

# 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 (coefficient 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  $i3$  par  $inf$ . Quel sont les ordres choisis ? 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  $i3$  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 `wooldridge` 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  0 ...  
 $ ww2   : int   0  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

```

```

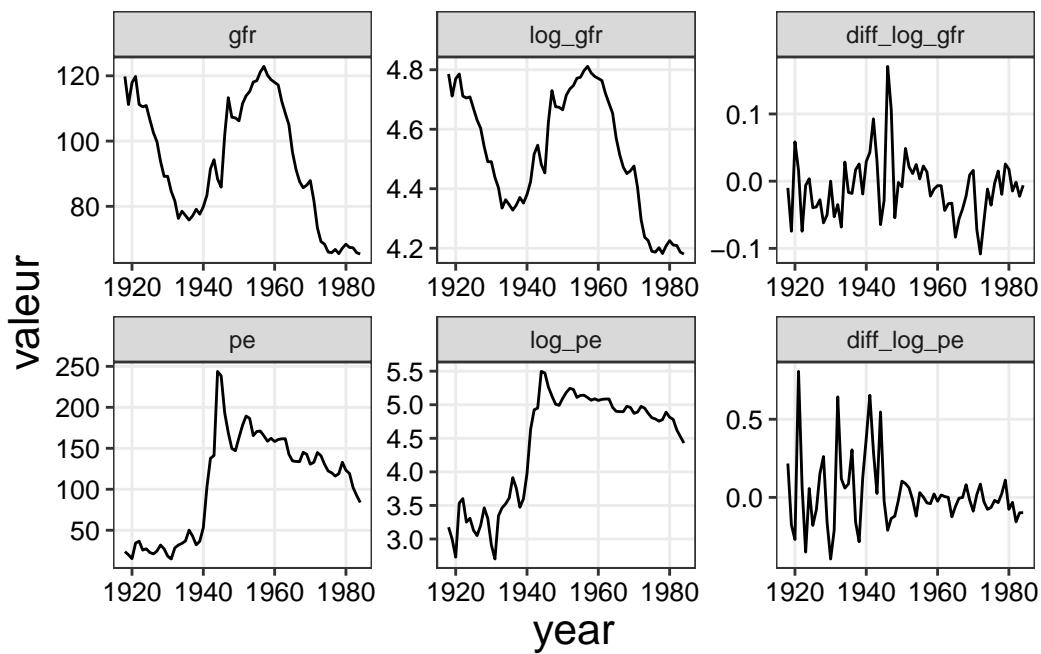
7 1924-01-01 111.  27.3    0    0    4.71  3.31    0.00361  0.0568
8 1925-01-01 107.  22.9    0    0    4.67  3.13   -0.0395  -0.179
9 1926-01-01 103.  21.1    0    0    4.63  3.05   -0.0382  -0.0783
10 1927-01-01 99.8  24.6    0    0    4.60  3.20   -0.0277  0.152
# i 57 more rows

```

```

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	0.00000000000243 ***
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: < 0.00000000000000022

```
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	0.000000000000000232 ***
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 '.' 0.1 ' ' 1

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

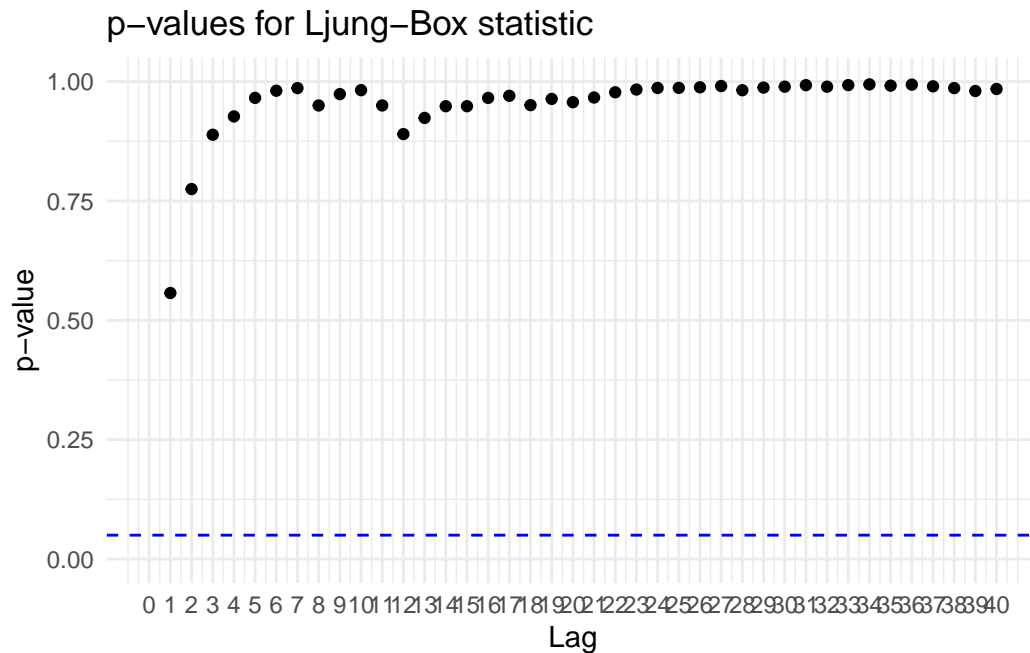
Coefficients:

	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
```

```
whitestrapped::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	

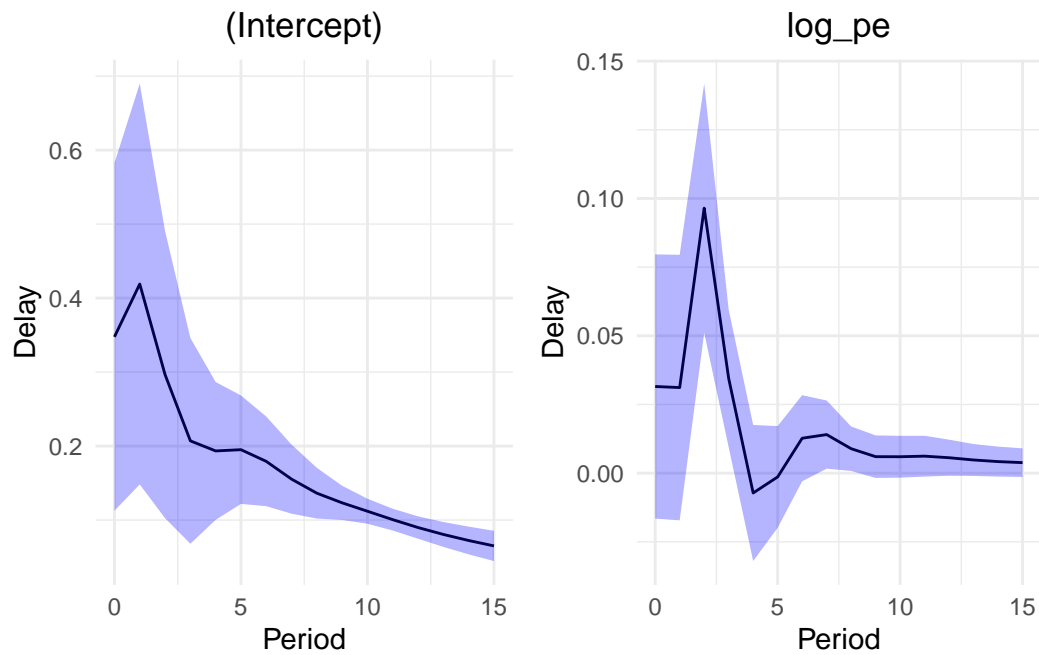
```
multipliers(model)
```

	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")
```

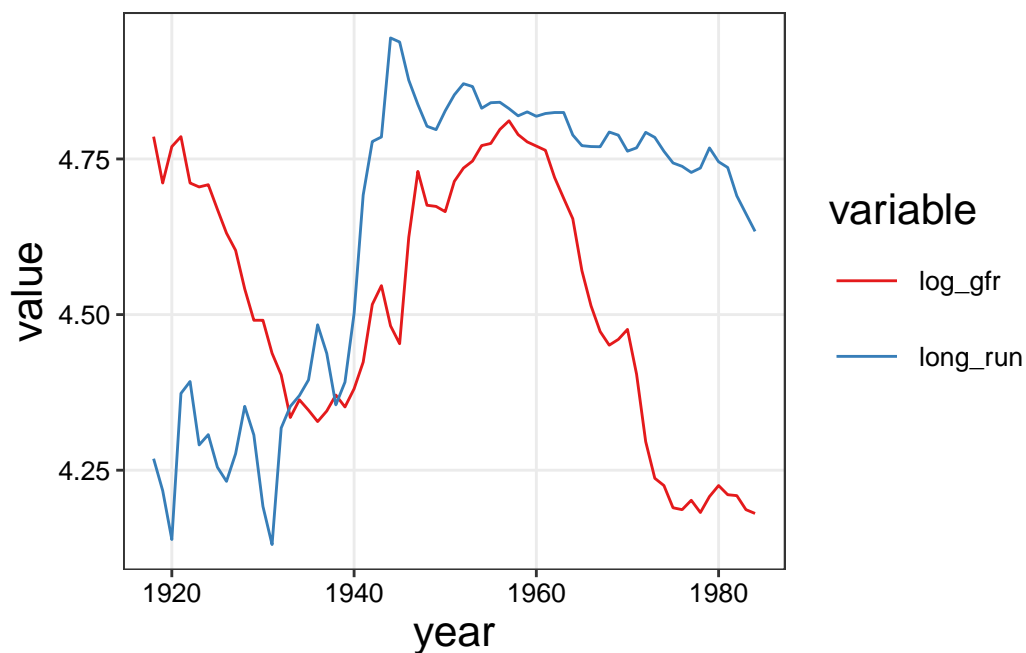
	Term	Estimate	Std. Error	t value	Pr(> t )
1	(Intercept)	0.34760704	0.11988194	2.899578	0.005392771
2	log_pe	0.03154554	0.02454042	1.285453	0.204119844
3	ww2	-0.04015317	0.02303021	-1.743500	0.086937561

```
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