

# The role of municipal and provincial social expenditure in reducing local income inequality: The case of Austria

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# Overview

#### Introduction

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- Data and Descriptives
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# Introduction

- Most welfare states have undergone major changes over the past decades.
- Population ageing, migration and rising within-inequality have put pressure on many welfare states.
- Generally, social expenditure per capita has been rising over the past decade:
  - EU: +23%, Austria: +25%
- Trend towards decentralization and focus on service provision.

# Decentralizing Social Policy

- In recent years: tendency of fiscal decentralization
- Sub-national governments fulfill certain tasks that meet societal needs, such as security, education, health and well-being.
- However, different understandings of the role of the state vs. role of municipalities.
  - It is argued that decentralization can improve service provision efficiency as well as quality (Hooghe and Marks 2001).
  - The effect of different levels responsible for public social expenditure on economic equality was, so far, given little attention in research.

## **Research Questions**

#### **Research Questions**

- What are the municipal, district-level and provincial characteristics influencing local income inequality?
- What are the contributions of municipal and provincial social public expenditure to the reduction of local income inequality?

# A Brief Overview of the Literature

• Economic structure and factors linked to globalization:

- Occupational switching
- Productivity of workers
- Job polarization
- Firm agglomeration...
- Demographic structure of a municipality:
  - Migration
  - Population aging
  - Share of women in the labour marktet...
- "Pro-devolution" fraction vs. sceptics
- Who does it better then?

# A Brief Overview of the Literature II

- Income inequality for disposable income is significantly lower than income inequality for market income.
  - In Austria, the Gini index for market-income is 0.36 and for disposable income 0.29.
- Public social expenditure can cushion income inequality, but investigations on the effect of different spending levels (e.g. provincial vs. municipal) are still rare.
  - Previous studies have showed that state level expenditure can lower inequality between states (see, e.g. Lee 2021, Moldogaziev et al. 2018).
  - Others focus on the role of municipal spending and hence, argue that the local level is crucial to reduce income inequalities (Lobao et al. 2021, Andreotti and Mingione 2016)
- There are a few studies that investigate the role of different governance levels in the alleviation of inequality with mixed results.

# The Austrian Case

#### • Austria is one of the world's 25 federal countries with 9 provinces.

- In terms of fiscal authority, all provinces can dispose of their own budgets.
- However, most provincial revenues stem from taxes levied by the federal government (revenue sharing system).
- Provinces have some legislative power (i.e. environmental protection, building regulations, regional planning, waste management and kindergartens).
- · Lowest level of administrative governance: municipalities
  - Typical responsibilities of municipalities: the administration of the municipal budget, spatial planning, road construction, school maintenance and provision.
  - However, there are certain voluntary tasks municipalities can fulfill, such as water and waste management, community housing and social welfare.

#### Table: Description of variables

Variable Name	Description
Municipal variables	
Gini index (dependent variable)	Gini index based on adjusted household income
Party	Dummy, Political party of the mayor, $\ddot{O}VP = Austrian People's Party (conservative), SP\ddot{O} = Social Democratic Party$
Social expenditure pc	Aggregated social expenditure per capita (10-year lag)
Average income	Average total income (all individual earnings are aggregated)
Leeway pc	Total revenue minus total expenditure and debt payments per capita
Unemployment rate	Unemployment rate (ILO definition)
Urban-rural Typology	Dummy, $1 = Urban$ center
	2 = Regional center
	3 = Rural area close to center
	4 = Rural area
Tourism	Dummy, $1 =$ more than 70 overnight visits per capita
Household size	Average householdsize
Number of children	Average number of children under 15 years
Single households with children	Share of households with one adult and at least one child under 15 years measured by all households
Dependency ratio	Share of people below 15 years and over 65 years measured by people be- tween 16 and 64 years
Share tertiary education	Share of people with tertiary education measured by total number of inhab- itants
Part time	Share of people working part time measured by all people in employment
Self-employment	Share of people that are self-employed measured by all people in employment
Female participation	Share of women in employment measured by all people in employment
Population change	Population change from 2008 to 2018

Variable Name	Description
District-level variables	
Start ups pc	Number of start ups per capita
Companies > 100  employees	Share of companies with more than 100 employees measured by all compa- nies
Secondary sector	Share of people working in the secondary sector measured by all people in employment
Tertiary sector	Share of people working in the tertiary sector measured by all people in employment
Share high-skilled sector	Share of people working in telecommunication, banking and finance, in- surance services, legal services, accounting and consulting, advertising and marketing, and engineering and architecture
Universities	Number of universities in the district
Provincial variables	
Prov. social expenditure pc RD expenditure pc GRP pc	Aggregated social expenditure per capita Provincial spending on research and development per capita Gross regional product per capita
Prov. Political party	Dummy, Political party of the governor, $\ddot{O}VP$ = Austrian People's Party (conservative), SP $\ddot{O}$ = Social Democratic Party



Figure: Distribution of ginis within provinces



Figure: Gini index (2018) and municipal social expenditure (2008)

# Method

- When conducting research on phenomena which are characterized by hierarchical data structures, employing a single-level model might lead to various estimation problems, such as the aggregation bias, misestimated precision and the "unit of analysis" problem. (Raudenbush and Bryk 2002).
- Multi-level (hierarchical) model:
  - Municipalities are nested within districts which are nested within provinces.
  - In a multi-level model each spatial level can be expressed as a sub-model.

## Method II

Level 1: 
$$y_{ikl} = \delta_{0kl} + \sum_{p=1}^{P} \delta_{pkl} x_{pikl} + e_{ikl}$$
 (1)

where  $y_{ikl}$  represents the inequality (Gini index) of municipality *i* in district *k* in province *l*,  $\delta_{0kl}$  is the intercept for district *k* in province *l*,  $x_{pikl}$  are p = 1, ..., P municipal characteristics that predict municipal inequality,  $\delta_{pkl}$  are the corresponding level-1 coefficients and  $e_{ikl}$  is a level-1 random effect referring to the deviation of municipality *ikl* from its predicted value.  $e_{ikl}$  is normally distributed with a mean of zero and variance  $\sigma^2$ .

Level 2: 
$$\delta_{pkl} = \beta_{p0l} + \sum_{q=1}^{Q_p} \beta_{pql} z_{qkl} + r_{pkl}$$
 (2)

where  $\beta_{p0l}$  is the intercept of province *l* in modelling the district effect,  $z_{qkl}$  are  $q = 1, ..., Q_p$  district characteristics,  $\beta_{ql}$  are the corresponding level-2 coefficients and  $r_{pkl}$  is a level-2 random effect following a multivariate normal distribution with mean vector of zeros and variance-covariance matrix  $\Sigma$ .

Level 3: 
$$\beta_{pql} = \phi_{pq0} + \sum_{s=1}^{S_{pq}} \phi_{pqs} v_{sl} + u_{pql}$$
 (3)

where  $\phi_{pq0}$  is the intercept term in the province-level model,  $v_{sl}$  are the  $s = 1, ..., S_{pq}$  province characteristics,  $\phi_{pqs}$  are the corresponding level-3 coefficients and  $u_{pql}$  is a random province effect following a multivariate normal distribution with mean vector of zeros and variance-covariance matrix  $\Omega$ .

# Method III

- We include both a spatial lag in the dependent and independent variables resulting in a Spatial Durbin Model (LeSage 2015).
- We define a spatial weights matrix with the dimension N × N where municipalities are defined as neighbors when:

$$w_{jm} = egin{cases} rac{1}{d_{jm}} & ext{if } j 
eq m ext{ and } d_{jm} \leq d_{max} \ 0 & ext{if } j = m ext{ and } d_{jm} > d_{max} \end{cases}$$

where  $w_{jm}$  refers to the spatial weight of the relationship between municipality j and its neighboring municipality m.  $d_{max}$  is specified as 15 minutes driving time between municipalities.

• The level one equation (1) can then be re-written as:

$$y_{j[ikl]} = \delta_{[0kl]} + \sum_{\rho=1}^{P} \delta_{\rho k l} x_{\rho j[ikl]} + \sum_{m=1}^{l} w_{jm} y_{m[ikl]} \rho + \sum_{m=1}^{l} w_{jm} x_{m[ikl]} \gamma + e_{j[ikl]}$$
(4)

where the notation stays the same as in equation (1). Additionally, the spatial weights  $w_{jm}$  and the respective spatial autoregressive parameters  $\rho$  and  $\gamma$  are included.

# Estimation

- "Bayesian Inference is a Way of Thinking, Not a Basket of 'Methods"' (Sims, 2007)
  - acknowlegdes uncertainty about the unknown parameters of a model
  - views unknown parameters as random variables
  - does so by utilizing elementary probability theory (e.g. the definition of conditional probability, Bayes' Theorem, and the law of total probability).
- The centerpiece of the Bayesian methodology is **Bayes theo**rem:

$$P(\theta|y) = \frac{P(y|\theta)P(\theta)}{P(y)}$$
(5)

- Bayesians treat p(θ|y) as being of fundamental interest: "Given the data, what do we know about θ?"
  - $p(\theta|y)$  is referred to as the posterior density.
  - $p(y|\theta)$  is the likelihood function.
  - $p(\theta)$  is the prior density.

# Estimation II

- Integrated nested Laplace approximation (INLA) first proposed by Rue et al. (2009)
  - more computationally efficient, provides more accurate results
  - available for latent Gaussian models
  - rather than estimating the joint posterior distribution of the model parameters: focusing on individual posterior marginals of the model parameters
- INLA uses a combination of analytical approximations and numerical integration to obtain approximated posterior distributions of the parameters.

## Results



## Results

Figure: Mean effect of municipal and provincial social expenditure pc and Gini indices for each province



# Conclusion and Outlook

## Findings

- Negative relationship between 10-year lag of social expenditure (on municipal as well as provincial level) and Gini index.
- There are large differences between the provinces in how a one percent increase in public spending translates into a reduction of inequality.
  - Similarities between the coefficients of municipal and provincial social public expenditure can be found. Accordingly, the average municipal effect is similar to the respective provincial effect suggesting that the effects of public spending are localized for each province.
- Relevant drivers of local inequality on all administrative levels.
  - This suggests that municipal features depend on wider regional characteristics due to the economic and social interdependence of municipalities nested in the same district or province.

# Limitations and Outlook

#### Limitations and Outlook

- Focus on income instead of wealth:
  - The effect of e.g. property as a long-term asset, protecting against short-term economic shocks and securing the social status of future generations, is not included as specific data is not available
- Prevailing measures of inequality tend to neglect that costs to cover basic needs differ between regions.
  - Residual income, measured as the post-tax disposable household income deducted by costs for necessities, could allow for a better understanding of the locally varying effect of income.

# Literature I

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# Appendix: Estimation III

- Integrated nested Laplace approximation (INLA) first proposed by Rue et al. (2009)
  - more computationally efficient, provides more accurate results
  - available for latent Gaussian models
    - = structured additive regression models. The response variable  $y_i$  is assumed to belong to an exponential family, where the mean  $\mu_i$  is linked to a structured additive predictor  $\eta_i$  through a link function  $g(\cdot)$ , so that  $g(\mu_i) = \eta_i$ .  $\eta_i$  accounts for the effects of the model parameters  $\psi$ , and x denotes a latent field,  $x = (\eta, \psi)$ .

Model: 
$$y_i | \eta_i, \theta_1 \sim N(\mu_i | \eta_i, \theta_1)$$
 (6)

Latent Gaussian prior for 
$$\eta$$
:  $\eta | \theta_2 \sim N(\mu_2(\psi \theta_2), \Sigma(\theta_2))$  (7)

Latent Gaussian prior for 
$$\psi$$
:  $\psi|\theta_3 \sim N(\mu_3(\theta_3), \Sigma(\theta_3))$  (8)

Prior for non-Gaussian 
$$\boldsymbol{\theta}$$
:  $\boldsymbol{\theta} = (\theta_1, \theta_2, \theta_3) \sim \Pi$  (9)

# Appendix: Estimation IV

- We are thus interested in the posterior marginals of the hyperparameters  $\theta$  and the latent field x.
- Aim: Approximate the marginal posteriors:
  - $\pi(x_i|\boldsymbol{ heta}, \boldsymbol{y}) \approx \tilde{\pi}(x_i|\boldsymbol{ heta}, \boldsymbol{y})$
  - $\pi(\boldsymbol{\theta}|\boldsymbol{y}) \approx \tilde{\pi}(\boldsymbol{\theta}|\boldsymbol{y})$

with the approximation

$$\tilde{\pi}(x_i|\boldsymbol{y}) = \int \tilde{\pi}(x_i|\boldsymbol{\theta}, \boldsymbol{y}) \tilde{\pi}(\boldsymbol{\theta}|\boldsymbol{y}) \ d\boldsymbol{\theta}$$
(10)

$$\widetilde{\pi}(\theta_j|\mathbf{y}) = \int \widetilde{\pi}(\boldsymbol{\theta}|\mathbf{y}) \ d\boldsymbol{\theta}_{-j}$$
(11)

 $\rightarrow$  Simplified Laplace approximation

	Mean	SD	0.025Q	0.75Q	0.975
Municipal variables					
Party = SPÖ	-0.00228	0.00019	-0.00266	-0.00215	-0.00190
log(Social expenditure pc) <sub>2008</sub>	-0.00043	0.00007	-0.00056	-0.00039	-0.00030
log(Leeway pc)	0.00001	0.00026	-0.00050	0.00018	0.00051
log(Average income)	0.02940	0.00081	0.02783	0.02885	0.03098
Unemployment rate	0.26941	0.00025	0.26892	0.26957	0.26988
Regional center	0.00506	0.00056	0.00397	0.00542	0.00621
Rural area close to center	-0.00198	0.00035	-0.00265	-0.00221	-0.00129
Rural area	-0.00058	0.00016	-0.00091	-0.00069	-0.00027
Tourism	-0.00201	0.00016	-0.00234	-0.00190	-0.00169
Household size	-0.00007	0.00030	-0.00147	0.00228	0.00268
Number of children	-0.02720	0.03100	-0.02115	0.02710	0.02892
Dependency ratio	0.03208	0.00039	0.03130	0.03234	0.03283
Share tertiary education	0.32146	0.00045	0.32057	0.32178	0.32232
Part time	0.06477	0.00042	0.06395	0.06505	0.06557
Female employment	-0.09563	0.00089	-0.09732	-0.09504	-0.09390
Self employment	0.35369	0.00070	0.35231	0.35417	0.35507
Population change	0.00779	0.00060	0.00662	0.00820	0.00898
Single households with children	0.22248	0.00014	0.22222	0.22257	0.22274

#### Table: Results: Mean, standard deviation and quantiles of coefficients

	Mean	SD	0.025Q	0.75Q	0.975
Municipal variables					
$W \times \log(Social expenditure pc)_{2008}$	-0.00069	0.00070	-0.00086	-0.00039	-0.00030
$W \times \log(\text{Leeway pc})$	0.00001	0.00073	-0.00138	0.00049	0.00135
$W \times log(Average income)$	0.00743	0.00045	0.00658	0.00773	0.00832
$W \times Unemployment rate$	0.05659	0.00012	0.05635	0.05682	0.05721
$W \times$ Household size	-0.00148	0.00042	-0.00270	-0.00106	0.00028
$W \times Number of children$	0.08752	0.00024	0.08706	0.08768	0.08797
$W \times Dependency ratio$	0.00324	0.00066	0.00197	0.00368	0.00452
$W \times$ Share tertiary education	0.13358	0.00073	0.13104	0.13504	0.13599
$W \times Part time$	0.07343	0.00013	0.07318	0.07369	0.0.07412
$W \times Female employment$	-0.11924	0.00088	-0.12101	-0.00733	-0.00629
$W \times Self employment$	0.10671	0.0.0065	0.10546	0.10794	0.11256
$W \times Population change$	-0.03004	0.00068	-0.03134	-0.02957	0.00287
W $ imes$ Single households with children	-0.17313	0.00077	-0.17465	-0.17253	-0.17167
Observations			2,094		

#### Table: Results: Mean, standard deviation and quantiles of coefficients

	Mean	SD	0.025Q	0.75Q	0.975
District-level variables					
Start ups pc Companies > 100 employees Secondary sector Tertiary sector Share high-skilled sector Universities	0.00005 0.00001 -0.08291 -0.03021 0.00032 0.00168	0.00070 0.00025 0.00033 0.00021 0.000018 0.00015	-0.00131 -0.00048 -0.08355 -0.03062 0.00009 0.00139	0.00053 0.00017 -0.08269 -0.03007 0.00047 0.00177	0.00145 0.00051 -0.08228 -0.02979 0.00057 0.00197
Observations			93		
Provincial variables					
Prov. Political party = SPÖ log(RD expenditure pc) log(GRP pc) log(Prov. social expenditure pc) <sub>2008</sub>	-0.00778 0.00829 0.02999 -0.00016	0.00023 0.00024 0.00030 0.00004	-0.00821 0.00782 0.02939 -0.00024	-0.00763 0.00845 0.02979 -0.00012	-0.00732 0.00877 0.03054 -0.00007
Observations			8		

Significant variables are indicated in **bold**.

Variabes	Direct	Indirect	Total
$Party = SP\ddot{O}$	-0.002338	-0.001837	-0.004175
log(Social expenditure pc) <sub>2008</sub>	-0.000030	-0.000009	-0.000039
log(Average income)	0.003172	0.002200	0.005372
log(Leeway pc)	0.000001	0.000002	0.000003
Unemployment rate	0.299726	0.099752	0.399478
Regional center	0.004191	-0.010564	-0.006373
Rural area close to center	0.001350	-0.009683	-0.008333
Rural area	0.000404	-0.008742	-0.008338
Tourism	-0.002649	-0.004180	-0.006829
Household size	0.000206	-0.019780	-0.019574
Number of children	-0.025600	0.092201	0.066601
Dependency ratio	0.031779	0.006477	0.038256
Share tertiary education	0.318672	0.066369	0.385040
Part time	0.080727	0.044671	0.125398
Female employment	-0.095139	-0.111670	-0.206809
Self-employment	0.375093	0.267935	0.643027
Population change	0.002632	-0.000474	0.002158
Single households with children	0.280366	-0.046524	0.233842
Rho		0.25	

#### Table: Mean direct, indirect and total effects

Note: Significant variables are indicated in **bold**.

