

```
require(pacman)

## Loading required package: pacman

## Warning: package 'pacman' was built under R version 4.1.3

#Load packages
pacman::p_load(datasets, tseries)
pacman::p_load(rio)
pacman::p_load(MASS)

#Membaca Data
masterdata <- import("____.xlsx") head(masterdata)

##      Penumpang
## 1      45980
## 2      36094
## 3      44646
## 4      39799
## 6      48620

tail(masterdata)

##      Penumpang
## 115     130138
## 116     95313
## 117     89222
## 118     96924
## 119     105504
## 120     116099

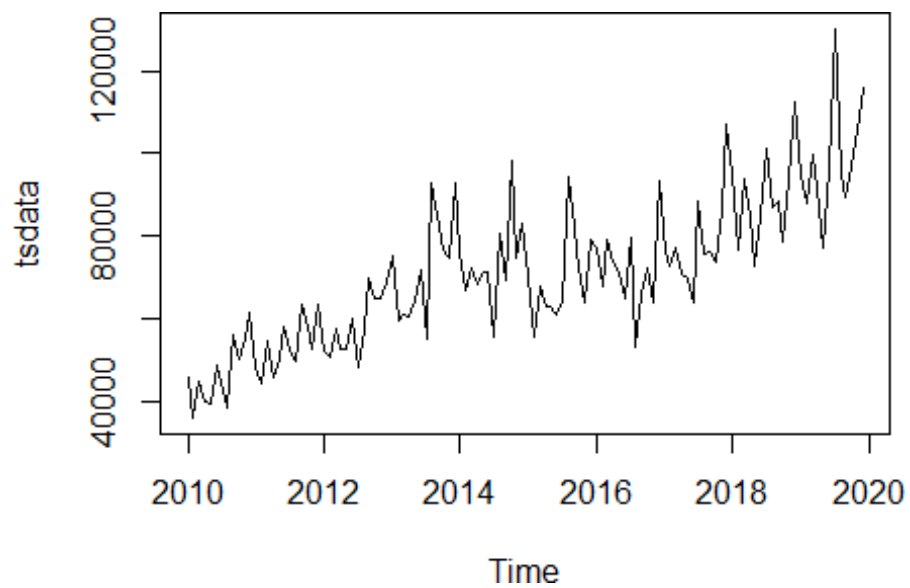
summary(masterdata)

##      Penumpang
## Min.      : 36094
## 1st Qu.: 57875
## Median : 70551
## Mean   : 71218
## 3rd Qu.: 82537
## Max.    : 130138

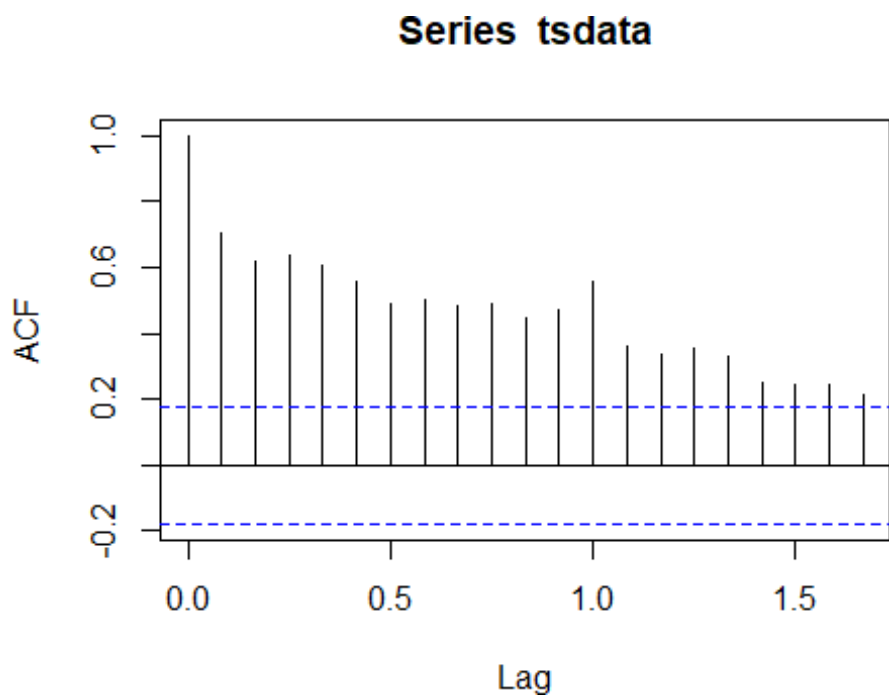
#Plot time series
tsdata <- ts(masterdata$Penumpang, frequency = 12, start = c(2010, 1))
tsdata
```

##		Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct
##	2010	45980	36094	44646	39799	39363	48620	45032	38335	55997	50353
##	2011	48350	44405	54560	45742	50127	58210	52658	49643	63390	58000
##	2012	52029	50785	57386	52720	52604	59982	48206	57501	69700	64800
##	2013	75029	59691	61132	60353	64384	71637	55163	92622	84801	76940
##	2014	76063	66733	72501	68315	71341	71210	55537	80665	69489	98232
##	2015	70408	55608	67923	63127	62931	60985	63755	94383	82329	72086
##	2016	76998	67737	78994	73520	71163	64703	79797	53219	66131	72468
##	2017	76162	72852	77381	70694	69750	64102	88245	75571	76091	73603
##	2018	95096	76779	94044	84798	72560	85175	101424	86919	88481	78428
##	2019	95023	87859	99910	91459	76965	96241	130138	95313	89222	96924
##		Nov	Dec								
##	2010	55043	61665								
##	2011	52940	63347								
##	2012	65061	68807								
##	2013	74757	92658								
##	2014	74874	83161								
##	2015	63856	78991								
##	2016	63731	93487								
##	2017	86459	107000								
##	2018	93862	112532								
##	2019	105504	116099								

```
plot(tsddata)
```

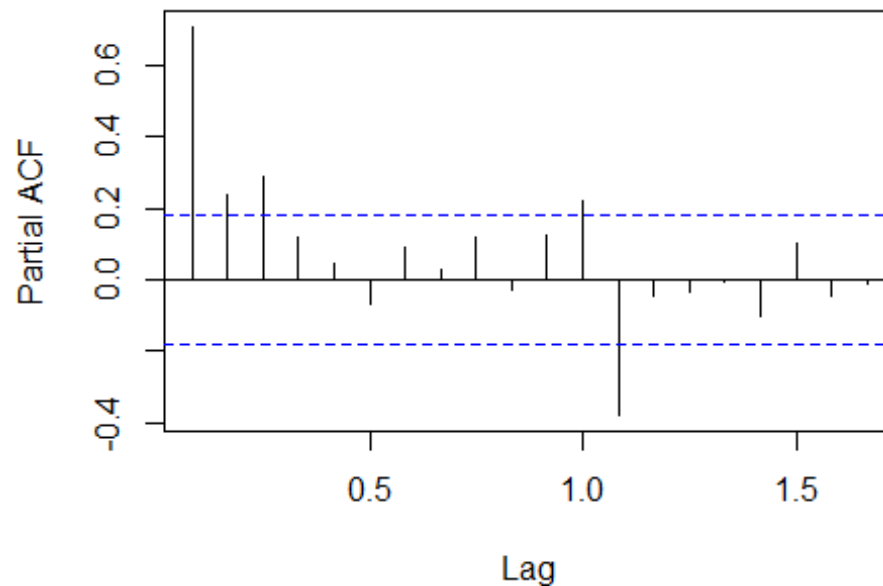


```
acf(tsdata)
```



```
pacf(tsdata)
```

### Series tsdata



```
#Cek kestasioneritasan Data
```

```
adf.test(tsdata)
```

```
##
```

```
## Augmented Dickey-Fuller Test
```

```
##
```

```
## data: tsdata
```

```
## Dickey-Fuller = -3.5351, Lag order = 4, p-value = 0.0421
```

```
## alternative hypothesis: stationary
```

```
#transformasi box-cox
```

```
library(TSA)
```

```
## Warning: package 'TSA' was built under R version 4.1.3
```

```
##
```

```
## Attaching package: 'TSA'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## acf, arima
```

```
## The following object is masked from 'package:utils':
```

```
##
```

```
## tar
```

[illegible]

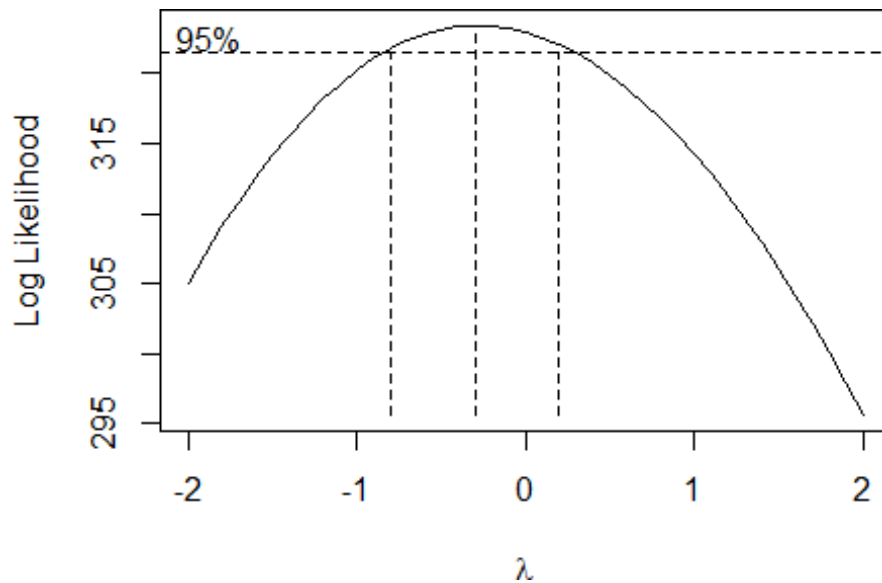
[illegible]

[illegible]

[illegible]



```
## Warning in arima0(x, order = c(i, 0L, 0L), include.mean = demean): possible
## convergence problem: optim gave code = 1
```



```
#Print data transformasi boxcox
```

```
boxcox <- bcPower(tsddata, 0)
boxcox
```

```
##           Jan      Feb      Mar      Apr      May      Jun      Jul      A
ug
## 2010 10.73596 10.49388 10.70652 10.59160 10.58058 10.79179 10.71513 10.554
12
## 2011 10.78622 10.70111 10.90706 10.73077 10.82232 10.97181 10.87157 10.812
61
## 2012 10.85956 10.83536 10.95756 10.87275 10.87055 11.00180 10.78324 10.959
56
## 2013 11.22563 10.99694 11.02079 11.00797 11.07262 11.17937 10.91805 11.436
28
## 2014 11.23932 11.10845 11.19136 11.13188 11.17523 11.17339 10.92480 11.298
06
## 2015 11.16206 10.92608 11.12613 11.05290 11.04979 11.01838 11.06280 11.455
12
## 2016 11.25153 11.12339 11.27713 11.20531 11.17273 11.07756 11.28724 10.882
17
## 2017 11.24062 11.19619 11.25650 11.16612 11.15267 11.06823 11.38787 11.232
83
## 2018 11.46264 11.24869 11.45152 11.34803 11.19217 11.35246 11.52707 11.372
```

73

```
## 2019 11.46187 11.38349 11.51203 11.42365 11.25111 11.47461 11.77635 11.46492
```

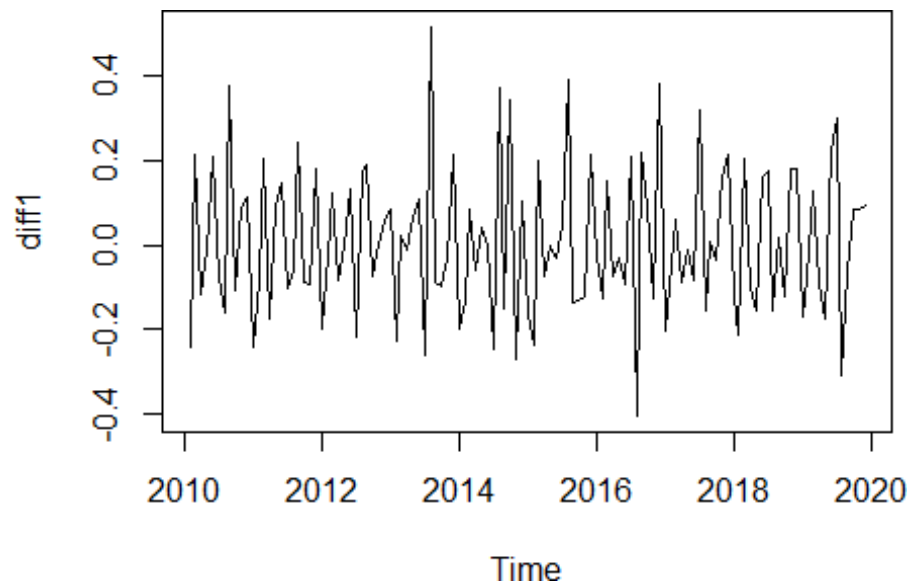
```
##          Sep      Oct      Nov      Dec
## 2010 10.93305 10.82681 10.91587 11.02947
## 2011 11.05706 10.96820 10.87691 11.05638
## 2012 11.15196 11.07906 11.08308 11.13906
## 2013 11.34806 11.25078 11.22200 11.43667
## 2014 11.14892 11.49509 11.22356 11.32853
## 2015 11.31848 11.18562 11.06439 11.27709
## 2016 11.09939 11.19090 11.06243 11.44558
## 2017 11.23969 11.20644 11.36743 11.58058
## 2018 11.39054 11.26994 11.44958 11.63099
## 2019 11.39888 11.48168 11.56650 11.66220
```

*#diff 1*

```
diff1 <- diff(boxcox, 1)
diff1
```

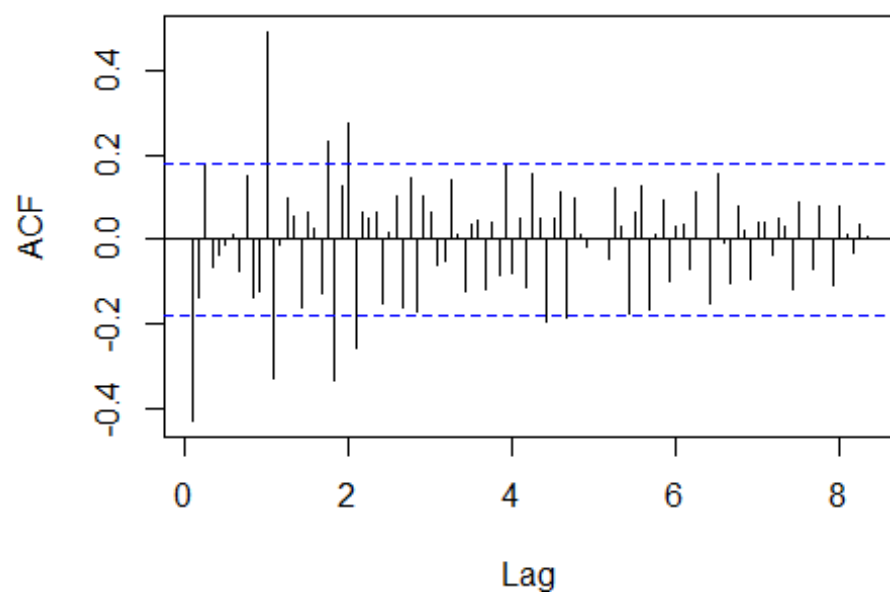
```
##          Jan      Feb      Mar      Apr      May
## 2010          -0.242079873  0.212638071 -0.114922931 -0.011015498
## 2011 -0.243250287 -0.085114146  0.205948938 -0.176284100  0.091542872
## 2012 -0.196826294 -0.024200220  0.122199336 -0.084805481 -0.002202728
## 2013  0.086569222 -0.228693450  0.023854205 -0.012824804  0.064654499
## 2014 -0.197353351 -0.130862361  0.082900772 -0.059470993  0.043341835
## 2015 -0.166471595 -0.235979817  0.200047635 -0.073226140 -0.003109682
## 2016 -0.025554474 -0.128146888  0.153739341 -0.071814422 -0.032584458
## 2017 -0.204959739 -0.044432665  0.060311286 -0.090380569 -0.013443283
## 2018 -0.117941927 -0.213955742  0.202831593 -0.103490800 -0.155858152
## 2019 -0.169118658 -0.078385711  0.128536524 -0.088378996 -0.172540011
##          Jun      Jul      Aug      Sep      Oct
## 2010  0.211208680 -0.076661621 -0.161010031  0.378934801 -0.106239918
## 2011  0.149497376 -0.100238987 -0.058960780  0.244448727 -0.088863110
## 2012  0.131252355 -0.218561023  0.176318844  0.192397979 -0.072894714
## 2013  0.106746545 -0.261319262  0.518234256 -0.088219359 -0.097281438
## 2014 -0.001837939 -0.248583793  0.373255311 -0.149136309  0.346163561
## 2015 -0.031410956  0.044419680  0.392313360 -0.136637557 -0.132863564
## 2016 -0.095165452  0.209678341 -0.405070434  0.217222148  0.091507462
## 2017 -0.084441856  0.319641472 -0.155044425  0.006857381 -0.033244207
## 2018  0.160294158  0.174601786 -0.154333099  0.017811189 -0.120606833
## 2019  0.223504689  0.301739964 -0.311429213 -0.066038567  0.082799520
##          Nov      Dec
## 2010  0.089056498  0.113601811
## 2011 -0.091283814  0.179468353
## 2012  0.004019688  0.055980193
## 2013 -0.028783044  0.214672442
## 2014 -0.271525327  0.104971787
## 2015 -0.121229302  0.212703374
## 2016 -0.128473985  0.383151289
## 2017  0.160984527  0.213158521
```

```
## 2018 0.179644612 0.181412007  
## 2019 0.084821701 0.095694409  
plot(diff1)
```



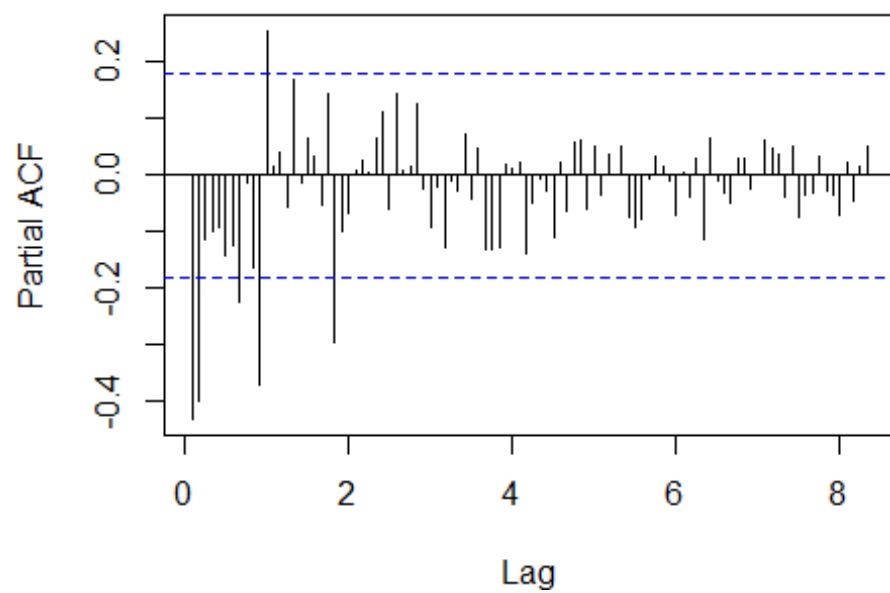
```
acf(diff1, lag.max = 100)
```

**Series diff1**



```
pacf(diff1, lag.max=100)
```

**Series diff1**



```
adf.test(diff1)
```

```
## Warning in adf.test(diff1): p-value smaller than printed p-value
```

```
##
```

```
## Augmented Dickey-Fuller Test
```

```
##
```

```
## data: diff1
```

```
## Dickey-Fuller = -6.6756, Lag order = 4, p-value = 0.01
```

```
## alternative hypothesis: stationary
```

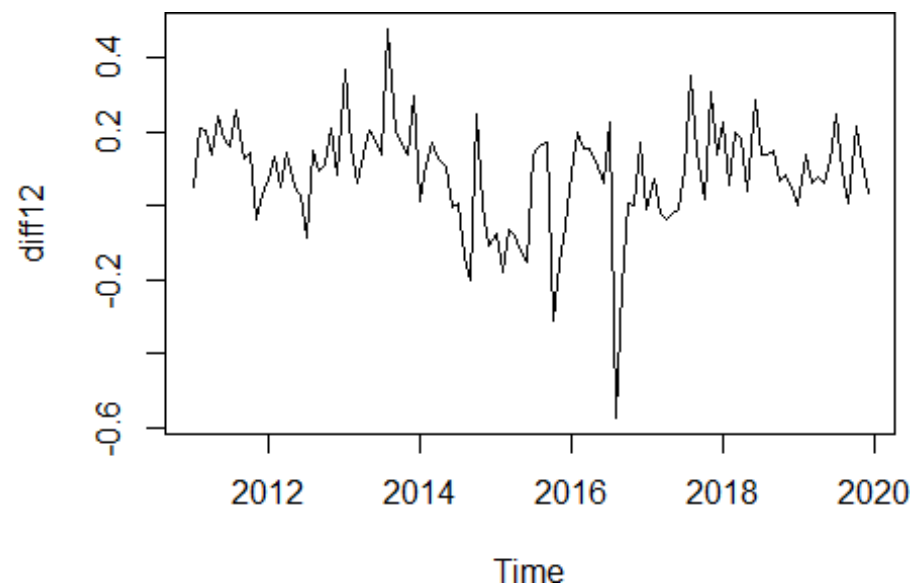
```
#diff 12
```

```
diff12 <- diff(boxcox, 12)
```

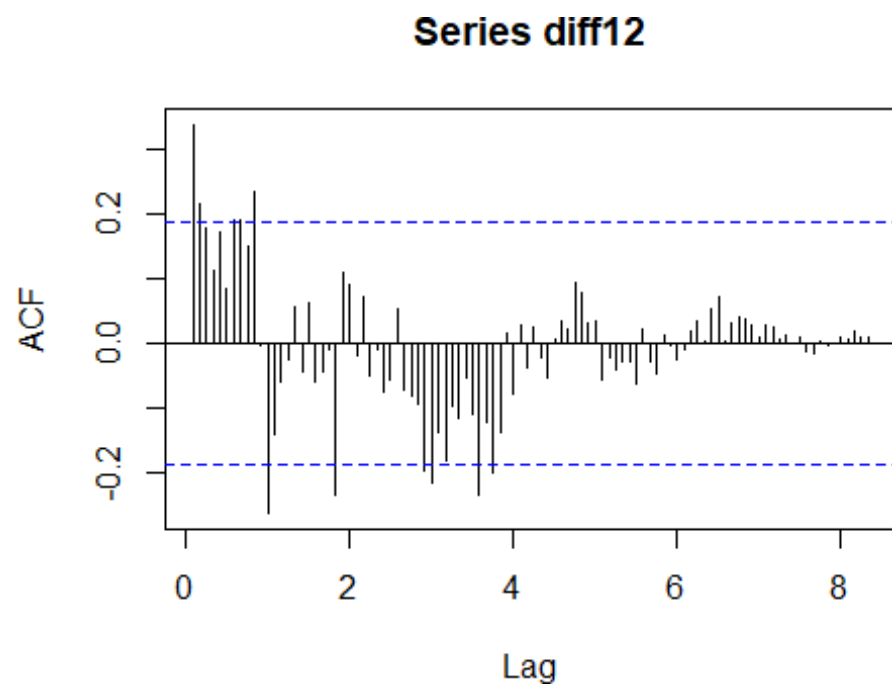
```
diff12
```

```
##           Jan           Feb           Mar           Apr           May
## 2011  0.0502597026  0.2072254292  0.2005362959  0.1391751267  0.2417334963
## 2012  0.0733350335  0.1342489597  0.0504993576  0.1419779772  0.0482323774
## 2013  0.3660734500  0.1615802199  0.0632350895  0.1352157661  0.2020729926
## 2014  0.0136872388  0.1115183279  0.1705648942  0.1239187055  0.1026060422
## 2015 -0.0772550512 -0.1823725074 -0.0652256442 -0.0789807915 -0.1254323091
## 2016  0.0894725544  0.1973054836  0.1509971896  0.1524089078  0.1229341323
## 2017 -0.0109167968  0.0727974265 -0.0206306278 -0.0391967744 -0.0200555997
## 2018  0.2220242570  0.0525011793  0.1950214854  0.1819112539  0.0394963851
## 2019 -0.0007679401  0.1348020915  0.0605070229  0.0756188268  0.0589369673
##           Jun           Jul           Aug           Sep           Oct
## 2011  0.1800221924  0.1564448257  0.2584940766  0.1240080024  0.1413848102
## 2012  0.0299873559 -0.0883346792  0.1469449453  0.0948941975  0.1108625928
## 2013  0.1775671831  0.1348089440  0.4767243555  0.1961070174  0.1717202940
## 2014 -0.0059784423  0.0067570266 -0.1382219183 -0.1991388685  0.2443061305
## 2015 -0.1550053257  0.1379981470  0.1570561961  0.1695549483 -0.3094721772
## 2016  0.0591796360  0.2244382974 -0.5729454967 -0.2190857917  0.0052852344
## 2017 -0.0093320036  0.1006311272  0.3506571363  0.1402923691  0.0155407007
## 2018  0.2842323989  0.1391927126  0.1399040388  0.1508578473  0.0634952207
## 2019  0.1221474988  0.2492856763  0.0921895622  0.0083398062  0.2117461598
##           Nov           Dec
## 2011 -0.0389555012  0.0269110407
## 2012  0.2061660946  0.0826779343
## 2013  0.1389175620  0.2976098115
## 2014  0.0015638474 -0.1081368075
## 2015 -0.1591761524 -0.0514445660
## 2016 -0.0019594479  0.1684884673
## 2017  0.3049992127  0.1350064453
## 2018  0.0821553053  0.0504087912
## 2019  0.1169232484  0.0312056498
```

```
plot(diff12)
```

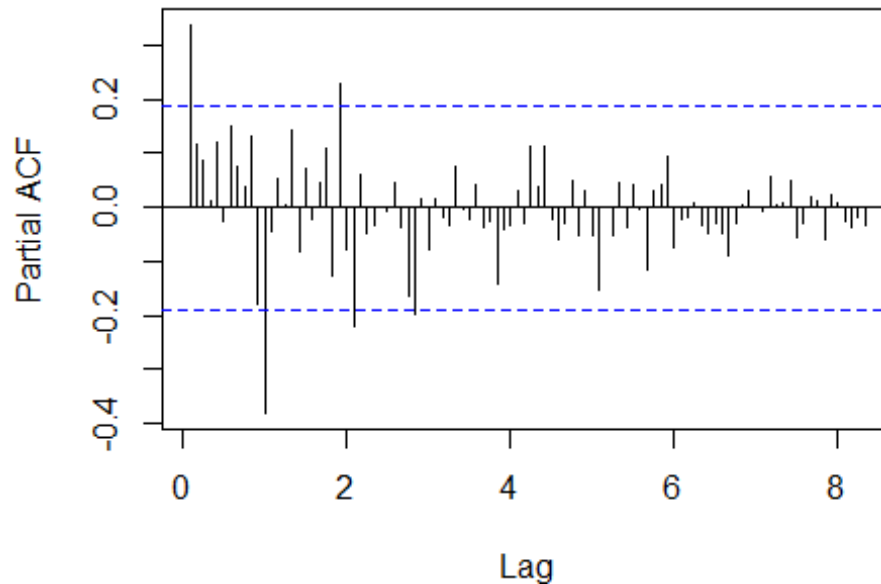


```
acf(diff12, lag.max = 100)
```



```
pacf(diff12, lag.max=100)
```

## Series diff12



```
adf.test(diff12)

##
## Augmented Dickey-Fuller Test
##
## data: diff12
## Dickey-Fuller = -3.1647, Lag order = 4, p-value = 0.0974
## alternative hypothesis: stationary

##Estimasi Model
library(astsa)

## Warning: package 'astsa' was built under R version 4.1.3

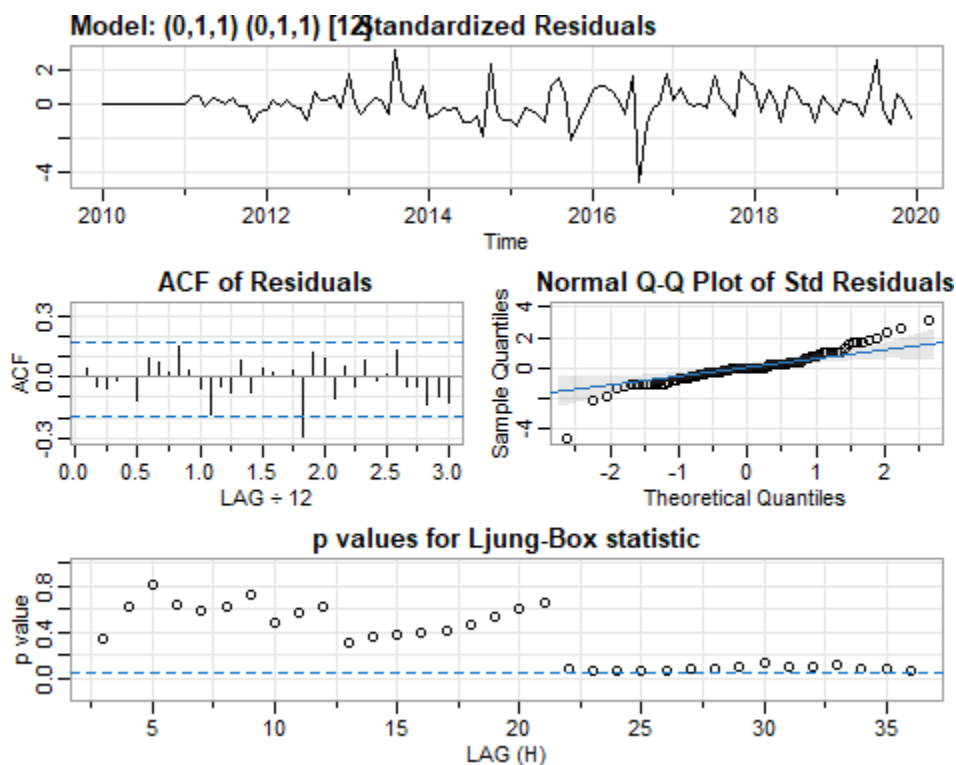
Imasima <- arima(boxcox, order = c(0,1,1), seasonal = list(order = c(0,1,1),
period = 12), include.mean = FALSE)
Imasima

##
## Call:
## arima(x = boxcox, order = c(0, 1, 1), seasonal = list(order = c(0, 1, 1),
period = 12),
## include.mean = FALSE)
##
## Coefficients:
##          ma1      sma1
##       -0.7919  -0.4352
## s.e.   0.0603   0.1064
```

```
##
## sigma^2 estimated as 0.01533: log likelihood = 69.92, aic = -135.84

sarima(tldata, 0,1,1, 0,1,1, 12)

## initial   value 9.413054
## iter      2 value 9.218819
## iter      3 value 9.182209
## iter      4 value 9.144682
## iter      5 value 9.133233
## iter      6 value 9.132268
## iter      7 value 9.131821
## iter      8 value 9.131492
## iter      9 value 9.131487
## iter      9 value 9.131487
## iter      9 value 9.131487
## final    value 9.131487
## converged
## initial   value 9.142397
## iter      2 value 9.141501
## iter      3 value 9.141477
## iter      4 value 9.141469
## iter      4 value 9.141469
## iter      4 value 9.141469
## final    value 9.141469
## converged
```





```

## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), p
period = S),
##      include.mean = !no.constant, transform.pars = trans, fixed = fixed, op
tim.control = list(trace = trc,
##          REPORT = 1, reltol = tol))
##
## Coefficients:
##          ma1      sma1
##      -0.8016  -0.4134
## s.e.    0.0590   0.1003
##
## sigma^2 estimated as 84456877: log likelihood = -1129.96, aic = 2265.93
##
## $degrees_of_freedom
## [1] 105
##
## $ttable
##      Estimate      SE t.value p.value
## ma1   -0.8016 0.0590 -13.5807  0e+00
## sma1  -0.4134 0.1003  -4.1212  1e-04
##
## $AIC
## [1] 21.17689
##
## $AICc
## [1] 21.17797
##
## $BIC
## [1] 21.25183

library(lmtest)

## Warning: package 'lmtest' was built under R version 4.1.3

## Loading required package: zoo

## Warning: package 'zoo' was built under R version 4.1.3

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric

#tes coef
coeftest(Imasima)

```

```
##
## z test of coefficients:
##
##      Estimate Std. Error  z value  Pr(>|z|)
## ma1  -0.791906    0.060322 -13.1280 < 2.2e-16 ***
## sma1 -0.435152    0.106372  -4.0908 4.298e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

#Uji white noise residual
Box.test(Imasima$residuals,type = "Ljung")

##
## Box-Ljung test
##
## data:  Imasima$residuals
## X-squared = 0.039957, df = 1, p-value = 0.8416

#Uji Normalitas residual
ks.test(Imasima$residuals,"pnorm",mean=0, sd=sd(Imasima$residuals))

##
## One-sample Kolmogorov-Smirnov test
##
## data:  Imasima$residuals
## D = 0.10395, p-value = 0.1495
## alternative hypothesis: two-sided

#Forecasting

library(forecast)

library(sarima)

forecasting <- forecast(boxcox, model = Imasima, h = 12)

forecasting

plot(forecasting, main = "Plot Forecasting")
```

Description: df [12 x 5]

	Point Forecast <dbl>	Lo 80 <dbl>	Hi 80 <dbl>	Lo 95 <dbl>	Hi 95 <dbl>
Jan 2020	11.58865	11.42998	11.74732	11.34599	11.83131
Feb 2020	11.47403	11.31197	11.63610	11.22617	11.72189
Mar 2020	11.61544	11.45004	11.78084	11.36249	11.86839
Apr 2020	11.52468	11.35602	11.69334	11.26674	11.78262
May 2020	11.38659	11.21473	11.55845	11.12375	11.64943
Jun 2020	11.53897	11.36396	11.71397	11.27132	11.80661
Jul 2020	11.79349	11.61540	11.97158	11.52112	12.06585
Aug 2020	11.55762	11.37649	11.73874	11.28061	11.83463
Sep 2020	11.53180	11.34769	11.71592	11.25023	11.81338
Oct 2020	11.54938	11.36233	11.73643	11.26331	11.83544

1-10 of 12 rows

Previous   Next

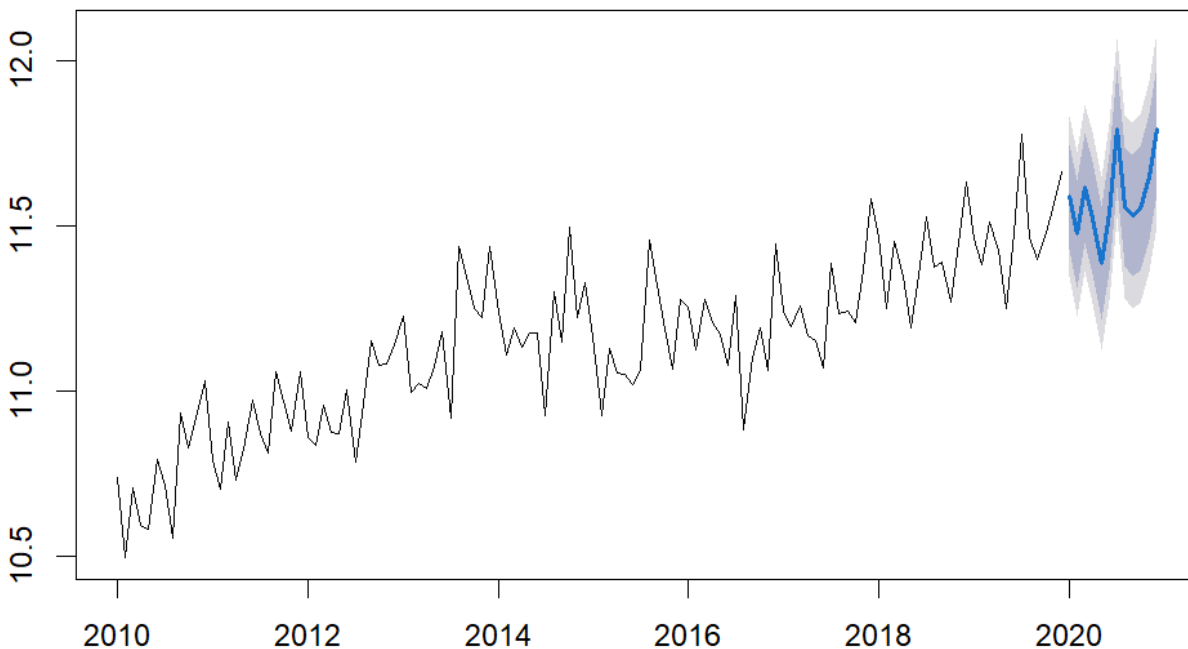
Description: df [12 x 5]

	Point Forecast <dbl>	Lo 80 <dbl>	Hi 80 <dbl>	Lo 95 <dbl>	Hi 95 <dbl>
Nov 2020	11.64770	11.45776	11.83764	11.35721	11.93819
Dec 2020	11.79332	11.60053	11.98611	11.49848	12.08817

11-12 of 12 rows

Previous   Next

## Plot Forecasting



```

#MAPE
tsdata.fitted <- fitted(Imasima)
nilaimape <- mean(abs(tsdata-(tsdata.fitted))/(tsdata))*100
nilaimape

## [1] 99.9834

#uji residual
t.test(Imasima$residuals, mu = 0, alternative = "two.sided")

##
## One Sample t-test
##
## data: Imasima$residuals
## t = -0.22795, df = 119, p-value = 0.8201
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.02366969 0.01878252
## sample estimates:
## mean of x
## -0.002443587

#Kesimpulan
print(Imasima)

##
## Call:
## arima(x = boxcox, order = c(0, 1, 1), seasonal = list(order = c(0, 1, 1),
## period = 12),
## include.mean = FALSE)
##
## Coefficients:
##          ma1          sma1
##       -0.7919       -0.4352
## s.e.    0.0603    0.1064
##
## sigma^2 estimated as 0.01533: log likelihood = 69.92, aic = -135.84

```