

# justiphone\_variables

[Code ▾](#)[Hide](#)

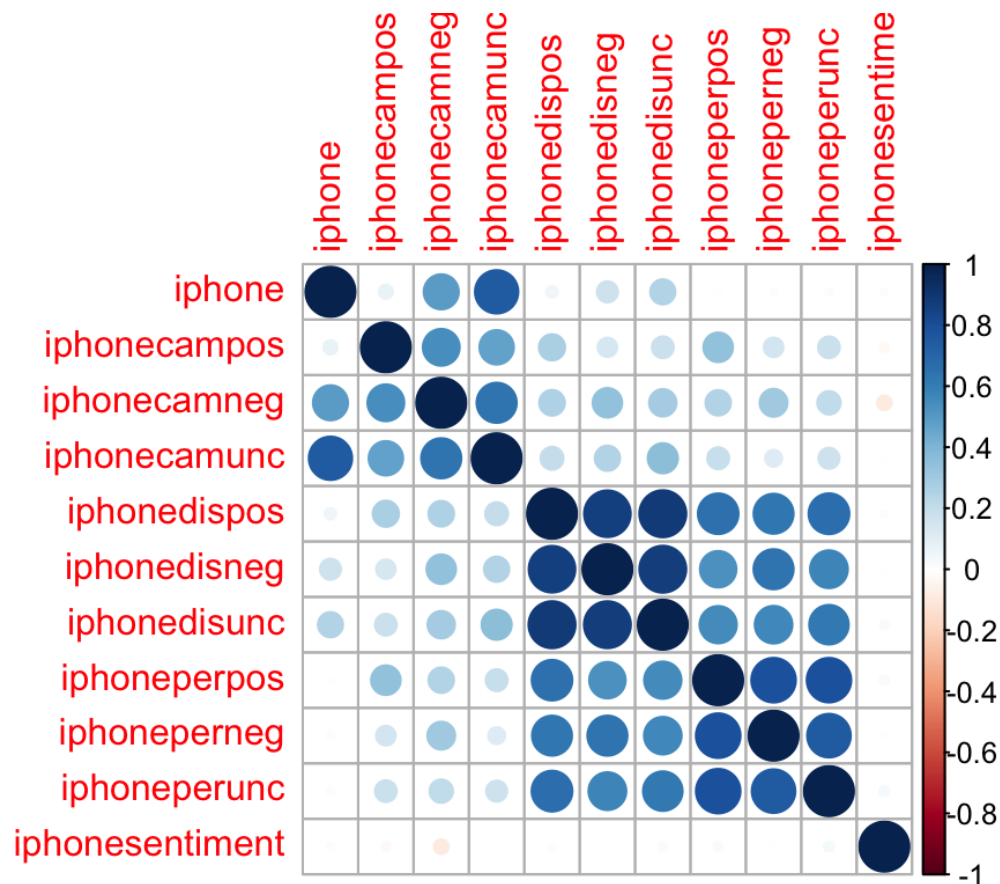
```
# keep only variables for iphones
toFilter <- grepl('iphone', colnames(iphone_smallMatrix))
iphones <- iphone_smallMatrix[toFilter]
colnames(iphones)
```

```
[1] "iphone"          "iphonecampos"    "iphonecamneg"    "iphonecamunc"    "iphonedispos"
[6] "iphonedisneg"    "iphonedisunc"    "iphoneperpos"    "iphoneperneg"    "iphoneperunc"
[11] "iphonesentiment"
```

## Correlation

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```
corr_matrix <- cor(iphones)
corr_plot <- corrplot(as.matrix(corr_matrix))
```

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corr\_plot

	iphone	iphonecampos	iphonecamneg	iphonecamunc	iphonedispos	iphonedisneg
isneg						
iphone 72621	1.000000000	0.07815733	0.49052359	0.750403174	0.05262462	0.1755
iphonecampos 50674	0.078157326	1.00000000	0.54133997	0.473266316	0.27258655	0.1486
iphonecamneg 78956	0.490523588	0.54133997	1.00000000	0.643460020	0.26198314	0.3468
iphonecamunc 53711	0.750403174	0.47326632	0.64346002	1.00000000	0.20900762	0.2532
iphonedispos 65387	0.052624621	0.27258655	0.26198314	0.209007616	1.00000000	0.8687
iphonedisneg 00000	0.175572621	0.14865067	0.34687896	0.253253711	0.86876539	1.0000
iphonedisunc 50578	0.250929821	0.18831003	0.29907429	0.361321734	0.88302623	0.8799
iphoneperpos 88336	-0.009507666	0.34833242	0.25756896	0.190248578	0.65935383	0.5308
iphoneperneg 95104	0.013863107	0.15191863	0.30887521	0.113175498	0.63776843	0.6409
iphoneperunc 44418	-0.016037424	0.18725962	0.21757939	0.174433158	0.66523752	0.5700
iphonesentiment 44905	0.014858654	-0.02973122	-0.08396314	0.001443485	0.01454682	0.0031

	iphonedisunc	iphoneperpos	iphoneperneg	iphoneperunc	iphonesentiment
iphone	0.25092982	-0.009507666	0.013863107	-0.01603742	0.014858654
iphonecampos	0.18831003	0.348332416	0.151918629	0.18725962	-0.029731217
iphonecamneg	0.29907429	0.257568960	0.308875213	0.21757939	-0.083963139
iphonecamunc	0.36132173	0.190248578	0.113175498	0.17443316	0.001443485
iphonedispos	0.88302623	0.659353827	0.637768430	0.66523752	0.014546824
iphonedisneg	0.87995058	0.530888336	0.640995104	0.57004442	0.003144905
iphonedisunc	1.00000000	0.554364879	0.564479458	0.62392944	0.027172723
iphoneperpos	0.55436488	1.000000000	0.794832452	0.79182763	0.029637900
iphoneperneg	0.56447946	0.794832452	1.000000000	0.75948372	-0.004804058
iphoneperunc	0.62392944	0.791827630	0.759483720	1.00000000	0.037199859
iphonesentiment	0.02717272	0.029637900	-0.004804058	0.03719986	1.000000000

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```
# run this for any features that are h
any_over_80 <- function(my_matrix) any(my_matrix > .8 & my_matrix < 1, na.rm = TRUE)
any_under_80 <- function(my_matrix) any(my_matrix < -.8 & my_matrix > -1, na.rm = TRUE)
```

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```
# remove features with collinearity, correlation greater than .8, FOR small corr_matrix
corr_matrix %>%
  focus_if(any_over_80, mirror = TRUE)
```

rowname <chr>	iphonedispos <dbl>	iphonedisneg <dbl>	iphonedisunc <dbl>
iphonedispos	NA	0.8687654	0.8830262
iphonedisneg	0.8687654	NA	0.8799506
iphonedisunc	0.8830262	0.8799506	NA
3 rows			

Let's drop these variables: iphonedispos, iphonedisneg, iphonedisunc

Remove columns 5,6,7

Not all models are affected by collinearity

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```
# keep columns only for iphone and samsunggalaxy
iphones_corr <- iphones[,-(5:7)]
```

## NZR

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#nearZeroVar() with saveMetrics = TRUE returns an object containing a table including: frequency ratio, percentage unique, zero variance and near zero variance

```
nzvMetrics <- nearZeroVar(iphones, saveMetrics = TRUE)
nzvMetrics
```

	freqRatio <dbl>	percentUnique <dbl>	zeroVar <lgl>	nzv <lgl>
iphone	5.041322	0.2081246	FALSE	FALSE
iphonecampos	10.524697	0.2312495	FALSE	FALSE
iphonecamneg	19.517529	0.1310414	FALSE	TRUE
iphonecamunc	16.764205	0.1618747	FALSE	FALSE
iphonedispos	6.792440	0.2466662	FALSE	FALSE
iphonedisneg	10.084428	0.1849996	FALSE	FALSE
iphonedisunc	11.471875	0.2081246	FALSE	FALSE
iphoneperpos	9.297834	0.1927079	FALSE	FALSE
iphoneperneg	11.054137	0.1695830	FALSE	FALSE
iphoneperunc	13.018349	0.1233331	FALSE	FALSE
1-10 of 11 rows			Previous	1 2 Next

[Hide](#)

```
# returns column 2, iphonecamunc, same as nvzMetrics
# nearZeroVar() with saveMetrics = FALSE returns an vector
nvz <- nearZeroVar(iphones, saveMetrics = FALSE)
nvz
```

```
[1] 3
```

[Hide](#)

```
# create a new data set and remove near zero variance features
iphones_nvz <- iphones[,-nvz]
str(iphones_nvz)
```

```
'data.frame': 12973 obs. of 10 variables:
 $ iphone      : int  1 1 1 1 1 41 1 1 1 1 ...
 $ iphonecampos : int  0 0 0 0 0 1 1 0 0 0 ...
 $ iphonecamunc : int  0 0 0 0 0 7 1 0 0 0 ...
 $ iphonedispos : int  0 0 0 0 0 1 13 0 0 0 ...
 $ iphonedisneg : int  0 0 0 0 0 3 10 0 0 0 ...
 $ iphonedisunc : int  0 0 0 0 0 4 9 0 0 0 ...
 $ iphoneperpos : int  0 1 0 1 1 0 5 3 0 0 ...
 $ iphoneperneg : int  0 0 0 0 0 0 4 1 0 0 ...
 $ iphoneperunc : int  0 0 0 1 0 0 5 0 0 0 ...
 $ iphonesentiment: Factor w/ 6 levels "0","1","2","3",...: 1 1 1 1 1 5 5 1 1 1 ...
```

visualize variable with nvz

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```
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphone, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonecampos, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonecamneg, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonecamunc, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonecamneg, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonedispos, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonedisneg, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonedisunc, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphoneperpos, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphoneperneg, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphoneperunc, type='histogram')
plot_ly(iphone_smallMatrix, x= ~iphone_smallMatrix$iphonesentiment, type='histogram')
```

#rfe

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```
# Let's sample the data before using RFE
iphoneSample <- iphones[sample(1:nrow(iphones), 1000, replace=FALSE),]

# Set up rfeControl with randomforest, repeated cross validation and no updates
ctrl <- rfeControl(functions = rfFuncs,
                    method = "repeatedcv",
                    repeats = 5,
                    verbose = FALSE)

# Use rfe and omit the response variable (attribute 11 iphonesentiment)
rfeResultsSMALL <- rfe(iphoneSample[,1:10],
                       iphoneSample$iphonesentiment,
                       sizes=(1:10),
                       rfeControl=ctrl)

# Get results
rfeResultsSMALL
```

Recursive feature selection

Outer resampling method: Cross-Validated (10 fold, repeated 5 times)

Resampling performance over subset size:

	<b>Variables</b>	<b>RMSE</b>	<b>Rsquared</b>	<b>MAE</b>	<b>RMSESD</b>	<b>RsquaredSD</b>	<b>MAESD</b>	<b>Selected</b>
	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>	<S3: AsIs>
1	1	1.567	0.2675	1.198	0.1407	0.10288	0.10058	
2	2	1.547	0.2876	1.203	0.1420	0.10489	0.09385	
3	3	1.555	0.2834	1.227	0.1359	0.10263	0.08864	
4	4	1.563	0.2811	1.239	0.1347	0.10285	0.08785	
5	5	1.572	0.2753	1.250	0.1333	0.10264	0.08771	
6	6	1.552	0.2815	1.182	0.1425	0.09936	0.09356	
7	7	1.550	0.2833	1.180	0.1439	0.09888	0.09569	
8	8	1.546	0.2872	1.179	0.1414	0.09685	0.09200	
9	9	1.541	0.2920	1.154	0.1407	0.09595	0.09551	
10	10	1.540	0.2929	1.157	0.1425	0.09848	0.09683	*

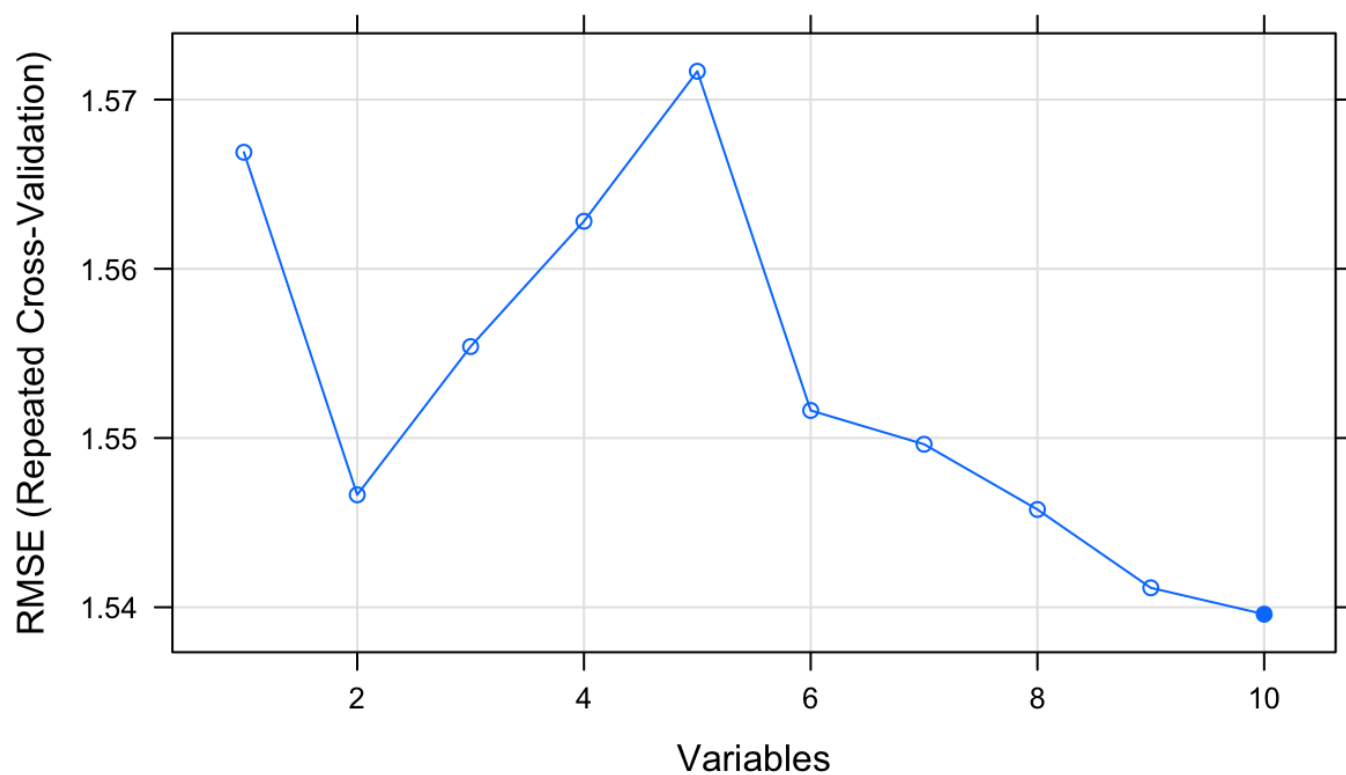
1-10 of 10 rows

The top 5 variables (out of 10):

iphone, iphonecamneg, iphoneperunc, iphonedisneg, iphonedisunc

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```
# Plot results
plot(rfeResultsSMALL, type=c("g", "o"))
```

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```
# create new data set with rfe recommended features
iphones_RFE <- iphones[,predictors(rfeResultsSMALL)]

# add the dependent variable to iphoneRFE
iphones_RFE$iphonesentiment <- iphones$iphonesentiment

# review outcome
str(iphones_RFE)
```

```
'data.frame': 12973 obs. of 11 variables:
 $ iphone      : int  1 1 1 1 1 41 1 1 1 1 ...
 $ iphonecamneg : int  0 0 0 0 0 3 1 0 0 0 ...
 $ iphoneperunc : int  0 0 0 1 0 0 5 0 0 0 ...
 $ iphonedisneg : int  0 0 0 0 0 3 10 0 0 0 ...
 $ iphonedisunc : int  0 0 0 0 0 4 9 0 0 0 ...
 $ iphonedispos : int  0 0 0 0 0 1 13 0 0 0 ...
 $ iphoneperneg : int  0 0 0 0 0 0 4 1 0 0 ...
 $ iphonecampos : int  0 0 0 0 0 1 1 0 0 0 ...
 $ iphoneperpos : int  0 1 0 1 1 0 5 3 0 0 ...
 $ iphonecamunc : int  0 0 0 0 0 7 1 0 0 0 ...
 $ iphonesentiment: int  0 0 0 0 0 4 4 0 0 0 ...
```

# Model for Regular Data: Iphones

[Hide](#)

```
# convert variable types, categorical
iphones$iphonesentiment <- as.factor(iphones$iphonesentiment)
```

Train and Test Set:

[Hide](#)

```
# Create Train and Test Set for iphoneDFBig
# create 75% sample of row indices
in_training <- createDataPartition(iphones$iphonesentiment, p = .7, list = FALSE)
# create 75% sample of data and save it to trainData
trainData_iphones <- iphones[in_training, ]
# create 25% sample of data and save it to test_data
testData_iphones <- iphones[-in_training, ]
# verify split percentages
nrow(trainData_iphones) / nrow(iphones)
```

```
[1] 0.7001465
```

[Hide](#)

```
#c5
c5_iphones <- train(iphonesentiment ~., data = trainData_iphones, method = "C5.0",
  trControl = fitControl)
```

[Hide](#)

```
# randomforest
rf_iphones <- train(iphonesentiment ~., data = trainData_iphones, method = "rf",
  trControl = fitControl)
```

[Hide](#)

```
# svm (kernlab)
svm_iphones <- train(iphonesentiment ~., data = trainData_iphones, method = "svmLinear",
  trControl = fitControl)
```

[Hide](#)

```
# kkn
kkn_iphones <- train(iphonesentiment ~., data = trainData_iphones, method = "kkn",
  trControl = fitControl)
```

[Hide](#)

```
# gbm
#gbm_iphones <- train(iphonesentiment ~., data = trainData_iphones, method = "gbm",
#                      trControl = fitControl)
```

Compare Accuracy on Prediction Results:

Hide

```
#c5
prediction_c5_iphones <- predict(c5_iphones, testData_iphones)
postResample(prediction_c5_iphones, testData_iphones$iphonesentiment)
```

Accuracy	Kappa
0.7167095	0.4224468

Hide

```
#randomforest
prediction_rf_iphones <- predict(rf_iphones, testData_iphones)
postResample(prediction_rf_iphones, testData_iphones$iphonesentiment)
```

Accuracy	Kappa
0.7226221	0.4350340

Hide

```
#svm
prediction_svm_iphones <- predict(svm_iphones, testData_iphones)
postResample(prediction_svm_iphones, testData_iphones$iphonesentiment)
```

Accuracy	Kappa
0.60385604	0.09009688

Hide

```
# kknk
prediction_kknn_iphones <- predict(kknn_iphones, testData_iphones)
postResample(prediction_kknn_iphones, testData_iphones$iphonesentiment)
```

Accuracy	Kappa
0.3017995	0.1179396

Hide

```
modelData_iphones <- resamples(list(C50 = c5_iphones, randomForest = rf_iphones, svmLine
ar = svm_iphones,kknn = kknn_iphones))
```

Hide



```
summary(modelData_iphones)
```

Call:

```
summary.resamples(object = modelData_iphones)
```

Models: C50, randomForest, svmLinear, kkn

Number of resamples: 10

Accuracy

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.6993392	0.7215974	0.7306211	0.7246528	0.7332593	0.7337734	0
randomForest	0.7183718	0.7228969	0.7272729	0.7276227	0.7297732	0.7436744	0
svmLinear	0.5984598	0.6075358	0.6112330	0.6100399	0.6126547	0.6211454	0
kkn	0.2797357	0.2845692	0.2990626	0.2981422	0.3071429	0.3241455	0

Kappa

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.37825534	0.4353286	0.4557795	0.4418250	0.4578631	0.4674848	0
randomForest	0.42670894	0.4380634	0.4467810	0.4484161	0.4505314	0.4863522	0
svmLinear	0.07693399	0.1001085	0.1071164	0.1072959	0.1135268	0.1434714	0
kkn	0.09471739	0.1099662	0.1199002	0.1196086	0.1278466	0.1513078	0

## Model for Correlated Data:

Hide

```
# convert variable types, categorical
iphones_corr$iphonesentiment <- as.factor(iphones_corr$iphonesentiment)
```

Train and Test Set:

Hide

```
# Create Train and Test Set for iphoneDFBig
# create 75% sample of row indices
in_training <- createDataPartition(iphones_corr$iphonesentiment, p = .7, list = FALSE)
# create 75% sample of data and save it to trainData
trainData_iphones_corr <- iphones_corr[in_training, ]
# create 25% sample of data and save it to test_data
testData_iphones_corr <- iphones_corr[-in_training, ]
# verify split percentages
nrow(trainData_iphones_corr) / nrow(iphones_corr)
```

```
[1] 0.7001465
```

Hide

```
#c5
c5_iphones_corr <- train(iphonesentiment ~., data = trainData_iphones_corr, method = "C
5.0",
                        trControl = fitControl)
```

[Hide](#)

```
# randomforest
rf_iphones_corr <- train(iphonesentiment ~., data = trainData_iphones_corr, method = "r
f",
                        trControl = fitControl)
```

No:

[Hide](#)

```
# svm    (kernlab)
svm_iphones_corr <- train(iphonesentiment ~., data = trainData_iphones_corr, method = "s
vmLinear",
                        trControl = fitControl)

# knn
kkn_iphones_corr <- train(iphonesentiment ~., data = trainData_iphones_corr, method =
"knn",
                        trControl = fitControl)

# gbm
gbm_iphones_corr <- train(iphonesentiment ~., data = trainData_iphones_corr, method =
"gbm",
                        trControl = fitControl)
```

Compare Accuracy on Prediction Results:

[Hide](#)

```
#c5
prediction_c5_iphones_corr <- predict(c5_iphones_corr, testData_iphones_corr)
postResample(prediction_c5_iphones_corr, testData_iphones_corr$iphonesentiment)
```

Accuracy	Kappa
0.6861183	0.3389545

[Hide](#)

```
#randomforest
prediction_rf_iphones_corr <- predict(rf_iphones_corr, testData_iphones_corr)
postResample(prediction_rf_iphones_corr, testData_iphones_corr$iphonesentiment)
```

Accuracy	Kappa
0.6884319	0.3455847

No:

[Hide](#)

```
#svm
prediction_svm_iphones_corr <- predict(svm_iphones_corr, testData_iphones_corr)
postResample(prediction_svm_iphones_corr, testData_iphones_corr$iphonesentiment)
# kknn
prediction_kknn_iphones_corr <- predict(kknn_iphones_corr, testData_iphones_corr)
postResample(prediction_kknn_iphones_corr, testData_iphones_corr$iphonesentiment)
```

[Hide](#)

```
modelData_iphones_corr <- resamples(list(C50 = c5_iphones_corr, randomForest = rf_iphone
s_corr))

# svMLinear = svm_iphones_corr, kknn = kknn_iphones_corr))
```

[Hide](#)

```
summary(modelData_iphones_corr)
```

```
Call:
summary.resamples(object = modelData_iphones_corr)
```

```
Models: C50, randomForest
Number of resamples: 10
```

Accuracy

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.6828194	0.6864297	0.6919692	0.6913998	0.6936418	0.7015419	0
randomForest	0.6824697	0.6835304	0.6870177	0.6930564	0.7044070	0.7106711	0

Kappa

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.3252032	0.3357033	0.3479342	0.3486942	0.3587904	0.3774182	0
randomForest	0.3275320	0.3382338	0.3438833	0.3576174	0.3828510	0.4027938	0

## Model for NZR Data:

[Hide](#)

```
# convert variable types, categorical
iphones_nvz$iphonesentiment <- as.factor(iphones_nvz$iphonesentiment)
```

Train and Test Set:

[Hide](#)

```
# Create Train and Test Set for iphoneDFBig
# create 75% sample of row indices
in_training <-createDataPartition(iphones_nvz$iphonesentiment, p = .7, list = FALSE)
# create 75% sample of data and save it to trainData
trainData_iphones_nvz <- iphones_nvz[in_training, ]
# create 25% sample of data and save it to test_data
testData_iphones_nvz <- iphones_nvz[-in_training, ]
# verify split percentages
nrow(trainData_iphones_nvz) / nrow(iphones_nvz)
```

```
[1] 0.7001465
```

[Hide](#)

```
#c5
c5_iphones_nvz <- train(iphonesentiment ~., data = trainData_iphones_nvz, method = "C5.0",
                        trControl = fitControl)
```

[Hide](#)

```
# randomforest
rf_iphones_nvz <- train(iphonesentiment ~., data = trainData_iphones_nvz, method = "rf",
                        trControl = fitControl)
```

No:

[Hide](#)

```
# svm (kernlab)
svm_iphones_nvz <- train(iphonesentiment ~., data = trainData_iphones_nvz, method = "svm
Linear",
                        trControl = fitControl)

# kknn
kknn_iphones_nvz <- train(iphonesentiment ~., data = trainData_iphones_nvz, method = "kk
nn",
                        trControl = fitControl)

# gbm
#gbm_iphones_nvz <- train(iphonesentiment ~., data = trainData_iphones_nvz, method = "gb
m",
#                        trControl = fitControl)
```

Compare Accuracy on Prediction Results:

[Hide](#)

```
#c5
prediction_c5_iphones_nvz <- predict(c5_iphones_nvz, testData_iphones_nvz)
postResample(prediction_c5_iphones_nvz, testData_iphones_nvz$iphonesentiment)
```

Accuracy      Kappa  
0.7239075 0.4427599

Hide

```
#randomforest  
prediction_rf_iphones_nvz <- predict(rf_iphones_nvz, testData_iphones_nvz)  
postResample(prediction_rf_iphones_nvz, testData_iphones_nvz$iphonesentiment)
```

Accuracy      Kappa  
0.7293059 0.4502492

No:

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```
#svm  
prediction_svm_iphones_nvz <- predict(svm_iphones_nvz, testData_iphones_nvz)  
postResample(prediction_svm_iphones_nvz, testData_iphones_nvz$iphonesentiment)  
# kkn  
prediction_kknn_iphones_nvz <- predict(kknn_iphones_nvz, testData_iphones_nvz)  
postResample(prediction_kknn_iphones_nvz, testData_iphones_nvz$iphonesentiment)
```

Hide

```
modelData_iphones_nvz <- resamples(list(C50 = c5_iphones_nvz, randomForest = rf_iphones_  
nvz))  
  
# svMLinear = svm_iphones_nvz,kknn = kknn_iphones_nvz))
```

Hide

```
summary(modelData_iphones_nvz)
```

Call:  
summary.resamples(object = modelData\_iphones\_nvz)

Models: C50, randomForest  
Number of resamples: 10

Accuracy

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.6993392	0.7166850	0.7238745	0.7226659	0.7318482	0.7386990	0
randomForest	0.6949339	0.7206163	0.7225277	0.7247615	0.7331127	0.7535754	0

Kappa

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.3851717	0.4245681	0.4389983	0.4383170	0.4608489	0.4750914	0
randomForest	0.3653074	0.4289318	0.4392082	0.4421312	0.4653474	0.5125422	0

# Models for RFE Data:

[Hide](#)

```
# convert variable types, categorical
iphones_RFE$iphonesentiment <- as.factor(iphones_RFE$iphonesentiment)
```

Train and Test Set:

[Hide](#)

```
# Create Train and Test Set for iphoneDFBig
# create 75% sample of row indices
in_training <- createDataPartition(iphones_RFE$iphonesentiment, p = .7, list = FALSE)
# create 75% sample of data and save it to trainData
trainData_iphones_RFE <- iphones_RFE[in_training, ]
# create 25% sample of data and save it to test_data
testData_iphones_RFE <- iphones_RFE[-in_training, ]
# verify split percentages
nrow(trainData_iphones_RFE) / nrow(iphones_RFE)
```

```
[1] 0.7001465
```

[Hide](#)

```
#c5
c5_iphones_RFE <- train(iphonesentiment ~., data = trainData_iphones_RFE, method = "C5.0",
                        trControl = fitControl)
```

[Hide](#)

```
# randomforest
rf_iphones_RFE <- train(iphonesentiment ~., data = trainData_iphones_RFE, method = "rf",
                        trControl = fitControl)
```

No:

[Hide](#)

```
# svm (kernlab)
svm_iphones_RFE <- train(iphonesentiment ~., data = trainData_iphones_RFE, method = "svm
Linear",
                        trControl = fitControl)

# kkn
kkn_iphones_RFE <- train(iphonesentiment ~., data = trainData_iphones_RFE, method = "kk
nn",
                        trControl = fitControl)

# gbm
#gbm_iphones_RFE <- train(iphonesentiment ~., data = trainData_iphones_RFE, method = "gb
m",
#                        trControl = fitControl)
```

Compare Accuracy on Prediction Results:

Hide

```
#c5
prediction_c5_iphones_RFE <- predict(c5_iphones_RFE, testData_iphones_RFE)
postResample(prediction_c5_iphones_RFE, testData_iphones_RFE$iphonesentiment)
```

Accuracy	Kappa
0.7269923	0.4465710

Hide

```
#randomforest
prediction_rf_iphones_RFE <- predict(rf_iphones_RFE, testData_iphones_RFE)
postResample(prediction_rf_iphones_RFE, testData_iphones_RFE$iphonesentiment)
```

Accuracy	Kappa
0.7295630	0.4506718

No:

Hide

```
#svm
prediction_svm_iphones_RFE <- predict(svm_iphones_RFE, testData_iphones_RFE)
postResample(prediction_svm_iphones_RFE, testData_iphones_RFE$iphonesentiment)
# kkn
prediction_kkn_iphones_RFE <- predict(kkn_iphones_RFE, testData_iphones_RFE)
postResample(prediction_kkn_iphones_RFE, testData_iphones_RFE$iphonesentiment)
```

Hide

```
modelData_iphones_RFE <- resamples(list(C50 = c5_iphones_RFE, randomForest = rf_iphones_
RFE))

# svMLinear = svm_iphones_RFE,kkn = kkn_iphones_RFE))
```

```
summary(modelData_iphones_RFE)
```

Call:

```
summary.resamples(object = modelData_iphones_RFE)
```

Models: C50, randomForest

Number of resamples: 10

Accuracy

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.6945976	0.7148753	0.7224670	0.7227723	0.7346563	0.7414741	0
randomForest	0.7051705	0.7178285	0.7222237	0.7250933	0.7320295	0.7477974	0

Kappa

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
C50	0.3669234	0.4215689	0.4399212	0.4385360	0.4653061	0.4829802	0
randomForest	0.3959714	0.4221486	0.4339812	0.4424809	0.4593732	0.5002932	0