# HOUSING MARKET AND MIGRATION REVISITED

A BAYESIAN MULTILEVEL GRAVITY MODEL FOR DUTCH MUNICIPALITIES

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#### Housing market and migration: why bother?

- Concerns about current housing market in the Netherlands:
  - large shortage of dwellings (especially in popular, urban, regions and for non-controlled rental housing)
  - large yearly prices increases (8% in 2018)
  - decrease in number of houses sold
  - large regional variation
- Policy debates about changes in housing market structure
  - → Increase home-ownership rates
- Large literature of external effects of home-ownership (Dietz and Haurin, 2003)
  - positive: savings, labor supply, health, maintenance, etc.
  - negative: migration (by increased moving costs) and on aggregate labour market performance (Oswald, 1996, 1999)

#### My contributions to the literature

- Large empirical (economic) literature on impact home-ownership on migration, but:
  - usually concerns marginal effect of home-ownership
  - less attention to predictions for the whole network
- Literature on impact of social renting on migration flows is scarce (De Graaff et al., 2009)
  - In the Netherlands social renting is a large phenomenon ( pprox 40% of total housing stock)
  - Social renting rights only valid within city (Boyle, 1998)
  - Social renting is an urban phenomenon ( $\approx$  40–50% in Amsterdam)

#### So, this paper

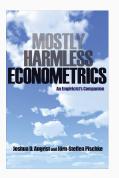
Does what? Revisits the impact of housing market structure (home-ownership and social renting rates) on within-country migration flows using a Bayesian multilevel gravity model (Congdon, 2010)

And Predict all changes in incoming and outcoming migration flows when housing market structure changes

# Background: two different cultures (Breiman, 2001)

#### In economics:

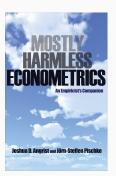
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#### Outside economics:

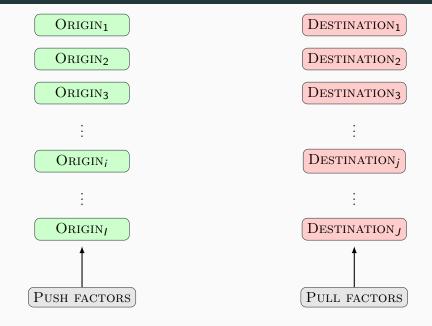
- model performance
- focuses on  $\hat{y}$
- prediction of total effect



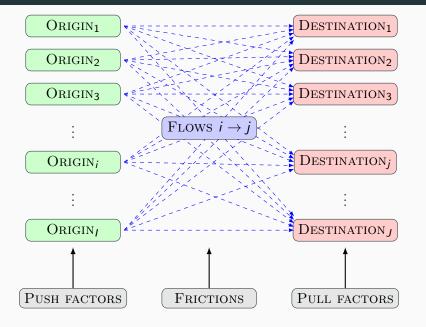
## Why a multilevel approach for the gravity model

 $\operatorname{CITY}_1$ CITY<sub>2</sub> CITY<sub>3</sub>  $CITY_i$ CITY

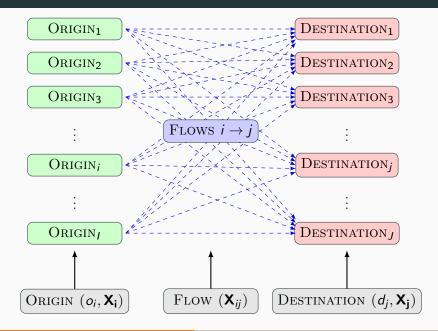
# Why a multilevel approach for the gravity model



#### Why a multilevel approach structure for the gravity model



## Why a multilevel approach for the gravity model



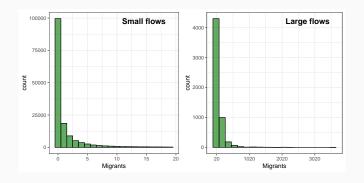
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  - no two-stage models anymore
  - precision (standard deviation) is correct

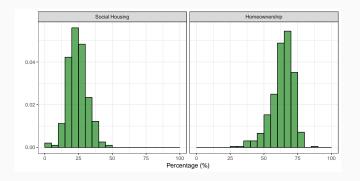
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- Partial pooling: origin and destination specific effects are draws from a distribution: usually  $\sim \text{Normal}(\alpha, \sigma)$ 
  - $\sigma \longrightarrow 0$  : complete pooling
  - $\sigma \longrightarrow \infty$  : no pooling (fixed effects)

#### **Data: migrations flows**



- Migration flows between 393 Dutch municipalities in  $2015 (\approx 150,000)$
- Variance is 4 times expectation: dispersion

#### Data: municipal housing structure



- Positive correlation between city size and social renting (0.4)
- Negative correlation between social renting and home-ownership (-0.84)

## Modeling framework: traditional gravity modeling

$$\log(\mathsf{migrants}_{ij}) = o_i + d_j + \log(\mathbf{X}_{ij})\beta + \gamma \log(\mathsf{dist}_{ij}) + \epsilon_{ij}$$

Origin and destination specific fixed effects for multilateral resistance (Anderson and Van Wincoop, 2003), but:

- what about zeros in migrants<sub>ij</sub>?
- how to incorporate housing structure in the presence of o<sub>i</sub> and d<sub>j</sub>?
- dispersion and heteroskedasticity (Silva and Tenreyro, 2006)

$$\mathsf{Migrants}_{\mathit{ij}} \sim \mathsf{GammaPoisson}(\lambda_{\mathit{ij}}, \tau)$$

(flow of migrants)

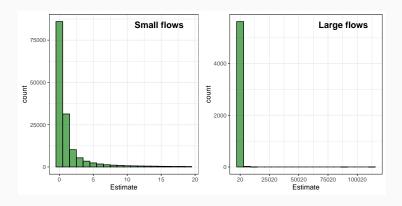
$$\begin{aligned} \mathsf{Migrants}_{ij} \sim \mathsf{GammaPoisson}(\lambda_{ij}, \tau) & \qquad & \mathsf{(flow\ of\ migrants)} \\ \mathsf{log}(\lambda_{ij}) = & \alpha + o_{\mathsf{mun}[i]} + d_{\mathsf{mun}[j]} + & \qquad & \mathsf{(regional\ effects)} \end{aligned}$$

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\begin{split} \mathsf{Migrants}_{ij} &\sim \mathsf{GammaPoisson}(\lambda_{ij}, \tau) & (\mathsf{flow} \ \mathsf{of} \ \mathsf{migrants}) \\ \log(\lambda_{ij}) = & \alpha + o_{\mathsf{mun}[i]} + d_{\mathsf{mun}[j]} + \\ & \beta_1 \log(\mathsf{pop}_i) + \beta_2 \log(\mathsf{pop}_j) + \\ & \beta_3 \log(\mathsf{home}_i) + \beta_4 \log(\mathsf{home}_j) + \\ & \beta_5 \log(\mathsf{soc}_i) + \beta_6 \log(\mathsf{soc}_j) + \\ & \beta_7 \log(\mathsf{dist}_{ij}) & (\mathsf{explanatory} \ \mathsf{variables}) \end{split}
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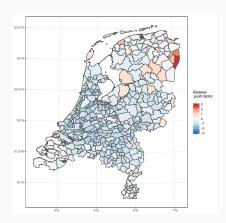
$$\begin{split} & \mathsf{Migrants}_{ij} \sim \mathsf{GammaPoisson}(\lambda_{ij}, \tau) & (\mathsf{flow of migrants}) \\ & \mathsf{log}(\lambda_{ij}) = \alpha + o_{\mathsf{mun}[i]} + d_{\mathsf{mun}[j]} + \\ & \beta_1 \, \mathsf{log}(\mathsf{pop}_i) + \beta_2 \, \mathsf{log}(\mathsf{pop}_j) + \\ & \beta_3 \, \mathsf{log}(\mathsf{home}_i) + \beta_4 \, \mathsf{log}(\mathsf{home}_j) + \\ & \beta_5 \, \mathsf{log}(\mathsf{soc}_i) + \beta_6 \, \mathsf{log}(\mathsf{soc}_j) + \\ & \beta_7 \, \mathsf{log}(\mathsf{dist}_{ij}) & (\mathsf{explanatory variables}) \\ & o_{\mathsf{mun}} \sim \mathsf{Normal}(0, \sigma_o) & (\mathsf{origin effects}) \\ & d_{\mathsf{mun}} \sim \mathsf{Normal}(0, \sigma_d) & (\mathsf{destination effects}) \\ & \alpha, \beta_1, \dots, \beta_7 \sim \mathsf{Normal}(0, 2) & (\mathsf{priors}) \\ & \sigma_o, \sigma_d \sim \mathsf{HalfCauchy}(0, 1) & (\mathsf{priors}) \\ & \tau \sim \mathsf{Gamma}(0.01, 0.01) & (\mathsf{prior}) \end{split}$$

#### Observed versus predicted data

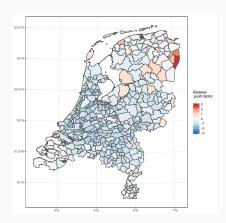


 Predictions have large outliers: caused by Rotterdam (large population, low percentage home-ownership)

#### $d_{\text{mun}}$ as measure of attractivity?



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#### Home-ownership rate in Amsterdam decreases with 10%?

#### **Conclusions**

#### Migration flows:

- ullet homeownership has an elasticity well below -1
- social renting has negative elasticity as well, but close to zero (Boyle, 1998)
- most migration dynamics outside the most popular areas

#### Bayesian multilevel gravity framework:

- flexible and powerful
- suitable for predictions within and outside sample
- with good model specification, estimation runs smoothly
- not very scalable; 154,056 obs.  $\approx$  6 hrs. of estimation

## **Supplementary materials**

Paper, presentation, data and code can be retrieved from the project's GitHub page:

https://github.com/Thdegraaff/migration\_gravity

Thank you!

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