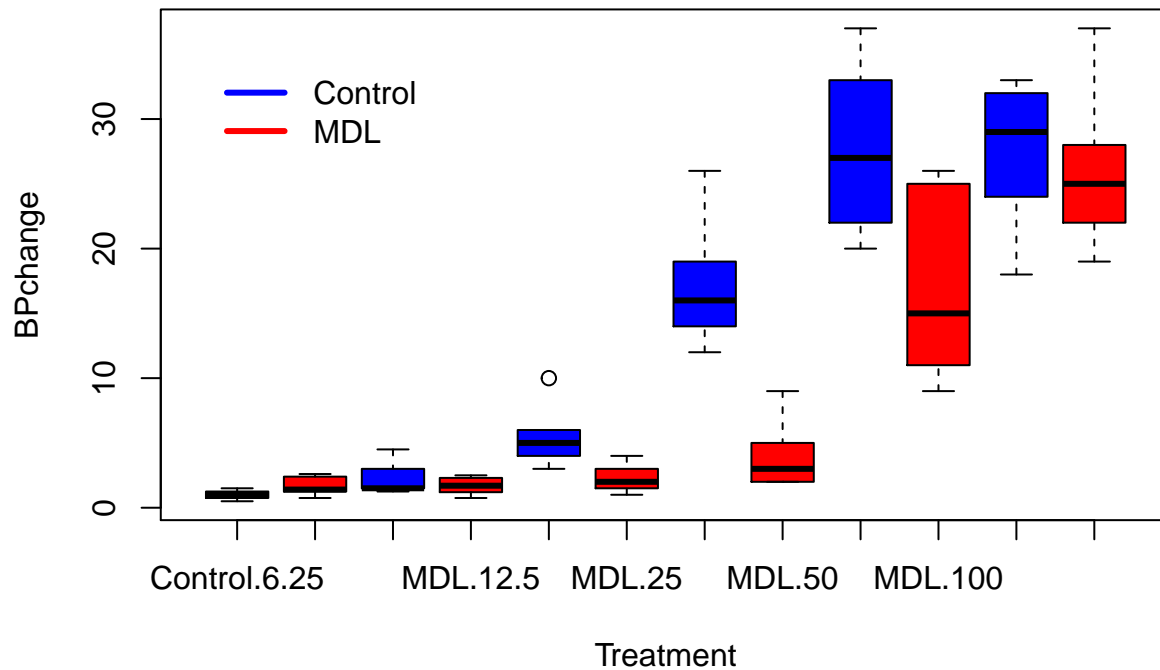


Change in blood pressure versus Treatment and by dose

This confirms that control group in colour blue (for dose = 12.5, 25, 50, 100, 200) have higher change in blood pressure compared to the MDL group in colour red. While reverse happens to be the case when dose= 6.25.

```
boxplot(BPchange~Treatment*Dose, data=Rabbit, col=c("blue","red")), main="Change in blood pressure vs '
legend(x=0.5, y=35, legend=c("Control", "MDL"), col=c("blue","red"), bty="n", lwd=3)
```

Change in blood pressure vs Treatment and by Dose

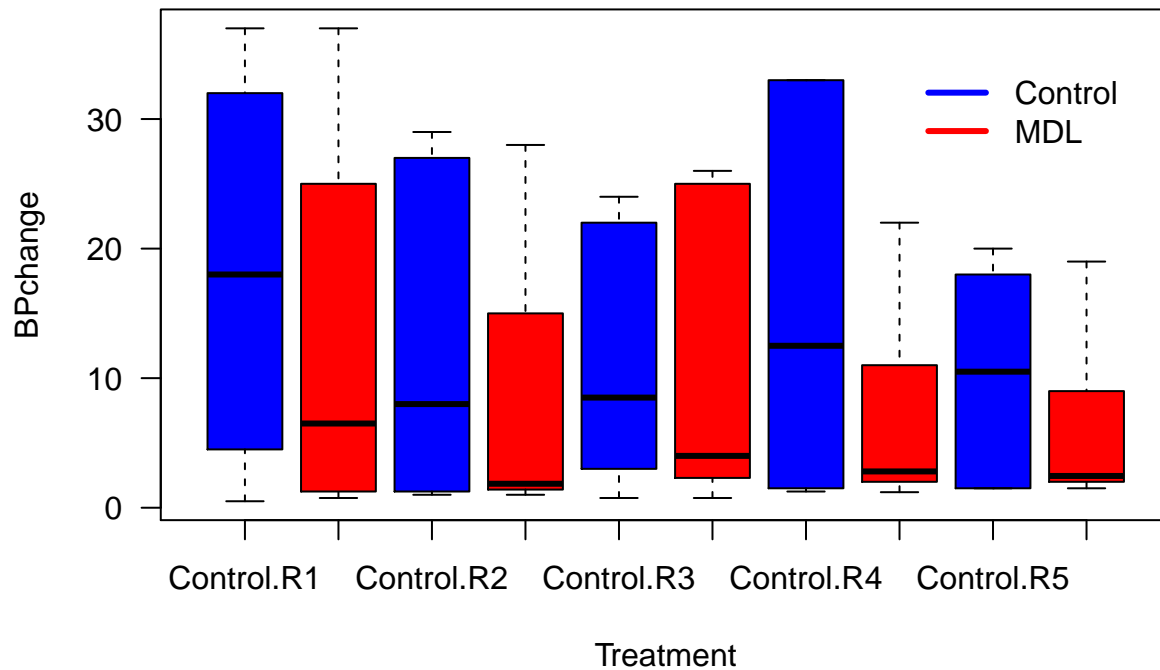


Relation between change in blood pressure and animal type

Animals in the control group generally have a higher change in blood pressure except for R_2 animals.

```
boxplot(BPchange~Treatment*Animal, ylab="BPchange", main="BPchange vs Treatment, stratified by Animal type",
legend(x=8, y=35, legend=c("Control", "MDL"), col=c(4,2), bty="n", lwd=3)
```

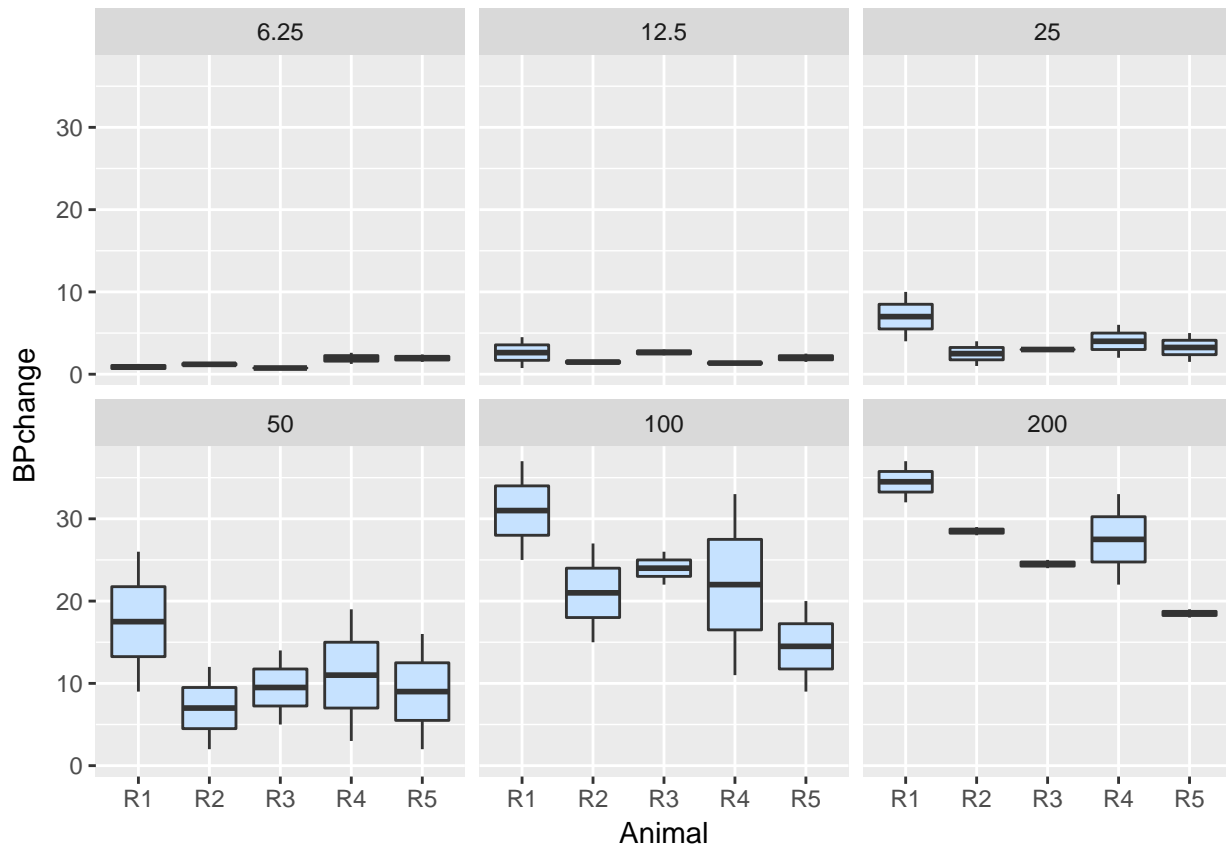
BPchange vs Treatment, stratified by Animal type



Relation between change in blood pressure and animal type, adjusting for dose

Animals on doses= 50, 100 experience a higher change in blood pressure.

```
ggplot(data = Rabbit, aes(x = Animal, y = BPchange)) +  
  geom_boxplot(fill = "slategray1") +  
  facet_wrap(~Dose)
```



Relation between change in blood pressure and dose use

There does seem to be a fairly strong linear relationship between BPchange and dose use based on this scatterplot and the pearson's correlation $\text{cor}(\text{BPchange}, \text{Dose}) = 0.831961$. The higher the dose, the larger the BPchange, i.e. Rabbits with higher doses have higher change in blood pressure, with dose= 100 giving the maximum effectiveness.

```
plot(Dose, BPchange, data=Rabbit, main="Scatter plot of BPchange vs Dose", xlab="Dose", ylab="BPchange")
```

```
## Warning in plot.window(...): "data" is not a graphical parameter
```

```
## Warning in plot.xy(xy, type, ...): "data" is not a graphical parameter
```

```
## Warning in axis(side = side, at = at, labels = labels, ...): "data" is not  
## a graphical parameter
```

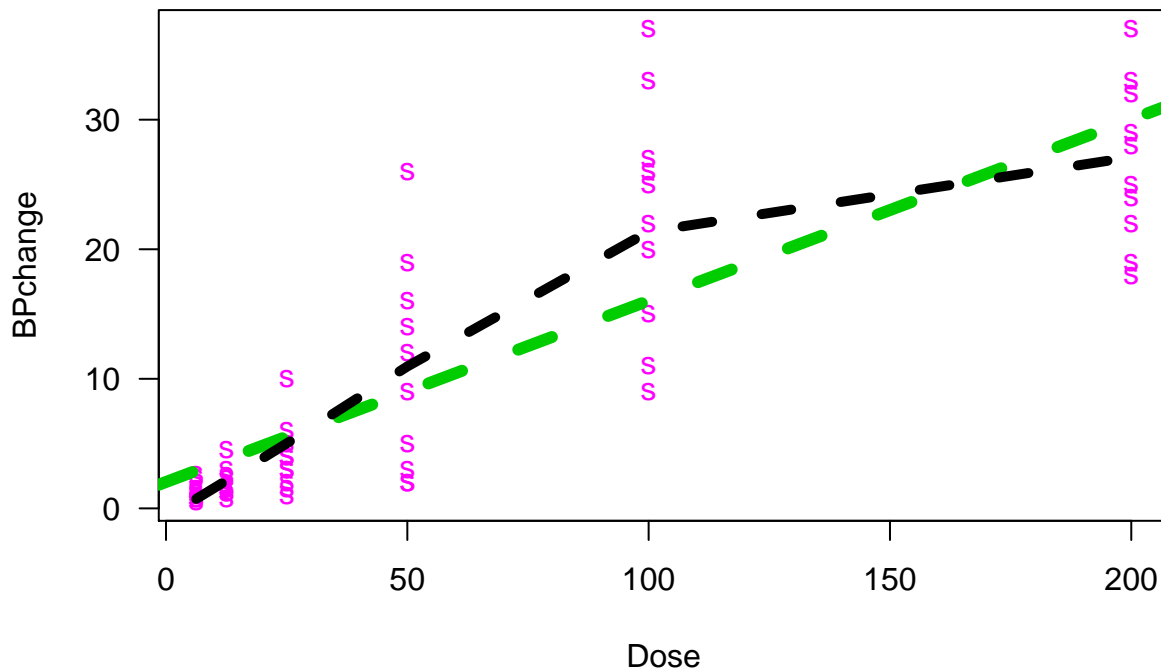
```
## Warning in axis(side = side, at = at, labels = labels, ...): "data" is not  
## a graphical parameter
```

```
## Warning in box(...): "data" is not a graphical parameter
```

```
## Warning in title(...): "data" is not a graphical parameter
```

```
abline(lm(BPchange~Dose), col=3, lty=2, lwd=6)
lines(smooth.spline(Dose,BPchange), lty=2, lwd=5)
```

Scatter plot of BPchange vs Dose



```
cor(BPchange,Dose)
```

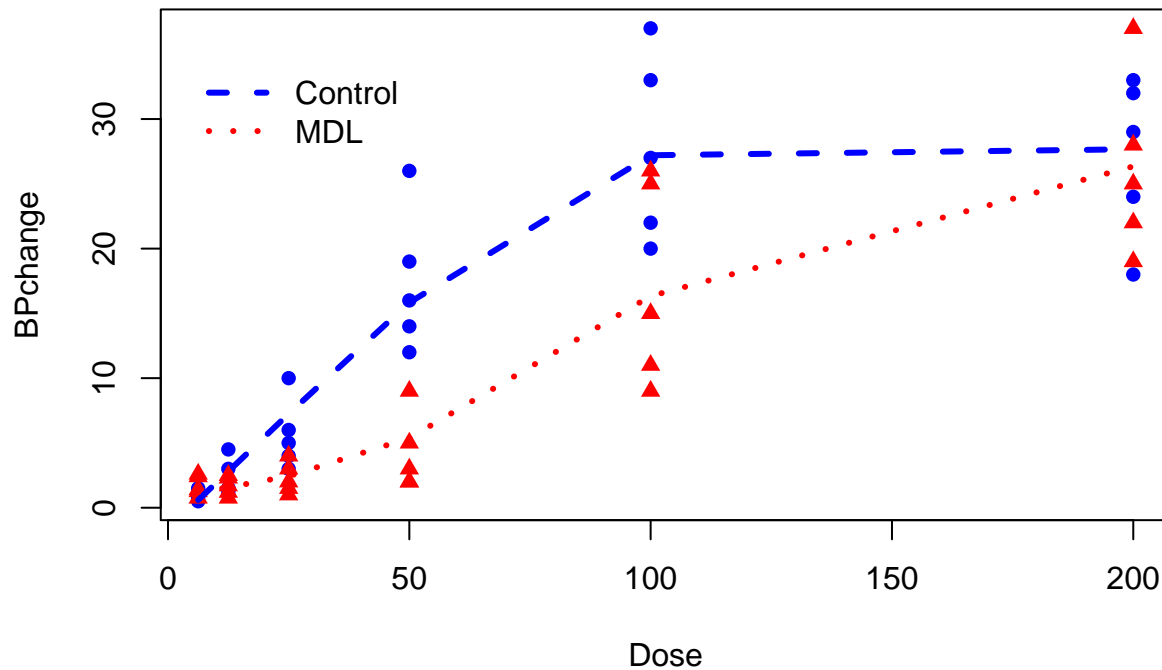
```
## [1] 0.831961
```

Scatter plot to show the change in blood pressure with dose use and by treatment type (Control/MDL)

There seems to be a wider variation in the change in blood pressure and especially in the middle. The control BPchange in blue seems higher than MDL in red. Then there seems to be no significant different in the change in blood pressure of the control group with doses 100 and 200.

```
plot(Dose[Treatment=="Control"], BPchange[Treatment=="Control"], col=4, xlab="Dose", ylab="BPchange", main="BPchange vs Dose by Treatment",
points(Dose[Treatment=="MDL"], BPchange[Treatment=="MDL"], col=2, pch=17)
lines(smooth.spline(Dose[Treatment=="Control"], BPchange[Treatment=="Control"]), col=4, lwd=3, lty=2)
lines(smooth.spline(Dose[Treatment=="MDL"], BPchange[Treatment=="MDL"]), col=2, lwd=3, lty=3)
legend(x=2.5, y=35, legend=c("Control", "MDL"), col=c(4,2), lty=c(2,3), bty="n", lwd=3)
```

Change in blood pressure vs Dose and by Treatment



Density plots of BPchange by Treatment

Dose= 6.25

Change in blood pressure with a higher density occurs among the control group.

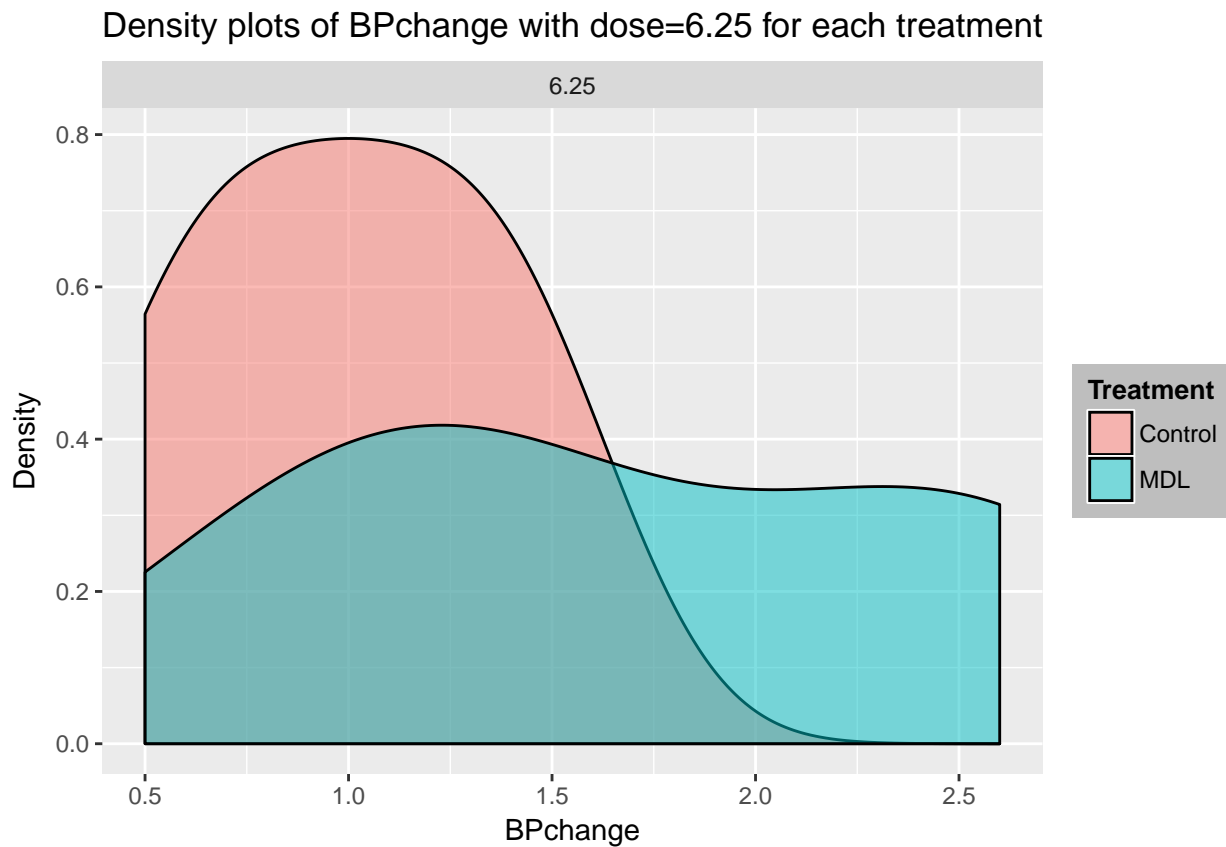
```
par(mfrow=c(3,2))
```

```
BP_6.25<- Rabbit %>%
  filter(Dose == 6.25)
```

```
## Warning: package 'bindrcpp' was built under R version 3.3.2
```

```
BP_6.25 %>%
```

```
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.3,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```

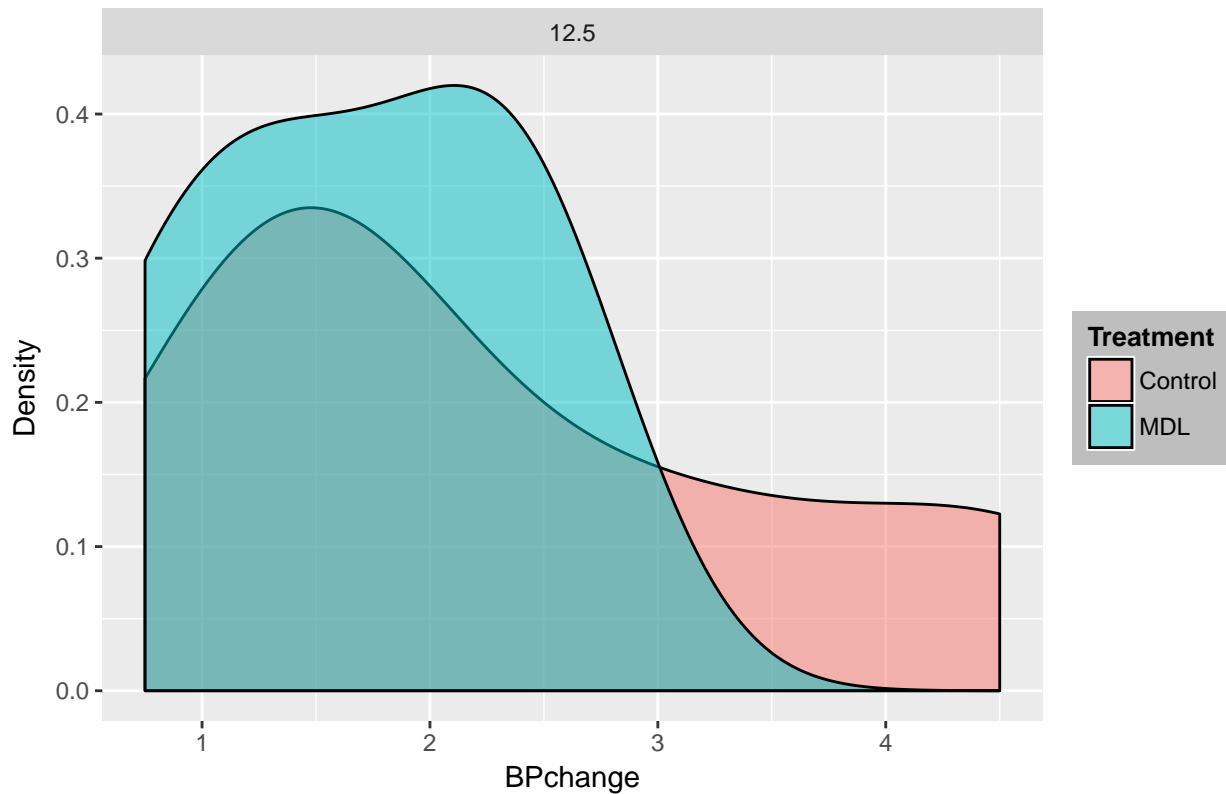


Dose= 12.50

Change in blood pressure with a higher density occurs among the MDL group.

```
BP_12.50<- Rabbit %>%
  filter(Dose == 12.50)
BP_12.50 %>%
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.3,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```

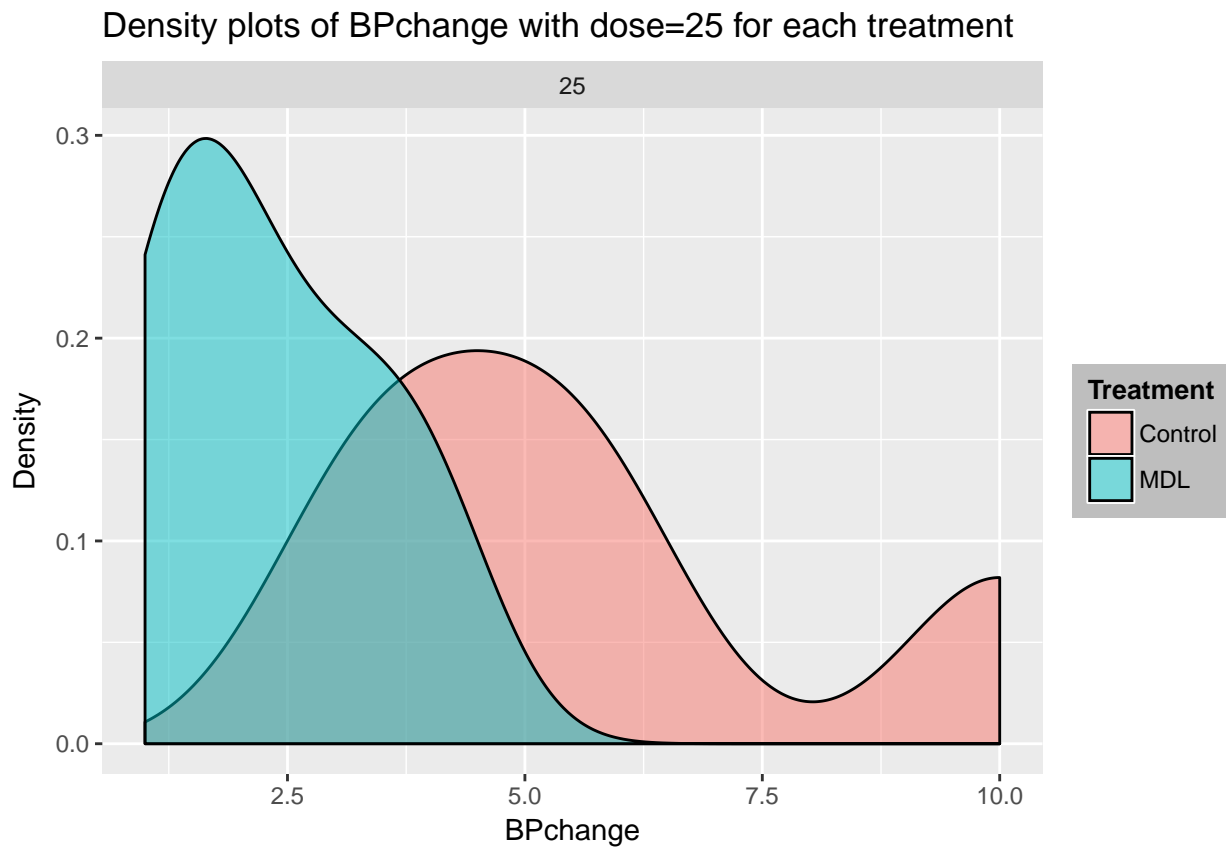

Density plots of BPchange with dose=12.50 for each treatment



Dose= 25

Change in blood pressure with a higher density occurs among the MDL group.

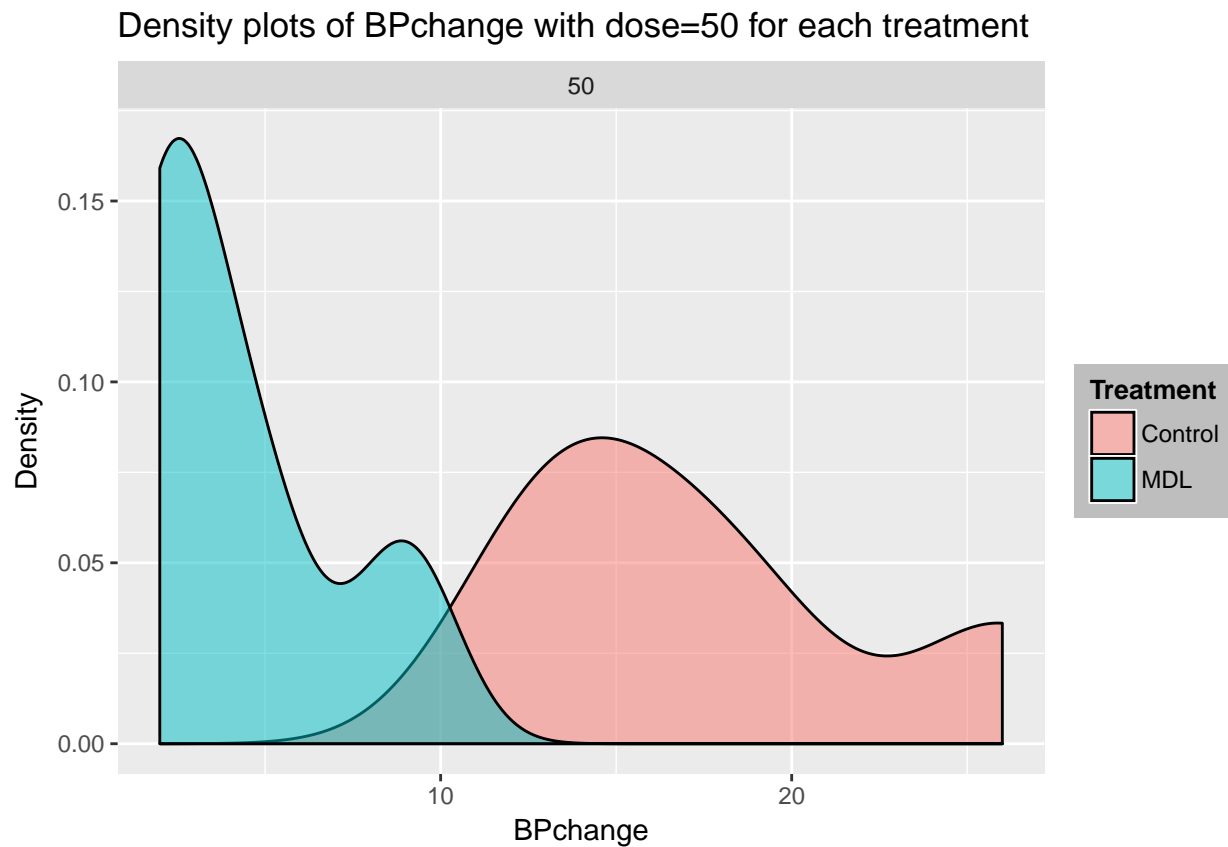
```
BP_25<- Rabbit %>%
  filter(Dose == 25)
BP_25 %>%
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.3,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```



Dose= 50

Change in blood pressure with a higher density occurs among the MDL group.

```
BP_50<- Rabbit %>%
  filter(Dose == 50)
BP_50 %>%
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.3,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```

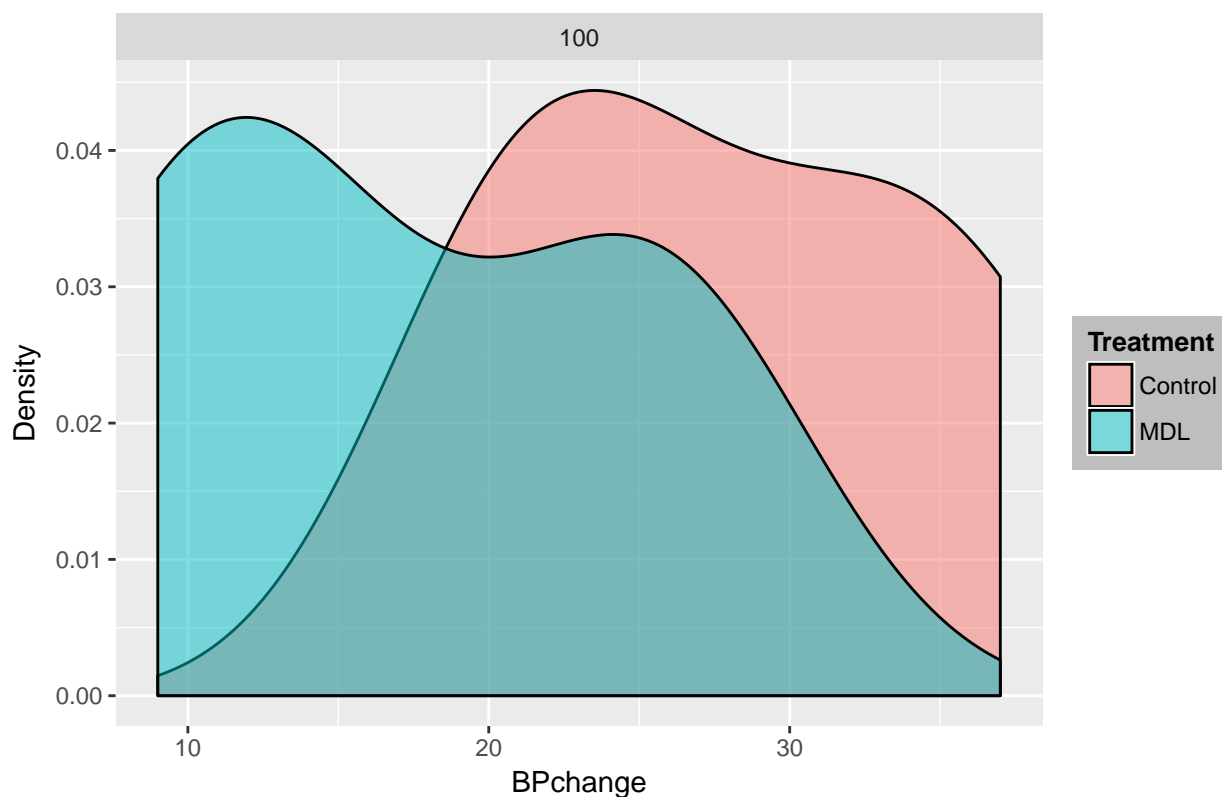


Dose= 100

There is no significantly difference in the density of the change in blood pressure among both group.

```
BP_100<- Rabbit %>%
  filter(Dose == 100)
BP_100 %>%
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.3,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```

Density plots of BPchange with dose=100 for each treatment

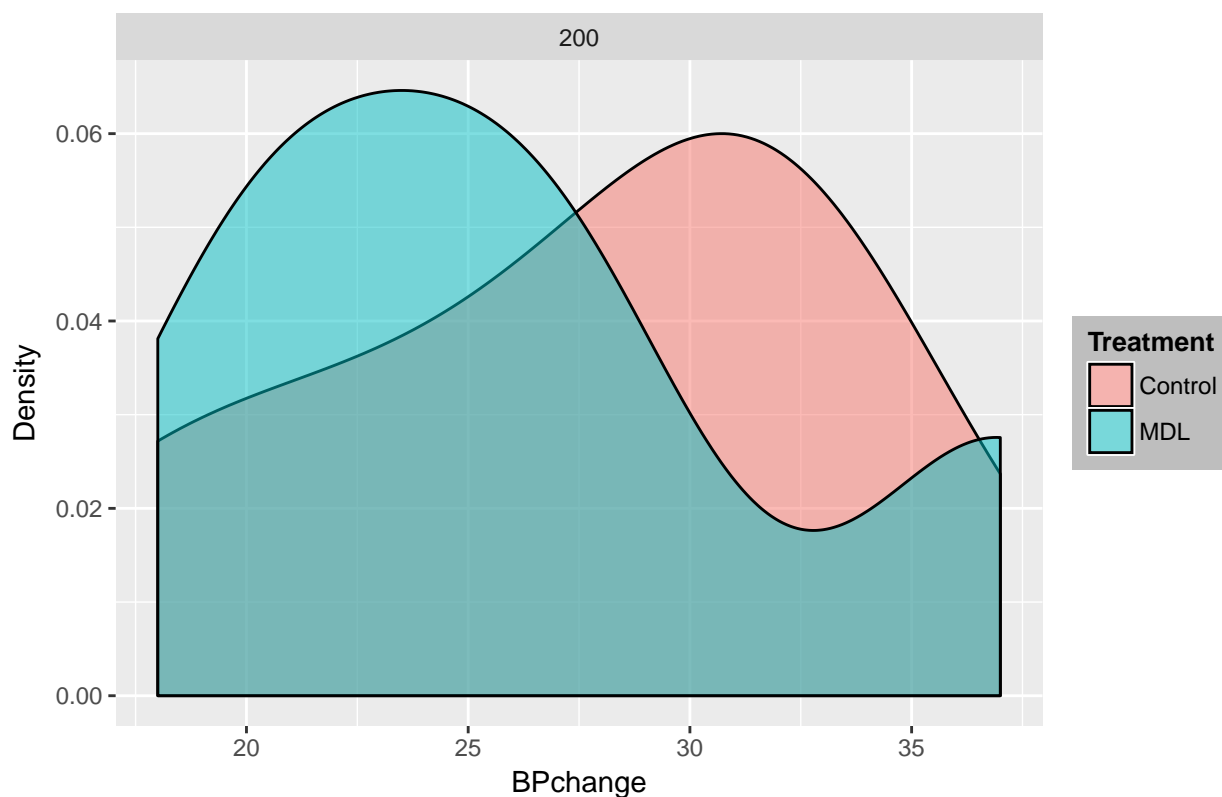


Dose= 200

MDL group have a higher density in change in blood pressure

```
BP_200<- Rabbit %>%
  filter(Dose == 200)
BP_200 %>%
  ggplot(aes(x = BPchange, fill = Treatment)) + facet_wrap(~Dose) + geom_density(alpha = 0.5) + ggtitle(
    size = 0.2,
    linetype = "dashed")) +
  labs(x="BPchange", y="Density")
```

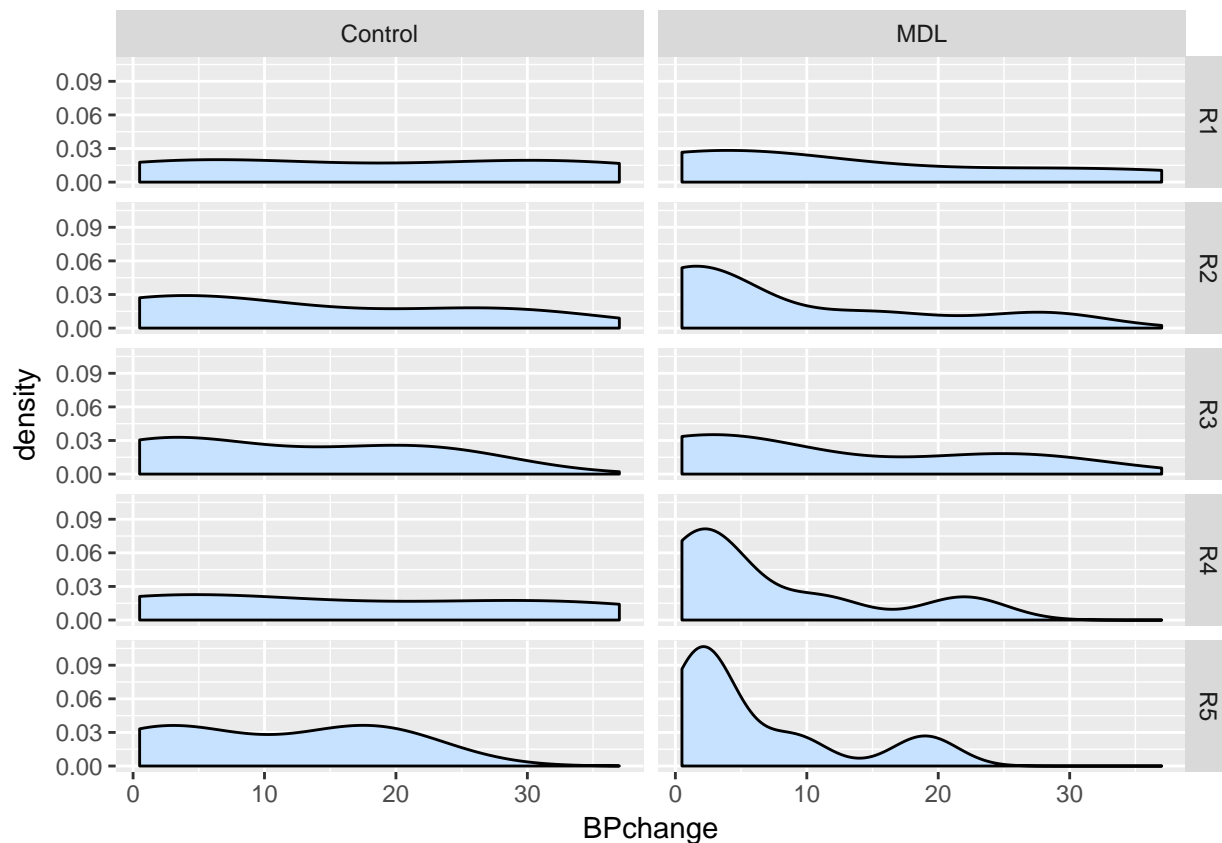
Density plots of BPchange with dose=200 for each treatment



Density plots of BPchange by Treatment

The density of animal with change in blood pressure is generally higher in MDL group relative to the control group.

```
Rabbit %>%  
  ggplot(aes(x = BPchange)) +  
    geom_density(fill = "slategray1") +  
    facet_grid(Animal ~ Treatment)
```

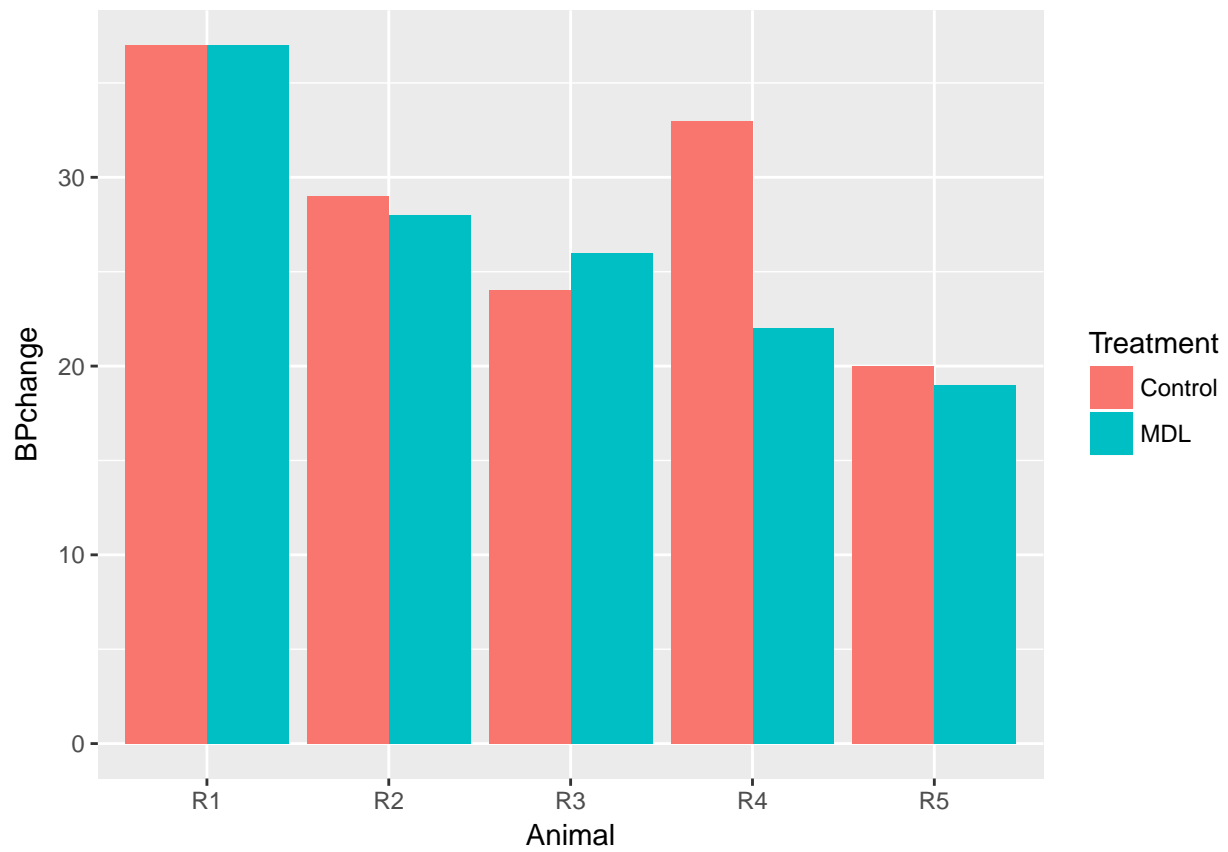


Barplot of change in blood pressure versus dose, adjusting for treatment type and animal

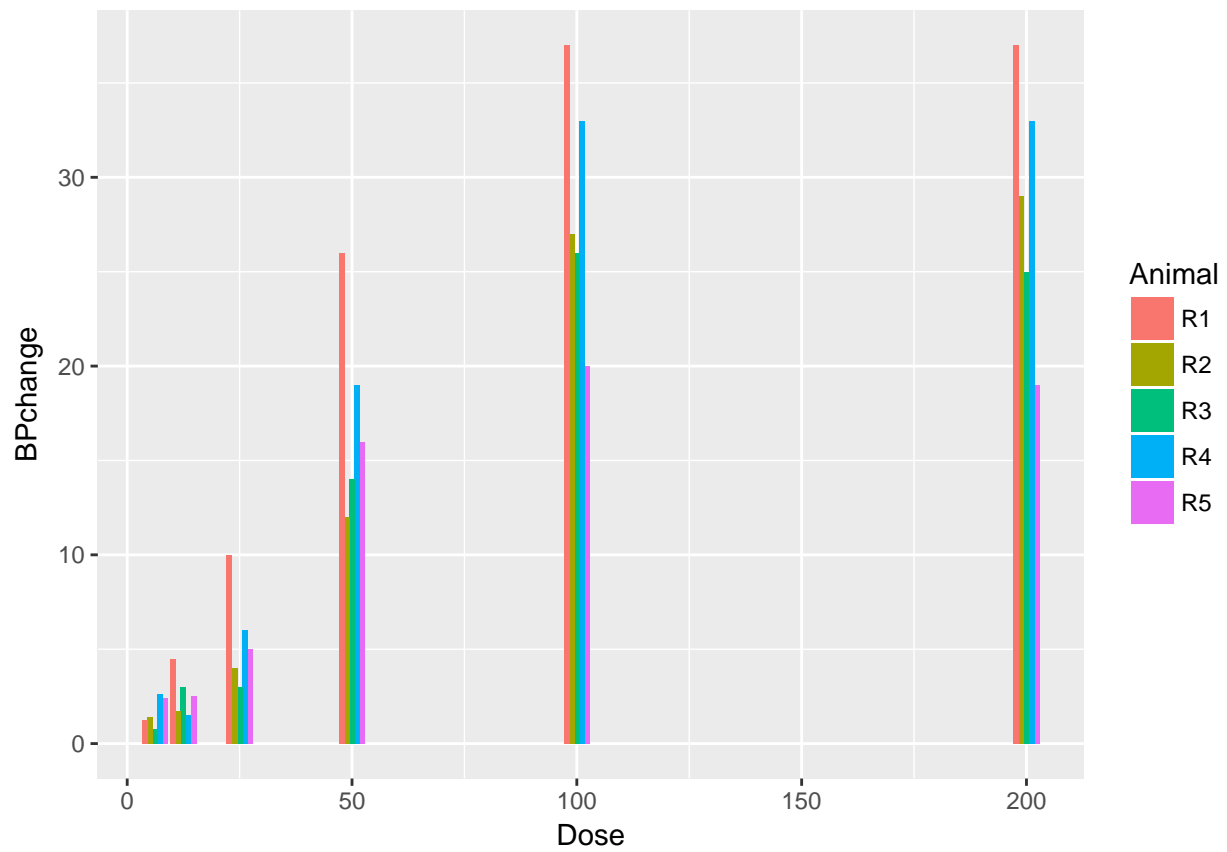
Among all animals, there seems to be a relationship between change in blood pressure and level of dose. BPchange increases with increase in dose among all rabbit for all treatment type. Animal R_1 have the highest BPchange and R_5 have the lowest. No significantly difference in the BPchange of animal R_3 when dose= 100, 200.

Also, since the p-value of the relationship between BPchange and treatment is $0.1151791 > 0.05$, we conclude that we have evidence to believe that there really is no significant difference in the change in blood pressure of the 5 groups.

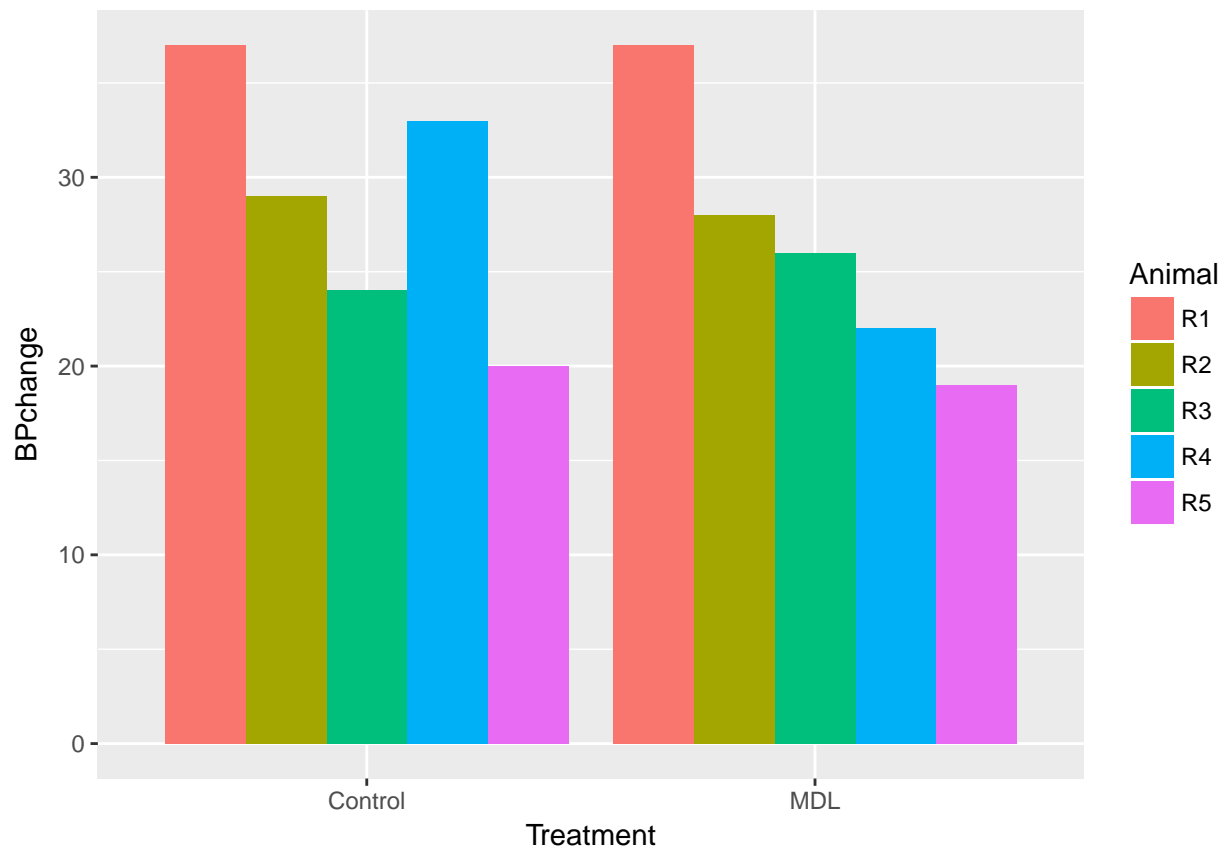
```
ggplot(Rabbit, aes(fill=Treatment, y=BPchange, x=Animal)) +
  geom_bar(position="dodge", stat="identity")
```



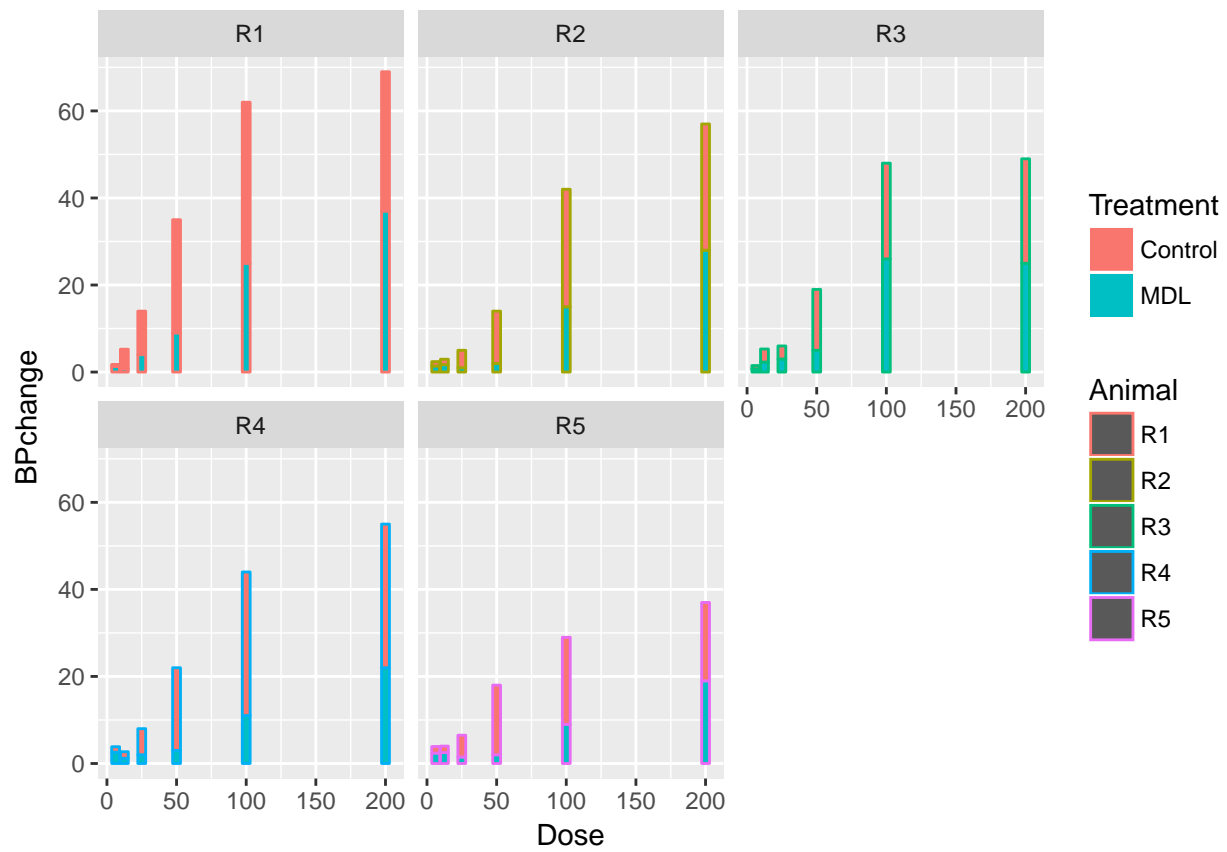
```
ggplot(Rabbit, aes(fill=Animal, y=BPchange, x=Dose)) +  
  geom_bar(position="dodge", stat="identity")
```



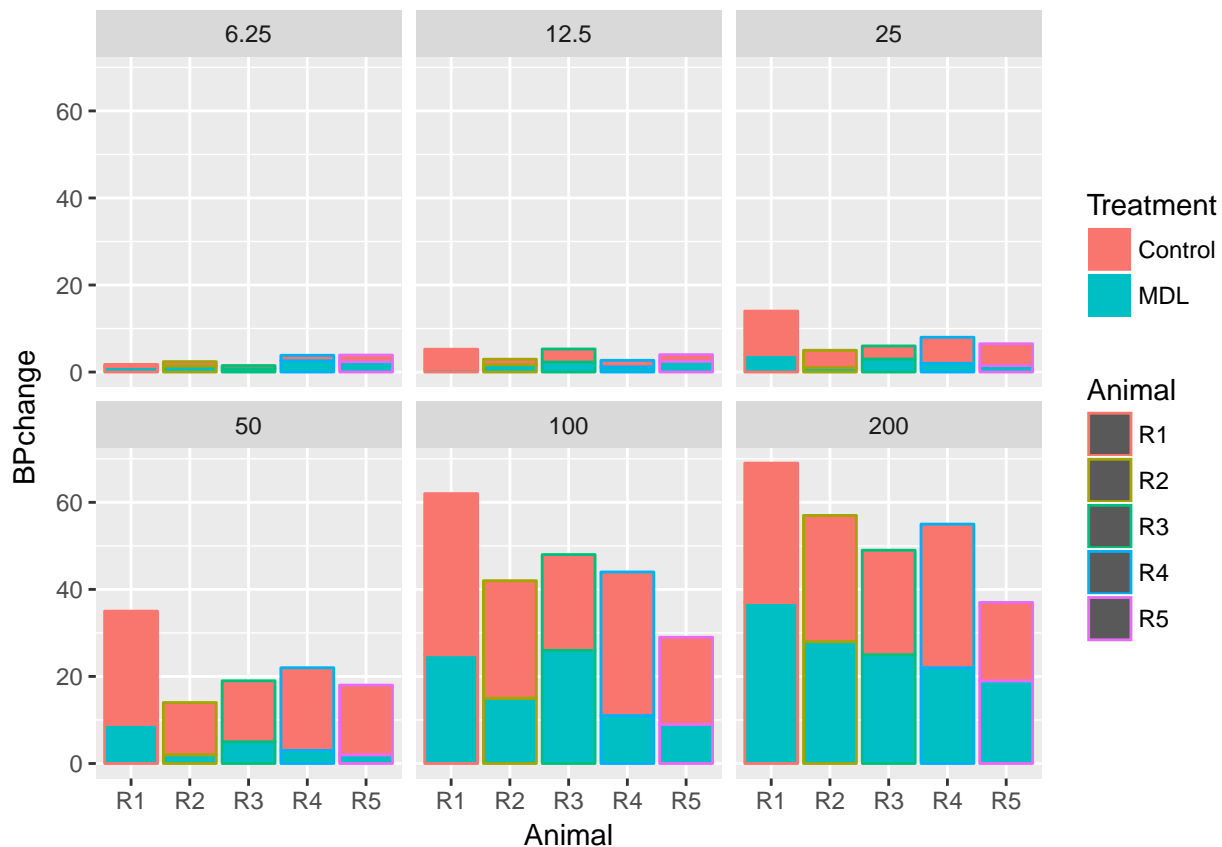
```
ggplot(Rabbit, aes(fill=Animal, y=BPchange, x=Treatment)) +  
  geom_bar(position="dodge", stat="identity")
```

```
ggplot(Rabbit, aes(y=BPchange, x=Dose, color=Animal, fill=Treatment)) +  
  geom_bar( stat="identity") +  
  facet_wrap(~Animal)
```



```
ggplot(Rabbit, aes(y=BPchange, x=Animal, color=Animal, fill=Treatment)) +
  geom_bar( stat="identity") +
  facet_wrap(~Dose)
```



```
t.test(Dose,BPchange, mu=0, alt="two.sided", paired=T, conf.level=0.95)
```

```
##
## Paired t-test
##
## data: Dose and BPchange
## t = 7.1378, df = 59, p-value = 1.581e-09
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 39.15438 69.65895
## sample estimates:
## mean of the differences
## 54.40667
```

```
Ttest=t.test(BPchange~Treatment)
Ttest
```

```
##
## Welch Two Sample t-test
##
## data: BPchange by Treatment
## t = 1.5999, df = 56.771, p-value = 0.1152
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.17821 10.53821
## sample estimates:
```

```
## mean in group Control      mean in group MDL
##           13.558333           8.878333
```

```
attributes(Ttest)
```

```
## $names
## [1] "statistic"  "parameter"  "p.value"    "conf.int"   "estimate"
## [6] "null.value" "alternative" "method"     "data.name"
##
## $class
## [1] "htest"
```

```
Ttest$conf.int
```

```
## [1] -1.17821 10.53821
## attr(,"conf.level")
## [1] 0.95
```

```
Ttest$p.value
```

```
## [1] 0.1151791
```