

Web mining – Term project

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"Effect of cross-border labor mobility on national real estate price"



Credit: https://www.fondation-idea.lu/2019/04/30/combien-y-a-t-il-vraiment-de-travailleurs-frontaliers-au-luxembourg/

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1. Introduction

Cross border labor mobility has drastically increased in Europe after creation of the EU and adoption of different treaties with its neighbors. Academic literature studied this phenomenon and its consequences in many regards including unemployment, competition, and price. The latter constitutes a core domain of research in this context and contains a whole subcategory in real estate, including Micheli, Rouwendal, Dekkers (2018), Dumeignil (2021) and Gonzalez, Ortega (2012) amongst other. Thanks to the heterogeneity dimension of political decision, most of those studies use changes in the legal framework as natural experiment for their analysis. In this case however, real time market data are gathered from the internet to isolate real estate price and its determinants in four different cities, two in France and two in Germany. half are located inland and the other half at the border of Switzerland to constitute the treatment group due to the number of cross-border commuters driven by minimum income discrepancy.

A naïve mean comparison between treated and non-treated is first used to test the effect. It returns poor results quality and does not suffice for causal claim. A linear regression is then performed to identify the weight of each determinant in price composition. Based on the results and backed by Grether, Mieszkowski (1973), footage appears to be the most important. Therefore, interaction effect of real estate footage and treatment are analyzed to test the impact of cross border labor mobility on real estate price. The assumption being that, by controlling for density, land tax, job attractiveness, and unemployment, any variation of footage's coefficient would reflect the treatment effect. A statistically significant difference is found for France between region near Geneva and Lyon but none for Germany comparing region near Basel and Stuttgart. Despite literature support, this research design doesn't allow to conclude in a consistent treatment effect.

2. Data

As mentioned, natural experiment is usually preferred for such analysis, but it is well known that adequate circumstances to perform it are rare. Therefore, web mining technics are used in this study to generate the dataset. Indeed, thanks to many real estate's announcement websites, it is easy to obtain market data such as price, size, number of rooms and type of estate. One French and one German property advertisement website were picked, namely, seloger.com and Wohnungsboerse.net.

While the power of web scrapping would allow to deal with enormous amount of data, nowadays many hosting servers are equipped with bot detection algorithm restricting the scope of possibilities. Indeed, seloger.com as the first French website per volume, automatically block the access after too many requests. It nevertheless remains the best option for remote location. In Germany, Immobilienscout24.de goes even further and has anti-bot protection preventing any attempt. Second best option called wohnungsboerse.net is therefore held.

As mentionned, first step of research design is focused on identifying two municipalities per country as similar as possible. The choice fell on Miribel and Gex in France and Leinfelden-Echterdingen and Lörrach in Germany. That selection process raises important risk of selection bias and reflect the weakness of this study. However, comparisons of many criteria were made to limit systematic error as much as possible. First, by choosing towns in the same state, most difference in land tax can be controlled as for France and Germany, it is managed on a national or state level. In the case of France, only two municipal taxes are applied: Property tax on property (TFPB) and property tax on non-buildings (TFPNB) and will be considered in the results. No such tax was found within German state of Baden-Württemberg (BW). Regarding unemployment rate, communal French latest figures available are for 2019 only, a period when economic conditions were very different. State numbers are therefore retained for both French cities with 5.4% (1).

 $^{^{\}rm 1}$ Data of the 4th trimester 2022 from INSEE, https://www.insee.fr/fr/statistiques/2012804

Baden-Württemberg state on the other hand provides more recent figures for Lörrach and Leinfelden-Echterdingen with 4.1% and 3.4% $^{(2)}$ respectively. It's worth comparing with neighboring regions in Switzerland to identify any structural difference. At the latest, Geneva's state announces 3.7% $^{(3)}$ against 3% $^{(4)}$ for Basel-Stadt. Those differences must be acknowledged.

Intimately connected, job attractivity of the location has its importance. Therefore, all cities are within a circle of 10-15 km of major agglomerations, namely Geneva or Lyon for France and Basel or Stuttgart for Germany. For that reason, the locations are assumed to be likely attractive in this regard. It should be noted that no discrimination in hiring foreign workers were assumed.

Finally, density of cities must be similar as land shortage remains one of the most important metrics when it comes to property price. The total number of inhabitant and inhabitants per squared kilometer have hence been controlled. According to Wikipedia as of 2020, Miribel has a population of 10,202 and a density of 420/km² while Gex announces 13,177 and 410/km². On the other hand, Leinfelden-Echterdingen counts 40,100 for 1,300/km² while Lörrach declares 49,318 inhabitants with 1,300/km² as well.

The table	helow	summarizes	those	characteristics:
THE LUDIC	201044	Juli 1111111112CJ	uiosc	cital acteristics.

Town	State	Unemployment	Distance to city	Nearest city	Population	Density
Miribel	Aim	5.4%	13.1 km	Lyon	10,202	420/km ²
Gex	Aim	5.4%	16.2 km	Geneva	13,177	410/km ²
Leinfelden	BW	3.4%	11.3 km	Stuttgart	40,100	1,300/km ²
Lörrach	BW	4.1%	13.7 km	Basel	49,318	1,300/km ²

Figure 1 – cities characteristics

From those 4 locations, every announcement available on the respective website are collected, including: Price (1), the type of estate (2), number of rooms (3) and footage (4). Properties other than apartment and villa are excluded to focus the research on so-called "traditional" estate. The same process is applied on Wohnungsboerse.net. At the moment of the study, a total of 309 property advertisements were collected for Germany including 176 in Leinfelden-Echertdingen against 133 in Lörrach and 288 for France including 95 for Miribel against 195 for Gex.



Figure 2 – announcement example

² Data of March 2022 from Statistisches Landsamt, https://www.statistik-bw.de/Arbeit/Arbeitslose/03033022.tab?R=KR336

³ Data of April 2023 from Republique et Canton de Genève, https://statistique.ge.ch/domaines/apercu.asp?dom=03_03

⁴ Data of April 2023 from Statistisches Amt, https://www.statistik.bs.ch/zahlen/tabellen/3-arbeit-erwerb/arbeitslose.html

The proportion of apartment for sale on seloger.com is of is of 52% in the non-treated group and 74% for the treated. For Germany, those figures are 60% and 58% respectively. Type is slightly unbalanced in the data but will be controlled in the research design.

Here's a summary of the three main characteristics collected for both countries:

```
> summary(data_d$Price)
                                                   > summary(data_f$Price)
  Min. 1st Qu. Median
                           Mean 3rd Qu.
                                                      Min. 1st Qu.
                                                                    Median
                                                                               Mean 3rd Qu.
                                            Max.
                                                                                                Max.
 12000 299900
                 460000
                         573453 750000 3100000
                                                    109900 252250
                                                                    337000
                                                                             425363 491250 2980000
> summary(data_d$Room)
                                                   > summary(data_f$Room)
                 Median
  Min. 1st Qu.
                           Mean 3rd Ou.
                                            Max.
                                                      Min. 1st Qu.
                                                                    Median
                                                                               Mean 3rd Ou.
                                                                                                Max.
 1.000
          3.000
                  4.000
                          4.655
                                  5.500
                                          30.000
                                                             1.000
                                                                      2,000
                                                                              2.597
                                                                                      3.250
> summarv(data_d$Footage)
                                                   > summary(data_f$Footage)
  Min. 1st Qu.
                 Median
                           Mean 3rd Qu.
                                            Max.
                                                                               Mean 3rd Qu.
                                                      Min. 1st Qu.
                                                                    Median
  10.0
           71.0
                  105.0
                          129.5
                                  155.0
                                           880.0
                                                                       77.5
                                                                              102.1
                                                                                     126.2
                                                      28.0
                                                              61.0
                                                                                               517.0
> print(sumup_d)
                                                   > print(sumup_f)
      Leinfelden Lörrach
                                                         Miribel Gex
Flat
              80
                     101
                                                   Flat
                                                              50 143
House
              53
                      75
                                                              45
                                                   House
```

Figure 3 – data statistics

3. Empirical Strategy/Methods

In 2014 Switzerland voted to adopt what would have been the highest minimum wage in the world at CHF 22. While the initiative was rejected at 76.3% ⁽⁵⁾, many states implemented similar laws within their jurisdiction. Notably, Geneva and Basel-Stadt with an amount of CHF 24 ⁽⁶⁾ and CHF 23⁽⁷⁾ respectively. On the other side of the borders, France, and Germany have also put in place similar measures but at a significant lower level with $11.52 \in {}^{(8)}$ and EUR $12 \in {}^{(9)}$. This discrepancy represents a serious arbitrage opportunity that many have figured out. Cross-border mobility labor is therefore very important in those areas implying major population movement.

To test the causal effect of this phenomena on real estate price, a regression discontinuity design could be used with border as the intervention. Indeed, by selecting two cities of similar characteristics in the neighbor countries with only one at the border, property price difference would reflect the treatment effect. While the risk for bias selection is important, controls can be made to limit the impact (see Data).

Two methods are used to determine the impact of cross-border labor mobility. First, simply comparing the average price of the treated and the non-treated for both geographical areas. An additional approach based on the property type is also tested to control for the effect of neighborhood as a potential confounder.

a) Naïve approach:

$$\widehat{ATE} = \frac{1}{N} \sum (price_1(i) - price_0(i))$$

b) Property type controlled:

$$\widehat{ATE} = \frac{1}{N} \sum (price_{1,t}(i) - price_{0,t}(i)) \text{ for } t = 1,2 \text{ (apartment or house)}$$

⁵ Volksinitiative vom 23.01.2012 «Für den Schutz fairer Löhne» https://www.bk.admin.ch/ch/d/pore/va/20140518/det583.html

⁶ Appliquer le salaire minimum genevois, https://www.ge.ch/appliquer-salaire-minimum-genevois

⁷ Gesetz über den kantonalen Mindestlohn (MiLoG), https://www.gesetzessammlung.bs.ch/app/de/texts_of_law/812.200

⁸ Smic (salaire minimum de croissance), https://www.service-public.fr/particuliers/vosdroits/F2300

⁹ Eine Frage des Respekts, https://www.bundesregierung.de/breg-de/aktuelles/mindestlohn-faq-1688186

The second method aims to be more subtle in its design and analyzes the effect of treatment indirectly. With the use of a linear regression, price determinants effect is decomposed while including the interaction effect with treatment to test its impact on coefficients.

c) <u>Linear regression with interaction effect</u>

$$Price(i) = \beta_0 + \beta_1 Type(i) + \beta_2 Room(i) + \beta_3 Footage(i) + \beta_4 Treatment(i)$$
$$+\beta_5 Type(i) * Treatment(i) + \beta_6 Room(i) * Treatment(i)$$
$$+\beta_7 Footage(i) * Treatment(i)$$

4. Results

d) ATE methods

France returns a price difference of 119,032 € without controlling for the property type. With apartment only, it decreases to 73,119€ while for houses it jumps to 426,188€. Those differences must be treated with cautious as the proportion of type in the dataset is unequal.

Germany, on the other hand, returns a price difference of 2,704€ without controlling for the property type. With apartment only, it turns negative with -95,428€ while for house it goes upside down at 112,063€.

Such differences betray a poor approach to testing the treatment effect. Indeed, with those little observations, the law of large numbers doesn't hold. As counterfactual can't be observed, the assumption that characteristics of the treatment group are, on average, identical the non-treated is hence weakly supported. In addition, using data from only one location per group is certainly prone to many biases.

e) Linear regression method:

France data returns the following table:

```
MODEL INFO:
Observations: 288
Dependent Variable: data_f$Price
Type: OLS linear regression
MODEL FIT:
F(7,280) = 124.93, p = 0.00
R^2 = 0.76
Adj. R^2 = 0.75
Standard errors: OLS
                                    Est. S.E. t val.
85461.42 40361.33 2.12 0.04
(Intercept)
data_f$Type2
                                 46856.63 50523.12 0.93 0.35
data_f$Room
                                 36022.87 19176.27 1.88 0.06
                                          582.15 2.36 0.02
data_f$Footage
                                 1376.30
data_f$Treatment1
                                -24226.12 48207.73 -0.50
                                                         0.62
data_f$Type2:data_f$Treatment1
                                64783.56 63161.52
                                                   1.03
                                                         0.31
                                         24148.55
data_f$Room:data_f$Treatment1
                                -43819.79
                                                   -1.81
                                                          0.07
data_f$Footage:data_f$Treatment1
                                  2477.30
                                          643.32
                                                   3.85 0.00
```

Figure 4 – France OLS results

Based on standard convention, only three coefficients seem to be statistically significant: β_0 (the intercept), β_2 (Footage) and β_7 (interaction term of treatment and footage). In this context, the constant term could be interpreted as the average price of land before construction and return a figure of 85,461 ϵ .

As expected, footage plays a relevant part with an average effect of $1,376 \in$ per additional square meter of built property (not to be confused with land). Finally, the interaction term of the treatment with footage shows an average effect of $2,477 \in$. This embodies the intended effect of the study.

In other words, labor mobility between Gex and Geneva has a positive impact on real estate prices in Gex. This supports the presence of a treatment effect from French working in Switzerland on the national real estate market. Before confirming this claim, the German results must be analyzed.

```
MODEL INFO:
observations: 309
Dependent Variable: data_d$Price
Type: OLS linear regression
MODEL FIT:
F(7,301) = 110.02, p = 0.00
R^2 = 0.72
Adj. R^2 = 0.71
Standard errors: OLS
                                             Est. S.E. t val.
                                       190859.14 36046.34 5.29 0.00
(Intercept)
data_d$Type2
                                        70498.13 50155.94 1.41 0.16
                                        -13847.64 14696.87 -0.94 0.35
data_d$Room
                                                               5.96 0.00
data_d$Footage
                                         3687.96 618.82
                                       -41476.85 44896.70
                                                              -0.92
data_d$Treatment1
                                                                       0.36
data_d$Type2:data_d$Treatment1
data_d$Room:data_d$Treatment1
data_d$Footage:data_d$Treatment1
                                      142112.04 64349.18
                                                                2.21
                                                                       0.03
                                       -15503.73 17513.34
                                                                -0.89
                                                                       0.38
                                         -327.98
                                                    677.46 -0.48
                                                                       0.63
```

Figure 5 – Germany OLS results

Here again, three coefficients are statistically significant: β_0 (the intercept), β_2 (Footage) and β_6 (interaction term of treatment and type). The average price of the land appears to be more than double at 190,859 ϵ . Same observation for the impact of any additional square meter at 3,687 ϵ . Finally, the interaction term of the treatment with property type rises to 142,112 ϵ . There's a lot more to draw attention here. To begin with, the negative "Room" coefficient which raises concern. Indeed, it doesn't make any sense to observe a decrease in price for any additional chamber. Relevance of those numbers should then be address cautiously. Secondly, the treatment effect appears to be negative here, even though below any statistically relevant threshold. When looking at its effect on price determinants however, footage shows a figure close to zero. According to hedonic pricing theory, external factors could explain those differences. For instance, neighborhood appealing features such as access and quality of school or public transport, presence of a downtown area, cultural offering, air quality, etc. Nevertheless, conclusion drawn for France must be challenged.

As the relevance of footage remains constant across both datasets, a plot comparison allows to visualize the difference in treatment effect.

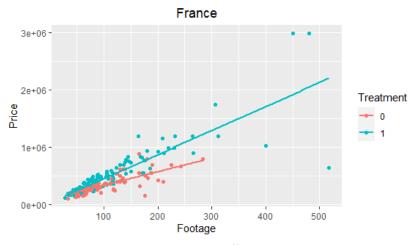


Figure 6 - Treatment effect, France

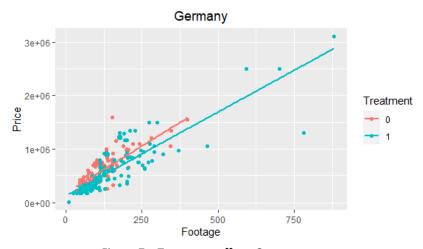


Figure 7 – Treatment effect, Germany

Despite contradictory results, it seems that this decomposition of the price makes sense. With adjusted R² of 0.75 and 0.71 for France and Germany respectively, this second method is more reliable. Represented in the graph above as the difference of regression lines' slopes, the interaction term can be distinguished as the closest form of treatment effect. Indeed, France dataset clearly shows a steeper relationship between footage and property price for the treated than the non-treated confirming regression tab's results. German's coefficients on the other hand are very similar which translates as almost parallel regression lines. To conclude, with all the assumptions made for this research design, it's not possible to assert the existence of a causal link between cross-border mobility and national property price.

5. Concluding remarks

Real estate market is highly location-dependent and one slight difference could have drastic consequences in attractivity, volume and price. While some controls were made in the city selection, countless others were forgotten. In addition, even amongst the retained criteria, finding a perfect match was impossible. Furthermore, the study didn't make distinctions between foreign workers coming to Switzerland and Swiss citizens moving across the border as both benefit from arbitrage opportunity. However, the composition of cross-border movement and its proportion would be a very interesting information to consider.

Finally, one should also keep in mind that any causality test relies on assumptions which hence exposes its conclusions to debate.

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