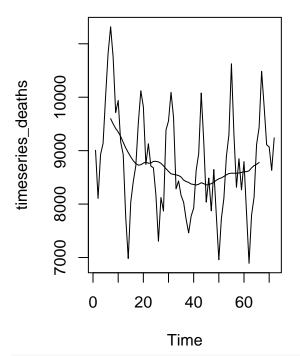
#### Homework 2

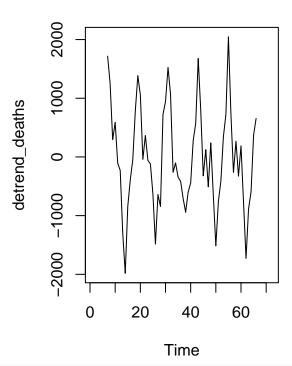
#### Yeshashvi Munagala

1/26/2022

# Questions 1 and 3 are attached at the end of the pdf! 2a, 2b)

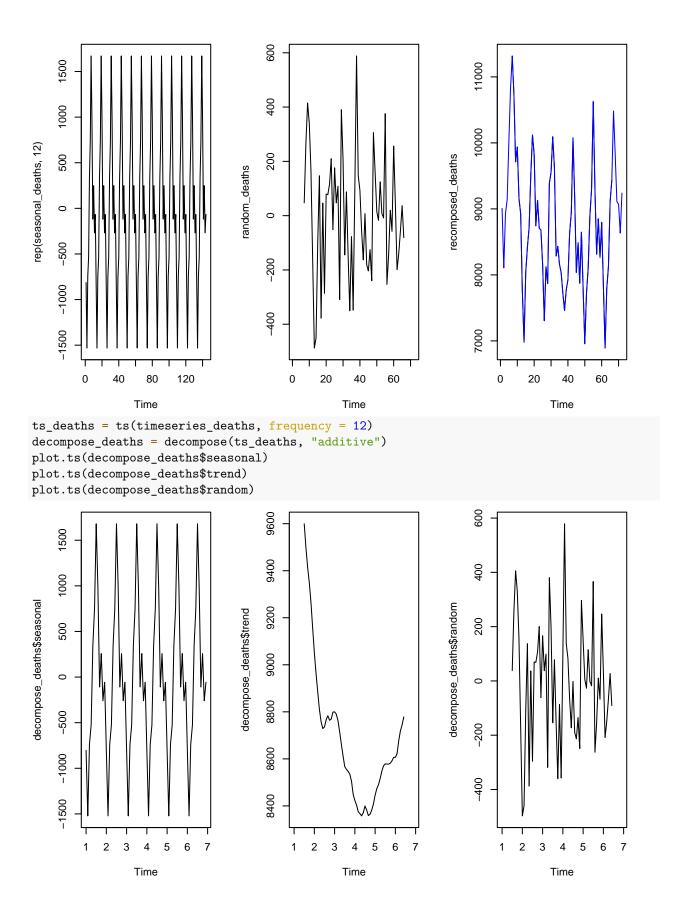
```
#library(TSstudio)
library(itsmr)
library(forecast)
## Warning: package 'forecast' was built under R version 4.1.1
## Registered S3 method overwritten by 'quantmod':
##
     method
                       from
##
     as.zoo.data.frame zoo
##
## Attaching package: 'forecast'
## The following object is masked from 'package:itsmr':
##
       forecast
timeseries_deaths = tail(head(deaths, 17*12+2),17*12-4)
trend_deaths = ma(timeseries_deaths, order = 12, centre = T)
detrend_deaths = timeseries_deaths - trend_deaths
par(mfrow = c(1, 2))
plot.ts(timeseries_deaths)
lines(trend_deaths)
plot.ts(detrend_deaths)
```



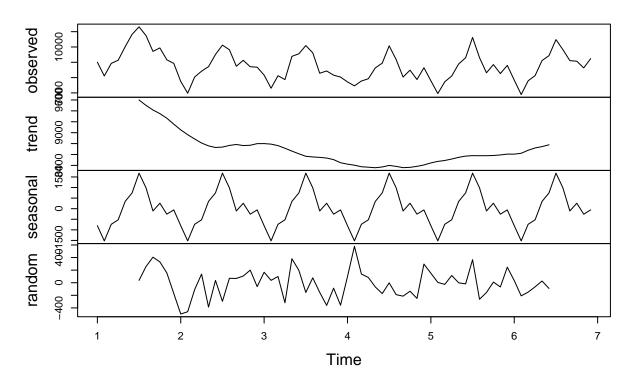


```
m_deaths = t(matrix(data = detrend_deaths, nrow = 12))
#m_deaths

seasonal_deaths = colMeans(m_deaths, na.rm = T)
random_deaths = timeseries_deaths - trend_deaths - seasonal_deaths
recomposed_deaths = trend_deaths + seasonal_deaths + random_deaths
par(mfrow = c(1, 3))
plot.ts(rep(seasonal_deaths, 12))
plot.ts(random_deaths)
plot.ts(recomposed_deaths)
lines(timeseries_deaths, col="blue")
```



#### **Decomposition of additive time series**



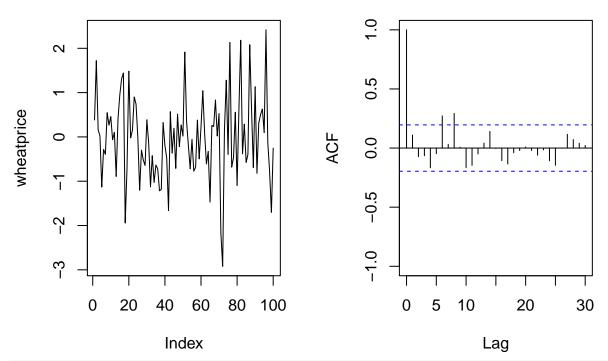
#### 4a)

```
library(fpp3)
```

```
## -- Attaching packages ------ fpp3 0.4.0 --
## v tibble
               3.1.2
                        v tsibble
                                     1.1.1
## v dplyr
                        v tsibbledata 0.4.0
               1.0.7
## v tidyr
               1.1.3
                                    0.2.2
                        v feasts
## v lubridate
               1.8.0
                        v fable
                                     0.3.1
## v ggplot2
               3.3.5
## Warning: package 'lubridate' was built under R version 4.1.1
## Warning: package 'tsibble' was built under R version 4.1.1
## Warning: package 'tsibbledata' was built under R version 4.1.1
## Warning: package 'feasts' was built under R version 4.1.1
## Warning: package 'fabletools' was built under R version 4.1.1
## Warning: package 'fable' was built under R version 4.1.1
## -- Conflicts ----- fpp3_conflicts --
## x lubridate::date()
                         masks base::date()
## x dplyr::filter()
                         masks stats::filter()
## x fabletools::forecast() masks forecast::forecast(), itsmr::forecast()
## x tsibble::intersect() masks base::intersect()
## x tsibble::interval() masks lubridate::interval()
```

```
## x dplyr::lag()
                              masks stats::lag()
## x tsibble::setdiff()
                              masks base::setdiff()
## x tsibble::union()
                              masks base::union()
wheatprice <- ts(prices$wheat, start = 1800, end = 1997,frequency = 1)</pre>
plot.ts(wheatprice)
      009
wheatprice
      400
      200
           1800
                              1850
                                                                   1950
                                                                                      2000
                                                 1900
                                                Time
wheatprice <- rnorm(100, mean=0, sd=1)</pre>
par(\overline{mfrow} = c(1, 2))
plot(wheatprice, type="1")
acf(wheatprice, lag.max=30, ylim=c(-1, 1))
```

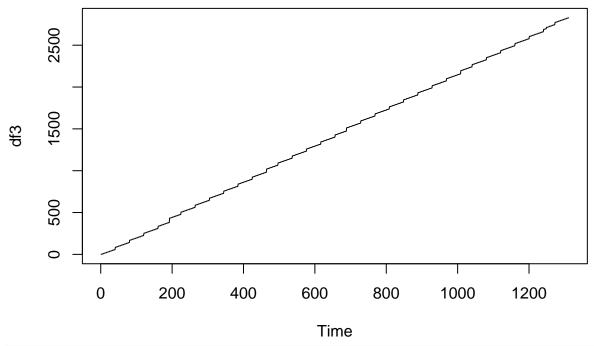
#### Series wheatprice



# There is clearly a decreasing trend in this time series. The model does a good job af representing th

#### 4b)

```
df <- data.frame(bank_calls)</pre>
dt <- seq(
  from = as.POSIXct("2003-03-03 07:00:00", tz = "GMT"),
  to = as.POSIXct("2003-10-24 21:00:00", tz = "GMT"),
  by = "7200 sec")
binned_dataset = bank_calls %>% mutate(BinnedTime = cut(DateTime, breaks=dt)) %>%
  group_by(BinnedTime,add=TRUE)
## Warning: The `add` argument of `group_by()` is deprecated as of dplyr 1.0.0.
## Please use the `.add` argument instead.
df2 <- data.frame(binned_dataset$Calls, binned_dataset$BinnedTime)</pre>
colnames(df2) <- c('Calls', 'BinnedTime')</pre>
df2 <- aggregate(df2$Calls,</pre>
                by = list(df2$BinnedTime),
                FUN = sum)
colnames(df2) <- c('DateTime','Aggregate Calls')</pre>
df3 \leftarrow df2[c(1:50),]
df3 <- ts(df2$DateTime)</pre>
plot.ts(df3)
```



```
#df2$DateTime <- as.datetime(df2$DateTime, '%Y-%m-%d %H:%M:%S')

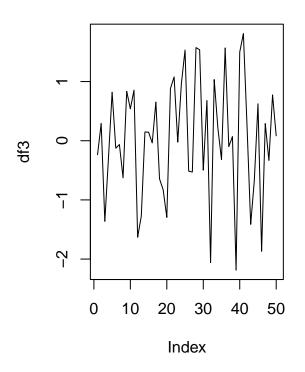
df3 <- rnorm(50, mean=0, sd=1)

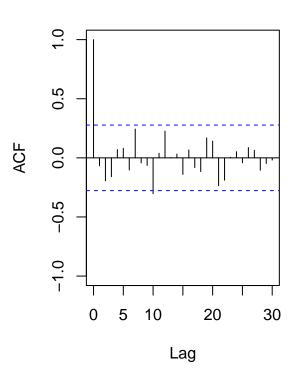
par(mfrow = c(1, 2))

plot(df3, type="l")

acf(df3, lag.max=30, ylim=c(-1, 1))
```

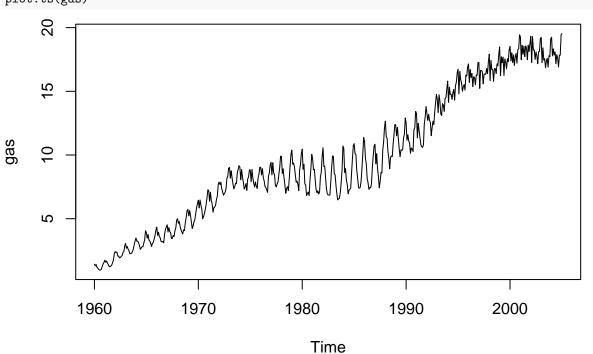
## Series df3





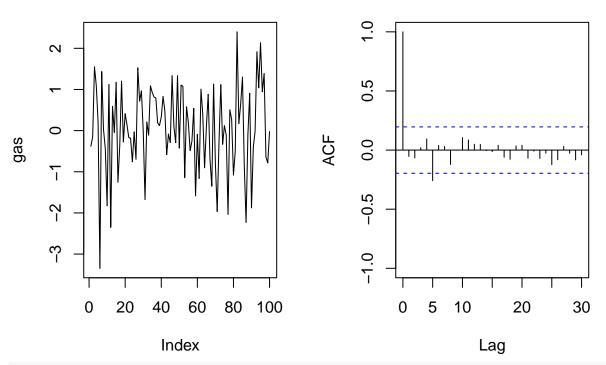
#### **4c**)

```
canadian_gas$Month <- as.Date(canadian_gas$Month)
gas <- ts(canadian_gas$Volume, start = 1960, end = 2005, frequency = 12)
plot.ts(gas)</pre>
```



```
gas <- rnorm(100, mean=0, sd=1)
par(mfrow = c(1, 2))
plot(gas, type="1")
acf(gas, lag.max=30, ylim=c(-1, 1))</pre>
```

### Series gas



# Here, I expected the correlogram to exhibits peaks at multiples of 12 because #this data contains monthly data; however, that is not the case. Based on #autocorrelation, it seems as though the model does a good job representing the #data set, as the lines in the ACF plot are mostly between the dashed blue lines.