

```
---
title: "BuildThePrior"
author: "Linyi Guo"
date: "2020/1/8"
output:
  html_document:
    toc: true
---
```

```
``{r include=FALSE}
set.seed(9483)
rm(list=ls())
``
```

```
``{r}
# install these packages if needed
#install.packages("seasonal")
#install.packages("forecast")
#install.packages("KFAS")
```

```
# import package
library(seasonal)
library(forecast)
library(KFAS)
``
```

```
# SIMULATION
```

To avoid potential mistakes, I only simulate the datasets from ARIMA(1,1,1)(0,1,1)

```
## Define a function to simulate data that we want
```

```
``{r}
simulation1 <- function(length){

  model <- Arima(ts(rnorm(120),start=c(1980,01),frequency =12), order=c(1,1,1),
                 seasonal=c(0,1,1), fixed=c(phi=runif(1), theta=runif(1),
                                             Theta=runif(1))
                 )

  data <- simulate(model, nsim=length)
```

```

    # because if we need to take log later, data must be positive
    if(min(data) <= 0) data <- data - min(data) + runif(1)
    else data <- data

    return(data)
}
'''

```

function to simulate a datalist with a lot of datasets

```

'''{r}
simlist1 <- function(n,length) {

    Datalist <- list()

    for (i in 1:n) Datalist[[i]] <- simulation1(length)

    return(Datalist)

}

'''

```

```

'''{r}
set.seed(9483)

datalist2 <- simlist1(100, 180)

# If time permits, you can try this one
# datalist2 <- simlist1(200, 180)

'''

```

Until this step, datalist is basically what I want to use in the following analysis, but we still have one more step: that is to preprocessing.

Preprocessing

```

```{r}
define a function for outliers and log-transformation
preprocess <- function(x11) {

 if(transformfunction(x11) == 'log')
 data <- log(series(x11, 'b1'))
 else
 data <- series(x11, 'b1')

 return(data)
}
```

```

```

```{r}
build model list
x11list2 <- lapply(datalist2, function(x) seas(x, x11=""))

```

```

```

```{r}
obtain the preprocessed datalist denoted as Datalist
Datalist2 <- lapply(x11list2, preprocess)

```

```

Now we finished this step, then we need to find the 'ideal' values of parameters in these datasets.

Building Our Prior

Searching for the 'best' value

****Loss function****

```

```{r}

Dif1 <- function(x11, ssm, data, sigma){

 x11_trend <- series(x11, 'd12')
 x11_seasonal <- series(x11, 'd10')
 x11_irregular <- series(x11, 'd13')

```

```

ssm_trend <- coef(ssm, states = 'trend')
ssm_seasonal <- -rowSums(coef(ssm, states='seasonal'))
ssm_irregular <- data[-1] - ssm_trend[-1] - ssm_seasonal[-length(data)]

D <- sum((x11_irregular[-1]-ssm_irregular)^2)/sigma[1] +
 sum((x11_trend-ssm_trend)^2)/sigma[2] +
 sum((x11_seasonal[-1]-ssm_seasonal[-length(data)])^2)/sigma[3]

return(D)
}
'''

Exhaustion function

'''{r}

exhaustion1 <- function(data){

 Difference <- c()
 index <- c()

 x11 <- seas(data, x11='')

 for (i in 1:100) {
 for (j in 1:100) {

 ssmm <- SSMModel(data ~ SSMtrend(1, Q=list(j*0.2)) +
 SSMseasonal(12, sea.type = 'dummy', Q = 1),
 H = i*0.2)
 ssm <- KFS(ssmm)

 sigma <- c(i*0.2, j*0.2, 1)

 dif <- Dif1(x11, ssm, data, sigma)

 Difference <- c(Difference, dif)

 index <- rbind(index, sigma)
 }
 }

 df <- data.frame(variance=index, difference = Difference)
 return(df)
}

```

```
}
'''
```

```
'''{r}
system.time(exhaustion1(Datalist1[[1]]))
user system elapsed
189.80 237.43 427.96
```

```
idevallist2 <- lapply(Datalist2, exhaustion1)
'''
```

```
'''{r}
idevalmat2 <- c()

for (i in 1:100){
 ideval <- idevallist2[[i]][which.min(idevallist2[[i]]$difference),c(1,2)]
 idevalmat2 <- rbind(idevalmat2, ideval)
}

'''
```

And then please output the data frame `idevalmat2`.

```
'''
write.csv(idevalmat2, "The position//idevalmat2.csv")
'''
```