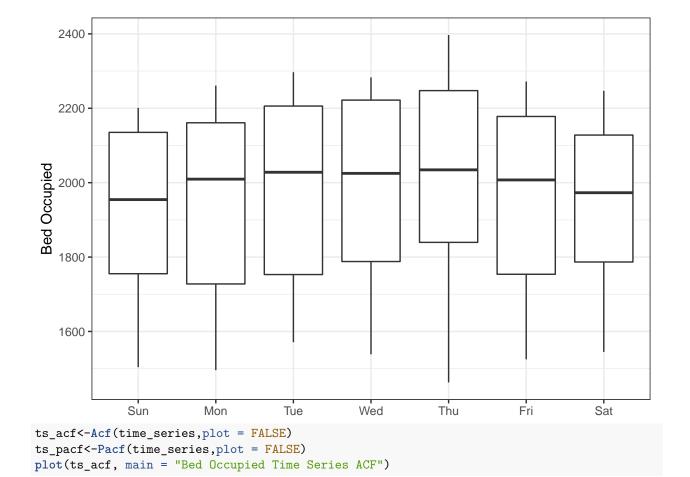
real

Joshua

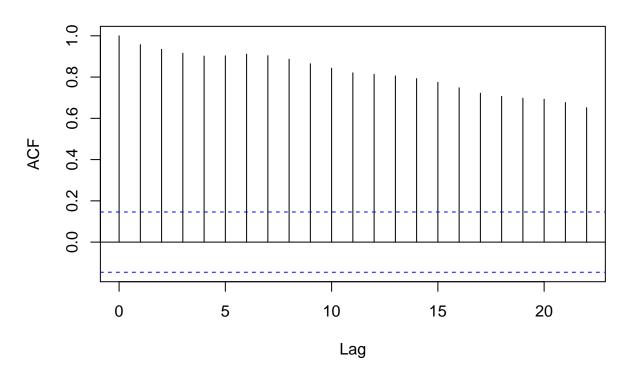
```
rm(list=ls())
library('knitr')
## Warning: package 'knitr' was built under R version 3.5.2
library('kableExtra')
## Warning: package 'kableExtra' was built under R version 3.5.2
library('forecast')
## Warning: package 'forecast' was built under R version 3.5.2
library('smooth')
## Warning: package 'smooth' was built under R version 3.5.2
## Loading required package: greybox
## Warning: package 'greybox' was built under R version 3.5.2
## Package "greybox", v0.5.8 loaded.
## This is package "smooth", v2.5.5
library('beamplot')
library('pastecs')
library('scales')
library('rbenchmark')
library('ggplot2')
library('readxl')
## Warning: package 'readxl' was built under R version 3.5.2
source<-read_excel('Covid.xlsx',2,range = cell_rows(151),col_names = FALSE)</pre>
## New names:
## * `` -> ...1
## * `` -> ...2
## * `` -> ...3
## * `` -> ...4
## * `` -> ...5
## * ... and 181 more problems
time_series<-ts(as.numeric(source[4:183]),frequency = 7, start = c(1, 5))</pre>
time_series<-tsclean(time_series)</pre>
horizon<-100
n<-80
ts_plot_season <- function(x = x) {</pre>
season <- cycle(x)</pre>
```

```
season.factor <- factor(season)</pre>
ggplot() +
  geom_boxplot(mapping = aes(x = season.factor,
                            y = x)) +
 labs(x = "", y = "Bed Occupied") +
 scale_x_discrete(labels=c("1" = "Sun", "2" = "Mon", "3" = "Tue", "4" = "Wed", "5" = "Thu", "6" = "Fri
 theme_bw()
}
data <- data.frame(</pre>
 day = as.Date("2020-04-02") + 0:179,
 value = as.numeric(time_series)
plot <- ggplot(data, aes(x=day, y=value)) +</pre>
 geom_line(size = 0.7) +
 xlab("")
plot + labs(x = "", y = "Bed Occupied") + theme_bw()
                          2400
   2200
Bed Occupied
   2000
   1800
   1600
                                                                                 Oct
         Apr
                                             Jul
ts_plot_season(time_series)
```

Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.

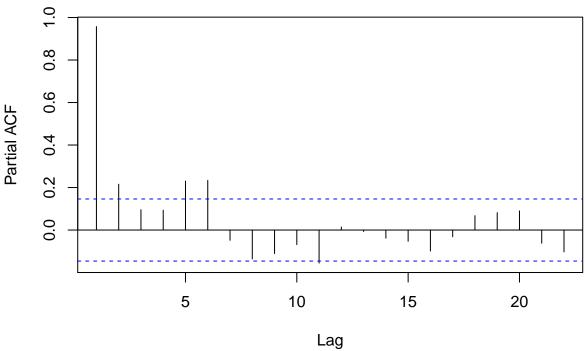


Bed Occupied Time Series ACF



```
plot(ts_pacf, main = "Bed Occupied Time Series PACF")
```

Bed Occupied Time Series PACF



```
cost_function<-function(order,demand){</pre>
        u<-runif(100,0,15)
         salvage < -mean((rep((order-demand), 100) >= u) * u + (rep((order-demand), 100) < u) * rep((order-demand), 100)) < u) * rep((order-demand), 100) < u) * rep
        over<-10*(order-demand)-4*salvage
        under<-(demand-order)^2
        cost<-(order>=demand)*over+(order<demand)*under</pre>
        return(cost)
}
art_demand<-rnorm(10000,700,50)
mini_function<-function(par,list_demand){</pre>
        record<-c()
        for (i in list_demand) {
               co<-cost_function(par,i)</pre>
               record<-c(record,co)
       }
        return(mean(record))
quan <-mean(optim(par = 700, fn = mini_function, method = 'L-BFGS-B', list_demand = art_demand) par > art_demand
mini_cost_1<-function(data,par){</pre>
        order<-par[1]+par[2]*data[,2]
        total_cost<-sum(apply(cbind(order,data[,1]),1,function(x) cost_function(x[1],x[2])))
        return(total_cost)
}
dj_record_1<-c()</pre>
```

```
imeo_record_1<-c()</pre>
for (i in 1:n){
  train<-ts(time_series[i:(horizon+i-1)]/3,frequency = 7)</pre>
  L1 train=lag(train,-1)
  set_train=cbind(train,L1_train)
  colnames(set train)<-c('train','L1')</pre>
  set_train=ts(set_train[8:horizon,])
  dj fit<-ssarima(train,frequency=7,orders=list(ar=c(1,0)),lags =c(1,7),constant=T)</pre>
  dj_value<-forecast(dj_fit,1,interval="parametric",level=quan*2-1)$upper
  dj_record_1<-c(dj_record_1,dj_value)</pre>
  coe<-lm(train ~., data=set_train)$coefficients</pre>
  imeo_fit<-optim(par = coe, fn = mini_cost_1,method = 'L-BFGS-B', data = set_train)$par</pre>
  imeo_value<-imeo_fit[1]+imeo_fit[2]*train[horizon]</pre>
  imeo_record_1<-c(imeo_record_1,imeo_value)</pre>
mini_cost_2<-function(data,par){</pre>
  order<-par[1]+par[2]*data[,2]+par[3]*data[,3]
  total_cost<-sum(apply(cbind(order,data[,1]),1,function(x) cost_function(x[1],x[2])))
  return(total cost)
}
dj_record_2<-c()
imeo record 2<-c()
for (i in 1:n){
  train<-ts(time_series[i:(horizon+i-1)]/3,frequency = 7)</pre>
  L1_train=lag(train,-1)
  L2_train=lag(train,-2)
  set_train=cbind(train,L1_train,L2_train)
  colnames(set train)<-c('train','L1','L2')</pre>
  set train=ts(set train[8:horizon,])
  dj_fit<-ssarima(train,frequency=7,orders=list(ar=c(2,0)),lags =c(1,7),constant=T)
  dj_value<-forecast(dj_fit,1,interval="parametric",level=quan*2-1)$upper
  dj_record_2<-c(dj_record_2,dj_value)</pre>
  coe<-lm(train ~., data=set train)$coefficients
  imeo fit<-optim(par = coe, fn = mini cost 2, method = 'L-BFGS-B', data = set train) par
  imeo value<-imeo fit[1]+imeo fit[2]*train[horizon]+imeo fit[3]*train[horizon-1]</pre>
  imeo_record_2<-c(imeo_record_2,imeo_value)</pre>
}
mini_cost_3<-function(data,par){</pre>
  order<-par[1]+par[2]*data[,2]+par[3]*data[,3]
  total_cost<-sum(apply(cbind(order,data[,1]),1,function(x) cost_function(x[1],x[2])))
  return(total_cost)
}
dj_record_3<-c()
imeo_record_3<-c()</pre>
for (i in 1:n){
  train<-ts(time_series[i:(horizon+i-1)]/3,frequency = 7)</pre>
  L1_train=lag(train,-1)
  L7_train=lag(train,-7)
  set_train=cbind(train,L1_train,L7_train)
  colnames(set_train)<-c('train','L1','L7')</pre>
```

```
set_train=ts(set_train[8:horizon,])
   dj_fit<-ssarima(train,frequency=7,orders=list(ar=c(1,1)),lags =c(1,7),constant=T)</pre>
   dj_value<-forecast(dj_fit,1,interval="parametric",level=quan*2-1)$upper
   dj_record_3<-c(dj_record_3,dj_value)</pre>
   coe<-lm(train ~., data=set_train)$coefficients</pre>
   imeo_fit<-optim(par = coe, fn = mini_cost_3,method = 'L-BFGS-B', data = set_train)$par</pre>
   imeo_value<-imeo_fit[1]+imeo_fit[2]*train[horizon]+imeo_fit[3]*train[horizon-6]
   imeo record 3<-c(imeo record 3,imeo value)</pre>
mini_cost_4<-function(data,par){</pre>
   order<-par[1]+par[2]*data[,2]+par[3]*data[,3]+par[4]*data[,4]
   total_cost<-sum(apply(cbind(order,data[,1]),1,function(x) cost_function(x[1],x[2])))
   return(total cost)
}
dj_record_4<-c()
imeo record 4<-c()
for (i in 1:n){
   train<-ts(time series[i:(horizon+i-1)]/3,frequency = 7)</pre>
   L1_train=lag(train,-1)
   L2_train=lag(train,-2)
   L7_train=lag(train,-7)
   set_train=cbind(train,L1_train,L2_train,L7_train)
   colnames(set_train)<-c('train','L1','L2','L7')</pre>
   set_train=ts(set_train[8:horizon,])
   dj_fit<-ssarima(train,frequency=7,orders=list(ar=c(2,1)),lags =c(1,7),constant=T)
   dj_value<-forecast(dj_fit,1,interval="parametric",level=quan*2-1)$upper
   dj_record_4<-c(dj_record_4,dj_value)</pre>
   coe<-lm(train ~., data=set_train)$coefficients</pre>
   imeo_fit<-optim(par = coe, fn = mini_cost_4,method = 'L-BFGS-B', data = set_train)$par</pre>
   imeo_value<-imeo_fit[1]+imeo_fit[2]*train[horizon]+imeo_fit[3]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horizon-1]+imeo_fit[4]*train[horiz
   imeo_record_4<-c(imeo_record_4,imeo_value)</pre>
test_series<-time_series[(horizon+1):(horizon+n)]/3
return_function<-function(forecast_list,real_list){</pre>
   cost<-apply(cbind(forecast_list,real_list),1,function(x) cost_function(x[1],x[2]))</pre>
   error<-forecast list-real list
   list<-list(cost,error)</pre>
   return(list)
7
dj_result_1<-return_function(dj_record_1,test_series)</pre>
dj_result_2<-return_function(dj_record_2,test_series)</pre>
dj_result_3<-return_function(dj_record_3,test_series)</pre>
dj_result_4<-return_function(dj_record_4,test_series)</pre>
imeo_result_1<-return_function(imeo_record_1,test_series)</pre>
imeo_result_2<-return_function(imeo_record_2,test_series)</pre>
imeo_result_3<-return_function(imeo_record_3,test_series)</pre>
imeo_result_4<-return_function(imeo_record_4,test_series)</pre>
method_app<-c(rep('DJ',4*n),rep('IMEO',4*n))</pre>
```

```
lag_length<-c(rep('p=1,P=0',n),rep('p=2,P=0',n),rep('p=1,P=1',n),rep('p=2,P=1',n),
              rep('p=1,P=0',n),rep('p=2,P=0',n),rep('p=1,P=1',n),rep('p=2,P=1',n))
cost_record<-c(dj_result_1[[1]],dj_result_2[[1]],dj_result_3[[1]],dj_result_4[[1]],</pre>
               imeo_result_1[[1]],imeo_result_2[[1]],imeo_result_3[[1]],imeo_result_4[[1]])
error_record<-c(dj_result_1[[2]],dj_result_2[[2]],dj_result_3[[2]],dj_result_4[[2]],
               imeo_result_1[[2]],imeo_result_2[[2]],imeo_result_3[[2]],imeo_result_4[[2]])
df<-data.frame(Method_app,lag_length=lag_length,cost_record=cost_record,error_record=error_record
df$lag length <- as.factor(df$lag length)</pre>
ggplot(df, aes(x=lag_length, y=cost_record, fill=Method)) +
  coord_cartesian(ylim = c(0, 1000)) +
  geom_boxplot(aes(middle = mean(cost_record))) + theme_bw() +
  labs(x = "Order", y = "Cost")
  1000
   750
                                                              •
                                                                               Method
   500
                                                                                   DJ
                                                                                   IMEO
   250
     0
                                                              p=2,P=1
                              p=1,P=1
             p=1,P=0
                                              p=2,P=0
                                       Order
ggplot(df, aes(x=lag_length, y=error_record, fill=Method)) +
  coord_cartesian(ylim = c(-100, 100)) +
  geom_boxplot(aes(middle = mean(error_record))) + theme_bw() +
```

labs(x = "Order", y = "Error")

