

Who benefits from Housing Subsidy?

Evidence from

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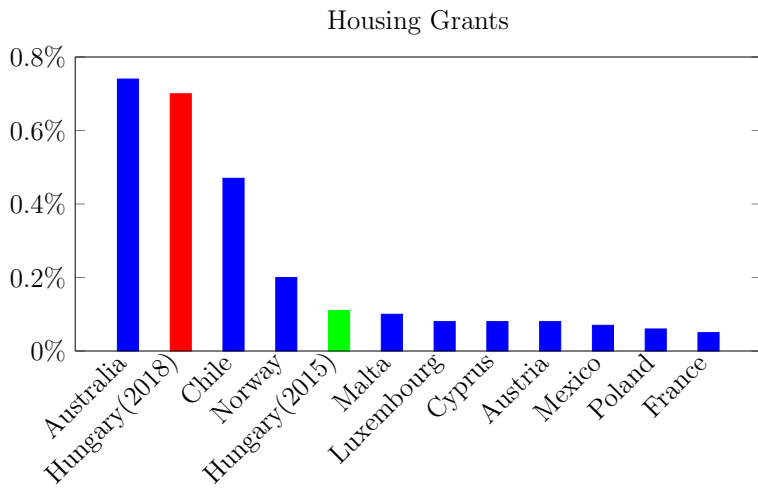
Motivation

Hungarian Expenditures on Housing Grants

1. 226 bln HUF in 2018
 - ▶ 1.2% of the budget
 - ▶ 0.7% of GDP
 - ▶ but... less than defense expenditures (427 bln HUF)
2. Second largest among 35 OECD countries (2015 numbers)
 - ▶ 16 have grant programs
 - ▶ Australia 1st (0.74%) and 3rd Chile (0.45%)

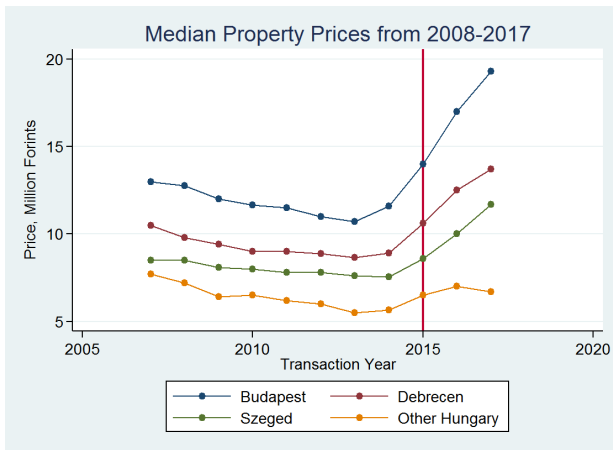
Introduction

Housing Grants as GPD Share, Selected OECD countries



Introduction

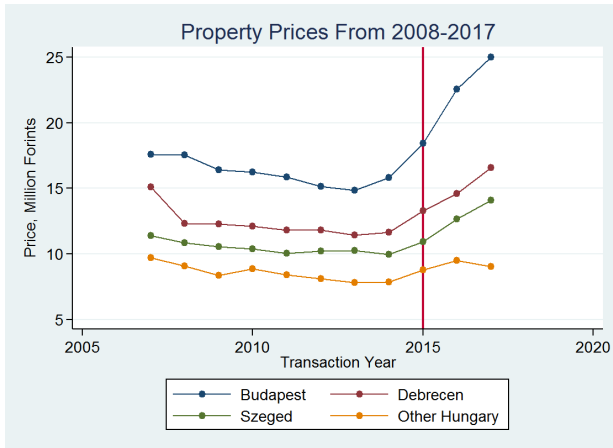
Median Property Prices in Hungary



Micro data from Hungarian Statistical Office (2017Q3)

Introduction

Average Property Prices in Hungary



Micro data from Hungarian Statistical Office (2017Q3)

Motivation

Who benefits from grants?

1. buyers or sellers?
2. more precisely
 - ▶ Buyers
 - ▶ First time buyers
 - ▶ Households with kids
 - ▶ Pensioners
 - ▶ ...
 - ▶ Supply side
 - ▶ Home Owners
 - ▶ Developers
 - ▶ ...

The answer depends on policy criteria

Housing Grant in Hungary

Family Housing Allowance (CSOK)

- ▶ from July 2015 to 2020
- ▶ supports households with children
 - ▶ $age < 25$ + enrolled in school/university
- ▶ applies to certain properties
- ▶ favors new construction
 - ▶ + VAT reduction in Jan 2016 (27% \rightarrow 5%)
- ▶ other issues
 - ▶ policy counts promised (not actual) kids
 - ▶ no criminal/tax arrears record
 - ▶ up to date with social insurance contributions
 - ▶ policy requirements changed several times

Subsidy Size

function of HH size and property characteristics

New Properties			Old Properties*	
Child	Area	Subsidy, M HUF	Area	Subsidy, M HUF
1 st	No	0.6	40m ²	0.6
2 nd	No	2.6	50m ²	1.43
3 rd	Apart.: 60m ²	10	60m ²	2.2
4 th +	House: 90m ²		70m ²	2.75

*Applies only to used properties priced below 35 M HUF

Does CSOK policy have any effect?

Empirical Strategy

Bunching or Regression Discontinuity

CSOK policy kinks:

1. 35 M HUF for used houses
2. $40m^2$, $50m^2$, $60m^2$, $70m^2$ area for used houses/apartments
3. $60m^2$ area for new apartments
4. $90m^2$ area for new houses

allow for either regression discontinuity design (RDD) or bunching.

Empirical Strategy

RDD

Consider $50m^2$ area restriction for used properties:

$$\max(\text{CSOK}_h) = \begin{cases} 1.43 \text{ M HUF,} & \text{if area} = 49m^2 \\ 2.2 \text{ M HUF,} & \text{if area} = 50m^2 \end{cases} \quad (1)$$

Proposition 1

If sellers appropriate part of CSOK, they should increase prices for $50m^2$ apartments relatively more than for $49m^2$ after CSOK implementation.

Empirical Strategy

Bunching

Consider 35 M HUF restriction for used properties:

$$\max(\text{CSOK}_h) = \begin{cases} 2.75 \text{ M HUF,} & \text{if price} = 34.9\text{M HUF} \\ 0 \text{ M HUF,} & \text{if area} = 35\text{M HUF} \end{cases} \quad (2)$$

Proposition 2

If CSOK is effective, we should observe substantially more properties purchased below the kink.

Empirical Strategy

Accounts for 1st endogeneity concern:

- ▶ State tax policy responds to price levels

but creates another one:

- ▶ Incentives for bunching
 - ▶ Example: NYC exemption
 - ▶ Retail price: \$110 → consumer price: \$119
 - ▶ Retail price: \$109.99 → consumer price: \$109.99
 - ▶ Solution — instrument from taxable income elasticity literature

More

Empirical Strategy - Instrument

Instrument: would-be tax rate applied to **predicted** item price

p_{im} :

$$p_{im} = \alpha + \gamma_{\text{category}} + \gamma_{\text{region}} + \mu_m + \text{season}_{im} + \epsilon_{im}$$

- ▶ Prediction sample: treatment states before 2000 and control states
- ▶ Category example: men's sweaters and vests

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Robustness Checks

This result is true for most subsamples:

Dependent Variable: Logarithm of Pre-tax Price				
	(1) < 2008	(2) > 2008	(3) Tax↓	(4) Tax↑
Tax Rate	-0.070* (0.037)	0.060 (0.167)	-0.003 (0.086)	0.074 (0.057)
F-statistic				
Sales Tax	0.634	0.467	0.366	0.571
Holiday	(0.630)	(3.18)	(0.967)	(0.293)
Item and month fixed effects are in all columns				
No. of Obs.	367,192	122,487	188,051	88,936
R^2	0.067	0.057	0.069	0.082
No. of Items	63,995	24,350	34,466	16,916

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

More

Robustness Checks 2

Except for some apparel groups for which demand is presumably more elastic:

Dependent Variable: Logarithm of Pre-tax Price

	(1) Men	(2) Women	(3) Non- Seasonal	(4) Seasonal
Tax Rate	-0.04 (0.077)	0.01 (0.124)	-0.21*** (0.050)	0.00 (0.077)
F-statistic				
Sales Tax	-0.729	-0.36	-0.302	-0.254
Holiday	(1.92)	(0.551)	(0.891)	(1.10)
Item and month fixed effects are in all columns				
No. of Obs.	141,911	164,016	184,579	324,193
R^2	0.036	0.121	0.010	0.075
No. of Items	11,780	27,306	15,466	7,651

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

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More

- ▶ Tax incidence (Empirics)
 - ▶ Apparel market: Besley and Rosen, 1999; Poterba, 1996
 - ▶ Other markets: Kosonen, 2015; DeCicca et al., 2013; Kopczuk et al., 2013; Harding et al., 2012; Doyle and Samphantharak, 2008
- ▶ Tax incidence (Theory)
 - ▶ Fabinger and Weyl, 2014; Anderson et al., 2001
- ▶ Elasticity of apparel expenditures:
 - ▶ Einav et al., 2014; Hu and Tang, 2014; Agarwal et al., 2013
- ▶ Sales tax and employment:
 - ▶ Burnes et al., 2013; Rohlin and Thompson, 2012; Billings, 2009; O'Keefe, 2004