

takeover_time_analysis

Oct 23, 2019

Takeover time versus lead time

```
setwd("C://Doc//resume//apply//TOPS")
require(ggplot2)
```

```
## Loading required package: ggplot2
```

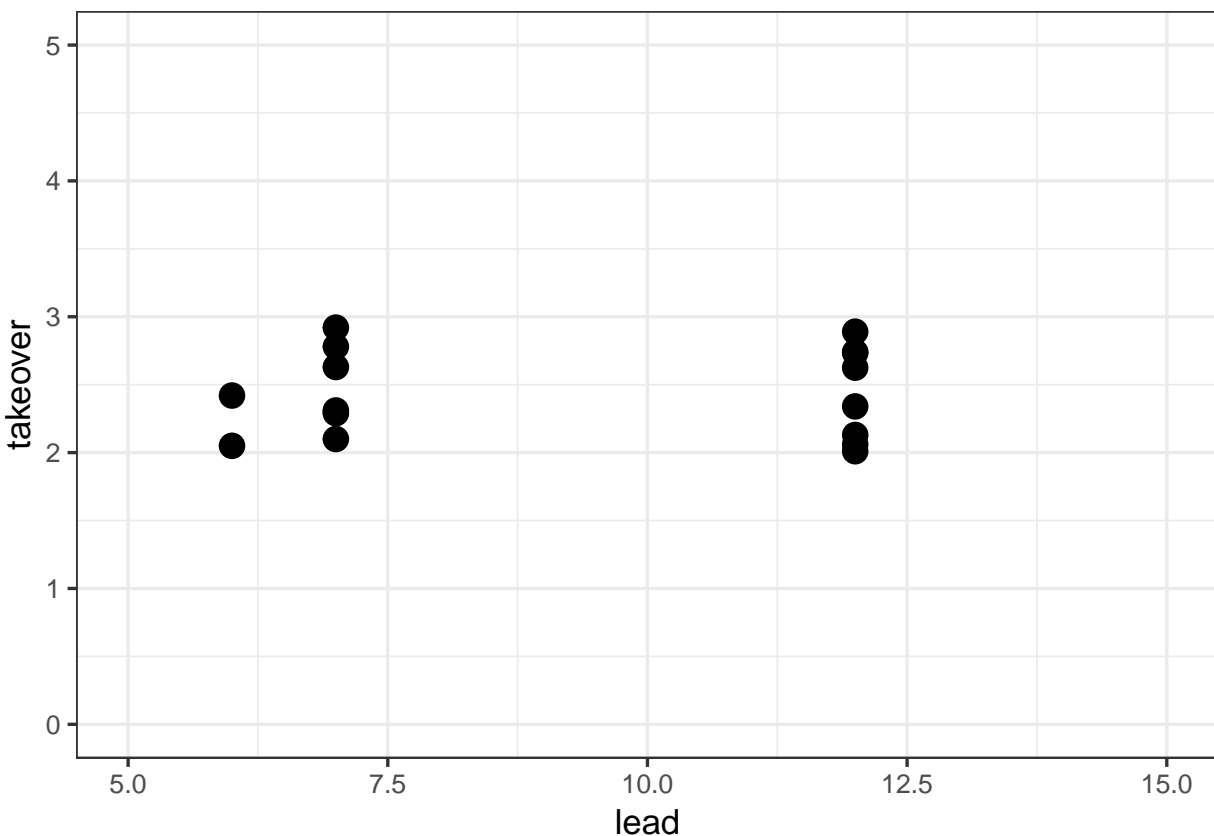
```
## Warning: package 'ggplot2' was built under R version 3.4.4
```

```
raw <- read.csv("takeover_time.csv", header = T)
raw=raw[-73,] #Remove outliers
```

Here 73rd sample point with 42s takeover time seems suspicious, thus it's removed.

```
takeover_time = sapply(1:nrow(raw), function(x)
  ifelse(is.na(raw$Takeover_Time[x]), mean(unlist(raw[x,c("Braking","Steering")])), na.rm = T), raw$Takeover_Time[x])
ggplot(data.frame(lead=raw$Lead_time, takeover=takeover_time), aes(x = lead, y = takeover))+geom_point()
```

```
## Warning: Removed 106 rows containing missing values (geom_point).
```

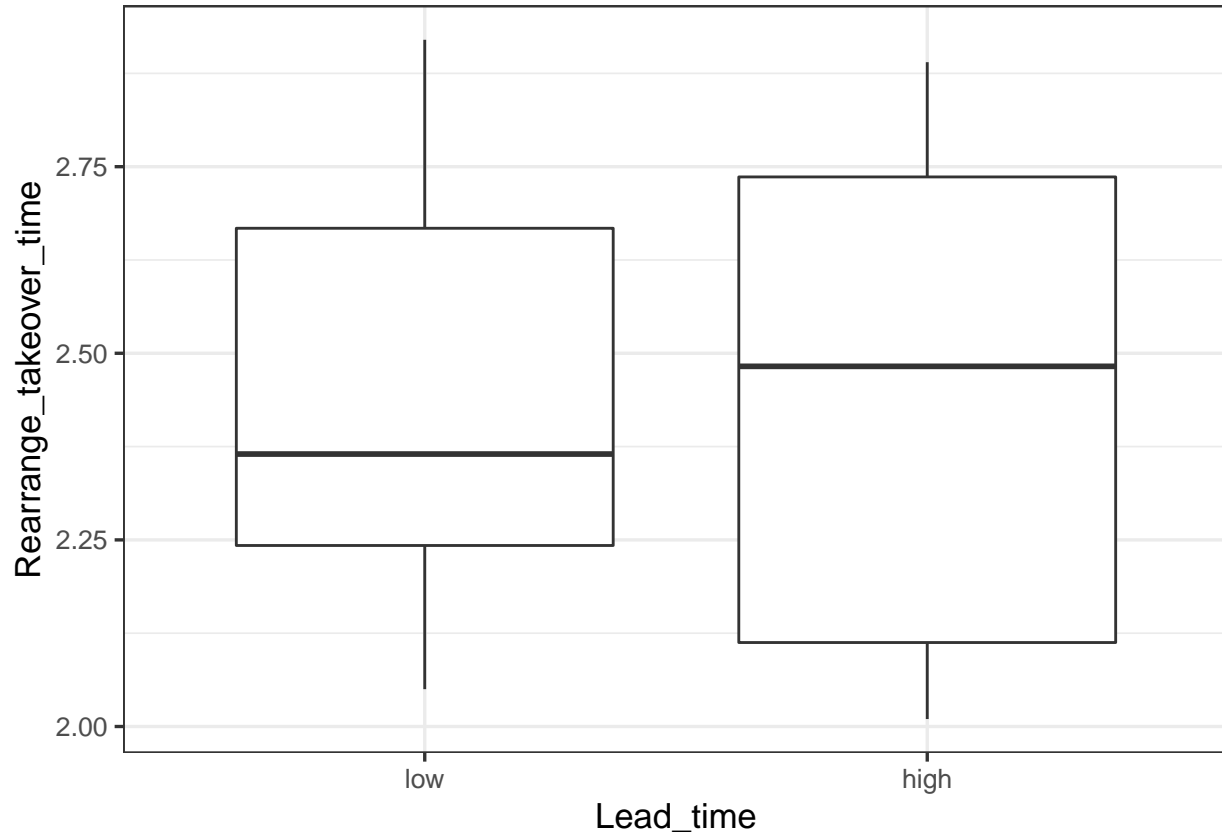


```
raw[, "Rearrange_takeover_time"] <- takeover_time
```

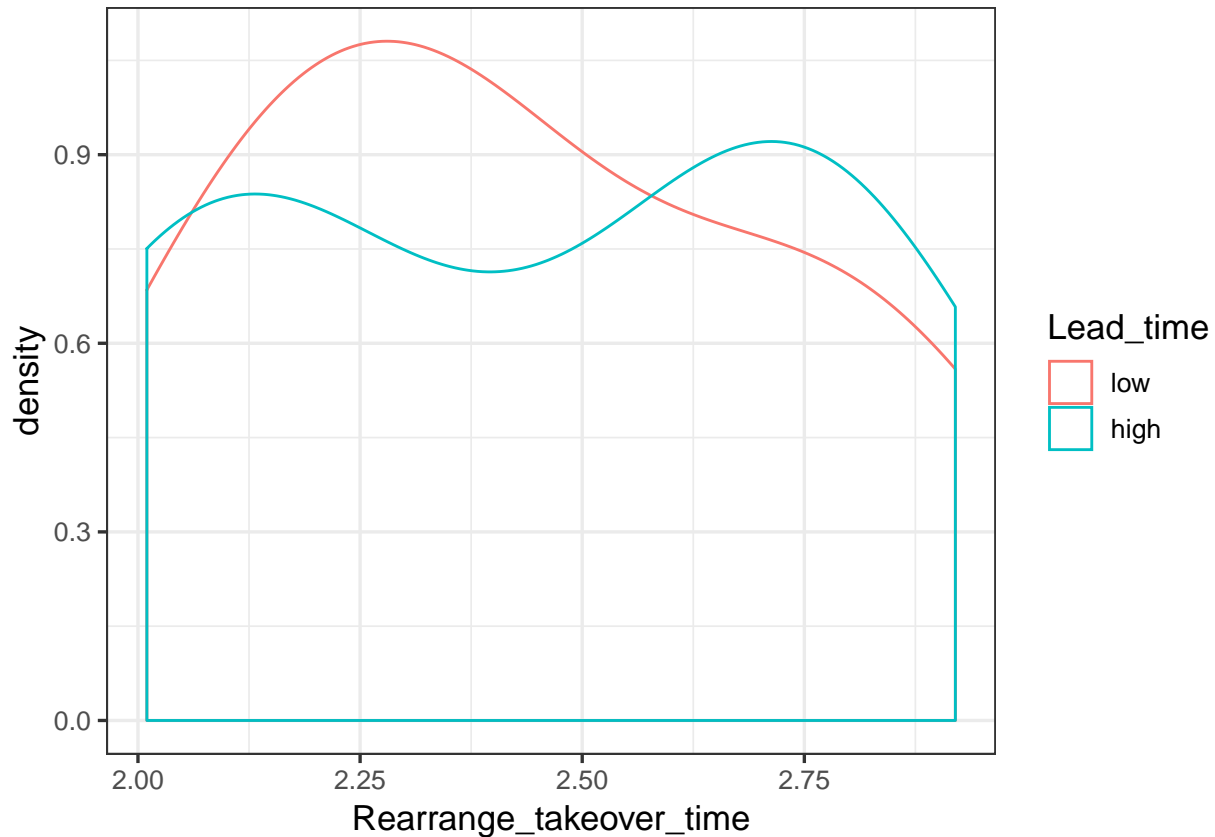
Here we recalculate the takeover time for more observations. If the takeover time is recorded, it won't be

changed. Otherwise, replace it by the average value of steering time and Braking time, ignoring the missing value. The scatter plot shows we still does not have enough observations for regression, especially the lead time. We only have 3 different lead time. Next we categorize the lead time 6 and 7 into low-level and 12 into high-level. Check the density plot first.

```
lead_vs_takeover <- raw[,c("Lead_time", "Rearrange_takeover_time")]
lead_vs_takeover <- lead_vs_takeover[apply(!is.na(lead_vs_takeover), 1, FUN = function(x) all(unlist(x))))
lead_vs_takeover$Lead_time[1:2] <- c(7,7)
lead_vs_takeover$Lead_time <- factor(lead_vs_takeover$Lead_time, labels=c("low", "high"))
ggplot(lead_vs_takeover, aes(x=Lead_time, y=Rearrange_takeover_time))+geom_boxplot()+theme_bw(base_size=
```



```
ggplot(lead_vs_takeover, aes(x=Rearrange_takeover_time, color=Lead_time))+geom_density()+theme_bw(base_s
```



obviously they have different peaks. It may support my assumption before. It's interesting that the box plot indicates that the high lead time group have greater deviation. I guess drivers have no choice but stopping the secondary task at once when the lead time is short. But drivers in the other group have more time to decide how to tackle with it. Consequently, they have various reaction. low level variance: 0.0985, high level variance: 0.121.

two sample t test

```
shapiro.test(lead_vs_takeover$Rearrange_takeover_time[lead_vs_takeover$Lead_time=="low"])

##
##  Shapiro-Wilk normality test
##
## data:  lead_vs_takeover$Rearrange_takeover_time[lead_vs_takeover$Lead_time ==      "low"]
## W = 0.94502, p-value = 0.661

shapiro.test(lead_vs_takeover$Rearrange_takeover_time[lead_vs_takeover$Lead_time=="high"])

##
##  Shapiro-Wilk normality test
##
## data:  lead_vs_takeover$Rearrange_takeover_time[lead_vs_takeover$Lead_time ==      "high"]
## W = 0.89051, p-value = 0.2366

t.test(Rearrange_takeover_time~Lead_time,data=lead_vs_takeover)

##
```

```
## Welch Two Sample t-test
##
## data: Rearrange_takeover_time by Lead_time
## t = -0.022623, df = 13.851, p-value = 0.9823
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.3596302 0.3521302
## sample estimates:
## mean in group low mean in group high
## 2.43750 2.44125
```

The two sample t test shows there is no significant difference between them. But it doesn't contradict to the result from density plot since t test takes the variance into consideration.

Add more variables

```
ancova_data <- raw[,c("Lead_time", "Rearrange_takeover_time", "Mean_age", "Modality", "NDRT")]
ancova_data <- ancova_data[apply(!is.na(ancova_data), 1, FUN = function(x) all(unlist(x))),]
ancova_data
```

	Lead_time	Rearrange_takeover_time	Mean_age	Modality	NDRT
## 2	6	2.050	44.60	A	N
## 3	6	2.420	44.60	V	N
## 4	12	2.890	25.70	HV	P
## 5	12	2.735	25.70	HV	P
## 6	12	2.740	25.70	HV	P
## 7	12	2.625	25.70	HV	P
## 8	12	2.010	25.70	HV	P
## 9	12	2.340	25.70	HV	P
## 10	12	2.060	25.70	HV	P
## 11	12	2.130	25.70	HV	P
## 13	7	2.310	30.10	AHV	M
## 14	7	2.290	30.10	AHV	M
## 15	7	2.630	30.10	AHV	M
## 16	7	2.100	30.10	AHV	M
## 17	7	2.920	31.36	AV	E
## 18	7	2.780	31.36	AV	E

The category of lead time and other factors are really similar. Thus we can't do regression or ANCOVA. We have to remove lead time.

```
anova_data <- raw[,c("Rearrange_takeover_time", "Modality", "NDRT")]
anova_data <- anova_data[apply(!(is.na(anova_data)|anova_data==""), 1, FUN = function(x) all(unlist(x))),]
print(anova_data)
```

	Rearrange_takeover_time	Modality	NDRT
## 2	2.050	A	N
## 3	2.420	V	N
## 4	2.890	HV	P
## 5	2.735	HV	P
## 6	2.740	HV	P
## 7	2.625	HV	P
## 8	2.010	HV	P
## 9	2.340	HV	P
## 10	2.060	HV	P

## 11	2.130	HV	P
## 13	2.310	AHV	M
## 14	2.290	AHV	M
## 15	2.630	AHV	M
## 16	2.100	AHV	M
## 17	2.920	AV	E
## 18	2.780	AV	E
## 19	2.770	AV	M
## 20	3.180	AV	M
## 21	3.210	AV	M
## 22	3.480	AV	M
## 23	4.580	AV	M
## 24	5.500	AV	M
## 25	4.510	AV	M
## 26	5.420	AV	M
## 27	3.230	AV	M
## 28	4.010	AV	M
## 29	2.940	AV	E
## 30	6.450	AV	E
## 31	3.400	AV	E
## 32	4.200	AV	E

Modality

Haptic

```
lapply(split(anova_data[11:30,1],anova_data[11:30,2]),function(x) {if(length(x)!=0){shapiro.test(x)$p.v
```

```
## [[1]]
## NULL
##
## $A
## NULL
##
## $AHV
## [1] 0.6098569
##
## $AV
## [1] 0.03805787
##
## $HV
## NULL
##
## $V
## NULL
```

```
bartlett.test(Rearrange_takeover_time~Modality,data=anova_data[11:30,])
```

```
##
## Bartlett test of homogeneity of variances
##
## data: Rearrange_takeover_time by Modality
## Bartlett's K-squared = 5.9142, df = 1, p-value = 0.01502
```

The shapiro test shows it doesnot satisfy the normality assumption. And bartlett test shows it doesnot satisfy the homoskedasticity assumption. Thus we tried the logarithm transformation.

```
anova_data$Rearrange_takeover_time <- log(anova_data$Rearrange_takeover_time)
lapply(split(anova_data[11:30,1],anova_data[11:30,2]),function(x) {if(length(x)!=0){shapiro.test(x)$p.v
```

```
## [[1]]
## NULL
##
## $A
## NULL
##
## $AHV
## [1] 0.6769054
##
## $AV
## [1] 0.1425158
##
## $HV
## NULL
##
## $V
## NULL

bartlett.test(Rearrange_takeover_time~Modality,data=anova_data[11:30,])

##
## Bartlett test of homogeneity of variances
##
## data: Rearrange_takeover_time by Modality
## Bartlett's K-squared = 3.1077, df = 1, p-value = 0.07792

m1 <- aov(Rearrange_takeover_time~Modality,data=anova_data[11:30,])
summary(m1)

##              Df Sum Sq Mean Sq F value    Pr(>F)
## Modality      1  0.7546   0.7546    12.52 0.00235 **
## Residuals    18  1.0847   0.0603
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

And the ANOVA table shows haptic warning is significant under 0.99 confidence level. It indicates that Haptic warning will decrease the takeover time significantly.

Audio

```
lapply(split(anova_data[3:14,1],anova_data[3:14,2]),function(x) {if(length(x)!=0){shapiro.test(x)$p.val
```

```
## [[1]]
## NULL
##
## $A
## NULL
##
## $AHV
## [1] 0.6769054
```

```
##
## $AV
## NULL
##
## $HV
## [1] 0.2178439
##
## $V
## NULL

bartlett.test(Rearrange_takeover_time~Modality,data=anova_data[3:14,])
```

```
##
## Bartlett test of homogeneity of variances
##
## data: Rearrange_takeover_time by Modality
## Bartlett's K-squared = 0.63998, df = 1, p-value = 0.4237

m2 <- aov(Rearrange_takeover_time~Modality,data=anova_data[3:14,])
summary(m2)
```

```
##           Df Sum Sq Mean Sq F value Pr(>F)
## Modality    1 0.00422 0.004216   0.245  0.631
## Residuals  10 0.17216 0.017216
```

The audio warning is not significant under 0.95 confidence level. Thus it won't have great influence on the takeover time.

NDRT

```
t.test(anova_data$Rearrange_takeover_time[anova_data$NDRT=="N"],anova_data$Rearrange_takeover_time[anova_data$NDRT=="Y"],var.equal=FALSE)

##
## Welch Two Sample t-test
##
## data: anova_data$Rearrange_takeover_time[anova_data$NDRT == "N"] and anova_data$Rearrange_takeover_time[anova_data$NDRT == "Y"]
## t = -0.84803, df = 1.865, p-value = 0.4912
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.5323895 0.3671117
## sample estimates:
## mean of x mean of y
## 0.8008037 0.8834425

t.test(anova_data$Rearrange_takeover_time[anova_data$NDRT=="N"],anova_data$Rearrange_takeover_time[anova_data$NDRT=="Y"],var.equal=FALSE)
```

```
##
## Welch Two Sample t-test
##
## data: anova_data$Rearrange_takeover_time[anova_data$NDRT == "N"] and anova_data$Rearrange_takeover_time[anova_data$NDRT == "Y"]
## t = -3.0985, df = 5.4528, p-value = 0.02398
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.87165552 -0.09186522
## sample estimates:
```

```
## mean of x mean of y
## 0.8008037 1.2825640

t.test(anova_data$Rearrange_takeover_time[anova_data$NDRT=="N"],anova_data$Rearrange_takeover_time[anova_data$NDRT=="Y"])

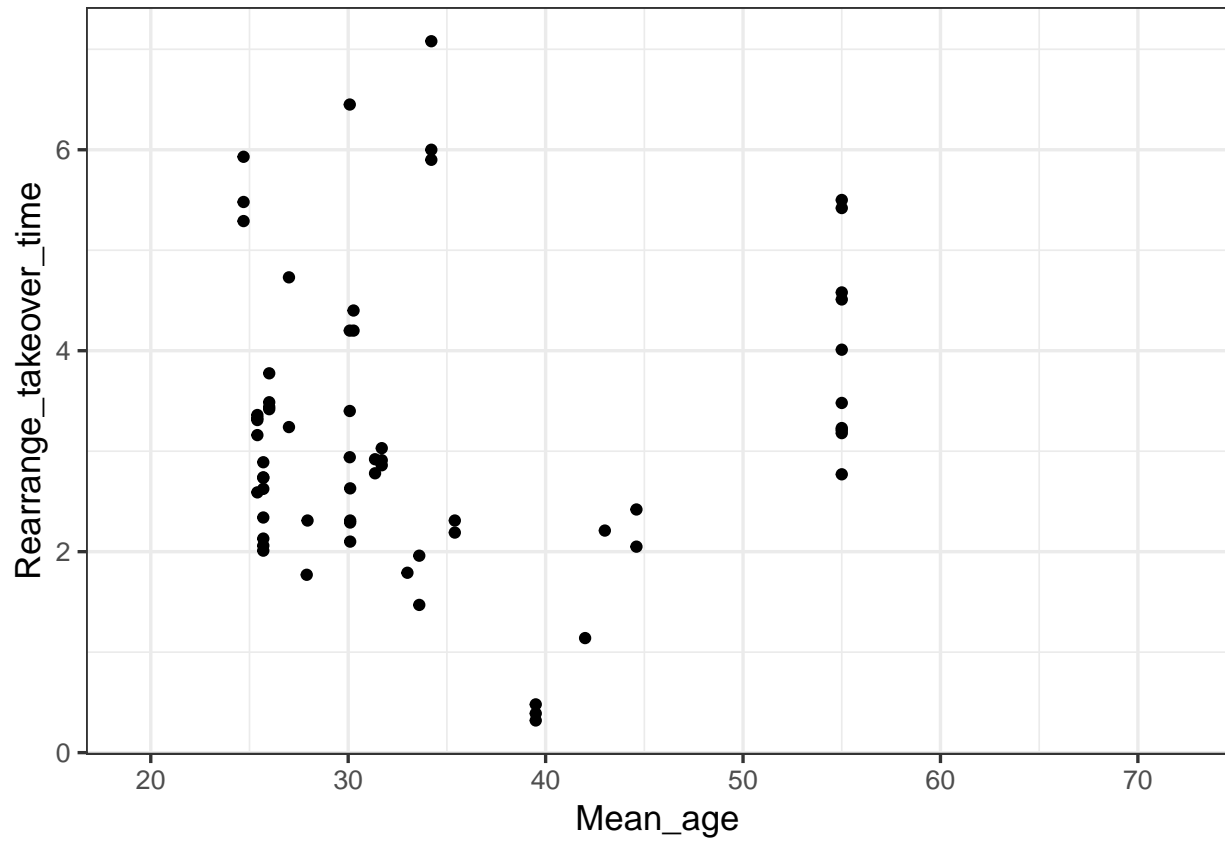
##
## Welch Two Sample t-test
##
## data: anova_data$Rearrange_takeover_time[anova_data$NDRT == "N"] and anova_data$Rearrange_takeover_time[anova_data$NDRT == "Y"]
## t = -3.4607, df = 3.8317, p-value = 0.02763
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.74427080 -0.07524554
## sample estimates:
## mean of x mean of y
## 0.8008037 1.2105618
```

The two sample t-test indicates that eye distraction task and miscellaneous task will significantly increase the takeover time compared with the no-secondary task test. But physical movement related seems has little influence on it.

Others

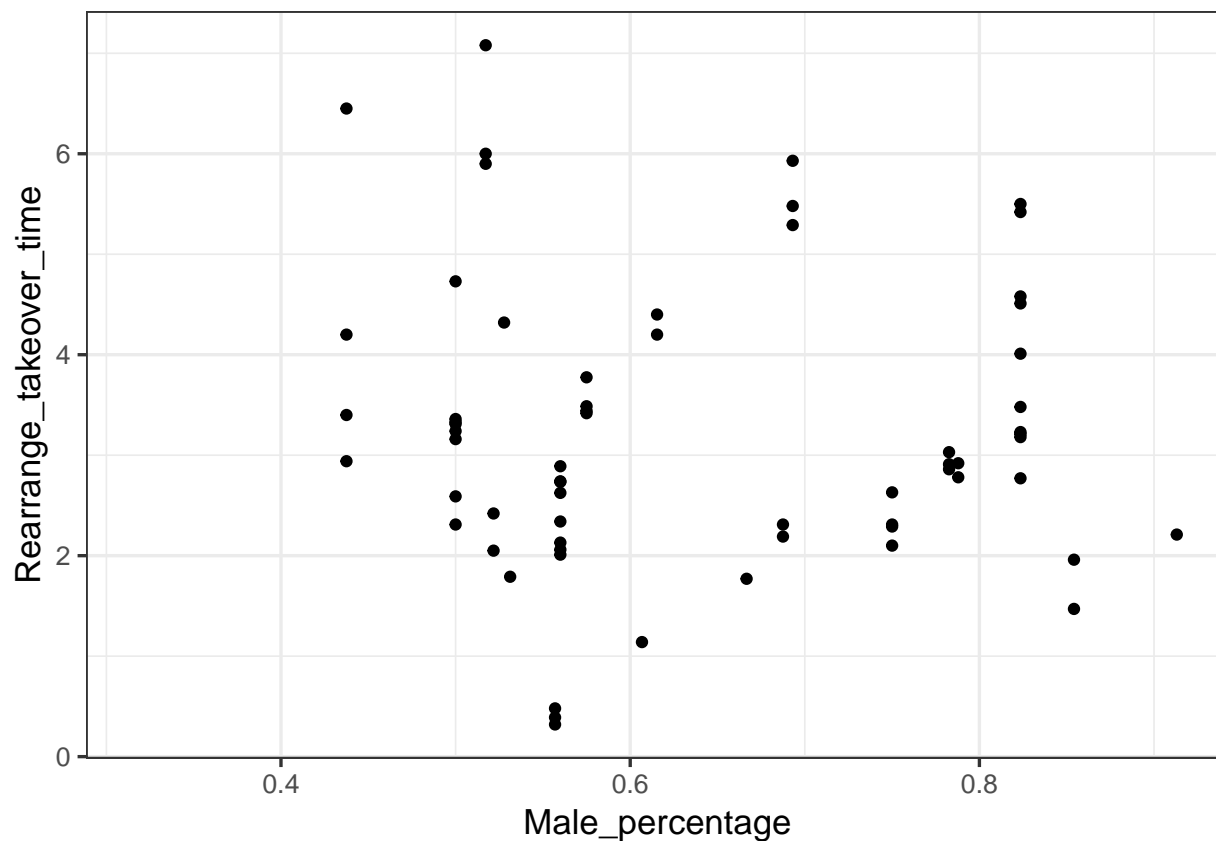
```
raw[, "Male_percentage"] <- raw[, "Male"] / (raw[, "Female"] + raw[, "Male"])
ggplot(data=raw, aes(x=Mean_age, y=Rearrange_takeover_time)) + geom_point() + theme_bw(base_size=13) + theme(text = element_size(13))

## Warning: Removed 58 rows containing missing values (geom_point).
```

```
ggplot(data=raw,aes(x=Male_percentage,y=Rearrange_takeover_time))+geom_point()+theme_bw(base_size=13)+t
```

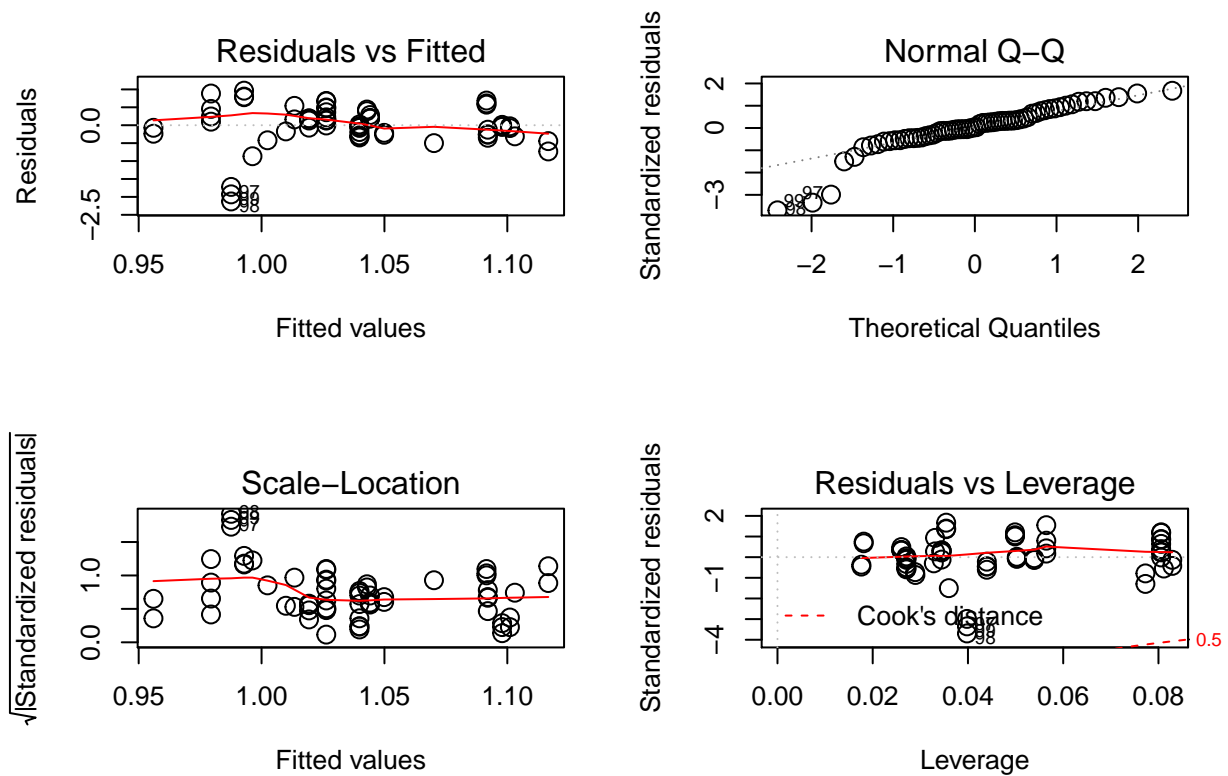
```
## Warning: Removed 57 rows containing missing values (geom_point).
```



```
m<-lm(log(Rearrange_takeover_time)~Male_percentage+Mean_age,data=raw)
summary(m)
```

```
##
## Call:
## lm(formula = log(Rearrange_takeover_time) ~ Male_percentage +
##     Mean_age, data = raw)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.12704 -0.24233  0.01605  0.30382  0.96439
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.933076   0.360756   2.586   0.0121 *
## Male_percentage 0.360664   0.668750   0.539   0.5916
## Mean_age      -0.003705   0.008853  -0.418   0.6771
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5875 on 61 degrees of freedom
## (58 observations deleted due to missingness)
## Multiple R-squared:  0.005015,    Adjusted R-squared:  -0.02761
## F-statistic: 0.1537 on 2 and 61 DF,  p-value: 0.8578
```

```
par(mfrow=c(2,2))
plot(m,cex=1.5)
```



Actually I didnot do much in this section. From the scatter plot we could find that they have no obvious pattern. But I still did the regression. All p-values are insignificant after removing the three suspicious outliers and trasforming the takeover time.

One thing to be concerned

anova_data

```
##      Rearrange_takeover_time Modality NDRT
## 2          0.7178398          A      N
## 3          0.8837675          V      N
## 4          1.0612565         HV      P
## 5          1.0061314         HV      P
## 6          1.0079579         HV      P
## 7          0.9650809         HV      P
## 8          0.6981347         HV      P
## 9          0.8501509         HV      P
## 10         0.7227060         HV      P
## 11         0.7561220         HV      P
## 13         0.8372475        AHV      M
## 14         0.8285518        AHV      M
## 15         0.9669838        AHV      M
## 16         0.7419373        AHV      M
## 17         1.0715836          AV      E
## 18         1.0224509          AV      E
```

## 19	1.0188473	AV	M
## 20	1.1568812	AV	M
## 21	1.1662709	AV	M
## 22	1.2470323	AV	M
## 23	1.5216990	AV	M
## 24	1.7047481	AV	M
## 25	1.5062972	AV	M
## 26	1.6900958	AV	M
## 27	1.1724821	AV	M
## 28	1.3887912	AV	M
## 29	1.0784096	AV	E
## 30	1.8640801	AV	E
## 31	1.2237754	AV	E
## 32	1.4350845	AV	E

Let's go back to the dataset. The NDRT and Modality has a similar part from No.4 to No.11. When we compare the difference of takeover time with one factor, the result may be distorted by the other one. That is to say, if Audio warning decrease the takeover time while the physical task increase the time, it may just offset the influence. Here I just ignore this problem. I searched but have no ideal how to deal with it.