Simulaciones

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En este documento se exponen múltiples simulaciones de la selección automática de variables con sus respectivos retardos usando una nueva propuesta. Se mostrarán ejemplos donde la selección automática trabaja sobre un conjunto de variables de las cuales sólo algunas inciden en la variable respuesta y con un retardo concreto (aunque siempre menor o igual a 0). A lo largo de las siguientes secciones se irán complicando escenarios con la finalidad de analizar cómo se comporta la nueva propuesta ante datos simulados e inferir a partir de ellos cómo se comportará en escenarios reales.

Nota: Para generar los datos de las simulaciones se usó el código arima_simulation.R, el cual permite generar de forma pseudo-aleatoria series temporales a partir de un proceso ARIMA. Este documento no muestra cómo generar las series (para evitar la aleatoriedad de los resultados), sino que, una vez generadas y guardadas, se cargan directamente de memoria.

Simulación de un modelo de regresión dinámica con errores 1 estacionarios

En esta sección veremos cómo se comporta la función de selección automática sobre ejemplos muy básicos donde los errores del modelo son estacionarios:

$$Y_t = \beta_0 + \beta_1 X_{t-r_1}^{(1)} + \beta_2 X_{t-r_2}^{(2)} + \dots + X_{t-r_p}^{(p)} + \eta_t, \qquad \eta_t \sim \mathsf{ARMA(p,q)}, \quad r_i \geq 0 \text{ para } i = 1, \dots, p$$

${\bf Modelo\ donde}\ r_i=0\ {\bf para}\ i=1,...,p$ 1.1

Supongamos un modelo de regresión dinámica de tres variables regresoras donde todos los retardos son igual a cero. En concreto, nuestro modelo tendrá la forma:

$$Y_t = \beta_0 + \beta_1 X_t^{(1)} + \beta_2 X_t^{(2)} + \beta_3 X_t^{(3)} + \eta_t \tag{1}$$

donde:

- $\eta_t \sim \mathsf{ARMA}(2,1)$, por tanto, los errores son estacionarios.

- $\begin{array}{l} \bullet \quad X_t^{(1)} \sim \mathsf{ARIMA}(2,\,1,\,3) \text{ y su coeficiente } \beta_1 = 2.8. \\ \bullet \quad X_t^{(2)} \sim \mathsf{ARIMA}(1,\,1,\,2) \text{ y su coeficiente } \beta_2 = -1.12. \\ \bullet \quad X_t^{(3)} \sim \mathsf{ARMA}(1,\,2) \text{ y su coeficiente } \beta_3 = -2.3. \end{array}$
- El intercept es $\beta_0 = 0.8$.

Supongamos otro conjunto de variables (que siguen también un proceso ARIMA) que no van a influir en la variable respuesta:

- $\begin{array}{l} \bullet \quad X_t^{(4)} \sim \mathsf{ARIMA}(1,\, 0,\, 3). \\ \bullet \quad X_t^{(5)} \sim \mathsf{ARIMA}(2,\, 1,\, 2). \\ \bullet \quad X_t^{(6)} \sim \mathsf{ARIMA}(2,\, 1,\, 1). \end{array}$

```
# Cargamos los datos sobre las variables regresoras
load(file='../simulations/X1 ~ ARIMA(2,1,3).RData')
                                                            # X1
load(file='../simulations/X2 ~ ARIMA(1,1,2).RData')
                                                            # X2
load(file='../simulations/X3 ~ ARIMA(1,0,2).RData')
                                                            # X3
load(file='../simulations/residuals ~ ARIMA(2,0,1).RData') # residuos
# Cargamos las variables independientes
load(file='../simulations/X4 ~ ARIMA(1,0,3).RData')
                                                            # X4
load(file='../simulations/X5 ~ ARIMA(2,1,2).RData')
                                                            # X5
load(file='../simulations/X6 ~ ARIMA(2,1,1).RData')
                                                            # X6
```

Se puede realizar una comprobación de que estas series siguen los procesos ARIMA mencionados. Para chequearlo, consulte el apéndice del documento.

Selección de variables y ajuste del modelo: Creamos el modelo y comprobamos la solución final de la función drm.select().

```
beta0 <- 0.8; beta1 <- 2.8; beta2 <- -1.12; beta3 <- -2.3
Y <- beta0 + beta1 * X1$X + beta2 * X2$X + beta3 * X3$X + residuals$X
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)
ajuste <- drm.select(Y, regresoras, show_info=T)</pre>
Covariate X1 has been tested [ic=-1135.19120552446, lag=0]
Covariate X2 has been tested [ic=-286.986760465024, lag=0]
Covariate X3 has been tested [ic=-640.026014781463, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-185.777061985974, lag=0]
Covariate X1 has been added [aicc=-1135.19120552446, lag=0]
Series: serie
Regression with ARIMA(4,1,1) errors
Coefficients:
        ar1
                ar2
                         ar3
                                 ar4
                                          ma1
                                                 xreg
     0.3013 0.2546 -0.1522 -0.0784 -0.9035 2.7997
s.e. 0.0378 0.0342 0.0335 0.0347
                                       0.0218 0.0485
sigma^2 = 0.0184: log likelihood = 574.65
              AICc=-1135.19
AIC=-1135.31
                             BIC=-1101.03
_____
Covariate X2 has been tested [ic=-1396.16397504914, lag=0]
Covariate X3 has been tested [ic=-2213.72165014961, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
Covariate X3 has been added [aicc=-2213.72165014961, lag=0]
Series: serie
Regression with ARIMA(0,1,3) errors
Coefficients:
         ma1 ma2
                     ma3
                               X1
                                        ХЗ
     -0.5907 0 -0.1460 2.7962 -2.2890
s.e. 0.0274 0 0.0269 0.0350
                                    0.0505
sigma^2 = 0.006201: log likelihood = 1111.89
AIC=-2213.78 AICc=-2213.72
                             BIC=-2189.3
Covariate X2 has been tested [ic=-3163.18780840121, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-2213.72165025041, lag=-3]
Covariate X2 has been added [aicc=-3163.18780840121, lag=0]
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
         ar1
                 ar2
                        ma1 intercept
                                            Х1
                                                     ХЗ
                                                              Х2
```

-0.1989 0.4054 0.4443 0.8056 2.7982 -2.2719 -1.1084

0.0702 0.0302 0.0746 0.0034 0.0092 s.e. 0.0320 0.0108 sigma^2 = 0.002375: log likelihood = 1589.67 AIC=-3163.33 AICc=-3163.19 BIC=-3124.15 Significative correlation with lag<=0 could not be found for X4 Covariate X5 has been tested [ic=-3163.18780355806, lag=-8] Significative correlation with lag<=0 could not be found for X6 No more variables will be added Historical of added covariates to the model (ndiff=0) var lag X1 0 -1135.19120552446 0 -2213.72165014961 ХЗ

Series: serie

Regression with ARIMA(2,0,1) errors

0 -3163.18780840121

Coefficients:

```
X2
    ar1
            ar2
                    ma1
                         intercept
                                        Х1
                                                 ХЗ
-0.1989 0.4054 0.4443
                            0.8056
                                    2.7982
                                            -2.2719
                                                    -1.1084
0.0702 0.0302 0.0746
                            0.0034
                                    0.0092
                                             0.0320
```

```
sigma^2 = 0.002375: log likelihood = 1589.67
AIC=-3163.33 AICc=-3163.19 BIC=-3124.15
```

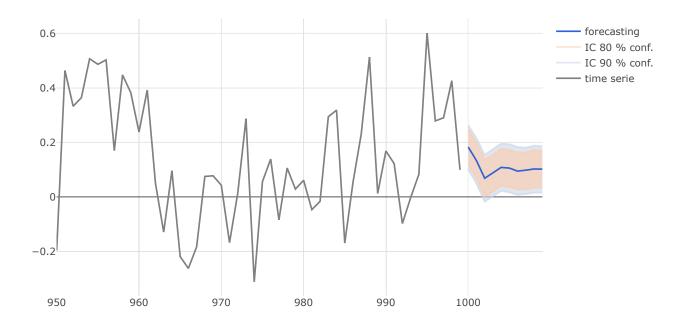
En el *output* de la función vemos cuál ha sido el proceso de selección de variables regresoras:

- 1. En la primera iteración se añade la variable $X_t^{(1)}$ con retardo nulo (algo que es correcto teniendo en cuenta cómo se ha creado el modelo en 1) y obteniendo un AICc=-1135.19120552446.
- 2. En la segunda iteración se introduce la variable $X_t^{(3)}$ con un retardo nulo y mejorando el AICc del modelo anterior (que sólo contaba con la variable $X_t^{(1)}$) con un AICc=-2213.72165014961.
- 3. En la tercera iteración se añade la variable $X_t^{(2)}$ con retardo nulo y mejorando el criterio de información del modelo anterior (que tenía las variables $X_t^{(1)}$ y $X_t^{(1)}$) con un AICc=-3163.18780840121.
- 4. En la siguiente iteración no se encuentran correlaciones significativas con ningún retardo para las variables $X_t^{(4)}$ y $X_t^{(6)}$, por lo que éstas no se pueden añadir al modelo. Para la varialbe $X_t^{(5)}$, se encuentra un retardo significativo en k=-8, pero no se mejora el AICc del modelo anterior (con las tres variables regresoras), por lo que se detiene el bucle para añadir variables.

Como el modelo resultante de haber añadido de forma iterativa todas las variables tiene errores estacionarios (siguen un ARIMA(2,0,1), lo que coincide con la simulación realizada), se considera como modelo válido para definir la relación entre la variable respuesta Y y el conjunto de variables regresoras seleccionadas $(X_t^{(1)}, X_t^{(2)} \text{ y } X_t^{(3)})$. Podemos observar cómo el valor del *intercept* y de los coeficientes de regresión se aproximan bastante a los coeficientes seleccionados al construir de forma artificial la variable regresora Y en Oref(eq:ejemplo1).

Predicción: Finalmente realizamos las predicciones puntuales:

```
preds <- forecast model(Y, regresoras, ajuste, h=10, mode='bootstrap')
display(plot_forecast(preds, rang=c(950, 1009)), name='ejemplo1')
```



Modelo donde $r_i \geq 0$ para i = 1, ..., p1.2

Supongamos un modelo de regresión dinámica parecido al del primer ejemplo, utilizando las mismas variables, pero donde los retardos sean menores o iguales a 0 (que haya "variedad" en los retardos).

$$Y_t = \beta_0 + \beta_1 X_{t-r_1}^{(1)} + \beta_2 X_{t-r_2}^{(2)} + \beta_3 X_{t-r_3}^{(3)} + \eta_t$$
 (2)

donde:

- $\eta_t \sim \mathsf{ARMA}(\mathsf{2,\,1}).$
- $\begin{array}{l} \text{ } & X_t^{(1)} \sim \text{ARIMA(2, 1)}. \\ \text{ } & X_t^{(1)} \sim \text{ARIMA(2, 1, 3)} \text{ y su retardo } r_1 = 2. \\ \text{ } & X_t^{(2)} \sim \text{ARMA(1, 1, 2)} \text{ y su retardo } r_2 = 0. \\ \text{ } & X_t^{(3)} \sim \text{ARMA(1, 0, 2)} \text{ y su retardo } r_3 = 3. \end{array}$

```
# Construimos el modelo
beta0 <- -0.6; beta1 <- 1.7; beta2 <- -2.2; beta3 <- 1.3
r1 <- 2; r3 <- 3
Y \leftarrow beta0 + beta1 * lag(X1$X, -r1) + beta2 * X2$X + beta3 * lag(X3$X, -r3) +
    residuals$X
```

Selección de variables y ajuste del modelo:

```
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)
ajuste <- drm.select(Y, regresoras, show_info=T, st_method='adf.test')</pre>
```

Covariate X1 has been tested [ic=-989.555684098221, lag=-2]

Covariate X2 has been tested [ic=-1156.68486061937, lag=0] Covariate X3 has been tested [ic=-802.904991565362, lag=-3] Covariate X4 has been tested [ic=-580.329461807603, lag=-1] Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 Covariate X2 has been added [aicc=-1156.68486061937, lag=0] Series: serie

Regression with ARIMA(4,1,0) errors

Coefficients:

ar1 ar2 ar3 ar4 xreg -0.4245 -0.3382 -0.1395 0.1144 -2.2607 s.e. 0.0319 0.0344 0.0344 0.0320 0.0789

sigma^2 = 0.01767: log likelihood = 584.39 AIC=-1156.77 AICc=-1156.68 BIC=-1127.5

Covariate X1 has been tested [ic=-2171.66958134745, lag=-2] Covariate X3 has been tested [ic=-1561.28469740141, lag=-3] Significative correlation with lag<=0 could not be found for X4 Covariate X5 has been tested [ic=-1156.6848606207, lag=-21] Significative correlation with lag<=0 could not be found for X6 Covariate X1 has been added [aicc=-2171.66958134745, lag=-2] Series: serie

Regression with ARIMA(3,0,0) errors

Coefficients:

ar3 intercept X2 ar1 ar2 0.3022 0.3526 -0.1198 -0.6039 -2.2393 1.7268 s.e. 0.0318 0.0313 0.0319 0.0065 0.0200 0.0171

sigma^2 = 0.006229: log likelihood = 1092.89 AIC=-2171.79 AICc=-2171.67 BIC=-2137.62

Covariate X3 has been tested [ic=-3108.15443209894, lag=-3] Covariate X4 has been tested [ic=-2171.66942313677, lag=-15] Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 Covariate X3 has been added [aicc=-3108.15443209894, lag=-3] Series: serie

Regression with ARIMA(0,0,4) errors

Coefficients:

ma2 ma3 ma4 intercept X2 X1 ХЗ ma1 s.e. 0.0304 0.0302 0 0.0300 0.0033 0.0105 0.0089 0.0320

sigma^2 = 0.002377: log likelihood = 1562.15 AIC=-3108.3 AICc=-3108.15 BIC=-3069.26

Significative correlation with lag<=0 could not be found for X4 Covariate X5 has been tested [ic=-3108.15429253356, lag=-8] Significative correlation with lag<=0 could not be found for X6 No more variables will be added

```
| Historical of added covariates to the model (ndiff=0) |
var lag ic
X2 0 -1156.68486061937
X1 -2 -2171.66958134745
X3 -3 -3108.15443209894
```

Series: serie

Regression with ARIMA(0,0,4) errors

Coefficients:

```
ma1
                ma2
                     ma3
                             ma4
                                   intercept
      0.2498 0.3360
                        0
                          0.1589
                                     -0.5947
                                              -2.1868
                                                       1.6949
                                                               1.3083
     0.0304
            0.0302
                          0.0300
                                                              0.0320
                                      0.0033
                                               0.0105
                                                      0.0089
sigma^2 = 0.002377: log likelihood = 1562.15
AIC=-3108.3
              AICc=-3108.15
                             BIC=-3069.26
```

De nuevo, observando el output de la función podemos analizar la selección de variables:

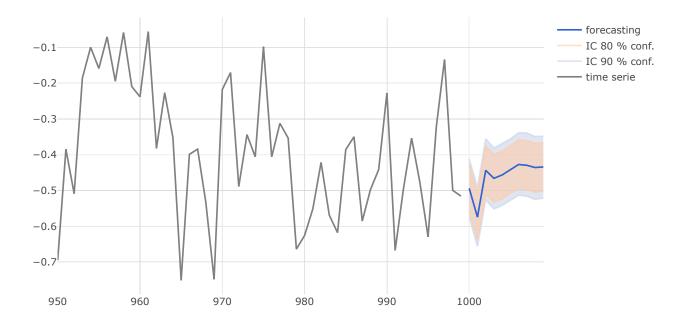
- 1. Primero se selecciona la variable $X_t^{(2)}$ con retardo nulo, construyendo un modelo de regresión con errores que siguen un ARIMA(4,1,0) y un AICc=-1156.68486061937.
- 2. En la siguiente iteración se selecciona la variable $X_t^{(1)}$ con un retardo $r_1=2$, mejorando el AICc del modelo anterior con un nuevo AICc=-2171.66958134745.
- 3. En la siguiente iteración se añade la variable $X_t^{(3)}$ con retardo $r_3=3$ al construir un modelo de regresión con variables regresoras $X_t^{(1)}$, $X_t^{(2)}$ y $X_t^{(3)}$ con respectivos retardos $r_1=2$, $r_2=0$ y $r_3=3$, consiguiendo un AICc=-3108.15443209894.
- 4. En la siguiente iteración no se encuentran retardos significativos para $X_t^{(4)}$ y $X_t^{(6)}$, pero sí para $X_t^{(5)}$ con un retardo $r_5=8$.. No obstante, añadir esta variable al modelo no supone mejorar el AICc del modelo anterior (sólo se consigue un AICc=-3106.1490878263). Por tanto se detiene la selección de varaibles.

Como el mejor modelo conseguido (el de la tercera iteración) ya tiene errores estacionarios (siguen un ARIMA(0,0,4)), se escoge dicho ajuste para modelizar la dependencia entre Y y las variables regresoras escogidas.

Podemos observar que los valores de los coeficientes de regresión y del *intercept* se aproximan bien a los valores verdaderos que se seleccionaron en la construcción del modelo. No obstante, los errores son modelizables con un ARIMA(0,0,4), no con un ARIMA(2,0,1).

Predicción:

```
# Podemos mostrar las predicciones puntuales
preds <- forecast_model(Y, regresoras, ajuste, h=10, mode='bootstrap')
display(plot_forecast(preds, rang=c(950, 1009)), name='ejemplo2')</pre>
```



Simulación de un modelo de regresión dinámico con errores 2 ARIMA ($d \ge 1$)

En esta sección consideraremos modelos de regresión dinámica donde las innovaciones no son estacionarias:

$$Y_t = \beta_0 + \beta_1 X_{t-r_1}^{(1)} + \beta_2 X_{t-r_2}^{(2)} + \dots + X_{t-r_p}^{(p)} + \eta_t, \qquad \eta_t \sim \mathsf{ARIMA}(\mathsf{p,d,q})$$

${\bf Modelo\ donde}\ r_i=0\ {\bf para}\ i=1,...,p$ 2.1

Tomemos el mismo modelo que en el primer ejemplo pero con errores no estacionarios:

$$Y_t = \beta_0 + \beta_1 X_t^{(1)} + \beta_2 X_t^{(2)} + \beta_3 X_t^{(3)} + \eta_t, \qquad \eta_t \sim \text{ARIMA(1,2,2)}$$
 (3)

donde el conjunto de variables $\mathcal X$ sobre el que se realiza la selección está compuesto por las variables que sí influyen en Y:

- $\begin{array}{l} \bullet \quad X_t^{(1)} \sim \mathsf{ARIMA}(\mathsf{2,\,1,\,3}) \; \mathsf{y} \; \mathsf{su} \; \mathsf{coeficiente} \; \beta_1 = -1.3. \\ \bullet \quad X_t^{(2)} \sim \mathsf{ARIMA}(\mathsf{1,\,1,\,2}) \; \mathsf{y} \; \mathsf{su} \; \mathsf{coeficiente} \; \beta_2 = 2.12. \\ \bullet \quad X_t^{(3)} \sim \mathsf{ARMA}(\mathsf{1,\,2}) \; \mathsf{y} \; \mathsf{su} \; \mathsf{coeficiente} \; \beta_3 = 2.3. \end{array}$

- El intercept es $\beta_0 = 0.8$.

Y las variables que no interfieren en Y (las mismas que en el primer ejemplo).

cargamos únicamente los residuos no estacionarios load('../simulations/residuals ~ ARIMA(1,2,2).RData')

```
# volvemos a generar la variable respuesta
beta0 <- 0.8; beta1 <- -1.3; beta2 <- 2.12; beta3 <- 2.3
Y <- beta0 + beta1 * X1$X + beta2 * X2$X + beta3 * X3$X + 2.1*residuals$X
Selección de variables y ajuste del modelo: Ajustamos el modelo con las variables originales (no
diferenciamos ninguna):
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)
ajuste <- drm.select(Y, regresoras, show_info=T)</pre>
Covariate X1 has been tested [ic=59.5855339744035, lag=0]
Covariate X2 has been tested [ic=-181.525975728858, lag=0]
Covariate X3 has been tested [ic=-202.902218914428, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=199.408749112135, lag=-5]
Covariate X3 has been added [aicc=-202.902218914428, lag=0]
Series: serie
Regression with ARIMA(3,1,2) errors
Coefficients:
                  ar2
                          ar3
                                   ma1
                                            ma2
                                                   xreg
      1.0348 -0.3048 0.2549 -0.7430 0.3264 2.3714
s.e. 0.1407 0.1250 0.0532
                               0.1408 0.0863 0.0980
sigma^2 = 0.04716: log likelihood = 108.51
AIC=-203.02
            AICc=-202.9
                            BIC=-168.74
Covariate X1 has been tested [ic=-417.422002505725, lag=0]
Covariate X2 has been tested [ic=-884.596379075807, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-204.936549154963, lag=-5]
Covariate X2 has been added [aicc=-884.596379075807, lag=0]
Series: serie
Regression with ARIMA(4,1,1) errors
Coefficients:
      ar1
                      ar3
                              ar4
                                                ХЗ
                                                        X2
              ar2
                                      ma1
        0 \quad 0.4720 \quad 0.2733 \quad 0.2332 \quad 0.4715 \quad 2.3610 \quad 2.1784
        0 0.0323 0.0245 0.0299 0.0309 0.0658 0.0666
sigma^2 = 0.02366: log likelihood = 449.36
              AICc=-884.6
AIC=-884.71
                            BIC=-850.43
Covariate X1 has been tested [ic=-1613.80360585469, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
Covariate X1 has been added [aicc=-1613.80360585469, lag=0]
Series: serie
Regression with ARIMA(2,1,1) errors
```

```
Coefficients:
```

ХЗ Х2 ar1 ar2 ma1X 1 0.2266 0.7606 0.3070 2.3763 2.1588 -1.2784 s.e. 0.0299 0.0295 0.0399 0.0447 0.0391 0.0336

sigma^2 = 0.01131: log likelihood = 813.96 AIC=-1613.92 AICc=-1613.8 BIC=-1579.64

Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 No more variables will be added

The global model does not have stationary errors Trying to adjust a model that do have stationary errors No valid model with stationary errors could be optimized

Applying regular differentiation (ndiff=1) and calling again the function

_____ ______

Covariate X1 has been tested [ic=61.2589226747805, lag=0]

Covariate X2 has been tested [ic=-179.890806223957, lag=0]

Covariate X3 has been tested [ic=-201.85732451794, lag=0]

Significative correlation with lag<=0 could not be found for X4

Significative correlation with lag<=0 could not be found for X5

Covariate X6 has been tested [ic=201.298523732159, lag=-5]

Covariate X3 has been added [aicc=-201.85732451794, lag=0]

Series: serie

Regression with ARIMA(3,1,0) errors

Coefficients:

ar1 ar2 ar3 xreg -0.7053 -0.4727 -0.1673 2.3747 s.e. 0.0314 0.0355 0.0314 0.0993

sigma^2 = 0.04741: log likelihood = 105.96 AIC=-201.92 AICc=-201.86 BIC=-177.44

Covariate X1 has been tested [ic=-415.897154793375, lag=0] Covariate X2 has been tested [ic=-624.559112508249, lag=0] Significative correlation with lag<=0 could not be found for X4

Significative correlation with lag<=0 could not be found for X5

Covariate X6 has been tested [ic=-206.141817506691, lag=-5]

Covariate X2 has been added [aicc=-624.559112508249, lag=0]

Series: serie

Regression with ARIMA(0,1,0) errors

Coefficients:

ХЗ 2.3654 2.1730 s.e. 0.0557 0.0599

sigma^2 = 0.03099: log likelihood = 315.29 AIC=-624.58 AICc=-624.56 BIC=-609.9

Covariate X1 has been tested [ic=-1209.59028329543, lag=0]

```
Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 Covariate X1 has been added [aicc=-1209.59028329543, lag=0]
```

Series: serie

Regression with ARIMA(0,1,0) errors

Coefficients:

X3 X2 X1 2.3776 2.1642 -1.2493 s.e. 0.0414 0.0445 0.0441

sigma^2 = 0.01712: log likelihood = 608.82 AIC=-1209.63 AICc=-1209.59 BIC=-1190.05

Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 No more variables will be added

The global model does not have stationary errors $% \left(1\right) =\left(1\right) +\left(1\right) +\left$

Trying to adjust a model that do have stationary errors

Historical of added covariates to the model (ndiff=1)

var lag ic

X3 0 -201.85732451794 X2 0 -624.559112508249 X1 0 -1209.59028329543

Series: serie

Regression with ARIMA(2,0,1) errors

Coefficients:

ar1 ar2 ma1 X3 X2 X1 0.2265 0.7606 0.3071 2.3763 2.1588 -1.2784 s.e. 0.0299 0.0295 0.0399 0.0447 0.0391 0.0336

sigma^2 = 0.01131: log likelihood = 813.96 AIC=-1613.92 AICc=-1613.8 BIC=-1579.64

Si volvemos a analizar el *output* de la consola observamos que se ha realizado una diferenciación regular a los datos (dobles líneas horizontales) para conseguir un ajuste en el que los errores fuesen estacionarios.

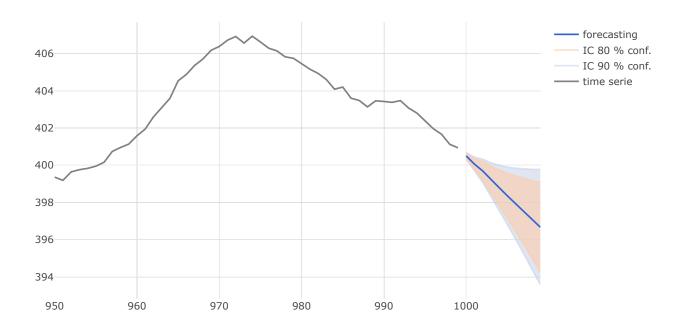
1. En las primeras iteraciones se añaden, en este orden, las variables $X_t^{(3)}$, $X_t^{(2)}$ y $X_t^{(1)}$ con retardos nulos, mejorando el AICc iterativamente hasta alcanzar un valor de AICc=-1613.80360585469. Como el modelo ajustado con estas tres variables regresoras no tiene errores estacionarios (se ajustó un ARIMA(2,1,1) para los residuos), se intenta ajustar un modelo donde el orden de d sea nulo. Como no se puede optimizar ("No se ha podido encontrar un modelo válido con errores estacionarios..."), se procede a aplicar una diferenciación regular a todos los datos, tanto variable respuesta como conjunto de posibles variables regresoras, y se vuelve a llamar a la función drm.select() con los datos diferenciados.

2. En la siguiente llamada a la función se consigue añadir al modelo, en este orden, las variables $X_t^{(3)}$, $X_t^{(2)}$ y $X_t^{(1)}$ con retardos nulos, y se ajusta un ARIMA(0,1,0) para los errores de regresión. No obstante, como este ajuste no cumple la condición de errores estacionarios, se intenta ajustar un ARIMA para los errores donde d=0. En este caso sí se consigue optimizar un modelo que respeta dicha condición, obteniendo un ajuste donde los residuos son estacionarios y el AICc=-1613.8.

Obsérvese que habiendo diferenciado los datos, se ha conseguido seleccionar las 3 variables regresoras que realmente tienen una influencia en la construcción de la variable respuesta Y con los retardos correctos.

Predicción: Una vez obtenido el modelo, se pueden realizar predicciones puntuales. Cuando se detecta que el ajuste se corresponde a un ajuste de los datos diferenciados, la función forecast model() lanza un aviso de que se utilizarán los datos en unidades originales para realizar las predicciones.

```
preds <- forecast_model(Y, regresoras, ajuste, h=10, mode='bootstrap')</pre>
Returning predictions in original scale...
display(plot_forecast(preds, rang=c(950, 1009)), name='ejemplo3')
```



Modelo donde $r_i \geq 0$ para i = 1, ..., p

Podemos alterar el ejemplo anterior para que las variables regresoras influyan en Y con cierto retardo.

- $\begin{tabular}{ll} \blacksquare & \mbox{La variable } X_t^{(1)} \mbox{ se introduce con retardo } r_1=2. \\ \blacksquare & \mbox{La variable } X_t^{(3)} \mbox{ se introduce con retardo } r_3=1. \\ \end{tabular}$

```
beta0 <- 0.8; beta1 <- -1.3; beta2 <- 2.12; beta3 <- 2.3
r1 <- 2; r3 <- 1
Y \leftarrow beta0 + beta1 * lag(X1$X, -r1) + beta2 * X2$X +
    beta3 * lag(X3$X, -r3) + 1.5*residuals$X
```

Selección de variables y ajuste del modelo:

```
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)
ajuste <- drm.select(Y, regresoras, show_info=T, st_method='adf.test')</pre>
Covariate X1 has been tested [ic=-128.730307266973, lag=-2]
Covariate X2 has been tested [ic=-353.706447261812, lag=0]
Covariate X3 has been tested [ic=-435.193689069136, lag=-1]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
Covariate X3 has been added [aicc=-435.193689069136, lag=-1]
Series: serie
Regression with ARIMA(3,1,2) errors
Coefficients:
        ar1 ar2
                     ar3
                              ma1
                                     ma2
                                            xreg
              0 0.2839 -0.4997
                                   0.0790
                                          2.3230
     0.6985
s.e. 0.0440
               0 0.0422
                           0.0558
                                  0.0326 0.0903
sigma^2 = 0.03729: log likelihood = 223.64
             AICc=-435.19
AIC=-435.28
                            BIC=-405.92
Covariate X1 has been tested [ic=-653.565041178663, lag=-2]
Covariate X2 has been tested [ic=-1254.38354028493, lag=0]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-438.084375852223, lag=-9]
Covariate X2 has been added [aicc=-1254.38354028493, lag=0]
Series: serie
Regression with ARIMA(3,1,2) errors
Coefficients:
        ar1
                 ar2
                         ar3
                                 ma1
                                         ma2
                                                  ХЗ
                                                          X2
     1.0735 -0.5681 0.4771 -0.6904 0.5167 2.2884 2.1250
s.e. 0.0699 0.0736 0.0450 0.0701 0.0504 0.0530 0.0562
sigma^2 = 0.0162: log likelihood = 635.27
             AICc=-1254.38
                             BIC=-1215.38
AIC=-1254.53
_____
Covariate X1 has been tested [ic=-2269.80243895166, lag=-2]
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
Covariate X1 has been added [aicc=-2269.80243895166, lag=-2]
Series: serie
Regression with ARIMA(2,1,1) errors
Coefficients:
        ar1
                ar2
                        ma1
                                 Х3
                                        X2
                                                 X1
     0.2283 0.7589 0.3025 2.2676 2.1484 -1.3085
s.e. 0.0299 0.0294 0.0396 0.0321 0.0280
sigma^2 = 0.005786: log likelihood = 1141.96
```

BIC=-2235.66

AIC=-2269.92

AICc=-2269.8

Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 No more variables will be added

The global model does not have stationary errors

Trying to adjust a model that do have stationary errors

No valid model with stationary errors could be optimized

Applying regular differentiation (ndiff=1) and calling again the function

Covariate X1 has been tested [ic=-127.111191500974, lag=-2]

Covariate X2 has been tested [ic=-350.746172271828, lag=0]

Covariate X3 has been tested [ic=-432.584774965942, lag=-1]

Significative correlation with lag<=0 could not be found for X4

Significative correlation with lag<=0 could not be found for X5

Significative correlation with lag<=0 could not be found for X6

Covariate X3 has been added [aicc=-432.584774965942, lag=-1]

Series: serie

Regression with ARIMA(4,1,0) errors

Coefficients:

ar1 ar2 ar3 ar4 xreg -0.7949 -0.6217 -0.2668 -0.0734 2.3192 s.e. 0.0318 0.0397 0.0398 0.0320 0.0907

sigma^2 = 0.03744: log likelihood = 222.34 AIC=-432.67 AICc=-432.58 BIC=-403.31

Covariate X1 has been tested [ic=-651.671017504022, lag=-2] Covariate X2 has been tested [ic=-1252.41874509484, lag=0]

Significative correlation with lag<=0 could not be found for X4

Significative correlation with lag<=0 could not be found for X5

Covariate X6 has been tested [ic=-437.623697388088, lag=-9]

Covariate X2 has been added [aicc=-1252.41874509484, lag=0]

Series: serie

Regression with ARIMA(2,1,2) errors

Coefficients:

ar1 ar2 ma1 ma2 X3 X2 0 -0.4898 -0.6156 0.4634 2.2891 2.1249 s.e. 0 0.0430 0.0308 0.0362 0.0531 0.0564

sigma^2 = 0.01629: log likelihood = 632.25 AIC=-1252.5 AICc=-1252.42 BIC=-1223.15

Covariate X1 has been tested [ic=-1863.55961789977, lag=-2]

Significative correlation with lag<=0 could not be found for X4

Significative correlation with lag<=0 could not be found for X5

Significative correlation with lag<=0 could not be found for X6

Covariate X1 has been added [aicc=-1863.55961789977, lag=-2]

Series: serie

Regression with ARIMA(0,1,0) errors

```
Coefficients:
```

```
X3 X2 X1
2.2783 2.1512 -1.3106
s.e. 0.0297 0.0319 0.0316
```

```
sigma<sup>2</sup> = 0.008783: log likelihood = 935.8
AIC=-1863.6 AICc=-1863.56 BIC=-1844.03
```

Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 No more variables will be added

The global model does not have stationary errors

Trying to adjust a model that do have stationary errors

| Historical of added covariates to the model (ndiff=1) |

```
var lag ic

X3 -1 -432.584774965942

X2 0 -1252.41874509484

X1 -2 -1863.55961789977
```

Series: serie

Regression with ARIMA(2,0,1) errors

Coefficients:

```
ar1 ar2 ma1 X3 X2 X1 0.2283 0.7589 0.3024 2.2677 2.1484 -1.3086 s.e. 0.0299 0.0295 0.0396 0.0321 0.0280 0.0241
```

```
sigma^2 = 0.005786: log likelihood = 1141.96
AIC=-2269.92 AICc=-2269.8 BIC=-2235.66
```

Como podemos observar, el *output* es muy similar al obtenido en el ejemplo anterior. En las primeras iteraciones se añaden las variables $X_t^{(3)}$, $X_t^{(2)}$ y $X_t^{(1)}$, en ese orden, con retardos $r_3=1$, $r_2=0$ y $r_1=1$ (correcto según la definición del modelo), pero al no poder ajustar un modelo con errores estacionarios (se obtiene un modelo con errores ARIMA(2,1,1)), se tienen que diferenciar los datos y volver a llamar a la función.

En la siguiente llamada, se vuelven a añadir las mismas variables con los retardos correctos, pero esta vez sí se consigue ajustar un modelo con errores estacionarios (un ARIMA(2,0,1)) con un AICc=-2269.51.

3 Comparativa del método de preblanqueado

3.1 Con errores estacionarios

```
load(file='.../simulations/residuals ~ ARIMA(2,0,1).RData') # residuals
beta0 <- -0.1; beta1 <- 3.2; beta2 <- -2.5
r1 <- 2; r2 <- 3
Y <- beta0 + beta1 * lag(X1$X, -r1) + beta2 * lag(X2$X, -r2) + residuals$X
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)
Ajustamos un modelo usando como método para chequear estacionariedad la función auto.arima:
ajuste <- drm.select(Y, regresoras, show_info=T,</pre>
```

```
st method='auto.arima')
Covariate X1 has been tested [ic=-1088.61845492454, lag=-2]
Covariate X2 has been tested [ic=-595.912721666589, lag=-3]
Covariate X3 has been tested [ic=-146.378165277729, lag=-5]
Significative correlation with lag<=0 could not be found for X4
Covariate X5 has been tested [ic=-142.384016382745, lag=-21]
Covariate X6 has been tested [ic=-142.384016382745, lag=-2]
Covariate X1 has been added [aicc=-1088.61845492454, lag=-2]
Series: serie
Regression with ARIMA(0,1,2) errors
Coefficients:
         ma1
               ma2
                         xreg
     -0.4887 -0.1949 3.1621
s.e. 0.0314 0.0316 0.0684
sigma^2 = 0.01904: log likelihood = 548.33
             AICc=-1088.62
AIC=-1088.66
                             BIC=-1069.13
Covariate X2 has been tested [ic=-3114.07541439909, lag=-3]
Covariate X3 has been tested [ic=-1088.61845591082, lag=-6]
Covariate X4 has been tested [ic=-1088.61845475434, lag=-9]
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-1088.61845508187, lag=-8]
Covariate X2 has been added [aicc=-3114.07541439909, lag=-3]
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
                       ma1 intercept
         ar1
                 ar2
                                           X1
     -0.1927 0.4046 0.4424 -0.0938 3.1927 -2.4812
s.e. 0.0713 0.0307 0.0757 0.0035 0.0093
                                                0.0109
sigma^2 = 0.002381: log likelihood = 1564.1
AIC=-3114.19
             AICc=-3114.08
                             BIC=-3080.01
Significative correlation with lag<=0 could not be found for X3
```

Significative correlation with lag<=0 could not be found for X4

```
Significative correlation with lag<=0 could not be found for X6
No more variables will be added
               Historical of added covariates to the model (ndiff=0)
 var lag
 X1 -2 -1088.61845492454
 X2 -3 -3114.07541439909
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
                 ar2
                        ma1 intercept
                                           X1
                                                     Х2
         ar1
                                -0.0938 3.1927 -2.4812
     -0.1927 0.4046 0.4424
s.e. 0.0713 0.0307 0.0757
                               0.0035 0.0093 0.0109
sigma^2 = 0.002381: log likelihood = 1564.1
             AICc=-3114.08 BIC=-3080.01
AIC=-3114.19
Ajustamos un modelo usando como método para chequear estacionariedad el adf.test:
ajuste <- drm.select(Y, regresoras, show_info=T, st_method='adf.test')
Covariate X1 has been tested [ic=-1088.61845492454, lag=-2]
Covariate X2 has been tested [ic=-595.912721666589, lag=-3]
Covariate X3 has been tested [ic=-146.378165277729, lag=-5]
Significative correlation with lag<=0 could not be found for X4
Covariate X5 has been tested [ic=-142.384016382745, lag=-21]
Covariate X6 has been tested [ic=-142.384016382745, lag=-2]
Covariate X1 has been added [aicc=-1088.61845492454, lag=-2]
Series: serie
Regression with ARIMA(0,1,2) errors
Coefficients:
         ma1
                 \mathtt{ma2}
                         xreg
     -0.4887 -0.1949 3.1621
s.e. 0.0314 0.0316 0.0684
sigma^2 = 0.01904: log likelihood = 548.33
AIC=-1088.66
             AICc=-1088.62 BIC=-1069.13
Covariate X2 has been tested [ic=-3114.07541439909, lag=-3]
Covariate X3 has been tested [ic=-1088.61845591082, lag=-6]
Covariate X4 has been tested [ic=-1088.61845475434, lag=-9]
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-1088.61845508187, lag=-8]
Covariate X2 has been added [aicc=-3114.07541439909, lag=-3]
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
                 ar2 ma1 intercept X1
         ar1
     -0.1927 0.4046 0.4424 -0.0938 3.1927 -2.4812
```

Covariate X5 has been tested [ic=-3114.07521810619, lag=-8]

```
s.e. 0.0713 0.0307 0.0757 0.0035 0.0093 0.0109
sigma^2 = 0.002381: log likelihood = 1564.1
            AICc=-3114.08 BIC=-3080.01
AIC=-3114.19
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Covariate X5 has been tested [ic=-3114.07521810619, lag=-8]
Significative correlation with lag<=0 could not be found for X6
No more variables will be added
              Historical of added covariates to the model (ndiff=0)
var lag
 X1 -2 -1088.61845492454
 X2 -3 -3114.07541439909
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
                ar2 ma1 intercept X1
                                                   X2
         ar1
     -0.1927 0.4046 0.4424 -0.0938 3.1927 -2.4812
    0.0713 0.0307 0.0757
                               0.0035 0.0093 0.0109
s.e.
sigma^2 = 0.002381: log likelihood = 1564.1
AIC=-3114.19 AICc=-3114.08 BIC=-3080.01
```

3.2 Con errores no estacionarios

```
load(file='../simulations/residuals ~ ARIMA(1,2,2).RData') # residuals
beta0 <- -0.1; beta1 <- 3.2; beta2 <- -2.5
r1 <- 2; r2 <- 3
Y <- beta0 + beta1 * lag(X1$X, -r1) + beta2 * lag(X2$X, -r2) + residuals$X
regresoras <- cbind(X1=X1$X, X2=X2$X, X3=X3$X, X4=X4$X, X5=X5$X, X6=X6$X)</pre>
```

Ajustamos un modelo usando como método para chequear estacionariedad la función auto.arima:

```
ajuste <- drm.select(Y, regresoras, show_info=T, st_method='auto.arima')

Covariate X1 has been tested [ic=-649.682587578322, lag=-2]

Covariate X2 has been tested [ic=-434.459021738045, lag=-3]

Covariate X3 has been tested [ic=57.7067437758955, lag=-1]

Covariate X4 has been tested [ic=53.6412837880618, lag=-9]

Covariate X5 has been tested [ic=57.7067437758955, lag=-23]

Covariate X6 has been tested [ic=60.6278346700799, lag=-6]

Covariate X1 has been added [aicc=-649.682587578322, lag=-2]

Series: serie

Regression with ARIMA(2,1,3) errors

Coefficients:

ar1 ar2 ma1 ma2 ma3 xreg
```

```
0.5931 0.3925 -0.5505 -0.3111 0.2199 3.1695
s.e. 0.1656 0.1644 0.1633 0.1478 0.0367 0.0884
sigma^2 = 0.02971: log likelihood = 331.9
AIC=-649.8 AICc=-649.68 BIC=-615.64
Covariate X2 has been tested [ic=-3025.1467782192, lag=-3]
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-652.50410291203, lag=-8]
Covariate X2 has been added [aicc=-3025.1467782192, lag=-3]
Series: serie
Regression with ARIMA(2,1,1) errors
Coefficients:
                              X1
                                      X2
       ar1
              ar2
                      ma1
     0.2283 0.7584 0.3004 3.1983 -2.4964
     0.0299 0.0295 0.0395 0.0163
sigma^2 = 0.002584: log likelihood = 1518.62
AIC=-3025.23
            AICc=-3025.15
                           BIC=-2995.95
______
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
No more variables will be added
The global model does not have stationary errors
Trying to adjust a model that do have stationary errors
No valid model with stationary errors could be optimized
Applying regular differentiation (ndiff=1) and calling again the function
______
Covariate X1 has been tested [ic=-646.897656850379, lag=-2]
Covariate X2 has been tested [ic=-430.04365565032, lag=-3]
Covariate X3 has been tested [ic=60.6278158177817, lag=-1]
Covariate X4 has been tested [ic=53.6412661805395, lag=-9]
Covariate X5 has been tested [ic=60.6278158177817, lag=-23]
Covariate X6 has been tested [ic=60.6278158177817, lag=-6]
Covariate X1 has been added [aicc=-646.897656850379, lag=-2]
Series: serie
Regression with ARIMA(1,1,3) errors
Coefficients:
        ar1
                ma1
                        ma2
                               ma3
                                      xreg
     -0.4113 -0.5402 -0.3262 0.2186 3.1688
s.e. 0.1622 0.1597 0.1446 0.0361 0.0882
sigma^2 = 0.02984: log likelihood = 329.49
AIC=-646.98 AICc=-646.9 BIC=-617.71
______
Covariate X2 has been tested [ic=-3087.52651374318, lag=-3]
```

Significative correlation with lag<=0 could not be found for X3

```
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-651.270885560183, lag=-8]
Covariate X2 has been added [aicc=-3087.52651374318, lag=-3]
Series: serie
Regression with ARIMA(1,1,2) errors
Coefficients:
         ar1 ma1 ma2
                            X1
                                      X2
     -0.4461 0 0.3683 3.2043 -2.4945
s.e. 0.0306 0 0.0308 0.0129 0.0165
sigma^2 = 0.002427: log likelihood = 1548.79
AIC=-3087.59 AICc=-3087.53 BIC=-3063.19
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
No more variables will be added
The global model does not have stationary errors
Trying to adjust a model that do have stationary errors
______
              Historical of added covariates to the model (ndiff=1)
var lag
 X1 -2 -646.897656850379
 X2 -3 -3087.52651374318
Series: serie
Regression with ARIMA(2,0,1) errors
Coefficients:
              ar2 ma1 X1
                                        X2
        ar1
     0.2283 0.7584 0.3004 3.1983 -2.4964
s.e. 0.0299 0.0295 0.0395 0.0163
sigma^2 = 0.002584: log likelihood = 1518.62
AIC=-3025.23
            AICc=-3025.15
                             BIC=-2995.95
Ajustamos un modelo usando como método para chequear estacionariedad el adf.test:
ajuste <- drm.select(Y, regresoras, show info=T, st method='adf.test')</pre>
Covariate X1 has been tested [ic=-649.682587578322, lag=-2]
Covariate X2 has been tested [ic=-434.459021738045, lag=-3]
Covariate X3 has been tested [ic=57.7067437758955, lag=-1]
Covariate X4 has been tested [ic=53.6412837880618, lag=-9]
Covariate X5 has been tested [ic=57.7067437758955, lag=-23]
Covariate X6 has been tested [ic=60.6278346700799, lag=-6]
Covariate X1 has been added [aicc=-649.682587578322, lag=-2]
Series: serie
Regression with ARIMA(2,1,3) errors
```

Coefficients:

```
ar2
                     ma1
                              ma2
                                     ma3
     0.5931 0.3925 -0.5505 -0.3111 0.2199 3.1695
s.e. 0.1656 0.1644 0.1633 0.1478 0.0367 0.0884
sigma^2 = 0.02971: log likelihood = 331.9
AIC=-649.8 AICc=-649.68 BIC=-615.64
______
Covariate X2 has been tested [ic=-3025.1467782192, lag=-3]
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Covariate X6 has been tested [ic=-652.50410291203, lag=-8]
Covariate X2 has been added [aicc=-3025.1467782192, lag=-3]
Series: serie
Regression with ARIMA(2,1,1) errors
Coefficients:
        ar1
               ar2
                      ma1
                             Х1
     0.2283 0.7584 0.3004 3.1983 -2.4964
s.e. 0.0299 0.0295 0.0395 0.0163 0.0188
sigma^2 = 0.002584: log likelihood = 1518.62
AIC=-3025.23
           AICc=-3025.15 BIC=-2995.95
Significative correlation with lag<=0 could not be found for X3
Significative correlation with lag<=0 could not be found for X4
Significative correlation with lag<=0 could not be found for X5
Significative correlation with lag<=0 could not be found for X6
No more variables will be added
The global model does not have stationary errors
Trying to adjust a model that do have stationary errors
No valid model with stationary errors could be optimized
Applying regular differentiation (ndiff=1) and calling again the function
-----
Covariate X1 has been tested [ic=-646.897656850379, lag=-2]
Covariate X2 has been tested [ic=-430.04365565032, lag=-3]
Covariate X3 has been tested [ic=60.6278158177817, lag=-1]
Covariate X4 has been tested [ic=53.6412661805395, lag=-9]
Covariate X5 has been tested [ic=60.6278158177817, lag=-23]
Covariate X6 has been tested [ic=60.6278158177817, lag=-6]
Covariate X1 has been added [aicc=-646.897656850379, lag=-2]
Series: serie
Regression with ARIMA(1,1,3) errors
Coefficients:
        ar1
                ma1
                        ma2
                                ma3
                                      xreg
     -0.4113 -0.5402 -0.3262 0.2186 3.1688
      0.1622  0.1597  0.1446  0.0361  0.0882
sigma^2 = 0.02984: log likelihood = 329.49
AIC=-646.98 AICc=-646.9 BIC=-617.71
_____
```

Covariate X2 has been tested [ic=-3087.52651374318, lag=-3]

Significative correlation with lag<=0 could not be found for X3 Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Covariate X6 has been tested [ic=-651.270885560183, lag=-8] Covariate X2 has been added [aicc=-3087.52651374318, lag=-3] Series: serie

Regression with ARIMA(1,1,2) errors

Coefficients:

Significative correlation with lag<=0 could not be found for X3 Significative correlation with lag<=0 could not be found for X4 Significative correlation with lag<=0 could not be found for X5 Significative correlation with lag<=0 could not be found for X6 No more variables will be added
The global model does not have stationary errors

Trying to adjust a model that do have stationary errors

Historical of added covariates to the model (ndiff=1)

var lag ic X1 -2 -646.897656850379 X2 -3 -3087.52651374318

Series: serie

Regression with ARIMA(2,0,1) errors

Coefficients:

ar1 ar2 ma1 X1 X2 0.2283 0.7584 0.3004 3.1983 -2.4964 s.e. 0.0299 0.0295 0.0395 0.0163 0.0188

sigma² = 0.002584: log likelihood = 1518.62 AIC=-3025.23 AICc=-3025.15 BIC=-2995.95

4 Apéndice

ARIMA(2,0,1) with zero mean

En esta sección se muestra la comprobación con la función auto.fit.arima de que las muestras cargadas cumplen con los requisitos mencionados.

Para el primer ejempo, las muestras simuladas eran las siguientes:

```
{\tt load(file='.../simulations/residuals~~ARIMA(2,0,1).RData')~\#~residuals}
Si la función auto.fit.arima y observamos los outputs, vemos que siguen el proceso ARIMA anotado:
auto.fit.arima(X1$X, show_info=F)
Series: serie
ARIMA(2,1,3)
Coefficients:
      ar1
              ar2
                       ma1
                                ma2
                                        ma3
       0 -0.4505 -0.0720 -0.1633 0.4002
          0.0465
                   0.0321
                             0.0443 0.0304
sigma^2 = 0.002651: log likelihood = 1547.33
              AICc=-3084.61
AIC=-3084.67
                             BIC=-3060.13
auto.fit.arima(X2$X, show_info=F)
Series: serie
ARIMA(1,1,2)
Coefficients:
         ar1 ma1
                       ma2
     -0.4570 0 -0.4718
     0.0317
                0 0.0315
s.e.
sigma^2 = 0.002683: log likelihood = 1540.68
AIC=-3075.37
              AICc=-3075.34
                              BIC=-3060.65
auto.fit.arima(X3$X, show_info=F)
Series: serie
ARIMA(1,0,2) with zero mean
Coefficients:
        ar1 ma1
                     ma2
             0 0.3508
      0.3602
s.e. 0.0311 0 0.0313
sigma^2 = 0.002282: log likelihood = 1623.08
AIC=-3240.16
              AICc=-3240.14
                              BIC=-3225.44
auto.fit.arima(residuals$X, show_info=F)
Series: serie
```

```
Coefficients:
         ar1
              ar2
                         ma1
     -0.2356 0.4323 0.4563
s.e. 0.0626 0.0295 0.0668
sigma^2 = 0.002469: log likelihood = 1584.27
              AICc=-3160.49
AIC=-3160.53
                             BIC=-3140.9
auto.fit.arima(X4$X, show_info=F)
Series: serie
ARIMA(1,0,3) with zero mean
Coefficients:
                 ma1 ma2
         ar1
                             ma3
     -0.5708 0.1396 0 0.4028
s.e. 0.0633 0.0650 0 0.0319
sigma^2 = 0.002529: log likelihood = 1572.27
AIC=-3136.55 AICc=-3136.51
                            BIC=-3116.92
auto.fit.arima(X5$X, show_info=F)
Series: serie
ARIMA(2,1,2)
Coefficients:
         ar1
                 ar2 ma1
                             ma2
     -0.0623 0.4527 0 0.4725
s.e. 0.0283 0.0373 0 0.0377
sigma^2 = 0.002489: log likelihood = 1577.98
AIC=-3147.95
             AICc=-3147.91 BIC=-3128.32
auto.fit.arima(X6$X, show_info=F)
Series: serie
ARIMA(2,1,1)
Coefficients:
         ar1
                ar2
                         ma1
     -0.2399 0.4477 0.4440
s.e. 0.0597 0.0290 0.0645
sigma^2 = 0.002464: log likelihood = 1583.7
AIC=-3159.4 AICc=-3159.36 BIC=-3139.78
Podemos hacer la misma comprobación con los residuos del ejemplo 2.
load('../simulations/residuals ~ ARIMA(1,2,2).RData')
auto.fit.arima(residuals$X, show_info=F)
Series: serie
ARIMA(1,2,2)
Coefficients:
```

 $\begin{array}{ccccc} & & ar1 & ma1 & ma2 \\ & -0.4430 & 0 & 0.3639 \\ s.e. & 0.0303 & 0 & 0.0305 \end{array}$

sigma^2 = 0.002414: log likelihood = 1591.83 AIC=-3177.65 AICc=-3177.63 BIC=-3162.93