

Model for pooled data

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Data Agriculture climate

```
library(readxl)
library(stargazer)

Data_AgricultureClimate <- read_excel("~/Videos/Inverno 2019/Aula 4/Data_AgricultureClimate.xls")

head(Data_AgricultureClimate)
```

```
## # A tibble: 6 x 7
##   ano periodo   vttotal areatotal tempano precano   munic
##   <dbl>   <dbl>   <dbl>     <dbl>   <dbl>   <dbl>   <dbl>
## 1 1990     0 1264259.  1388970   22.9   1316. 3500105
## 2 1991     0  737240.  1235817   23.2   1415. 3500105
## 3 1992     0 1381728.  1059597   22.7   1390. 3500105
## 4 1993     0 1547279.   941193   23.2   1320. 3500105
## 5 1994     0 1383900.   991584   23.6   1160. 3500105
## 6 1995     0 1632009.   972081   23.6   1492. 3500105
```

```
summary(Data_AgricultureClimate)
```

```
##      ano      periodo      vttotal      areatotal
## Min.   :1990   Min.   :0.00   Min.   :      4   Min.   :      0
## 1st Qu.:1996   1st Qu.:0.00   1st Qu.: 591117   1st Qu.: 227997
## Median :2002   Median :0.00   Median : 1598384   Median : 649044
## Mean   :2002   Mean   :0.36   Mean   : 2851582   Mean   : 1191138
## 3rd Qu.:2008   3rd Qu.:1.00   3rd Qu.: 3598698   3rd Qu.: 1487277
## Max.   :2014   Max.   :1.00   Max.   :45904552   Max.   :14584086
##
##      NA's      :1552      NA's      :375
##      tempano      precano      munic
## Min.   :14.34   Min.   : 155.3   Min.   :3500105
## 1st Qu.:20.53   1st Qu.:1226.1   1st Qu.:3514576
## Median :21.59   Median :1373.8   Median :3528552
## Mean   :21.79   Mean   :1406.4   Mean   :3528625
## 3rd Qu.:23.27   3rd Qu.:1542.8   3rd Qu.:3542753
## Max.   :25.19   Max.   :4692.0   Max.   :3557204
##
```

Regression 1

```
loglinear = lm(log(vtotal) ~ log(areatotal) + (tempano) + (precano),
               data = Data_AgricultureClimate)
```

Regression 2

```
loglinear2 = lm(log(vtotal) ~ log(areatotal) + (tempano) + (precano) + periodo,
               data = Data_AgricultureClimate)
```

Mudança no coefficients

```
int_areatotal = Data_AgricultureClimate$areatotal*Data_AgricultureClimate$periodo

loglinear3 = lm(log(vtotal) ~ log(areatotal) + (tempano) + (precano) + periodo
               + (int_areatotal),
               data = Data_AgricultureClimate)

stargazer(loglinear, loglinear2, loglinear3, type = "text", digits = 5,
          column.labels = c("", "", ""), keep.stat = c('n', 'rsq', 'adj.rsq', 'f'),
          out = "mrd.txt")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               log(vtotal)
##
##                               (1)                (2)                (3)
## -----
## log(areatotal)          0.95245***          0.93920***          0.94798***
##                        (0.00375)          (0.00360)          (0.00401)
##
## tempano                 -0.01594***          -0.02095***          -0.02201***
##                        (0.00336)          (0.00320)          (0.00321)
##
## precano                 0.00004*            0.00002            0.00002
##                        (0.00002)          (0.00002)          (0.00002)
##
## periodo                 0.37914***          0.37914***          0.41733***
##                        (0.01069)          (0.01069)          (0.01316)
##
## int_areatotal           -0.0000000***       -0.0000000***       -0.0000000***
##                        (0.00000)          (0.00000)          (0.00000)
##
## Constant                1.70869***          1.88977***          1.79947***
##                        (0.09271)          (0.08854)          (0.09031)
## -----
## Observations            12,588              12,588              12,588
## R2                      0.83962              0.85421              0.85449
## Adjusted R2             0.83958              0.85416              0.85444
## F Statistic             21,960.18000*** (df = 3; 12584) 18,431.20000*** (df = 4; 12583) 14,777.63000*** (df =
## =====
## Note:                                                            *p<0.1; **p<0.05;
```

Data Panel

L'objectif des données de panel est donc de contrôler les variables que la régression linéaire ne saura pas réaliser. Plusieurs variables pour éviter les biais d'informations dans la régression. On peut donc créer une variable binaire pour contrôler les scénarios.

```
library(plm)

panelregre = plm(log(vtotal) ~ log(areatotal) + tempano + precano, index = c("munic", "ano"),
                 model = "within", data = Data_AgricultureClimate)

stargazer(panelregre, type = "text", digits = 5, column.labels = c(""),
           keep.stat = NULL, out = "mrd.txt")

##
## =====
##                      Dependent variable:
##                      -----
##                      log(vtotal)
##
## -----
## log(areatotal)          0.93367***
##                      (0.00964)
##
## tempano                 0.18908***
##                      (0.00733)
##
## precano                -0.00006***
##                      (0.00002)
##
## -----
## Observations              12,588
## R2                       0.44493
## Adjusted R2              0.41951
## F Statistic    3,215.85900*** (df = 3; 12036)
## =====
## Note:          *p<0.1; **p<0.05; ***p<0.01
panelregre2 = plm(log(vtotal) ~ log(areatotal) + tempano + precano, index = c("munic", "ano"),
                  effect = "twoways", model = "within", data = Data_AgricultureClimate)

stargazer(panelregre, panelregre2, type = "text", digits = 5, column.labels = c("", ""),
           keep.stat = NULL, out = "mrd.txt")

##
## =====
##                      Dependent variable:
##                      -----
##                      log(vtotal)
##
##                      (1)                      (2)
## -----
## log(areatotal)          0.93367***          0.86365***
##                      (0.00964)          (0.00763)
##
```

```
## tempano          0.18908***          0.00611
##                  (0.00733)          (0.00700)
##
## precano          -0.00006***          0.00003*
##                  (0.00002)          (0.00002)
##
## -----
## Observations      12,588            12,588
## R2                0.44493            0.51927
## Adjusted R2       0.41951            0.49626
## F Statistic       3,215.85900*** (df = 3; 12036) 4,325.06500*** (df = 3; 12012)
## =====
## Note:                                *p<0.1; **p<0.05; ***p<0.01
```

Random Effects and Hausman

L'avantage des données en panel ce que le degré de liberté augmente $p(n - k) = s.e$, mais on perd l'efficacité avec $E(\hat{\beta}) = 0$. En donnée de panel, on inclut *dummy* pour capter toute les informations constantes au long du temps. C'est à dire on contrôle toute les variables omis. (voir aula 4, p. 18, slides). n égal recette = taille du modèle = amostra et k égal numéro de variables.

Dans le modèle *Random Effects* e_{it} mesure la variance entre la ligne de régression et C_i qui est la variable *dummy*, mesure la variance générale entre toute le ligne (aula 5, p. 3, slides).

```
random1 = plm(log(vtotal) ~ log(areatotal) + tempano, data = Data_AgricultureClimate,
              index = c("munic", "ano"), model = "random")

random2 = plm(log(vtotal) ~ log(areatotal) + tempano + as.factor(ano),
              data = Data_AgricultureClimate, index = c("munic", "ano"), model = "random")

stargazer(random1, random2, type = "text", digits = 5, column.labels = c("", ""),
           keep.stat = NULL, out = "mrd.txt")
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               log(vtotal)
##
##                               (1)                (2)
## -----
## log(areatotal)              0.92678***          0.88993***
##                             (0.00712)          (0.00604)
##
## tempano                     0.12371***          -0.00297
##                             (0.00606)          (0.00582)
##
## as.factor(ano)1991          -0.22795***
##                             (0.02057)
##
## as.factor(ano)1992          0.45106***
##                             (0.02060)
##
## as.factor(ano)1993          0.61571***
```

##	(0.02069)
##	
## as.factor(ano)1994	0.59632***
##	(0.02091)
##	
## as.factor(ano)1995	0.54243***
##	(0.02098)
##	
## as.factor(ano)1996	0.57063***
##	(0.02101)
##	
## as.factor(ano)1997	0.53620***
##	(0.02093)
##	
## as.factor(ano)1998	0.62035***
##	(0.02101)
##	
## as.factor(ano)1999	0.55050***
##	(0.02088)
##	
## as.factor(ano)2000	0.65207***
##	(0.02076)
##	
## as.factor(ano)2001	0.85050***
##	(0.02120)
##	
## as.factor(ano)2002	0.95138***
##	(0.02233)
##	
## as.factor(ano)2003	0.82917***
##	(0.02115)
##	
## as.factor(ano)2004	0.81623***
##	(0.02088)
##	
## as.factor(ano)2005	0.91050***
##	(0.02121)
##	
## as.factor(ano)2006	1.05524***
##	(0.02120)
##	
## as.factor(ano)2007	1.05637***
##	(0.02173)
##	
## as.factor(ano)2008	0.90107***
##	(0.02098)
##	
## as.factor(ano)2009	0.83407***
##	(0.02128)
##	
## as.factor(ano)2010	1.00665***
##	(0.02117)
##	
## as.factor(ano)2011	1.05223***

```
## (0.02107)
##
## as.factor(ano)2012 0.98717***
## (0.02149)
##
## as.factor(ano)2013 0.91078***
## (0.02099)
##
## as.factor(ano)2014 0.84232***
## (0.02188)
##
## Constant -0.96151*** 1.59258***
## (0.16036) (0.14846)
## -----
## Observations 12,588 12,588
## R2 0.58845 0.74666
## Adjusted R2 0.58838 0.74614
## F Statistic 8,993.91100*** (df = 2; 12585) 1,423.47200*** (df = 26; 12561)
## =====
## Note: *p<0.1; **p<0.05; ***p<0.01
```

Hausman test

Ce test compare la consistance de différents estimateurs alternatifs. Grâce à cela, ce test permet de vérifier si le modèle économétrique est approprié au cas traité par l'économiste.

```
phtest(panelregre, random1)
```

```
##
## Hausman Test
##
## data: log(vtotal) ~ log(areatotal) + tempano + precano
## chisq = 276.3, df = 2, p-value < 2.2e-16
## alternative hypothesis: one model is inconsistent
```

```
phtest(panelregre2, random2)
```

```
##
## Hausman Test
##
## data: log(vtotal) ~ log(areatotal) + tempano + precano
## chisq = 53.106, df = 2, p-value = 2.939e-12
## alternative hypothesis: one model is inconsistent
```

Dans le deux cas, on accepte H_0 .