

04-exercices-ARDI-cointégration

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Exercice 1 : L'Hypothèse de Fisher et la Cointégration

Données :intdef du package wooldridge.

*Tester la relation de long terme entre le taux d'intérêt des bons du trésor à 3 mois (*i3*) et le taux d'inflation (*inf*). Selon l'effet Fisher, une augmentation de l'inflation devrait se traduire par une augmentation équivalente des taux d'intérêt nominaux à long terme (coefficients de 1).*

Question 0

Charger le setup de l'analyse

Question 1

*Importer les données *intdef*. Créer un nouvel objet *TS* contenant les variables *i3* et *inf*. Représenter graphiquement les deux séries sur le même graphique. Semblent-elles évoluer ensemble ?*

Question 2

Réaliser les tests de stationnarité sur les deux variables. Sont-elles intégrées du même ordre ? Peut-on réaliser la procédure d'Engle-Granger pour la cointégration ? Si oui tester la cointégration entre ces deux variables avec cette procédure. Estimer le modèle à correction d'erreur lié. Que peut-on en conclure ?

Question 3

Estimer un modèle ARDL pour expliquer i_3 par inf . Quel sont les ordres choisi ? Pourquoi ?

Question 4

Réaliser le test des bornes (Bounds Test) sur le modèle estimé. Peut-on conclure à l'existence d'une relation de cointégration entre les taux d'intérêt et l'inflation au seuil de 5% ? Au seuil de 1% ?

Question 5

Si la cointégration est avérée, estimer les coefficients de long terme. Le coefficient associé à l'inflation est-il statistiquement différent de 1 ? L'hypothèse de Fisher est-elle validée sur cette période ?

Question 6

Estimer le modèle à correction d'erreur (ECM) associé. Quelle est la vitesse d'ajustement vers l'équilibre de long terme ? Commentez son signe et sa significativité.

Question 7

Représenter l'équation de long-terme de i_3 et la comparer aux données réelles. Que pouvez-vous en dire ?

Question 8

Le modèle satisfait-il les hypothèses des MCO ? Si non comment pourriez-vous corriger cela ?

Exercice 2 : La courbe de Phillips

Le trésor américain vous demande d'utiliser les données *phillips* du package *woolridge* afin de tester l'existence de la courbe de Phillips, c'est à dire l'existence d'une relation négative entre l'inflation et le chômage. Définissez une stratégie et une modélisation pour tester cette relation. Cette relation existe-t-elle à court- et long-terme ? Comment se situe l'inflation réelle par rapport à son niveau de long-terme ? Que pourriez-vous proposer au trésor américain pour améliorer l'analyse ?

Exercice 3 : Fertilité

Vous venez de récupérer le script R d'un économètre senior qui a quitté le bureau. Il étudie la relation entre le taux de fertilité (*gfr*) et l'exonération fiscale personnelle réelle (*pe*) aux États-Unis sur la période 1913-1984. Expliquez et interprétez les différentes étapes. Que pouvez-vous conclure ? Quelles améliorations et suites pouvez-vous proposer ?

```
source(here::here("02-codes", "utils", "setup.R"))
```

```
wooldridge::fertil3 |>
  str()

'data.frame': 72 obs. of 24 variables:
 $ gfr    : num  125 127 125 123 121 ...
 $ pe     : num  0 0 0 0 19.3 ...
 $ year   : int  1913 1914 1915 1916 1917 1918 1919 1920 1921 1922 ...
 $ t      : int  1 2 3 4 5 6 7 8 9 10 ...
 $ tsq   : int  1 4 9 16 25 36 49 64 81 100 ...
 $ pe_1  : num  NA 0 0 0 0 ...
 $ pe_2  : num  NA NA 0 0 0 ...
 $ pe_3  : num  NA NA NA 0 0 ...
 $ pe_4  : num  NA NA NA NA 0 ...
 $ pill  : int  0 0 0 0 0 0 0 0 0 ...
 $ ww2   : int  0 0 0 0 0 0 0 0 0 ...
 $ tcu   : num  1 8 27 64 125 216 343 512 729 1000 ...
 $ cgfr  : num  NA 1.9 -1.6 -1.6 -2.4 ...
 $ cpe   : num  NA 0 0 0 19.3 ...
 $ cpe_1 : num  NA NA 0 0 0 ...
 $ cpe_2 : num  NA NA NA 0 0 ...
 $ cpe_3 : num  NA NA NA NA 0 ...
 $ cpe_4 : num  NA NA NA NA NA ...
```

```

$ gfr_1 : num  NA 125 127 125 123 ...
$ cgfr_1: num  NA NA 1.9 -1.6 -1.6 ...
$ cgfr_2: num  NA NA NA 1.9 -1.6 ...
$ cgfr_3: num  NA NA NA NA 1.9 ...
$ cgfr_4: num  NA NA NA NA NA ...
$ gfr_2 : num  NA NA 125 127 125 ...
- attr(*, "time.stamp")= chr "25 Jun 2011 23:03"

```

```

df <-
  wooldridge::fertil3 |>
  as_tibble() |>
  mutate(
    year = lubridate::make_date(year)
  ) |>
  select(year, gfr, pe, pill, ww2) |>
  mutate(
    across(
      .cols = c(gfr, pe),
      .fns = \(var) log(var),
      .names = "log_{.col}"
    ),
    across(
      .cols = c(gfr, pe),
      .fns = \(var) log(var) - log(lag(var)),
      .names = "diff_log_{.col}"
    ),
    across(
      .cols = everything(),
      .fns = \(var) if_else(is.infinite(var), NA, var)
    )
  ) |>
  drop_na() |>
  print()

```

```

# A tibble: 67 x 9
  year      gfr     pe   pill   ww2 log_gfr log_pe diff_log_gfr diff_log_pe
  <date>    <dbl>  <dbl> <int> <int>   <dbl>   <dbl>        <dbl>        <dbl>
1 1918-01-01 120.   23.9     0     0    4.79    3.18     -0.00997    0.217
2 1919-01-01 111.   20.1     0     0    4.71    3.00     -0.0745    -0.176
3 1920-01-01 118.   15.3     0     0    4.77    2.73      0.0585    -0.269
4 1921-01-01 120.   34.3     0     0    4.79    3.54      0.0160    0.806
5 1922-01-01 111.   36.7     0     0    4.71    3.60     -0.0745    0.0657
6 1923-01-01 110.   25.8     0     0    4.71    3.25     -0.00631   -0.350

```

```

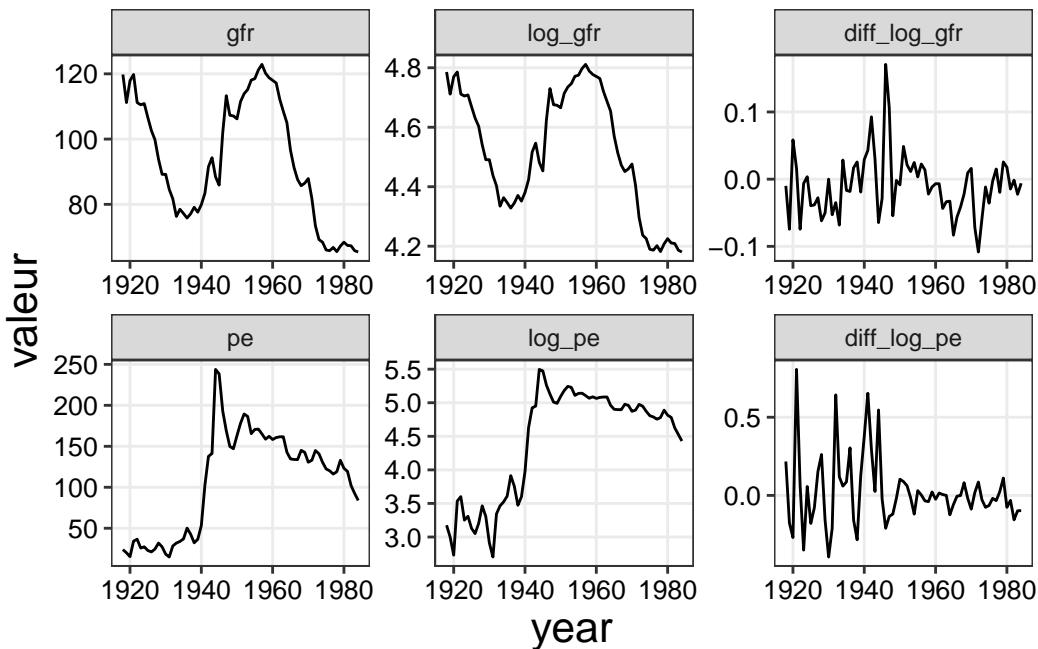
7 1924-01-01 111.    27.3      0      0     4.71     3.31      0.00361     0.0568
8 1925-01-01 107.    22.9      0      0     4.67     3.13     -0.0395    -0.179
9 1926-01-01 103.    21.1      0      0     4.63     3.05     -0.0382    -0.0783
10 1927-01-01 99.8    24.6     0      0     4.60     3.20     -0.0277    0.152
# i 57 more rows

```

```

df |>
pivot_longer(
  cols = !c(year, pill, ww2),
  names_to = "nom",
  values_to = "valeur"
) |>
mutate(
  nom = factor(nom, levels = c("gfr", "log_gfr", "diff_log_gfr", "pe",
  "log_pe", "diff_log_pe"))
) |>
ggplot(aes(x = year, y = valeur)) +
geom_line() +
facet_wrap("nom", scales= "free")

```



```

df_ts <-
df |>

```

```

timetk::tk_zooreg(
  select = !year,
  start = df$year[1] |> year(),
  frequency = 1
)

head(df_ts)

      gfr     pe pill ww2 log_gfr   log_pe diff_log_gfr diff_log_pe
1918 119.8 23.94     0    0 4.785824 3.175551 -0.009966834  0.2170012
1919 111.2 20.07     0    0 4.711330 2.999226 -0.074493357 -0.1763246
1920 117.9 15.33     0    0 4.769837 2.729812  0.058506466 -0.2694145
1921 119.8 34.32     0    0 4.785824 3.535728  0.015986891  0.8059166
1922 111.2 36.65     0    0 4.711330 3.601413 -0.074493357  0.0656852
1923 110.5 25.83     0    0 4.705016 3.251537 -0.006314833 -0.3498769

```

```

urca::ur.df(df$log_gfr, "trend", lags = 20, selectlags = "AIC") |>
  summary()

```

```

#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####

```

Test regression trend

Call:
`lm(formula = z.diff ~ z.lag.1 + 1 + tt + z.diff.lag)`

Residuals:

Min	1Q	Median	3Q	Max
-0.066296	-0.024084	0.002171	0.023550	0.123953

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.345161	0.159414	2.165	0.03640 *
z.lag.1	-0.062428	0.032269	-1.935	0.06014 .
tt	-0.001536	0.000596	-2.576	0.01378 *
z.diff.lag1	0.507677	0.149127	3.404	0.00152 **
z.diff.lag2	-0.476835	0.150441	-3.170	0.00293 **
z.diff.lag3	0.213956	0.148141	1.444	0.15645

```
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.03916 on 40 degrees of freedom
Multiple R-squared: 0.4223, Adjusted R-squared: 0.3501
F-statistic: 5.848 on 5 and 40 DF, p-value: 0.0003817
```

```
Value of test-statistic is: -1.9346 2.4844 3.6485
```

```
Critical values for test statistics:
```

	1pct	5pct	10pct
tau3	-4.04	-3.45	-3.15
phi2	6.50	4.88	4.16
phi3	8.73	6.49	5.47

```
urca::ur.df(df$log_gfr, "drift", lags = 20, selectlags = "AIC") |>
  summary()
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
```

```
Test regression drift
```

```
Call:
```

```
lm(formula = z.diff ~ z.lag.1 + 1 + z.diff.lag)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-0.05585	-0.02457	-0.00752	0.02520	0.13205

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.20538	0.14039	1.463	0.151507
z.lag.1	-0.04588	0.03106	-1.477	0.147623
z.diff.lag1	0.55211	0.15172	3.639	0.000792 ***
z.diff.lag2	-0.42706	0.17601	-2.426	0.019974 *
z.diff.lag3	0.35730	0.18457	1.936	0.060155 .
z.diff.lag4	0.02794	0.17506	0.160	0.874023
z.diff.lag5	0.29329	0.15744	1.863	0.070025 .

```
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.04035 on 39 degrees of freedom
Multiple R-squared: 0.402, Adjusted R-squared: 0.31
F-statistic: 4.369 on 6 and 39 DF, p-value: 0.00182
```

```
Value of test-statistic is: -1.4773 1.1382
```

```
Critical values for test statistics:
```

	1pct	5pct	10pct
tau2	-3.51	-2.89	-2.58
phi1	6.70	4.71	3.86

```
urca::ur.df(df$log_gfr, "none", lags = 20, selectlags = "AIC") |>
  summary()
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
```

```
Test regression none
```

```
Call:
```

```
lm(formula = z.diff ~ z.lag.1 - 1 + z.diff.lag)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-0.068179	-0.019933	-0.004234	0.014968	0.150982

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
z.lag.1	-0.0005542	0.0013622	-0.407	0.68621
z.diff.lag1	0.6243398	0.1462812	4.268	0.00011 ***
z.diff.lag2	-0.4763187	0.1589243	-2.997	0.00456 **
z.diff.lag3	0.3128714	0.1458749	2.145	0.03780 *

```
---
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.04154 on 42 degrees of freedom
```

```
Multiple R-squared:  0.3223,    Adjusted R-squared:  0.2578
F-statistic: 4.994 on 4 and 42 DF,  p-value: 0.002196
```

```
Value of test-statistic is: -0.4068
```

```
Critical values for test statistics:
```

```
 1pct  5pct 10pct
tau1 -2.6 -1.95 -1.61
```

```
urca::ur.df(df$log_pe, "trend", lags = 20, selectlags = "AIC") |>
  summary()
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
```

```
Test regression trend
```

```
Call:
```

```
lm(formula = z.diff ~ z.lag.1 + 1 + tt + z.diff.lag)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-0.144412	-0.049409	-0.009865	0.029167	0.194965

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.314598	0.237152	5.543	0.00000503 ***
z.lag.1	-0.211401	0.049373	-4.282	0.000175 ***
tt	-0.005685	0.001908	-2.979	0.005687 **
z.diff.lag1	0.240831	0.134138	1.795	0.082670 .
z.diff.lag2	-0.033309	0.133387	-0.250	0.804508
z.diff.lag3	0.103594	0.120654	0.859	0.397370
z.diff.lag4	-0.186664	0.121600	-1.535	0.135247
z.diff.lag5	0.108018	0.126511	0.854	0.399973
z.diff.lag6	-0.148250	0.119611	-1.239	0.224791
z.diff.lag7	-0.210392	0.111473	-1.887	0.068817 .
z.diff.lag8	0.074421	0.090530	0.822	0.417535
z.diff.lag9	0.123085	0.090692	1.357	0.184847
z.diff.lag10	-0.031785	0.093143	-0.341	0.735294

```
z.diff.lag11 -0.114178  0.092077 -1.240  0.224574
z.diff.lag12  0.137761  0.092844  1.484  0.148297
z.diff.lag13 -0.202049  0.094045 -2.148  0.039873 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 0.09584 on 30 degrees of freedom
Multiple R-squared:  0.7681,    Adjusted R-squared:  0.6522
F-statistic: 6.624 on 15 and 30 DF,  p-value: 0.000006003
```

Value of test-statistic is: -4.2817 11.7878 17.5893

Critical values for test statistics:

	1pct	5pct	10pct
tau3	-4.04	-3.45	-3.15
phi2	6.50	4.88	4.16
phi3	8.73	6.49	5.47

```
urca::ur.df(df$log_pe, "drift", lags = 20, selectlags = "AIC") |>
  summary()
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
```

Test regression drift

Call:
lm(formula = z.diff ~ z.lag.1 + 1 + z.diff.lag)

Residuals:

Min	1Q	Median	3Q	Max
-0.181373	-0.061134	0.000062	0.059044	0.294933

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.24738	0.24410	5.110	0.0000115 ***
z.lag.1	-0.25430	0.04963	-5.124	0.0000110 ***
z.diff.lag1	0.31018	0.13412	2.313	0.0267 *
z.diff.lag2	0.02953	0.13826	0.214	0.8321

```

z.diff.lag3  0.23517   0.12406   1.896   0.0663 .
z.diff.lag4 -0.10299   0.12641   -0.815   0.4207
z.diff.lag5  0.12871   0.12615   1.020   0.3146
z.diff.lag6 -0.11107   0.12677   -0.876   0.3869
z.diff.lag7 -0.06423   0.10963   -0.586   0.5617
z.diff.lag8  0.18862   0.09437   1.999   0.0535 .
z.diff.lag9  0.18969   0.09734   1.949   0.0594 .
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

Residual standard error: 0.108 on 35 degrees of freedom
Multiple R-squared:  0.6564,    Adjusted R-squared:  0.5582
F-statistic: 6.686 on 10 and 35 DF,  p-value: 0.0000105

```

Value of test-statistic is: -5.1236 13.1255

Critical values for test statistics:

	1pct	5pct	10pct
tau2	-3.51	-2.89	-2.58
phi1	6.70	4.71	3.86

```

urca::ur.df(df$log_pe, "none", lags = 20, selectlags = "AIC") |>
  summary()

```

```

#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####


```

Test regression none

Call:

```
lm(formula = z.diff ~ z.lag.1 - 1 + z.diff.lag)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.26757	-0.06048	-0.01021	0.03256	0.52866

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
z.lag.1	0.001358	0.004451	0.305	0.76174

```

z.diff.lag 0.434901   0.131256   3.313  0.00185 **  

---  

Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
  

Residual standard error: 0.1477 on 44 degrees of freedom  

Multiple R-squared:  0.2057,    Adjusted R-squared:  0.1696  

F-statistic: 5.697 on 2 and 44 DF,  p-value: 0.006305

```

Value of test-statistic is: 0.3051

Critical values for test statistics:
1pct 5pct 10pct
tau1 -2.6 -1.95 -1.61

```

lm_statique <- lm(log_gfr ~ log_pe + pill + ww2, data = df)  
  

summary(lm_statique)

```

Call:
lm(formula = log_gfr ~ log_pe + pill + ww2, data = df)

Residuals:
Min 1Q Median 3Q Max
-0.27429 -0.11687 -0.00896 0.09777 0.32559

Coefficients:
Estimate Std. Error t value Pr(>|t|)
(Intercept) 4.23016 0.10649 39.723 < 0.0000000000000002 ***
log_pe 0.09509 0.02533 3.754 0.000381 ***
pill -0.35257 0.04355 -8.096 0.000000000243 ***
ww2 -0.23036 0.07469 -3.084 0.003031 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1483 on 63 degrees of freedom
Multiple R-squared: 0.5113, Adjusted R-squared: 0.488
F-statistic: 21.97 on 3 and 63 DF, p-value: 0.0000000007464

```

lm_ar <- dynlm::dynlm(log_gfr ~ L(log_gfr) + log_pe + pill + ww2, data =
df_ts)

```

```
summary(lm_ar)
```

Time series regression with "zooreg" data:

Start = 1919, End = 1984

Call:

```
dynlm::dynlm(formula = log_gfr ~ L(log_gfr) + log_pe + pill +  
ww2, data = df_ts)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.08281	-0.02649	-0.00136	0.02233	0.12749

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.387244	0.144230	2.685	0.00933 **
L(log_gfr)	0.892653	0.033087	26.979 < 0.0000000000000002	***
log_pe	0.025479	0.007623	3.342	0.00142 **
pill	-0.066777	0.015977	-4.180	0.0000947 ***
ww2	-0.022759	0.021984	-1.035	0.30462

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.04086 on 61 degrees of freedom

Multiple R-squared: 0.9631, Adjusted R-squared: 0.9607

F-statistic: 398.1 on 4 and 61 DF, p-value: < 0.0000000000000022

```
model_selection <-  
  ARDL::auto_ardl(  
    log_gfr ~ log_pe | ww2 + pill,  
    data = df_ts,  
    max_order = 5,  
    selection = "BIC"  
)  
  
model_selection$top_orders
```

	log_gfr	log_pe	BIC
1	3	3	-221.7543
2	2	3	-217.1013
3	3	4	-216.9120

```

4      1      0 -214.8433
5      4      3 -214.6643
6      4      4 -213.1556
7      1      1 -212.8469
8      2      4 -211.6981
9      2      2 -204.4353
10     5      5 -200.7002

```

```

model <- model_selection$best_model

summary(model)

```

Time series regression with "zooreg" data:
 Start = 1921, End = 1984

Call:

```
dynlm::dynlm(formula = full_formula, data = data, start = start,
end = end)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.070097	-0.020754	0.000676	0.022199	0.077559

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.347607	0.119882	2.900	0.005393 **
L(log_gfr, 1)	1.205398	0.109869	10.971	0.0000000000000232 ***
L(log_gfr, 2)	-0.600023	0.157120	-3.819	0.000348 ***
L(log_gfr, 3)	0.290663	0.102951	2.823	0.006642 **
log_pe	0.031546	0.024540	1.285	0.204120
L(log_pe, 1)	-0.006867	0.031203	-0.220	0.826636
L(log_pe, 2)	0.077847	0.031029	2.509	0.015148 *
L(log_pe, 3)	-0.072247	0.023639	-3.056	0.003480 **
ww2	-0.040153	0.023030	-1.743	0.086938 .
pill	-0.063087	0.014505	-4.349	0.00006093999116124 ***

Signif. codes:	0 ***	0.001 **	0.01 *	0.05 .
	''	'	'	'

Residual standard error: 0.03259 on 54 degrees of freedom
 Multiple R-squared: 0.9784, Adjusted R-squared: 0.9748
 F-statistic: 271.5 on 9 and 54 DF, p-value: < 0.00000000000000022

```
uecm_model <- ARDL::uecm(model)
```

```
summary(uecm_model)
```

Time series regression with "zooreg" data:

Start = 1921, End = 1984

Call:

```
dynlm::dynlm(formula = full_formula, data = data, start = start,
end = end)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.070097	-0.020754	0.000676	0.022199	0.077559

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.347607	0.119882	2.900	0.005393 **
L(log_gfr, 1)	-0.103962	0.028118	-3.697	0.000511 ***
L(log_pe, 1)	0.030279	0.007151	4.234	0.0000897 ***
d(L(log_gfr, 1))	0.309360	0.103443	2.991	0.004187 **
d(L(log_gfr, 2))	-0.290663	0.102951	-2.823	0.006642 **
d(log_pe)	0.031546	0.024540	1.285	0.204120
d(L(log_pe, 1))	-0.005600	0.022250	-0.252	0.802232
d(L(log_pe, 2))	0.072247	0.023639	3.056	0.003480 **
ww2	-0.040153	0.023030	-1.743	0.086938 .
pill	-0.063087	0.014505	-4.349	0.0000609 ***

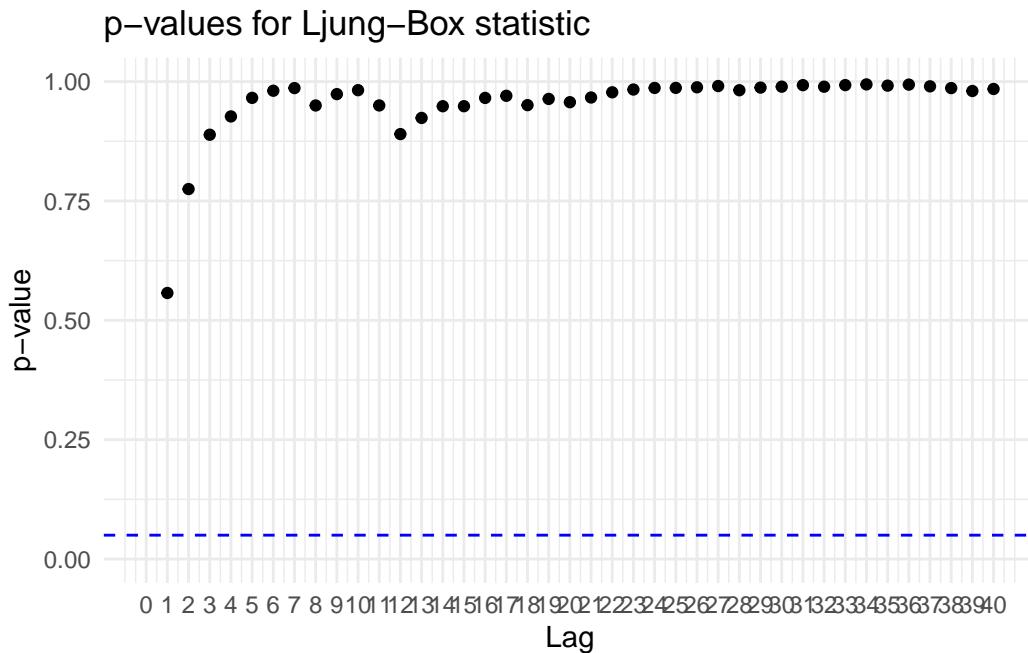
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.03259 on 54 degrees of freedom

Multiple R-squared: 0.5541, Adjusted R-squared: 0.4798

F-statistic: 7.457 on 9 and 54 DF, p-value: 0.0000005507

```
LSTS::Box.Ljung.Test(uecm_model$residuals, lag = 40)
```



```
jarque.test(as.numeric(uecm_model$residuals))
```

Jarque-Bera Normality Test

```
data: as.numeric(uecm_model$residuals)
JB = 0.14977, p-value = 0.9279
alternative hypothesis: greater
```

```
lmtest::bptest(uecm_model)
```

studentized Breusch-Pagan test

```
data: uecm_model
BP = 13.072, df = 9, p-value = 0.1594
```

```
whitestrap::white_test(uecm_model)
```

White's test results

Null hypothesis: Homoskedasticity of the residuals

```

Alternative hypothesis: Heteroskedasticity of the residuals
Test Statistic: 8.69
P-value: 0.012945

```

```
fDMA::archtest(as.numeric(uecm_model$residuals**2), lag = 20)
```

Engle's LM ARCH Test

```

data: as.numeric(uecm_model$residuals^2)
statistic = 16.389, lag = 20, p-value = 0.6922
alternative hypothesis: ARCH effects of order 20 are present

```

```
bounds_f_test(uecm_model, case = 2, alpha = 0.05)[["tab"]]
```

	statistic	Lower-bound I(0)	Upper-bound I(1)	alpha	p.value
F	7.630195	3.620887	4.135148	0.05	0.0009516391

```
bounds_t_test(uecm_model, case = 3, alpha = 0.05)[["tab"]]
```

	statistic	Lower-bound I(0)	Upper-bound I(1)	alpha	p.value
t	-3.697361	-2.860562	-3.224002	0.05	0.01465199

```
multipliers(model)
```

	Term	Estimate	Std. Error	t value	Pr(> t)
1	(Intercept)	3.343593	0.3438624	9.723637	0.000000000000182664
2	log_pe	0.291247	0.0786929	3.701058	0.000504920857463079

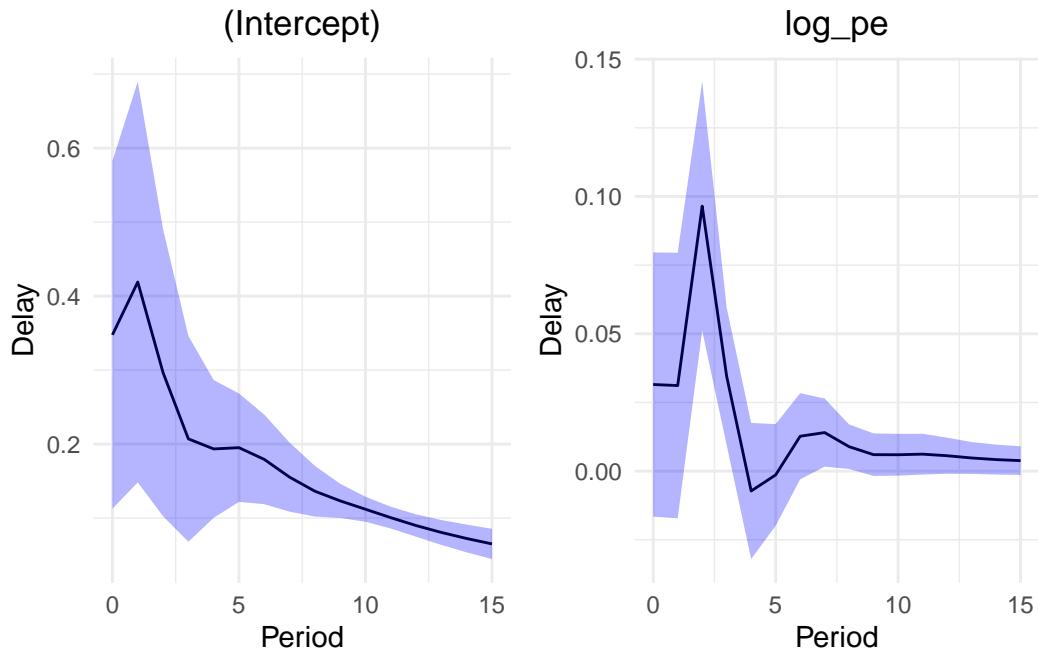
```
multipliers(model, type = "sr")
```

	Term	Estimate	Std. Error	t value	Pr(> t)
1	(Intercept)	0.34760704	0.11988194	2.899578	0.005392771
2	log_pe	0.03154554	0.02454042	1.285453	0.204119844
3	ww2	-0.04015317	0.02303021	-1.743500	0.086937561

```

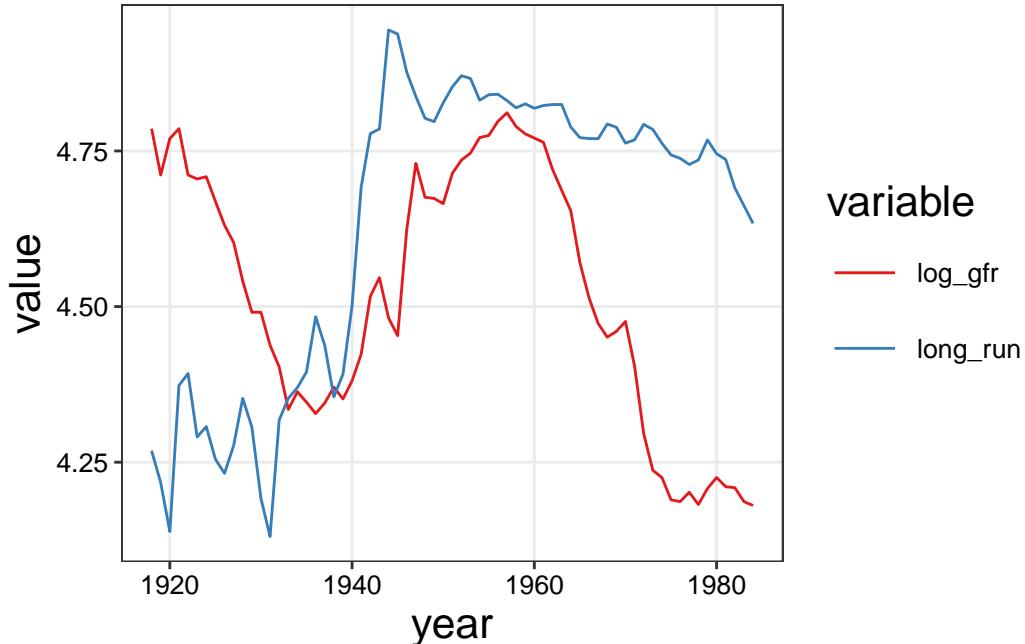
ARDL::multipliers(
  model,
  type = 15,
  se = TRUE
) |>
  ARDL::plot_delay(interval = 0.95)

```



```
long_run_eq <- ARDL::coint_eq(model, case = 2)

tibble(
  log_gfr = as.numeric(model$data$log_gfr),
  long_run = as.numeric(long_run_eq),
  year = time(model$data)
) |>
  pivot_longer(
    cols = !year,
    names_to = "variable",
    values_to = "value"
  ) |>
  ggplot(aes(x = year, y = value, color = variable)) +
  geom_line() +
  scale_color_brewer(palette = "Set1")
```



```
recm_model <- recm(model, case = 3)

summary(recm_model)
```

Time series regression with "zooreg" data:
 Start = 1921, End = 1984

Call:
`dynlm::dynlm(formula = full_formula, data = data, start = start,
 end = end)`

Residuals:

Min	1Q	Median	3Q	Max
-0.070097	-0.020754	0.000676	0.022199	0.077559

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.34761	0.07439	4.672	0.0000196 ***
d(L(log_gfr, 1))	0.30936	0.10189	3.036	0.00366 **
d(L(log_gfr, 2))	-0.29066	0.09979	-2.913	0.00517 **
d(log_pe)	0.03155	0.02386	1.322	0.19161
d(L(log_pe, 1))	-0.00560	0.02204	-0.254	0.80036

```
d(L(log_pe, 2))  0.07225   0.02339   3.089   0.00315 **  
ww2              -0.04015   0.02282  -1.760   0.08403 .  
pill             -0.06309   0.01426  -4.423  0.0000464 ***  
ect              -0.10396   0.02190  -4.748  0.0000151 ***  
---  
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 0.03229 on 55 degrees of freedom

(0 observation effacée parce que manquante)

Multiple R-squared: 0.5541, Adjusted R-squared: 0.4893

F-statistic: 8.544 on 8 and 55 DF, p-value: 0.0000001788