

Análisis de la serie Daily Female Births

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Información de contacto

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```
## 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:astsa':
```

```
##
```

```
##      gas
```

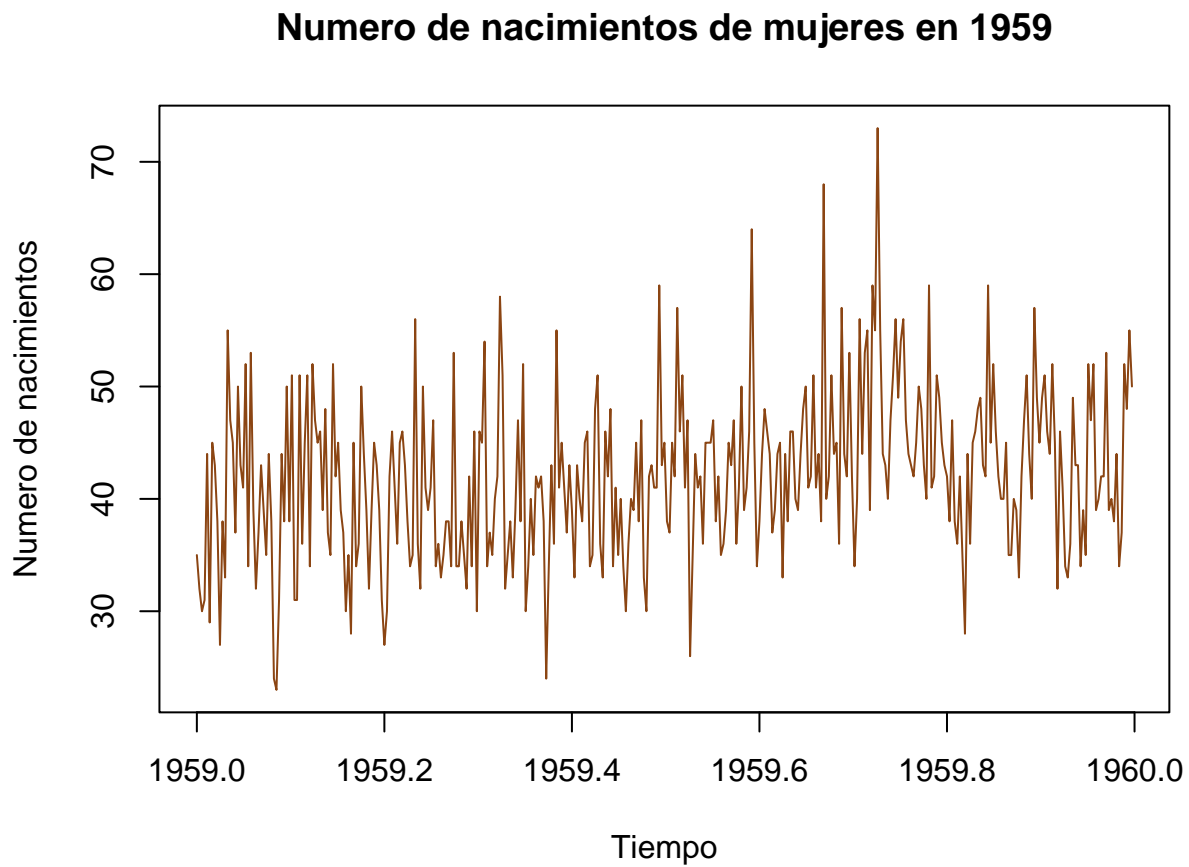
Modelando la serie DFBIC

Descripción

En esta parte se hará un análisis de la serie de tiempo “Daily Female births in California”. Cuya descripción citare

“Un conjunto de datos de series de tiempo que representa el número total de nacimientos de mujeres registrados en California, EE. UU. Durante el año de 1959”

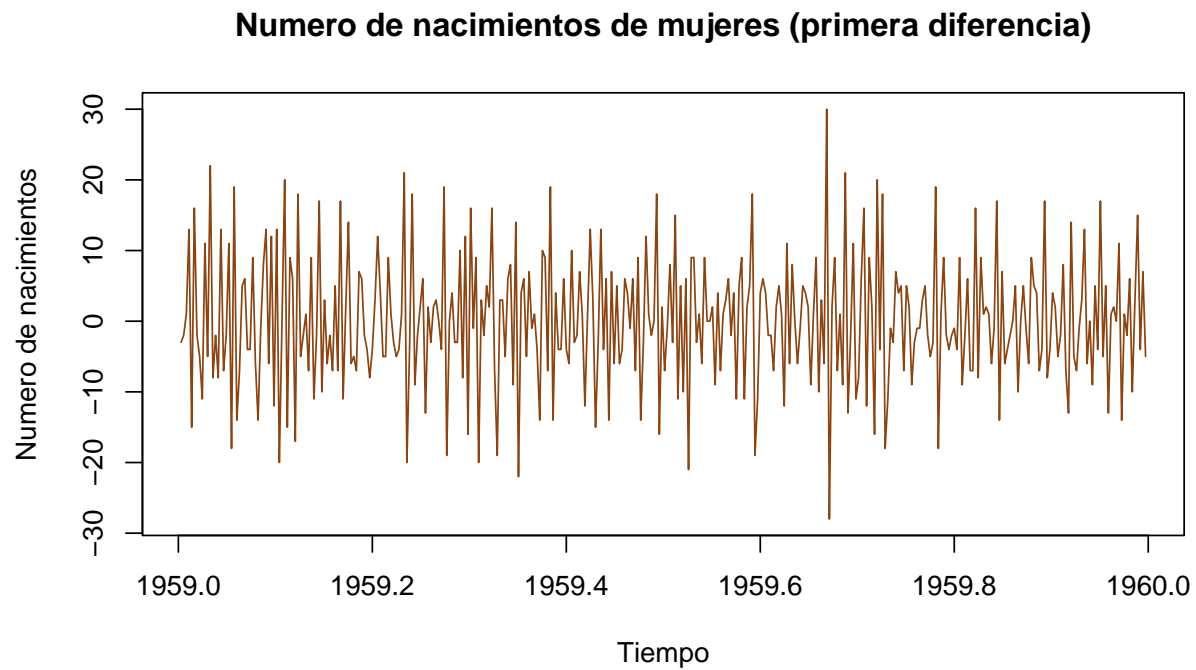
Visualización



Test de Box-Pierce

```
##  
## Box-Pierce test  
##  
## data: birth.ts  
## X-squared = 36.391, df = 5.8999, p-value = 2.088e-06
```

Primera diferencia

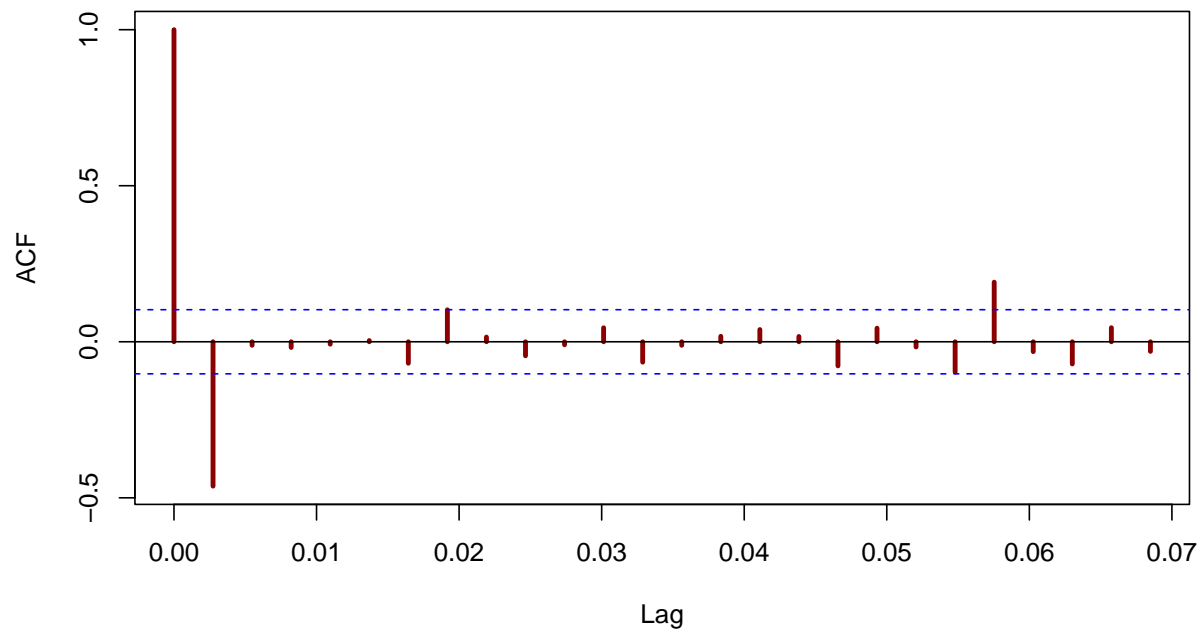


Test de Box-Pierce

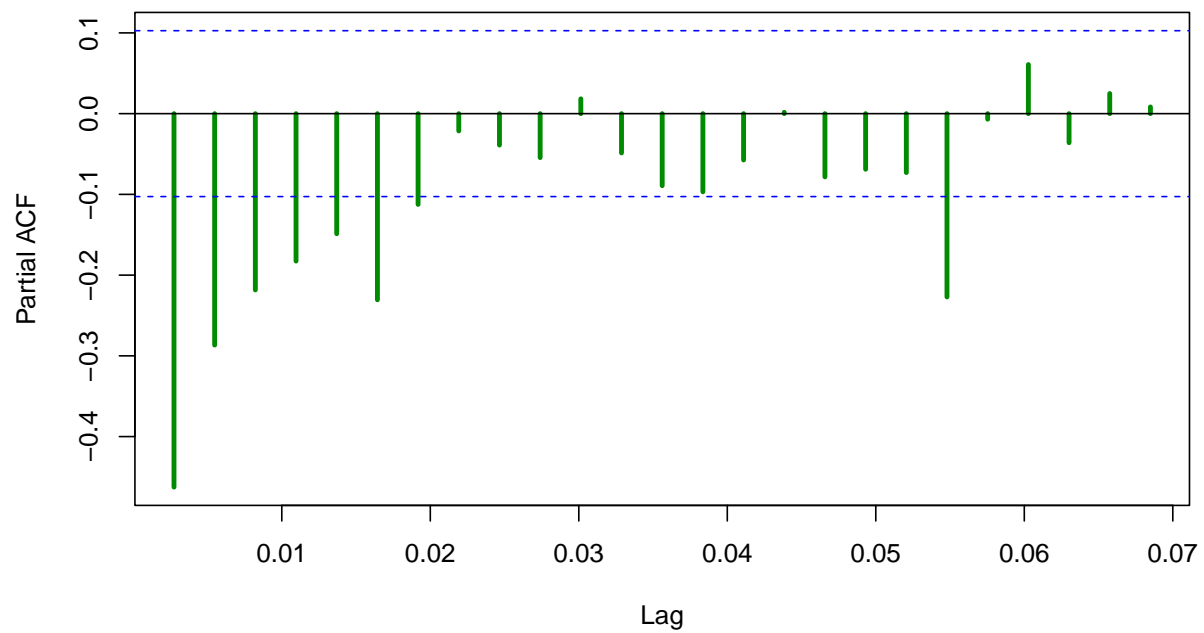
```
##  
## Box-Pierce test  
##  
## data: birth.ts_diff  
## X-squared = 78.094, df = 5.8972, p-value = 7.661e-15
```

ACF y PACF

ACF – Nacimientos de mujeres(primer diferencia)



PACF – Nacimientos de mujeres(primer diferencia)



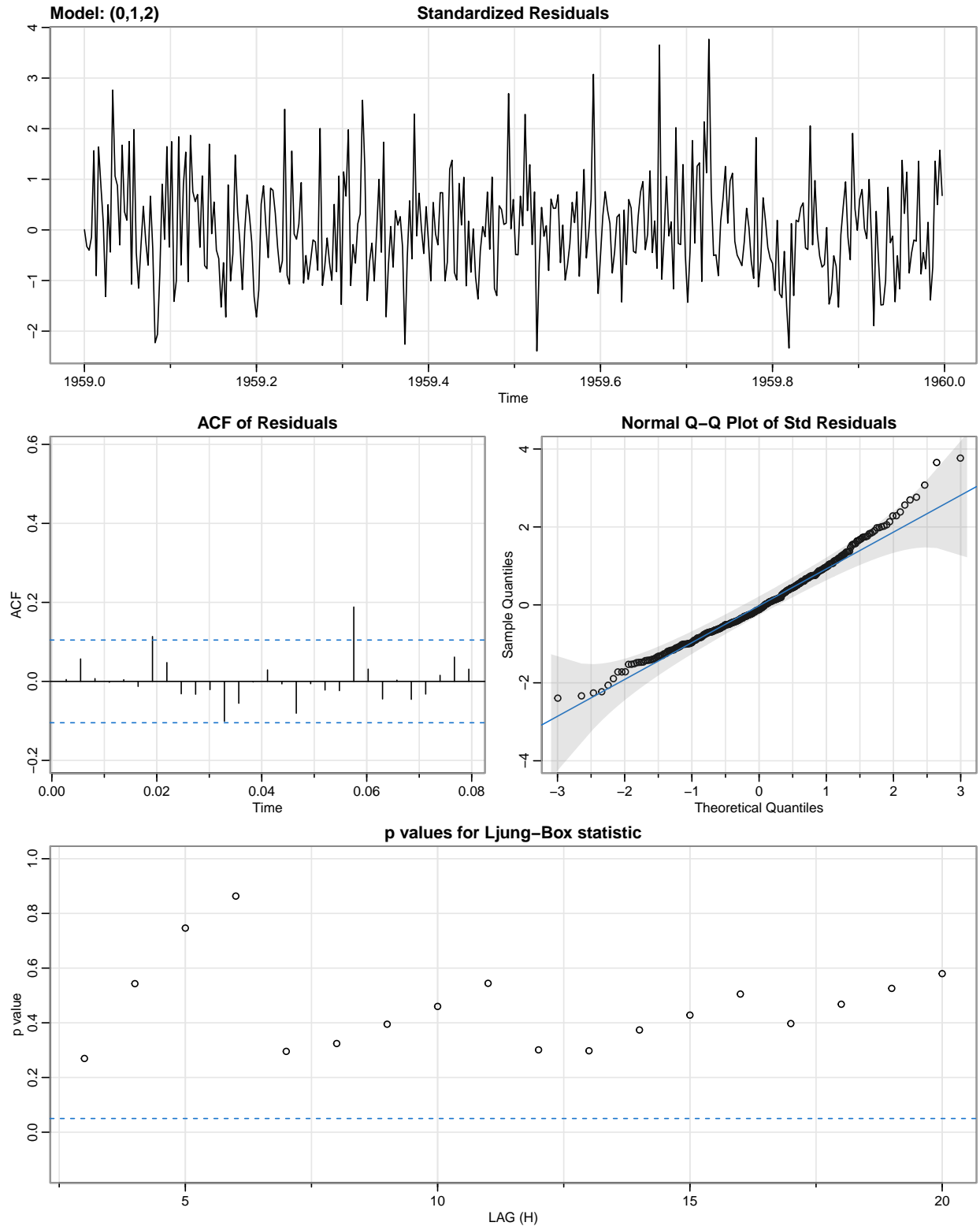
Ajustando modelos

Se proponen diferentes modelos y se llega a:

##	Arima(0,1,1)	Arima(0,1,2)	Arima(7,1,1)	Arima(7,1,2)
## AIC	2462.2207021	2459.5705306	2464.8827225	2466.6664136
## SSE	18148.4561632	17914.6513437	17584.3902548	17574.0578118
## p-value	0.5333604	0.9859227	0.9999899	0.9999929

Siendo el modelo final:

```
## initial value 2.216721
## iter 2 value 2.047518
## iter 3 value 1.974780
## iter 4 value 1.966955
## iter 5 value 1.958906
## iter 6 value 1.952299
## iter 7 value 1.951439
## iter 8 value 1.950801
## iter 9 value 1.950797
## iter 10 value 1.950650
## iter 11 value 1.950646
## iter 12 value 1.950638
## iter 13 value 1.950635
## iter 13 value 1.950635
## iter 13 value 1.950635
## final value 1.950635
## converged
## initial value 1.950708
## iter 2 value 1.950564
## iter 3 value 1.950290
## iter 4 value 1.950196
## iter 5 value 1.950185
## iter 6 value 1.950185
## iter 7 value 1.950185
## iter 7 value 1.950185
## iter 7 value 1.950185
## final value 1.950185
## converged
```



```
## $fit
##
## Call:
## arima(x = xdata, order = c(p, d, q), seasonal = list(order = c(P, D, Q), period = S),
```

```
##      xreg = constant, transform.pars = trans, fixed = fixed, optim.control = list(trace = trc,
##      REPORT = 1, reltol = tol))
##
## Coefficients:
##          ma1          ma2    constant
##      -0.8511  -0.1113     0.015
## s.e.    0.0496    0.0502     0.015
##
## sigma^2 estimated as 49.08:  log likelihood = -1226.36,  aic = 2460.72
##
## $degrees_of_freedom
## [1] 361
##
## $ttable
##      Estimate      SE  t.value p.value
## ma1      -0.8511 0.0496 -17.1448 0.0000
## ma2      -0.1113 0.0502  -2.2164 0.0273
## constant   0.0150 0.0150   1.0007 0.3176
##
## $AIC
## [1] 6.760225
##
## $AICc
## [1] 6.760408
##
## $BIC
## [1] 6.803051
```

Para así finalmente obtener:

$$x_t - x_{t-1} = 0.015 + Z_t - 0.8511Z_{t-1} - 0.1113Z_{t-2}$$

$$Z_t \sim N(0, 49.08)$$