

test1

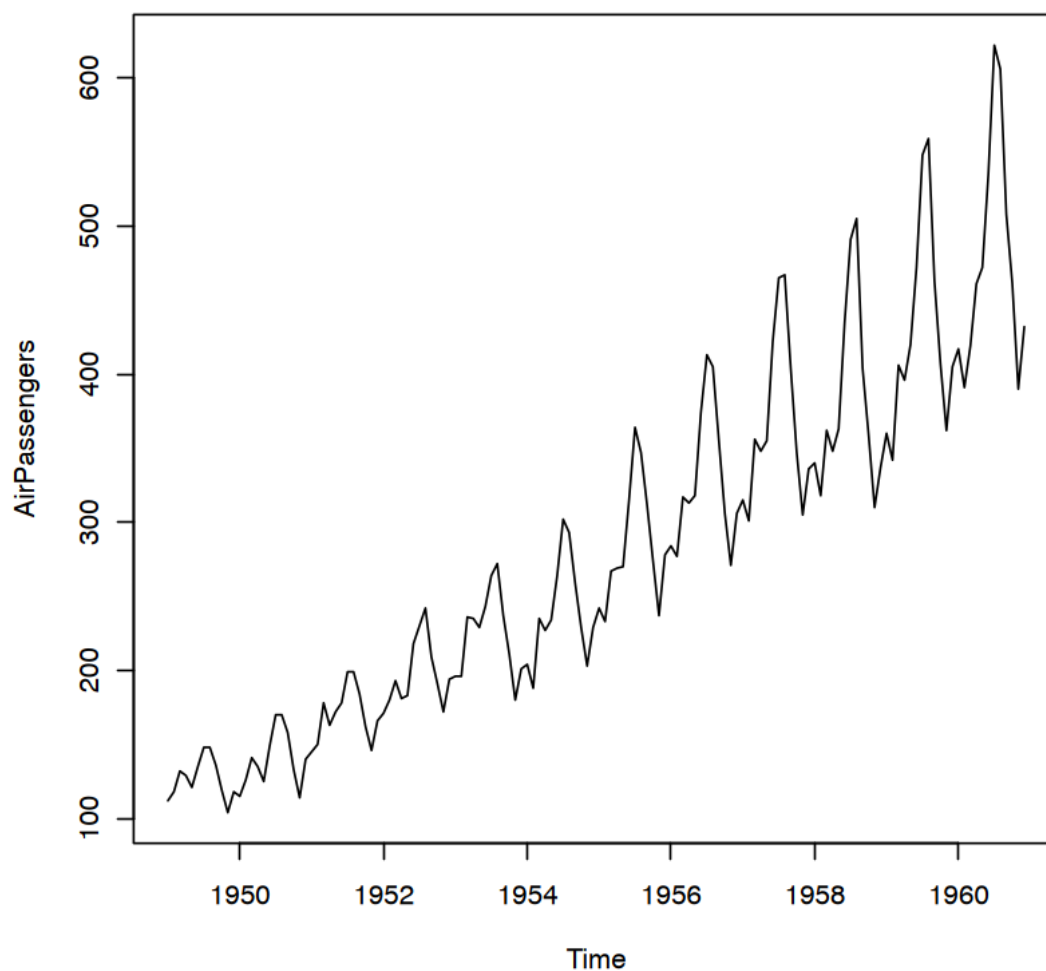
November 26, 2024

```
[2]: library(astsa)

library(datasets)

## dataset
data("AirPassengers")
summary(AirPassengers)
plot(AirPassengers)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
104.0	180.0	265.5	280.3	360.5	622.0



```
[3]: data <- ts(AirPassengers)
      sarima.for(data, xreg = data.frame(time = 1:length(data)), newxreg = data.
      ↪frame(time = 1:24 + length(data)), 24, 1, 0, 0, 0, 0, 1, 12)
```

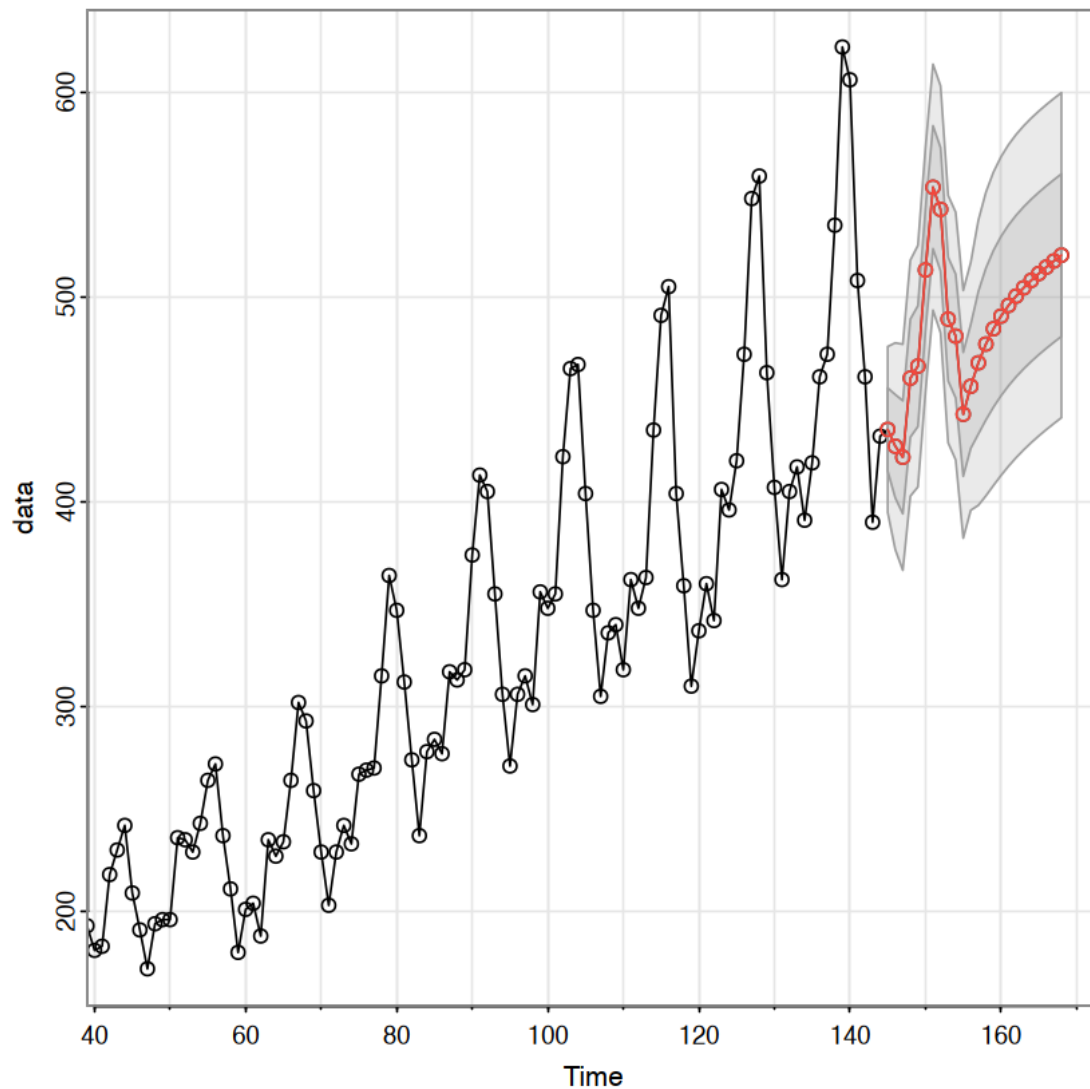
\$pred A Time Series:

```
1. 435.296404448951 2. 427.132067810041 3. 421.740058102871 4. 460.354001689686
5. 466.215317459882 6. 513.152786883938 7. 553.530511194164 8. 542.737439328013
9. 489.167182676446 10. 480.895509645266 11. 442.72359302425 12. 456.381396774694
13. 467.826636698122 14. 476.9821798826 15. 484.434921417891 16. 490.621323438804
17. 495.865974101629 18. 500.410263123371 19. 504.433707206489 20. 508.069809330573
21. 511.417852990222 22. 514.551673333763 23. 517.526180083603 24. 520.382208501587
```

\$se A Time Series:

```
1. 20.2369465077224 2. 25.2153420831165 3. 27.5854874380948 4. 28.8126935603087
```

5. 29.4694463939624 6. 29.8264137556358 7. 30.0219320156396 8. 30.1293703559624
 9. 30.1883551446478 10. 30.2204386342557 11. 30.2372840889865 12. 30.2449986195653
 13. 34.7770323714392 14. 37.0461712998146 15. 38.2433623059643 16. 38.8896560011889
 17. 39.2425241796212 18. 39.436325454688 19. 39.5431012580231 20. 39.6020310364572
 21. 39.6345851062503 22. 39.6525779796707 23. 39.6625256114465 24. 39.6680261766189



```
[4]: DIR <- "/Users/gianlucastrantonio/Dropbox (Politecnico di Torino Staff)/
      ↳didattica/DataSpaces/dispense/Time Series/"

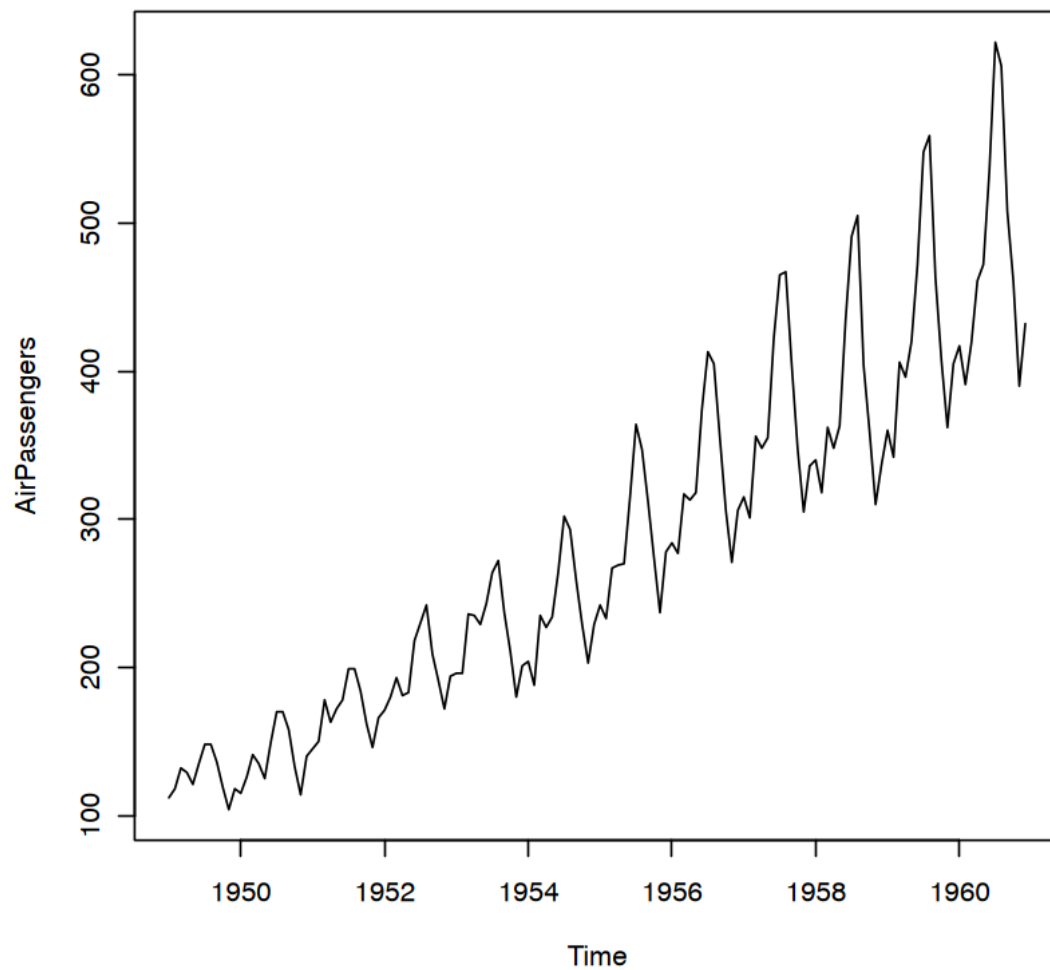
      # remotes::install_github("nickpoison/astsa/astsa_build")
      library(astsa)
      library(astsa)
      library(datasets)
```

```
## dataset
data("AirPassengers")
summary(AirPassengers)

# prendiamo lo stesso dataset precedente e confrontiamo modelli ARIMA

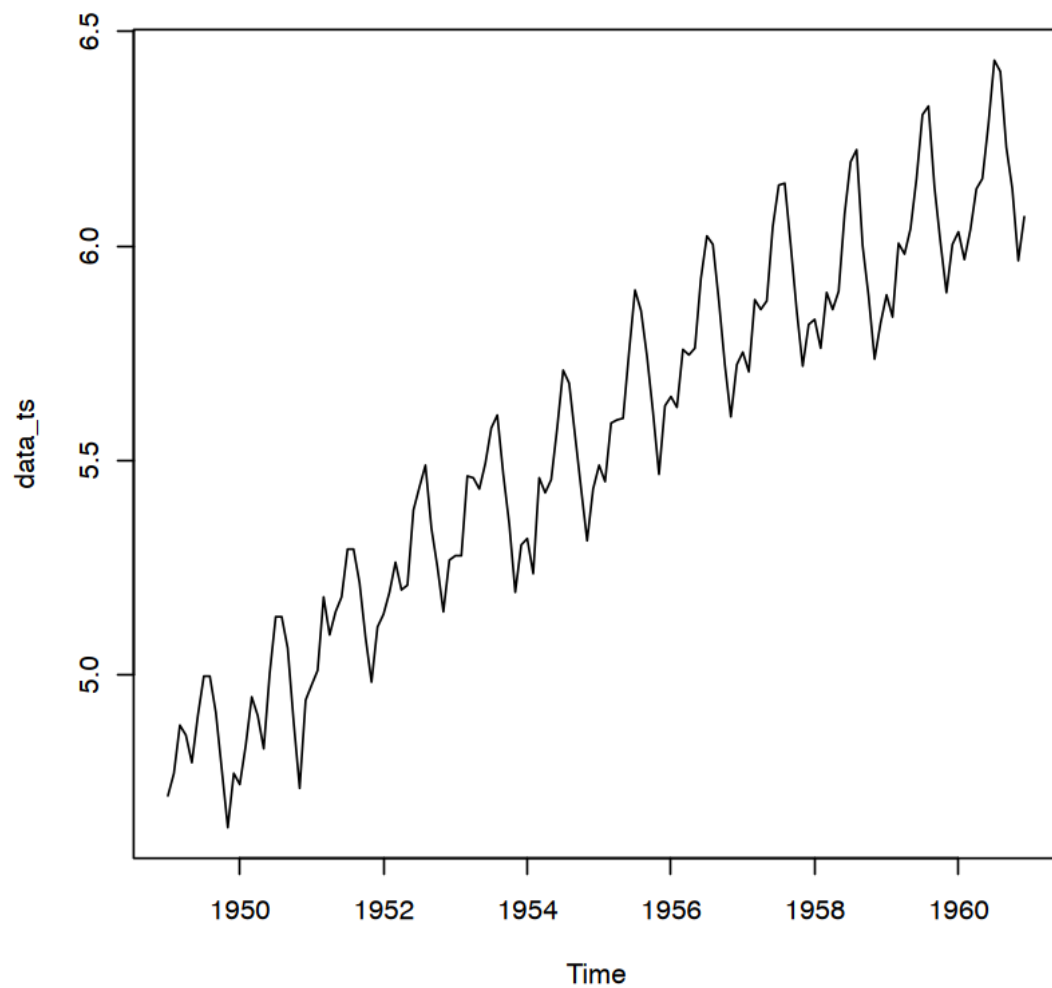
# vediamo dei plot
plot(AirPassengers)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
104.0	180.0	265.5	280.3	360.5	622.0



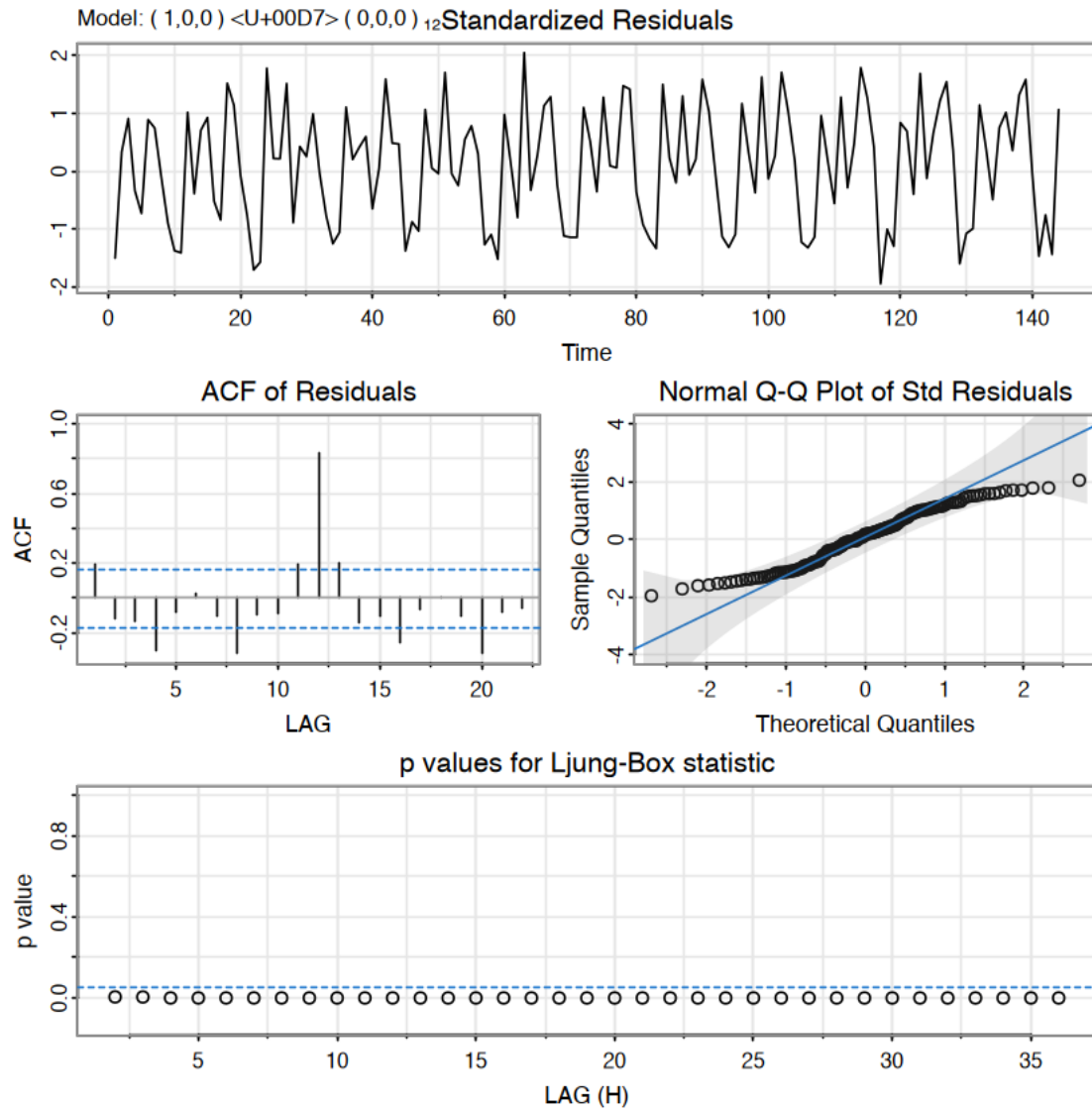
Applico la trasformazione logartmica

```
[5]: data_ts = log(AirPassengers)
      plot(data_ts)
```



```
[6]: data_ts <- ts(data_ts)
      sarima(data_ts,
      1, # AR
      0, # I
      0, # MA
      0, # AR_s
      0, # I_s
      0, # MA_s
      12)
```

```
initial value -0.829999
iter 2 value -2.248560
iter 3 value -2.250003
iter 4 value -2.250493
iter 5 value -2.250645
iter 6 value -2.251785
iter 7 value -2.252303
iter 8 value -2.252413
iter 9 value -2.252541
iter 10 value -2.252943
iter 11 value -2.253590
iter 12 value -2.253823
iter 13 value -2.253841
iter 14 value -2.253859
iter 15 value -2.253919
iter 16 value -2.254000
iter 17 value -2.254051
iter 18 value -2.254055
iter 19 value -2.254063
iter 20 value -2.254082
iter 21 value -2.254113
iter 22 value -2.254129
iter 23 value -2.254130
iter 24 value -2.254132
iter 25 value -2.254136
iter 26 value -2.254143
iter 27 value -2.254147
iter 28 value -2.254147
iter 29 value -2.254148
iter 30 value -2.254149
iter 31 value -2.254151
iter 32 value -2.254152
iter 32 value -2.254152
final value -2.254152
converged
initial value -2.222794
iter 2 value -2.224612
iter 3 value -2.228908
iter 4 value -2.229816
iter 5 value -2.230136
iter 6 value -2.230206
iter 7 value -2.230224
iter 8 value -2.231597
iter 9 value -2.231670
iter 10 value -2.231714
iter 11 value -2.231729
iter 12 value -2.231741
```

```
[7]: mod <- sarima(
  data_ts,
  1, # AR
  0, # I
  0, # MA
  1, # AR_s
  0, # I_s
  0, # MA_s
  12
)
```

```
initial value -0.912435
iter 2 value -3.106051
```



```
iter    3 value -3.118288
iter    4 value -3.118682
iter    5 value -3.118841
iter    6 value -3.118922
iter    7 value -3.119519
iter    8 value -3.121222
iter    9 value -3.122354
iter   10 value -3.122715
iter   11 value -3.123006
iter   12 value -3.123424
iter   13 value -3.123719
iter   14 value -3.127010
iter   15 value -3.142954
iter   16 value -3.152645
iter   17 value -3.163801
iter   18 value -3.173404
iter   19 value -3.180734
iter   20 value -3.181914
iter   21 value -3.182794
iter   22 value -3.183073
iter   23 value -3.183084
iter   24 value -3.183338
iter   25 value -3.186227
iter   26 value -3.188607
iter   27 value -3.191853
iter   28 value -3.192674
iter   29 value -3.193119
iter   30 value -3.193122
iter   31 value -3.193131
iter   32 value -3.193137
iter   33 value -3.193138
iter   34 value -3.193154
iter   35 value -3.193178
iter   36 value -3.193223
iter   37 value -3.193225
iter   38 value -3.193248
iter   39 value -3.193253
iter   40 value -3.193254
iter   41 value -3.193267
iter   42 value -3.193278
iter   43 value -3.193290
iter   44 value -3.193291
iter   45 value -3.193292
iter   46 value -3.193293
iter   46 value -3.193293
iter   46 value -3.193293
final  value -3.193293
converged
```

```
initial value -2.333650
iter 2 value -2.982095
iter 3 value -2.988231
iter 4 value -3.017026
iter 5 value -3.018893
iter 6 value -3.020718
iter 7 value -3.021906
iter 8 value -3.024465
iter 9 value -3.024555
iter 10 value -3.024714
iter 11 value -3.024751
iter 12 value -3.024864
iter 13 value -3.024942
iter 14 value -3.024985
iter 15 value -3.025017
iter 16 value -3.025037
iter 17 value -3.025210
iter 18 value -3.025257
iter 19 value -3.025364
iter 20 value -3.025422
iter 21 value -3.025507
iter 22 value -3.025607
iter 23 value -3.025686
iter 24 value -3.025814
iter 25 value -3.025843
iter 26 value -3.025957
iter 27 value -3.026058
iter 28 value -3.026131
iter 29 value -3.026181
iter 30 value -3.026214
iter 31 value -3.026529
iter 32 value -3.026567
iter 33 value -3.026660
iter 34 value -3.026705
iter 35 value -3.026822
iter 36 value -3.027039
iter 37 value -3.027199
iter 38 value -3.027353
iter 39 value -3.027372
iter 40 value -3.027471
iter 41 value -3.027632
iter 42 value -3.027950
iter 43 value -3.028316
iter 44 value -3.028456
iter 45 value -3.029088
iter 46 value -3.029223
iter 47 value -3.029538
iter 48 value -3.029624
```

```
iter 49 value -3.029652
iter 50 value -3.029763
iter 51 value -3.029803
iter 52 value -3.030163
iter 53 value -3.030200
iter 54 value -3.030426
iter 55 value -3.030635
iter 56 value -3.031067
iter 57 value -3.031960
iter 58 value -3.032545
iter 59 value -3.033705
iter 60 value -3.033902
iter 61 value -3.034408
iter 62 value -3.034738
iter 63 value -3.034781
iter 64 value -3.034848
iter 65 value -3.034883
iter 66 value -3.035200
iter 67 value -3.035299
iter 68 value -3.035548
iter 69 value -3.035819
iter 70 value -3.036228
iter 71 value -3.036968
iter 72 value -3.037472
iter 73 value -3.038425
iter 74 value -3.038632
iter 75 value -3.038890
iter 76 value -3.039009
iter 77 value -3.039086
iter 78 value -3.039114
iter 79 value -3.039137
iter 80 value -3.039286
iter 81 value -3.039349
iter 82 value -3.039634
iter 83 value -3.040038
iter 84 value -3.040541
iter 85 value -3.040854
iter 86 value -3.041001
iter 87 value -3.041532
iter 88 value -3.041584
iter 89 value -3.041587
iter 90 value -3.041588
iter 91 value -3.041589
iter 92 value -3.041591
iter 93 value -3.041591
iter 94 value -3.041593
iter 95 value -3.041594
iter 96 value -3.041598
```

```

iter 97 value -3.041602
iter 98 value -3.041608
iter 99 value -3.041613
iter 100 value -3.041614
final value -3.041614
stopped after 100 iterations

```

```

Warning message in arima(xdata, order = c(p, d, q), seasonal = list(order = c(P,
:
"possibile errore di convergenza: optim ha restituito codice = 1"

```

```

<><><><><><><><><><><><><><><>

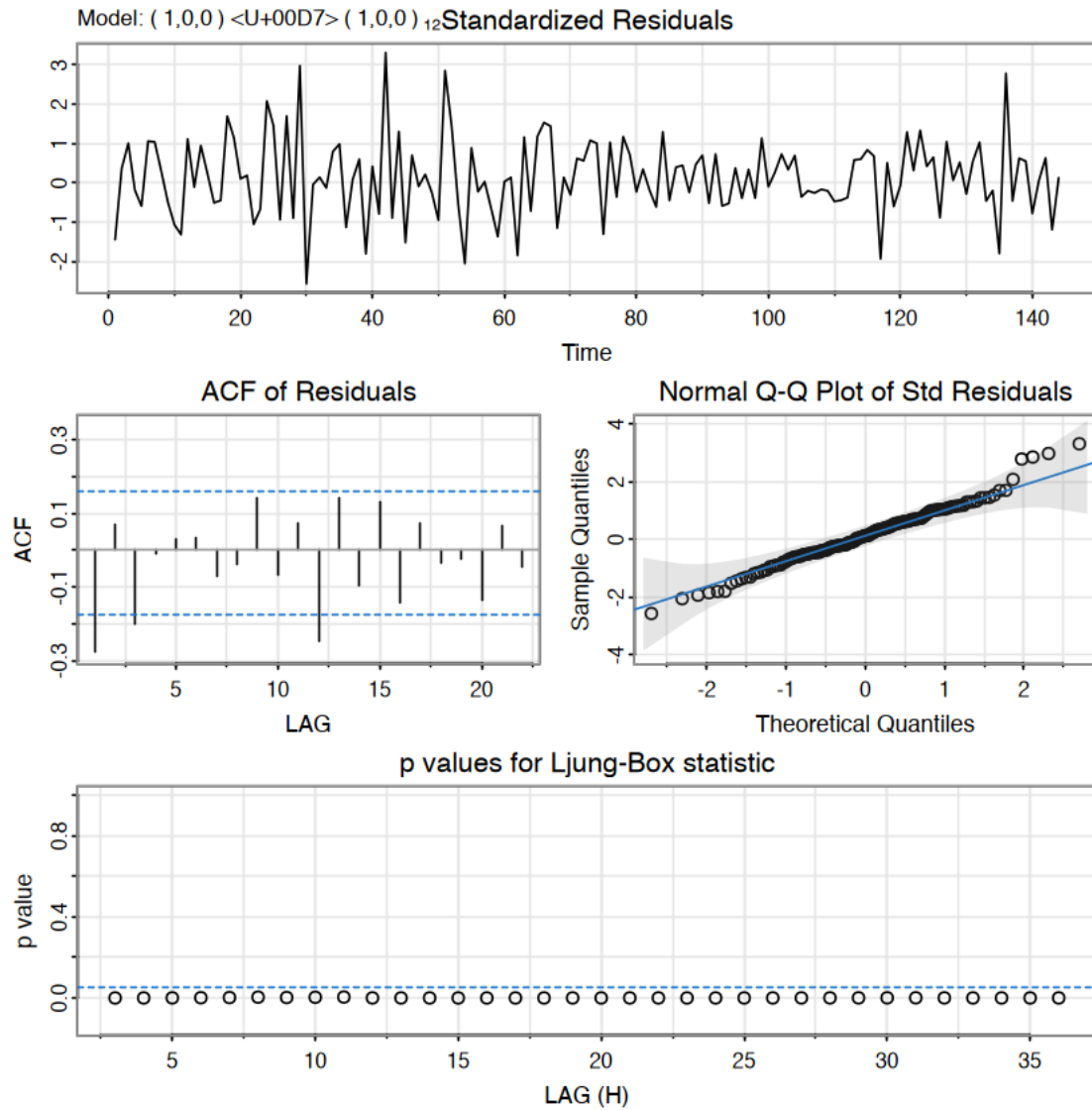
```

Coefficients:

	Estimate	SE	t.value	p.value
ar1	0.9488	0.0220	43.1396	0
sar1	0.9094	0.0271	33.5774	0
xmean	5.5325	0.4599	12.0306	0

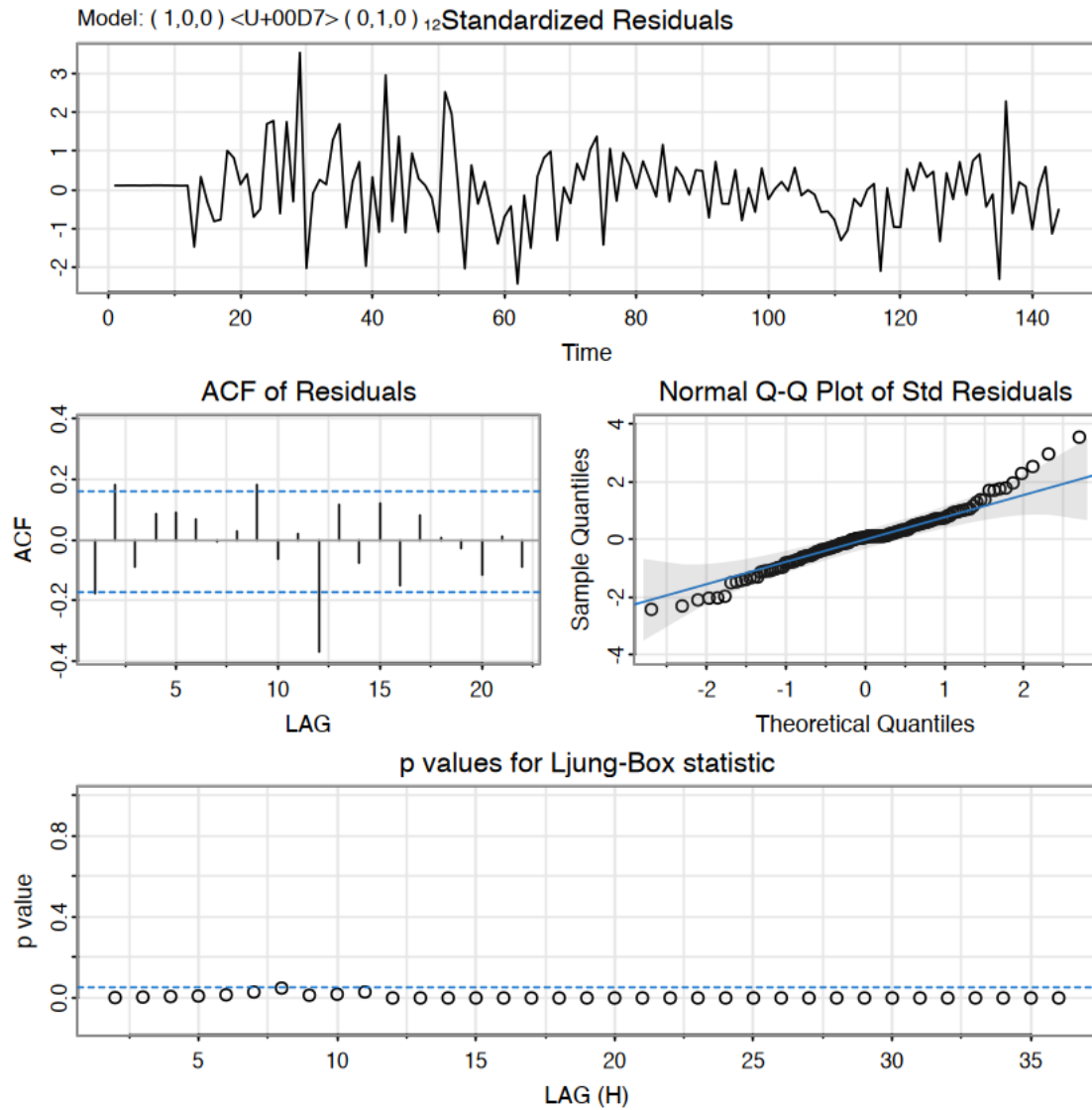
sigma^2 estimated as 0.001921555 on 141 degrees of freedom

AIC = -3.189796 AICc = -3.188606 BIC = -3.107301



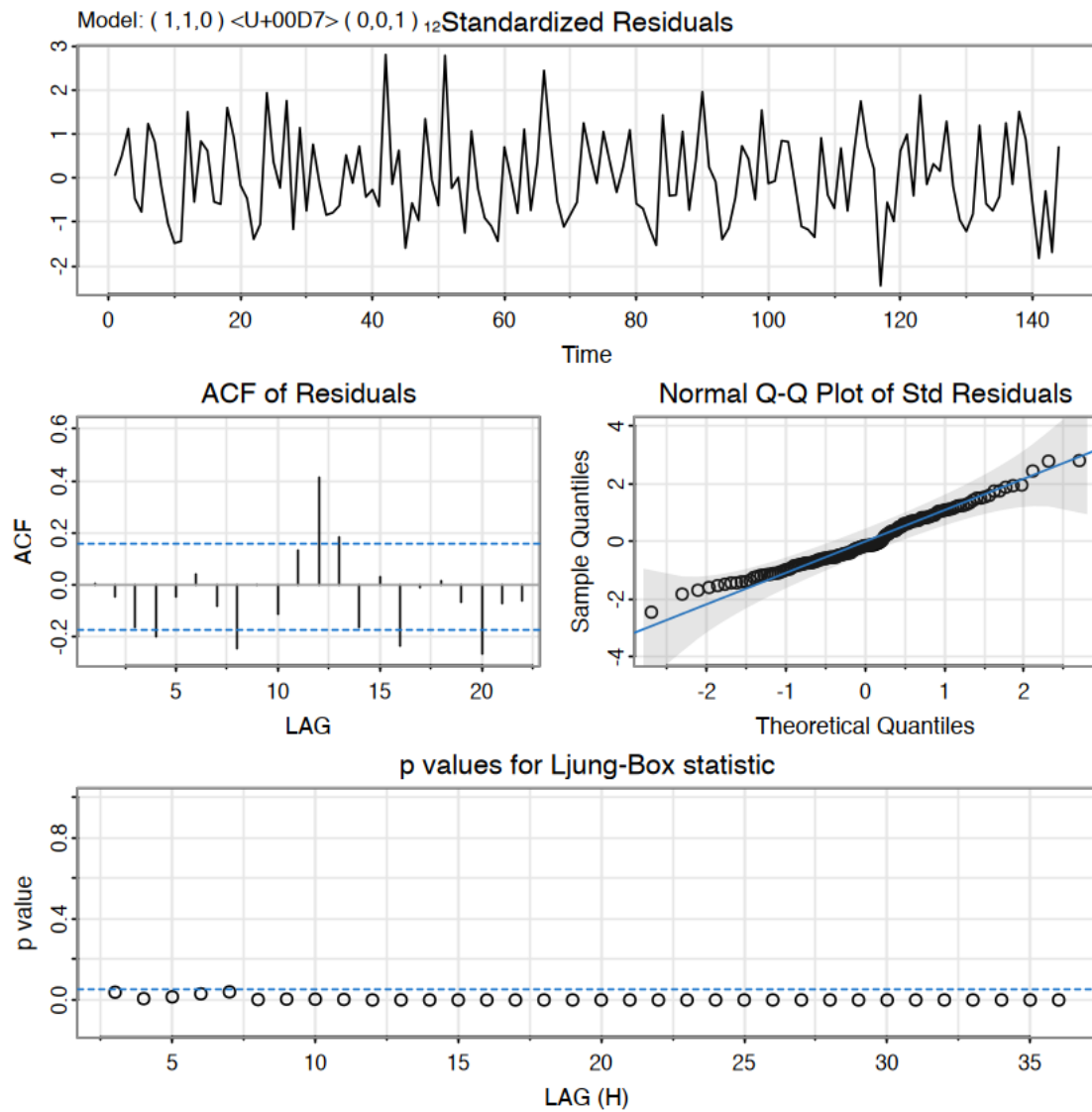
```
[8]: mod1 <- sarima(
  data_ts,
  1, # AR
  0, # I
  0, # MA
  0, # AR_s
  1, # I_s
  0, # MA_s
  12
)
```

```
initial value -2.795206
iter 2 value -3.163852
```

```
[9]: mod2 <- sarima(
  data_ts,
  1, # AR
  1, # I
  0, # MA
  0, # AR_s
  0, # I_s
  1, # MA_s
  12
)
```

```
initial value -2.239650
iter 2 value -2.556407
```

```
[10]: mod1$fit$aic
      mod2$fit$aic
```

```
-452.646059544319
```

```
-331.262213554599
```

```
[11]: n = length(data_ts)
      n
      data_cv <- ts(data_ts[1:120])
      #data_ts

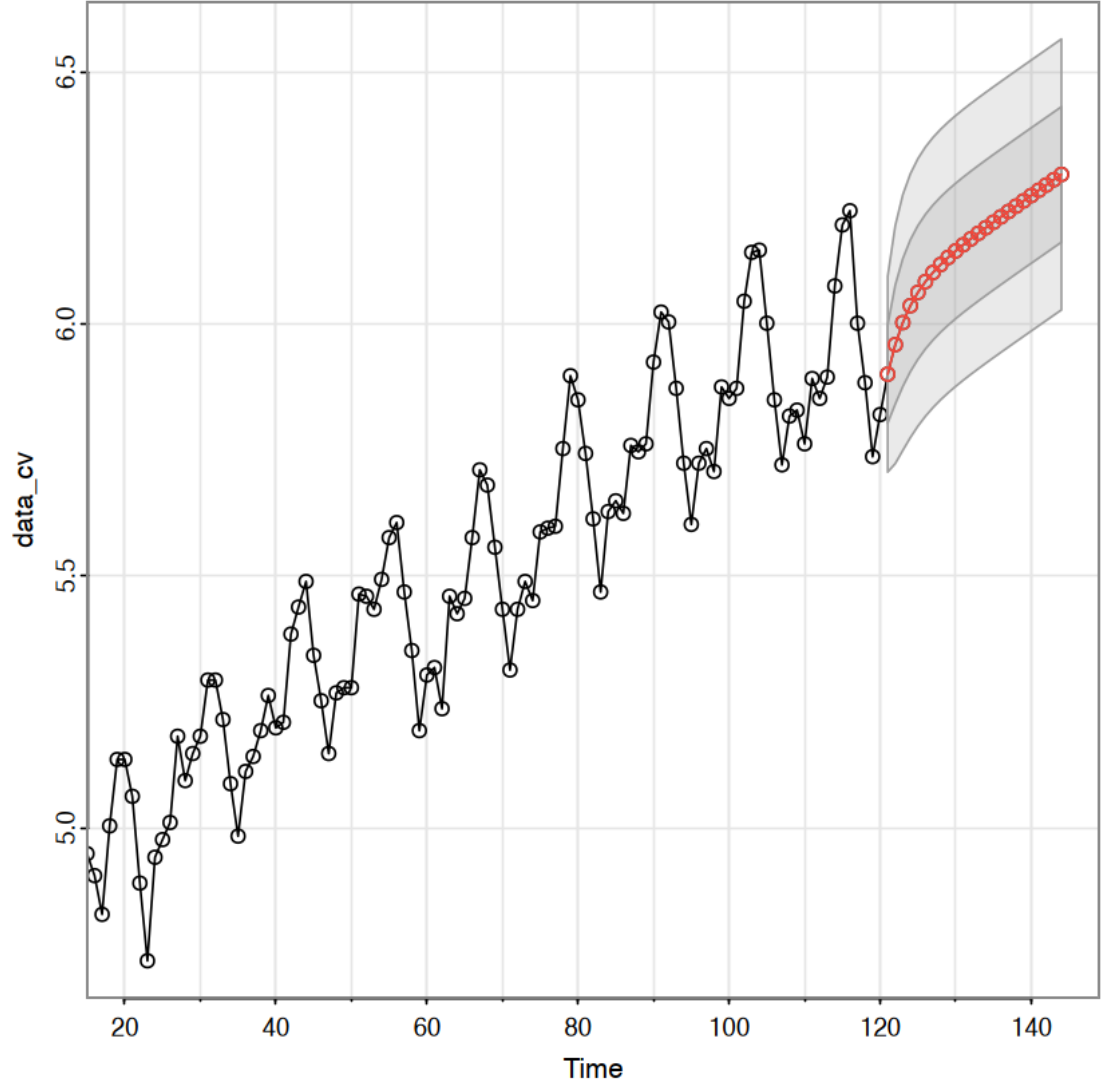
      prev <- mod2 <- sarima.for(
        data_cv,
```

```

xreg = data.frame(time = 1:120),
newxreg = data.frame(time = 121:144),
24,
1, # AR
0, # I
0, # MA
0, # AR_s
0, # I_s
0, # MA_s
12
)

```

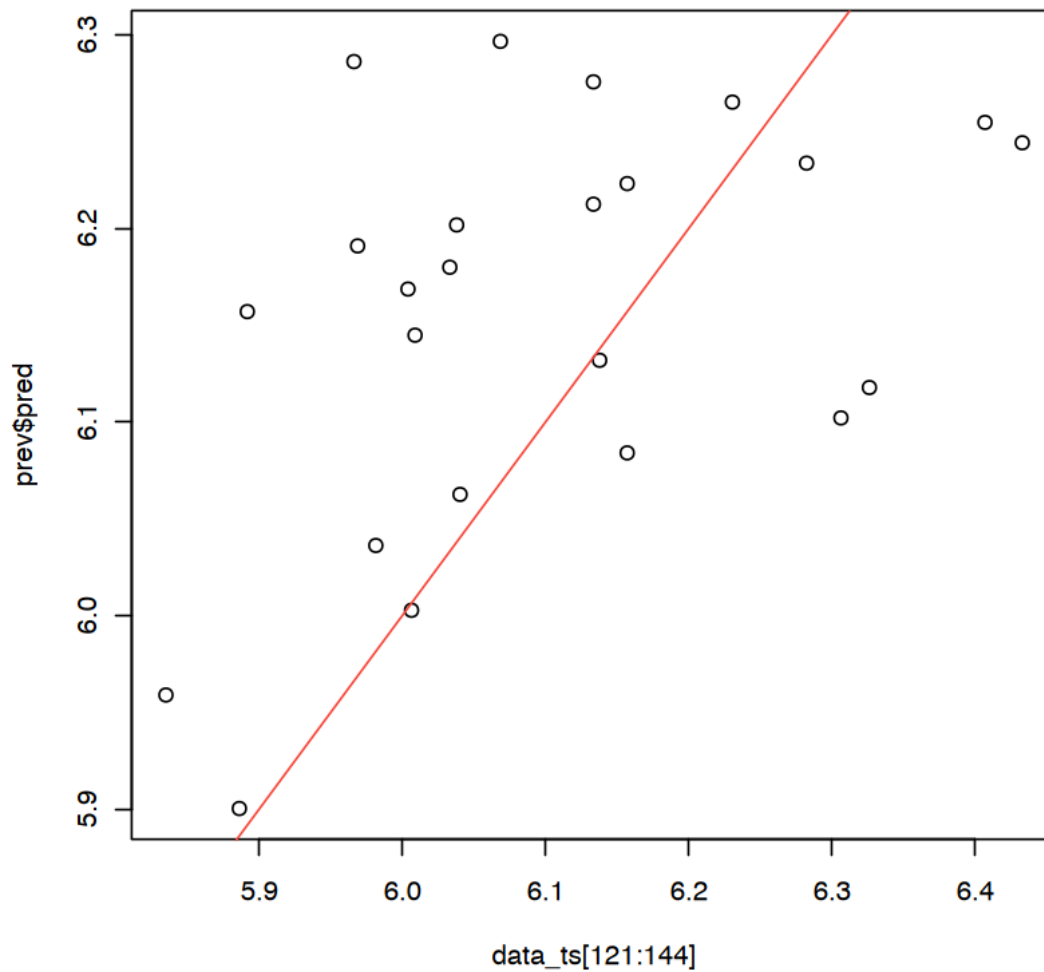
144



```
[12]: plot(data_ts[121:144],prev$pred)
      abline(a=0, b = 1, col=2)

      mean((data_ts[121:144] - prev$pred)^2)
```

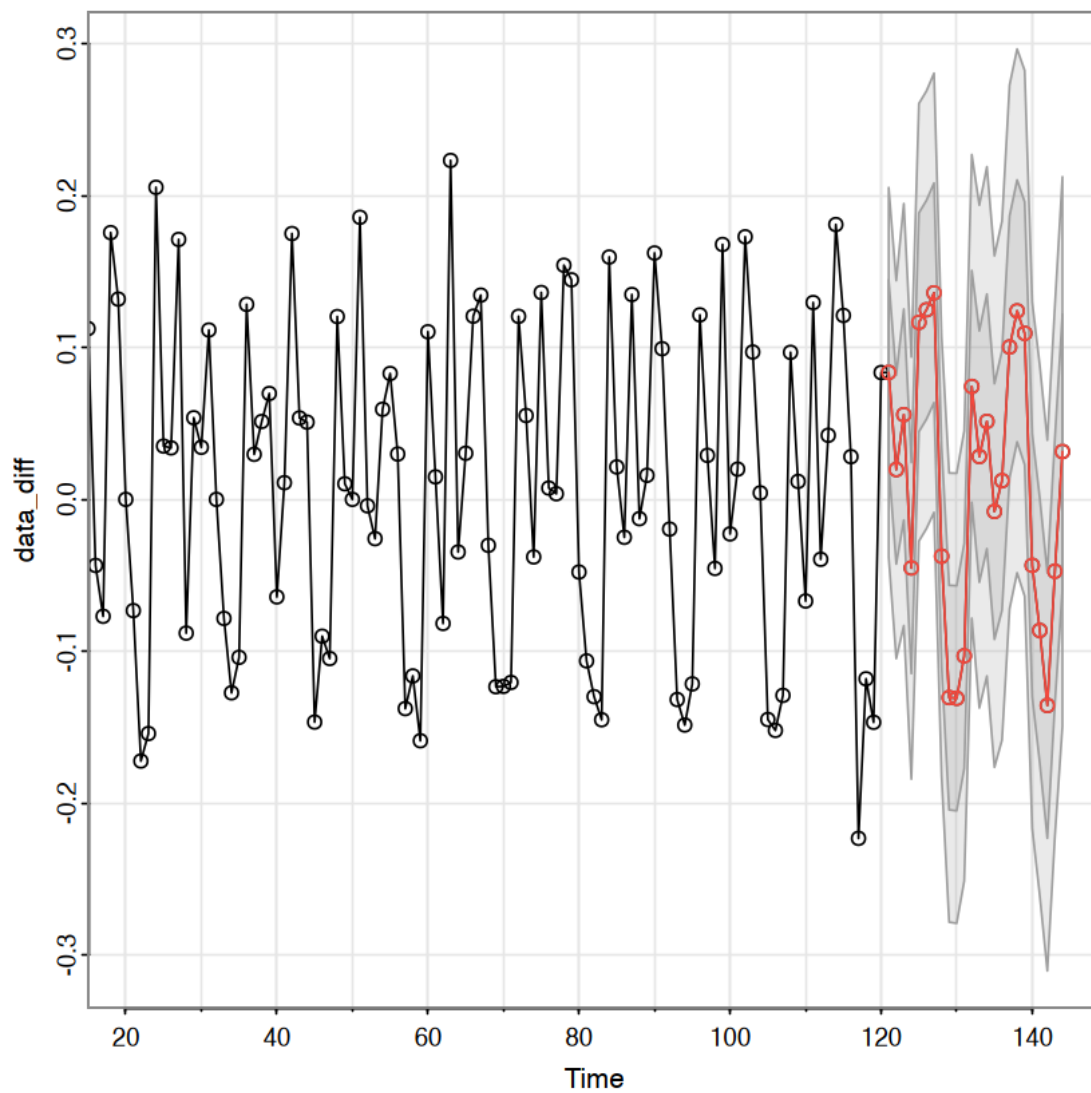
0.0237953299801296



```
[13]: data_diff <- diff(data_cv)
      #plot(data_diff)

      prev <- mod2 <- sarima.for(
        data_diff,
```

```
24,  
10, # AR  
0, # I  
0, # MA  
0, # AR_s  
0, # I_s  
0, # MA_s  
12  
)
```



1 GP

Simuliamo un GP con

$$C() = \sigma^2 \exp(-\phi h)$$

o

$$C() = \sigma^2 \exp(-\frac{h}{\phi})$$

e media 0.

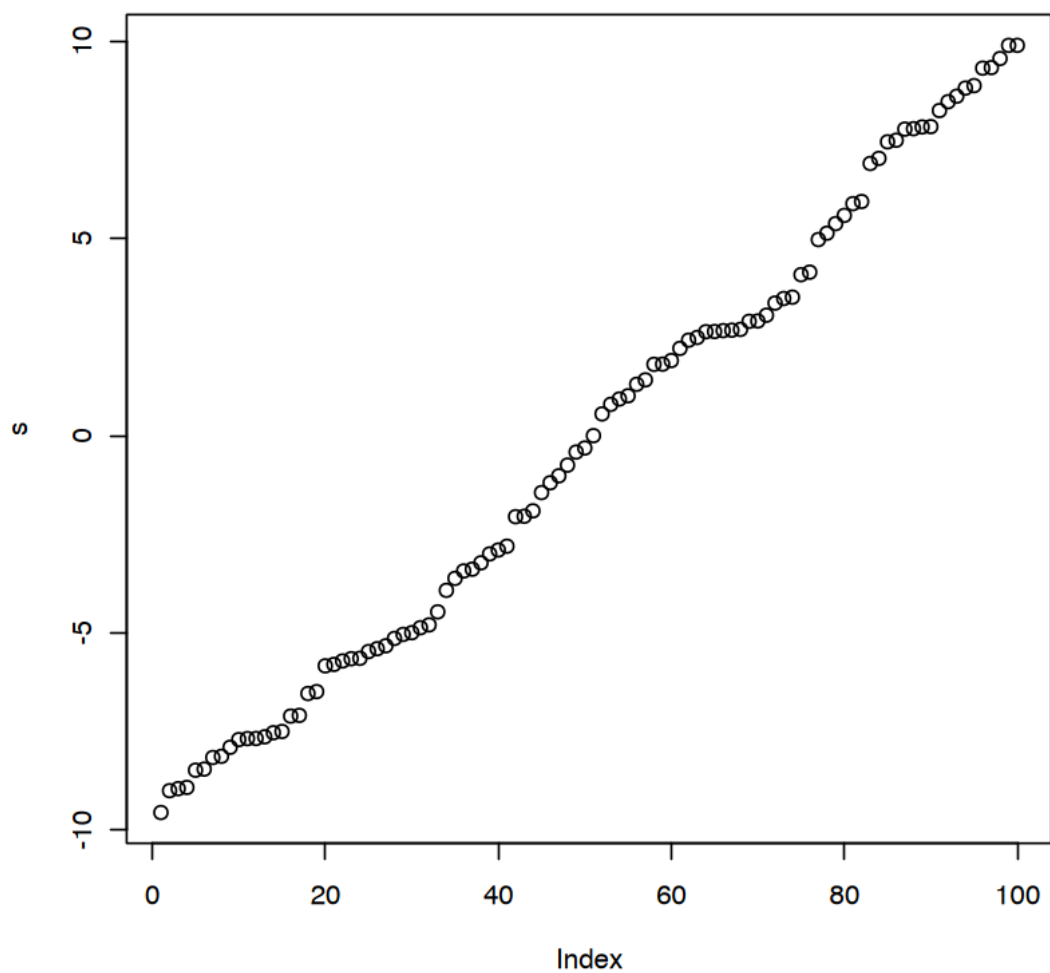
Assumiamo di aver osservato il processo nei punti

$$s_1, \dots, s_n$$

con

$$s_i \sim R$$

```
[14]: n = 100  
s = runif(n, -10,10)  
  
s = s[order(s)]  
plot(s)
```



```
[18]: mean_function = rep(0, n)
Sigma = matrix(NA, nrow=n , ncol= n)
sigma2 = 1
phi = 3/10

for(i in 1:n)
{
  for(j in 1:n)
  {
    Sigma[i, j] = sigma2 * exp(-phi * abs(s[i] - s[j]))
  }
}
```

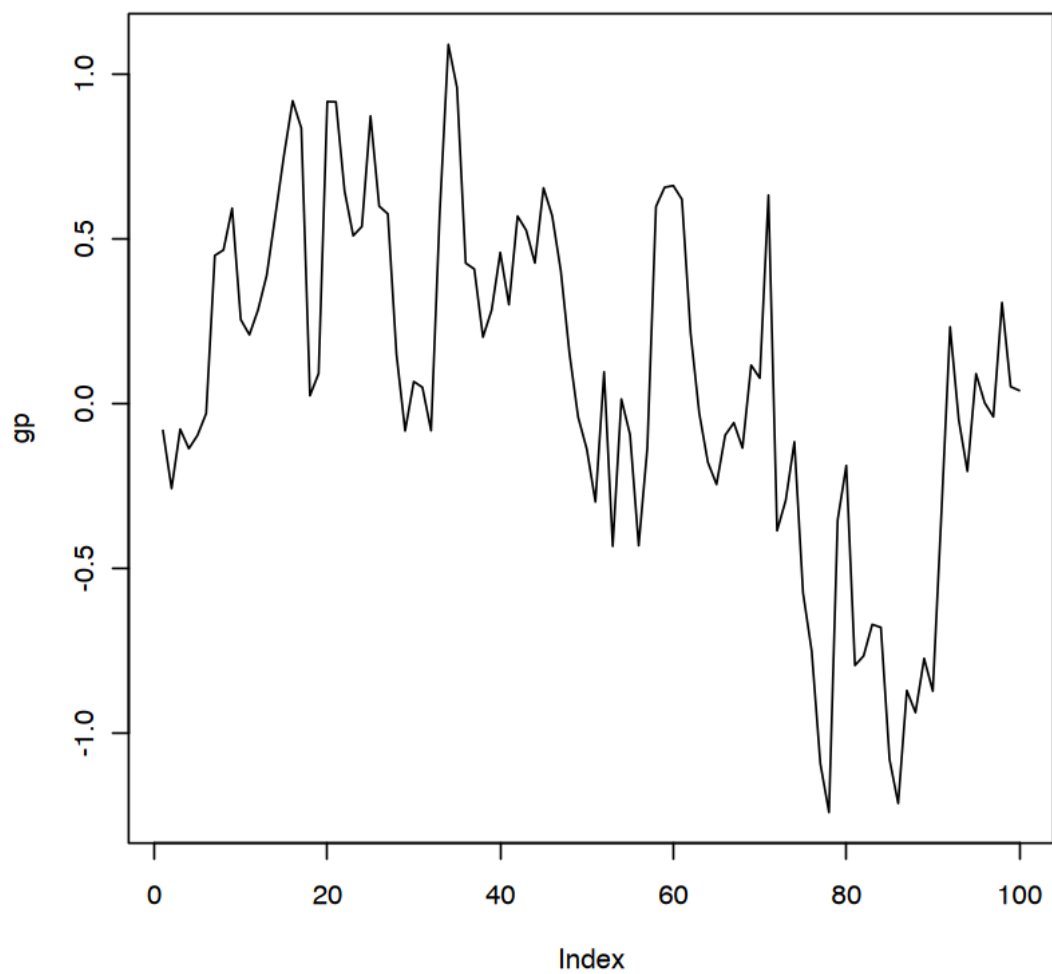
```

Sigma_chol = t(chol(Sigma))

gp <- mean_function + Sigma_chol%% matrix(rnorm(n,0,1), ncol=1)

plot(gp, type = "l")

```



[]:

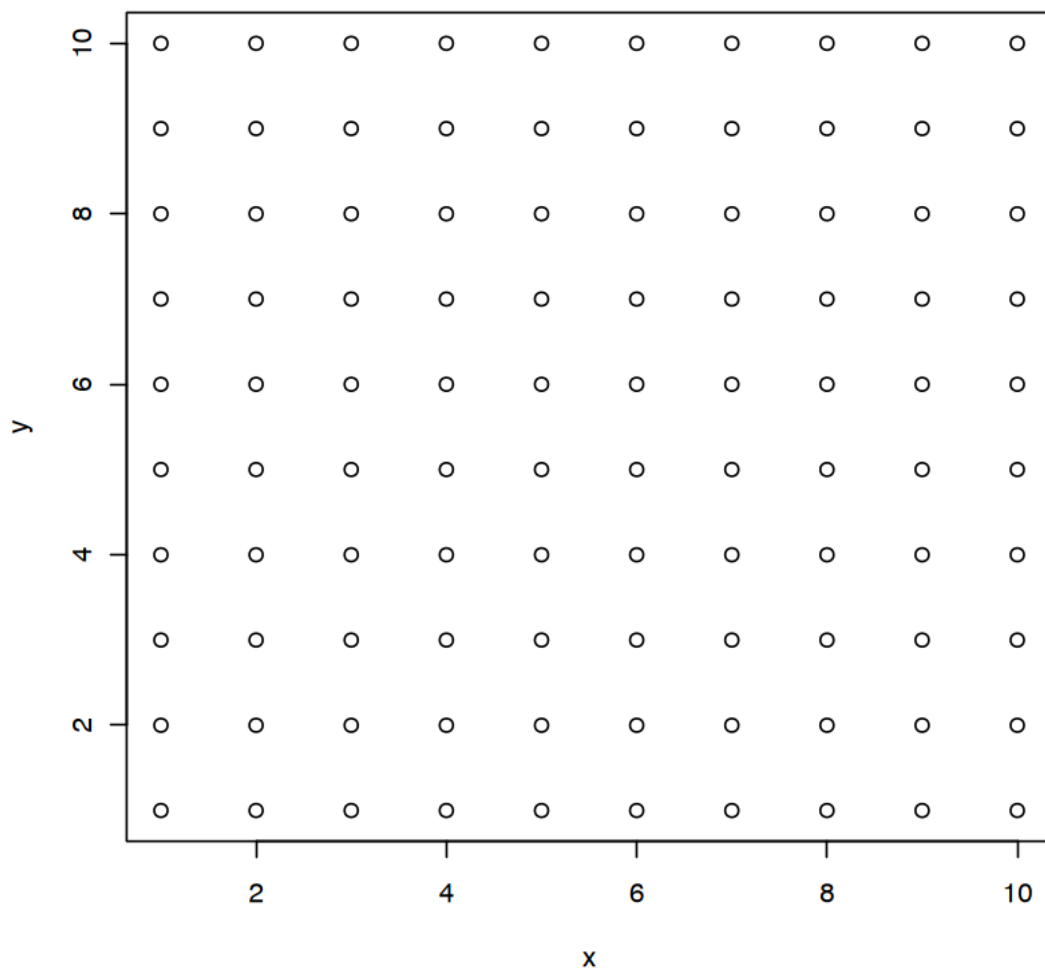
[]:

Simulazione di un processo gaussiano definito su \mathcal{R}^2 con funzione di covarianza esponenziale

```
[19]: n_grid = 10
x_grid = seq(1, n_grid, by = 1)
y_grid = seq(1, n_grid, by = 1)

data_grid = data.frame(x = rep(x_grid, times = n_grid), y = rep(y_grid, each =
  ↪n_grid))

plot(data_grid)
```



```
[21]: n = n_grid^2
mu = rep(0, n)
```



```

sigma2 = 0.2
phi = 3/5
Sigma = matrix(NA, ncol=n, nrow=n)
dist_mat = as.matrix(dist(data_grid))
for(i in 1:n)
{
  for(j in 1:n)
  {
    Sigma[i,j] = sigma2 * exp(- phi * sqrt((data_grid[i, 1] - data_grid[j, 1])^2 + (data_grid[i, 2] - data_grid[j, 2])^2))
  }
}

gp2 <- mu + t(chol(Sigma))%*%matrix(rnorm(n,0,1), ncol=1)

```

```
[24]: image(matrix(gp2, ncol=n_grid))
```

