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# Empirical link between firmlevel indicators and target variables of competitiveness: Bayesian Model Averaging approach

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#### Motivation to incorporate firm level information

- Di Giovanni and Levchenko (2010): aggregate competitiveness outcomes driven by the largest and most productive firms, not by the average firm
  - Focusing on average/representative firm may yield biased policy conclusions
- Melitz and Redding (2013): firms at opposite tails of the distribution react differently to policy intervention
  - Firm heterogeneity may explain cross-country differences in policy outcomes
- Di Mauro and Pappada (2014): real exchange rate movements are underestimated when cross-country differences in productivity distributions are ignored
- Berman et al. (2012): firm heterogeneity affects the response to exchange rate movements

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## **Econometric Approach**

## Panel regression wit country- and time-fixed effects

$$y_{it} = \alpha + x'_{i,t-1}\beta + \gamma_i + \delta_t + \varepsilon_{i,t}$$

 Problem of model uncertainty arises due to large set of candidate regressors

#### **BMA** approach

- · Estimate models for all possible covariate combinations
- Posterior model probabilities conditional on the data for a given model yield
  - Posterior inclusion probability (PIP) for a given variable
  - Model-weighted posterior mean coefficient and standard deviation
- Two criteria to confirm statistical significance of our indicators:
  - 1. PIP > prior inclusion probability
  - 2. Posterior mean is different from zero at 10% significance level

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## Key results: regression with HCI interaction terms

		OP per capita	TFP (Solow residual)		Export market shares	
	8	rowth	g	rowth	growth	
	PIP	Posterior	PIP	Posterior	PIP	Posterior
	PIP	Mean	PIP	Mean	PIP	Mean
Exogenous variables (first lag):						
HCI (ULC based)	0.478	-0.214**	0.720	-0.297***	0.830	-0.355***
x Labour productivity, skewness	0.289	0.147**	0.196	0.152*	0.336	0.183
x TFP, IQR	0.840	0.223***	0.441	0.173**	0.503	0.204**
x Capital intensity, IQR	0.219	-0.125°	0.059	-0.046	0.174	-0.037
Change in GVC position	0.154	0.126	0.132	0.135	0.862	0.401***
New overlap with China	0.700	0.518***	0.075	0.082	0.288	0.268
Labour with tertiary education	0.184	-0.080°	0.449	-0.314***	0.224	0.124
Labour productivity, skewness	0.361	0.174**	0.204	0.156**	0.225	0.089
Existent overlap with China	0.343	-0.449	0.216	-0.367°	0.211	0.085
Labour with secondary education	0.175	0.093**	0.144	0.131	0.413	0.233**
TFP, IQR	0.144	0.070	0.258	0.157*	0.320	0.160*
SAFE index	0.427	-0.203**	0.063	-0.032	0.204	-0.093
Part-time employment	0.317	0.185**	0.114	0.133	0.175	-0.011
GVC position	0.253	-0.177*	0.097	-0.111	0.253	-0.132
RCA in high-tech industries	0.325	0.135**	0.058	0.023	0.173	-0.013
Labour productivity, IQR	0.141	0.084*	0.180	0.131**	0.218	0.104
Legal system and property rights	0.230	-0.124**	0.096	-0.092	0.169	-0.034
Time dummies:						
2004	0.578	-0.242***	0.083	-0.087	0.311	-0.186*
2005	0.205	0.108	0.056	-0.012	0.207	-0.068
2006	0.535	0.214**	0.128	0.156	0.381	0.163**
2007	0.515	-0.177***	0.866	-0.262***	0.205	-0.021
2008	0.957	-0.427***	0.988	-0.597***	0.303	0.216*
2009	0.696	0.337***	0.657	0.283***	0.849	-0.505***
2010	0.178	0.130	0.129	-0.022	0.275	0.158
2011	0.262	-0.065	0.388	-0.198**	0.698	-0.278***
Summary statistics:						
Mean number of regressors	11.740		8.2515		13.067	
Model space		00E+13		0E+13		0E+13
Number of models visited		579302		49411		326723
Percent of model space visited	0.0	1000038	0.0	000030	0.0	000063
Percent of total PMP covered by top 10,000 models		18		45		12
Correlation between analytical and sample PMP	0.9939		0.9996		0.9901	
Average posterior shrinkage factor	(	0.9590	0	.9640	0	.8120
Number of observations		75		75		75

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#### Introduction

## Research topic

- 1. What is the empirical link between firm-level indicators and competitiveness outcomes on a macro level?
- 2. What can we learn from firm-level distributions?
- 3. How important are firm-level indicators above and beyond macroeconomic variables in explaining competitiveness outcomes?

### **Empirical approach**

- Panel model for 9 EU-countries over the period 2003 to 2011 (Belgium, Estonia, Finland, France, Germany, Italy, Lithuania, Slovenia, Spain)
- Novel CompNet dataset bridging the macro and micro dimension
- Econometric approach: Bayesian model averaging (BMA)

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#### Dataset

#### Three dependent variables

 Target variables of Competitiveness: Real GDP/capita growth, TFP growth, Export market share growth

#### **Independent variables**

- 1. CompNet firm-level indicators
  - Inter-quartile range (IQR) and skewness for firm size (# of employees),
    TFP growth, labour productivity growth, capital intensity
  - Share of credit constrained firms (SAFE Index)
- 2. CompNet macroeconomic indicators
  - Global value chains (GVC), revealed comparative advantage (RCA) in high-tech industries, competitive pressures from China
- 3. Traditional indicators
  - Macroeconomic environment, labour market, institutional and legal framework, human capital, demographics

Real GDP per capita TFP (Solow residual) Export market shares

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## Key results: benchmark regression

		growth		growth		growth	
	5						
	PIP	Posterior	PIP	Posterior	PIP	Posterior	
	PIP	Mean	PIP	Mean	PIP	Mean	
Exogenous variables (first lag):							
HCI (ULC based)	0.936	-0.283***	0.887	-0.324***	0.962	-0.416**	
Change in GVC position	0.332	0.180*	0.153	0.136	0.907	0.408**	
Existent overlap with China	0.790	-0.776***	0.341	-0.473**	0.216	-0.006	
Labour productivity, skewness	0.578	0.196***	0.348	0.179**	0.287	0.126	
Labour with tertiary education	0.291	0.174	0.450	-0.313***	0.227	0.093	
Part-time employment	0.571	0.238**	0.162	0.151	0.182	-0.002	
SAFE index	0,606	-0.278**	0.069	-0.014	0.222	-0.099	
New overlap with China	0.462	0.391**	0.076	0.023	0.297	0.275	
Labour with secondary education	0.203	0.030	0.175	0.138	0.452	0.237**	
RCA in high-tech industries	0.562	0.158**	0.065	0.022	0.186	0.002	
GVC position	0.452	-0.260*	0.094	-0.086	0.251	-0.087	
Labour productivity, IQR	0.183	0.051	0.219	0.135**	0.250	0.115	
Temporary employees	0.314	0.140*	0.088	0.070	0.178	0.011	
Legal system and property rights	0.271	-0.113*	0.091	-0.073	0.185	-0.020	
Time dummies:							
2004	0.620	-0.230**	0.116	-0.103	0.369	-0.2084	
2005	0.225	0.084	0.073	-0.049	0.243	-0.106	
2006	0.744	0.281***	0.195	0.213	0.392	0.161**	
2007	0.425	-0.050	0.828	-0.255***	0.225	-0.026	
2008	0.805	-0.335**	0.961	-0.576***	0.335	0.232	
2009	0.668	0.446**	0.701	0.299***	0.868	-0.510**	
2010	0.410	0.403	0.172	0.091	0.305	0.174	
2011	0.437	0.244	0.374	-0.125*	0.719	-0.268**	
Summary statistics:							
Mean number of regressors		13.752		7.920		11.9023	
Model space		2.70E+11		2.70E+11		2.70E+11	
Number of models visited	25	2907114		2154804		4433389	
Percent of model space visited	- (	0.0011		0.00078		0.0016	
Percent of total PMP covered by top 10,000 models	1	19		56		18	
Correlation between analytical and sample PMP	(	0.9973		0.9999		0.9989	
Average posterior shrinkage factor	- (	0.9566		0.9646		0.8235	
Number of observations		75		75		75	

## Conclusions

#### Conclusions

- Firm-level information has significant explanatory power above and beyond traditional macroeconomic variables
- Real GDP/capita and TFP growth are driven by the most productive firms in the economy
  - Skewness of labour productivity is one of the most important indicators in the BMA analysis
- Tighter financial constraints of firms tend to dampen real GDP/capita growth
- Real effective exchange rate is the single most important driver for all three target variables of competitiveness
- Response to exchange-rate movements depends crucially on distribution of firm size and productivity
  - Fatter right tale of the productivity distribution is associated with a smaller impact of HCI, i.e. highly productive firms are less vulnerable to changes in relative costs.
- Results are robust across different Bayesian prior assumptions