

Building Multi-Variate Models on London BikeSharing Data

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```
library(alr4)
```

```
## Loading required package: car
```

```
## Loading required package: carData
```

```
## Loading required package: effects
```

```
## Registered S3 methods overwritten by 'lme4':  
##   method                      from  
##   cooks.distance.influence.merMod car  
##   influence.merMod             car  
##   dfbeta.influence.merMod      car  
##   dfbetas.influence.merMod     car
```

```
## lattice theme set by effectsTheme()  
## See ?effectsTheme for details.
```

```
library(leaps)  
library(MASS)  
library(pls)
```

```
##  
## Attaching package: 'pls'
```

```
## The following object is masked from 'package:stats':  
##  
##   loadings
```

```
library(trafo)
```

```
## Registered S3 method overwritten by 'pryr':  
##   method      from  
##   print.bytes Rcpp
```

```
##  
## Attaching package: 'trafo'
```

```
## The following object is masked from 'package:MASS':  
##  
##      boxcox
```

```
library(VGAM)
```

```
## Loading required package: stats4
```

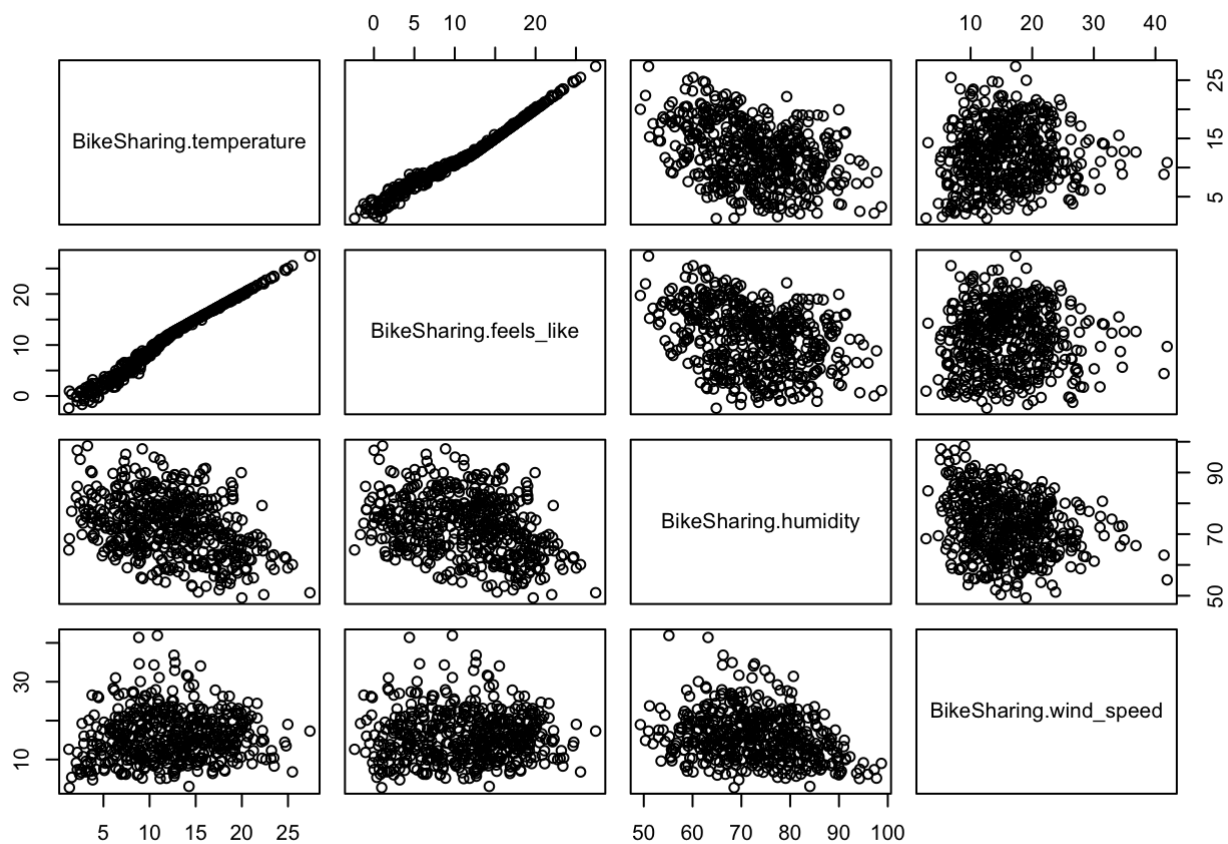
```
## Loading required package: splines
```

```
##  
## Attaching package: 'VGAM'
```

```
## The following object is masked from 'package:trafo':  
##  
##      reciprocal
```

```
## The following object is masked from 'package:car':  
##  
##      logit
```

```
BikeSharing <- read.delim("bikesharing18.txt",sep = "")  
pairs(data.frame(BikeSharing$temperature,BikeSharing$feels_like,BikeSharing$humidity,BikeSharing$wind_speed))
```



```
m1 = lm(N_bikes~temperature+feels_like+humidity+wind_speed+holiday+weekend+season,data
= BikeSharing)
summary(m1)
```

```
##
## Call:
## lm(formula = N_bikes ~ temperature + feels_like + humidity +
##     wind_speed + holiday + weekend + season, data = BikeSharing)
##
## Residuals:
```

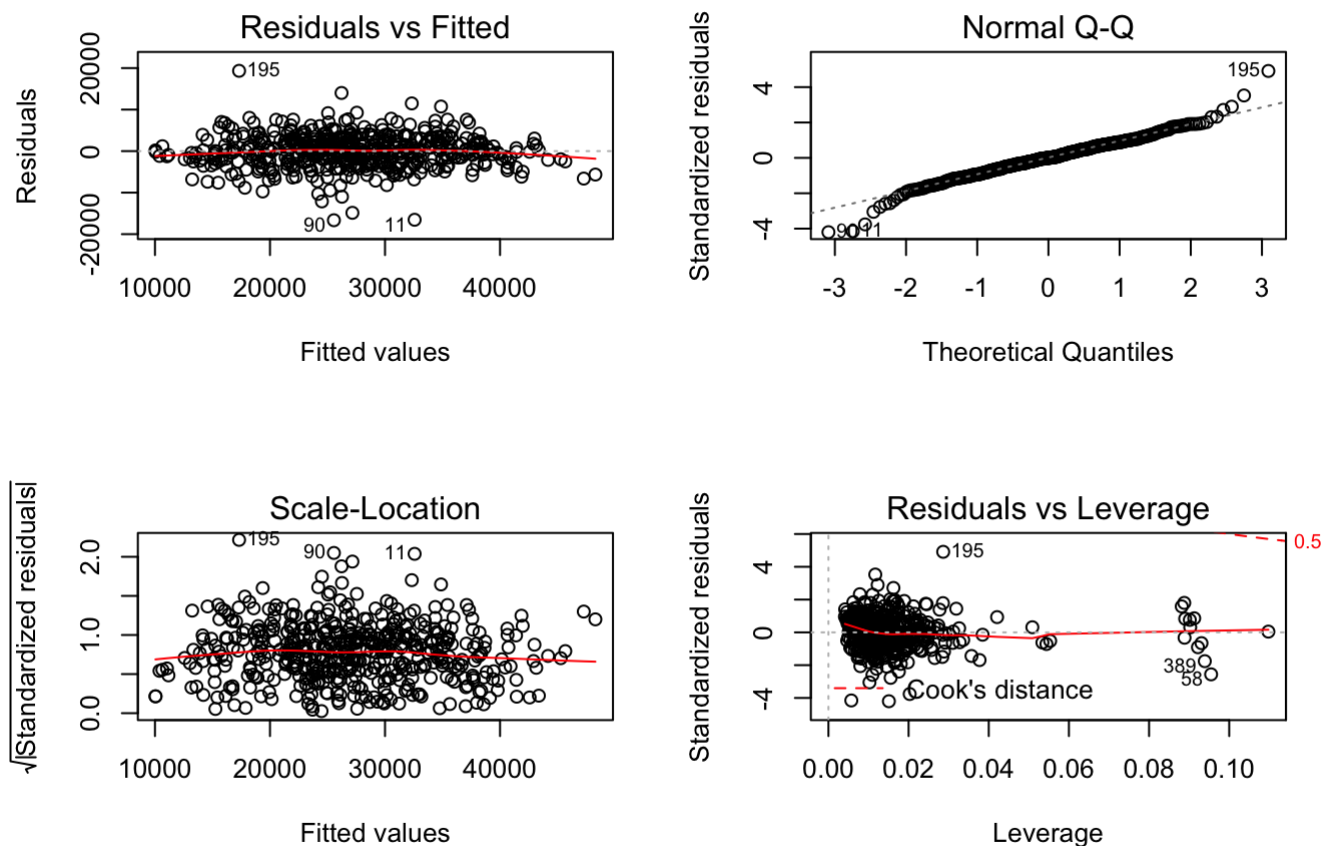
	Min	1Q	Median	3Q	Max
	-16648.7	-2430.3	7.2	2610.3	19338.3

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	51893.71	2175.84	23.850	< 2e-16 ***
temperature	226.95	321.35	0.706	0.4804
feels_like	545.91	265.94	2.053	0.0406 *
humidity	-363.12	22.68	-16.008	< 2e-16 ***
wind_speed	-357.92	32.07	-11.162	< 2e-16 ***
holiday	-8153.36	1178.70	-6.917	1.44e-11 ***
weekend	-5237.73	400.70	-13.072	< 2e-16 ***
season	-87.24	183.91	-0.474	0.6354

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3994 on 492 degrees of freedom
## Multiple R-squared:  0.7722, Adjusted R-squared:  0.7689
## F-statistic: 238.2 on 7 and 492 DF,  p-value: < 2.2e-16
```

```
par(mfrow=c(2,2))
plot(m1)
```



```
m_all = lm(N_bikes~feels_like+humidity+wind_speed+holiday+weekend+feels_like*holiday+feels_like*weekend+wind_speed*holiday+humidity*holiday+humidity*weekend+wind_speed*weekend,
data = BikeSharing)
Anova(m_all,type='II')
```

	Sum Sq <dbl>	Df <dbl>	F value <dbl>	Pr(>F) <dbl>
feels_like	8984527787	1	590.69052053	4.316673e-86
humidity	5224559626	1	343.49026663	1.894593e-58
wind_speed	2292817818	1	150.74200696	2.204192e-30
holiday	822036833	1	54.04506240	8.323053e-13
weekend	2713945757	1	178.42919184	6.602646e-35
feels_like:holiday	2559828	1	0.16829668	6.818101e-01
feels_like:weekend	135348499	1	8.89852839	2.996902e-03
wind_speed:holiday	1065859	1	0.07007519	7.913391e-01
humidity:holiday	54368322	1	3.57446194	5.926661e-02
humidity:weekend	76597248	1	5.03590944	2.527455e-02

1-10 of 12 rows

Previous 1 2 Next

Best Subset Selection

```
m.select = regsubsets(N_bikes~temperature+feels_like+humidity+wind_speed+holiday+weekend+
season+
                        temperature*holiday+temperature*weekend+temperature*season+
                        feels_like*holiday+feels_like*weekend+feels_like*season+
                        wind_speed*holiday+wind_speed*weekend+wind_speed*season+
                        humidity*holiday+humidity*weekend+humidity*season,data=BikeSharin
g,nvmax=20)
reg.summary = summary(m.select)
reg.summary
```

```

## Subset selection object
## Call: regsubsets.formula(N_bikes ~ temperature + feels_like + humidity +
##      wind_speed + holiday + weekend + season + temperature * holiday +
##      temperature * weekend + temperature * season + feels_like *
##      holiday + feels_like * weekend + feels_like * season + wind_speed *
##      holiday + wind_speed * weekend + wind_speed * season + humidity *
##      holiday + humidity * weekend + humidity * season, data = BikeSharing,
##      nvmax = 20)
## 19 Variables (and intercept)
##
##      Forced in Forced out
## temperature      FALSE      FALSE
## feels_like       FALSE      FALSE
## humidity         FALSE      FALSE
## wind_speed       FALSE      FALSE
## holiday          FALSE      FALSE
## weekend           FALSE      FALSE
## season           FALSE      FALSE
## temperature:holiday FALSE      FALSE
## temperature:weekend FALSE      FALSE
## temperature:season FALSE      FALSE
## feels_like:holiday FALSE      FALSE
## feels_like:weekend FALSE      FALSE
## feels_like:season  FALSE      FALSE
## wind_speed:holiday FALSE      FALSE
## wind_speed:weekend FALSE      FALSE
## wind_speed:season  FALSE      FALSE
## humidity:holiday  FALSE      FALSE
## humidity:weekend  FALSE      FALSE
## humidity:season    FALSE      FALSE
## 1 subsets of each size up to 19
## Selection Algorithm: exhaustive
##
##      temperature feels_like humidity wind_speed holiday weekend season
## 1  ( 1 )  " "      "*"      " "      " "      " "      " "
## 2  ( 1 )  " "      "*"      "*"      " "      " "      " "
## 3  ( 1 )  " "      "*"      "*"      " "      " "      " "
## 4  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 5  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 6  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 7  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 8  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 9  ( 1 )  " "      "*"      "*"      "*"      " "      " "
## 10 ( 1 )  " "      "*"      "*"      "*"      " "      "*"
## 11 ( 1 )  " "      "*"      "*"      "*"      "*"      "*"
## 12 ( 1 )  " "      "*"      "*"      "*"      "*"      " "
## 13 ( 1 )  " "      "*"      "*"      "*"      "*"      "*"
## 14 ( 1 )  " "      "*"      "*"      "*"      "*"      " "
## 15 ( 1 )  "*"      "*"      "*"      "*"      "*"      "*"      "*"
## 16 ( 1 )  " "      "*"      "*"      "*"      "*"      "*"      " "
## 17 ( 1 )  "*"      "*"      "*"      "*"      "*"      "*"      " "
## 18 ( 1 )  "*"      "*"      "*"      "*"      "*"      "*"      "*"
## 19 ( 1 )  "*"      "*"      "*"      "*"      "*"      "*"      "*"
##
##      temperature:holiday temperature:weekend temperature:season
## 1  ( 1 )  " "      " "      " "

```

```

## 2 ( 1 ) " " " "
## 3 ( 1 ) " " " "
## 4 ( 1 ) " " " "
## 5 ( 1 ) " " " "
## 6 ( 1 ) " " " "
## 7 ( 1 ) " " " "
## 8 ( 1 ) " " " "
## 9 ( 1 ) " " " "
## 10 ( 1 ) " " " "
## 11 ( 1 ) " " " "
## 12 ( 1 ) " " " * "
## 13 ( 1 ) " " " * "
## 14 ( 1 ) " " " * "
## 15 ( 1 ) " " " "
## 16 ( 1 ) " * " " * "
## 17 ( 1 ) " * " " * "
## 18 ( 1 ) " * " " * "
## 19 ( 1 ) " * " " * "
##
##      feels_like:holiday feels_like:weekend feels_like:season
## 1 ( 1 ) " " " "
## 2 ( 1 ) " " " "
## 3 ( 1 ) " " " "
## 4 ( 1 ) " " " "
## 5 ( 1 ) " " " "
## 6 ( 1 ) " " " * "
## 7 ( 1 ) " " " * "
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## 14 ( 1 ) " " " * "
## 15 ( 1 ) " " " * "
## 16 ( 1 ) " * " " * "
## 17 ( 1 ) " * " " * "
## 18 ( 1 ) " * " " * "
## 19 ( 1 ) " * " " * "
##
##      wind_speed:holiday wind_speed:weekend wind_speed:season
## 1 ( 1 ) " " " "
## 2 ( 1 ) " " " "
## 3 ( 1 ) " " " * "
## 4 ( 1 ) " " " "
## 5 ( 1 ) " " " "
## 6 ( 1 ) " " " "
## 7 ( 1 ) " " " * "
## 8 ( 1 ) " " " * "
## 9 ( 1 ) " " " * "
## 10 ( 1 ) " " " * "
## 11 ( 1 ) " " " * "
## 12 ( 1 ) " " " * "
## 13 ( 1 ) " " " * "
## 14 ( 1 ) " " " * "
## 15 ( 1 ) " * " " * "

```



```
## 16 ( 1 ) " " " * " " * "
## 17 ( 1 ) " " " * " " * "
## 18 ( 1 ) " " " * " " * "
## 19 ( 1 ) " * " " * " " * "
##      humidity:holiday humidity:weekend humidity:season
## 1 ( 1 ) " " " " " "
## 2 ( 1 ) " " " " " "
## 3 ( 1 ) " " " " " "
## 4 ( 1 ) " " " * " " "
## 5 ( 1 ) " * " " * " " "
## 6 ( 1 ) " * " " * " " "
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## 8 ( 1 ) " * " " * " " "
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## 11 ( 1 ) " * " " * " " "
## 12 ( 1 ) " * " " * " " * "
## 13 ( 1 ) " * " " * " " * "
## 14 ( 1 ) " * " " * " " * "
## 15 ( 1 ) " * " " * " " "
## 16 ( 1 ) " * " " * " " * "
## 17 ( 1 ) " * " " * " " * "
## 18 ( 1 ) " * " " * " " * "
## 19 ( 1 ) " * " " * " " * "
```

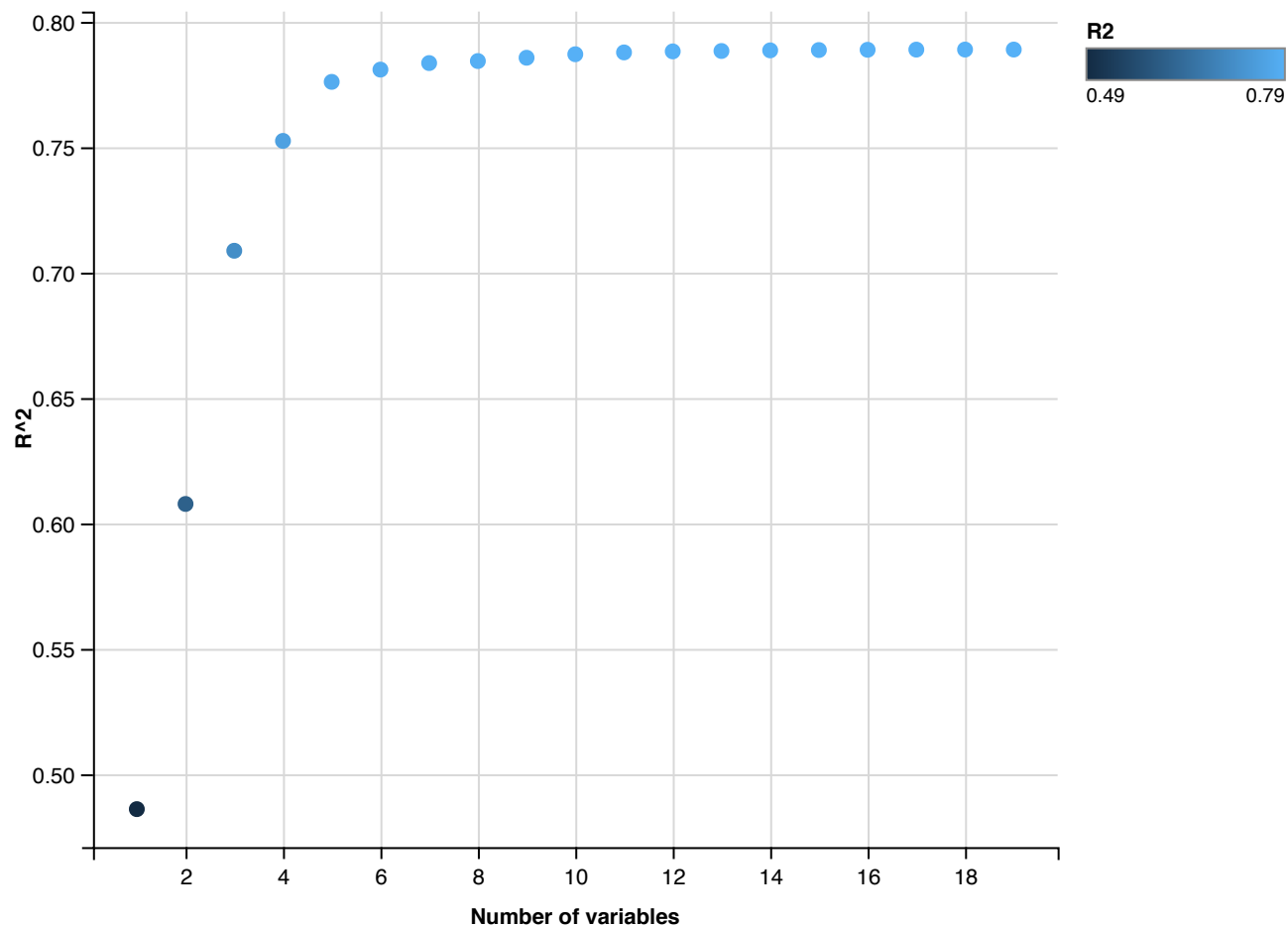
```
reg.summary$adjr2
```

```
## [1] 0.4850612 0.6062035 0.7069845 0.7505448 0.7738142 0.7783012 0.7804956
## [8] 0.7808679 0.7817611 0.7827414 0.7830404 0.7830140 0.7827299 0.7825421
## [15] 0.7822240 0.7818762 0.7814979 0.7810676 0.7806115
```

```
c(which.max(summary(m.select)$adjr2),max(summary(m.select)$adjr2))
```

```
## [1] 11.0000000 0.7830404
```

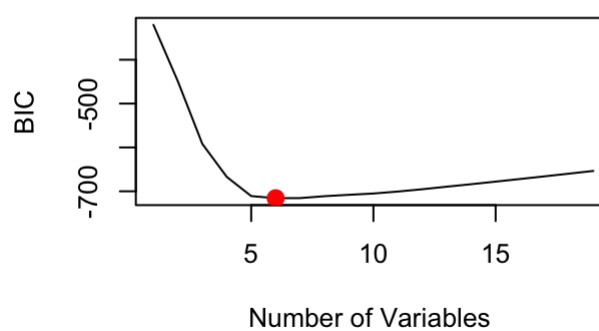
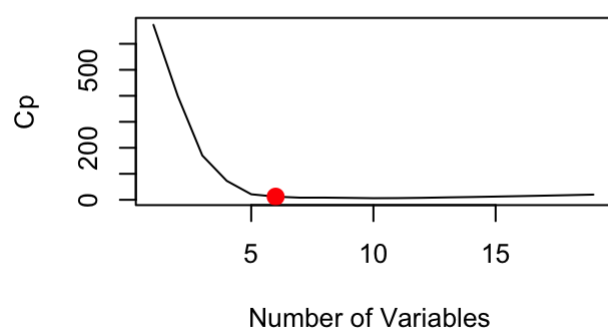
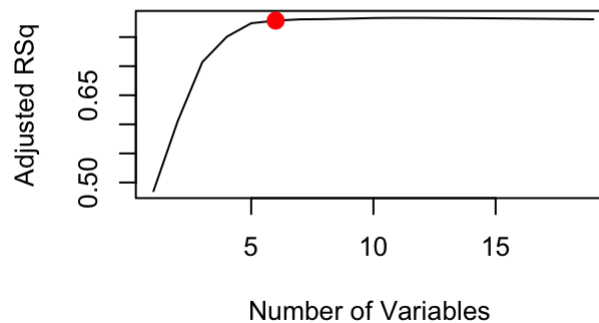
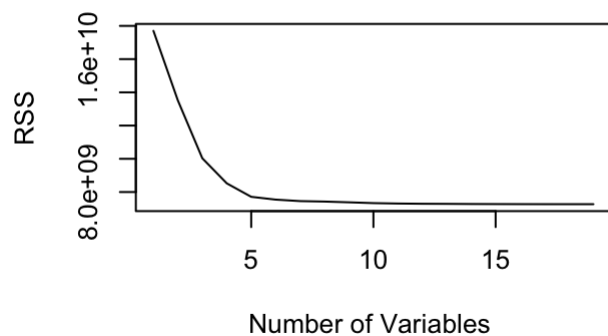
```
library(ggvis)
rsq <- as.data.frame(reg.summary$rsq)
names(rsq) <- "R2"
rsq %>%
  ggvis(x=~ c(1:nrow(rsq)), y=~R2 ) %>%
  layer_points(fill = ~ R2 ) %>%
  add_axis("y", title = "R^2") %>%
  add_axis("x", title = "Number of variables")
```



```

par(mfrow=c(2,2))
plot(reg.summary$rss ,xlab="Number of Variables ",ylab="RSS",type="l")
plot(reg.summary$adjr2 ,xlab="Number of Variables ", ylab="Adjusted RSq",type="l")
# which.max(reg.summary$adjr2)
points(6,reg.summary$adjr2[6], col="red",cex=2,pch=20)
plot(reg.summary$cp ,xlab="Number of Variables ",ylab="Cp", type='l')
# which.min(reg.summary$cp )
points(6,reg.summary$cp [6],col="red",cex=2,pch=20)
plot(reg.summary$bic ,xlab="Number of Variables ",ylab="BIC",type='l')
# which.min(reg.summary$bic )
points(6,reg.summary$bic [6],col="red",cex=2,pch=20)

```



```
#The model fitted from best subset selection
m_sub = lm(N_bikes~feels_like + humidity + wind_speed +feels_like:weekend + humidity:holiday + humidity:weekend,
           data = BikeSharing)
summary(m_sub)
```

```
##
## Call:
## lm(formula = N_bikes ~ feels_like + humidity + wind_speed + feels_like:weekend +
##      humidity:holiday + humidity:weekend, data = BikeSharing)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16829.0  -2460.1   191.7   2501.9  19307.3
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    51185.810    1763.072   29.032 < 2e-16 ***
## feels_like       678.121      34.271   19.787 < 2e-16 ***
## humidity        -338.182      20.192  -16.749 < 2e-16 ***
## wind_speed      -354.322      29.012  -12.213 < 2e-16 ***
## feels_like:weekend  186.135      56.126    3.316 0.000979 ***
## humidity:holiday  -112.115      15.087   -7.431 4.79e-13 ***
## humidity:weekend  -100.522       9.931  -10.123 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3912 on 493 degrees of freedom
## Multiple R-squared:  0.781, Adjusted R-squared:  0.7783
## F-statistic: 293 on 6 and 493 DF, p-value: < 2.2e-16
```

```
m_max = lm(N_bikes~temperature+feels_like+humidity+wind_speed+holiday+weekend+season+
            temperature*holiday+temperature*weekend+temperature*season+
            feels_like*holiday+feels_like*weekend+feels_like*season+
            wind_speed*holiday+wind_speed*weekend+wind_speed*season+
            humidity*holiday+humidity*weekend+humidity*season,data=BikeSharin
g)
Anova(m_max,type='II')
```

	Sum Sq <dbl>	Df <dbl>	F value <dbl>	Pr(>F) <dbl>
temperature	1.092693e+07	1	7.215994e-01	3.960432e-01
feels_like	5.065073e+07	1	3.344906e+00	6.803362e-02
humidity	3.718521e+09	1	2.455661e+02	5.370569e-45
wind_speed	1.729032e+09	1	1.141829e+02	4.705424e-24
holiday	8.642165e+08	1	5.707169e+01	2.141800e-13
weekend	2.685205e+09	1	1.773273e+02	1.184349e-34
season	7.082474e+05	1	4.677170e-02	8.288710e-01
temperature:holiday	2.643925e+06	1	1.746012e-01	6.762414e-01
temperature:weekend	1.167517e+07	1	7.710124e-01	3.803434e-01

	Sum Sq <dbl>	Df <dbl>	F value <dbl>	Pr(>F) <dbl>
temperature:season	1.307281e+06	1	8.633107e-02	7.690213e-01

1-10 of 20 rows

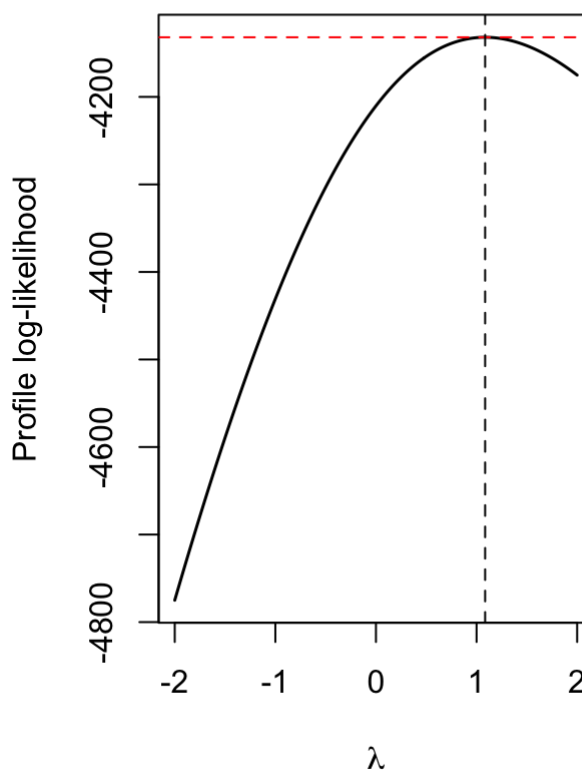
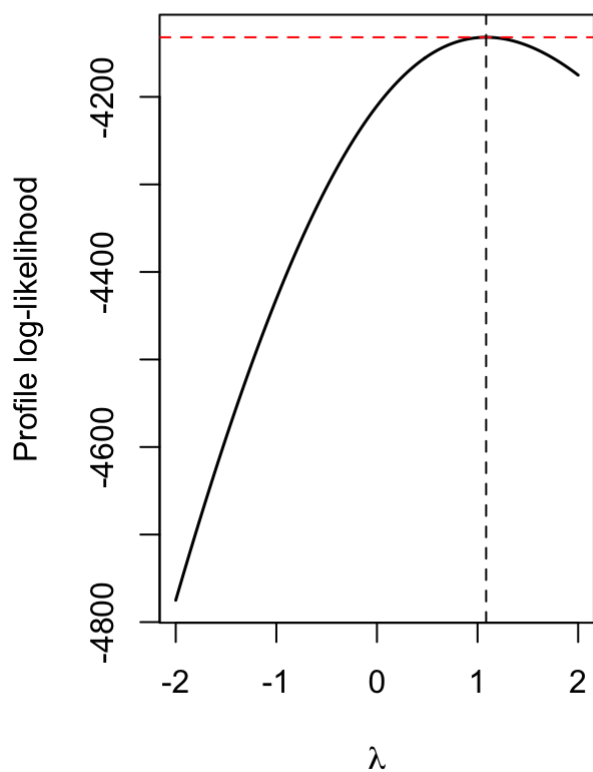
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Box-Cox and Yeo-Jhonson

```
par(mfrow=c(1,2))
boxcox(m_sub,plotit=TRUE)
```

```
## Box-Cox Transformation
##
## Estimation method:  ml
## Optimal parameter:  1.085186
## Loglike:  4131.914
##
## Summary of transformed variables
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  12486   47730   58991   59748   73933  105844
```

```
best.lambda_bc = boxcox(m_sub)$lambda.hat
```



```
best.lambda_bc
```

```
## [1] 1.085186
```

```
m_boxcox = lm(((N_bikes^best.lambda_bc-1)/best.lambda_bc)~feels_like + humidity + wind_s
peed +feels_like:weekend + humidity:holiday + humidity:weekend,data = BikeSharing)
summary(m_boxcox)
```

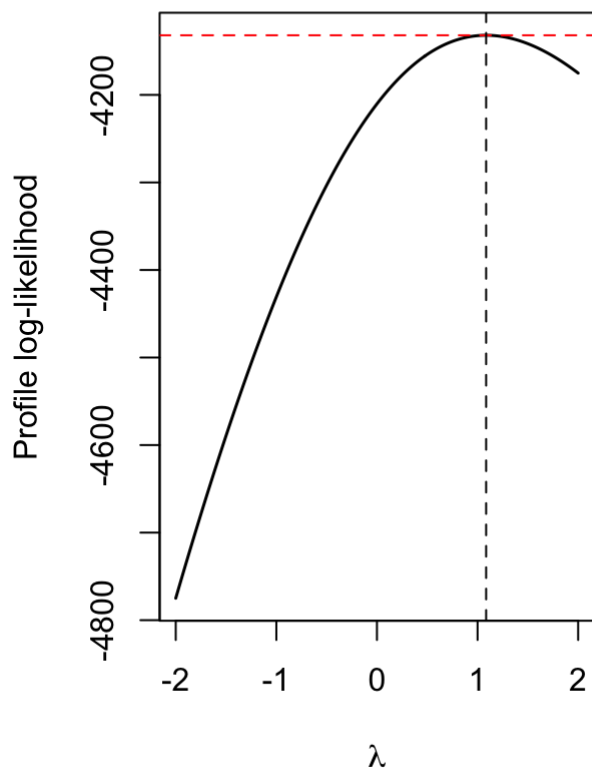
```
##
## Call:
## lm(formula = ((N_bikes^best.lambda_bc - 1)/best.lambda_bc) ~
##     feels_like + humidity + wind_speed + feels_like:weekend +
##     humidity:holiday + humidity:weekend, data = BikeSharing)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -38738  -5801    411    5839  45856
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    117124.56    4182.15  28.006 < 2e-16 ***
## feels_like      1631.63      81.29  20.071 < 2e-16 ***
## humidity        -808.68     47.90 -16.884 < 2e-16 ***
## wind_speed     -840.19     68.82 -12.209 < 2e-16 ***
## feels_like:weekend  401.90    133.14   3.019  0.00267 **
## humidity:holiday  -259.41     35.79  -7.248 1.64e-12 ***
## humidity:weekend  -229.83     23.56  -9.757 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9279 on 493 degrees of freedom
## Multiple R-squared:  0.7814, Adjusted R-squared:  0.7788
## F-statistic: 293.7 on 6 and 493 DF, p-value: < 2.2e-16
```

```
best.lambda_yeo = yeojohnson(m_sub)$lambdahat
best.lambda_yeo
```

```
## [1] 1.085192
```

```
m_yeo =lm(yeo.johnson(N_bikes,best.lambda_yeo)~feels_like + humidity + wind_speed +feels
_like:weekend + humidity:holiday + humidity:weekend,data=BikeSharing)
summary(m_yeo)
```

```
##
## Call:
## lm(formula = yeo.johnson(N_bikes, best.lambda_yeo) ~ feels_like +
##      humidity + wind_speed + feels_like:weekend + humidity:holiday +
##      humidity:weekend, data = BikeSharing)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -38741  -5801    411    5839  45859
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   117134.75    4182.45  28.006 < 2e-16 ***
## feels_like     1631.75     81.30  20.071 < 2e-16 ***
## humidity      -808.74     47.90 -16.884 < 2e-16 ***
## wind_speed    -840.25     68.82 -12.209 < 2e-16 ***
## feels_like:weekend  401.93    133.15   3.019  0.00267 **
## humidity:holiday -259.43     35.79  -7.248 1.64e-12 ***
## humidity:weekend -229.84     23.56  -9.757 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9280 on 493 degrees of freedom
## Multiple R-squared:  0.7814, Adjusted R-squared:  0.7788
## F-statistic: 293.7 on 6 and 493 DF, p-value: < 2.2e-16
```



Principle Component Regression

```
# #Without standardizing the variables
# m.pcr1 = pcr(N_bikes~.,data= BikeSharing,ncomp=7,validation='CV')
# prc1 = princomp(BikeSharing[,colnames(BikeSharing)!='N_bikes'])
# summary(m.pcr1)
# screeplot(prc1)
# m.pcr1$loadings

## With standardizing the variables
# BikeSharing[,colnames(BikeSharing)!='N_bikes'] = data.frame(scale(BikeSharing[,colnames(BikeSharing)!='N_bikes']))
#
# m.pcr2 = pcr(N_bikes~.,data= BikeSharing,ncomp=7,validation='CV')
# prc2 = princomp(BikeSharing[,colnames(BikeSharing)!='N_bikes'])
# summary(m.pcr2)
# screeplot(prc2)
# m.pcr2$loadings
#
#
# #Pick the non-standardized variable model, with ncomp = 3
# BikeSharing <- read.delim("bikesharing18.txt",sep = "")
# m.pcr = pcr(N_bikes~.,data= BikeSharing,ncomp=3,validation='CV')
# MSEP(m.pcr)$val
# summary(m.pcr)
```

```
library(caret)
```

```
## Loading required package: lattice
```

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

```
##
## Attaching package: 'ggplot2'
```

```
## The following object is masked from 'package:ggvis':
##
## resolution
```

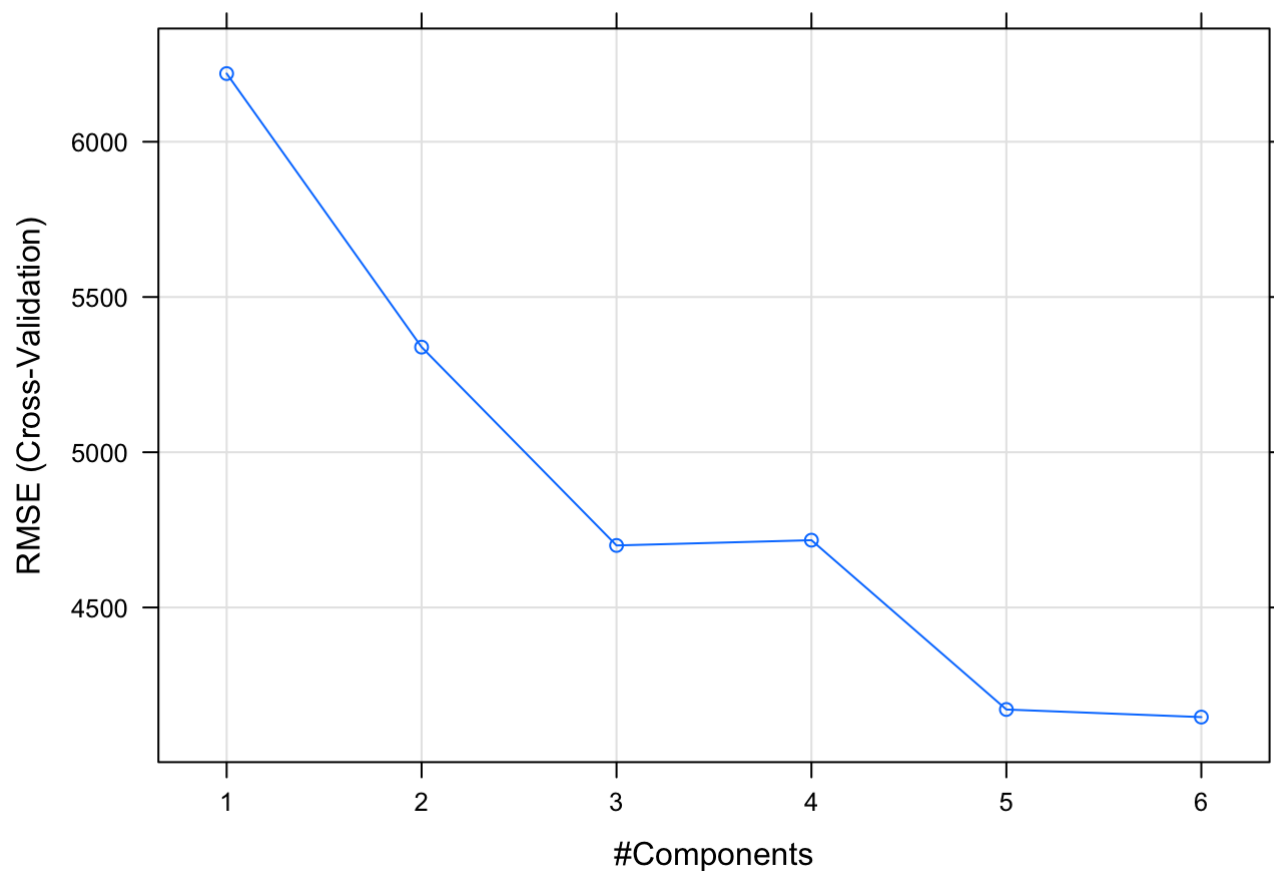
```
##
## Attaching package: 'caret'
```

```
## The following object is masked from 'package:VGAM':
##
## predictors
```



```
## The following object is masked from 'package:pls':  
##  
##      R2
```

```
set.seed(123)  
training.samples <- BikeSharing$N_bikes %>%  
  createDataPartition(p = 0.66, list = FALSE)  
train.data <- BikeSharing[training.samples, ]  
test.data <- BikeSharing[-training.samples, ]  
  
set.seed(123)  
model <- train(  
  N_bikes~., data = train.data, method = "pcr",  
  scale = FALSE,  
  trControl = trainControl("cv", number = 7),  
  tuneLength = 7  
)  
# Plot model RMSE vs different values of components  
plot(model)
```



```
# Print the best tuning parameter ncomp that  
# minimize the cross-validation error, RMSE  
model$bestTune
```

ncomp
<dbl>

6	6
---	---

1 row

```
summary(model$finalModel)
```

```
## Data:      X dimension: 332 7
## Y dimension: 332 1
## Fit method: svdpc
## Number of components considered: 6
## TRAINING: % variance explained
##           1 comps  2 comps  3 comps  4 comps  5 comps  6 comps
## X           58.91   82.26   99.32   99.80   99.91   99.99
## .outcome    43.40   58.77   68.06   68.07   74.96   75.19
```

```
# Make predictions
predictions <- model %>% predict(test.data)
# Model performance metrics
data.frame(
  RMSE = caret::RMSE(predictions, test.data$N_bikes),
  Rsquare = caret::R2(predictions, test.data$N_bikes)
)
```

RMSE
<dbl>

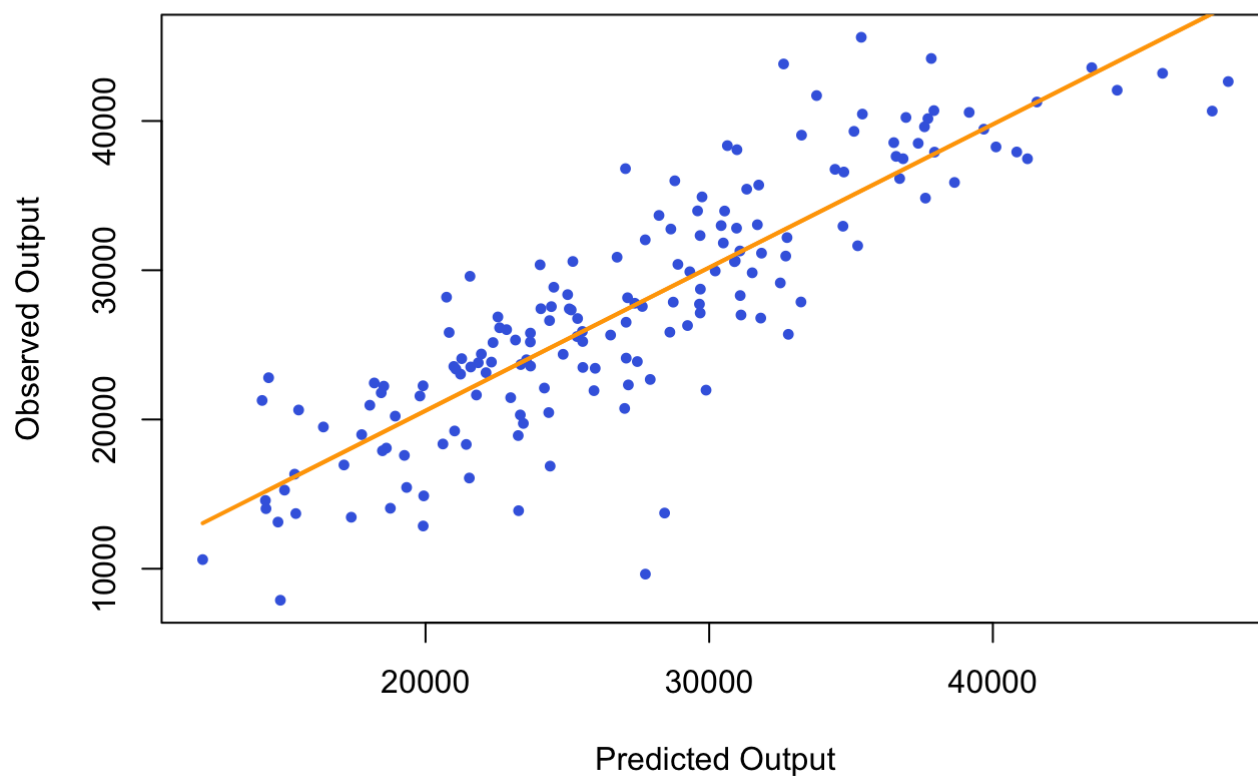
Rsquare
<dbl>

4248.386	0.7394426
----------	-----------

1 row

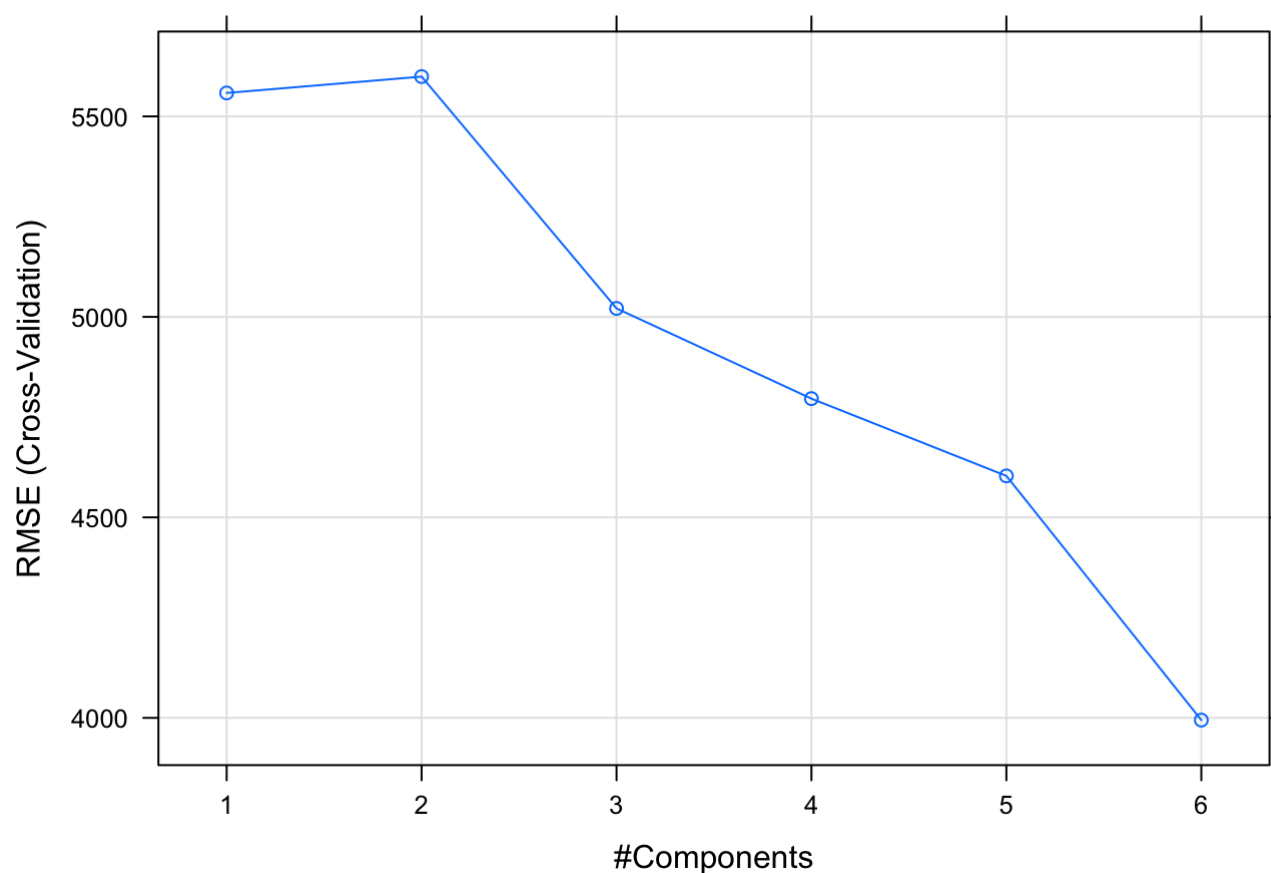
```
plot(predictions, test.data$N_bikes, pch=16, col="royalblue", cex=0.75,
xlab="Predicted Output",
ylab="Observed Output",
main="Principle Component Regression: Observed vs. Predicted")
lines(predictions, lm(a~b, data=data.frame(a=test.data$N_bikes, b=predictions))$fitted,
  lwd=2, col="orange")
```

Principle Component Regression: Observed vs. Predicted



```
#Normalizing data
BikeSharing[,colnames(BikeSharing)!='N_bikes'] = data.frame(scale(BikeSharing[,colnames
(BikeSharing)!='N_bikes'])))
set.seed(123)
training.samples <- BikeSharing$N_bikes %>%
  createDataPartition(p = 0.66, list = FALSE)
train.data <- BikeSharing[training.samples, ]
test.data <- BikeSharing[-training.samples, ]

set.seed(123)
model <- train(
  N_bikes~., data = train.data, method = "pcr",
  scale = FALSE,
  trControl = trainControl("cv", number = 7),
  tuneLength = 7
)
# Plot model RMSE vs different values of components
plot(model)
```



```
# Print the best tuning parameter ncomp that
# minimize the cross-validation error, RMSE
model$bestTune
```

	ncomp <dbl>
6	6

1 row

```
summary(model$finalModel)
```

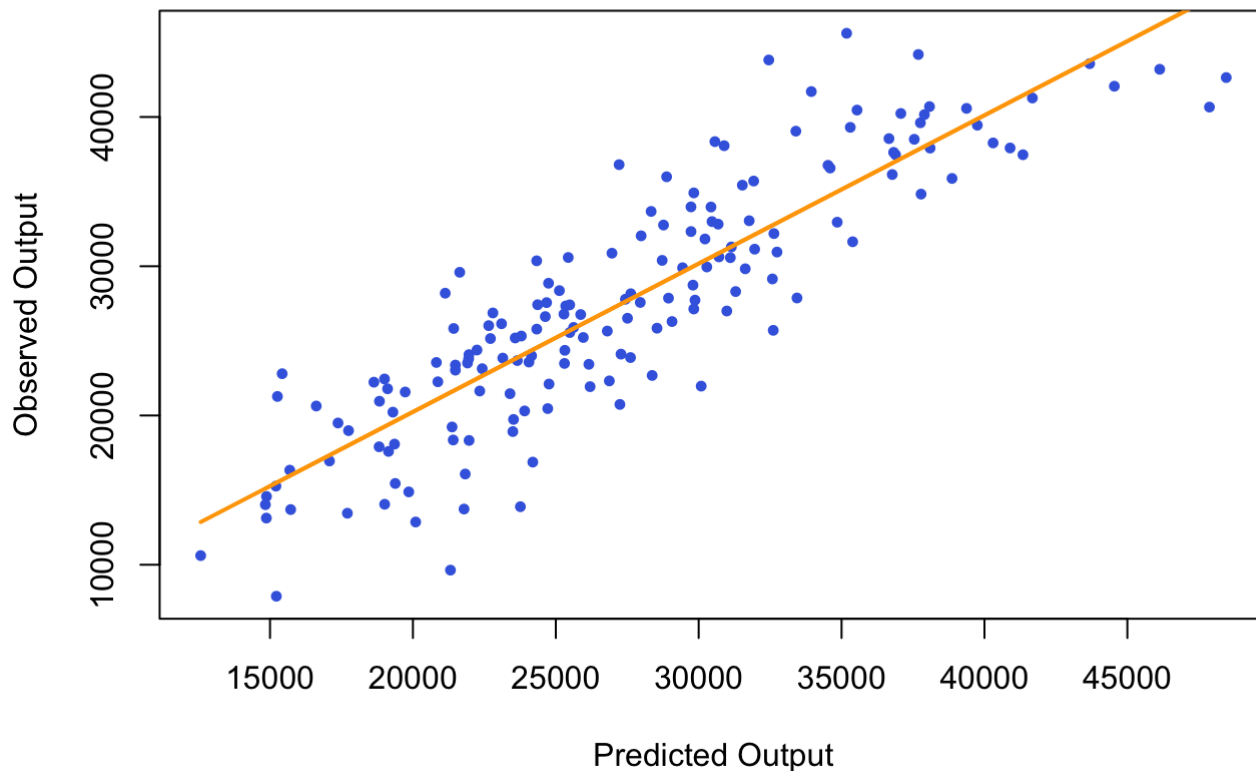
```
## Data:      X dimension: 332 7
## Y dimension: 332 1
## Fit method: svdpc
## Number of components considered: 6
## TRAINING: % variance explained
##           1 comps  2 comps  3 comps  4 comps  5 comps  6 comps
## X           34.51   51.75   66.67   80.07   92.80   99.93
## .outcome    54.88   55.04   64.41   67.46   69.92   76.83
```

```
# Make predictions
predictions <- model %>% predict(test.data)
# Model performance metrics
data.frame(
  RMSE = caret::RMSE(predictions, test.data$N_bikes),
  Rsquare = caret::R2(predictions, test.data$N_bikes)
)
```

	RMSE <dbl>	Rsquare <dbl>
	3947.463	0.7733957
1 row		

```
plot(predictions, test.data$N_bikes, pch=16, col="royalblue", cex=0.75,
  xlab="Predicted Output",
  ylab="Observed Output",
  main="Principle Component Regression: Observed vs. Predicted")
lines(predictions, lm(a~b, data=data.frame(a=test.data$N_bikes, b=predictions))$fitted,
  lwd=2, col="orange")
```

Principle Component Regression: Observed vs. Predicted



```
library(xgboost)
library(Metrics)
```

```
##  
## Attaching package: 'Metrics'
```

```
## The following objects are masked from 'package:caret':  
##  
##      precision, recall
```

```
training.samples <- BikeSharing$N_bikes %>%  
  createDataPartition(p = 0.66, list = FALSE)  
train.data <- BikeSharing[training.samples, ]  
test.data <- BikeSharing[-training.samples, ]  
y_train = train.data[,1]  
X_train = train.data[,-1]  
X_train <- as.matrix(X_train)  
  
y_test = test.data[,1]  
X_test = test.data[,-1]  
X_test <- as.matrix(X_test)  
  
fit_xgb <- xgboost(X_train, y_train  
                  , max_depth = 10  
                  , eta = 0.02  
                  , nthread = 4  
                  , nrounds = 800  
                  , subsample = .7  
                  , colsample_bytree = .7  
                  , booster = "gbtree"  
                  , eval_metric = "rmse"  
                  , objective="reg:linear")
```

```
## [1] train-rmse:27743.531250
## [2] train-rmse:27221.800781
## [3] train-rmse:26712.132812
## [4] train-rmse:26208.347656
## [5] train-rmse:25720.515625
## [6] train-rmse:25240.978516
## [7] train-rmse:24762.300781
## [8] train-rmse:24305.730469
## [9] train-rmse:23853.441406
## [10] train-rmse:23408.259766
## [11] train-rmse:22971.455078
## [12] train-rmse:22538.347656
## [13] train-rmse:22119.998047
## [14] train-rmse:21704.425781
## [15] train-rmse:21292.203125
## [16] train-rmse:20901.816406
## [17] train-rmse:20518.195312
## [18] train-rmse:20134.470703
## [19] train-rmse:19762.724609
## [20] train-rmse:19399.187500
## [21] train-rmse:19047.445312
## [22] train-rmse:18700.113281
## [23] train-rmse:18371.136719
## [24] train-rmse:18032.853516
## [25] train-rmse:17703.134766
## [26] train-rmse:17384.726562
## [27] train-rmse:17066.320312
## [28] train-rmse:16754.599609
## [29] train-rmse:16447.916016
## [30] train-rmse:16150.884766
## [31] train-rmse:15861.286133
## [32] train-rmse:15570.099609
## [33] train-rmse:15294.421875
## [34] train-rmse:15028.643555
## [35] train-rmse:14757.707031
## [36] train-rmse:14486.216797
## [37] train-rmse:14222.744141
## [38] train-rmse:13968.128906
## [39] train-rmse:13726.092773
## [40] train-rmse:13484.792969
## [41] train-rmse:13244.374023
## [42] train-rmse:13008.515625
## [43] train-rmse:12783.118164
## [44] train-rmse:12559.764648
## [45] train-rmse:12336.700195
## [46] train-rmse:12115.701172
## [47] train-rmse:11901.664062
## [48] train-rmse:11695.286133
## [49] train-rmse:11497.737305
## [50] train-rmse:11304.179688
## [51] train-rmse:11108.521484
## [52] train-rmse:10922.160156
## [53] train-rmse:10731.785156
```

```
## [54] train-rmse:10539.979492
## [55] train-rmse:10361.614258
## [56] train-rmse:10184.431641
## [57] train-rmse:10009.558594
## [58] train-rmse:9836.306641
## [59] train-rmse:9669.949219
## [60] train-rmse:9503.525391
## [61] train-rmse:9345.850586
## [62] train-rmse:9185.360352
## [63] train-rmse:9037.725586
## [64] train-rmse:8886.983398
## [65] train-rmse:8734.384766
## [66] train-rmse:8589.279297
## [67] train-rmse:8453.056641
## [68] train-rmse:8313.802734
## [69] train-rmse:8185.420410
## [70] train-rmse:8059.378418
## [71] train-rmse:7923.378418
## [72] train-rmse:7796.852539
## [73] train-rmse:7671.351562
## [74] train-rmse:7553.328613
## [75] train-rmse:7436.678711
## [76] train-rmse:7320.436035
## [77] train-rmse:7200.841797
## [78] train-rmse:7089.483887
## [79] train-rmse:6976.525879
## [80] train-rmse:6873.501953
## [81] train-rmse:6767.017090
## [82] train-rmse:6666.192871
## [83] train-rmse:6567.474121
## [84] train-rmse:6462.243652
## [85] train-rmse:6358.239258
## [86] train-rmse:6259.083496
## [87] train-rmse:6161.566895
## [88] train-rmse:6064.799805
## [89] train-rmse:5974.864258
## [90] train-rmse:5890.081543
## [91] train-rmse:5805.908691
## [92] train-rmse:5719.312500
## [93] train-rmse:5635.772461
## [94] train-rmse:5552.648438
## [95] train-rmse:5469.339844
## [96] train-rmse:5400.510254
## [97] train-rmse:5325.775879
## [98] train-rmse:5250.012207
## [99] train-rmse:5180.275879
## [100]   train-rmse:5116.121582
## [101]   train-rmse:5045.099121
## [102]   train-rmse:4975.862305
## [103]   train-rmse:4906.013184
## [104]   train-rmse:4835.144043
## [105]   train-rmse:4766.580078
## [106]   train-rmse:4698.137695
## [107]   train-rmse:4627.566895
```



```
## [108] train-rmse:4564.226074
## [109] train-rmse:4504.202637
## [110] train-rmse:4450.030273
## [111] train-rmse:4389.895020
## [112] train-rmse:4332.625488
## [113] train-rmse:4276.886719
## [114] train-rmse:4218.433105
## [115] train-rmse:4161.882324
## [116] train-rmse:4111.832031
## [117] train-rmse:4061.996826
## [118] train-rmse:4010.291504
## [119] train-rmse:3953.902832
## [120] train-rmse:3909.545166
## [121] train-rmse:3853.952637
## [122] train-rmse:3807.298828
## [123] train-rmse:3754.910156
## [124] train-rmse:3703.069580
## [125] train-rmse:3658.929443
## [126] train-rmse:3610.255371
## [127] train-rmse:3565.666260
## [128] train-rmse:3517.803467
## [129] train-rmse:3471.909668
## [130] train-rmse:3429.885498
## [131] train-rmse:3393.559814
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## [139] train-rmse:3089.548340
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## [141] train-rmse:3014.766113
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## [144] train-rmse:2917.207764
## [145] train-rmse:2886.307861
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## [147] train-rmse:2829.162598
## [148] train-rmse:2800.322754
## [149] train-rmse:2770.081055
## [150] train-rmse:2740.758301
## [151] train-rmse:2710.939453
## [152] train-rmse:2685.288818
## [153] train-rmse:2654.124023
## [154] train-rmse:2627.987061
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## [156] train-rmse:2574.392822
## [157] train-rmse:2548.642334
## [158] train-rmse:2524.929443
## [159] train-rmse:2500.286865
## [160] train-rmse:2472.302979
## [161] train-rmse:2449.703369
```

```
## [162] train-rmse:2421.910156
## [163] train-rmse:2400.241455
## [164] train-rmse:2374.399170
## [165] train-rmse:2351.449219
## [166] train-rmse:2327.297607
## [167] train-rmse:2301.128418
## [168] train-rmse:2279.024414
## [169] train-rmse:2259.474121
## [170] train-rmse:2242.081543
## [171] train-rmse:2222.509277
## [172] train-rmse:2202.005127
## [173] train-rmse:2178.126953
## [174] train-rmse:2159.520996
## [175] train-rmse:2139.286621
## [176] train-rmse:2116.710693
## [177] train-rmse:2094.339355
## [178] train-rmse:2077.511963
## [179] train-rmse:2059.480713
## [180] train-rmse:2041.257812
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## [182] train-rmse:2008.536987
## [183] train-rmse:1990.047363
## [184] train-rmse:1970.418457
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## [200] train-rmse:1723.855835
## [201] train-rmse:1708.864868
## [202] train-rmse:1695.851074
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## [204] train-rmse:1663.069092
## [205] train-rmse:1652.534058
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## [207] train-rmse:1625.977783
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## [209] train-rmse:1599.584961
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## [211] train-rmse:1577.029541
## [212] train-rmse:1563.318481
## [213] train-rmse:1549.054443
## [214] train-rmse:1538.256104
## [215] train-rmse:1525.852051
```

```
## [216] train-rmse:1512.236206
## [217] train-rmse:1499.499390
## [218] train-rmse:1485.424927
## [219] train-rmse:1472.894287
## [220] train-rmse:1459.263184
## [221] train-rmse:1446.096191
## [222] train-rmse:1436.820557
## [223] train-rmse:1424.882812
## [224] train-rmse:1413.234497
## [225] train-rmse:1402.587891
## [226] train-rmse:1392.371460
## [227] train-rmse:1383.109131
## [228] train-rmse:1376.226685
## [229] train-rmse:1364.655029
## [230] train-rmse:1353.670776
## [231] train-rmse:1342.640625
## [232] train-rmse:1335.910278
## [233] train-rmse:1326.055664
## [234] train-rmse:1315.624878
## [235] train-rmse:1305.260010
## [236] train-rmse:1297.424683
## [237] train-rmse:1286.511108
## [238] train-rmse:1276.915527
## [239] train-rmse:1267.100830
## [240] train-rmse:1258.540283
## [241] train-rmse:1248.128052
## [242] train-rmse:1239.351562
## [243] train-rmse:1230.786743
## [244] train-rmse:1223.972534
## [245] train-rmse:1217.058228
## [246] train-rmse:1207.585327
## [247] train-rmse:1198.423828
## [248] train-rmse:1189.002686
## [249] train-rmse:1179.613647
## [250] train-rmse:1170.701294
## [251] train-rmse:1162.062500
## [252] train-rmse:1152.641113
## [253] train-rmse:1142.710815
## [254] train-rmse:1132.348511
## [255] train-rmse:1126.051025
## [256] train-rmse:1117.499023
## [257] train-rmse:1111.070923
## [258] train-rmse:1102.713745
## [259] train-rmse:1096.026978
## [260] train-rmse:1086.269287
## [261] train-rmse:1077.637573
## [262] train-rmse:1068.922363
## [263] train-rmse:1064.730957
## [264] train-rmse:1055.606323
## [265] train-rmse:1047.673950
## [266] train-rmse:1042.374634
## [267] train-rmse:1036.812988
## [268] train-rmse:1030.233154
## [269] train-rmse:1023.082153
```

```
## [270] train-rmse:1014.900818
## [271] train-rmse:1007.253052
## [272] train-rmse:1002.228271
## [273] train-rmse:996.697571
## [274] train-rmse:990.601501
## [275] train-rmse:985.431580
## [276] train-rmse:981.181641
## [277] train-rmse:974.408691
## [278] train-rmse:968.434204
## [279] train-rmse:965.172485
## [280] train-rmse:960.383789
## [281] train-rmse:953.143127
## [282] train-rmse:946.859741
## [283] train-rmse:939.425415
## [284] train-rmse:933.087646
## [285] train-rmse:926.878540
## [286] train-rmse:921.696716
## [287] train-rmse:915.811401
## [288] train-rmse:911.903809
## [289] train-rmse:907.006226
## [290] train-rmse:900.148315
## [291] train-rmse:894.573669
## [292] train-rmse:888.859680
## [293] train-rmse:883.806213
## [294] train-rmse:878.357483
## [295] train-rmse:872.805725
## [296] train-rmse:867.807678
## [297] train-rmse:863.119751
## [298] train-rmse:859.031128
## [299] train-rmse:854.806152
## [300] train-rmse:850.860168
## [301] train-rmse:844.665344
## [302] train-rmse:839.771240
## [303] train-rmse:834.503296
## [304] train-rmse:828.773132
## [305] train-rmse:824.137634
## [306] train-rmse:819.328735
## [307] train-rmse:813.786255
## [308] train-rmse:808.183899
## [309] train-rmse:805.742493
## [310] train-rmse:801.410950
## [311] train-rmse:796.990051
## [312] train-rmse:791.335754
## [313] train-rmse:785.102661
## [314] train-rmse:781.866272
## [315] train-rmse:777.517334
## [316] train-rmse:771.622131
## [317] train-rmse:767.515991
## [318] train-rmse:763.874451
## [319] train-rmse:758.611389
## [320] train-rmse:755.924255
## [321] train-rmse:751.644958
## [322] train-rmse:748.244202
## [323] train-rmse:743.013611
```

```
## [324] train-rmse:738.568787
## [325] train-rmse:734.080688
## [326] train-rmse:731.502319
## [327] train-rmse:727.385437
## [328] train-rmse:723.855652
## [329] train-rmse:719.661560
## [330] train-rmse:716.766846
## [331] train-rmse:710.849426
## [332] train-rmse:706.254211
## [333] train-rmse:702.621765
## [334] train-rmse:698.704041
## [335] train-rmse:694.002991
## [336] train-rmse:691.618225
## [337] train-rmse:687.534302
## [338] train-rmse:684.415222
## [339] train-rmse:680.462708
## [340] train-rmse:678.083862
## [341] train-rmse:675.511230
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## [343] train-rmse:668.237915
## [344] train-rmse:664.607788
## [345] train-rmse:661.414246
## [346] train-rmse:657.721375
## [347] train-rmse:655.312256
## [348] train-rmse:652.114136
## [349] train-rmse:649.788452
## [350] train-rmse:648.471680
## [351] train-rmse:644.893311
## [352] train-rmse:641.398621
## [353] train-rmse:638.242065
## [354] train-rmse:633.905823
## [355] train-rmse:629.340149
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## [357] train-rmse:621.204346
## [358] train-rmse:618.642700
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## [361] train-rmse:609.276184
## [362] train-rmse:605.911926
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## [364] train-rmse:599.114441
## [365] train-rmse:595.777161
## [366] train-rmse:593.307800
## [367] train-rmse:591.145264
## [368] train-rmse:587.473572
## [369] train-rmse:585.214966
## [370] train-rmse:581.388062
## [371] train-rmse:578.013489
## [372] train-rmse:575.280151
## [373] train-rmse:572.318665
## [374] train-rmse:569.902832
## [375] train-rmse:566.112854
## [376] train-rmse:563.149475
## [377] train-rmse:559.802490
```

```
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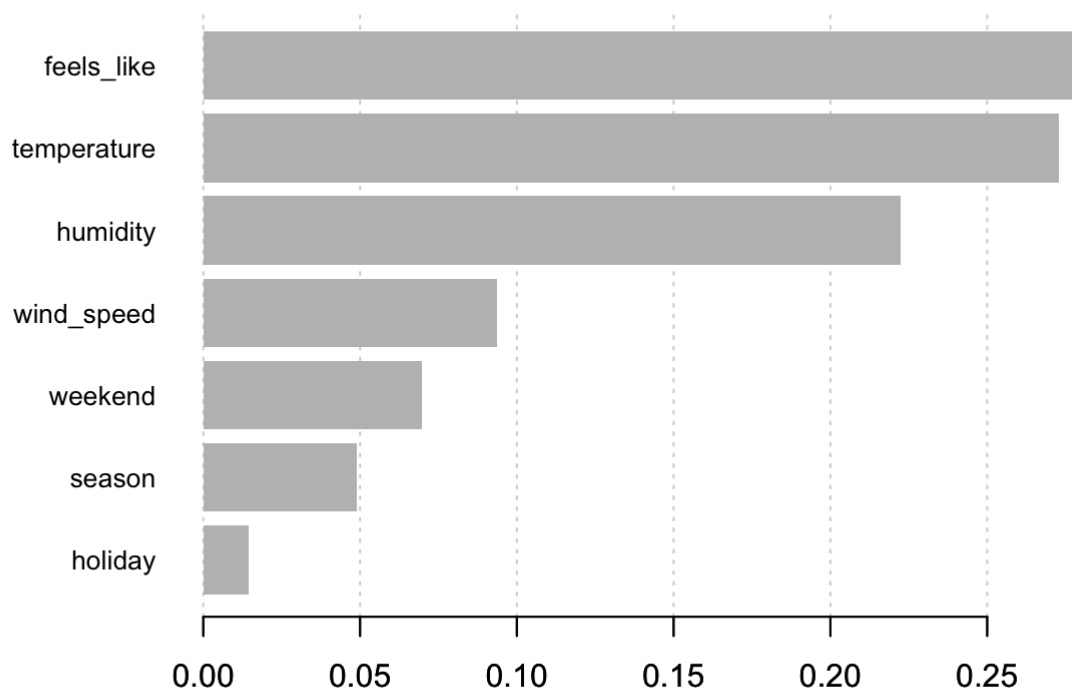
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## [798] train-rmse:69.887207
## [799] train-rmse:69.684853
## [800] train-rmse:69.356483
```

```
y_hat_xgb <- predict(fit_xgb, X_test)
```

```
## Plot the feature importance
```

```
importance_matrix <- xgb.importance(colnames(X_train), model = fit_xgb)
xgb.plot.importance(importance_matrix = importance_matrix[1:7])
```



```
data.frame(
  RMSE = RMSE(y_hat_xgb,y_test),
  Rsquare = R2(y_hat_xgb,y_test)
)
```

	RMSE <dbl>	Rsquare <dbl>
1 row	4372.736	0.7307053

```
plot(y_hat_xgb, y_test, pch=16, col="royalblue", cex=0.75,
xlab="Predicted Output",
ylab="Observed Output",
main="XGBOOST: Observed vs. Predicted")
lines(y_hat_xgb, lm(a~b, data=data.frame(a=y_test, b=y_hat_xgb))$fitted, lwd=2, col="orange")
```

XGBOOST: Observed vs. Predicted

