

Introduction to plm

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1 Introduction

The aim of package `plm` is to provide an easy way to estimate panel models. Some panel models may be estimated with package `nlme` (*non-linear mixed effect models*), but not in an intuitive way for an econometrician. `plm` provides methods to read panel data, to estimate a wide range of models and to make some tests. This library is loaded using :

```
> library(plm)
```

This document illustrates the features of `plm`, using data available in package `Ecdat`.

```
> library(Ecdat)
```

These data are used in BALTAGI (2001).

2 Reading data

With `plm`, data are stored in an object of class `pdata.frame`, which is a `data.frame` with additional attributes describing the structure of the data set. A `pdata.frame` may be created from an ordinary `data.frame` using the `pdata.frame` function or from a text file using the `pread.table` function.

2.1 Reading the data from a data.frame

We illustrate the use of the `pdata.frame` function with the `Produc` data :

```
> data(Produc)
> pdata.frame(Produc, "state", "year", "pprod")
```

The `pdata.frame` function has 4 arguments :

- the name of the `data.frame`,
- `id` : the individual index,

- `time` : the time index,
- `name` : the name under which the `pdata.frame` will be stored.

Observations are assumed to be sorted by individuals first, and by period. The third argument is optional, if `NULL` a new variable called `time` is added. The fourth argument is also optional, if `NULL` the `pdata.frame` is stored under the same name as the `data.frame`.

```
> data(Hedonic)
> pdata.frame(Hedonic, "townid")
```

In case of a balanced panel, the `id` may be the number of individuals. In this case, two new variables (called `id` and `time`) are added.

```
> data(Wages)
> pdata.frame(Wages, 595)
```

A description of the data is obtained using the `summary` method :

```
> summary(Hedonic)
```

```
-----
----- Indexes -----
-----
Individual index : townid
Time index      : time
-----

----- Panel Dimensions -----
-----
Unbalanced Panel
Number of Individuals      : 92
Number of Time Obserbations : from 1 to 30
Total Number of Observations : 506
-----

----- Time/Individual Variation -----
-----
no time variation : zn indus rad tax ptratio
-----

----- Descriptive Statistics -----
-----
              mv              crim              zn              indus              chas
Min.   : 8.517   Min.   : 0.00632   Min.   : 0.00   Min.   : 0.46   no :471
1st Qu.: 9.742   1st Qu.: 0.08205   1st Qu.: 0.00   1st Qu.: 5.19   yes: 35
Median : 9.962   Median : 0.25651   Median : 0.00   Median : 9.69
Mean   : 9.942   Mean   : 3.61352   Mean   : 11.36   Mean   :11.14
3rd Qu.:10.127   3rd Qu.: 3.67708   3rd Qu.: 12.50   3rd Qu.:18.10
Max.   :10.820   Max.   :88.97620   Max.   :100.00   Max.   :27.74

              nox              rm              age              dis
Min.   :14.82   Min.   :12.68   Min.   : 2.90   Min.   :0.1219
1st Qu.:20.16   1st Qu.:34.64   1st Qu.: 45.02   1st Qu.:0.7420
```

Median :28.94	Median :38.55	Median : 77.50	Median :1.1655
Mean :32.11	Mean :39.99	Mean : 68.57	Mean :1.1880
3rd Qu.:38.94	3rd Qu.:43.87	3rd Qu.: 94.07	3rd Qu.:1.6464
Max. :75.86	Max. :77.09	Max. :100.00	Max. :2.4954

rad	tax	ptratio	blacks
Min. :0.000	Min. :187.0	Min. :12.60	Min. :0.00032
1st Qu.:1.386	1st Qu.:279.0	1st Qu.:17.40	1st Qu.:0.37538
Median :1.609	Median :330.0	Median :19.05	Median :0.39144
Mean :1.868	Mean :408.2	Mean :18.46	Mean :0.35667
3rd Qu.:3.178	3rd Qu.:666.0	3rd Qu.:20.20	3rd Qu.:0.39623
Max. :3.178	Max. :711.0	Max. :22.00	Max. :0.39690

lstat	townid	time
Min. :-4.0582	29 : 30	1 : 92
1st Qu.: -2.6659	84 : 23	2 : 75
Median : -2.1747	5 : 22	3 : 60
Mean : -2.2342	83 : 19	4 : 50
3rd Qu.: -1.7744	41 : 18	5 : 39
Max. : -0.9684	28 : 15	6 : 33
	(Other):379	(Other):157

The printing consists on four sections :

- `indexes` indicates the names of the index variables,
- `panel dimensions` gives information about the dimension of the panel,
- `Time/individual variation` indicates whether some variables have only individual or time variation,
- `Descriptive statistics` gives descriptive statistics about the variables.

2.2 Reading the data from a text file

`pread.table` reads panel data from a text file, with the following syntax :

```
pread.table("c:/mes documents/essai/mydata.txt",
            "firm", "year", "dataname", header=T, sep=";", dec=",")
```

The arguments of `pread.table` are :

- the text file,
- `id` : the individual index,
- `time` : the time index,

- `name` : the name under which the `pdata.frame` will be stored (if `NULL`, the name of the `pdata.frame` is the name of the file without the path and the extension),
- further arguments that will be passed to `read.table`.

3 Model estimation

A panel model is estimated with the `plm` function.

3.1 Basic use of `plm`

There are two ways to use `plm` : the first one is to estimate a list of models (the default behavior), the second to estimate just one model. In the first case, the estimated models are :

- the fixed effects model (`within`),
- the pooling model (`pooling`),
- the between model (`between`),
- the error components model (`random`).

The basic use of `plm` is to indicate the model formula and the `pdata.frame`¹ :

```
> zz <- plm(log(gsp) ~ log(pcap) + log(pc) + log(emp) + unemp,
+          data = pprod)
```

The result of the estimation is stored in a `plms` object which is a list of 4 estimated models, each of them being objects of class `plm`. Each individual model can be easily extracted :

```
> zz$within <- zz$within
```

A particular model to be estimated may also be indicated by filling the `model` argument of `plm`.

```
> zzra <- plm(log(gsp) ~ log(pcap) + log(pc) + log(emp) + unemp,
+          data = pprod, model = "random")
```

Objects of class `plm` and `plms` have a `print` method.

```
> print(zzra)
```

¹The following example is from BALTAGI (2001), pp. 25–28.

Model Formula: $\log(\text{gsp}) \sim \log(\text{pcap}) + \log(\text{pc}) + \log(\text{emp}) + \text{unemp}$

Coefficients:

(intercept)	$\log(\text{pcap})$	$\log(\text{pc})$	$\log(\text{emp})$	unemp
2.1354110	0.0044386	0.3105484	0.7296705	-0.0061725

There is also a `summary` method :

- for `plms` objects, coefficients and standard errors of the fixed effects and the error components models are printed,
- for `plm` object, the table of coefficients and some statistics are printed.

> `summary(zz)`

```

-----
----- Model Description -----
Oneway (individual) effect

Model Formula      : log(gsp) ~ log(pcap) + log(pc) + log(emp) +
                      unemp

-----
----- Panel Dimensions -----
Balanced Panel
Number of Individuals      : 48
Number of Time Observations : 17
Total Number of Observations : 816

-----
----- Coefficients -----
              within      wse      random      rse
(intercept)          .          .  2.13541100 0.1335
log(pcap)    -0.02614965 0.02900158 0.00443859 0.0234
log(pc)       0.29200693 0.02511967 0.31054843 0.0198
log(emp)      0.76815947 0.03009174 0.72967053 0.0249
unemp        -0.00529774 0.00098873 -0.00617247 0.0009

-----
----- Tests -----
Hausman Test      : chi2(4) = 9.525416 (p.value=0.04922762)
F Test            : F(47,764) = 75.8204 (p.value=0)
Lagrange Multiplier Test : chi2(1) = 4134.961 (p.value=0)

```

> `summary(zzra)`

```

-----
----- Model Description -----
Oneway (individual) effect

```

Random Effect Model (Swamy-Arora's transformation)
Model Formula : log(gsp) ~ log(pcap) + log(pc) +
log(emp) + unemp

----- Panel Dimensions -----

Balanced Panel

Number of Individuals : 48
Number of Time Observations : 17
Total Number of Observations : 816

----- Effects -----

	var	std.dev	share
idiosyncratic	0.0014544	0.0381371	0.1754
individual	0.0068377	0.0826905	0.8246
theta	: 0.88884		

----- Residuals -----

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-1.07e-01	-2.46e-02	-2.37e-03	-9.93e-19	2.17e-02	2.00e-01

----- Coefficients -----

	Estimate	Std. Error	z-value	Pr(> z)
(intercept)	2.13541100	0.13346149	16.0002	< 2.2e-16 ***
log(pcap)	0.00443859	0.02341732	0.1895	0.8497
log(pc)	0.31054843	0.01980475	15.6805	< 2.2e-16 ***
log(emp)	0.72967053	0.02492022	29.2803	< 2.2e-16 ***
unemp	-0.00617247	0.00090728	-6.8033	1.023e-11 ***

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

----- Overall Statistics -----

Total Sum of Squares	: 29.209
Sum of Squares Residuals	: 1.1879
Rsq	: 0.95933
F	: 4782.77
P(F>0)	: 8.76231e-08

For a random model, the summary method gives information about the variance of the components of the errors.

plm's can be updated using the update method :

```
> zzwithmod <- update(zzwith, . ~ . - unemp - log(emp) + emp)
> zzmod <- update(zz, . ~ . - unemp - log(emp) + emp)
> summary(zzwithmod)
```

```

-----
----- Model Description -----
Oneway (individual) effect

Model Formula      : log(gsp) ~ log(pcap) + log(pc) + emp
-----
----- Panel Dimensions -----
Balanced Panel
Number of Individuals      : 48
Number of Time Observations : 17
Total Number of Observations : 816
-----
----- Coefficients -----
              within      wse      random      rse
(intercept)      .      . 7.1982e-01      0.1846
log(pcap)  1.7888e-01 4.0690e-02 3.4357e-01      0.0322
log(pc)    6.9975e-01 2.9154e-02 6.0369e-01      0.0256
emp        3.7909e-05 8.7824e-06 5.0924e-05 8.218e-06
-----
----- Tests -----
Hausman Test      : chi2(3) = 80.35868 (p.value=0)
F Test           : F(47,765) = 101.9109 (p.value=0)
Lagrange Multiplier Test : chi2(1) = 4355.292 (p.value=0)
-----

```

Fixed effects may be extracted easily from a `plms` or a `plm` object using `FE` :

```
> FE(zzmod)
```

ALABAMA	ARIZONA	ARKANSAS	CALIFORNIA	COLORADO
1.1717531	1.3062389	1.1877004	1.6191982	1.4582149
CONNECTICUT	DELAWARE	FLORIDA	GEORGIA	IDAHO
1.7060341	1.2035746	1.5564969	1.4460171	1.1002049
ILLINOIS	INDIANA	IOWA	KANSAS	KENTUCKY
1.5496106	1.3451714	1.2323038	1.1735476	1.3492604
LOUISIANA	MAINE	MARYLAND	MASSACHUSETTS	MICHIGAN
1.1652834	1.2659480	1.6011871	1.7384231	1.5290312
MINNESOTA	MISSISSIPPI	MISSOURI	MONTANA	NEBRASKA
1.3654287	1.1545345	1.4809262	0.7960951	1.0905033
NEVADA	NEW_HAMPSHIRE	NEW_JERSEY	NEW_MEXICO	NEW_YORK
1.0627992	1.4138235	1.7420589	1.0925399	1.6694387
NORTH_CAROLINA	NORTH_DAKOTA	OHIO	OKLAHOMA	OREGON
1.5048751	0.7663694	1.4985974	1.2784660	1.3345094
PENNSYLVANIA	RHODE_ISLAND	SOUTH_CAROLINA	SOUTH_DAKOTA	TENNESSE
1.4972243	1.5948140	1.2344011	0.8705826	1.3123010
TEXAS	UTAH	VERMONT	VIRGINIA	WASHINGTON
1.3230328	1.2464927	1.1804339	1.6175357	1.3492922

WEST_VIRGINIA	WISCONSIN	WYOMING
1.0129871	1.4860561	0.7842841

3.2 Options for the random effect model

The random effect model is obtained as a linear estimation on quasi-differentiated data. The parameter of this transformation is obtained using preliminary estimations. Four estimators of this parameter are available, depending on the value of the argument `theta.method` :

- `swar` : from SWAMY and ARORA (1972), the default value,
- `walhus` : from WALLACE and HUSSAIN (1969),
- `amemiya` : from AMEMIYA (1971),
- `nerlove` : from NERLOVE (1971).

For example, to use the `amemiya` estimator :

```
> zzra <- plm(log(gsp) ~ log(pcap) + log(pc) + log(emp) + unemp,
+ data = pprod, model = "random", theta.method = "amemiya")
```

3.3 Choosing the effects

The default behavior of `plm` is to introduce individual effects. Using the `effect` argument, one may also introduce :

- time effects (`effect="time"`),
- individual and time effects (`effect="twoways"`).

For example, to estimate a two-ways effect model for the `Grunfeld` data :

```
> data(Grunfeld)
> pdata.frame(Grunfeld, "firm", "year")
> z <- plm(inv ~ value + capital, data = Grunfeld, effect = "twoways",
+ theta.method = "amemiya")
> summary(z$random)
```

```
-----
----- Model Description -----
Twoways effects
Random Effect Model (Swamy-Arora's transformation)
Model Formula      : inv ~ value + capital
-----
----- Panel Dimensions -----
Balanced Panel
Number of Individuals      : 10
```



```
Number of Time Obserbations : 20
Total Number of Observations : 200
```

```
-----
                        Effects -----
      var  std.dev  share
idiosyncratic 2675.426  51.725 0.2738
individual    7095.252  84.233 0.7262
time          0.000    0.000 0.0000
theta : 0.86397 (id) 0 (time) 0 (total)
-----

                        Residuals -----
      Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
-1.77e+02 -1.98e+01  4.60e+00  8.77e-16  1.95e+01  2.53e+02
-----

                        Coefficients -----
      Estimate Std. Error z-value Pr(>|z|)
(intercept) -57.865377  29.393359 -1.9687  0.04899 *
value        0.109790   0.010528 10.4285 < 2e-16 ***
capital      0.308190   0.017171 17.9483 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

-----
                        Overall Statistics -----
Total Sum of Squares      : 2376000
Sum of Squares Residuals  : 547910
Rsqr                      : 0.7694
F                          : 328.647
P(F>0)                    : 0.0030381
-----
```

In the “effects” section of the result is printed now the variance of the three elements of the error term and the three parameters used in the transformation.

The two-ways effect model is for the moment only available for balanced panels.

3.4 Hausman–Taylor’s model

HAUSMAN–TAYLOR’s model may be estimated with `plm` by equating the `model` argument to “ht” and filling the second argument `instruments` with a formula indicating the variables used as instruments.

```
> data(Wages)
> pdata.frame(Wages, 595)
> form = lwage ~ wks + south + smsa + married + exp + I(exp^2) +
+       bluecol + ind + union + sex + black + ed
> ht = plm(form, ~sex + black + bluecol + south + smsa + ind, data = Wages,
```

```
+      model = "ht")
> summary(ht)
```

```
-----
----- Model Description -----
Oneway (individual) effect
Hausman-Taylor Model
Model Formula      : lwage ~ wks + south + smsa + married +
                    : exp + I(exp^2) + bluecol + ind +
                    : union + sex + black + ed
Instrumental Variables : ~sex + black + bluecol + south +
                    : smsa + ind
Time--Varying Variables
  exogenous variables : bluecolyes,southyes,smsayes,ind
  endogenous variables : wks,marriedyes,exp,I(exp^2),unionyes
Time--Invariant Variables
  exogenous variables : sexmale,blackyes
  endogenous variables : ed
-----
----- Panel Dimensions -----
Balanced Panel
Number of Individuals      : 595
Number of Time Obserbations : 7
Total Number of Observations : 4165
-----
----- Effects -----
              var std.dev share
idiosyncratic 0.023044 0.151803 0.0253
individual    0.886993 0.941803 0.9747
theta       : 0.93919
-----
----- Residuals -----
      Min.    1st Qu.    Median      Mean    3rd Qu.      Max.
-1.92e+00 -7.07e-02  6.57e-03 -2.46e-17  7.97e-02  2.03e+00
-----
----- Coefficients -----
              Estimate Std. Error z-value Pr(>|z|)
(intercept)  2.7818e+00 3.0768e-01  9.0411 < 2.2e-16 ***
wks          8.3740e-04 5.9981e-04  1.3961  0.16268
southyes     7.4398e-03 3.1959e-02  0.2328  0.81592
smsayes      -4.1833e-02 1.8960e-02 -2.2064  0.02736 *
marriedyes   -2.9851e-02 1.8982e-02 -1.5726  0.11582
exp          1.1313e-01 2.4713e-03 45.7795 < 2.2e-16 ***
I(exp^2)     -4.1886e-04 5.4605e-05 -7.6709 1.710e-14 ***
bluecolyes   -2.0705e-02 1.3783e-02 -1.5022  0.13304
ind          1.3604e-02 1.5239e-02  0.8927  0.37202
```

```

unionyes      3.2771e-02  1.4910e-02  2.1979   0.02796 *
sexmale       1.3092e-01  1.2667e-01  1.0335   0.30135
blackyes      -2.8575e-01  1.5572e-01 -1.8350   0.06651 .
ed            1.3794e-01  2.1251e-02  6.4912  8.518e-11 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

-----
----- Overall Statistics -----
Total Sum of Squares      : 243.04
Sum of Squares Residuals  : 95.947
Rsq                       : 0.60522
F                         : 530.318
P(F>0)                   : 2.88658e-15
-----

```

3.5 Instrumental variables estimation

One or all of the models may be estimated using instrumental variables by indicating the list of the instrumental variables. This can be done using one of the two following techniques :

- specifying the total list of instruments (using the `instruments` argument of `plm`),
- specifying, on the one hand the external instruments in the argument `instrument` and on the other hand the variables of the model that are assumed to be endogenous in the argument `endog`.

The instrumental variables estimator used may be indicated with the `inst.method` argument :

- `bvk`, from BALESTRA et VARADHARAJAN–KRISHNAKUMAR (1987), the default value,
- `baltagi`, from BALTAGI (1981).

We illustrate instrumental variables estimation with the `Crime` data². The same estimation is done using the first syntax (`cr1`) and the second (`cr2`). The `prbarr` and `polpc` variables are assumed to be endogenous and there are two external instruments `taxpc` and `mix` :

```

> data(Crime)
> pdata.frame(Crime, "county", "year")
> form = log(crmrte) ~ log(prbarr) + log(polpc) + log(prbconv) +
+   log(prbpris) + log(avgsen) + log(density) + log(wcon) + log(wtuc) +
+   log(wtrd) + log(wfir) + log(wser) + log(wmfg) + log(wfed) +

```

²See BALTAGI (2001), pp.119–120.

```

+      log(wsta) + log(wloc) + log(pctymle) + log(pctmin) + region +
+      smsa + year
> inst = ~log(prbconv) + log(prbpris) + log(avgsen) + log(density) +
+      log(wcon) + log(wtuc) + log(wtrd) + log(wfir) + log(wser) +
+      log(wmfg) + log(wfed) + log(wsta) + log(wloc) + log(pctymle) +
+      log(pctmin) + region + smsa + log(taxpc) + log(mix) + year
> inst2 = ~log(taxpc) + log(mix)
> endog = ~log(prbarr) + log(polpc)
> cr = plm(form, data = Crime)
> cr1 = plm(form, inst, data = Crime)
> cr2 = plm(form, inst2, endog, data = Crime)
> summary(cr2$random)

-----
----- Model Description -----
Oneway (individual) effect
Random Effect Model (Swamy-Arora's transformation)
Instrumental variable estimation (Balestra-Varadharajan-Krishnakumar's transformation)
Model Formula          : log(crmrte) ~ log(prbarr) + log(polpc) +
                        log(prbconv) + log(prbpris) +
                        log(avgsen) + log(density) +
                        log(wcon) + log(wtuc) + log(wtrd) +
                        log(wfir) + log(wser) + log(wmfg) +
                        log(wfed) + log(wsta) + log(wloc) +
                        log(pctymle) + log(pctmin) +
                        region + smsa + year
Endogenous Variables    : ~log(prbarr) + log(polpc)
Instrumental Variables   : ~log(taxpc) + log(mix)

-----
----- Panel Dimensions -----
Balanced Panel
Number of Individuals    : 90
Number of Time Obserbations : 7
Total Number of Observations : 630

-----
----- Effects -----
var  std.dev share
idiosyncratic 0.022269 0.149228 0.326
individual    0.046036 0.214561 0.674
theta       : 0.74576

-----
----- Residuals -----
Min.    1st Qu.    Median      Mean    3rd Qu.      Max.
-5.02e+00 -4.76e-01  2.73e-02  7.11e-16  5.26e-01  3.19e+00

-----
----- Coefficients -----

```

	Estimate	Std. Error	z-value	Pr(> z)
(intercept)	-0.4538241	1.7029840	-0.2665	0.789864
log(prbarr)	-0.4141200	0.2210540	-1.8734	0.061015 .
log(polpc)	0.5049285	0.2277811	2.2167	0.026642 *
log(prbconv)	-0.3432383	0.1324679	-2.5911	0.009567 **
log(prbpris)	-0.1900437	0.0733420	-2.5912	0.009564 **
log(avgsen)	-0.0064374	0.0289406	-0.2224	0.823977
log(density)	0.4343519	0.0711528	6.1045	1.031e-09 ***
log(wcon)	-0.0042963	0.0414225	-0.1037	0.917392
log(wtuc)	0.0444572	0.0215449	2.0635	0.039068 *
log(wtrd)	-0.0085626	0.0419822	-0.2040	0.838387
log(wfir)	-0.0040302	0.0294565	-0.1368	0.891175
log(wser)	0.0105604	0.0215822	0.4893	0.624620
log(wmfg)	-0.2017917	0.0839423	-2.4039	0.016220 *
log(wfed)	-0.2134634	0.2151074	-0.9924	0.321023
log(wsta)	-0.0601083	0.1203146	-0.4996	0.617362
log(wloc)	0.1835137	0.1396721	1.3139	0.188884
log(pctymle)	-0.1458448	0.2268137	-0.6430	0.520214
log(pctmin)	0.1948760	0.0459409	4.2419	2.217e-05 ***
regionwest	-0.2281780	0.1010317	-2.2585	0.023916 *
regioncentral	-0.1987675	0.0607510	-3.2718	0.001068 **
smsayes	-0.2595423	0.1499780	-1.7305	0.083535 .
year82	0.0132140	0.0299923	0.4406	0.659518
year83	-0.0847676	0.0320008	-2.6489	0.008075 **
year84	-0.1062004	0.0387893	-2.7379	0.006184 **
year85	-0.0977398	0.0511685	-1.9102	0.056113 .
year86	-0.0719390	0.0605821	-1.1875	0.235045
year87	-0.0396520	0.0758537	-0.5227	0.601153

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

```

-----
Overall Statistics
-----
Total Sum of Squares      : 1354.7
Sum of Squares Residuals  : 557.64
Rsquared                   : 0.58836
F                          : 33.1494
P(F>0)                    : 7.77156e-16
-----

```

3.6 Variable coefficients model

If there is enough time observations for each individual, the model may be estimate for each individual. The `nopool` function provide this kind of estimation. It can be done using :

- directly the `nopool` function,

- `plm` with the argument `np` fixed to `TRUE`.

With the `Grunfeld` data, we get :

```
> znp <- nopool(inv ~ value + capital, data = Grunfeld)

or

> z <- plm(inv ~ value + capital, data = Grunfeld, np = TRUE)
> znp <- z$nopool
> print(znp)
```

	(intercept)	value	capital
1	-149.7824533	0.119280833	0.371444807
2	-49.1983219	0.174856015	0.389641889
3	-9.9563065	0.026551189	0.151693870
4	-6.1899605	0.077947821	0.315718185
5	22.7071160	0.162377704	0.003101737
6	-8.6855434	0.131454842	0.085374274
7	-4.4995344	0.087527198	0.123781407
8	-0.5093902	0.052894126	0.092406492
9	-7.7228371	0.075387943	0.082103558
10	0.1615186	0.004573432	0.437369190

```
> summary(znp)
```

	(intercept)	value	capital
Min.	:-149.782	Min. :0.004573	Min. :0.003102
1st Qu.:	-9.639	1st Qu.:0.058518	1st Qu.:0.087132
Median :	-6.956	Median :0.082738	Median :0.137738
Mean :	-21.368	Mean :0.091285	Mean :0.205264
3rd Qu.:	-1.507	3rd Qu.:0.128411	3rd Qu.:0.357513
Max. :	22.707	Max. :0.174856	Max. :0.437369

The result is an object of class `nopool`. The `print` method presents the coefficients estimated for each individual. The `summary` method gives descriptive statistics for these coefficients.

3.7 Unbalanced panel

`plm` offers limited support for unbalanced panels. The following example is based on the `Hedonic` data³:

```
> form = mv ~ crim + zn + indus + chas + nox + rm + age + dis +
+       rad + tax + ptratio + blacks + lstat
> ba = plm(form, data = Hedonic)
> summary(ba$random)
```

³See BALTAGI (2001), p. 174.

```

-----
----- Model Description -----
Oneway (individual) effect
Random Effect Model (Swamy-Arora's transformation)
Model Formula      : mv ~ crim + zn + indus + chas + nox +
                    rm + age + dis + rad + tax +
                    ptratio + blacks + lstat

-----
----- Panel Dimensions -----
Unbalanced Panel
Number of Individuals      : 92
Number of Time Observations : from 1 to 30
Total Number of Observations : 506

-----
----- Effects -----
var  std.dev share
idiosyncratic 0.016965 0.130249 0.502
individual    0.016832 0.129738 0.498
theta :
      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
      0.2915 0.5904 0.6655 0.6499 0.7447 0.8197

-----
----- Residuals -----
      Min.    1st Qu.    Median    Mean    3rd Qu.    Max.
-0.641000 -0.066100 -0.000519 -0.001990 0.069800 0.527000

-----
----- Coefficients -----
      Estimate Std. Error z-value Pr(>|z|)
(intercept)  9.6778e+00 2.0714e-01 46.7207 < 2.2e-16 ***
crim         -7.2338e-03 1.0346e-03 -6.9921 2.707e-12 ***
zn           3.9575e-05 6.8778e-04 0.0575 0.9541153
indus        2.0794e-03 4.3403e-03 0.4791 0.6318706
chasyes      -1.0591e-02 2.8960e-02 -0.3657 0.7145720
nox          -5.8630e-03 1.2455e-03 -4.7074 2.509e-06 ***
rm           9.1773e-03 1.1792e-03 7.7828 7.105e-15 ***
age          -9.2715e-04 4.6468e-04 -1.9952 0.0460159 *
dis          -1.3288e-01 4.5683e-02 -2.9088 0.0036279 **
rad           9.6863e-02 2.8350e-02 3.4168 0.0006337 ***
tax          -3.7472e-04 1.8902e-04 -1.9824 0.0474298 *
ptratio      -2.9723e-02 9.7538e-03 -3.0473 0.0023089 **
blacks       5.7506e-01 1.0103e-01 5.6920 1.256e-08 ***
lstat        -2.8514e-01 2.3855e-02 -11.9533 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

-----
----- Overall Statistics -----

```

```

Total Sum of Squares      : 893.08
Sum of Squares Residuals  : 8.6843
Rsq                       : 0.99028
F                         : 3854.18
P(F>0)                   : 0

```

4 Tests

4.1 Tests of poolability

`pooltest` tests the hypothesis that the same coefficients apply to each individual. It is a standard F test, based on the comparison of a model obtained for the full sample and a model based on the estimation of an equation for each individual. The main argument of `pooltest` is a `plm` object. If the model has been estimated with the argument `np=F`, one has to indicate a second argument of class `nopool`. A third argument `effect` should be fixed to `FALSE` if the intercepts are assumed to be identical (the default value) or `TRUE` if not⁴.

```

> form = inv ~ value + capital
> znp = nopool(form, data = Grunfeld)
> zplm = plm(form, data = Grunfeld)
> pooltest(zplm, znp)

```

F statistic

```

data:  zplm
F = 27.7486, df1 = 27, df2 = 170, p-value < 2.2e-16

```

```

> pooltest(zplm, znp, effect = T)

```

F statistic

```

data:  zplm
F = 5.7805, df1 = 18, df2 = 170, p-value = 1.219e-10

```

```

> z = plm(form, data = Grunfeld, effect = "time", np = TRUE)
> pooltest(z, effect = F)

```

F statistic

```

data:  z
F = 1.1204, df1 = 57, df2 = 140, p-value = 0.2928

```

⁴The following examples are from BALTAGI (2001), pp. 57–58.

4.2 Tests for individual and time effects

4.2.1 Lagrange multiplier tests

`plmtest` implements tests of individual or/and time effects based on the results of the pooling model. Its main argument is a `plm` object (the result of a pooling model) or a `plms` object.

Two additional arguments can be added to indicate the kind of test to be computed. The argument `type` is whether :

- `bp` : BREUSCH–PAGAN (1980), the default value,
- `honda` : HONDA (1985),
- `kw` : KING and WU (1997).

The effects tested are indicated with the `effect` argument :

- `individual` for individual effects (the default value),
- `time` for time effects,
- `twoways` for individuals and time effects.

Some examples of the use of `plmtest` are shown below⁵:

```
> library(Ecdat)
> g <- plm(inv ~ value + capital, data = Grunfeld)
> plmtest(g)
```

Lagrange Multiplier Test - individual effects (Breush-Pagan)

```
data:  g
chi2 = 798.1615, df = 1, p-value < 2.2e-16
```

```
> plmtest(g, effect = "time")
```

Lagrange Multiplier Test - time effects (Breush-Pagan)

```
data:  g
chi2 = 6.4539, df = 1, p-value = 0.01107
```

```
> plmtest(g, type = "honda")
```

Lagrange Multiplier Test - individual effects (Honda)

```
data:  g
normal = 28.2518, p-value < 2.2e-16
```

⁵See BALTAGI (2001), p. 65.

```
> plmtest(g, type = "ghm", effect = "twoways")

Lagrange Multiplier Test - two-ways effects (Gourierroux, Holly and
Monfort)

data:  g
chi2 = 798.1615, df = 2, p-value < 2.2e-16

> plmtest(g, type = "kw", effect = "twoways")

Lagrange Multiplier Test - two-ways effects (King and Wu)

data:  g
normal = 21.8322, df = 2, p-value < 2.2e-16
```

4.2.2 F tests

`pFtest` computes F tests of effects based on the comparison of the `within` and the pooling models. Its arguments are whether a `plms` object or two `plm` objects (the results of a pooling and a within model). Some examples of the use of `pFtest` are shown below⁶:

```
> library(Ecdat)
> gi <- plm(inv ~ value + capital, data = Grunfeld)
> gt <- plm(inv ~ value + capital, data = Grunfeld, effect = "time")
> gd <- plm(inv ~ value + capital, data = Grunfeld, effect = "twoways")
> pFtest(gi)
```

F test for effects

```
data:  gi
F = 49.1766, df1 = 9, df2 = 188, p-value < 2.2e-16

> pFtest(gi$within, gi$pooling)
```

F test for effects

```
data:  gi$within and gi$pooling
F = 49.1766, df1 = 9, df2 = 188, p-value < 2.2e-16

> pFtest(gt)
```

F test for effects

```
data:  gt
F = 0.5229, df1 = 9, df2 = 188, p-value = 0.8569
```

⁶Voir BALTAGI (2001), p. 65.

```
> pFtest(gd)

      F test for effects

data:  gd
F = 17.4031, df1 = 28, df2 = 169, p-value < 2.2e-16
```

4.3 Hausman's test

`phtest` computes the HAUSMAN's test which is based on the comparison of two models. It's main argument may be :

- a `plms` object. In this case, the two models used in the test are the `within` and the `random` models (the most usual case with panel data),
- two `plm` objects.

Some examples of the use of `phtest` are shown below ⁷:

```
> g <- plm(inv ~ value + capital, data = Grunfeld)
> phtest(g)
```

Hausman Test

```
data:  g
chi2 = 2.3304, df = 2, p-value = 0.3119
```

```
> phtest(g$between, g$random)
```

Hausman Test

```
data:  g$between and g$random
chi2 = 2.1314, df = 3, p-value = 0.5456
```

5 Bibliographie

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⁷See BALTAGI (2001), p. 71.

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