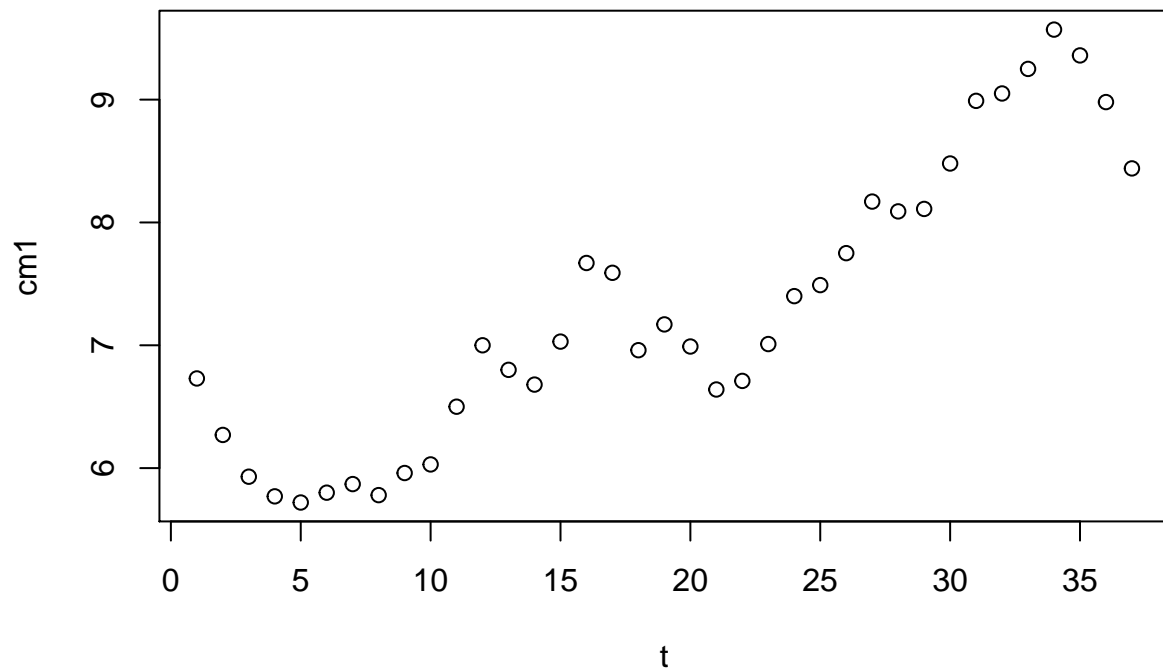


Exp 8-5 Edit

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2019-02-14

```
# Example 8.6 ARM problem  
  
#1. Read and preview  
# read data (change path for your computer)  
cm1=scan("cm1.txt")  
n=length(cm1)  
t=seq(1,n,1)  
# plot data  
#par(mfcol=c(2,2))  
plot(t,cm1)
```



```
#2. Data manipulation  
#repeat for Autoregression involving 1:3 previous terms
```

```

#(to avoid copying and pasting lag1=...,fit1...)

cm1df = data.frame(t=t,cm1=cm1)
cm1df = within(cm1df,{
  for(i in 1:3) assign(paste0('lag',i),c(rep(NA,i),cm1[1:(n-i)]))#assign lag1:lag3
  remove(i)
})
#for(i in 1:3) assign(paste0('lag',i),c(rep(NA,i),cm1[1:(n-i)]))#assign lag1:lag3

fit1 = lm(cm1~t+lag1,data=cm1df)
fit2 = lm(cm1~t+lag1+lag2,data = cm1df)
fit3 = lm(cm1~t+lag1+lag2+lag3,data = cm1df)

cm1df = within(cm1df,{
  for(i in 1:3) assign(paste0('fit',i),c(rep(NA,i),predict.lm(eval(parse(text = paste0('fit',i))))))#as
  remove(i)
})
cm1df =within(cm1df,{
  for(i in 1:3) assign(paste0('res',i),cm1 - get(paste0('fit',i)))#assign lag1:lag3
  remove(i)
})
#cm1df
head(cm1df)

```

```

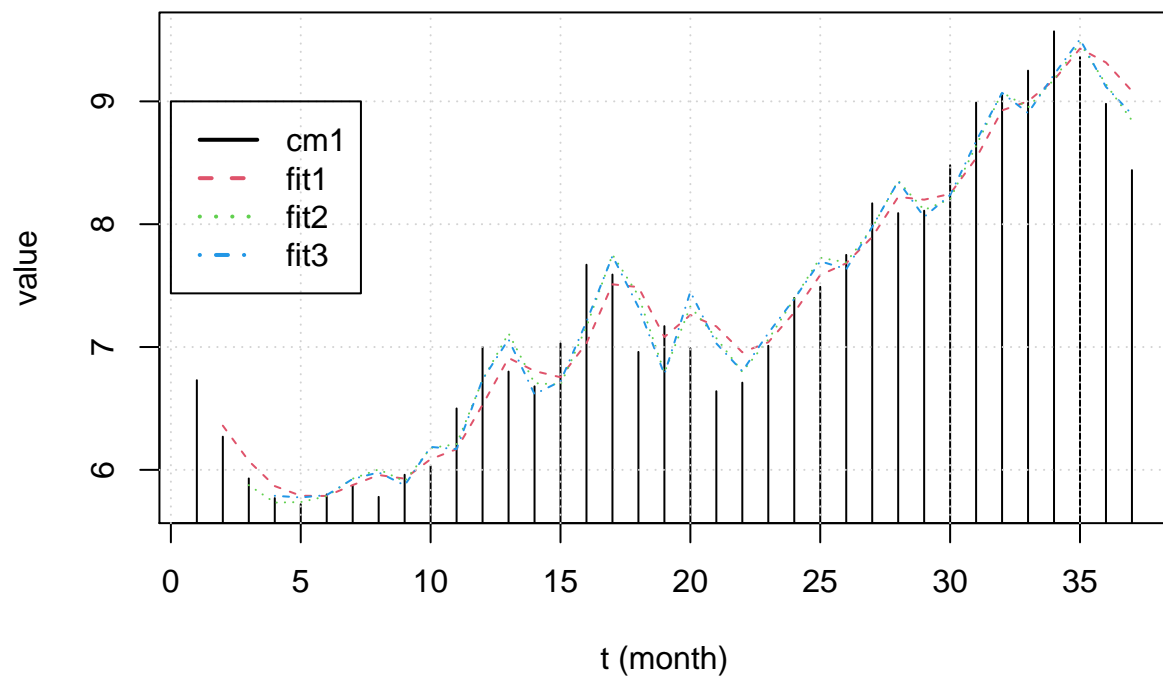
##   t  cm1 lag3 lag2 lag1      fit3      fit2      fit1      res3      res2
## 1 1 6.73   NA   NA   NA        NA        NA        NA        NA        NA
## 2 2 6.27   NA   NA 6.73        NA        NA 6.359942        NA        NA
## 3 3 5.93   NA 6.73 6.27        NA 5.875437 6.072003        NA 0.05456274
## 4 4 5.77 6.73 6.27 5.93 5.788715 5.733346 5.867784 -0.018714877 0.03665432
## 5 5 5.72 6.27 5.93 5.77 5.775334 5.737694 5.789145 -0.055333863 -0.01769442
## 6 6 5.80 5.93 5.77 5.72 5.794679 5.785128 5.787249 0.005321005 0.01487194
##           res1
## 1           NA
## 2 -0.08994206
## 3 -0.14200271
## 4 -0.09778354
## 5 -0.06914464
## 6 0.01275077

```

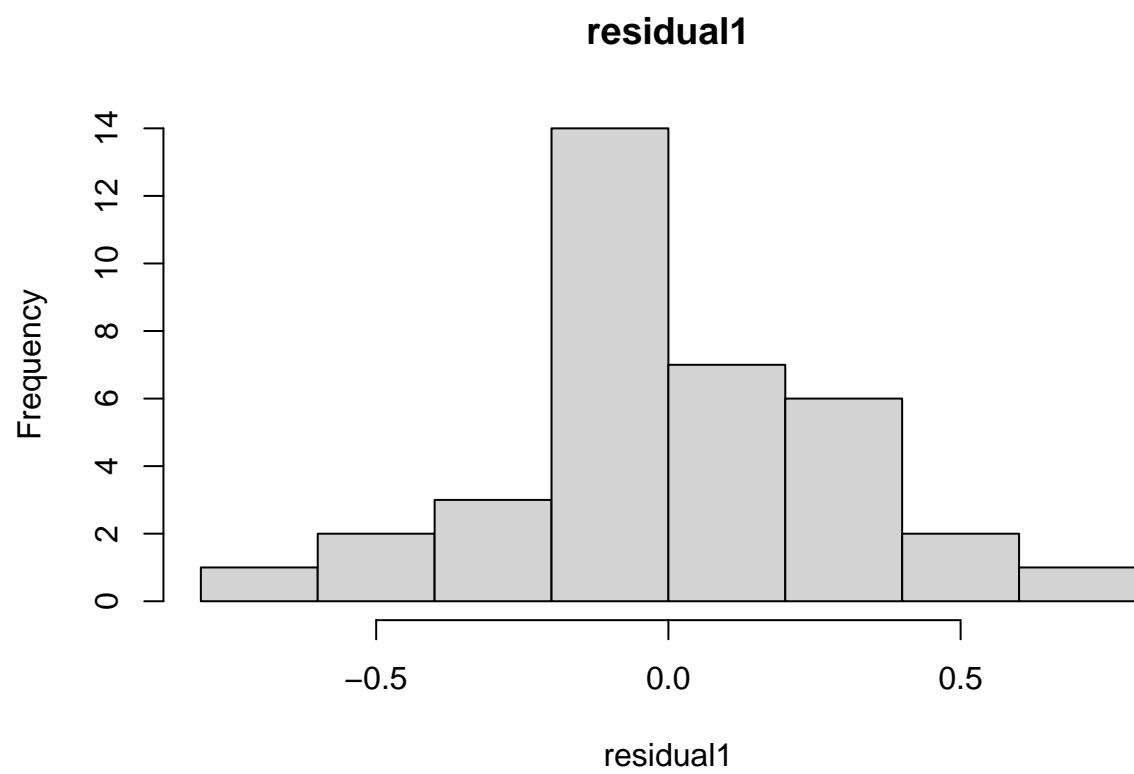
```

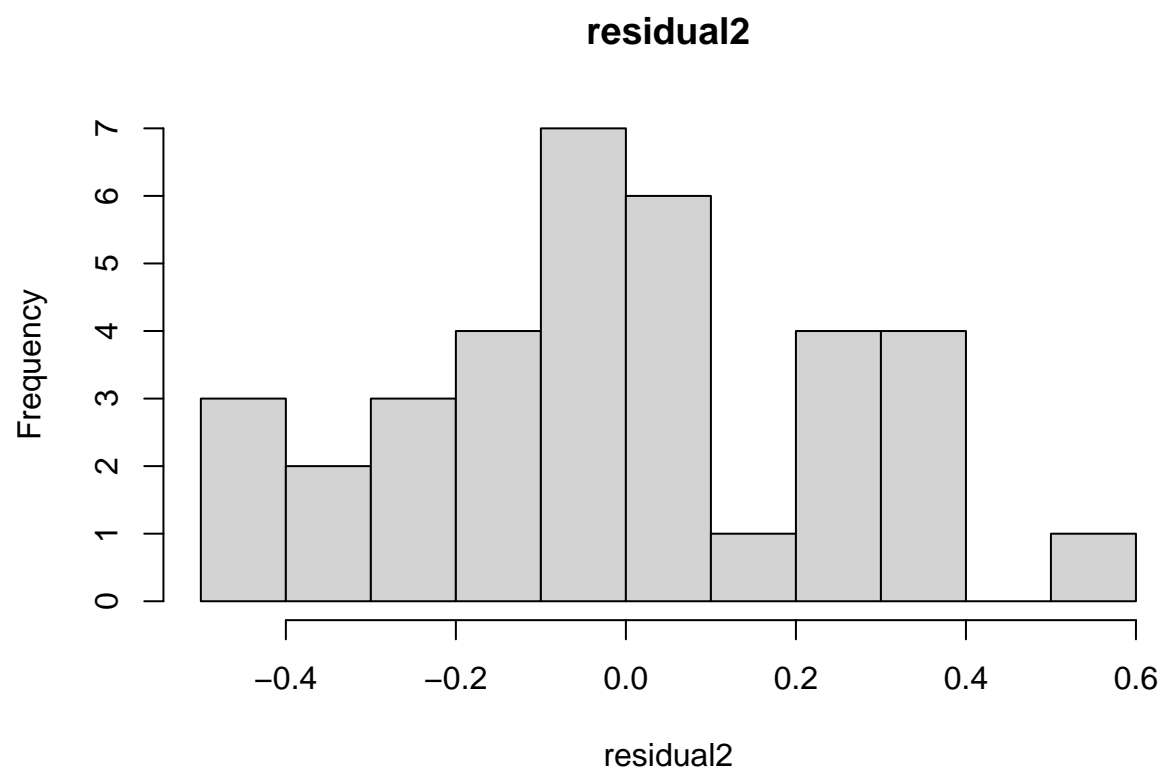
layout(1)
cm1pred = c('cm1','fit1','fit2','fit3')
matplot(x=cm1df[, 't'],y=cm1df[,cm1pred],type = c('h','l','l','l'),lty = 1:4,col =1:4,
        xlab = 't (month)',ylab = 'value')
grid()
legend(0,9,legend = cm1pred,lwd=2,col = 1:4,lty = 1:4)

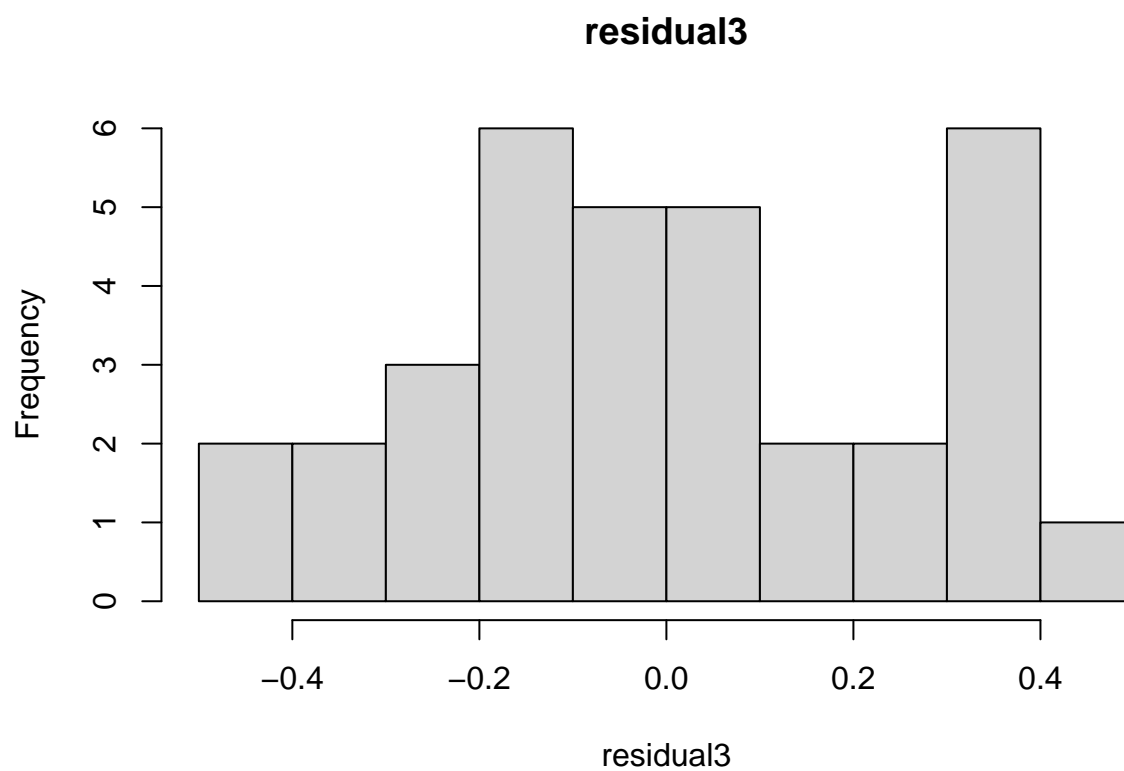
```



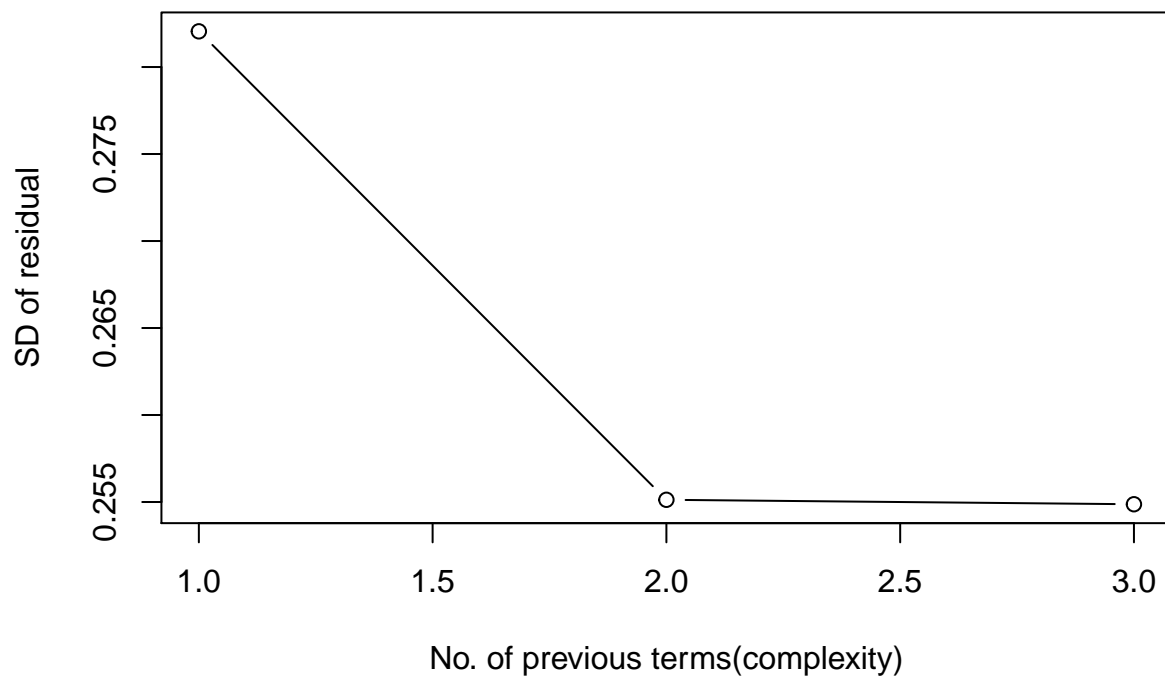
```
#layout(1:3)
for(i in 1:3) hist(cm1df[,as.character(paste0('res',i))],xlab = paste0('residual',i),main = paste0('residual',i))
```







```
layout(1)
sd_error=numeric(3)
for(i in 1:3) sd_error[i] = sd(cm1df[,as.character(paste0('res',i))],na.rm =T)
plot(c(1:3), sd_error,
     xlab = 'No. of previous terms(complexity)',ylab = 'SD of residual',type = 'b')
```



#plot shows how residuals get lowered at the expense of greater complexity

#uncomment the code below to avoid copying and pasting

###Note that this is usually not encouraged because it is unintuitive

##(to avoid copying and pasting lag1=...,fit1...)

#for(i in 1:3){

indp = paste0('lag',1:i)

sapply(indp,paste,sep = "+")

assign(paste0('fit',i),lm(as.formula(paste('cm1 ~ t +',indp))))

#}

for(i in 1:3) print(summary(eval(parse(text = paste0('fit',i)))))

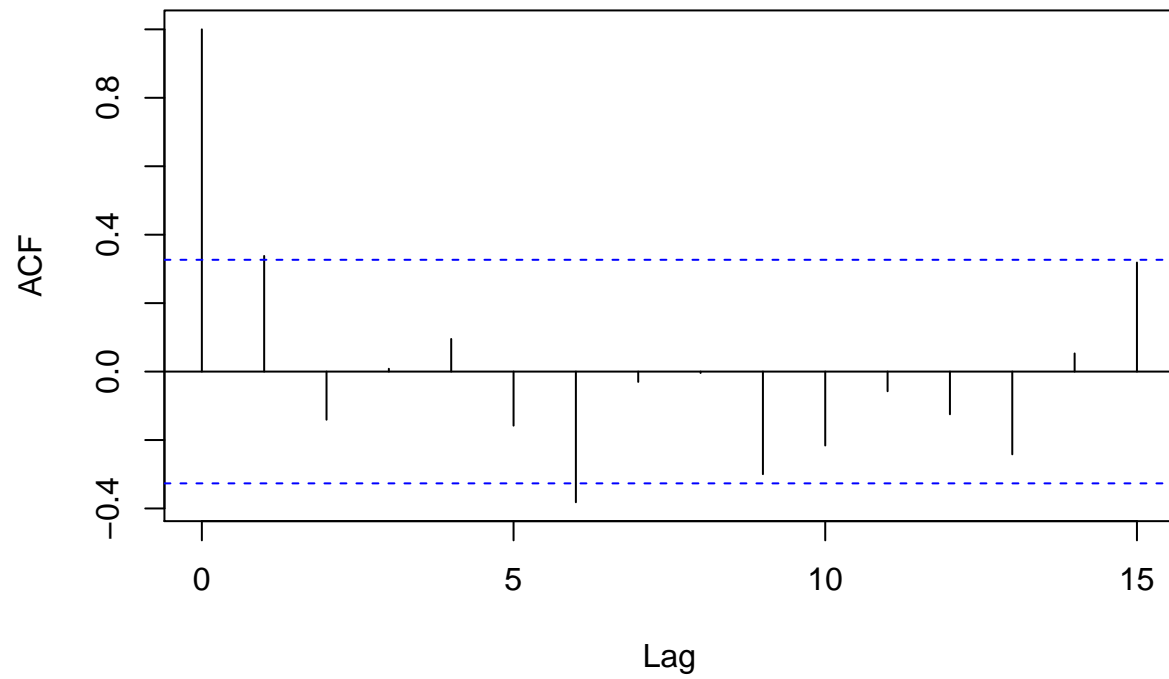
plot autocovariance function for these residuals

#require(graphics)

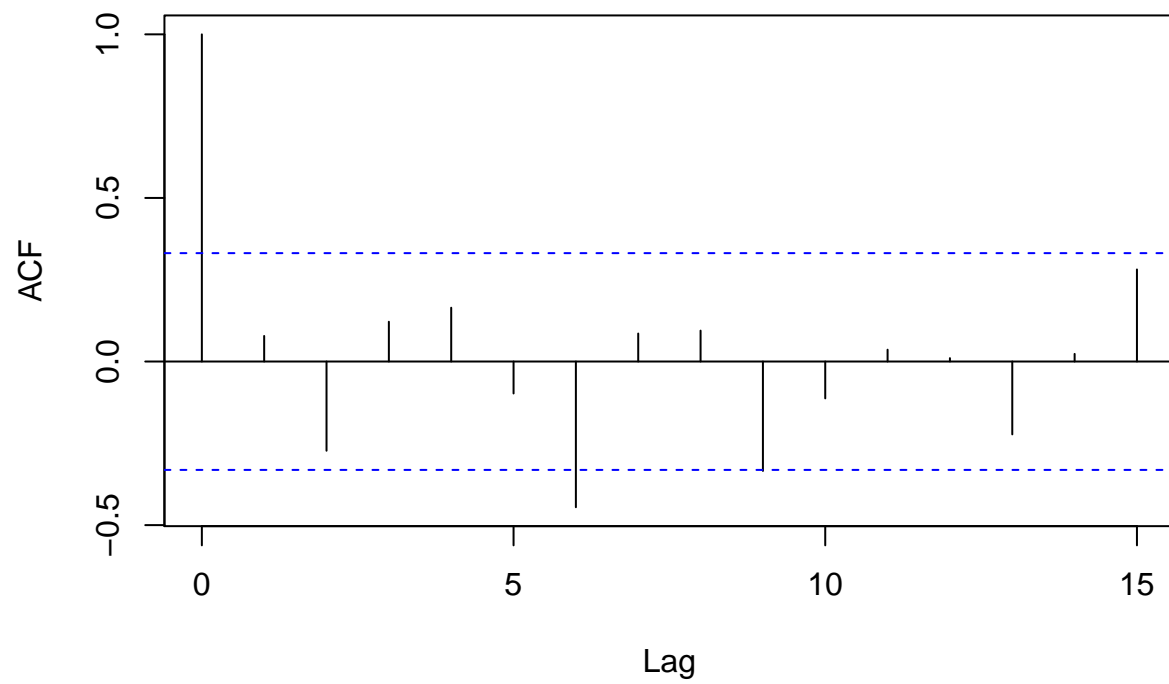
#layout(1:3)

for(i in 1:3) acf(cm1df[(i+1):nrow(cm1df),paste0('res',i)])

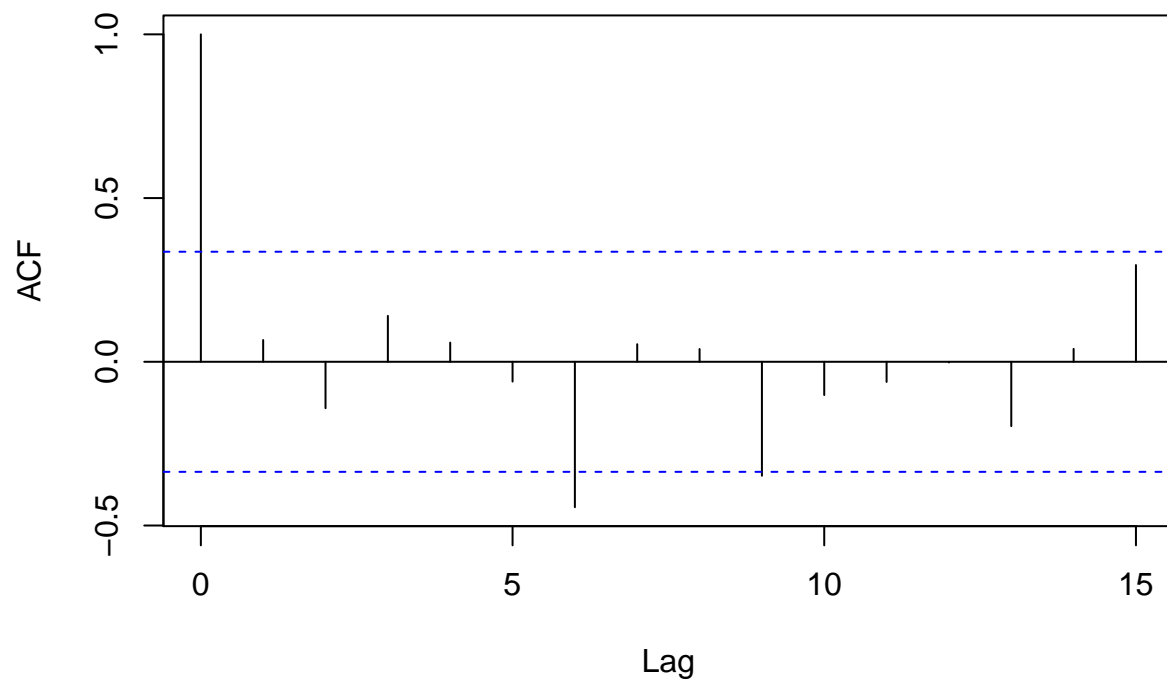
Series cm1df[(i + 1):nrow(cm1df), paste0("res", i)]



Series cm1df[(i + 1):nrow(cm1df), paste0("res", i)]



Series cm1df[(i + 1):nrow(cm1df), paste0("res", i)]



```
#acf(e1)
```

```
# forecast cm1 for May 1990
```

```
for(i in 1:3) {
  assign(paste0('coef',i),coef(get(paste0('fit',i))))
  print(get(paste0('coef',i)))
}
```

```
## (Intercept)      t      lag1
##  1.5986595  0.0329880  0.6976681
## (Intercept)      t      lag1      lag2
##  1.78482453  0.03300389  1.10359733 -0.43506006
## (Intercept)      t      lag1      lag2      lag3
##  1.63315690  0.03188948  1.17110722 -0.64645712  0.16888861
```

```
nf=11#n of forecast
```

```
t2=seq(1,n+nf,1)
```

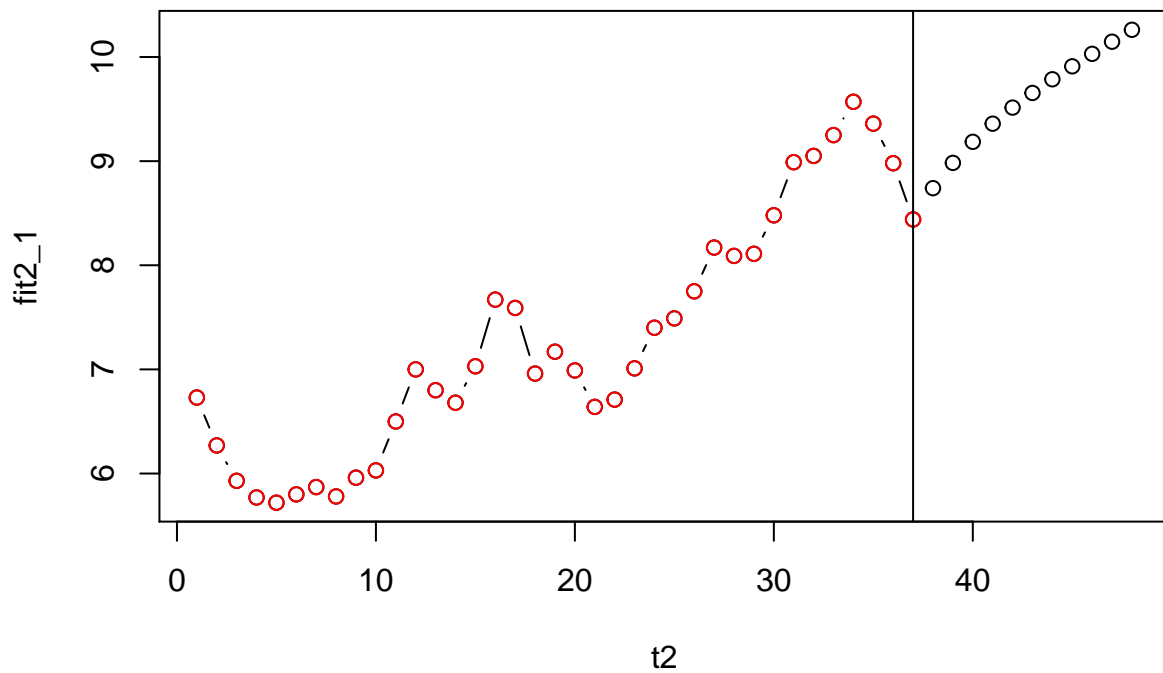
```
for(i in 1:3){
  #for every model
  cm1f = c(cm1,numeric(nf))
  #for(k in 1:i){
    # #for every lag
    # coeflag = c(coeflag,get(paste0('coef',k))[paste0('lag',k)])
  #}
```

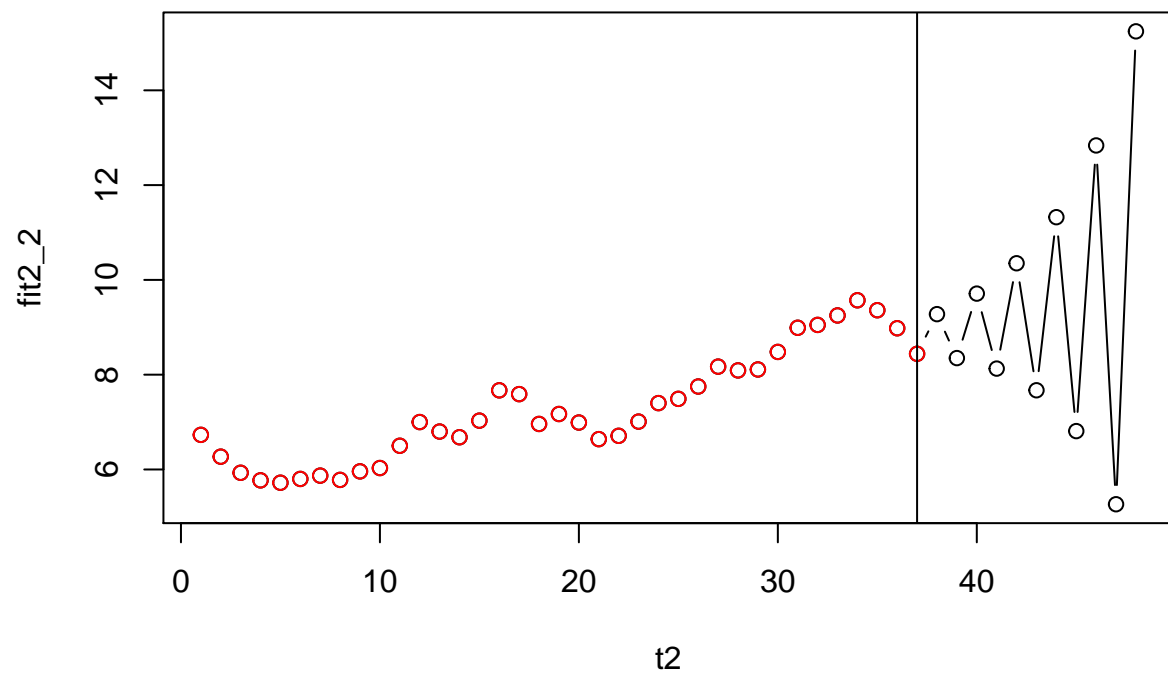
```

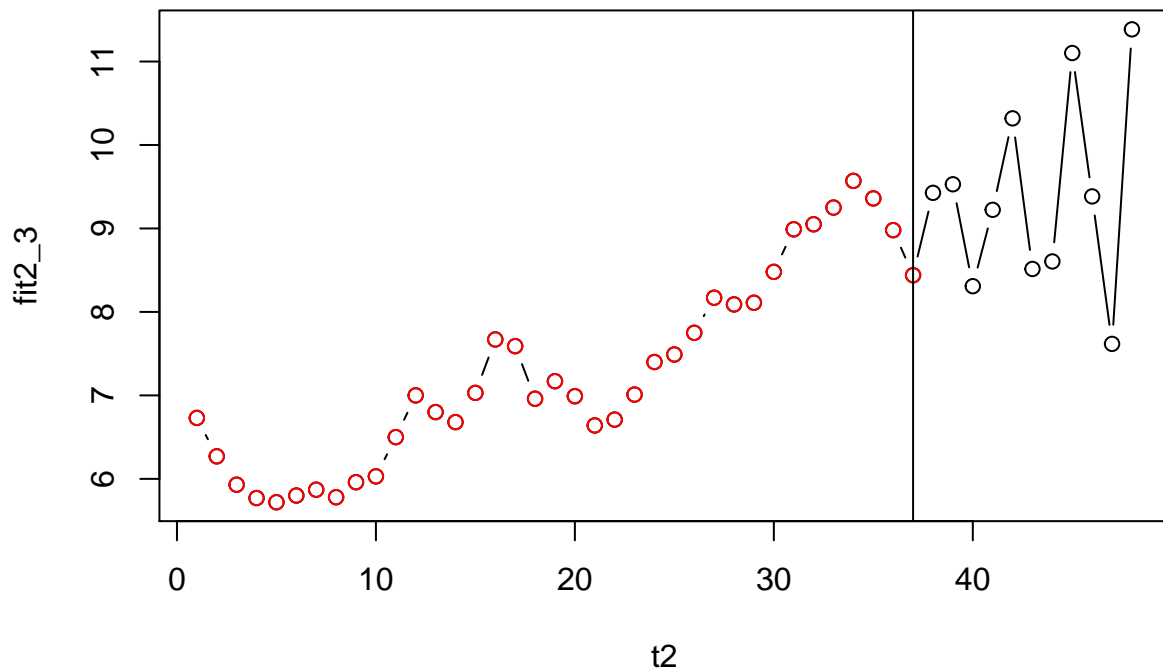
for(j in n+1:nf){
  #for every ts term

  #cm1flag = NULL
  #for(k in 1:i){
    #for every lag
    cm1flag= cm1f[(j-i):(j-1)]
    cm1f[j] = sum(get(paste0('coef',i))*c(1,t2[j],cm1flag))
  }
  assign(paste0('cm1f',i),cm1f)
  #fit2 = numeric(length(cm1f))
  #for(j in n+i:nf){
    # fit2[j] = sum(get(paste0('coef',i))*c(1,t2[j],cm1f[(j-i+1):(j)]))
  #}
  assign(paste0('fit2_',i),fit2)
}
#layout(1:3)
for(i in 1:3){
  plot(t2,get(paste0('cm1f',i)),ylab=paste0('fit2_',i), type='b')
  points(t,cm1,col='red')
  abline(v=n)
}

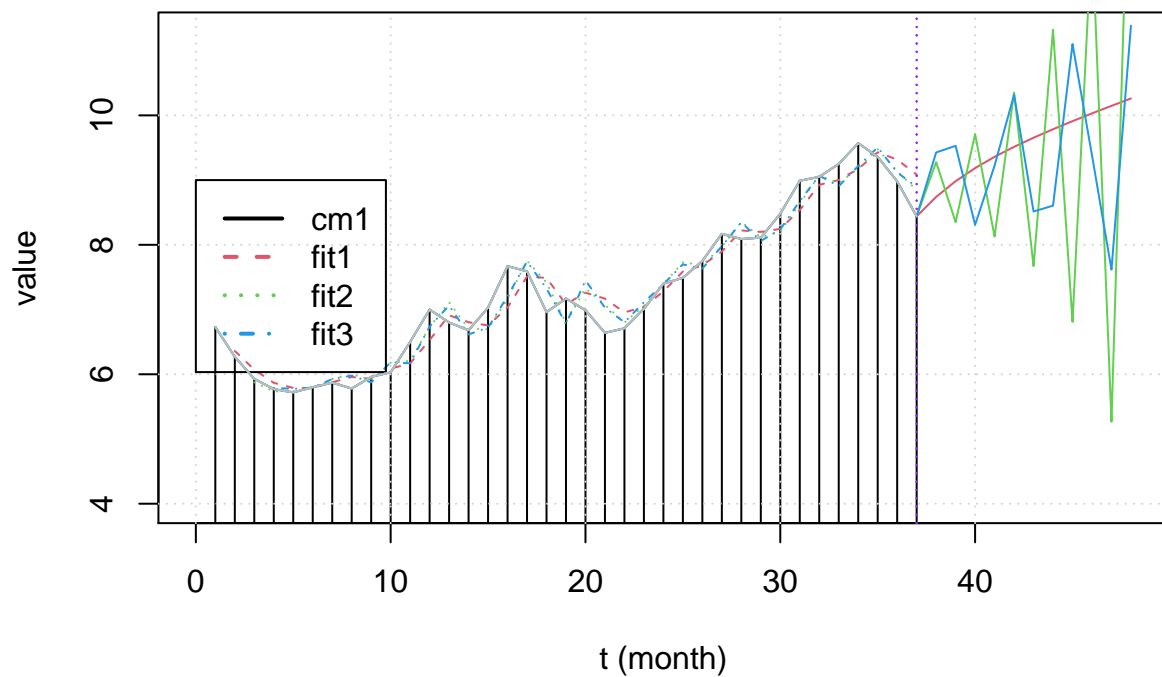
```







```
layout(1)
cm1pred = c('cm1', 'fit1', 'fit2', 'fit3')
matplot(x=cm1df[, 't'], y=cm1df[, cm1pred], type = c('h', 'l', 'l', 'l'), lty = 1:4, col = 1:4,
        xlab = 't (month)', ylab = 'value',
        xlim = c(0, n+nf),
        ylim = c(min(cm1) - 1.5*sd(cm1),
                  (max(cm1)+1.5*sd(cm1))))
grid()
legend(0, 9, legend = cm1pred, lwd=2, col = 1:4, lty = 1:4)
for(i in 1:3){
  lines(t2, get(paste0('cm1f', i)), ylab=paste0('fit2_', i), type = 'l', col = i+1)
  #points(t, cm1, col = 'red')
  abline(v = n, col = 'purple', lty = 3)
}
lines(t, cm1, col = 'grey')
```



```
for(i in 1:3) {
  print(i)
  print(get(paste0('cm1f',i)))
}
```

```
## [1] 1
## [1] 6.730000 6.270000 5.930000 5.770000 5.720000 5.800000 5.870000
## [8] 5.780000 5.960000 6.030000 6.500000 7.000000 6.800000 6.680000
## [15] 7.030000 7.670000 7.590000 6.960000 7.170000 6.990000 6.640000
## [22] 6.710000 7.010000 7.400000 7.490000 7.750000 8.170000 8.090000
## [29] 8.110000 8.480000 8.990000 9.050000 9.250000 9.570000 9.360000
## [36] 8.980000 8.440000 8.740522 8.983175 9.185455 9.359566 9.514027
## [43] 9.654777 9.785961 9.910473 10.030328 10.146936 10.261277
## [1] 2
## [1] 6.730000 6.270000 5.930000 5.770000 5.720000 5.800000 5.870000
## [8] 5.780000 5.960000 6.030000 6.500000 7.000000 6.800000 6.680000
## [15] 7.030000 7.670000 7.590000 6.960000 7.170000 6.990000 6.640000
## [22] 6.710000 7.010000 7.400000 7.490000 7.750000 8.170000 8.090000
## [29] 8.110000 8.480000 8.990000 9.050000 9.250000 9.570000 9.360000
## [36] 8.980000 8.440000 9.277369 8.350125 9.710655 8.128442 10.351280
## [43] 7.671090 11.323256 6.809497 12.836778 5.266181 15.244541
## [1] 3
## [1] 6.730000 6.270000 5.930000 5.770000 5.720000 5.800000 5.870000
## [8] 5.780000 5.960000 6.030000 6.500000 7.000000 6.800000 6.680000
## [15] 7.030000 7.670000 7.590000 6.960000 7.170000 6.990000 6.640000
## [22] 6.710000 7.010000 7.400000 7.490000 7.750000 8.170000 8.090000
```

##	[29]	8.110000	8.480000	8.990000	9.050000	9.250000	9.570000	9.360000
##	[36]	8.980000	8.440000	9.426756	9.529363	8.308289	9.223218	10.319165
##	[43]	8.514676	8.604805	11.101913	9.384021	7.617056	11.385450	