Homework 3

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Problem 4

There are some programming skills in notations, syntax, functions and organisation to make code clean and clear:

- Using lowercase and underscore in the variable names
- Placing spaces around operators and before left parentheses, except in a function call.
- The opening curly brace should never go on its own line and should always followed by a new line. A closing curly brace should always go on its own line, unless it's followed by else.
- Each lines shouldn't exceed 80 characters.
- When function definition runs over multiple lines, indent the second line where the definition starts.
- Commenting code and use - or = to seperate code into readable chunks.

I will pay attention to improve the identation and comments.

Problem 5

```
library(lintr)
lint(filename = "../02_data_munging_summarizing_R_git/HW2_Wang_Yueyao.Rmd")
```

- Pay attention that lines shouldn't more than 80 characters.
- Variable and function names should be all lowercase.
- Only use double-quotes for text.
- Commas should always have a space after.

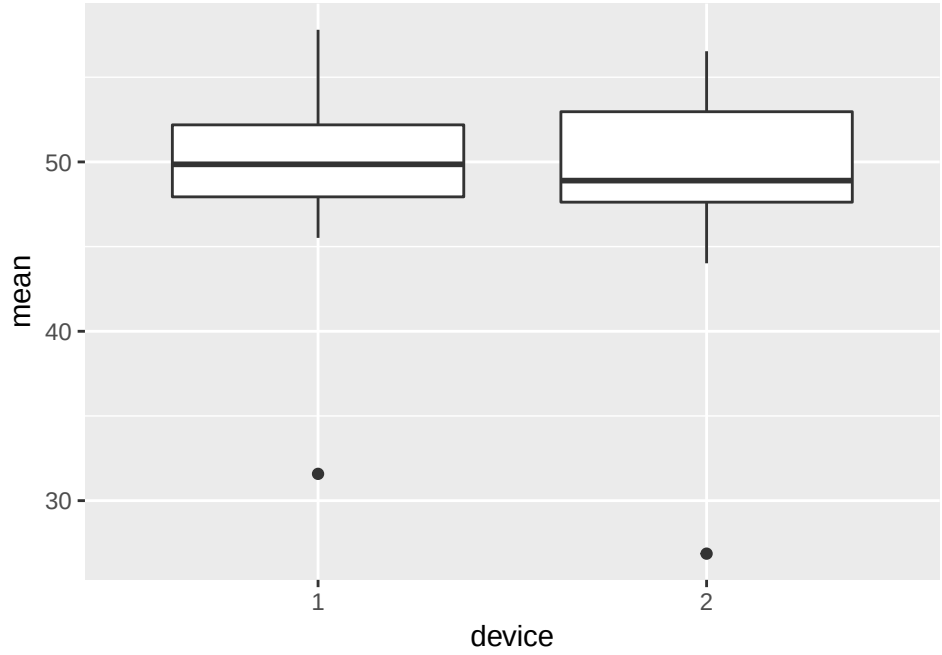
Problem 6

- The table containing means, sds and correlations of 13 observers is following:

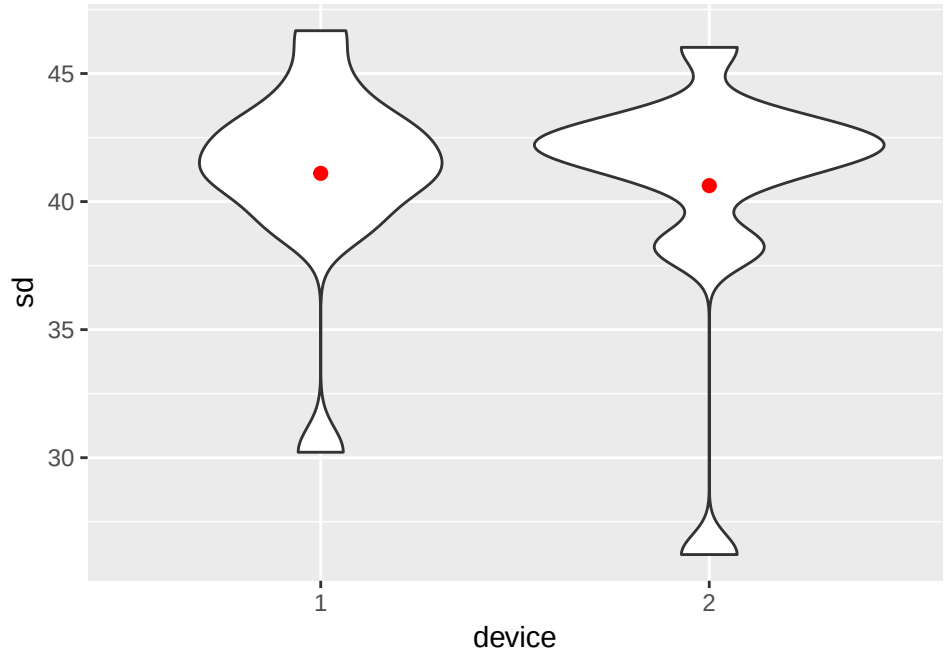
Table 1: Summary of data by Obersvers

	mean_dev1	mean_dev2	sd_dev1	sd_dev2	correlation
Observer1	31.5807	26.8694	30.2125	26.2174	-0.0641
Observer2	45.5146	48.8248	40.9673	42.6652	-0.0686
Observer3	46.0901	44.0151	39.0910	38.3621	-0.0683
Observer4	52.1880	50.5214	46.6719	46.0212	-0.0645
Observer5	49.0051	48.2065	42.9738	42.0988	-0.0603
Observer6	49.8582	49.6300	42.2135	42.1337	-0.0617
Observer7	49.5061	47.0308	39.0883	38.0867	-0.0685
Observer8	52.7384	53.5297	44.6718	43.0503	-0.0690
Observer9	50.6455	48.8953	43.1962	42.2014	-0.0686
Observer10	53.3652	53.4703	41.1557	41.3107	-0.0630
Observer11	51.5681	52.9641	41.1221	43.3672	-0.0694
Observer12	57.8013	56.5352	42.3666	40.7425	-0.0666
Observer13	47.9368	47.6229	40.5966	41.8410	-0.0656

- The boxplot of mean for device1 and device2 is following:



- The violin plot of sd for device1 and device2 is following:



Problem 7

The issue with this dataset is that:

- Column headers are values: the columns Devs and Docs in the original dataset should be the value of a variable that indicating the pressure read by devices or doctors.

- Multiple variables are stored in one column: Also we should have a variable of ID number indicating which device or doctors read this value, not combining them togehter.

So I would like to clean and munge the data into 4 variables:

- Day: the day recording those values
- Type: indicating the value is read by doctor or device
- ID: indicating which device or doctor read this value
- Pressure: the corresponding blood pressure

The tidying process:

- Use the gather function to convert those columns header Devs and Docs into a variable read_by_type
- Separate this characters and numbers in read_by_type variable into two columns Type and ID via mutate and gsub function
- Select the columns that we want to keep and arrange the rows by the ascending order of Day

So the first 10 rows of the tidy data is following:

Table 2: 10 Rows of Blood Pressure Data after Tidying

Day	type	ID	pressure
1	Dev	1	133.34
1	Dev	2	133.36
1	Dev	3	133.45
1	Doc	1	126.54
1	Doc	2	127.36
1	Doc	3	131.88
2	Dev	1	110.94
2	Dev	2	110.85
2	Dev	3	110.92
2	Doc	1	124.69

Problem 8

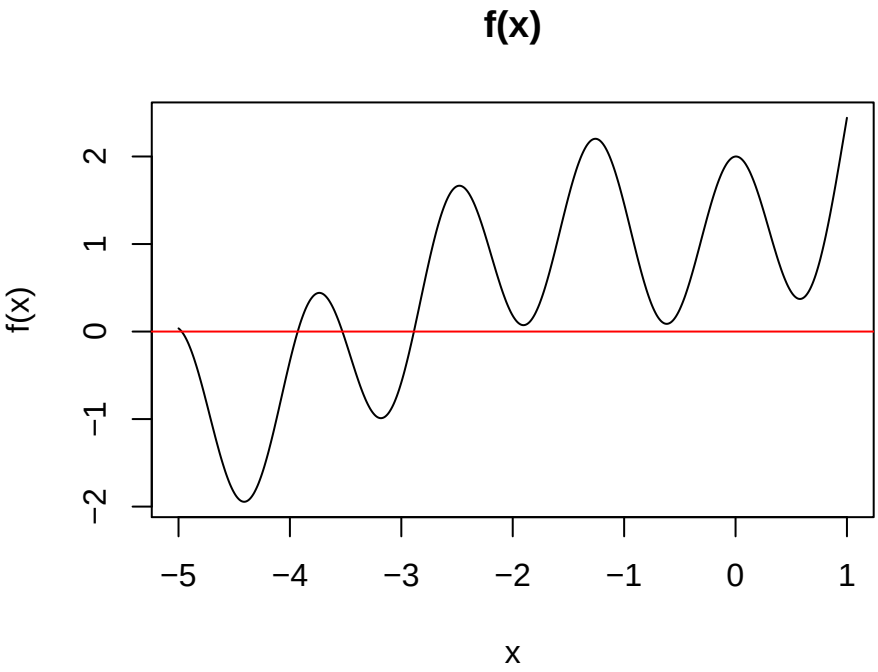
We need to calculate the derivative function of $f(x)$ in the Newton's method which is:

$$f^{'}(x) = 3^{x}log(3) - cos(x) - 5sin(5x)$$

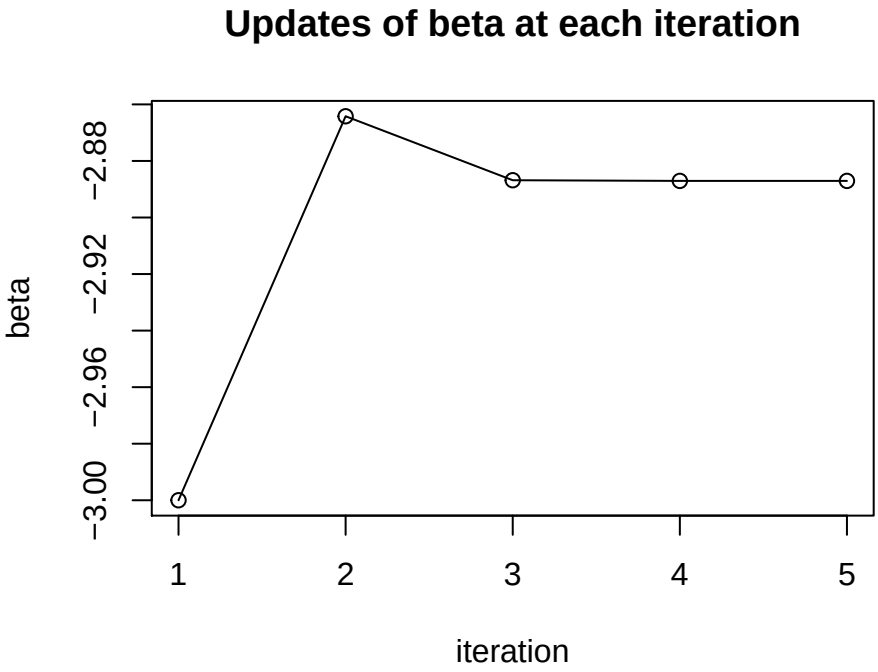
The procedure Newton's method is following:

set initial value of x denoted as x_0 and tolerance ϵ
repeat:
  update $x_{new} = x_0 - f(x_0)/f^{'}(x_0)$
until $|x_0 - x_{new}| \leq \epsilon$

Before directly applying this method, we can plot this function to choose an initial value. Because a bad initial point is easy to cause unconvergency.



From the above plot we could see that there isn't unique root for this function because the trigonometric function. We can just find the biggest negative root in this problem which is between -3 and -2 from the plot.



So from the above procedure we could know, when the tolerance is 1e-5, the biggest negative root of this function is -2.88706

Problem 9

d.how many DIFFERENT makes and models of cars you end with (?unique ?distinct ?duplicated) considering only year 2017

The different makes of cars is 503 and the different models of cars is 33470.

e. report a table of the 5 most frequent defects (translated) and the top make/models having that defect (?count) again considering only year 2017

Table 3: Top 5 mose frequent defects and the top coresponding make

defect_code	make	make_frequency	defect_frequency
AC1	VOLKSWAGEN	43490	385634
K04	PEUGEOT	27704	271273
RA2	OPEL	29089	223171
205	VOLKSWAGEN	18395	171255
497	PEUGEOT	16316	139978

f. use function lm to test for a relationship between number of defects observed by make, report both the coefficient and anova tables (2017 only)

	Estimate	Std. Error	t value	Pr(>|t|)
make	0.04069	0.03509	1.159	0.2468
(Intercept)	41.95	10.21	4.111	4.611e-05

Table 5: Fitting linear model: number_defect ~ make

Observations	Residual Std. Error	R^2	Adjusted R^2
503	114.3	0.002676	0.0006854

Table 6: Analysis of Variance Table

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
make	1	17555	17555	1.344	0.2468
Residuals	501	6542619	13059	NA	NA

g.repeat (f) by model (2017 only)

	Estimate	Std. Error	t value	Pr(>|t|)
model	0.0001302	1.92e-05	6.781	1.217e-11
(Intercept)	15.56	0.371	41.93	0

Table 8: Fitting linear model: number_defect ~ model

Observations	Residual Std. Error	R^2	Adjusted R^2
33470	33.94	0.001372	0.001342

Table 9: Analysis of Variance Table

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
model	1	52948	52948	45.98	1.217e-11
Residuals	33468	38543639	1152	NA	NA

h.comment on this workflow and how you might be more computationally efficient

I think save those important intermediate dataset for those summarise analysis would improve the efficiency a lot.

Appendix: R code

```
####### Problem6: construct summarise function and
####### table#############
data <- readRDS("./HW3_data.rds")
#-------------------Construct functions----------------------------------------
# input: x is a dataframe for all samples from 1
# Observer return:a dataframe of descriptive statstics
descrip_stats <- function(x) {
    mean_dev1_dev2 <- apply(x, MARGIN = 1, FUN = mean)
    sd_dev1_dev2 <- apply(x, MARGIN = 1, FUN = sd)
    correlation <- cor(x[, 2], x[, 3])
    d <- data.frame(mean_dev1 = mean_dev1_dev2[1], mean_dev2 = mean_dev1_dev2[2],
        sd_dev1 = sd_dev1_dev2[1], sd_dev2 = sd_dev1_dev2[2],
        correlation = correlation)
    return(d)
}
#----------loop through the Observers via the descrip_stats function-----------
obs_1 <- data[data$Observer == 1, ]
com_df <- descrip_stats(obs_1)
for (i in 2:13) {
    obs_i <- data[data$Observer == i, ]
    des_df <- descrip_stats(obs_i)
    com_df <- rbind(com_df, des_df)
}
rownames(com_df) <- paste("Observer", 1:13, sep = "")
kable(com_df, caption = "Summary of data by Obersvers",
    digits = 4) %>% kable_styling(full_width = T)

#-------------create boxplot for dev1 and dev2 mean-------------------------------
device_mean <- c(com_df[, 1], com_df[, 2])
device_index <- rep(c(1, 2), each = 13)
mean_summary <- data.frame(mean = device_mean, device = as.factor(device_index))
ggplot(data = mean_summary, aes(x = device, y = mean), main = "boxplot of the device mean") +
    geom_boxplot()

#-----------------create violin plot fro dev1 and dev2 sd-------------------------------
device_sd <- c(com_df[, 3], com_df[, 4])
sd_summary <- data.frame(sd = device_sd, device = as.factor(device_index))
ggplot(data = sd_summary, aes(x = device, y = sd), main = "violin plot for standard deviation") +
    geom_violin() + stat_summary(fun.y = mean, geom = "point",
```

```
    color = "red", size = 2)

#### Problem 8:Plot the function to choose initial value
#### #####
f <- function(x) {
    value <- 3^x - sin(x) + cos(5 * x)
    return(value)
}
x <- seq(-5, 1, by = 0.01)
plot(x, f(x), main = "f(x)", type = "l")
abline(h = 0, col = "red")
######## Problem 8: Construct the function ########
derv_f <- function(x) {
    value <- 3^x * log(3) - cos(x) - 5 * sin(5 * x)
    return(value)
}
## input: initial value: set it as -3 default tolerance:
## 1e-5 as the default output: a vector of the updates of
## beta at each iteration
newton_method <- function(epsilon = 1e-05, beta_0 = -3) {
    beta <- c(beta_0)
    while (TRUE) {
        beta_new <- beta_0 - f(beta_0)/derv_f(beta_0)
        beta <- c(beta, beta_new)
        if (abs(beta_new - beta_0) < epsilon) {
            beta_est <- beta_new
            break
        } else {
            beta_0 <- beta_new
        }
    }
    return(beta)
}
beta <- newton_method()
plot(1:length(beta), beta, xlab = "iteration", main = "Updates of beta at each iteration")
lines(1:length(beta), beta)
######## Problem9: Cleaning and Combining data ###########
#---------------a: load datasets into R  ---------------------
# this had the defect code and description
car_gebreken_select <- fread(input = "../02_data_munging_summarizing_R_git/Open_Data_RDW__Gebreken.csv"
    header = T, select = c(1, 6), showProgress = F)
# this has the license plate, inspection date and defect
# code
car_geconstat_select <- fread(input = "../02_data_munging_summarizing_R_git/Open_Data_RDW__Geconstateerd
    header = T, select = c(1, 3, 5), showProgress = F)
# this has the license plate, make and model of vehicle
car_person_select <- fread(input = "../02_data_munging_summarizing_R_git/Personenauto_basisdata.csv",
    header = T, showProgress = F, select = c(1, 3, 4))
#--------------- b: join them together -------------------------
### First join car_person_select and car_Geconstat_select
### by license plate Then join the merged one with
### car_gebreken_select by defect code
merged_geconstat_person <- merge(car_geconstat_select, car_person_select,
```

```
    by = "Kenteken", all = T)
merged_all <- merge(merged_geconstat_person, car_gebreken_select,
    by = "Gebrek identificatie", all = T)
#------------------ c: clean the data  ----------------------------
### translate the variables and remove observations
### containg missing values for any of the variables
colnames(merged_all) <- c("defect_code", "liscence_plate",
    "inspection_date", "make", "model", "defect_description")
merged_tidy <- na.omit(merged_all)


####### Problem9:d ######
merged_tidy_2017 <- merged_tidy[grep("2017", merged_tidy$inspection_date),
    ]
diff_makes <- n_distinct(merged_tidy_2017$make)
diff_models <- n_distinct(merged_tidy_2017$model)


####### Problem9:f ######
summary2 <- merged_tidy_2017 %>% group_by(make) %>% summarise(number_defect = n_distinct(defect_code)) %
    mutate(make = as.numeric(as.factor(make)))
fit1 <- lm(number_defect ~ make, data = summary2)
pander(summary(fit1))
pander(anova(fit1))


####### Problem9:g ######
summary3 <- merged_tidy_2017 %>% group_by(model) %>% summarise(number_defect = n_distinct(defect_code))
    mutate(model = as.numeric(as.factor(model)))
fit2 <- lm(number_defect ~ model, data = summary3)
pander(summary(fit2))
pander(anova(fit2))
```