HW9

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```
library(ggplot2)
library(reshape2)
library(stringr)
library(scales)
library(plyr)
library(lubridate)
##
## Attaching package: 'lubridate'
## The following object is masked from 'package:plyr':
##
##
       here
## The following object is masked from 'package:base':
##
##
       date
library(kernlab)
##
## Attaching package: 'kernlab'
## The following object is masked from 'package:scales':
##
##
       alpha
## The following object is masked from 'package:ggplot2':
##
##
       alpha
# For Grid Arrange
library(gridExtra)
## Warning: package 'gridExtra' was built under R version 3.5.3
```

```
library(grid)
library(ggplot2)
library(lattice)
```

```
rm(list=ls())
# Step 1: Load the data
my_aq<- airquality

# Step 2: Clean the data
my_aq<-na.omit(my_aq)

my_aq$Month<-as.factor(my_aq$Month)

my_aq$Day<-as.factor(my_aq$Day)

my_aq<-cbind.data.frame(my_aq,"date"=as.Date(gsub(" ","",paste(str_pad(my_aq$Month,2,side="left",pad = "0"),"-",str_pad(my_aq$Day,2,side="left",pad = "0"),"-1973")),"%m-%d-%Y"))

my_aq.m <- melt(my_aq,id.vars = "date", measure.vars = c("Ozone", "Solar.R","Wind","Temp"))
my_aq.m <- ddply(my_aq.m, .(variable), transform, rescale = rescale(value))

str(my_aq)</pre>
```

```
## 'data.frame': 111 obs. of 7 variables:
## $ Ozone : int 41 36 12 18 23 19 8 16 11 14 ...
## $ Solar.R: int 190 118 149 313 299 99 19 256 290 274 ...
## $ Wind : num 7.4 8 12.6 11.5 8.6 13.8 20.1 9.7 9.2 10.9 ...
## $ Temp : int 67 72 74 62 65 59 61 69 66 68 ...
## $ Month : Factor w/ 5 levels "5","6","7","8",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ Day : Factor w/ 31 levels "1","2","3","4",..: 1 2 3 4 7 8 9 12 13 14 ...
## $ date : Date, format: "1973-05-01" "1973-05-02" ...
```

summary(my_aq)

```
##
        0zone
                      Solar.R
                                         Wind
                                                         Temp
                                                                   Month
##
   Min.
         : 1.0
                          : 7.0
                                           : 2.30
                                                                    5:24
                   Min.
                                   Min.
                                                   Min.
                                                           :57.00
##
   1st Qu.: 18.0
                   1st Qu.:113.5
                                    1st Qu.: 7.40
                                                    1st Qu.:71.00
                                                                    6: 9
   Median : 31.0
                   Median :207.0
                                   Median: 9.70
                                                    Median :79.00
                                                                   7:26
##
   Mean
         : 42.1
                   Mean :184.8
                                    Mean : 9.94
                                                    Mean
                                                          :77.79
                                                                    8:23
##
                   3rd Qu.:255.5
##
   3rd Qu.: 62.0
                                    3rd Qu.:11.50
                                                    3rd Qu.:84.50
                                                                    9:29
##
   Max.
          :168.0
                   Max.
                           :334.0
                                    Max.
                                          :20.70
                                                    Max.
                                                           :97.00
##
##
         Day
                      date
##
   7
          : 5
                Min.
                       :1973-05-01
##
   9
          : 5
                1st Qu.:1973-06-14
          : 5
##
   13
                Median :1973-07-28
          : 5
                       :1973-07-22
##
   16
                Mean
   17
           : 5
                3rd Qu.:1973-09-01
##
##
   18
           : 5
                Max.
                       :1973-09-30
##
    (Other):81
```

```
#Create trainign and test data set
nrows <-nrow(my_aq)
nrows
```

```
## [1] 111
```

```
cutpoint <-floor(nrows/3*2)
cutpoint</pre>
```

```
## [1] 74
```

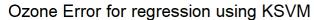
```
rand <- sample(1:nrows)

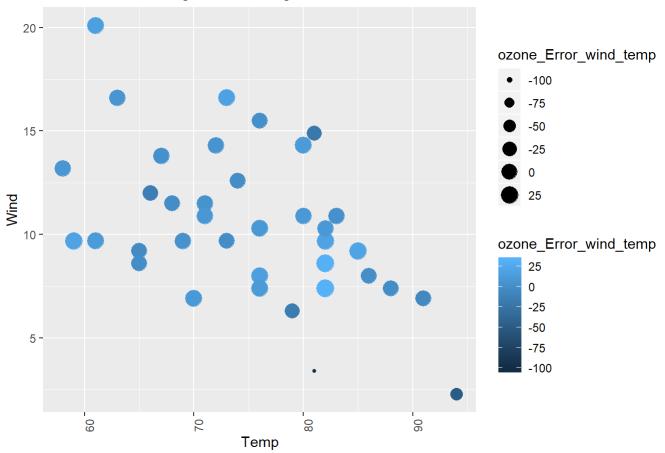
my_aq.train <- my_aq[rand[1:cutpoint],]

my_aq.test <- my_aq[rand[(cutpoint+1):nrows],]</pre>
```

```
#Build a model by using all variable
mq aq.model.all.var <- ksvm(Ozone ~ Wind+Temp+Solar.R+Month, data=my aq.train)
#mq aq.model.all.var
# Test the model using test dataset
my aq.test$ozone pred all var <- predict(mq aq.model.all.var,my aq.test)</pre>
# store the prediction results for test dataset
my aq.test.results.all.var <- table(my aq.test$ozone pred all var ,my aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_all_var <- (my_aq.test$ozone_pred_all_var-my_aq.test$0zone)^2
mse.ozone.all.var <- mean(my aq.test$ozone squaredError all var)</pre>
#Build a model by using wind, temp, solar variable
mq aq.model.wind.temp.solar <- ksvm(Ozone ~ Wind+Temp+Solar.R, data=my aq.train)
#mq aq.model.all.var
# Test the model using test dataset
my aq.test$ozone pred wind temp solar <- predict(mq aq.model.wind.temp.solar,my aq.test)
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.wind.temp.solar<- table(my aq.test$ozone pred wind temp solar,my aq.test$Ozon
e)
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredErrorwind_temp_solar <- (my_aq.test$ozone_pred_wind_temp_solar-my_aq.tes
t$0zone)^2
mse.ozone.wind.temp.solar <- mean(my_aq.test$ozone_squaredErrorwind_temp_solar)</pre>
#Build a model by using Temp and Wind variable
mq aq.model.wind.temp <- ksvm(Ozone ~ Wind+Temp, data=my aq.train)
#mq aq.model.wind.temp
# Test the model using test dataset
my aq.test$ozone wind temp <- predict(mq aq.model.wind.temp,my aq.test)
#table(pred,my_aq.test$0zone)
# store the prediction results for test dataset
my_aq.test.results.wind.temp <- table(my_aq.test$ozone_wind_temp,my_aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_wind_temp <- (my_aq.test$ozone_wind_temp-my_aq.test$Ozone)^2
my aq.test$ozone Error wind temp <- (my aq.test$ozone wind temp-my aq.test$Ozone)
mse.ozone.wind.temp <- mean(my ag.test$ozone squaredError wind temp)</pre>
#Build a model by using Temp variable
mq_aq.model.temp <- ksvm(Ozone ~ Temp, data=my_aq.train)</pre>
#mq_aq.model.temp
# Test the model using test dataset
my_aq.test$ozone_temp <- predict(mq_aq.model.temp,my_aq.test)</pre>
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.temp <- table(my aq.test$ozone temp,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError temp <- (my aq.test$ozone temp-my aq.test$Ozone)^2
mse.ozone.temp <- mean(my_aq.test$ozone_squaredError_temp)</pre>
#Build a model by using Wind variable
```

```
mq aq.model.wind <- ksvm(Ozone ~ Wind, data=my aq.train)
#mg ag.model.wind
# Test the model using test dataset
my aq.test$ozone wind <- predict(mq aq.model.wind,my aq.test)</pre>
#table(pred,my_aq.test$0zone)
# store the prediction results for test dataset
my aq.test.results.wind <- table(my aq.test$ozone wind,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_wind <- (my_aq.test$ozone_wind-my_aq.test$0zone)^2</pre>
mse.ozone.wind <- mean(my aq.test$ozone squaredError wind)</pre>
#Build a model by using Solar.R variable
mq aq.model.solar <- ksvm(Ozone ~ Solar.R, data=my aq.train)
#mg ag.model.solar
# Test the model using test dataset
my_aq.test$ozone_solar <- predict(mq_aq.model.solar,my_aq.test)</pre>
#table(pred,my aq.test$0zone)
# store the prediction results for test dataset
my_aq.test.results.solar <- table(my_aq.test$ozone_solar,my_aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError solar <- (my aq.test$ozone solar-my aq.test$Ozone)^2
mse.ozone.solar <- mean(my aq.test$ozone squaredError solar)</pre>
#scatter
            chart (using
                            ggplot geom_point),
                                                     with
                                                             the x-axis representing
                                                                                          the tem
     the
р,
#y-axis representing
                      the wind, the size and color of each
                                                                     dot representing
                                                                                          the erro
ggplot(my aq.test) +
  geom point(aes(x=Temp,y=Wind,size=ozone Error wind temp,color=ozone Error wind temp)) + theme
(legend.position = "right",axis.text.x = element_text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "Ozone Error for regression using KSVM")
```





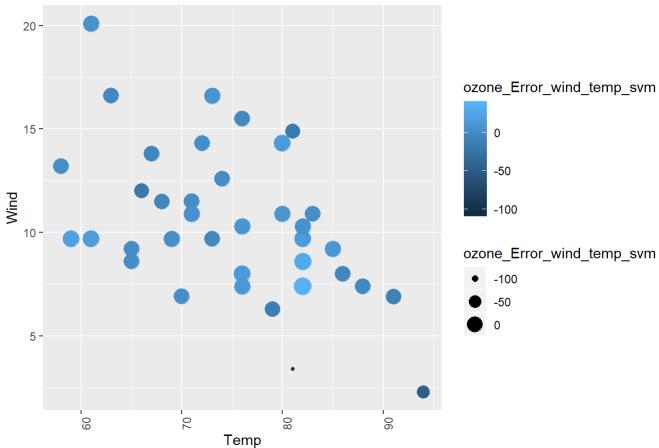
library(e1071)

Warning: package 'e1071' was built under R version 3.5.3

```
#Build a model by using all variable
mq aq.svm.model.all.var <- svm(Ozone ~ Wind+Temp+Solar.R+Month, data=my aq.train)
#mq aq.svm.model.all.var
# Test the model using test dataset
my_aq.test$ozone_pred_all_var_svm <- predict(mq_aq.svm.model.all.var,my_aq.test)</pre>
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.all.var.svm <- table(my aq.test$ozone pred all var svm,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_all_var_svm <- (my_aq.test$ozone_pred_all_var_svm-my_aq.test$Ozon
e)^2
mse.ozone.all.var.svm <- mean(my aq.test$ozone squaredError all var svm)</pre>
#Build a model by using wind, temp, solar variable
mq aq.svm.model.wind.temp.solar <- svm(Ozone ~ Wind+Temp+Solar.R, data=my aq.train)
#mq aq.model.all.var
# Test the model using test dataset
my aq.test$ozone pred wind temp solar svm <- predict(mq aq.svm.model.wind.temp.solar,my aq.test)
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.wind.temp.solar.svm<- table(my aq.test$ozone pred wind temp solar svm,my aq.t
est$Ozone)
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredErrorwind_temp_solar_svm <- (my_aq.test$ozone_pred_wind_temp_solar_svm-m
y aq.test$0zone)^2
mse.ozone.wind.temp.solar.svm <- mean(my_aq.test$ozone_squaredErrorwind_temp_solar_svm)</pre>
#Build a model by using Temp and Wind variable
mq aq.svm.model.wind.temp <- svm(Ozone ~ Wind+Temp, data=my aq.train)
#mq aq.model.wind.temp
# Test the model using test dataset
my_aq.test$ozone_wind_temp_svm <- predict(mq_aq.svm.model.wind.temp,my_aq.test)</pre>
#table(pred,my aq.test$0zone)
# store the prediction results for test dataset
my aq.test.results.wind.temp.svm <- table(my aq.test$ozone wind temp svm,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError wind temp svm <- (my aq.test$ozone wind temp svm-my aq.test$Ozone)
^2
my aq.test$ozone Error wind temp svm <- (my aq.test$ozone wind temp svm-my aq.test$Ozone)
mse.ozone.wind.temp.svm <- mean(my_aq.test$ozone_squaredError_wind_temp_svm)</pre>
#Build a model by using Temp variable
mq aq.svm.model.temp <- svm(Ozone ~ Temp, data=my aq.train)</pre>
#mq aq.model.temp
# Test the model using test dataset
my_aq.test$ozone_temp_svm <- predict(mq_aq.svm.model.temp,my_aq.test)</pre>
#table(pred,my aq.test$0zone)
# store the prediction results for test dataset
my aq.test.results.temp.svm <- table(my aq.test$ozone temp svm,my aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
```

```
my aq.test$ozone squaredError temp svm <- (my aq.test$ozone temp svm-my aq.test$Ozone)^2
mse.ozone.temp.svm <- mean(my_aq.test$ozone_squaredError_temp_svm)</pre>
#Build a model by using Wind variable
mq_aq.svm.model.wind <- svm(Ozone ~ Wind, data=my aq.train)
#mg ag.model.wind
# Test the model using test dataset
my ag.test$ozone wind svm <- predict(mg ag.svm.model.wind,my ag.test)
#table(pred,my_aq.test$0zone)
# store the prediction results for test dataset
my_aq.test.results.wind.svm <- table(my_aq.test$ozone_wind_svm,my_aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError wind svm <- (my aq.test$ozone wind svm-my aq.test$Ozone)^2
mse.ozone.wind.svm <- mean(my ag.test$ozone squaredError wind svm)</pre>
#Build a model by using Solar.R variable
mq aq.svm.model.solar <- svm(Ozone ~ Solar.R, data=my aq.train)
#mg ag.model.solar
# Test the model using test dataset
my aq.test$ozone solar svm <- predict(mq aq.svm.model.solar,my aq.test)</pre>
#table(pred,my aq.test$0zone)
# store the prediction results for test dataset
my_aq.test.results.solar.svm <- table(my_aq.test$ozone_solar_svm,my_aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_solar_svm <- (my_aq.test$ozone_solar_svm-my_aq.test$Ozone)^2</pre>
mse.ozone.solar.svm <- mean(my ag.test$ozone squaredError solar svm)</pre>
            chart (using
#scatter
                            ggplot geom point),
                                                     with
                                                             the x-axis representing
                                                                                          the tem
     the
р,
                        the wind, the size and color of each
#y-axis representing
                                                                     dot representing
                                                                                          the erro
ggplot(my aq.test) +
  geom_point(aes(x=Temp,y=Wind,size=ozone_Error_wind_temp_svm,color=ozone_Error_wind_temp_svm))
 + theme(legend.position = "right",axis.text.x = element text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "Ozone Error for regression using SVM")
```

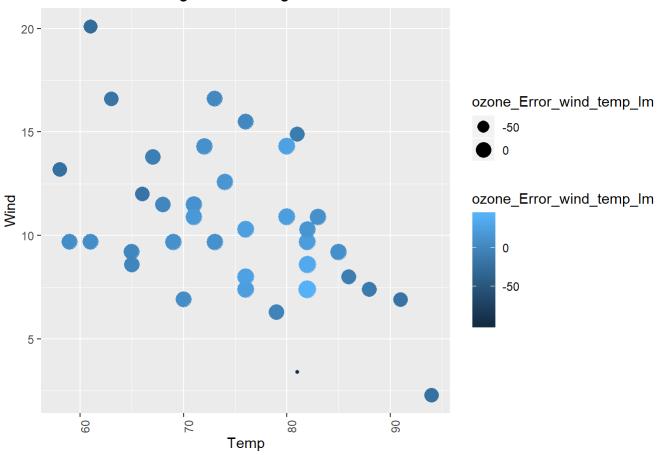




```
#Build a model by using all variable
mq aq.lm.model.all.var <- lm(Ozone ~ Wind+Temp+Solar.R+Month, data=my aq.train)
#mg ag.lm.model.all.var
# Test the model using test dataset
#not able to predict the test data with lm using all variable
#my aq.test$ozone pred all var lm <- predict(mq aq.lm.model.all.var,my aq.test)</pre>
#table(pred,my_aq.test$0zone)
# store the prediction results for test dataset
#my aq.test.results.all.var.lm <- table(pred,my aq.test$0zone)</pre>
# Square of Error - Support Vector Machine
#my aq.test$ozone squaredError all var lm <- (my aq.test$ozone pred all var lm-my aq.test$Ozone)</pre>
^2
#mse.ozone.all.var.lm <- mean(my aq.test$ozone squaredError all var lm)</pre>
#Build a model by using wind, temp, solar variable
mq_aq.lm.model.wind.temp.solar <- svm(Ozone ~ Wind+Temp+Solar.R, data=my_aq.train)</pre>
#mq aq.model.all.var
# Test the model using test dataset
my_aq.test$ozone_pred_wind_temp_solar_lm <- predict(mq_aq.lm.model.wind.temp.solar,my_aq.test)</pre>
#table(pred,my aq.test$0zone)
# store the prediction results for test dataset
my aq.test.results.wind.temp.solar.lm<- table(my aq.test$ozone pred wind temp solar lm,my aq.tes
t$0zone)
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredErrorwind_temp_solar_lm <- (my_aq.test$ozone_pred_wind_temp_solar_lm-my_</pre>
aq.test$0zone)^2
mse.ozone.wind.temp.solar.lm <- mean(my aq.test$ozone squaredErrorwind temp solar lm)</pre>
#Build a model by using Temp and Wind variable
mq aq.lm.model.wind.temp <- lm(Ozone ~ Wind+Temp, data=my aq.train)
#mq_aq.model.wind.temp
# Test the model using test dataset
my_aq.test$ozone_wind_temp_lm <- predict(mq_aq.lm.model.wind.temp,my_aq.test)</pre>
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.wind.temp.lm <- table(my aq.test$ozone wind temp lm,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError wind temp lm <- (my aq.test$ozone wind temp lm-my aq.test$Ozone)^2
my_aq.test$ozone_Error_wind_temp_lm <- (my_aq.test$ozone_wind_temp_lm-my_aq.test$Ozone)</pre>
mse.ozone.wind.temp.lm <- mean(my_aq.test$ozone_squaredError_wind_temp_lm)</pre>
#Build a model by using Temp variable
mq aq.lm.model.temp <- lm(Ozone ~ Temp, data=my aq.train)
#ma aa.model.temp
# Test the model using test dataset
my aq.test$ozone temp lm <- predict(mq aq.lm.model.temp,my aq.test)</pre>
#table(pred,my_aq.test$0zone)
# store the prediction results for test dataset
my_aq.test.results.temp.lm <- table(my_aq.test$ozone_temp_lm,my_aq.test$Ozone)</pre>
```

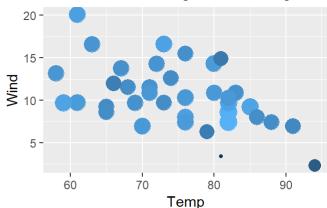
```
# Square of Error - Support Vector Machine
my_aq.test$ozone_squaredError_temp_lm <- (my_aq.test$ozone_temp_lm-my_aq.test$Ozone)^2</pre>
mse.ozone.temp.lm <- mean(my aq.test$ozone squaredError temp lm)</pre>
#Build a model by using Wind variable
mq aq.lm.model.wind <- lm(Ozone ~ Wind, data=my aq.train)
#mq_aq.model.wind
# Test the model using test dataset
my_aq.test$ozone_wind_lm <- predict(mq_aq.lm.model.wind,my_aq.test)</pre>
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.wind.lm <- table(my aq.test$ozone wind lm,my aq.test$Ozone)
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError wind lm <- (my aq.test$ozone wind lm-my aq.test$Ozone)^2
mse.ozone.wind.lm <- mean(my aq.test$ozone squaredError wind lm)</pre>
#Build a model by using Solar.R variable
mq_aq.lm.model.solar <- lm(Ozone ~ Solar.R, data=my_aq.train)</pre>
#mg ag.model.solar
# Test the model using test dataset
my aq.test$ozone solar lm <- predict(mg aq.lm.model.solar,my aq.test)</pre>
#table(pred,my aq.test$Ozone)
# store the prediction results for test dataset
my aq.test.results.solar.lm <- table(my aq.test$ozone solar lm,my aq.test$Ozone)</pre>
# Square of Error - Support Vector Machine
my aq.test$ozone squaredError solar lm <- (my aq.test$ozone solar lm-my aq.test$Ozone)^2
mse.ozone.solar.lm <- mean(my_aq.test$ozone_squaredError_solar_lm)</pre>
#scatter
            chart (using
                             gqplot geom point),
                                                     with
                                                             the x-axis representing
                                                                                          the tem
     the
#y-axis representing
                        the wind, the size and color of each
                                                                      dot representing
                                                                                          the erro
ggplot(my_aq.test) +
  geom point(aes(x=Temp,y=Wind,size=ozone Error wind temp lm,color=ozone Error wind temp lm)) +
 theme(legend.position = "right",axis.text.x = element text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "Ozone Error for regression using LM")
```

Ozone Error for regression using LM



```
ggplot_ksvm <-ggplot(my_aq.test) +</pre>
  geom point(aes(x=Temp,y=Wind,size=ozone Error wind temp,color=ozone Error wind temp)) +
  labs(x = "Temp", y="Wind", size="Error", color="Error", title = "Ozone Error for regression usin
g KSVM") + guides(size=FALSE, color=FALSE)
ggplot_svm <-ggplot(my_aq.test) +</pre>
  geom_point(aes(x=Temp,y=Wind,size=ozone_Error_wind_temp_svm,color=ozone_Error_wind_temp_svm))
  labs(x = "Temp", y="Wind", size="Error", color="Error", title = "Ozone Error for regression using
SVM") + guides(size=FALSE, color=FALSE)
ggplot_lm <-ggplot(my_aq.test) +</pre>
  geom point(aes(x=Temp,y=Wind,size=ozone Error wind temp lm,color=ozone Error wind temp lm)) +
  labs(x = "Temp", y="Wind", size="Error", color="Error", title = "Ozone Error for regression using
LM") + guides(size=FALSE, color=FALSE)
#Arrange all the scatter plots from different models in one Grid
text <-textGrob("This Grid showing the Error plot (predicted - Actual) from various regression m
ethod. Regression method used is KSVM, SVM and LM ")
#leq <- ggplot(my ag.test) + theme(aes(size=ozone Error wind temp lm,color=ozone Error wind temp
_lm),legend.position = "right",axis.text.x = element_text(angle = 90, hjust = 1))
grid.arrange(text, ggplot_ksvm, ggplot_svm, ggplot_lm, ncol=2)
```

Ozone Error for regression using KSVI



- Actual) from various regression method. Regre ⋛ 10

Ozone Error for regression using SVM 201560 70 80 90 Temp

20 -

80

Temp

90

70

Ozone Error for regression using LM

#Calculate Mean Ozone mean ozone <- mean(my aq.train\$Ozone)</pre> my_aq.train\$mean_oz <- mean_ozone</pre> # function to determine wheather the ozone level is good/bad flag_good_ozone <-function(Ozone) {</pre> if(Ozone < mean_ozone) 0 else 1</pre> } flag_pred_good_ozone <-function(goodOzone) {</pre> if(goodOzone < 0.5) 0 else 1</pre> } flag pred good ozone nb <-function(goodOzone) {</pre> if(goodOzone ==1) 0 else 1 } flag pred results <-function(goodOzone,goodOzonePred) {</pre> if(goodOzone == goodOzonePred) 0 else 1 } # Create a varaiable to house good/nbad ozone indicator my_aq.train\$goodOzone <- sapply(my_aq.train\$Ozone,flag_good_ozone)</pre> my_aq.test\$goodOzone <- sapply(my_aq.test\$Ozone,flag_good_ozone)</pre>

60

```
#Build a model by using all variable
mq aq.goz.model.all.var <- ksvm(goodOzone ~ Wind+Temp+Solar.R+ Month, data=my aq.train)
#mq aq.model.all.var
# Test the model using test dataset
my_aq.test$goodOzone_pred_all_var <- predict(mq_aq.goz.model.all.var,my_aq.test)</pre>
my aq.test$goodOzone pred all var <- sapply(my aq.test$goodOzone pred all var,flag pred good ozo
ne)
my aq.test$goodOzone pred all var results <- mapply(flag pred results,my aq.test$goodOzone,my a
q.test$goodOzone pred all var)
#table(pred,my_aq.test$goodOzone)
# store the prediction results for test dataset
#my aq.goz.test.results.all.var <- table(pred,my aq.test$goodOzone)</pre>
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_all_var <- (my_aq.test$goodOzone_pred_all_var-my_aq.test$goodO</pre>
zone)^2
mse.goodOzone.all.var <- mean(my_aq.test$goodOzone_squaredError_all_var)</pre>
# % success in prediction
length(my aq.test[my aq.test[seodOzone pred all var results==0,1])/length(my aq.test[,1])*100
```

[1] 91.89189

```
#Build a model by using wind, temp, solar variable
mq_aq.goz.model.wind.temp.solar <- ksvm(goodOzone ~ Wind+Temp+Solar.R, data=my_aq.train)</pre>
#mq aq.model.all.var
# Test the model using test dataset
my aq.test$goodOzone pred wind temp solar <- predict(mq aq.goz.model.wind.temp.solar,my aq.test)</pre>
my_aq.test$goodOzone_pred_wind_temp_solar <- sapply(my_aq.test$goodOzone_pred_wind_temp_solar,fl</pre>
ag pred good ozone)
my_aq.test$goodOzone_pred_wind_temp_solar_results <- mapply(flag_pred_results,my_aq.test$goodOzo</pre>
ne,my aq.test$goodOzone pred wind temp solar)
#table(pred,my aq.test$good0zone)
# store the prediction results for test dataset
#my aq.qoz.test.results.wind.temp.solar<- table(pred,my aq.test$qood0zone)</pre>
# Square of Error - Support Vector Machine
my aq.test$goodOzone squaredErrorwind temp solar <- (my aq.test$goodOzone pred wind temp solar-m
y aq.test$goodOzone)^2
mse.goodOzone.wind.temp.solar <- mean(my aq.test$goodOzone squaredErrorwind temp solar)</pre>
# % success in prediction
length(my aq.test[my aq.test[seodOzone pred wind temp solar results==0,1])/length(my aq.test[,1
])*100
```

[1] 97.2973

```
#Build a model by using Temp and Wind variable
mq aq.goz.model.wind.temp <- ksvm(goodOzone ~ Wind+Temp, data=my aq.train)
#mq aq.model.wind.temp
# Test the model using test dataset
my_aq.test$goodOzone_wind_temp <- predict(mq_aq.goz.model.wind.temp,my_aq.test)</pre>
my aq.test$goodOzone wind temp <- sapply(my aq.test$goodOzone wind temp,flag pred good ozone)
my_aq.test$goodOzone_pred_wind_temp_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_</pre>
aq.test$goodOzone wind temp)
#table(pred,my aq.test$goodOzone)
# store the prediction results for test dataset
#my aq.qoz.test.results.wind.temp <- table(pred,my aq.test$qood0zone)</pre>
# Square of Error - Support Vector Machine
my aq.test$goodOzone squaredError wind temp <- (my aq.test$goodOzone wind temp-my aq.test$goodOz
one)^2
my aq.test$goodOzone Error wind temp <- (my aq.test$goodOzone wind temp-my aq.test$goodOzone)
mse.goodOzone.wind.temp <- mean(my_aq.test$goodOzone_squaredError_wind_temp)</pre>
# % success in prediction
length(my aq.test[my aq.test[sgoodOzone pred wind temp results==0,1])/length(my aq.test[,1])*100
```

[1] 94.59459

```
#Build a model by using Temp variable
mq_aq.goz.model.temp <- ksvm(goodOzone ~ Temp, data=my_aq.train)</pre>
#mq aq.model.temp
# Test the model using test dataset
my aq.test$goodOzone temp <- predict(mq aq.goz.model.temp,my aq.test)</pre>
my_aq.test$goodOzone_temp <- sapply(my_aq.test$goodOzone_temp,flag_pred_good_ozone)</pre>
my_aq.test$goodOzone_pred_temp_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_aq.te</pre>
st$goodOzone temp)
#table(pred,my aq.test$goodOzone)
# store the prediction results for test dataset
#my_aq.goz.test.results.temp <- table(pred,my_aq.test$good0zone)</pre>
# Square of Error - Support Vector Machine
my aq.test$goodOzone squaredError temp <- (my aq.test$goodOzone temp-my aq.test$goodOzone)^2
mse.goodOzone.temp <- mean(my_aq.test$goodOzone_squaredError_temp)</pre>
# % success in prediction
length(my aq.test[my aq.test[sood0zone pred temp results==0,1])/length(my aq.test[,1])*100
```

[1] 91.89189

```
#Build a model by using Wind variable
mq aq.goz.model.wind <- ksvm(goodOzone ~ Wind, data=my aq.train)
#mg ag.model.wind
# Test the model using test dataset
my_aq.test$goodOzone_wind <- predict(mq_aq.goz.model.wind,my_aq.test)</pre>
my_aq.test$goodOzone_wind <- sapply(my_aq.test$goodOzone_wind,flag_pred_good_ozone)</pre>
my_aq.test$goodOzone_pred_wind_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_aq.te</pre>
st$goodOzone wind)
#table(pred,my aq.test$good0zone)
# store the prediction results for test dataset
#my aq.qoz.test.results.wind <- table(pred,my aq.test$qoodOzone)</pre>
# Square of Error - Support Vector Machine
my aq.test$goodOzone squaredError wind <- (my aq.test$goodOzone wind-my aq.test$goodOzone)^2
mse.goodOzone.wind <- mean(my_aq.test$goodOzone_squaredError_wind)</pre>
# % success in prediction
length(my_aq.test[my_aq.test$good0zone_pred_wind_results==0,1])/length(my_aq.test[,1])*100
```

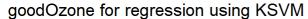
[1] 86.48649

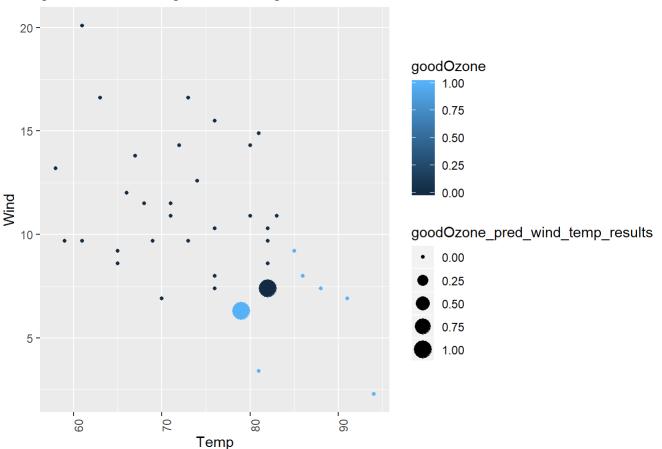
```
#Build a model by using Solar.R variable
mq_aq.goz.model.solar <- ksvm(goodOzone ~ Solar.R, data=my_aq.train)</pre>
#mq_aq.model.solar
# Test the model using test dataset
my_aq.test$goodOzone_solar <- predict(mq_aq.goz.model.solar,my_aq.test)</pre>
my aq.test$goodOzone solar <- sapply(my aq.test$goodOzone solar,flag pred good ozone)</pre>
my_aq.test$goodOzone_pred_solar_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_aq.t</pre>
est$goodOzone solar)
#table(pred,my_aq.test$good0zone)
# store the prediction results for test dataset
#my_aq.goz.test.results.solar <- table(pred,my_aq.test$good0zone)</pre>
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_solar <- (my_aq.test$goodOzone_solar-my_aq.test$goodOzone)^2</pre>
mse.goodOzone.solar <- mean(my_aq.test$goodOzone_squaredError_solar)</pre>
# % success in prediction
length(my_aq.test[my_aq.test[sgoodOzone_pred_solar_results==0,1])/length(my_aq.test[,1])*100
```

[1] 70.27027

```
#scatter chart (using ggplot geom_point), with the x-axis representing the tem
p, the
#y-axis representing the wind, the size and color of each dot representing the erro
r

ggplot(my_aq.test) +
   geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_results,color=goodOzone)) + theme(l
egend.position = "right",axis.text.x = element_text(angle = 90, hjust = 1)) +
   labs(x = "Temp", y="Wind",title = "goodOzone for regression using KSVM")
```





```
library(e1071)
#Build a model by using all variable
mq_aq.goz.svm.model.all.var <- svm(goodOzone ~ Wind+Temp+Solar.R+ Month, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_pred_all_var_svm <- predict(mq_aq.goz.svm.model.all.var,my_aq.test)
my_aq.test$goodOzone_pred_all_var_svm <- sapply(my_aq.test$goodOzone_pred_all_var_svm,flag_pred_
good_ozone)
my_aq.test$goodOzone_pred_all_var_results_svm <- mapply(flag_pred_results,my_aq.test$goodOzone,m
y_aq.test$goodOzone_pred_all_var_svm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_all_var_svm <- (my_aq.test$goodOzone_pred_all_var_svm-my_aq.te
st$goodOzone)^2
mse.goodOzone.all.var.svm <- mean(my_aq.test$goodOzone_squaredError_all_var_svm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_all_var_results_svm==0,1])/length(my_aq.test[,1])*10
0</pre>
```

[1] 91.89189

```
#Build a model by using wind, temp, solar variable
mq aq.goz.svm.model.wind.temp.solar <- svm(goodOzone ~ Wind+Temp+Solar.R, data=my aq.train)
# Test the model using test dataset
my_aq.test$goodOzone_pred_wind_temp_solar_svm <- predict(mq_aq.goz.svm.model.wind.temp.solar,my_</pre>
aq.test)
my_aq.test$goodOzone_pred_wind_temp_solar_svm <- sapply(my_aq.test$goodOzone_pred_wind_temp_sola</pre>
r svm,flag pred good ozone)
my aq.test$goodOzone pred wind temp solar svm results <- mapply(flag pred results,my aq.test$goo
dOzone,my_aq.test$goodOzone_pred_wind_temp_solar_svm)
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredErrorwind_temp_solar_svm <- (my_aq.test$goodOzone_pred_wind_temp_sol</pre>
ar svm-my aq.test$goodOzone)^2
mse.goodOzone.wind.temp.solar_svm <- mean(my_aq.test$goodOzone_squaredErrorwind_temp_solar_svm)</pre>
# % success in prediction
length(my aq.test[my aq.test$good0zone pred wind temp solar svm results==0,1])/length(my aq.test
[,1])*100
```

[1] 97.2973

```
#Build a model by using Temp and Wind variable
mq aq.goz.svm.model.wind.temp <- svm(goodOzone ~ Wind+Temp, data=my aq.train)
# Test the model using test dataset
my_aq.test$goodOzone_wind_temp_svm <- predict(mq_aq.goz.svm.model.wind.temp,my_aq.test)</pre>
my aq.test$goodOzone wind temp svm <- sapply(my aq.test$goodOzone wind temp svm,flag pred good o
zone)
my aq.test$goodOzone pred wind temp svm results <- mapply(flag pred results,my aq.test$goodOzon
e,my_aq.test$goodOzone_wind_temp_svm)
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_temp_svm <- (my_aq.test$goodOzone_wind_temp_svm-my_aq.tes</pre>
t$good0zone)^2
my_aq.test$goodOzone_Error_wind_temp_svm <- (my_aq.test$goodOzone_wind_temp_svm-my_aq.test$goodO
mse.goodOzone.wind.temp_svm <- mean(my_aq.test$goodOzone_squaredError_wind_temp_svm)</pre>
# % success in prediction
length(my aq.test[my aq.test[goodOzone pred wind temp svm results==0,1])/length(my aq.test[,1])*
100
```

[1] 94.59459

```
#Build a model by using Temp variable
mq_aq.goz.svm.model.temp <- svm(goodOzone ~ Temp, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_temp_svm <- predict(mq_aq.goz.svm.model.temp,my_aq.test)
my_aq.test$goodOzone_temp_svm <- sapply(my_aq.test$goodOzone_temp_svm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_temp_svm_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_temp_svm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_temp_svm <- (my_aq.test$goodOzone_temp_svm-my_aq.test$goodOzone
e)^2
mse.goodOzone.temp_svm <- mean(my_aq.test$goodOzone_squaredError_temp_svm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_temp_svm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 91.89189

```
#Build a model by using Wind variable
mq_aq.goz.svm.model.wind <- svm(goodOzone ~ Wind, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_wind_svm <- predict(mq_aq.goz.svm.model.wind,my_aq.test)
my_aq.test$goodOzone_wind_svm <- sapply(my_aq.test$goodOzone_wind_svm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_wind_svm_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_wind_svm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_svm <- (my_aq.test$goodOzone_wind_svm-my_aq.test$goodOzone
e)^2
mse.goodOzone.wind_svm <- mean(my_aq.test$goodOzone_squaredError_wind_svm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_wind_svm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 86.48649

```
#Build a model by using Solar.R variable
mq_aq.goz.svm.model.solar <- svm(goodOzone ~ Solar.R, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_solar_svm <- predict(mq_aq.goz.svm.model.solar,my_aq.test)
my_aq.test$goodOzone_solar_svm <- sapply(my_aq.test$goodOzone_solar_svm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_solar_svm_results <- mapply(flag_pred_results,my_aq.test$goodOzone_solar_svm)

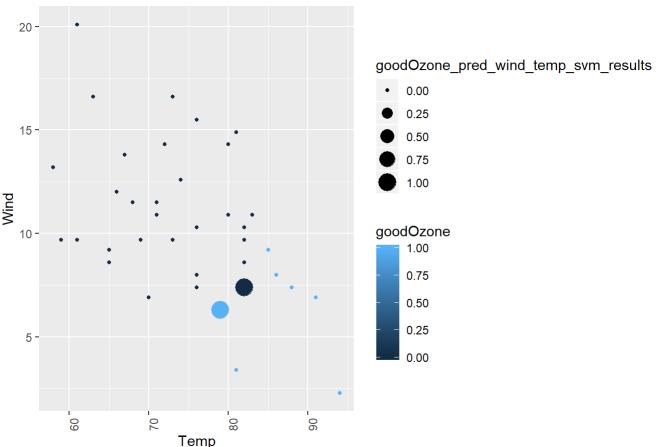
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_solar_svm <- (my_aq.test$goodOzone_solar_svm-my_aq.test$goodOzone)^2
mse.goodOzone.solar_svm <- mean(my_aq.test$goodOzone_squaredError_solar_svm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_solar_svm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 81.08108

```
#scatter
            chart (using
                                                            the x-axis representing
                            gqplot geom point),
                                                    with
                                                                                        the tem
     the
р,
#y-axis representing
                        the wind,
                                    the size and color of each
                                                                    dot representing
                                                                                        the erro
r
ggplot(my aq.test) +
  geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_svm_results,color=goodOzone)) + the
me(legend.position = "right",axis.text.x = element text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "goodOzone for regression using SVM")
```

goodOzone for regression using SVM



```
#Build a model by using all variable
mq_aq.goz.lm.model.all.var <- lm(goodOzone ~ Wind+Temp+Solar.R+ Month, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_pred_all_var_lm <- predict(mq_aq.goz.lm.model.all.var,my_aq.test)
my_aq.test$goodOzone_pred_all_var_lm <- sapply(my_aq.test$goodOzone_pred_all_var_lm,flag_pred_go
od_ozone)
my_aq.test$goodOzone_pred_all_var_results_lm <- mapply(flag_pred_results,my_aq.test$goodOzone,my
aq.test$goodOzone_pred_all_var_lm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_all_var_lm <- (my_aq.test$goodOzone_pred_all_var_lm-my_aq.test
$goodOzone)^2
mse.goodOzone.all.var.lm <- mean(my_aq.test$goodOzone_squaredError_all_var_lm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_all_var_results_lm==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 83.78378

```
#Build a model by using wind, temp, solar variable
mq aq.goz.lm.model.wind.temp.solar <- lm(goodOzone ~ Wind+Temp+Solar.R, data=my aq.train)
# Test the model using test dataset
my_aq.test$goodOzone_pred_wind_temp_solar_lm <- predict(mq_aq.goz.lm.model.wind.temp.solar,my_a</pre>
q.test)
my_aq.test$goodOzone_pred_wind_temp_solar_lm <- sapply(my_aq.test$goodOzone_pred_wind_temp_solar_</pre>
lm,flag pred good ozone)
my aq.test$goodOzone pred wind temp solar lm results <- mapply(flag pred results,my aq.test$good
Ozone, my_aq.test$goodOzone_pred_wind_temp_solar_lm)
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredErrorwind_temp_solar_lm <- (my_aq.test$goodOzone_pred_wind_temp_sola</pre>
r_lm-my_aq.test$goodOzone)^2
mse.goodOzone.wind.temp.solar_lm <- mean(my_aq.test$goodOzone_squaredErrorwind_temp_solar_lm)</pre>
# % success in prediction
length(my aq.test[my aq.test$goodOzone pred wind temp solar lm results==0,1])/length(my aq.test
[,1])*100
```

[1] 91.89189

```
#Build a model by using Temp and Wind variable
mq aq.goz.lm.model.wind.temp <- lm(goodOzone ~ Wind+Temp, data=my aq.train)
# Test the model using test dataset
my_aq.test$goodOzone_wind_temp_lm <- predict(mq_aq.goz.lm.model.wind.temp,my_aq.test)</pre>
my aq.test$goodOzone wind temp lm <- sapply(my aq.test$goodOzone wind temp lm,flag pred good ozo
ne)
my aq.test$goodOzone pred wind temp lm results <- mapply(flag pred results,my aq.test$goodOzone,
my_aq.test$goodOzone_wind_temp_lm)
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_temp_lm <- (my_aq.test$goodOzone_wind_temp_lm-my_aq.test</pre>
$goodOzone)^2
my_aq.test$goodOzone_Error_wind_temp_lm <- (my_aq.test$goodOzone_wind_temp_lm-my_aq.test$goodOzo</pre>
mse.goodOzone.wind.temp_lm <- mean(my_aq.test$goodOzone_squaredError_wind_temp_lm)</pre>
# % success in prediction
length(my aq.test[my aq.test[sood0zone pred wind temp lm results==0,1])/length(my aq.test[,1])*1
00
```

[1] 94.59459

```
#Build a model by using Temp variable
mq_aq.goz.lm.model.temp <- lm(goodOzone ~ Temp, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_temp_lm <- predict(mq_aq.goz.lm.model.temp,my_aq.test)
my_aq.test$goodOzone_temp_lm <- sapply(my_aq.test$goodOzone_temp_lm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_temp_lm_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_temp_lm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_temp_lm <- (my_aq.test$goodOzone_temp_lm-my_aq.test$goodOzone)
^2
mse.goodOzone.temp_lm <- mean(my_aq.test$goodOzone_squaredError_temp_lm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_temp_lm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 91.89189

```
#Build a model by using Wind variable
mq_aq.goz.lm.model.wind <- lm(goodOzone ~ Wind, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_wind_lm <- predict(mq_aq.goz.lm.model.wind,my_aq.test)
my_aq.test$goodOzone_wind_lm <- sapply(my_aq.test$goodOzone_wind_lm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_wind_lm_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_wind_lm)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_lm <- (my_aq.test$goodOzone_wind_lm-my_aq.test$goodOzone)
^2
mse.goodOzone.wind_lm <- mean(my_aq.test$goodOzone_squaredError_wind_lm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_wind_lm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

```
## [1] 86.48649
```

```
#Build a model by using Solar.R variable
mq_aq.goz.lm.model.solar <- lm(goodOzone ~ Solar.R, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_solar_lm <- predict(mq_aq.goz.lm.model.solar,my_aq.test)
my_aq.test$goodOzone_solar_lm <- sapply(my_aq.test$goodOzone_solar_lm,flag_pred_good_ozone)
my_aq.test$goodOzone_pred_solar_lm_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_solar_lm)

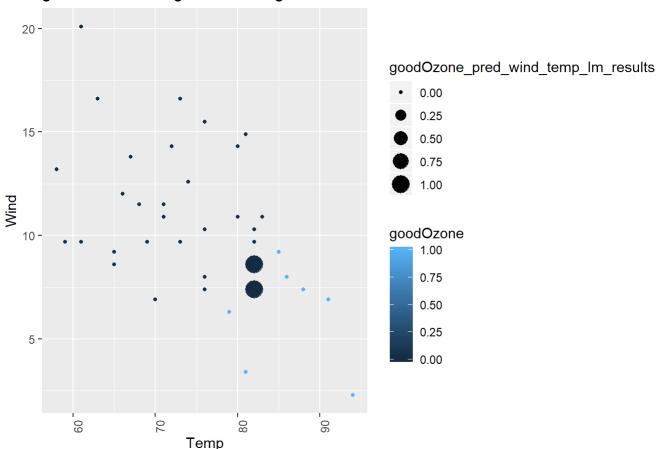
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_solar_lm <- (my_aq.test$goodOzone_solar_lm-my_aq.test$goodOzon
e)^2
mse.goodOzone.solar_lm <- mean(my_aq.test$goodOzone_squaredError_solar_lm)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_solar_lm_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 70.27027

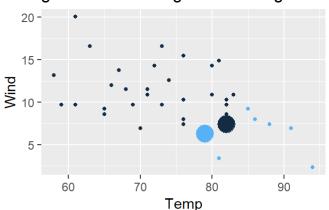
```
#scatter
            chart (using
                            gqplot geom point),
                                                    with
                                                            the x-axis representing
                                                                                        the tem
     the
р,
#y-axis representing
                        the wind,
                                    the size and color of each
                                                                    dot representing
                                                                                        the erro
r
ggplot(my aq.test) +
  geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_lm_results,color=goodOzone)) + them
e(legend.position = "right",axis.text.x = element text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "goodOzone for regression using LM")
```

goodOzone for regression using LM



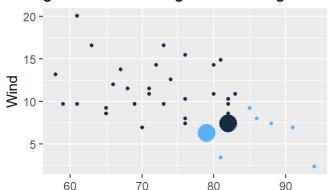
```
ggplot_ksvm_goodOzone <-ggplot(my_aq.test) +</pre>
  geom point(aes(x=Temp,y=Wind,size=goodOzone pred wind temp results,color=goodOzone)) +
  labs(x = "Temp", y="Wind",title = "goodOzone for regression using KSVM")+ guides(size=FALSE, c
olor=FALSE)
ggplot_svm_goodOzone <-ggplot(my_aq.test) +</pre>
  geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_svm_results,color=goodOzone)) +
  labs(x = "Temp", y="Wind", title = "goodOzone for regression using SVM") + guides(size=FALSE, c
olor=FALSE)
ggplot_lm_goodOzone <-ggplot(my_aq.test) +</pre>
  geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_lm_results,color=goodOzone)) +
  labs(x = "Temp", y="Wind",title = "goodOzone for regression using LM") + guides(size=FALSE, co
lor=FALSE)
#Arrange all the scatter plots from different models in one Grid
text <-textGrob("This Grid showing the accuracy of the goodOzone prediction between KSVM, SVM an
d LM ")
#leg <- ggplot(my_aq.test) + theme(aes(size=ozone_Error_wind_temp_lm,color=ozone_Error_wind_temp</pre>
lm), legend.position = "right", axis.text.x = element text(angle = 90, hjust = 1))
grid.arrange(text, ggplot_ksvm_good0zone, ggplot_svm_good0zone, ggplot_lm_good0zone, ncol=2)
```

goodOzone for regression using KSVN



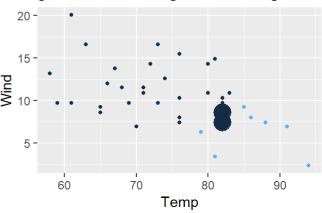
accuracy of the goodOzone prediction between

goodOzone for regression using SVM



Temp

goodOzone for regression using LM



library(e1071)

#Build a model by using all variable

mq_aq.goz.nb.model.all.var <- naiveBayes(goodOzone ~ Wind+Temp+Solar.R+Month, data=my_aq.train)</pre>

Test the model using test dataset

my_aq.test\$goodOzone_pred_all_var_nb <- predict(mq_aq.goz.nb.model.all.var,my_aq.test[,1:5],type
="raw")</pre>

my_aq.test\$goodOzone_pred_all_var_nb <- sapply(max.col(my_aq.test\$goodOzone_pred_all_var_nb,"las
t"),flag pred good ozone nb)</pre>

my_aq.test\$goodOzone_pred_all_var_results_nb <- mapply(flag_pred_results,my_aq.test\$goodOzone,my_aq.test\$goodOzone_pred_all_var_nb)</pre>

Square of Error - Support Vector Machine

my_aq.test\$goodOzone_squaredError_all_var_nb <- (my_aq.test\$goodOzone_pred_all_var_nb-my_aq.test \$goodOzone)^2

mse.goodOzone.all.var.nb <- mean(my_aq.test\$goodOzone_squaredError_all_var_nb)</pre>

% success in prediction

length(my_aq.test[my_aq.test[sood0zone_pred_all_var_results_nb==0,1])/length(my_aq.test[,1])*100

[1] 91.89189

```
#Build a model by using wind, temp, solar variable
mq aq.goz.nb.model.wind.temp.solar <- naiveBayes(goodOzone ~ Wind+Temp+Solar.R, data=my aq.trai
n)
# Test the model using test dataset
my aq.test$goodOzone pred wind temp solar nb <- predict(mq aq.goz.nb.model.wind.temp.solar,my a
q.test[,2:4],type="raw")
my aq.test$goodOzone pred wind temp solar nb <- sapply(max.col(my aq.test$goodOzone pred wind te
mp solar nb,"last"),flag pred good ozone nb)
my_aq.test$goodOzone_pred_wind_temp_solar_nb_results <- mapply(flag_pred_results,my_aq.test$good
Ozone, my aq.test$goodOzone pred wind temp solar nb)
# Square of Error - Support Vector Machine
my aq.test$goodOzone squaredErrorwind temp solar nb <- (my aq.test$goodOzone pred wind temp sola</pre>
r nb-my aq.test$goodOzone)^2
mse.goodOzone.wind.temp.solar nb <- mean(my aq.test$goodOzone squaredErrorwind temp solar nb)</pre>
# % success in prediction
length(my aq.test[my aq.test$goodOzone pred wind temp solar nb results==0,1])/length(my aq.test
[,1])*100
```

[1] 97.2973

```
#Build a model by using Temp and Wind variable
mq aq.goz.nb.model.wind.temp <- naiveBayes(goodOzone ~ Wind+Temp, data=my aq.train)</pre>
# Test the model using test dataset
my aq.test$goodOzone wind temp nb <- predict(mg aq.goz.nb.model.wind.temp,my aq.test[,3:4],type=
"raw")
my aq.test$goodOzone wind temp nb <- sapply(max.col(my aq.test$goodOzone wind temp nb, "last"),fl
ag pred good ozone nb)
my_aq.test$goodOzone_pred_wind_temp_nb_results <- mapply(flag_pred_results,my_aq.test$goodOzone,</pre>
my aq.test$goodOzone wind temp nb)
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_temp_nb <- (my_aq.test$goodOzone_wind_temp_nb-my_aq.test</pre>
$goodOzone)^2
my_aq.test$goodOzone_Error_wind_temp_nb <- (my_aq.test$goodOzone_wind_temp_nb-my_aq.test$goodOzo</pre>
ne)
mse.goodOzone.wind.temp_nb <- mean(my_aq.test$goodOzone_squaredError_wind_temp_nb)</pre>
# % success in prediction
length(my aq.test[my aq.test[sood0zone pred wind temp nb results==0,1])/length(my aq.test[,1])*1
00
```

[1] 94.59459

```
#Build a model by using Temp variable
mq_aq.goz.nb.model.temp <- naiveBayes(goodOzone ~ Temp, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_temp_nb <- predict(mq_aq.goz.nb.model.temp,my_aq.test[,4],type="raw")</pre>
```

```
## Warning in predict.naiveBayes(mq_aq.goz.nb.model.temp, my_aq.test[, 4], :
## Type mismatch between training and new data for variable 'Temp'. Did you
## use factors with numeric labels for training, and numeric values for new
## data?
```

```
my_aq.test$goodOzone_temp_nb <- sapply(max.col(my_aq.test$goodOzone_temp_nb, "last"),flag_pred_go
od_ozone_nb)
my_aq.test$goodOzone_pred_temp_nb_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_temp_nb)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_temp_nb <- (my_aq.test$goodOzone_temp_nb-my_aq.test$goodOzone)
^2
mse.goodOzone.temp_nb <- mean(my_aq.test$goodOzone_squaredError_temp_nb)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_temp_nb_results==0,1])/length(my_aq.test[,1])*100</pre>
```

```
## [1] 81.08108
```

```
#Build a model by using Wind variable
mq_aq.goz.nb.model.wind <- naiveBayes(goodOzone ~ Wind, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_wind_nb <- predict(mq_aq.goz.nb.model.wind,my_aq.test[,3],type="raw")</pre>
```

```
## Warning in predict.naiveBayes(mq_aq.goz.nb.model.wind, my_aq.test[, 3], :
## Type mismatch between training and new data for variable 'Wind'. Did you
## use factors with numeric labels for training, and numeric values for new
## data?
```

```
my_aq.test$goodOzone_wind_nb <- sapply(max.col(my_aq.test$goodOzone_wind_nb,"last"),flag_pred_go
od_ozone_nb)
my_aq.test$goodOzone_pred_wind_nb_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_wind_nb)

# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_wind_nb <- (my_aq.test$goodOzone_wind_nb-my_aq.test$goodOzone)
^2
mse.goodOzone.wind_nb <- mean(my_aq.test$goodOzone_squaredError_wind_nb)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_wind_nb_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 81.08108

```
#Build a model by using Solar.R variable
mq_aq.goz.nb.model.solar <- naiveBayes(goodOzone ~ Solar.R, data=my_aq.train)

# Test the model using test dataset
my_aq.test$goodOzone_solar_nb <- predict(mq_aq.goz.nb.model.solar,my_aq.test[,2],type="raw")</pre>
```

```
## Warning in predict.naiveBayes(mq_aq.goz.nb.model.solar, my_aq.test[, 2], :
## Type mismatch between training and new data for variable 'Solar.R'. Did you
## use factors with numeric labels for training, and numeric values for new
## data?
```

```
my_aq.test$goodOzone_solar_nb <- sapply(max.col(my_aq.test$goodOzone_solar_nb, "last"),flag_pred_
good_ozone_nb)
my_aq.test$goodOzone_pred_solar_nb_results <- mapply(flag_pred_results,my_aq.test$goodOzone,my_a
q.test$goodOzone_solar_nb)

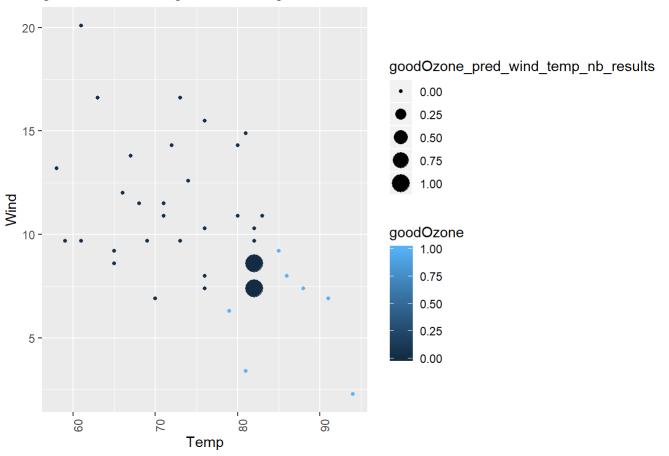
# Square of Error - Support Vector Machine
my_aq.test$goodOzone_squaredError_solar_nb <- (my_aq.test$goodOzone_solar_nb-my_aq.test$goodOzone
e)^2
mse.goodOzone.solar_nb <- mean(my_aq.test$goodOzone_squaredError_solar_nb)

# % success in prediction
length(my_aq.test[my_aq.test$goodOzone_pred_solar_nb_results==0,1])/length(my_aq.test[,1])*100</pre>
```

[1] 81.08108

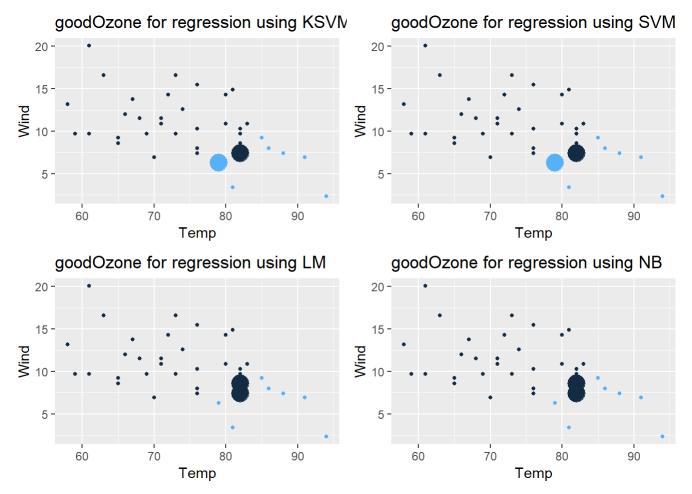
```
#scatter
            chart (using
                            ggplot geom_point),
                                                                                        the tem
                                                    with
                                                            the x-axis representing
     the
р,
#y-axis representing
                        the wind,
                                    the size and color of each
                                                                    dot representing
                                                                                        the erro
ggplot(my_aq.test) +
  geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_nb_results,color=goodOzone)) + them
e(legend.position = "right",axis.text.x = element text(angle = 90, hjust = 1)) +
  labs(x = "Temp", y="Wind",title = "goodOzone for regression using NB")
```

goodOzone for regression using NB



```
ggplot_nb_goodOzone <- ggplot(my_aq.test) +
   geom_point(aes(x=Temp,y=Wind,size=goodOzone_pred_wind_temp_nb_results,color=goodOzone)) +
   labs(x = "Temp", y="Wind",title = "goodOzone for regression using NB") + guides(size=FALSE, co
lor=FALSE)

# Now Looking at the difference for all modeLs
grid.arrange( ggplot_ksvm_goodOzone, ggplot_svm_goodOzone, ggplot_lm_goodOzone,ggplot_nb_goodOzone, ncol=2)</pre>
```



After comparing all the models to predict good and bad ozone . It appears that KSVM and SVM mo del is giving better results with predicting good and bad Ozone