

Report on panel model results

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Abstract

These notes relate panel data models used for data with two levels (city and brand) additional to time.

The objective is to relate the models that can describe the data. There are several usual features to consider in a model with panel data, for example, changes on parameters for time or individual. Also, specification on error term is relevant for interpretation.

1 Data plots

Here is a summary of the data available for the analysis.

Figure 1 presents the initial data points used for the analysis.

To decide which brands to include, I considered the number of observations on the period previous to the tax implementation, the tax started in January 2020, I made an exploratory analysis on the December 2019 data. This would ease the estimation, brands with few observations would have difficulties to calculate most estimates. Data for brands with few observations can be analyzed after defining a criteria to form brand groups.

Figure 2 only considers the 7 most frequent brands. The graph provides some guidance on what to consider for the proposed descriptive model. In particular, there is clear trend over time and there are price adjustments in almost every January. Results only consider data for the 7 brands. Brand labels are alphabetic, as shown in the graph: 1, Benson; 2, Camel; 3, Chesterfield; 4, Lucky; 5, Marlboro; 6, Montana; 7, Pall Mall.

Except for prices of other products, there are no other potential regressors to consider at the same level of the data.

2 Dummies for each level: city, brand, time

Estimations in this section use `areg`, it imposes fixed effects. Using this method there is one category with parameters "absorbed", which are not estimated as a result of the procedure.

Figure 1: All brands average price per unit

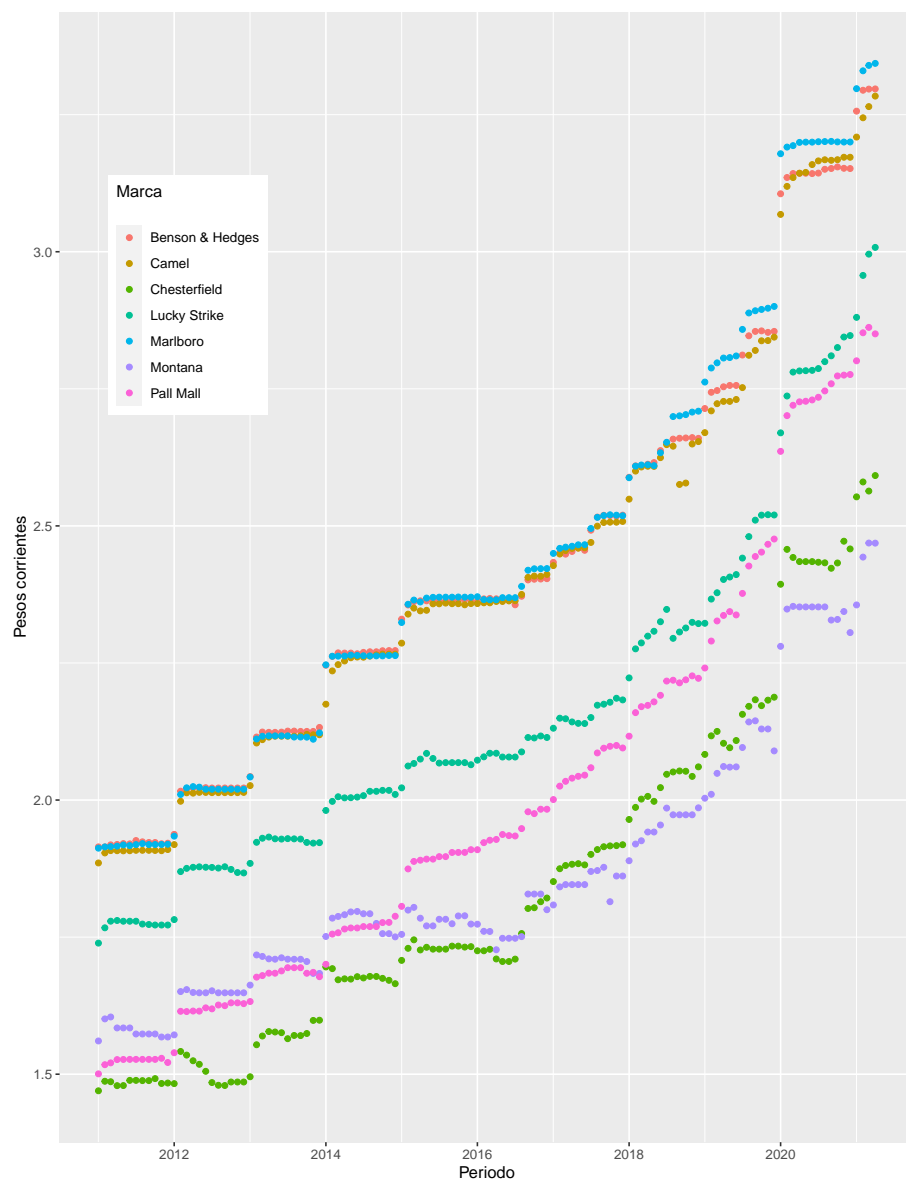


Figure 2: Seven brands average price per unit

Specification with indicators for city, brand and trend time, with the same tax effect on all the brands.

$$y_{ctm} = \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_1' tax2020 + \beta_1' tax2021 + u_{ctm}; \quad (2.1)$$

$$c = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$$

Specification with indicators for city, brand and trend time, with interacted effect for each brand.

$$y_{ctm} = \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_{1m}' tax2020 + \beta_{2m}' tax2021 + u_{itm}; \quad (2.2)$$

$$c = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$$

Dummy variables for time month and year, with same tax for all brands.

$$y_{ctm} = \alpha_i^* + \gamma_m^* + month + year + \beta_0' jan + \beta_1' tax2020 + \beta_2' tax2021 + u_{itm}; \quad (2.3)$$

Dummy variables for time month and year, with different tax for each brand.

$$y_{ctm} = \alpha_i^* + \gamma_m^* + month + year + \beta_0' jan + \beta_{1m}' tax2020 + \beta_{2m}' tax2021 + u_{itm}; \quad (2.4)$$

The columns 1 and 3 consider the same effect for each brand (2.1) and (2.3), the columns 2 and 4 estimate a different effect for each brand (2.2) and (2.4). The columns 1 and 2 consider a time trend, columns 3 and 4 use a combination of dummy variables for year and month.

Table 1: Fixed Effects for city and brand combination

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu
jan20	0.209*** (0.011)	0.238*** (0.024)	-0.017* (0.009)	0.012 (0.018)
jan21	0.242*** (0.011)	0.285*** (0.024)	-0.011 (0.010)	0.032* (0.019)
jan	-0.024*** (0.004)	-0.024*** (0.004)	-0.069*** (0.004)	-0.069*** (0.004)
jan20#2.marca		-0.007 (0.042)		-0.012 (0.032)
jan20#3.marca		-0.111** (0.043)		-0.112*** (0.033)
jan20#4.marca		-0.147*** (0.039)		-0.150*** (0.030)
jan20#5.marca		0.052 (0.032)		0.051** (0.024)

jan20#6.marca	-0.239***			-0.235***
	(0.062)			(0.048)
jan20#7.marca	-0.028			-0.024
	(0.034)			(0.026)
jan21#2.marca	-0.035			-0.040
	(0.040)			(0.031)
jan21#3.marca	-0.116**			-0.118***
	(0.046)			(0.035)
jan21#4.marca	-0.132***			-0.134***
	(0.038)			(0.029)
jan21#5.marca	0.022			0.021
	(0.032)			(0.024)
jan21#6.marca	-0.361***			-0.357***
	(0.063)			(0.048)
jan21#7.marca	-0.036			-0.031
	(0.034)			(0.026)
2.marca	-0.012***	-0.011***	-0.006**	-0.006**
	(0.003)	(0.003)	(0.003)	(0.003)
3.marca	-0.595***	-0.593***	-0.592***	-0.591***
	(0.004)	(0.004)	(0.003)	(0.003)
4.marca	-0.270***	-0.268***	-0.268***	-0.266***
	(0.003)	(0.003)	(0.003)	(0.003)
5.marca	0.011***	0.010***	0.011***	0.010***
	(0.003)	(0.003)	(0.002)	(0.002)
6.marca	-0.524***	-0.521***	-0.527***	-0.524***
	(0.005)	(0.005)	(0.004)	(0.004)
7.marca	-0.435***	-0.435***	-0.441***	-0.440***
	(0.003)	(0.003)	(0.002)	(0.002)
ym	0.009***	0.009***		
	(0.000)	(0.000)		
Constant	-3.925***	-3.926***	1.991***	1.990***
	(0.018)	(0.018)	(0.004)	(0.004)
Observations	24,010	24,010	24,010	24,010
R-squared	0.897	0.897	0.940	0.940

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

Results only consider data for the 7 brands. In the results they have a number as label assigned in alphabetic order, the same shown in the graph: 1, Benson; 2, Camel; 3, Chesterfield; 4, Lucky; 5, Marlboro; 6, Montana; 7, Pall Mall.

Although the model with dummy time variables fits the data better, R-squared, for ease of analysis of the tax effect the model with trend will be used subsequently. Because the tax effect in the last models is spread in the combination of year and month, together with the explicit coefficients for january

2020 and january 2021.

2.1 Comparisons by segment

This section separates the previous estimation of interactions by brand type or segment.

Results by brand type: premium, columns 1 and 2; medium, columns 3 and 4; low, columns 5 and 6. The omitted identifier corresponds to the reference brand.

Table 2: Fixed Effects by brand type

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu	(6) ppu
jan20	0.219*** (0.012)	0.198*** (0.019)	0.182*** (0.021)	0.119*** (0.033)	0.191*** (0.034)	0.225*** (0.040)
jan21	0.238*** (0.012)	0.235*** (0.019)	0.235*** (0.021)	0.188*** (0.033)	0.231*** (0.036)	0.310*** (0.042)
jan	-0.018*** (0.004)	-0.018*** (0.004)	-0.040*** (0.007)	-0.040*** (0.007)	-0.013 (0.009)	-0.013 (0.009)
jan20#2.marca		-0.009 (0.032)				
jan20#5.marca		0.051** (0.025)				
jan20#6.marca						-0.119* (0.072)
jan20#7.marca				0.102** (0.041)		
jan21#2.marca		-0.037 (0.031)				
jan21#5.marca		0.022 (0.025)				
jan21#6.marca						-0.252*** (0.073)
jan21#7.marca				0.076* (0.040)		
2.marca	-0.008*** (0.003)	-0.007*** (0.003)				
5.marca	0.008*** (0.002)	0.007*** (0.002)				
6.marca					0.101*** (0.007)	0.102*** (0.007)
7.marca			-0.165*** (0.004)	-0.166*** (0.004)		
ym	0.010***	0.010***	0.009***	0.009***	0.007***	0.007***

	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Constant	-4.429***	-4.429***	-4.040***	-4.041***	-2.892***	-2.895***
	(0.019)	(0.019)	(0.037)	(0.037)	(0.051)	(0.051)
Observations	13,396	13,396	6,700	6,700	3,914	3,914
R-squared	0.917	0.918	0.846	0.846	0.760	0.761

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

Brand labels are alphabetic: 1, Benson; 2, Camel; 3, Chesterfield; 4, Lucky; 5, Marlboro; 6, Montana; 7, Pall Mall.

Table 3 shows results to tests coefficient equality between brands. The first two rows correspond to the complete sample (Table 1, columns 1 and 2). Rows third and fourth test for difference in premium brands (Table 2, columns 1 and 2). Rows fifth and sixth test for difference in medium brands (Table 2, column 3 and 4). Rows seventh and eighth test for difference in lower priced brands (Table 2, column 5 and 6). Columns 1 to 3 indicate the parameters and result for the the test of fixed effects or individual effects by brand. Columns 4 to 6 show the values to test for the equality of the effect by brand of the tax in 2020. Columns 7 to 9 show the values and test result for the equality of the effect by brand of the tax in 2021.

Table 3: F tests for equality of coefficients

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	Equality of Intercept			Equality of Tax 2020			Equality of Tax 2021		
Equation	Numerator	Denominator	F	Numerator	Denominator	F	Numerator	Denominator	F
(2.1)	6	23954	10163.44						
(2.2)	6	23942	10006.08	6	23942	8.15	6	23942	9.14
(2.1)	2	13344	19.05						
(2.2)	2	13340	16.81	2	13340	2.89	2	13340	1.88
(2.1)	1	6649	1604.66						
(2.2)	1	6647	1613.81	1	6647	6.09	1	6647	3.52
(2.1)	1	3867	187						
(2.2)	1	3865	193.13	1	3865	2.74	1	3865	11.8
All tests are significant at 1 percent level									

3 Constant Parameters over time

This section presents estimations using xtreg. The first subsection presents static estimations, the second subsection contains dynamic estimates.

3.1 Static models

The first model includes the effect of the price change in every january, january 2020 and january 2021 and trend.

$$y_{it} = \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_1' tax2020 + \beta_1' tax2021 + u_{it}; \quad (4.1)$$

$c = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$

The second model includes interactions for the effect of the price change in january 2020 and january 2021, for different brand-types.

$$y_{it} = \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_{1m}' tax2020 + \beta_{2m}' tax2021 + u_{it}; \quad (4.2)$$

$i = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$

Because there are many omitted variables captured in the individual effects, there is the question of the relevance of them as fixed or random.

The result of the Hausman test for fixed effects does not rule out the non systematic difference in coefficients, this is in favour of the random effects model: $\text{Chi2}(4) = 0.60, \text{Prob} \geq \text{chi2} = 0.9628$

Similarly the Breusch-Pagan test for random effects does rule out the alternative of OLS : $\text{chibar2}(01) = 7.5\text{e}+05, \text{Prob} \geq \text{chibar2} = 0.0000$

The test of unit root, using the Fischer type estimation from Choi (2001): $\text{Inverse chi-squared}(500) = 1112.8056, \text{Prob} \geq \text{chi2} = 0.0000$, does not rule out the presence of unit root for any panel (defined as a combination of city and brand), except for the model that includes a drift. The result suggests to consider different trends for each brand or city, there is an estimation by brand to test for unit roots by specifications of the panel.

Next are the results from the complete sample, columns (1,2,4,5) are estimated using random effects. The third column presents fixed individual effects estimates, since the estimates by brand (presented next) suggest that most brands have fixed effects for city.

Column (1) corresponds to the equation (4.1), column (4) restricts the estimation in column (1) to zero trend. Column (2) corresponds to equation (4.2) column (4) restricts the estimation in column (4) to zero trend. Column (3) corresponds to column (2) estimated using fixed effects.

Table 4: Fixed Effects for brand and city by brand type

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu
jan20	0.202*** (0.010)	0.230*** (0.021)	0.230*** (0.021)	-0.015** (0.007)	0.012 (0.015)
jan21	0.232*** (0.010)	0.276*** (0.021)	0.275*** (0.021)	-0.012 (0.008)	0.027* (0.016)
jan	-0.023*** (0.003)	-0.023*** (0.003)	-0.023*** (0.003)	-0.070*** (0.003)	-0.070*** (0.003)
2.marca#jan20		-0.012 (0.037)	-0.011 (0.037)		-0.011 (0.026)
3.marca#jan20		-0.109*** (0.039)	-0.108*** (0.039)		-0.103*** (0.027)
4.marca#jan20		-0.125*** (0.034)	-0.124*** (0.034)		-0.129*** (0.024)
5.marca#jan20		0.056** (0.028)	0.056** (0.028)		0.052*** (0.020)
6.marca#jan20		-0.194*** (0.056)	-0.191*** (0.056)		-0.195*** (0.039)
7.marca#jan20		-0.052* (0.030)	-0.053* (0.030)		-0.040* (0.021)
2.marca#jan21		-0.047 (0.036)	-0.051 (0.036)		-0.039 (0.025)
3.marca#jan21		-0.101** (0.041)	-0.101** (0.041)		-0.100*** (0.029)
4.marca#jan21		-0.117*** (0.034)	-0.118*** (0.034)		-0.113*** (0.024)
5.marca#jan21		0.027 (0.028)	0.027 (0.028)		0.022 (0.020)
6.marca#jan21		-0.364*** (0.056)	-0.364*** (0.056)		-0.360*** (0.039)
7.marca#jan21		-0.053* (0.030)	-0.051* (0.030)		-0.033 (0.021)
2.marca	-0.013 (0.024)	-0.013 (0.022)		-0.009 (0.018)	-0.008 (0.017)
3.marca	-0.605*** (0.024)	-0.604*** (0.022)		-0.603*** (0.018)	-0.602*** (0.017)
4.marca	-0.274*** (0.024)	-0.272*** (0.021)		-0.269*** (0.017)	-0.268*** (0.017)
5.marca	0.004 (0.022)	0.004 (0.020)		0.007 (0.016)	0.007 (0.016)
6.marca	-0.502*** (0.028)	-0.500*** (0.026)		-0.505*** (0.021)	-0.503*** (0.020)
7.marca	-0.426*** (0.023)	-0.425*** (0.021)		-0.443*** (0.017)	-0.442*** (0.017)

ym	0.009*** (0.000)	0.009*** (0.000)	0.009*** (0.000)		
Constant	-3.940*** (0.023)	-3.940*** (0.022)	-4.137*** (0.017)	1.989*** (0.012)	1.989*** (0.012)
Observations	23,926	23,926	23,926	23,926	23,926
Number of gr_marca_ciudad	263	263	263	263	263
R-squared			0.866		

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

Next table present the results by brand type: premium, columns 1 and 2; medium, columns 3 and 4; low, columns 5 and 6. The omitted identifier corresponds to the reference brand. Columns (1,3,5) correspond to the equation (4.1), columns (2,4,6) correspond to equation (4.2).

Table 5: Fixed Effects/Random Effects for brand and city by brand type

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu	(6) ppu
jan20	0.218*** (0.011)	0.196*** (0.018)	0.176*** (0.019)	0.132*** (0.030)	0.190*** (0.032)	0.212*** (0.037)
jan21	0.235*** (0.011)	0.233*** (0.018)	0.228*** (0.019)	0.251*** (0.024)	0.217*** (0.034)	0.295*** (0.040)
jan	-0.018*** (0.004)	-0.018*** (0.004)	-0.039*** (0.006)	-0.039*** (0.006)	-0.012 (0.008)	-0.012 (0.008)
jan20#2.marca		-0.012 (0.031)				
jan20#5.marca		0.053** (0.024)				
jan20#6.marca						-0.078 (0.068)
jan20#7.marca				0.071* (0.037)		
jan21#2.marca		-0.049 (0.030)				
jan21#4.marca				-0.062* (0.037)		
jan21#5.marca		0.024 (0.024)				
jan21#6.marca						-0.245*** (0.069)
4.marca			-4.013*** (0.041)	0.150*** (0.028)		
7.marca			-4.162***			

			(0.041)			
ym	0.010***	0.010***	0.009***	0.009***	0.007***	0.007***
	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)	(0.000)
Constant	-4.417***	-4.416***		-4.164***	-2.961***	-2.963***
	(0.019)	(0.019)		(0.041)	(0.050)	(0.049)
Observations	13,396	13,396	6,620	6,620	3,910	3,910
R-squared	0.916	0.916			0.720	0.721
Number of gr_marca_ciudad	126	126	80	80	57	57

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

By brand type, for premium(1) and low(3) brand types the Hausman test (see Table 16) rejects the alternative of random effects in favour of fixed individual effects. The test for type 1 and 3 are $\text{Chi2}(4) = 0.60$, $\text{Prob} \geq \text{chi2} = 0.9628$ and $\text{Chi2}(4) = 0.60$, $\text{Prob} \geq \text{chi2} = 0.9628$, respectively.

Similarly the Breusch-Pagan test for random effects does rule out the alternative of OLS : $\text{chibar2}(01) = 7.5\text{e}+05$, $\text{Prob} \geq \text{chibar2} = 0.0000$

The test of unit root, using the Fischer type estimation from Choi (2001): Inverse chi-squared(500) = 1112.8056 , $\text{Prob} \geq \text{chi2} = 0.0000$, does not rule out the presence of unit root for any panel (defined as a combination of city and brand), except for the model that includes a drift. The result suggests to consider different trends for each brand or city, there is an estimation by brand to test for unit roots by specifications of the panel. See Table 17.

The next table presents F or Chi-squared tests for equality of coefficients, F corresponds to models by fixed effects, Chi-squared corresponds to models estimated by random effects.

The distribution is the same used in Table 3, first two rows correspond to the complete sample. Rows third and fourth test for difference in premium brands. Rows fifth and sixth test for difference in medium brands. Rows seventh and eighth test for difference in lower priced brands. Columns 1 to 3 indicate the parameters and result for the the test of fixed effects or individual effects by brand. Columns 4 to 6 show the values to test for the equality of the effect by brand of the tax in 2020. Columns 7 to 9 show the values and test result for the equality of the effect by brand of the tax in 2021.

Except for the effect of the tax in 2020 for lower priced brands, based on these F tests, the preferred model includes interactions by brand.

Table 6: F tests for equality of coefficients

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Equation	Num/gl-chi2	Equality of Intercept Den	F/Chi2	Num/gl-chi2	Equality of Tax 2020 Den	F/Chi2	Num/gl-chi2	Equality of Tax 2021 Den	F/Chi2
(4.1)	6		1287.39***						
(4.2)	6		1547.03***	6		51.86***	6		67.31***
(4.1)	125	13266	22.32***						
(4.2)	125	13262	22.32***	2	13262	3.51**	2	13262	3.08***
(4.1)	2		11224.48***						
(4.2)	1		28.63***	1		3.62*	1		2.85*
(4.1)	56	3849	41.93***						
(4.2)	56	3847	42.17***	1	3847	1.32	1	3847	12.43*

*** p<0.01, ** p<0.05, * p<0.1

The next table presents the static results by brand. The model was estimated using random or fixed effects for each brand, based on the Hausman test. The column number corresponds to the brand label: 1, Benson; 2, Camel; 3, Chesterfield; 4, Lucky; 5, Marlboro; 6, Montana; 7, Pall Mall.

Table 7: Fixed/Random individual effects for each brand

VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	ppu	ppu	ppu	ppu	ppu	ppu	ppu
jan20	0.193*** (0.018)	0.219*** (0.027)	0.192*** (0.040)	0.169*** (0.030)	0.232*** (0.018)	0.182*** (0.048)	0.177*** (0.024)
jan21	0.231*** (0.018)	0.224*** (0.025)	0.270*** (0.043)	0.238*** (0.030)	0.237*** (0.018)	0.113** (0.048)	0.218*** (0.024)
jan	-0.013** (0.006)	-0.031*** (0.007)	-0.007 (0.011)	-0.033*** (0.009)	-0.013** (0.006)	-0.023** (0.012)	-0.045*** (0.009)
ym	0.010*** (0.000)	0.010*** (0.000)	0.007*** (0.000)	0.008*** (0.000)	0.010*** (0.000)	0.006*** (0.000)	0.010*** (0.000)
Observations	4,650	3,207	2,700	3,213	5,539	1,210	3,407
R-squared	0.921	0.896	0.717	0.792	0.922	0.752	0.874
Number of gr_marca_ciudad	44	36	35	38	46	22	42

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

3.2 Dynamic models

Since we cannot rule the presence of unit roots for each panel, by city, except for a model with drift, an alternative is to consider dynamics in the equation, in particular the lagged dependent variable.

$$y_{it} = \delta * y_{i,t-1} + \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_1' tax2020 + \beta_1' tax2021 + u_{it}; \quad (4.3)$$

$$c = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$$

Similar to the previous section, estimations include interactions, to consider the effect of the price change in every january, january 2020 and january 2021.

$$y_{it} = \delta * y_{i,t-1} + \alpha_i^* + \gamma_m^* + \lambda * trend + \beta_0' jan + \beta_{1m}' tax2020 + \beta_{2m}' tax2021 + u_{it}; \quad (4.4)$$

$$i = 1, \dots, N; t = 1, \dots, T; m = 1, \dots, M.$$

Table 8 shows the results preferred by the Hausman test (see: Table 18), in this case fixed effects. It also shows the estimates by type. The first two columns correspond to the model estimation without interactions. Third and fourth columns include interactions by brand. Fifth and sixth columns consider interactions by brand type. Columns 2, 4 and 6 present, for reference, an estimation without lagged dependent variable.

The results show that the lag effect is captured, when omitted, by the parameter of the linear trend and the effect of the tax in january 2021.

The estimation by brand type is extended in Table 9, (1) and (2) show the estimates for premium brands, (3) and (4) show the estimates for medium brands, (5) and (6) show the estimates for lower brands. Columns (2,4,6) show results with interactions for brand and tax coefficients and columns (1,3,5) without interactions. In the next table, the dependent variable was transformed to cents.

The coefficient for jan20 and jan21 correspond to the value of the reference brand within a brand type. The coefficient for the other brands indicate the difference to that reference brand for the correspondent brand. The labels correspond to the alphabetical order, 1 is Benson, reference for premium brands, 2 is Camel, 3 is Chesterfield, reference for lower brands, 4 is Lucky Strike, reference for medium brands, 5 is Marlboro, 6 is Montana and 7 is Pall Mall.

With premium for the first label, results shows that the medium brands have the smallest tax impact, although, counterintuitively the lowest impact is estimated for the medium brands with a decrease of 8.9 cents while the lower brand only decreased 5 cents, both with respect to the premium brands average.

The distribution in the next table is the same used in Table 3 and Table 6. F tests for equality of coefficients, significant indicates difference between coefficients, the first two rows correspond to the complete sample. The third and fourth to the premium brands, where the equality is still relevant. Fifth and sixth rows are medium brands, where only is difference for the impact of

the tax in 2021. The last two rows correspond to the lower priced brands, here there is no difference in the coefficients for each brand.

Columns 1 to 3 indicate the parameters and result for the the test individual effects by brand. Columns 4 to 6 show the values to test for the equality of the effect by brand of the tax in 2020. Columns 7 to 9 show the values and test result for the equality of the effect by brand of the tax in 2021.

The results show that exception of premium brands, brand type or segment is relevant for equality of all the coefficients considered: intercept, tax2020 and tax2021. When considered in one equation the preferred estimation is with complete interactions and lag dependent variable (Column 3, Table 8).

Table 8: Fixed/Random individual effects

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu	(6) ppu
jan20	0.182*** (0.003)	0.202*** (0.010)	0.207*** (0.006)	0.230*** (0.021)		0.253*** (0.013)
jan21	0.044*** (0.003)	0.232*** (0.010)	0.083*** (0.006)	0.275*** (0.021)		0.278*** (0.013)
jan	0.022*** (0.001)	-0.023*** (0.003)	0.022*** (0.001)	-0.023*** (0.003)	0.022*** (0.001)	-0.023*** (0.003)
2.marca#jan20			-0.018* (0.010)	-0.011 (0.037)		
3.marca#jan20			-0.043*** (0.010)	-0.108*** (0.039)		
4.marca#jan20			-0.088*** (0.009)	-0.124*** (0.034)		
5.marca#jan20			0.031*** (0.008)	0.056** (0.028)		
6.marca#jan20			-0.049*** (0.015)	-0.191*** (0.056)		
7.marca#jan20			-0.073*** (0.008)	-0.053* (0.030)		
2.tipo#jan20					-0.089*** (0.006)	
3.tipo#jan20					-0.055*** (0.008)	
2.marca#jan21			-0.069*** (0.010)	-0.051 (0.036)		
3.marca#jan21			-0.023** (0.011)	-0.101** (0.041)		
4.marca#jan21			-0.083*** (0.009)	-0.118*** (0.034)		
5.marca#jan21			-0.013* (0.008)	0.027 (0.028)		
6.marca#jan21			-0.059*** (0.015)	-0.364*** (0.056)		
7.marca#jan21			-0.069*** (0.008)	-0.051* (0.030)		
2.tipo#jan21					-0.056*** (0.006)	
3.tipo#jan21					-0.016* (0.009)	
L.ppu	0.967*** (0.002)		0.966*** (0.002)		0.967*** (0.002)	
ym	0.000*** (0.000)	0.009*** (0.000)	0.000*** (0.000)	0.009*** (0.000)	0.000*** (0.000)	0.009*** (0.000)
Constant	-0.170*** (0.009)	-4.138*** (0.017)	-0.171*** (0.009)	-4.137*** (0.017)	-0.170*** (0.009)	-4.137*** (0.017)
Observations	23,616	23,926	23,616	23,926	23,616	23,926
R-squared	0.990	0.865	0.990	0.866	0.990	0.866
Number of gr_marca_ciudad	262	263	262	263	262	263

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

Table 9: Fixed individual effects by brand type, interacted

VARIABLES	(1) ppu100	(2) ppu100	(3) ppu100	(4) ppu100	(5) ppu100	(6) ppu100
m1_20	20.974*** (0.316)	19.904*** (0.496)	14.040*** (0.539)	13.084*** (0.845)	16.867*** (0.933)	17.046*** (1.091)
1.m1_20#2.marca		-1.762** (0.864)				
1.m1_20#5.marca		3.065*** (0.658)				
1.m1_20#6.marca						-0.620 (1.981)
1.m1_20#7.marca				1.534 (1.042)		
m1_21	5.700*** (0.320)	7.592*** (0.505)	1.948*** (0.544)	1.081 (0.833)	5.568*** (0.983)	6.670*** (1.175)
1.m1_21#2.marca		-6.874*** (0.840)				
1.m1_21#5.marca		-1.311** (0.663)				
1.m1_21#6.marca						-3.465* (2.023)
1.m1_21#7.marca				1.421 (1.030)		
m1	2.960*** (0.104)	2.960*** (0.104)	0.846*** (0.191)	0.844*** (0.191)	1.862*** (0.261)	1.859*** (0.261)
ym	0.044*** (0.003)	0.044*** (0.003)	0.038*** (0.004)	0.038*** (0.004)	0.030*** (0.004)	0.030*** (0.004)
L.ppu	96.071*** (0.245)	96.065*** (0.244)	96.872*** (0.351)	96.853*** (0.351)	96.663*** (0.479)	96.618*** (0.480)
Constant	-19.521*** (1.204)	-19.513*** (1.199)	-18.151*** (1.744)	-18.251*** (1.744)	-13.425*** (2.032)	-13.580*** (2.034)
Observations	13,255	13,255	6,524	6,524	3,837	3,837
R-squared	0.994	0.994	0.987	0.987	0.976	0.976
Number of gr_marca_ciudad	125	125	80	80	57	57

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

Table 10: F tests by brand type, interacted

Equation	(1) Num	(2) Den	(3) Equality of Intercept F	(4) Num	(5) Den	(6) Equality of Tax 2020 F	(7) Num	(8) Den	(9) Equality of Tax 2021 F
(4.3)	261	23349	1.6***	6	23337	47.36***	6	23337	25.88***
(4.4)	261	23337	1.62***						
(4.3)	124	13125	0.59						
(4.4)	124	13121	0.6	2	13121	20.34***	2	13121	35.05***
(4.3)	79	6439	0.92						
(4.4)	79	6437	0.93	1	6437	2.17	1	6437	1.9***
(4.3)	56	3775	0.94						
(4.4)	56	3773	0.95	1	3773	0.1	1	3773	2.93

*** p<0.01, ** p<0.05, * p<0.1

The dynamic model with full interactions, i.e. also january, trend and lag, returns F tests that show significant different results for each brand. See Table 11

Table 11: F tests by brand type, Fully interacted

HERE THE TABLE WITH F TESTS FOR DIFFERENCE BY BRAND OF JAN, Trend and Lag

The table 12 shows the estimated coefficients by brand in columns.

Table 12: Fixed/Random individual effects for each brand

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu	(6) ppu	(7) ppu
jan20	0.193*** (0.005)	0.201*** (0.008)	0.166*** (0.011)	0.127*** (0.009)	0.223*** (0.005)	0.174*** (0.017)	0.150*** (0.007)
jan21	0.069*** (0.005)	0.028*** (0.007)	0.061*** (0.012)	0.006 (0.009)	0.056*** (0.005)	0.044** (0.017)	0.030*** (0.007)
jan	0.037*** (0.002)	0.009*** (0.002)	0.023*** (0.003)	0.013*** (0.003)	0.036*** (0.002)	0.009** (0.005)	0.003 (0.003)
ym	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)
L.ppu	0.964*** (0.004)	0.957*** (0.005)	0.971*** (0.005)	0.974*** (0.005)	0.960*** (0.004)	0.944*** (0.011)	0.959*** (0.005)
Constant	-0.180*** (0.019)	-0.190*** (0.025)	-0.133*** (0.025)	-0.143*** (0.022)	-0.209*** (0.019)	-0.158*** (0.036)	-0.244*** (0.028)
Observations	4,601	3,163	2,655	3,167	5,491	1,182	3,357
R-squared	0.994	0.991	0.979	0.983	0.994	0.968	0.989
Number of gr_marca_ciudad	44	35	35	38	46	22	42

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

3.2.1 Test for different time windows in tax definition

To test if there are any price adjustment due to the announcement of the tax implementation at the end of october, the next table (13) shows estimation results with dummy variables for november and december 2019 and 2020, to consider earlier effect on price for tax adjustment on 2020 and 2021, respectively.

Estimates these wider tax effect show are negative or not statistically significant, this implies that the impact of the tax adjustment occurs in january.

Table 13: Wide Time window for tax adjustment

VARIABLES	(1) ppu	(2) ppu	(3) ppu	(4) ppu	(5) ppu	(6) ppu	(7) ppu
jan	0.036*** (0.002)	0.009*** (0.002)	0.023*** (0.003)	0.013*** (0.003)	0.036*** (0.002)	0.009** (0.005)	0.003 (0.003)
ym	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)
nov19_dic19	-0.011*** (0.003)	-0.008 (0.005)	0.001 (0.008)	-0.014** (0.006)	-0.006* (0.003)	-0.026** (0.012)	-0.008* (0.005)
jan20	0.192*** (0.005)	0.201*** (0.008)	0.166*** (0.011)	0.126*** (0.009)	0.223*** (0.005)	0.172*** (0.017)	0.149*** (0.007)
nov20_dic20	-0.007** (0.003)	-0.004 (0.005)	0.003 (0.008)	-0.009 (0.006)	-0.002 (0.003)	-0.030** (0.012)	-0.009* (0.005)
jan21	0.068*** (0.005)	0.027*** (0.007)	0.061*** (0.012)	0.005 (0.009)	0.056*** (0.005)	0.041** (0.017)	0.029*** (0.007)
L_ppu	0.965*** (0.004)	0.958*** (0.005)	0.971*** (0.005)	0.976*** (0.005)	0.960*** (0.004)	0.948*** (0.011)	0.961*** (0.005)
Constant	-0.184*** (0.019)	-0.191*** (0.025)	-0.133*** (0.025)	-0.147*** (0.022)	-0.212*** (0.019)	-0.172*** (0.036)	-0.244*** (0.028)
Observations	4,601	3,163	2,655	3,167	5,491	1,182	3,357
R-squared	0.994	0.991	0.979	0.983	0.994	0.969	0.989
Number of gr_marca_ciudad	44	35	35	38	46	22	42

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

3.2.2 Effects Including Other brands

Shown in the next table (14) are models of price adjustment including data from brands with fewer observations. Instead of getting an estimation by brand, the estimates are by type including the seven brands with more observations.

Columns 1, 3, 5 and 7 present the estimates for each brand type (premium, medium, low and other, respectively). Columns 2, 4, 6 show the estimates considering only the 7 principal brands by brand type (premium, medium and low). Column 8 presents the estimates for brand type low and other, to consider differences between the cheapest brands.

Table 14: Includes data for other brands

VARIABLES	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	ppu	ppu	ppu	ppu	ppu	ppu	ppu	ppu
jan20	0.211*** (0.003)	0.210*** (0.003)	0.140*** (0.005)	0.140*** (0.005)	0.156*** (0.009)	0.169*** (0.009)	0.157*** (0.009)	0.157*** (0.007)
jan21	0.058*** (0.003)	0.057*** (0.003)	0.018*** (0.005)	0.019*** (0.005)	0.057*** (0.010)	0.056*** (0.010)	0.057*** (0.010)	0.057*** (0.010)
jan	0.029*** (0.001)	0.030*** (0.001)	0.010*** (0.002)	0.008*** (0.002)	0.018*** (0.003)	0.019*** (0.003)	0.020*** (0.004)	0.018*** (0.002)
ym	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)	0.000*** (0.000)
L.ppu	0.960*** (0.002)	0.961*** (0.002)	0.969*** (0.003)	0.969*** (0.004)	0.958*** (0.005)	0.967*** (0.005)	0.951*** (0.011)	0.958*** (0.004)
Constant	-0.199*** (0.012)	-0.195*** (0.012)	-0.181*** (0.017)	-0.182*** (0.017)	-0.167*** (0.020)	-0.134*** (0.020)	-0.238*** (0.052)	-0.173*** (0.019)
Observations	13,731	13,255	6,998	6,524	4,479	3,837	658	5,137
R-squared	0.993	0.994	0.987	0.987	0.972	0.976	0.973	0.972
Number of gr_marca_ciudad	139	125	90	80	76	57	26	102

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

4 Consistent estimation for Variable Intercept

This models are based on Andrews, et al. (2006). The initial model comes from the transformation of:

$$y_{it} = x_{it}\beta_i + w_{j(i,t)}\gamma + u_{it}\eta + q_{j(i,t)}\rho + \alpha_i + \phi_{j(i,t)} + \mu_t + \epsilon_{i,t};$$

$$i = 1, \dots, N; t = 1, \dots, T$$

Given the interest only on the fixed independent variables, we can define an heterogeneity measure on brand and city (s), take the averages at that level, and make the transformation of variables, following:

$$y_{it} - \bar{y}_s = (x_{it} - \bar{x}_s)\beta_i + (w_{j(i,t)} - \bar{w}_s)\gamma + (\epsilon_{i,t} - \bar{\epsilon}_s);$$

$$i = 1, \dots, N; t = 1, \dots, T$$

Results in the left have the same estimate for the effect. Estimates in second column correspond to the first labeled brand, Benson.

Table 15: Transformation for consistency

VARIABLES	(1) dm_ppu_cm	(2) dm_ppu_cm
dm_m1_20_cm	0.202*** (0.010)	0.230*** (0.021)
dm_m1_21_cm	0.232*** (0.010)	0.275*** (0.021)
dm_m1_cm	-0.023*** (0.003)	-0.023*** (0.003)
ym	0.009*** (0.000)	0.009*** (0.000)
Observations	23,926	23,926
R-squared	0.865	0.866
Number of gr_marca_ciudad	263	263

Standard errors in parentheses

*** p<0.01, ** p<0.05, * p<0.1

5 Different parameters for each brand

Uses xtsur, user-defined, command. One estimate of each parameter for each brand. The intention was to make a unique model of Seemingly Unrelated Regressions to test the coefficients of the tax change for equality. Unfortunately, it is impossible (using the xtsur routine, in a 4th gen i7 with 16ram) to make

the estimation based on the complete sample, with 7 brands. I present the test based on three groups of brands.

$$y_{itm} = \alpha_i^* + \lambda_1 * t + \beta_0' janDummy_m + \beta_1' taxDummy_m + u_{itm};$$

$$i = 1, \dots, N; t = 1, \dots, T, m = 1, 2, \dots, 7$$

5.1 Comparisons by segment

Results for premium brands

VARIABLES	(1) ppu1	(2) ppu2	(3) ppu5
m1_20	0.156*** (0.018)	0.135*** (0.018)	0.151*** (0.012)
m1_21	0.077 (0.000)	0.027 (0.000)	0.070 (0.000)
m1	0.024*** (0.004)	0.000 (0.004)	0.003 (0.002)
ym	0.009*** (0.000)	0.009*** (0.000)	0.009*** (0.000)
Observations	2,543	2,543	2,543
Number of cve_ciudad	45	45	45

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

Results for lower segment brands

VARIABLES	(1) ppu3	(2) ppu6
m1_20	0.322*** (0.047)	0.204*** (0.030)
m1_21	0.364*** (0.047)	0.446*** (0.030)
m1	0.009 (0.015)	-0.035*** (0.010)
ym	0.007*** (0.000)	0.005*** (0.000)
Observations	614	614
Number of cve_ciudad	43	43

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

Results for mid-range segment brands

VARIABLES	(1) ppu4	(2) ppu7
m1_20	0.150*** (0.018)	0.071*** (0.018)
m1_21	0.050*** (0.018)	0.191*** (0.018)
m1	-0.013*** (0.005)	-0.018*** (0.005)
ym	0.009*** (0.000)	0.010*** (0.000)
Observations	2,112	2,112
Number of cve_ciudad	43	43

Standard errors in parentheses
*** p<0.01, ** p<0.05, * p<0.1

A

Appendix A: Tests tables

Results for static models Hausman tests

Table 16: Hausman tests static models

Name	Interactions	(1) chi2	(2) df	(3) p
(4.3)	No interactions	0.60265	5	0.96277***
(4.4)	Interactions	83.08106	16	0***
(4.4)	Premium	41.63451	5	0***
(4.4)	Medium	6.55265	5	0.1615
(4.4)	Lower	13.74536	5	0.00815***

*** p<0.01, ** p<0.05, * p<0.1

Results for Unit Root tests

Table 17: Unit Root tests

Unit Root Model/Test	Statistic(df,p-value)		
	(1)	(2)	(3)
Trend/Inverse chi-squared (df)[P]	No interactions	Premium	Medium
Trend/Inverse normal[Z]	289.457(508,1)	119.86(248,1)	81.862(152,1)
Trend/Inverse logit t(df)[L*]	11.905(,1)	9.88(,1)	7.002(,1)
Trend/Modified inv. chi-squared[Pm]	11.893(1234,1)	10.013(624,1)	6.935(369,1)
Drift/Inverse chi-squared (df)[P]	-6.856(,1)	-5.754(,1)	-4.023(,0.99997)
Drift/Inverse normal[Z]	1112.806(500,0)***	576.02(248,0)***	317.181(148,0)***
Drift/Inverse logit t(df)[L*]	-16.884(,0)***	-13.303(,0)***	-8.623(,0)***
Drift/Modified inv. chi-squared[Pm]	-17.034(1254,0)***	-13.16(624,0)***	-8.681(374,0)***
	19.379(,0)***	14.729(,0)***	9.833(,0)***
	*** p<0.01, ** p<0.05, * p<0.1		
			136.492(108,0.03326)
			2.173(,0.98509)
			0.525(244,0.69998)
			1.939(,0.02627)
			288.84(104,0)***
			-9.417(,0)***
			-10.066(264,0)***
			12.816(,0)***

Results for dynamic models Hausman tests

Table 18: Hausman tests dynamic models

		(1)	(2)	(3)
Name	Interactions	chi2	df	p
(4.3)	No interactions	330.23783	5	0
(4.4)	Interactions	141.36743	16	0
(4.4)	Segment	155.26197	9	0
(4.4)	Premium	59.99117	5	0
(4.4)	Medium	49.7208	5	0
(4.4)	Lower	24.90764	5	0.00015
All tests are significant at 1%				