

Assignment-3

3_1.

a. Implementing linear kernel model:

Type 'q()' to quit R.

[Previously saved workspace restored]

```
> library('e1071')
Warning message:
package 'e1071' was built under R version 3.4.2
> kweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kweather<-kweath[,2:9]
> head(kweather)
  Temp.F Dew_Point.F Humidity.percentage Sea_Level_Press.in Visibility.mi Wind.mph Precip.in Events
1     26           12                73           30.19           5           9      0.03   Snow
2     31           18                68           29.95           7          11      0.01   Snow
3     10            1                63           30.24           5          14      0.02   Snow
4     38           35                90           29.70           6           5      0.00   Rain
5     40           30                75           29.80           9           7      0.00   Rain
6     49           29                51           29.64          10          10      0.00   Rain
> n=366
> nt=290
> neval=n-nt
> rep=100
> accuracy=dim(rep)
> precision_Rain=dim(rep)
> Recall_Rain=dim(rep)
> precision_Rain_Thunderstorm=dim(rep)
> Recall_Rain_Thunderstorm=dim(rep)
> precision_Snow=dim(rep)
> Recall_Snow=dim(rep)
> for(k in 1:rep)
+ {
+   Tkweather = sample(1:n,nt)
+   kweather.Train = kweather[Tkweather,]
+   kweather.Test = kweather[-Tkweather,]
+   svmfit = svm(Events~.,data=kweather.Train,kernel='linear',cost=100)
+   p=predict(svmfit,kweather.Test)
+   cmatrix = table(p,kweather.Test$Events)
+   accuracy[k] = sum(diag(cmatrix))/sum(cmatrix)
```

```

+ svmfit = svm(Events~.,data=kcweather.Train,kernel='linear',cost=100)
+ p=predict(svmfit,kcweather.Test)
+ cmatrix = table(p,kcweather.Test$Events)
+ accuracy[k] = sum(diag(cmatrix))/sum(cmatrix)
+ precision_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[2,1]+cmatrix[3,1])
+ precision_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[1,2]+cmatrix[2,2]+cmatrix[3,2])
+ precision_Snow[k] = cmatrix[3,3]/(cmatrix[1,3]+cmatrix[2,3]+cmatrix[3,3])
+ Recall_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[1,2]+cmatrix[1,3])
+ Recall_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[2,1]+cmatrix[2,2]+cmatrix[2,3])
+ Recall_Snow[k] = cmatrix[3,3]/(cmatrix[3,1]+cmatrix[3,2]+cmatrix[3,3])
+ }
> summary(svmfit)

```

Call:

```
svm(formula = Events ~ ., data = kcweather.Train, kernel = "linear", cost = 100)
```

Parameters:

```

SVM-Type: C-classification
SVM-Kernel: linear
cost: 100
gamma: 0.1428571

```

Number of Support Vectors: 139

```
( 60 69 10 )
```

Number of Classes: 3

Levels:

```
Rain Rain_Thunderstorm Snow
```

```

> mean(accuracy)
[1] 0.7746053
> mean(precision_Rain)
[1] 0.7266815
> mean(Recall_Rain)
[1] 0.7853335
> mean(precision_Rain_Thunderstorm)
[1] 0.7819636
> mean(Recall_Rain_Thunderstorm)
[1] 0.7224564
> mean(precision_Snow)
[1] 0.9205422
> mean(Recall_Snow)
[1] 0.9038673
> AccySR_svm_err = qt(0.975, df = rep-1) * sd(accuracy) / sqrt(rep)
> mean(accuracy) - AccySR_svm_err; mean(accuracy) + AccySR_svm_err
[1] 0.7655562
[1] 0.7836544
> summary(accuracy)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6711 0.7500 0.7763 0.7746 0.8026 0.8947
> summary(precision_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5333 0.6828 0.7222 0.7267 0.7635 0.9333
> summary(Recall_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5455 0.7385 0.7858 0.7853 0.8287 0.9643
> summary(precision_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5652 0.7333 0.7795 0.7820 0.8409 0.9667
> summary(Recall_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5429 0.6667 0.7188 0.7225 0.7803 0.9310
> summary(precision_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6667 0.8643 0.9231 0.9205 1.0000 1.0000
> summary(Recall_Snow)

```

```

> summary(precision_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6667 0.8643 0.9231 0.9205 1.0000 1.0000
> summary(Recall_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6667 0.8571 0.9091 0.9039 1.0000 1.0000
> svm_tune <- tune(svm,
+   Events~.,data=kweather.Train,kernel='linear', ranges=list(cost=10^(-1:2), gamma=c(.5,1,2)))
> summary(svm_tune)

Parameter tuning of 'svm':

- sampling method: 10-fold cross validation

- best parameters:
  cost gamma
    10   0.5

- best performance: 0.2275862

- Detailed performance results:
  cost gamma    error dispersion
1    0.1   0.5 0.2551724 0.05677742
2    1.0   0.5 0.2586207 0.08004821
3   10.0   0.5 0.2275862 0.06337494
4  100.0   0.5 0.2379310 0.07350932
5    0.1   1.0 0.2551724 0.05677742
6    1.0   1.0 0.2586207 0.08004821
7   10.0   1.0 0.2275862 0.06337494
8  100.0   1.0 0.2379310 0.07350932
9    0.1   2.0 0.2551724 0.05677742
10   1.0   2.0 0.2586207 0.08004821
11  10.0   2.0 0.2275862 0.06337494
12 100.0   2.0 0.2379310 0.07350932

> |

```

Inference:

The total number of support vectors are 139 with three classes.

The svm tuning results shows the best parameters are with cost 10 and gamma=0.5

b. Implementing Radial Kernel:

```
> library('e1071')
> kcweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kcweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kcweather<-kcweath[,2:9]# kcweather$Events<-as.character(kcweather$Events)
> n=366
> nt=290
> neval=n-nt
> rep=100
> accuracy=dim(rep)
> precision_Rain=dim(rep)
> Recall_Rain=dim(rep)
> precision_Rain_Thunderstorm=dim(rep)
> Recall_Rain_Thunderstorm=dim(rep)
> precision_Snow=dim(rep)
> Recall_Snow=dim(rep)
>
> for(k in 1:rep)
+ {
+   Tkcweather = sample(1:n,nt)
+   kcweather.Train = kcweather[Tkcweather,]
+   kcweather.Test = kcweather[-Tkcweather,]
+   svmfit = svm(Events~.,data=kcweather.Train,kernel='radial',cost=100)
+   p=predict(svmfit,kcweather.Test)
+   cmatrix = table(p,kcweather.Test$Events)
+   accuracy[k] = sum(diag(cmatrix))/sum(cmatrix)
+   precision_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[2,1]+cmatrix[3,1])
+   precision_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[1,2]+cmatrix[2,2]+cmatrix[3,2])
+   precision_Snow[k] = cmatrix[3,3]/(cmatrix[1,3]+cmatrix[2,3]+cmatrix[3,3])
+   Recall_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[1,2]+cmatrix[1,3])
+   Recall_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[2,1]+cmatrix[2,2]+cmatrix[2,3])
+   Recall_Snow[k] = cmatrix[3,3]/(cmatrix[3,1]+cmatrix[3,2]+cmatrix[3,3])
+ }
> summary(svmfit)

Call:
svm(formula = Events ~ ., data = kcweather.Train, kernel = "radial", cost = 100)
|
```



```
Call:
svm(formula = Events ~ ., data = kcweather.Train, kernel = "radial", cost = 100)
```

```
Parameters:
  SVM-Type:  C-classification
  SVM-Kernel: radial
      cost:  100
   gamma:   0.1428571
```

```
Number of Support Vectors: 144
```

```
( 69 57 18 )
```

```
Number of Classes: 3
```

```
Levels:
Rain Rain_Thunderstorm Snow
```

```
> mean(accuracy)
[1] 0.7559211
> mean(precision_Rain)
[1] 0.7557304
> mean(Recall_Rain)
[1] 0.7415518
> mean(precision_Rain_Thunderstorm)
[1] 0.7376141
> mean(Recall_Rain_Thunderstorm)
[1] 0.7417431
> mean(precision_Snow)
[1] 0.8211042
> mean(Recall_Snow)
[1] 0.8614867
> AccySR_svm_err = qt(0.975, df = rep-1) * sd(accuracy) / sqrt(rep)
```

```

> AccySR_svm_err = qt(0.975, df = rep-1) * sd(accuracy) / sqrt(rep)
> mean(accuracy) - AccySR_svm_err; mean(accuracy) + AccySR_svm_err
[1] 0.746472
[1] 0.7653701
> summary(accuracy)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6447  0.7237  0.7566  0.7559  0.7895  0.8553
> summary(precision_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5714  0.7037  0.7436  0.7557  0.8058  0.9412
> summary(Recall_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5526  0.6911  0.7432  0.7416  0.7895  0.9091
> summary(precision_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5000  0.6936  0.7419  0.7376  0.7863  0.9167
> summary(Recall_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5758  0.6772  0.7407  0.7417  0.8014  0.9545
> summary(precision_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.4000  0.7500  0.8258  0.8211  0.9091  1.0000
> summary(Recall_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5000  0.7885  0.8889  0.8615  0.9464  1.0000
>
> svm_tune <- tune(svm,Events~.,data=kcweather.Train,kernel='radial', ranges=list(cost=10^(-1:2), gamma=c(.5,1,2)))
> summary(svm_tune)

Parameter tuning of 'svm':

- sampling method: 10-fold cross validation

- best parameters:
  cost gamma
    1    0.5

> summary(Recall_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5000  0.7885  0.8889  0.8615  0.9464  1.0000
>
> svm_tune <- tune(svm,Events~.,data=kcweather.Train,kernel='radial', ranges=list(cost=10^(-1:2), gamma=c(.5,1,2)))
> summary(svm_tune)

Parameter tuning of 'svm':

- sampling method: 10-fold cross validation

- best parameters:
  cost gamma
    1    0.5

- best performance: 0.2482759

- Detailed performance results:
  cost gamma  error dispersion
1    0.1    0.5  0.3862069  0.03914801
2    1.0    0.5  0.2482759  0.07761813
3   10.0    0.5  0.2655172  0.07457990
4  100.0    0.5  0.3068966  0.08821418
5    0.1    1.0  0.5206897  0.05254772
6    1.0    1.0  0.2517241  0.09057880
7   10.0    1.0  0.2931034  0.10690273
8  100.0    1.0  0.3103448  0.10534657
9    0.1    2.0  0.5241379  0.04539866
10   1.0    2.0  0.3068966  0.08969937
11  10.0    2.0  0.3103448  0.09337975
12 100.0    2.0  0.3206897  0.09759968

```

The number of support vectors are 144 with the best parameters cost=1, gamma=0.5

c. Let the svm choose the kernel:

```
> library('e1071')
> kcweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kcweath<-read.csv(file="G:/Fall Semester 2017/ISL/Assignment-2/kc_weather_srt.csv",head=T,sep=",")
> kcweather<-kcweath[,2:9]
> head(kcweather)
  Temp.F Dew_Point.F Humidity.percentage Sea_Level_Press.in Visibility.mi Wind.mph Precip.in Events
1    26         12           73         30.19           5         9      0.03    Snow
2    31         18           68         29.95           7        11      0.01    Snow
3    10          1           63         30.24           5        14      0.02    Snow
4    38         35           90         29.70           6         5      0.00    Rain
5    40         30           75         29.80           9         7      0.00    Rain
6    49         29           51         29.64          10        10      0.00    Rain
> n=366
> nt=290
> neval=n-nt
> rep=100
> accuracy=dim(rep)
> precision_Rain=dim(rep)
> Recall_Rain=dim(rep)
> precision_Rain_Thunderstorm=dim(rep)
> Recall_Rain_Thunderstorm=dim(rep)
> precision_Snow=dim(rep)
> Recall_Snow=dim(rep)
> |
> for(k in 1:rep)
+ {
+   Tkweather = sample(1:n,nt)
+   kcweather.Train = kcweather[Tkweather,]
+   kcweather.Test = kcweather[-Tkweather,]
+   svmfit = svm(Events~.,data=kcweather.Train)
+   p=predict(svmfit,kcweather.Test)
+   cmatrix = table(p,kcweather.Test$Events)
+   accuracy[k] = sum(diag(cmatrix))/sum(cmatrix)
+   precision_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[2,1]+cmatrix[3,1])
+   precision_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[1,2]+cmatrix[2,2]+cmatrix[3,2])
+   precision_Snow[k] = cmatrix[3,3]/(cmatrix[1,3]+cmatrix[2,3]+cmatrix[3,3])
+   Recall_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[1,2]+cmatrix[1,3])
+   Recall_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[2,1]+cmatrix[2,2]+cmatrix[2,3])
+   Recall_Snow[k] = cmatrix[3,3]/(cmatrix[3,1]+cmatrix[3,2]+cmatrix[3,3])
+ }
> summary(svmfit)
```

Call:
svm(formula = Events ~ ., data = kcweather.Train)

Parameters:
SVM-Type: C-classification
SVM-Kernel: radial
cost: 1
gamma: 0.1428571


```

> for(k in 1:rep)
+ {
+   Tkweather = sample(1:n,nt)
+   kweather.Train = kweather[Tkweather,]
+   kweather.Test = kweather[-Tkweather,]
+   svmfit = svm(Events~.,data=kweather.Train)
+   p=predict(svmfit,kweather.Test)
+   cmatrix = table(p,kweather.Test$Events)
+   accuracy[k] = sum(diag(cmatrix))/sum(cmatrix)
+   precision_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[2,1]+cmatrix[3,1])
+   precision_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[1,2]+cmatrix[2,2]+cmatrix[3,2])
+   precision_Snow[k] = cmatrix[3,3]/(cmatrix[1,3]+cmatrix[2,3]+cmatrix[3,3])
+   Recall_Rain[k] = cmatrix[1,1]/(cmatrix[1,1]+cmatrix[1,2]+cmatrix[1,3])
+   Recall_Rain_Thunderstorm[k] = cmatrix[2,2]/(cmatrix[2,1]+cmatrix[2,2]+cmatrix[2,3])
+   Recall_Snow[k] = cmatrix[3,3]/(cmatrix[3,1]+cmatrix[3,2]+cmatrix[3,3])
+ }
> summary(svmfit)

```

Call:

```
svm(formula = Events ~ ., data = kweather.Train)
```

Parameters:

```

  SVM-Type:  C-classification
  SVM-Kernel: radial
      cost:  1
  gamma:  0.1428571

```

Number of Support Vectors: 178

(87 71 20)

Number of Classes: 3

Levels:

Rain Rain_Thunderstorm Snow

```

> mean(accuracy)
[1] 0.7843421
> mean(precision_Rain)
[1] 0.7820518
> mean(Recall_Rain)
[1] 0.7739886
> mean(precision_Rain_Thunderstorm)
[1] 0.7497093
> mean(Recall_Rain_Thunderstorm)
[1] 0.7664871
> mean(precision_Snow)
[1] 0.8949755
> mean(Recall_Snow)
[1] 0.879679
> AccySR_svm_err = qt(0.975, df = rep-1) * sd(accuracy) / sqrt(rep)
> mean(accuracy) - AccySR_svm_err; mean(accuracy) + AccySR_svm_err

```

```

> summary(accuracy)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6842 0.7500 0.7895 0.7843 0.8158 0.9079
> summary(precision_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6000 0.7434 0.7910 0.7821 0.8235 0.9444
> summary(Recall_Rain)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6154 0.7365 0.7714 0.7740 0.8112 0.9474
> summary(precision_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5161 0.7000 0.7500 0.7497 0.7956 0.9583
> summary(Recall_Rain_Thunderstorm)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.5926 0.7241 0.7649 0.7665 0.8148 0.9500
> summary(precision_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6000 0.8333 0.9091 0.8950 1.0000 1.0000
> summary(Recall_Snow)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.6667 0.8000 0.8889 0.8797 0.9559 1.0000
> svm_tune <- tune(svm,
+                  Events~.,data=kcweather.Train)
> summary(svm_tune)

Error estimation of 'svm' using 10-fold cross validation: 0.1931034

```

Here we could see that the accuracy is 0.784. with the total number of svms=178.

Inference:

From the previous assignment comparing the values of different models like LDA, QDA, KNN with SVM, we understand that SVM is a better model than KNN and all others.