

Massachusetts_Case

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Case 1: Massachusetts Data Security Law

```
library(plyr)
library(here)

## here() starts at C:/Users/karl_000/Documents/SpiderOak Hive/Dissertation/Dissertation_Code
##
## Attaching package: 'here'

## The following object is masked from 'package:plyr':
##
##     here
library(tidyr)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:plyr':
##
##     arrange, count, desc, failwith, id, mutate, rename, summarise,
##     summarise
## The following objects are masked from 'package:stats':
##
##     filter, lag
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
library(lubridate)

##
## Attaching package: 'lubridate'

## The following objects are masked from 'package:base':
##
##     date, intersect, setdiff, union
library(tseries)

## Registered S3 method overwritten by 'quantmod':
##   method           from
```

```

##   as.zoo.data.frame zoo
library(TTR)

AllStateClean <- read.table(here::here("Data","Other_data","AllStateClean.txt"),sep=";")

AllStateClean$MA_affected_residents <- gsub("\\,.\\*", "", AllStateClean$MA_affected_residents) # this sele
AllStateClean$NC_affected_residents <- gsub("\\,.\\*", "", AllStateClean$NC_affected_residents) # this sele

AllStateClean$Massachusetts[!is.na(AllStateClean$Massachusetts)]<-1
AllStateClean$New_Hampshire[!is.na(AllStateClean$New_Hampshire)]<-1
AllStateClean$North_Carolina[!is.na(AllStateClean$North_Carolina)]<-1
AllStateClean$Massachusetts[is.na(AllStateClean$Massachusetts)]<-0
AllStateClean$New_Hampshire[is.na(AllStateClean$New_Hampshire)]<-0
AllStateClean$North_Carolina[is.na(AllStateClean$North_Carolina)]<-0

```

Produce Tables on Overlapping Co-occurrence of Breach Incidents

```

CoveredDays <- AllStateClean

CoveredDays$reported_date <- substr(CoveredDays$reported_date, start=1, stop=10)

CoveredDays <- subset(CoveredDays, reported_date > as.Date("2008-03-22") )
CoveredDays <- subset(CoveredDays, reported_date < as.Date("2010-09-01") )

table(CoveredDays$Massachusetts) # 0 = 313, 1 = 858

##
##    0    1
## 313 858

table(CoveredDays$New_Hampshire) # 0 = 949, 1 = 222

##
##    0    1
## 949 222

table(CoveredDays$North_Carolina) # 0 = 819, 1 = 352

##
##    0    1
## 819 352

table(CoveredDays$Massachusetts,CoveredDays$New_Hampshire) # 0 = 205 + 108 ; 1 = 744 + 114

##
##          0    1
##    0 205 108
##    1 744 114

table(CoveredDays$Massachusetts,CoveredDays$North_Carolina) # 0 = 99 + 214 ; 1 = 720 + 138

##
##          0    1
##    0 99 214

```

```

##    1 720 138

CoveredDays$MA_affected_residents <- as.numeric(CoveredDays$MA_affected_residents)
CoveredDays$NC_affected_residents <- as.numeric(CoveredDays$NC_affected_residents)

CoveredDays$MA_BigBreach <- CoveredDays$MA_affected_residents > 1000
CoveredDays$NC_BigBreach <- CoveredDays$NC_affected_residents > 1000

# Need to resolve this R issue to get to a fixed solution

# table(CoveredDays$MA_BigBreach,CoveredDays$NC_BigBreach)

```

Create Population Time Series for Matching with Incident Frequency

```

# Creating blank frequency starting with earliest date

dat2 <- data.frame(seq(as.Date("2006-06-01"), by="1 month", length.out=174)) # treatment date
names(dat2) <- "yearmonth"
dat2 <- format(dat2, "%Y/%m")

# Population
pop <- read.csv(here::here("Data","Other_data","populations.csv")) # starts at 2000.04.01
pop <- pop[c(22,30,34),] # Massachusetts row 22, New Hampshire row 30, North Carolina row 34 # removes

datforpop <- data.frame(seq(as.Date("2000-04-01"), by="1 month", length.out=(length(pop)-1)))
names(datforpop) <- "yearmonth"
datforpop <- format(datforpop, "%Y/%m")
datforpop <- rbind("yearmonth",datforpop)
row.names(datforpop) <- 1:nrow(datforpop)

pop <- cbind(datforpop,t(pop))
colnames(pop) <- pop[1,]
pop <- pop[-1,]
rownames(pop) <- seq(1:nrow(pop))
pop <- as.data.frame(pop)

```

Identifying treatment and control options

```

# Case 1: Experiment 1
massachusetts <- dplyr::filter(AllStateClean,Massachusetts==1)
massachusetts$reported_date <- substr(massachusetts$reported_date, start=1, stop=10)

new_hampshire <- dplyr::filter(AllStateClean,New_Hampshire==1)
new_hampshire$reported_date <- substr(new_hampshire$reported_date, start=1, stop=10)

# Case 1: Experiment 2
massachusetts$MA_affected_residents <- as.numeric(massachusetts$MA_affected_residents)
massachusetts1000 <- subset(massachusetts, massachusetts$MA_affected_residents > 1000)

```

```

n_carolina <- dplyr::filter(AllStateClean, North_Carolina==1)
n_carolina$NC_affected_residents <- as.numeric(n_carolina$NC_affected_residents)
n_carolina1000 <- subset(n_carolina, n_carolina$NC_affected_residents > 1000)

```

Experiment 1: Create control and treatment populations (Massachusetts v New Hampshire) with Total Incidents

```

treatment <- massachusetts # This Must Be Filled in to Work Properly!
control <- new_hampshire # This Must Be Filled in to Work Properly!

# Format treatment dates into months
treatment$date_formatted <- format(as.Date(treatment$reported_date, "%Y-%m-%d"), "%Y/%m") # Alternative
treatment_freq <- treatment %>%
  dplyr::group_by(treatment$date_formatted) %>%
  dplyr::summarise(frequency = n(),)

## `summarise()` ungrouping output (override with ` `.groups` argument)
names(treatment_freq)[1] <- "yearmonth"
treatment_freq$frequency[is.na(treatment_freq$frequency)] <- 0

# Format control dates into months
control$date_formatted <- format(as.Date(control$reported_date, "%Y-%m-%d"), "%Y/%m") # Alternative is
control_freq <- control %>%
  dplyr::group_by(control$date_formatted) %>%
  dplyr::summarise(frequency = n(),)

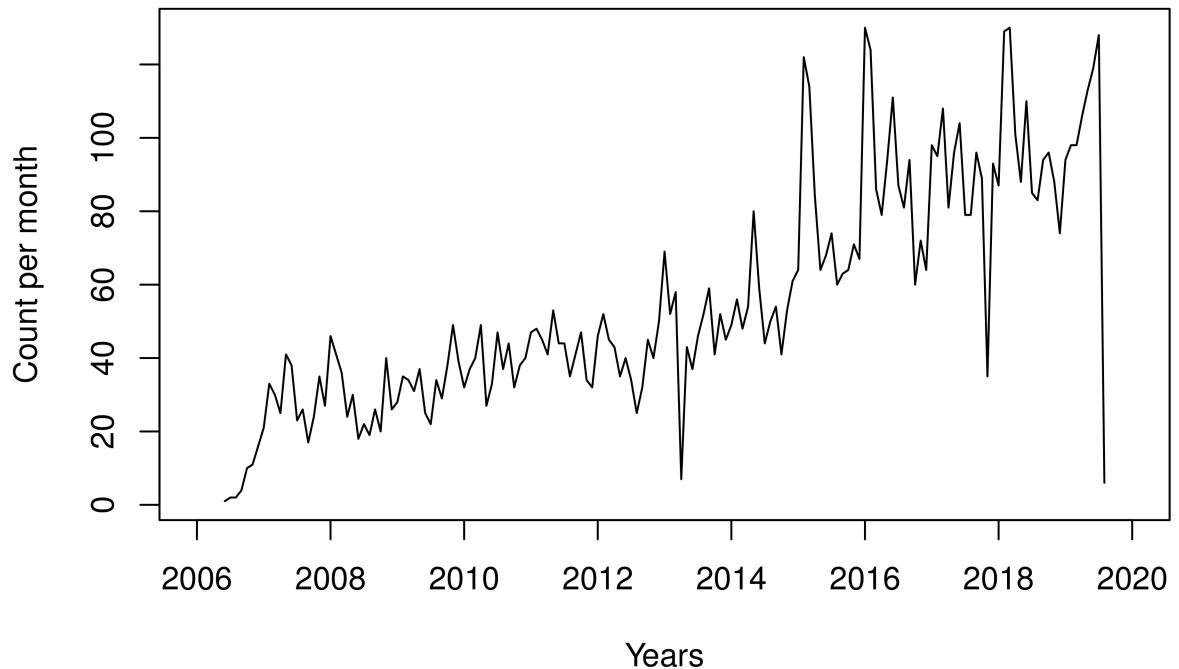
## `summarise()` ungrouping output (override with ` `.groups` argument)
names(control_freq)[1] <- "yearmonth"
control_freq$frequency[is.na(control_freq$frequency)] <- 0

treatment_ts <- ts(treatment_freq$frequency, frequency = 12, start = c(2006,6))
control_ts <- ts(control_freq$frequency, frequency = 12, start = c(2006,6))

plot.ts(treatment_ts, main = "Breaches over time", xlim=c(2006,2020), xlab = "Years", ylab = "Count per month")

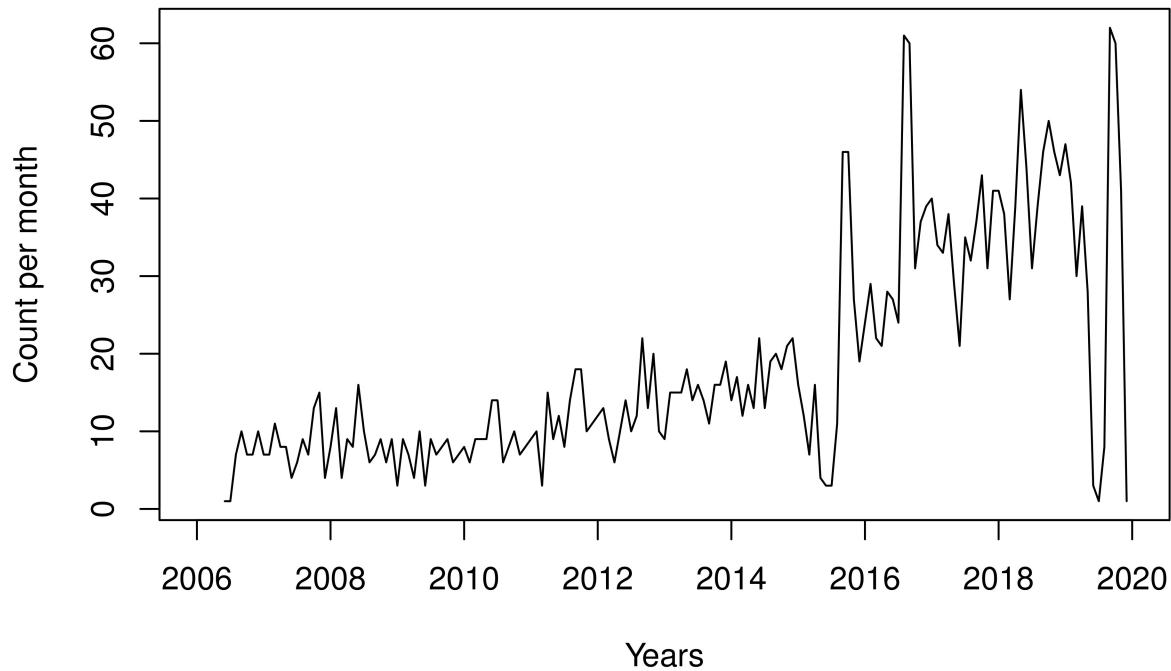
```

Breaches over time



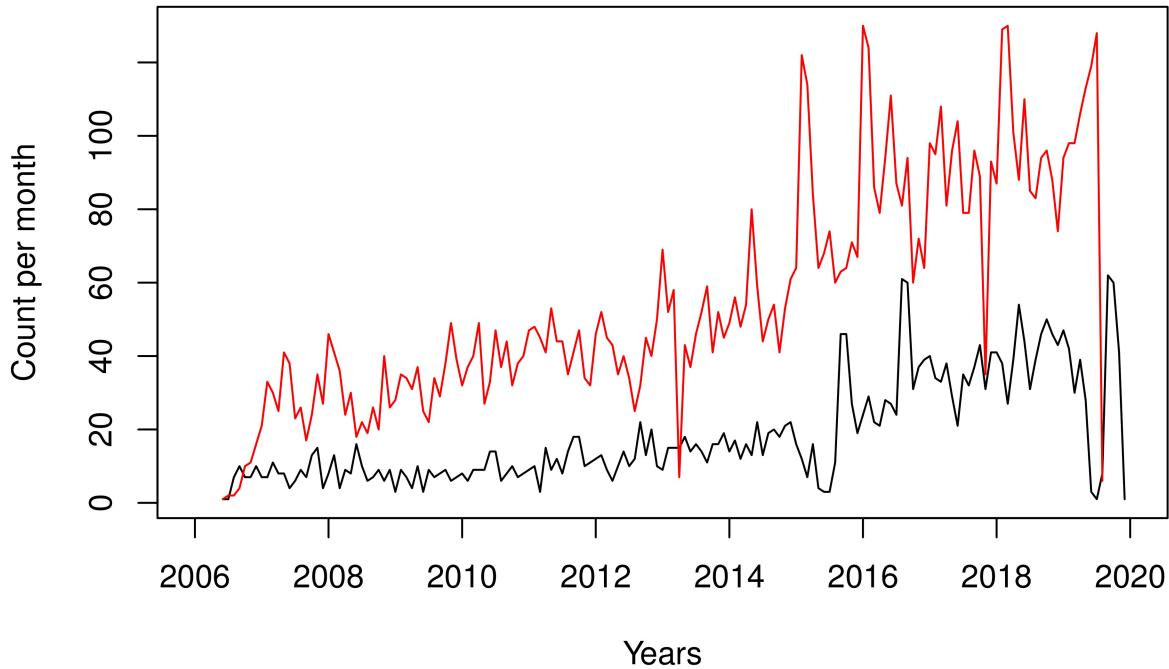
```
plot.ts(control_ts, main = "Breaches over time", xlim=c(2006,2020), xlab = "Years", ylab = "Count per m
```

Breaches over time



```
ts.plot(control_ts, treatment_ts, main = "Breaches over time", xlim=c(2006,2020), gpars = list(col = c(
```

Breaches over time



```
summary(treatment_ts)

##      Min. 1st Qu. Median      Mean 3rd Qu.      Max.
##     1.00   34.00  46.00    55.19  79.00  130.00

summary(control_ts)

##      Min. 1st Qu. Median      Mean 3rd Qu.      Max.
##     1.00   8.00  13.00   18.39  27.00  62.00
```

Create charts with breaches per million residents

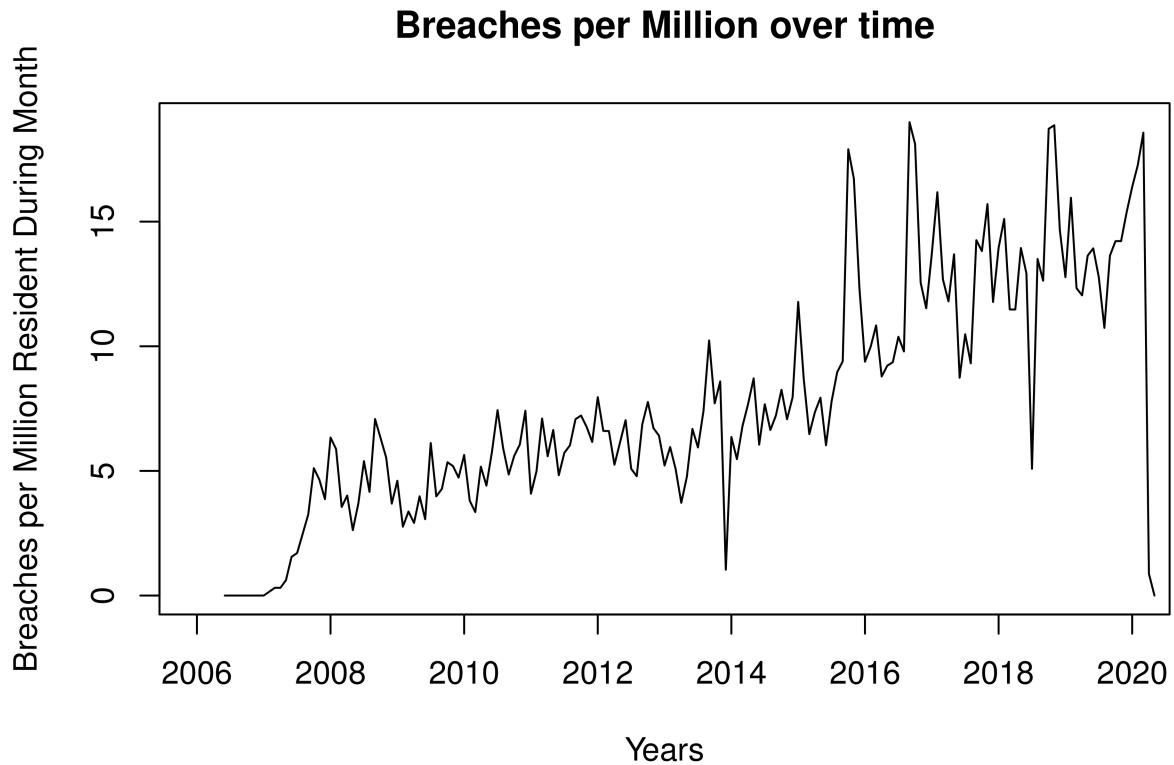
```
# Merge Treatment and Control Together
comb_ts <- merge(treatment_freq, control_freq, by="yearmonth", all=TRUE)

# Merge Combined Treatment and Control Together with Population Statistics
comb_ts <- merge(comb_ts, pop, by='yearmonth', all.x = TRUE)
comb_ts$frequency.x[is.na(comb_ts$frequency.x)]<-0
comb_ts$frequency.y[is.na(comb_ts$frequency.y)]<-0

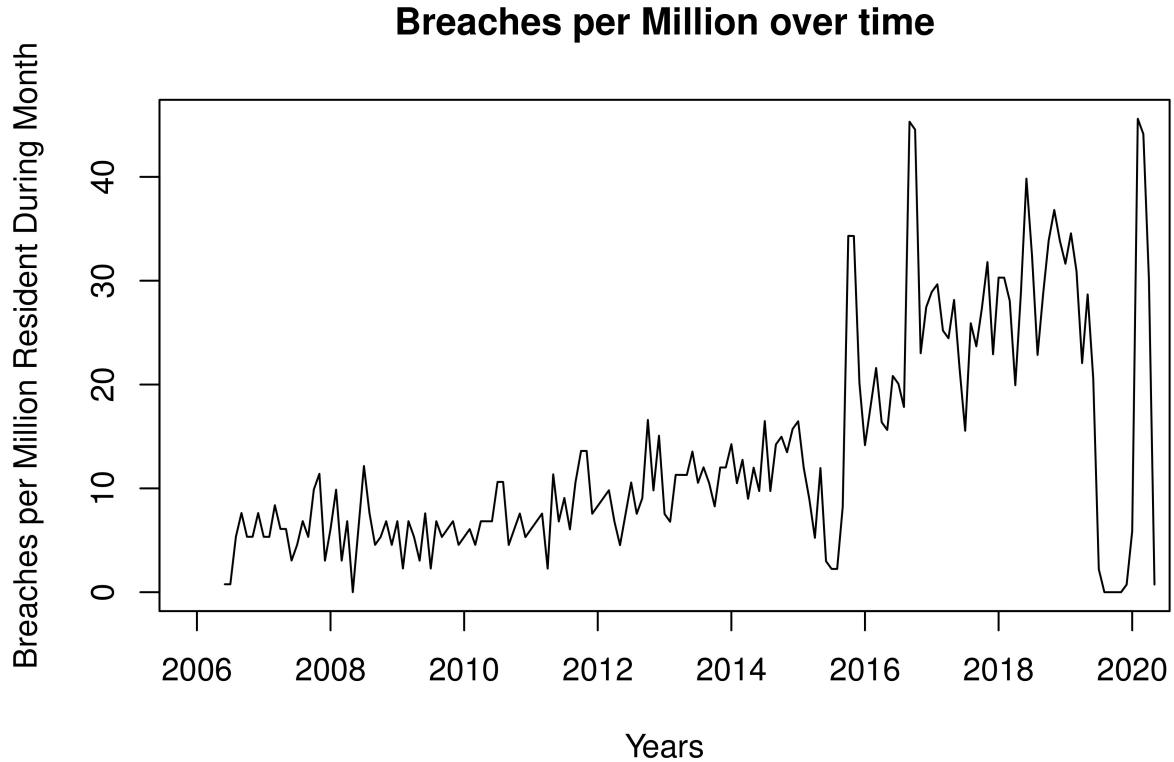
comb_ts$Massachusetts <- as.numeric(as.character(comb_ts$Massachusetts))
comb_ts$treatpermil <- comb_ts$frequency.x/(comb_ts$Massachusetts/1000000)
class(comb_ts$frequency.y)<"numeric"
comb_ts$`New Hampshire` <- as.numeric(as.character(comb_ts$`New Hampshire`))
comb_ts$controlpermil <- comb_ts$frequency.y/(comb_ts$`New Hampshire`/1000000)
```

```
treatment_tsM <- ts(comb_ts$treatpermil, frequency = 12, start = c(2006,6))
control_tsM <- ts(comb_ts$controlpermil, frequency = 12, start = c(2006,6))
```

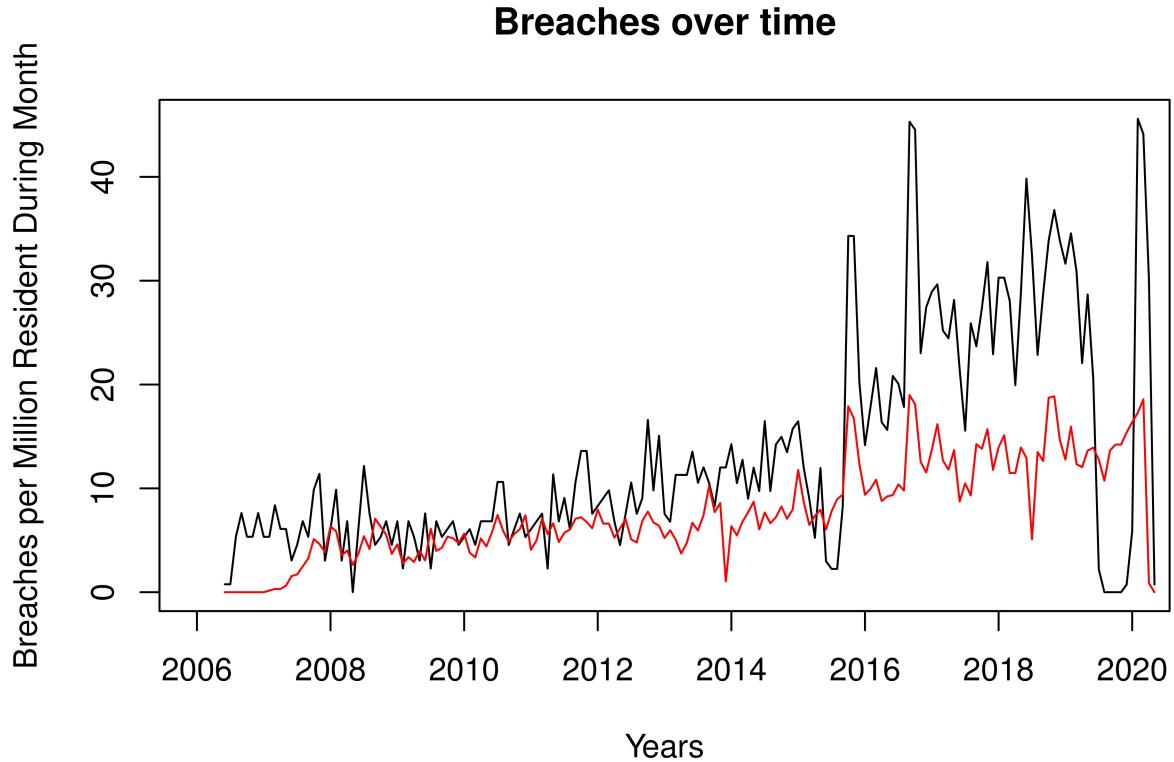
```
plot.ts(treatment_tsM, main = "Breaches per Million over time", xlim=c(2006,2020), xlab = "Years", ylab =
```



```
plot.ts(control_tsM, main = "Breaches per Million over time", xlim=c(2006,2020), xlab = "Years", ylab =
```



```
ts.plot(control_tsM, treatment_tsM, main = "Breaches over time", xlim=c(2006,2020), gpars = list(col = c
```



Identifying and subsetting relevant dates

```
# Legislation H.B. 4144 signed into law August 3, 2007
# Legislation H.B. 4144 becomes effective on October 31, 2007
# OCABR finalized the regulation on September 22, 2008

treatment_start <- as.Date("09/22/2008", "%m/%d/%Y") # Legislation H.B. 4144 becomes effective
treatment_start<- format(as.Date(as.character(treatment_start), origin = "1970-01-01"), "%Y/%m")

treatment_end <- as.Date("03/01/2010", "%m/%d/%Y") +180 # post 6 months after enforcement
treatment_end<- format(as.Date(as.character(treatment_end), origin = "1970-01-01"), "%Y/%m")

pretreat <- comb_ts[(which(comb_ts$yearmonth==treatment_start)-5):(which(comb_ts$yearmonth==treatment_start)+5)]
pretreat$type <- "pretest"

posttreat <- comb_ts[(which(comb_ts$yearmonth==treatment_end)): (which(comb_ts$yearmonth==treatment_end)+5)]
posttreat$type <- "posttest"

mean(posttreat$treatpermil) - mean(pretreat$treatpermil)

## [1] 0.6274941
mean(posttreat$controlpermil) - mean(pretreat$controlpermil)

## [1] 0.9999738
```

```

treatment_range <- comb_ts[(which(comb_ts$yearmonth==treatment_start)+1):(which(comb_ts$yearmonth==treatment_end))]
treatment_range$type <- "test"

experiment <- rbind(pretreat,treatment_range,posttreat)
experiment$treatpermil[is.na(experiment$treatpermil)]<-0
experiment$controlpermil[is.na(experiment$controlpermil)]<-0
experiment

##      yearmonth frequency.x frequency.y Massachusetts New Hampshire North Carolina
## 18    2008/04          30           15       6459615     1315064      9261596
## 19    2008/05          25            4       6462732     1315345      9277547
## 20    2008/06          41            8       6465850     1315626      9293498
## 21    2008/07          38           13       6468967     1315906      9309449
## 22    2008/08          23            4       6473021     1315922      9321125
## 23    2008/09          26            9       6477075     1315939      9332802
## 24    2008/10          17            0       6481128     1315955      9344478
## 25    2008/11          24            8       6485182     1315971      9356155
## 26    2008/12          35           16       6489236     1315988      9367831
## 27    2009/01          27           10       6493290     1316004      9379508
## 28    2009/02          46            6       6497344     1316020      9391184
## 29    2009/03          41            7       6501398     1316037      9402860
## 30    2009/04          36            9       6505452     1316053      9414537
## 31    2009/05          24            6       6509505     1316069      9426213
## 32    2009/06          30            9       6513559     1316086      9437890
## 33    2009/07          18            3       6517613     1316102      9449566
## 34    2009/08          22            9       6520948     1316143      9459112
## 35    2009/09          19            7       6524283     1316184      9468659
## 36    2009/10          26            4       6527618     1316225      9478205
## 37    2009/11          20           10       6530953     1316266      9487751
## 38    2009/12          40            3       6534289     1316306      9497298
## 39    2010/01          26            9       6537624     1316347      9506844
## 40    2010/02          28            7       6540959     1316388      9516390
## 41    2010/03          35            8       6544294     1316429      9525937
## 42    2010/04          34            9       6547629     1316470      9535483
## 43    2010/05          31            6       6553855     1316567      9548430
## 44    2010/06          37            7       6560081     1316665      9561376
## 45    2010/07          25            8       6566307     1316762      9574323
## 46    2010/08          22            6       6570247     1317049      9581262
## 47    2010/09          34            9       6574186     1317335      9588201
## 48    2010/10          29            9       6578126     1317622      9595140
## 49    2010/11          38            9       6582066     1317909      9602079
## 50    2010/12          49           14       6586005     1318195      9609018
## 51    2011/01          39           14       6589945     1318482      9615958

##      treatpermil controlpermil      type
## 18        4.644240    11.406289 pretest
## 19        3.868333    3.041027 pretest
## 20        6.341007    6.080755 pretest
## 21        5.874199    9.879125 pretest
## 22        3.553210    3.039694 pretest
## 23        4.014158    6.839223 pretest
## 24        2.623000    0.000000    test
## 25        3.700744    6.079161    test
## 26        5.393547   12.158166    test
## 27        4.158139    7.598761    test

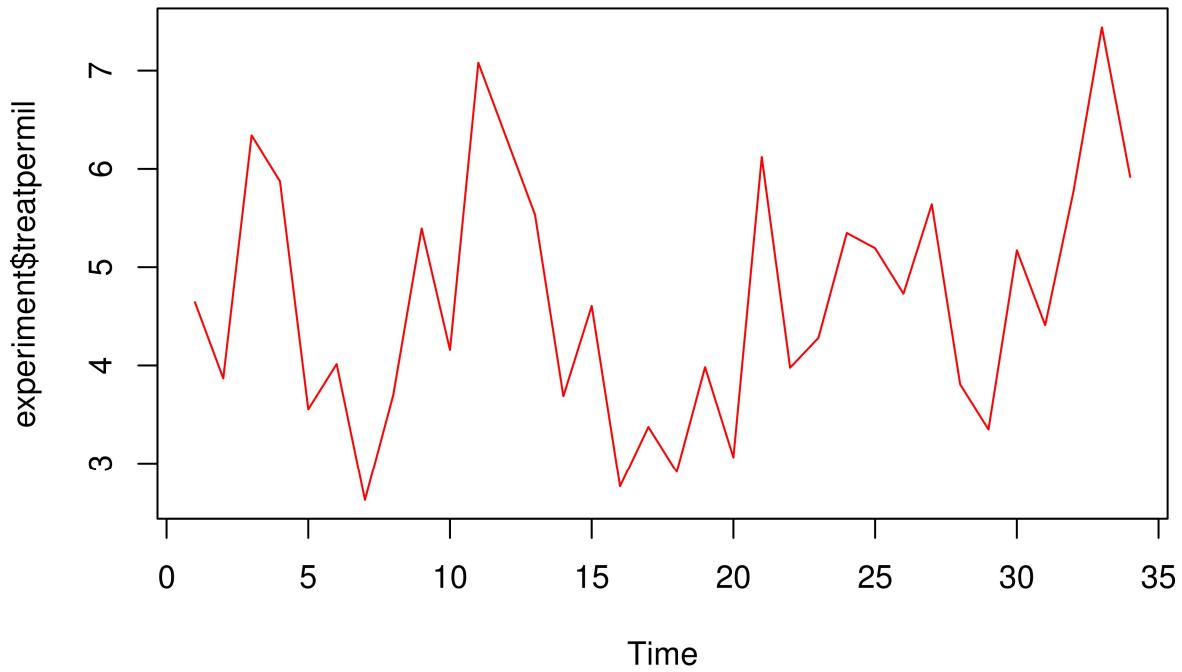
```

```

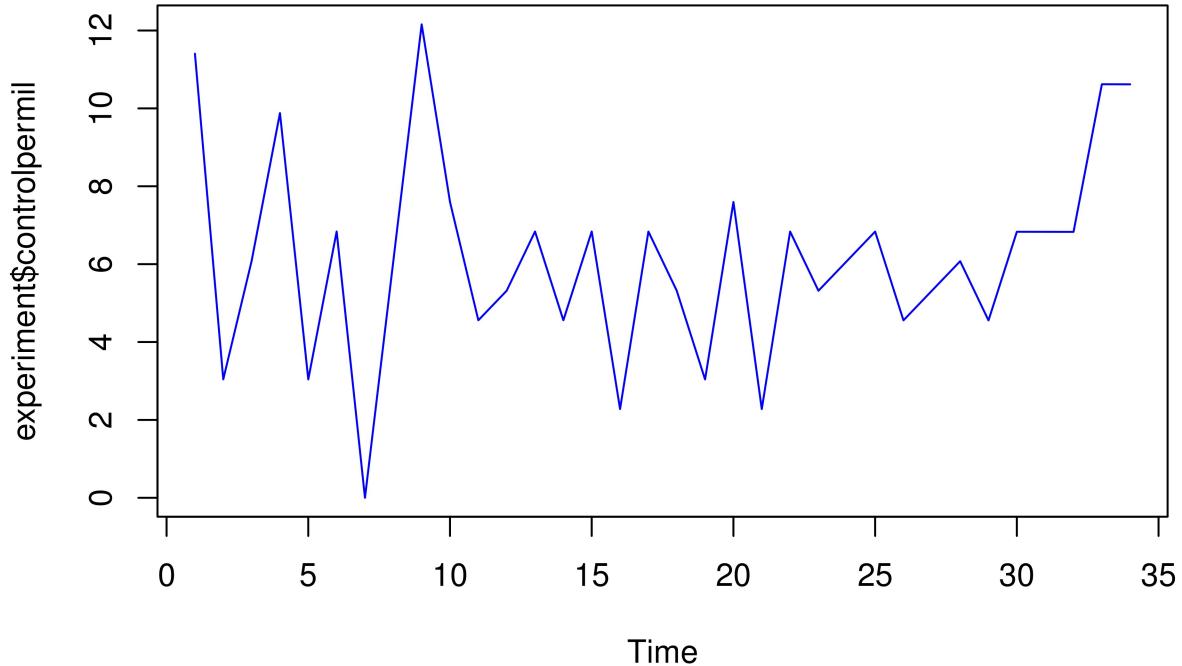
## 28    7.079816    4.559201    test
## 29    6.306336    5.318999    test
## 30    5.533820    6.838630    test
## 31    3.686916    4.559031    test
## 32    4.605777    6.838459    test
## 33    2.761747    2.279459    test
## 34    3.373743    6.838163    test
## 35    2.912197    5.318405    test
## 36    3.983076    3.038994    test
## 37    3.062340    7.597249    test
## 38    6.121554    2.279105    test
## 39    3.976980    6.837103    test
## 40    4.280718    5.317581    test
## 41    5.348170    6.077046    test
## 42    5.192719    6.836464    test
## 43    4.730041    4.557307    test
## 44    5.640174    5.316462    test
## 45    3.807315    6.075509    test
## 46    3.348428    4.555639    posttest
## 47    5.171743    6.831975    posttest
## 48    4.408550    6.830487    posttest
## 49    5.773263    6.829000    posttest
## 50    7.440019    10.620583   posttest
## 51    5.918107    10.618272   posttest

```

```
ts.plot(experiment$treatpermil, col = "red")
```



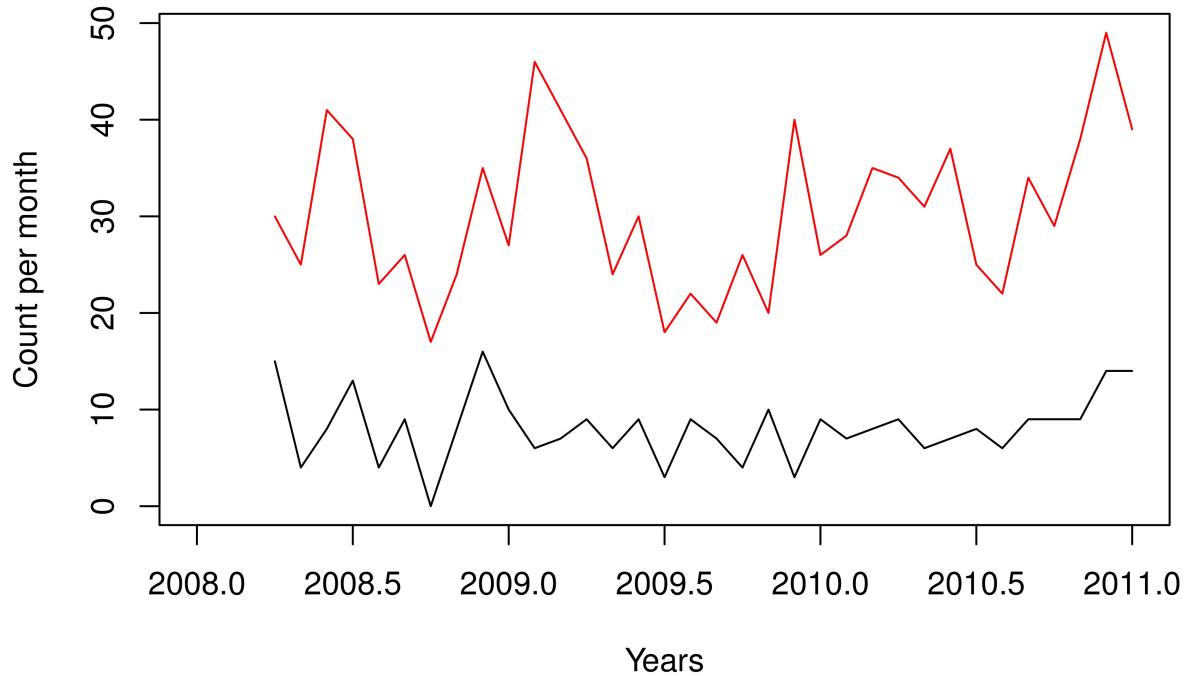
```
ts.plot(experiment$controlpermil, col = "blue")
```



```
# Look at Raw Frequency Counts
```

```
treatment_ts <- ts(experiment$frequency.x, frequency = 12, start = c(2008,4))
control_ts <- ts(experiment$frequency.y, frequency = 12, start = c(2008,4))
ts.plot(control_ts, treatment_ts, main = "Breaches over time", xlim=c(2008,2011), gpars = list(col = c(
```

Breaches over time



```
# Look at Treatment and Control per Million

treatment_tsM <- ts(experiment$treatpermil, frequency = 12, start = c(2008,4))
mean(treatment_tsM)

## [1] 4.66571
sd(treatment_tsM)

## [1] 1.253152

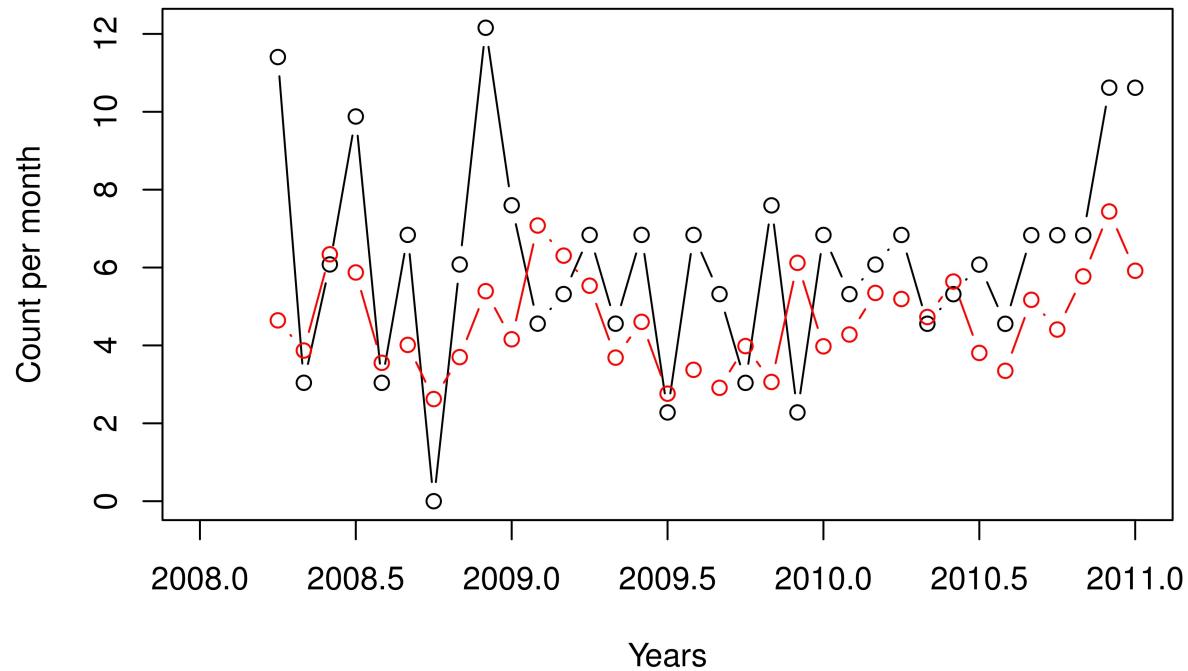
control_tsM <- ts(experiment$controlpermil, frequency = 12, start = c(2008,4))
mean(control_tsM)

## [1] 6.143863
sd(control_tsM)

## [1] 2.672772

ts.plot(control_tsM, treatment_tsM, main = "Breaches over time", xlim=c(2008,2011),
        gpars = list(col = c("black", "red")), type = "b", xlab = "Years", ylab = "Count pe
```

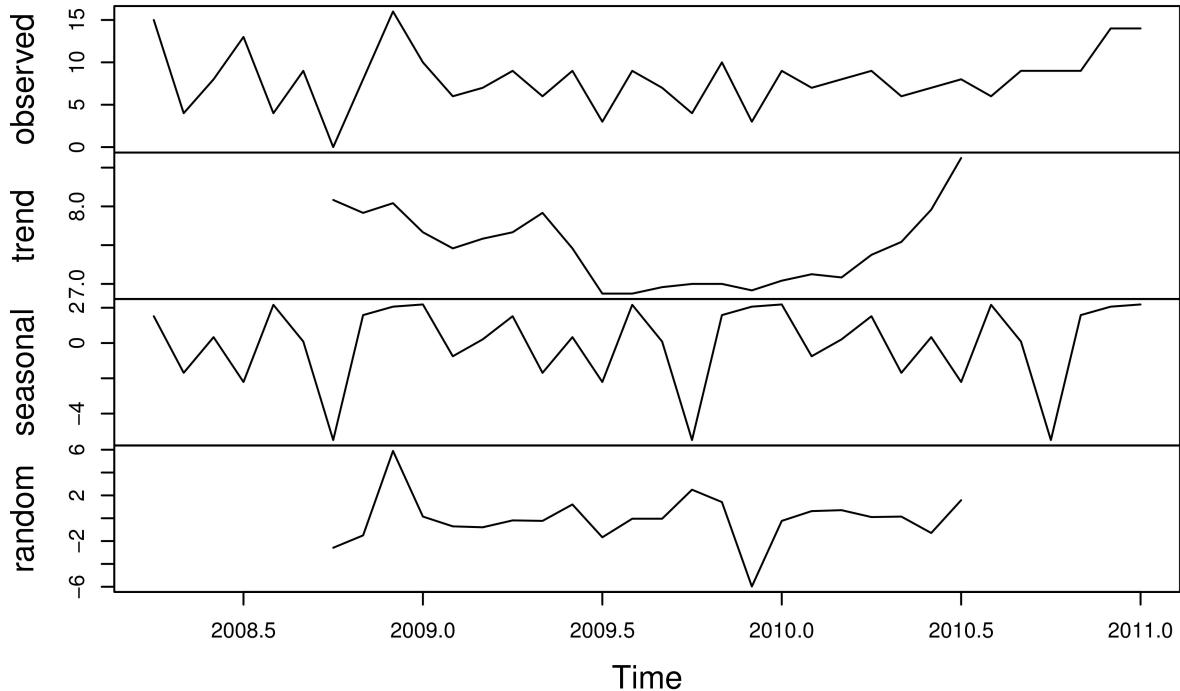
Breaches over time



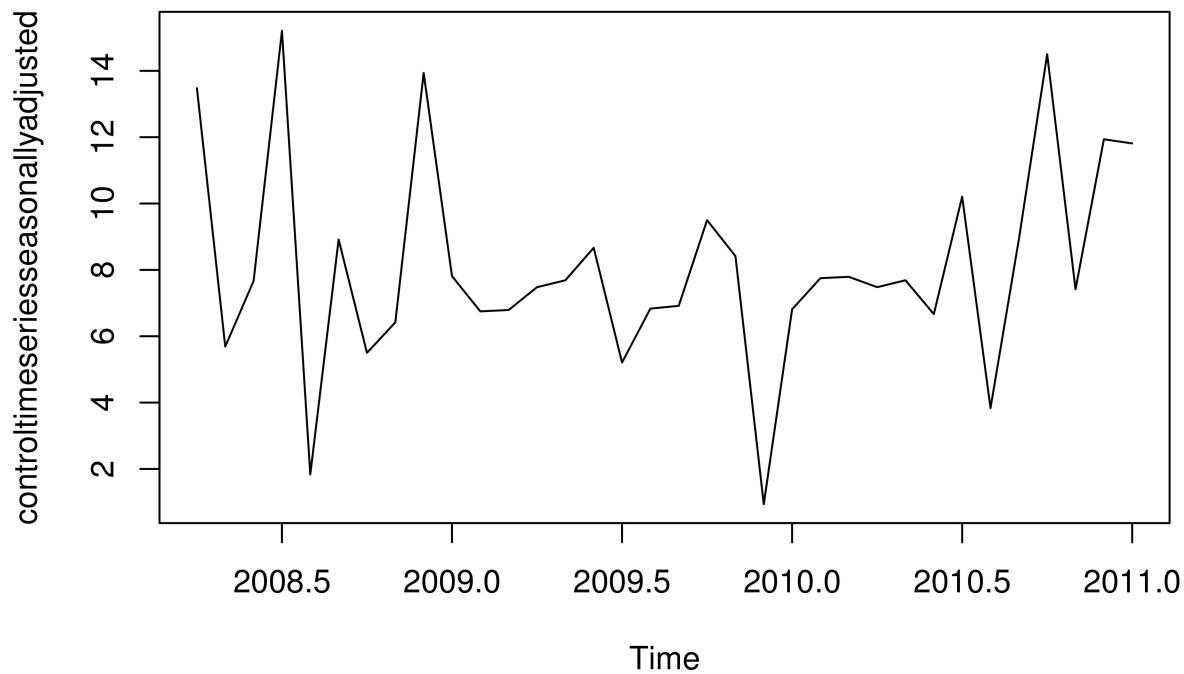
Decompose the Massachusetts Data to Find Seasonal Patterns

```
controltimeseriescomponents <- decompose(control_ts)
plot(controltimeseriescomponents)
```

Decomposition of additive time series



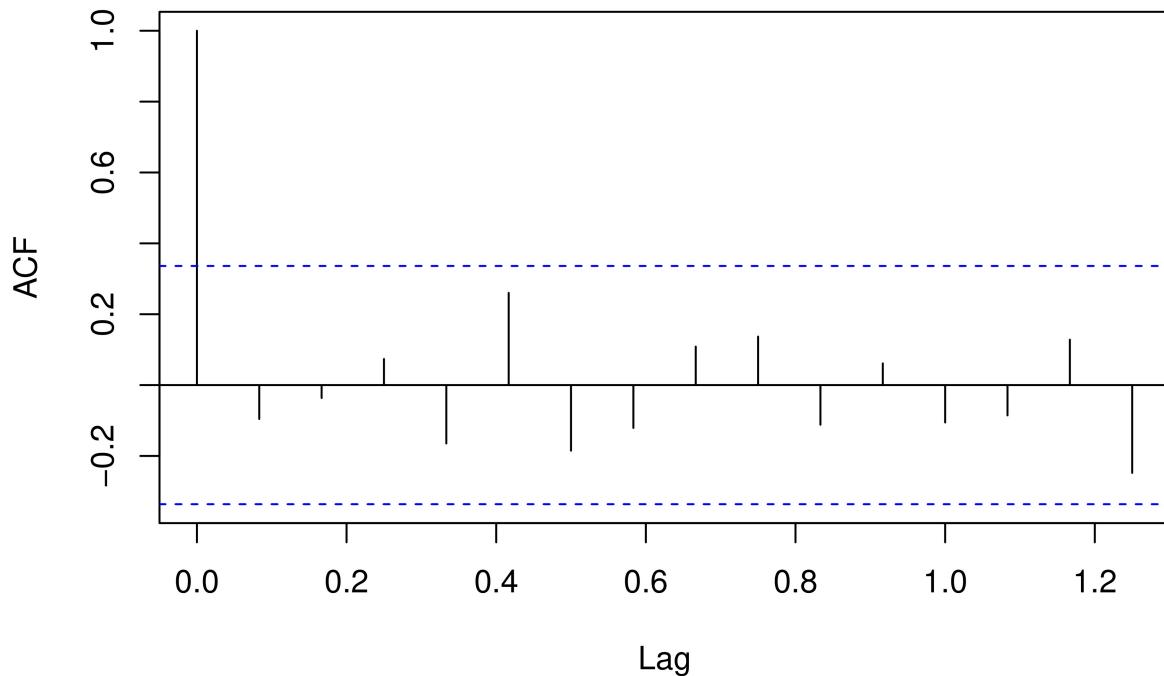
```
controltimeseriesseasonallyadjusted <- control_ts - controltimeseriescomponents$seasonal  
plot(controltimeseriesseasonallyadjusted)
```



Run Statistical Tests on Time Series for Stationarity

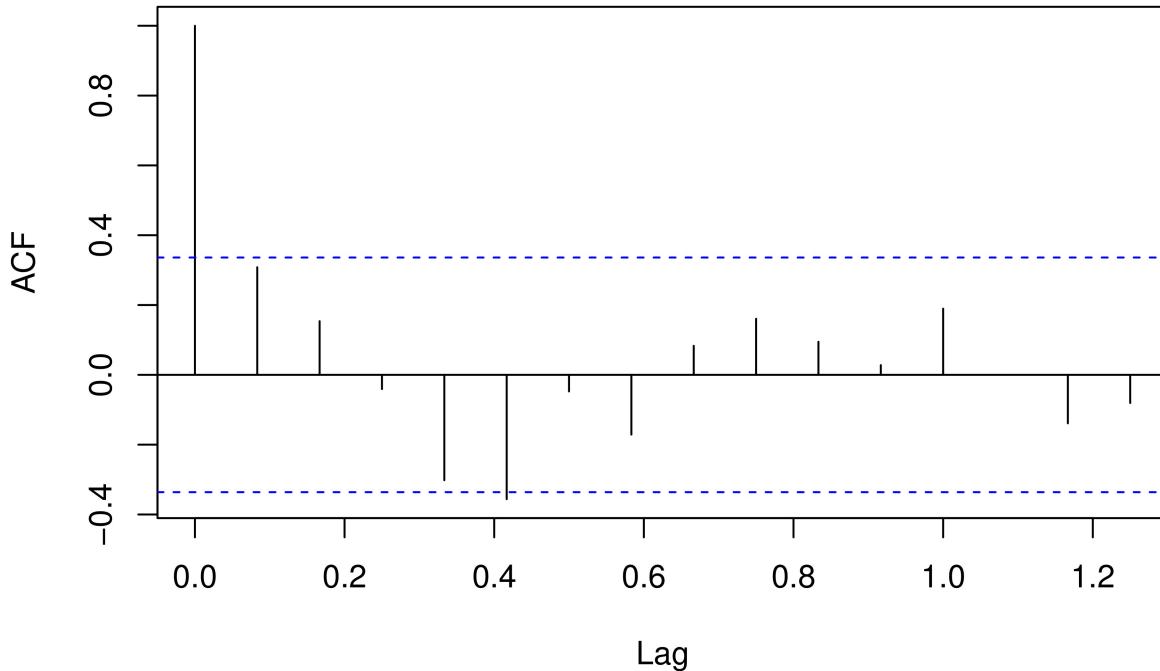
```
# source of statistical tests http://r-statistics.co/Time-Series-Analysis-With-R.html
acfcontrolNH <- acf(control_ts) # autocorrelation (i.e. a Time Series with lags of itself)
```

Series control_ts



```
acftreatmentMA <- acf(treatment_ts)
```

Series treatment_ts



```
# shows that the control time series is a "stationary time series"

png(here::here("Output","acfcontrolNH.png"))
plot(acfcontrolNH)

png(here::here("Output","acftreatmentMA.png"))
plot(acftreatmentMA)

pacfcontrolNH <- pacf(control_ts) # partial autocorrelation (i.e. correlation of the time series with
pacftreatmentMA <- pacf(treatment_ts) # partial autocorrelation (i.e. correlation of the time series w

plot(pacfcontrolNH)

png(here::here("Output","pacftreatmentNH.png"))
plot(pacftreatmentMA)

ccfRes <- ccf(control_ts, treatment_ts, ylab = "cross-correlation")
ccfRes

##  
## Autocorrelations of series 'X', by lag  
##  
## -1.0000 -0.9167 -0.8333 -0.7500 -0.6667 -0.5833 -0.5000 -0.4167 -0.3333 -0.2500  
## 0.094 0.055 0.060 -0.059 0.009 -0.098 -0.229 -0.117 -0.145 0.200  
## -0.1667 -0.0833 0.0000 0.0833 0.1667 0.2500 0.3333 0.4167 0.5000 0.5833  
## 0.295 0.074 0.412 0.267 -0.074 -0.045 -0.142 0.006 0.148 0.026  
## 0.6667 0.7500 0.8333 0.9167 1.0000
```

```

## -0.068 0.177 -0.038 0.075 0.024
# adf test is an Augmented Dickey-Fuller Test
adf.test(control_ts) # p-value < 0.05 indicates the TS is stationary

##
## Augmented Dickey-Fuller Test
##
## data: control_ts
## Dickey-Fuller = -2.7292, Lag order = 3, p-value = 0.2905
## alternative hypothesis: stationary
adf.test(treatment_ts)

##
## Augmented Dickey-Fuller Test
##
## data: treatment_ts
## Dickey-Fuller = -3.5501, Lag order = 3, p-value = 0.05258
## alternative hypothesis: stationary
kpss.test(control_ts) # Kwiatkowski-Phillips-Schmidt-Shin (KPSS) testz

## Warning in kpss.test(control_ts): p-value greater than printed p-value

##
## KPSS Test for Level Stationarity
##
## data: control_ts
## KPSS Level = 0.16838, Truncation lag parameter = 3, p-value = 0.1
kpss.test(treatment_ts)

## Warning in kpss.test(treatment_ts): p-value greater than printed p-value

##
## KPSS Test for Level Stationarity
##
## data: treatment_ts
## KPSS Level = 0.14572, Truncation lag parameter = 3, p-value = 0.1
# https://www.sas.com/content/dam/SAS/en\_ca/User%20Group%20Presentations/Health-User-Groups/ITS\_SAS.pdf

```

ITS analyses use regression-based techniques

```

quasiexp <- experiment[experiment$type != "test",]
quasiexp

##      yearmonth frequency.x frequency.y Massachusetts New Hampshire North Carolina
## 18    2008/04          30           15     6459615     1315064     9261596
## 19    2008/05          25            4     6462732     1315345     9277547
## 20    2008/06          41            8     6465850     1315626     9293498
## 21    2008/07          38           13     6468967     1315906     9309449
## 22    2008/08          23            4     6473021     1315922     9321125
## 23    2008/09          26            9     6477075     1315939     9332802
## 46    2010/08          22            6     6570247     1317049     9581262
## 47    2010/09          34            9     6574186     1317335     9588201

```

```

## 48 2010/10 29 9 6578126 1317622 9595140
## 49 2010/11 38 9 6582066 1317909 9602079
## 50 2010/12 49 14 6586005 1318195 9609018
## 51 2011/01 39 14 6589945 1318482 9615958
##   treatpermil controlpermil type
## 18 4.644240 11.406289 pretest
## 19 3.868333 3.041027 pretest
## 20 6.341007 6.080755 pretest
## 21 5.874199 9.879125 pretest
## 22 3.553210 3.039694 pretest
## 23 4.014158 6.839223 pretest
## 46 3.348428 4.555639 posttest
## 47 5.171743 6.831975 posttest
## 48 4.408550 6.830487 posttest
## 49 5.773263 6.829000 posttest
## 50 7.440019 10.620583 posttest
## 51 5.918107 10.618272 posttest

# Added dummy variables for ITS
control <- as.data.frame(t(rbind(quasiexp$yearmonth, quasiexp$controlpermil)))
control$treat <- as.vector(rep(0, nrow(control))) # Create example vector
time <- 1:nrow(control)
control$time <- as.vector(time)
control$z <- c(rep(0, 6), 1:(nrow(control)-6))

treatment <- as.data.frame(t(rbind(quasiexp$yearmonth, quasiexp$treatpermil)))
treatment$treat <- as.vector(rep(1, nrow(control))) # Create example vector
time <- 1:nrow(control)
treatment$time <- as.vector(time)
treatment$z <- c(rep(0, 6), 1:(nrow(control)-6))
treatment

##          V1          V2 treat time z
## 1 2008/04 4.64423963347661    1   1 0
## 2 2008/05 3.86833308266535    1   2 0
## 3 2008/06 6.34100698284062    1   3 0
## 4 2008/07 5.87419908000767    1   4 0
## 5 2008/08 3.55320954466238    1   5 0
## 6 2008/09 4.01415762516259    1   6 0
## 7 2010/08 3.34842814889608    1   7 1
## 8 2010/09 5.17174293517099    1   8 2
## 9 2010/10 4.40855039870018    1   9 3
## 10 2010/11 5.77326328845685   1  10 4
## 11 2010/12 7.44001864559775   1  11 5
## 12 2011/01 5.91810705552171   1  12 6

AppendITS <- rbind(treatment, control)
names(AppendITS) <- c("yearmonth", "incident_permil", "treat", "time", "z")
AppendITS$incident_permil <- as.numeric(as.character(AppendITS$incident_permil))
AppendITS$time <- as.numeric(as.character(AppendITS$time))
AppendITS$z <- as.numeric(as.character(AppendITS$z))
AppendITS

##      yearmonth incident_permil treat time z
## 1 2008/04        4.644240    1   1 0
## 2 2008/05        3.868333    1   2 0

```

```

## 3 2008/06 6.341007 1 3 0
## 4 2008/07 5.874199 1 4 0
## 5 2008/08 3.553210 1 5 0
## 6 2008/09 4.014158 1 6 0
## 7 2010/08 3.348428 1 7 1
## 8 2010/09 5.171743 1 8 2
## 9 2010/10 4.408550 1 9 3
## 10 2010/11 5.773263 1 10 4
## 11 2010/12 7.440019 1 11 5
## 12 2011/01 5.918107 1 12 6
## 13 2008/04 11.406289 0 1 0
## 14 2008/05 3.041027 0 2 0
## 15 2008/06 6.080755 0 3 0
## 16 2008/07 9.879125 0 4 0
## 17 2008/08 3.039694 0 5 0
## 18 2008/09 6.839223 0 6 0
## 19 2010/08 4.555639 0 7 1
## 20 2010/09 6.831975 0 8 2
## 21 2010/10 6.830487 0 9 3
## 22 2010/11 6.829000 0 10 4
## 23 2010/12 10.620583 0 11 5
## 24 2011/01 10.618272 0 12 6

factor_cols <- c("treat", "time", "z")

sapply(AppendITS, class)

##          yearmonth incident_permil          treat           time             z
##       "character"      "numeric"      "numeric"      "numeric"      "numeric"

regTest <- lm(incident_permil ~ time + treat + z, AppendITS)
summary(regTest)

## 
## Call:
## lm(formula = incident_permil ~ time + treat + z, data = AppendITS)
## 
## Residuals:
##    Min     1Q     Median      3Q     Max 
## -4.3229 -1.0957 -0.1703  1.1921  3.5534 
## 
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)    
## (Intercept)  8.3420    1.2881   6.476  2.58e-06 *** 
## time        -0.4890    0.2846  -1.719   0.1011    
## treat       -2.1847    0.7973  -2.740   0.0126 *  
## z            1.1721    0.4620   2.537   0.0196 *  
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 1.953 on 20 degrees of freedom
## Multiple R-squared:  0.4465, Adjusted R-squared:  0.3634 
## F-statistic: 5.377 on 3 and 20 DF,  p-value: 0.007037

regTest2 <- lm(incident_permil ~ time + treat + time*treat + z + z*time + z*treat + z*treat*time, AppendITS)
summary(regTest2)

```

```

## 
## Call:
## lm(formula = incident_permil ~ time + treat + time * treat +
##      z + z * time + z * treat + z * treat * time, data = AppendITS)
##
## Residuals:
##    Min     1Q Median     3Q    Max 
## -4.5044 -0.8769 -0.2531  1.2697  3.5121 
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)    
## (Intercept)  8.7239    1.9058   4.578  0.00031 ***  
## time        -0.5892    0.4735  -1.244  0.23131    
## treat       -3.3444    2.6952  -1.241  0.23254    
## z           -0.1885    2.7195  -0.069  0.94559    
## time:treat  0.3700    0.6697   0.553  0.58824    
## time:z       0.1466    0.2084   0.703  0.49190    
## treat:z      0.4889    3.8460   0.127  0.90043    
## time:treat:z -0.1162    0.2947  -0.394  0.69858  
## --- 
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 
##
## Residual standard error: 2.083 on 16 degrees of freedom
## Multiple R-squared:  0.4961, Adjusted R-squared:  0.2756 
## F-statistic:  2.25 on 7 and 16 DF,  p-value: 0.08494

```

AppendITS

	yearmonth	incident_permil	treat	time	z
## 1	2008/04	4.644240	1	1	0
## 2	2008/05	3.868333	1	2	0
## 3	2008/06	6.341007	1	3	0
## 4	2008/07	5.874199	1	4	0
## 5	2008/08	3.553210	1	5	0
## 6	2008/09	4.014158	1	6	0
## 7	2010/08	3.348428	1	7	1
## 8	2010/09	5.171743	1	8	2
## 9	2010/10	4.408550	1	9	3
## 10	2010/11	5.773263	1	10	4
## 11	2010/12	7.440019	1	11	5
## 12	2011/01	5.918107	1	12	6
## 13	2008/04	11.406289	0	1	0
## 14	2008/05	3.041027	0	2	0
## 15	2008/06	6.080755	0	3	0
## 16	2008/07	9.879125	0	4	0
## 17	2008/08	3.039694	0	5	0
## 18	2008/09	6.839223	0	6	0
## 19	2010/08	4.555639	0	7	1
## 20	2010/09	6.831975	0	8	2
## 21	2010/10	6.830487	0	9	3
## 22	2010/11	6.829000	0	10	4
## 23	2010/12	10.620583	0	11	5
## 24	2011/01	10.618272	0	12	6

```
# View(appendITS)
```