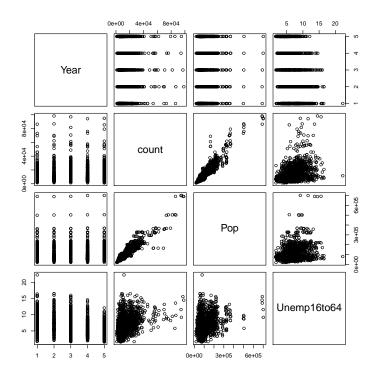
```
#Read in the total_crime dataset which contains the LAD values, year, and count
totCrime <- read.csv("total_crime_LAD_year.csv")</pre>
totCrime <- totCrime[,-1] #Drop first column, contains an index
#Read in the counts of type
totType <- read.csv("type_LAD_year.csv")</pre>
totType <- totType[,-1] #Drop first column, contains an index
#Get LAD/Year for data used in maps
shape <- read.csv("shape.csv")</pre>
#Rename column to match other files
names(shape) [names(shape) == "District"] <- "LAD_name"</pre>
#Load unemployment data
unemp <- read.csv("UnemploymentLAD.csv")</pre>
#Rename columns we are going to use to start
names(unemp)[names(unemp)=="local.authority..district...unitary..prior.to.April.2015."] <- "LAD_name"</pre>
names(unemp) [names(unemp) == "Date"] <- "Year"</pre>
names(unemp) [names(unemp) == "Unemployment.rate...aged.16.64"] <- "Unemp16to64"</pre>
names(unemp) [names(unemp) == "Denominator"] <- "Pop"</pre>
#Get rid of some of the extra columns
unemp <- unemp[,-grep("(Conf|Numerator|Denominator)",names(unemp))]</pre>
#Try the first regression
#Limit Unemployment data file to just the variables that we need
reg1.unemp <- unemp[,names(unemp) %in% c("LAD_name","Year","Unemp16to64","Pop")]
#Perform merge of unemployment data and crime data
reg1.data <- merge(totCrime, reg1.unemp, by=c("LAD_name","Year"), all=TRUE)</pre>
#Perform merge of merged unemp/crime and the shape file for maps
reg1.data <- merge(shape, reg1.data, by=c("LAD_name","Year"), all.x=TRUE)</pre>
#Remove observations with weird characters frm Unemp16to64
reg1.data <- reg1.data[!(reg1.data$Unemp16to64 %in% c("!","-")),]
#Remove column total rows
reg1.data <- reg1.data[reg1.data$LAD_name!="Column Total",]</pre>
#Change variable formats as needed
reg1.data$Year <- as.factor(reg1.data$Year)</pre>
reg1.data$Unemp16to64 <- as.numeric(levels(reg1.data$Unemp16to64))[reg1.data$Unemp16to64]
## Warning: NAs introduced by coercion
reg1.data$Pop <- as.numeric(levels(reg1.data$Pop))[reg1.data$Pop]</pre>
## Warning: NAs introduced by coercion
```

```
#Check pairs plot
pairs(reg1.data[,colnames(reg1.data) %in% c("Year","count","Pop","Unemp16to64")])
```



```
#First regression done
reg1 <- lm(count ~ Year + Unemp16to64 + Pop, data=reg1.data)
summary(reg1)
##
## Call:
## lm(formula = count ~ Year + Unemp16to64 + Pop, data = reg1.data)
##
## Residuals:
##
     Min
             1Q Median
                           3Q
## -15415 -1920
                 -228
                        1604 45249
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.935e+03 3.656e+02 -18.966 < 2e-16 ***
## Year2012
              2.313e+03 3.217e+02
                                     7.190 1.02e-12 ***
## Year2013
               1.802e+03 3.213e+02
                                      5.609 2.42e-08 ***
## Year2014
               2.103e+03 3.300e+02
                                     6.374 2.44e-10 ***
               3.193e+03 3.453e+02 9.248 < 2e-16 ***
## Year2015
## Unemp16to64 5.614e+02 3.965e+01 14.157 < 2e-16 ***
               1.427e-01 1.466e-03 97.367 < 2e-16 ***
## Pop
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 4007 on 1503 degrees of freedom
## Multiple R-squared: 0.891, Adjusted R-squared: 0.8906
## F-statistic: 2047 on 6 and 1503 DF, p-value: < 2.2e-16
#Get asterisks from regression
reg1sum <- summary(reg1)</pre>
```

```
pvals <- coef(reg1sum)[,colnames(coef(reg1sum))=="Pr(>|t|)"]
names(pvals) <- rownames(coef(reg1sum))
sig.pvals <- rep(NA,length(pvals))
sig.pvals[pvals<0.01] <- "***"
f.p <- pf(reg1sum$fstatistic[1],reg1sum$fstatistic[2],reg1sum$fstatistic[3],lower.tail=FALSE)
f.sig <- rep(NA,1)
f.sig[f.p<0.01] <- "***"
sig.pvals <- c(sig.pvals,f.sig)
names(sig.pvals) <- c(names(pvals),"fstat")</pre>
```

Table 1: Regression Results

Variables	OLS	
Unemployment Ages 16-64	561.398*** (14.16)	
Population Size	0.143*** (97.37)	
Year 2012	2312.862*** (7.19)	
Year 2013	1801.764*** (5.61)	
Year 2014	2103.244*** (6.37)	
Year 2015	3193.214*** (9.25)	
Adjusted $\mathbb{R}^2$	0.891	
F	2047.428***	
N	1510	

Notes: t/z-values of coefficients in parentheses, with level of significance shown as \*\*\* = (99%), \*\* = (95%), and \* = (90%). Data is at the Local Authority District level and covers England.

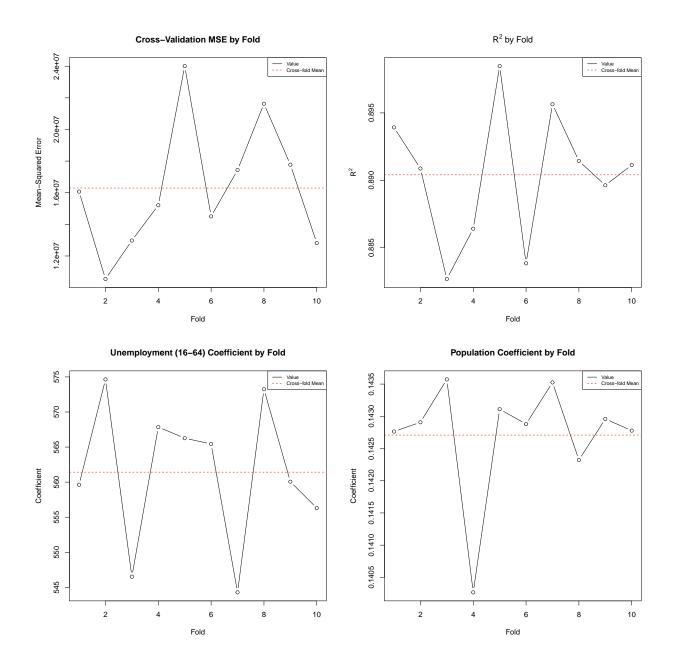
```
#Model validation
#Use cross-validation
k <- 10 #Number of cv folds

#Create index to identify folds
folds <- sample(1:k, nrow(reg1.data), replace=TRUE)

#Create matrix to store error values from each regression
cv.errors <- rep(NA,k)
r.squared <- rep(NA,k)
coeffs <- matrix(NA,k,7)</pre>
```

```
#Run cross-validation for best subset selection
for (j in 1:k) {
 #Get best subset for the fold
 lm.fit <- lm(count ~ Year + Unemp16to64 + Pop, data=reg1.data[folds!=j,])</pre>
 #Get prediction for this fold and i predictors
 pred <- predict(lm.fit, reg1.data[folds==j,])</pre>
  #Save mse
 cv.errors[j] <- mean((reg1.data$count[folds==j]-pred)^2)</pre>
 r.squared[j] <- summary(lm.fit)$adj.r.squared</pre>
  coeffs[j,] <- coefficients(lm.fit)</pre>
colnames(coeffs) <- names(coefficients(lm.fit))</pre>
#Take average of vector to get test mse
mean.cv.errors <- mean(cv.errors)</pre>
#CV Error
plot(cv.errors, type="b",main="Cross-Validation MSE by Fold",
     xlab="Fold",
     ylab="Mean-Squared Error")
abline(h=mean.cv.errors,col="red",lty=2) #Plot average fold
legend("topright",
       legend=c("Value","Cross-fold Mean"),
       col=c("black","red"),
       lwd=1, lty=c(1,2), cex=0.7)
plot(r.squared, type="b",main=bquote(R^2*" by Fold"),
     xlab="Fold",
     vlab=bquote(R^2))
abline(h=mean(r.squared),col="red",lty=2) #Plot average fold
legend("topright",
       legend=c("Value","Cross-fold Mean"),
       col=c("black", "red"),
       lwd=1, lty=c(1,2), cex=0.7)
#Unemployment
plot(coeffs[,colnames(coeffs)=="Unemp16to64"], type="b",
     main="Unemployment (16-64) Coefficient by Fold",
     xlab="Fold",
     vlab="Coefficient")
abline(h=mean(coeffs[,colnames(coeffs)=="Unemp16to64"]),
       col="red", lty=2) #Plot average fold
legend("topright",
       legend=c("Value", "Cross-fold Mean"),
       col=c("black", "red"),
       lwd=1, lty=c(1,2), cex=0.7)
#Unemployment
```

```
plot(coeffs[,colnames(coeffs)=="Pop"], type="b",
     main="Population Coefficient by Fold",
     xlab="Fold",
     vlab="Coefficient")
abline(h=mean(coeffs[,colnames(coeffs)=="Pop"]),
       col="red", lty=2) #Plot average fold
legend("topright",
       legend=c("Value", "Cross-fold Mean"),
       col=c("black", "red"),
       lwd=1, lty=c(1,2), cex=0.7)
ybounds <- c(min(coeffs[,colnames(coeffs)=="Year2012"],</pre>
                 coeffs[,colnames(coeffs)=="Year2013"],
                 coeffs[,colnames(coeffs)=="Year2014"],
                 coeffs[,colnames(coeffs) == "Year2015"]),
             max(coeffs[,colnames(coeffs)=="Year2012"],
                 coeffs[,colnames(coeffs)=="Year2013"],
                 coeffs[,colnames(coeffs)=="Year2014"],
                 coeffs[,colnames(coeffs)=="Year2015"]))
plot(coeffs[,colnames(coeffs)=="Year2012"], type="b",main="Year Coefficients by Fold",
     xlab="Fold",
     ylab="Year Coefficient",
     ylim=ybounds,
     col="red")
abline(h=mean(coeffs[,colnames(coeffs)=="Year2012"]),col="red",lty=2) #Plot average fold
lines(coeffs[,colnames(coeffs)=="Year2013"], type="b",col="blue")
abline(h=mean(coeffs[,colnames(coeffs)=="Year2013"]),col="blue",lty=2) #Plot average fold
lines(coeffs[,colnames(coeffs)=="Year2014"], type="b",col="green")
abline(h=mean(coeffs[,colnames(coeffs)=="Year2014"]),col="green",lty=2) #Plot average fold
lines(coeffs[,colnames(coeffs)=="Year2015"], type="b",col="purple")
abline(h=mean(coeffs[,colnames(coeffs)=="Year2015"]),col="purple",lty=2) #Plot average fold
legend(x=7.5, y=3000,
       legend=c("2012 Value","2012 Cross-fold Mean",
                "2013 Value", "2013 Cross-fold Mean",
                "2014 Value", "2014 Cross-fold Mean",
                "2015 Value", "2015 Cross-fold Mean"),
       col=c("red","red","blue","blue","green","green","purple","purple"),
       lwd=1, lty=c(1,2,1,2,1,2,1,2), cex=0.7)
```



## Year Coefficients by Fold

