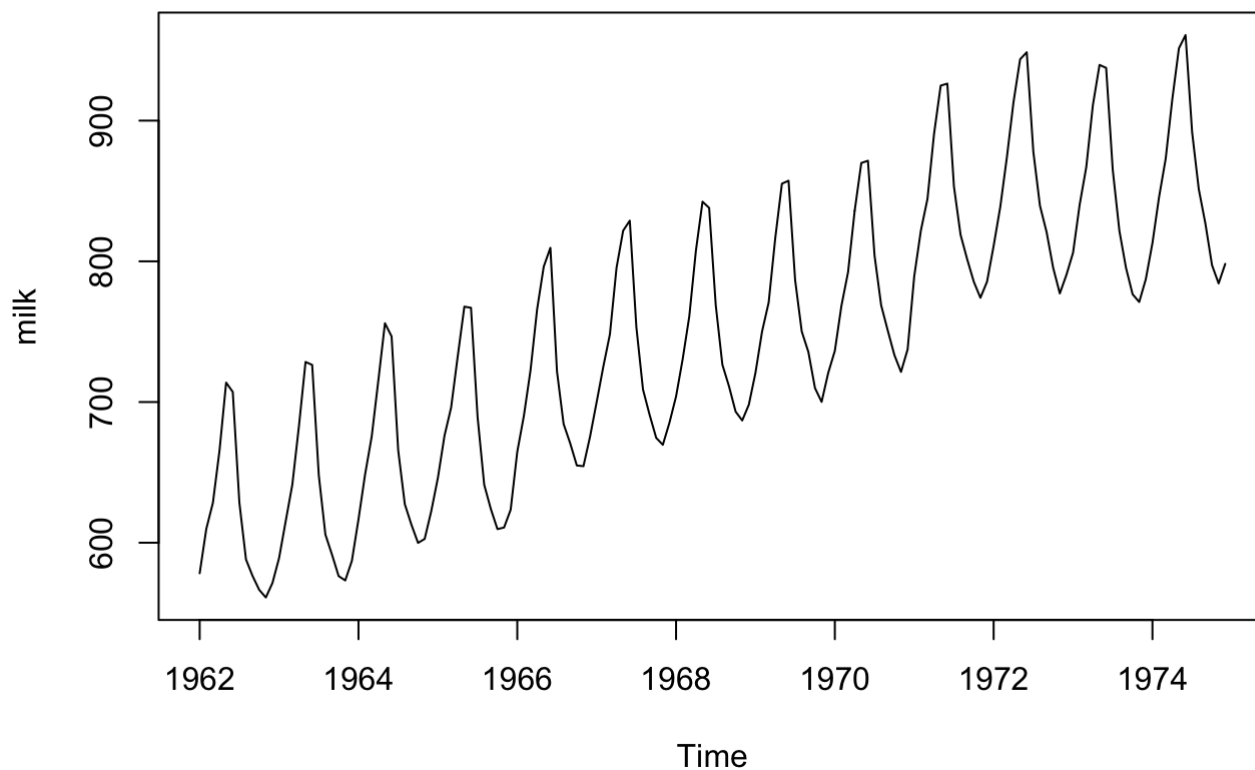


# pstat271\_lab05\_aoxu

AO XU

2022-10-31

```
library(tsd1)
milk <- subset(tsd1, 12, "Agriculture")[[3]]
plot(milk)
```



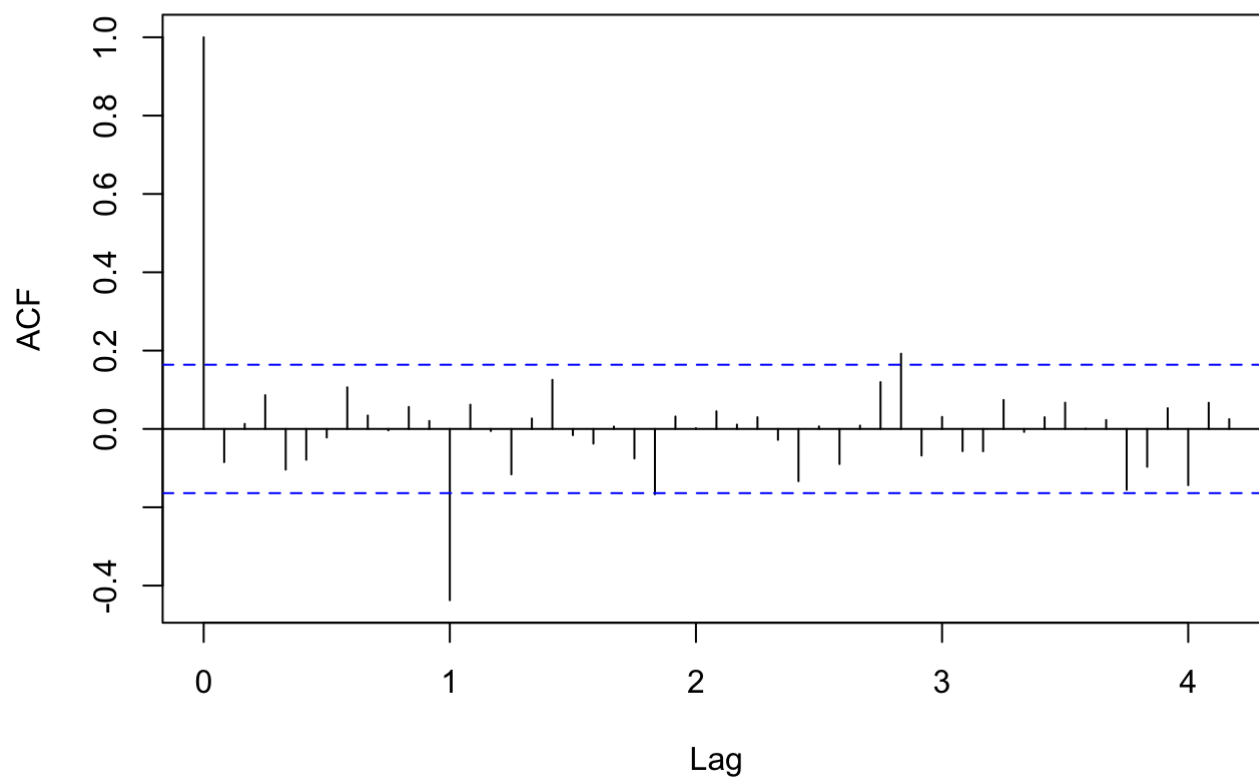
```
# To make it more stationary, we use the following code:
dmilk <- diff(milk, 12)
ddmilk <- diff(dmilk, 1)
```

a.

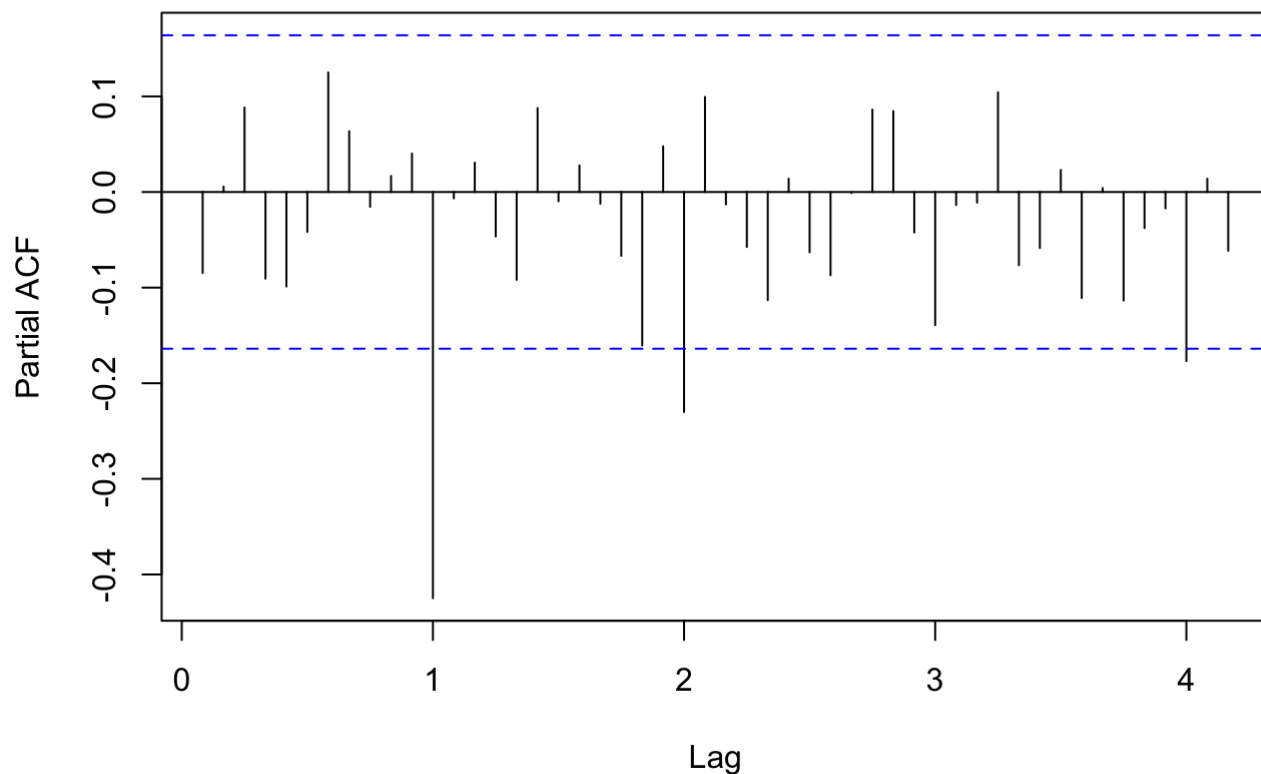
It has a positive linear trend and seasonality, then it is not stationary.

b.

```
dmilk <- diff(milk, 12)
ddmilk <- diff(dmilk, 1)
acf(ddmilk, lag.max = 50, main="")
```



```
pacf(ddmilk, lag.max = 50, main="")
```



c.

Since it's one seasonal differencing, then we get that  $D=1$  at lag  $s=12$ . And `ddmilk` applied one differencing to remove trend, so  $d=1$ . From `acf`, there is one peak at 1, so  $Q$  is 1. From `PACF`, there are peaks at 1, 2 or 4. So  $P$  could be 1 or 2 or 4. For 0-1 from both `ACF` and `PACF`, there are no peaks. Then we get

$\text{SARIMA}(0, 1, 0) \times (1, 1, 1)_{s=12}$

$\text{SARIMA}(0, 1, 0) \times (2, 1, 1)_{s=12}$

$\text{SARIMA}(0, 1, 0) \times (4, 1, 1)_{s=12}$

d.

```
library(astsa)
#library(sarima)
fit.i <- sarima( xdata = milk,
                 p = 0, d = 1, q = 0,
                 P = 1, D = 1, Q = 1, S = 12,
                 details = F)
print('Coefficients')
```

```
## [1] "Coefficients"
```

```
fit.i$fit$coef
```

```
##          sar1          sma1  
## 0.01876417 -0.68617816
```

```
plot(fit.i$fit$residuals)
```

