

Innovation and Access to Finance *(Capacitatea de inovare și accesul la finanțare)*

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Innovation and Access to Finance : Theoretical aspects

❖ Innovation drivers

- Supply-side factors (e.g. FDI) (Romer, 1990; Weitzman, 1998)
- Demand-side factors (e.g. entrepreneurship) (Schmookler, 1966)
- Structural factors (e.g. quality of institutions) (Acemoglu and Robinson, 2000)
- Financial constraints influence innovation capabilities (Hall, 2008; Hottenrott and Peters, 2012)

❖ Access to finance

- ✓ External finance plays a significant role in firm-level innovation (Ayyagari et al., 2011; Brown et al., 2012; Fernandez, 2017)
- ✓ Accessibility to finance is not uniformly available in all economies (Hewa et al., 2019; Lee et al., 2015)
- ✓ Bank financing has the most significant effect on firm-level innovation compared to non-bank financing (Wellalage and Fernandez, 2019)
- ✓ Most empirical works show that the access to finance enhances innovation activities (long-term horizon, affordable interest rates, etc.)

Questions, research hypotheses and contributions

❖ Questions

- How credit allocation fluctuations influence the innovation capacity (and related inputs/outputs)?
- How the innovation is impacted by positive and negative shocks in credit activity?
- Has the access to finance the same impact on high- or low-level innovation economies?

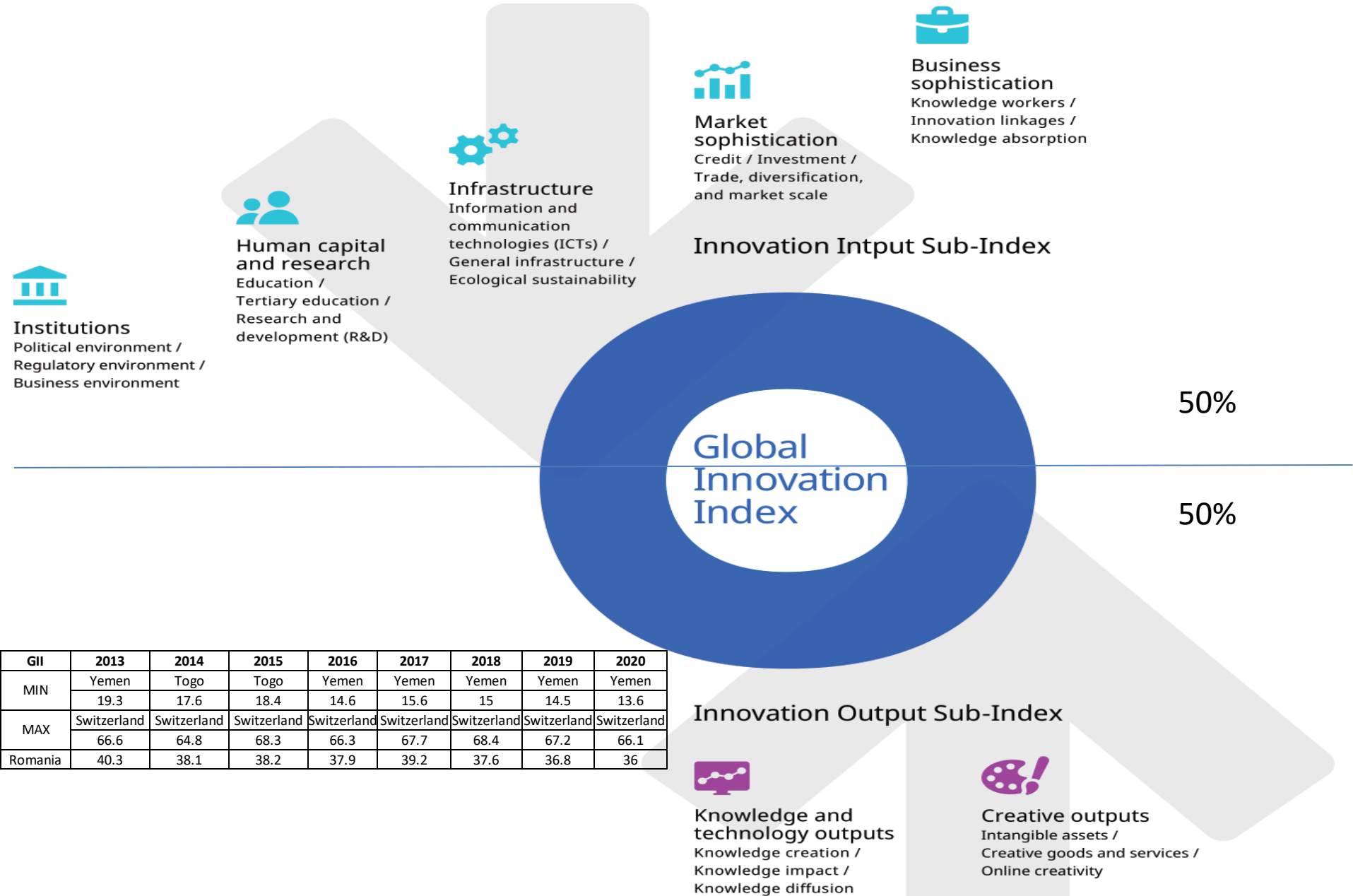
❖ Research hypotheses

- H1: The impact of credit activity fluctuations on innovation, can be either positive or negative (depending on the type of innovation – opportunity driven vs necessity driven)
- H2: The response of innovation capacity to credit activity fluctuations is asymmetric, depending on the level of innovation
- H3: The financial shocks have a stronger impact on innovation outputs compared to the innovation inputs

❖ Contributions

- Panel data quantile regression for 117 countries, over the period 2013-2020 (WIPO and WB data), bootstrap regression and fixed effect analysis for robustness (Canay, 2011)
- A new approach used for identifying positive and negative shocks in the credit activity (credit to GDP ratio decomposition)

Global Innovation Index and Sub-Indexes



GII	2013	2014	2015	2016	2017	2018	2019	2020
MIN	Yemen	Togo	Togo	Yemen	Yemen	Yemen	Yemen	Yemen
	19.3	17.6	18.4	14.6	15.6	15	14.5	13.6
MAX	Switzerland	Switzerland	Switzerland	Switzerland	Switzerland	Switzerland	Switzerland	Switzerland
	66.6	64.8	68.3	66.3	67.7	68.4	67.2	66.1
Romania	40.3	38.1	38.2	37.9	39.2	37.6	36.8	36

Data and methodology

❖ Data

- Global Innovation Index, Innovation Input Sub-index, Innovation Output Sub-index (WIPO database) – 117 countries with complete data for 2013-2020
- Credit activity (Domestic credit to private sector- of GDP) (WB database)
- GDP per capita PPP (international \$) in natural log – control
- 909 observations

❖ Methodology

- Quantile regression panel data models

- General equation

$$GII_{it} = \alpha_0 + \alpha_1 Credit_{it} + \alpha_2 GDP_{it-1} + \mu_i + \gamma_t + \varepsilon_{it}$$

- Panel bootstrap regression (Kapetanios, 2008)
- Fixed effects panel-quantile regression (Canay, 2011) - robustness

- Credit activity shocks

$$Credit_{it}^+ = IF(Credit_{it} > AVERAGE(Credit_{it-1}:Credit_{it-4}); Credit_{it} - AVERAGE(Credit_{it-1}:Credit_{it-4}); 0)$$

$$Credit_{it}^- = IF(Credit_{it} < AVERAGE(Credit_{it-1}:Credit_{it-4}); Credit_{it} - AVERAGE(Credit_{it-1}:Credit_{it-4}); 0)$$

Results (1/3)

❖ GII and credit activity

➤ Credit activity fluctuations (SD – rolling window $t : t-4$)

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_SD	0.0238	0.073	0.181**	0.247**	0.341***	0.308***	0.226***	0.084	0.017	-0.073
GDP	4.408***	5.251***	6.645***	7.497***	7.999***	8.270***	8.567***	8.872***	9.149***	9.756***
intercept	-16.39***	-21.53***	-32.40***	-38.83***	-42.18***	-43.21***	-43.89***	-43.67***	-43.16***	-43.68***

➤ Positive shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_PS	-0.092*	-0.056	-0.070	-0.094	-0.035	-0.075	-0.044	0.016	-0.045	-0.036
GDP	4.885***	5.316***	6.701***	7.705***	8.461***	8.535***	8.824***	8.845***	9.274***	9.735***
intercept	-20.21***	-21.56***	-31.98***	-39.42***	-45.01***	-43.86***	-44.91***	-42.88***	-43.99***	-43.55***

➤ Negative shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_NS	0.025*	0.023	0.050	-0.001	-0.041	-0.089**	-0.076**	-0.068**	-0.054**	-0.022
GDP	4.762***	5.360***	6.744***	7.623***	8.424***	8.345***	8.692***	8.798***	9.149***	9.666***
intercept	-18.88***	-21.96***	-32.47***	-38.95***	-44.83***	-42.62***	-43.71***	-42.74***	-43.21***	-43.14***

➤ Hypotheses validation

- ✓ H1 – for middle-level innovation countries, fluctuations in credit activity enhance innovation whereas the negative shocks negatively impact innovation at middle and upper quantiles (mixed findings)
- ✓ H2 – the impact of access to finance on innovation capacity is asymmetric

➤ Robustness checks confirm the results: (i) fixed effects quantile regression, (ii) no. of bootstraps (200 instead of 500)

Results (2/3)

❖ GII inputs and credit activity

➤ Credit activity fluctuations (SD – rolling window $t : t-4$)

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_SD	-0.107	0.091	0.150**	0.107*	0.148**	0.207***	0.129**	-0.050	-0.061	-0.001
GDP	5.688***	6.553***	7.534***	8.007***	8.439***	8.325***	8.590***	9.688***	10.20***	9.704***
intercept	-19.86***	-26.08***	-33.13***	-35.51***	-38.47***	-36.33***	-37.03***	-43.90***	-45.49***	-37.05***

➤ Positive shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_PS	-0.054	-0.163***	-0.237***	-0.192**	-0.090	-0.029	0.024	-0.002	0.022	-0.000
GDP	5.525***	6.833***	7.852***	8.331***	8.503***	8.500**	8.895***	9.525***	10.19***	9.703***
intercept	-18.52***	-27.51***	-34.79***	-37.57***	-38.20***	-36.80***	-39.13***	-42.81***	-45.64	-37.05***

➤ Negative shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_NS	-0.004	0.007	0.035	-0.012	-0.023***	-0.038***	-0.056	-0.035	-0.010	0.002
GDP	5.365***	6.650***	7.674***	8.049***	8.539***	8.423***	8.590***	9.386***	10.18***	9.705***
intercept	-17.24***	-26.60***	-33.78***	-35.43***	-38.87***	-36.34***	-36.53***	-41.79***	-45.54***	-37.06***

➤ Hypotheses validation

- ✓ H1 – for middle-level innovation countries, fluctuations in credit activity enhance innovation inputs
- ✓ H1 - Positive shocks negatively impact innovation at lower quantile whereas the negative shocks negatively impact innovation at middle and upper quantiles (mixed findings)
- ✓ H2 – the impact of access to finance on innovation capacity is asymmetric

➤ Robustness checks confirm the results: (i) fixed effects quantile regression, (ii) no. of bootstraps (200 instead of 500)

Results (3/3)

❖ GII **outputs** and credit activity

➤ Credit activity fluctuations (SD – rolling window $t : t-4$)

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_SD	0.091	0.117	0.0217**	0.312**	0.399***	0.319***	0.279***	0.172**	0.131	0.074
GDP	4.344***	5.096***	6.168***	7.243***	7.607***	7.506***	7.810***	8.307***	8.815***	9.327***
intercept	-25.53***	-29.03***	-36.60***	-44.15***	-46.13***	-42.90***	-43.23***	-44.98***	-46.41**	-44.99***

➤ Positive shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_PS	-0.089	-0.085	-0.083	-0.108	-0.038	-0.058	-0.071	-0.091	-0.030	0.170
GDP	4.199***	5.198***	6.483***	7.660***	8.222***	8.077***	8.476***	8.729***	9.033***	9.674***
intercept	-23.74***	-29.44***	-38.34***	-46.61***	-50.11***	-46.25***	-47.77***	-47.94***	-47.57***	-48.21***

➤ Negative shocks in credit activity

quantiles	lower quantiles			middle quantiles				upper quantiles		
	0.05	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.85	0.95
Credit_NS	0.039*	0.041	0.050	-0.016	-0.047	-0.113**	-0.106***	-0.087***	-0.090***	-0.051**
GDP	4.468***	5.184***	6.605***	7.677***	8.140***	7.987***	8.085***	8.589***	8.778***	9.147***
intercept	-26.16***	-29.25***	-39.58***	-47.15***	-49.56***	-46.00***	-44.85***	-47.26***	-45.90***	-43.47***

➤ Hypotheses validation

- ✓ H1 – for middle-level innovation countries, fluctuations in credit activity enhance innovation inputs; negative shocks negatively impact innovation at middle and upper quantiles (mixed findings)
- ✓ H2 – the impact of access to finance on innovation capacity is asymmetric
- ✓ H3 – the impact of credit activity is higher on innovation outputs compared with innovation inputs

➤ Robustness checks confirm the results: (i) fixed effects quantile regression, (ii) no. of bootstraps (200 instead of 500)

Conclusions

❖ Summary of results

- Credit activity volatility positively impact the national innovation capabilities for middle-level innovative countries
- No clear impact of positive shocks in credit activity
- Negative in credit activity shocks have a negative impact on innovation in middle- and upper-level innovative countries
- The negative effect is stronger in the case of innovation outputs compared with innovation inputs

❖ Results' interpretation and policy implications

- Credit fluctuations generates an increased uncertainty regarding the access to finance for middle-level innovative countries; in this context, the innovation activities intensify in order to identify new profit sources, to prevent profit decreases in the context of credit tightening, etc.
- No significant effect of the access to finance on innovation in the low-level innovative countries (less developed financial systems, the access to external financing sources is reduced)
- For middle- and upper-level innovative countries it is important to avoid negative shocks in credit activity (a credit crunch negatively impact knowledge and technology outputs, as well as creative inputs).

Smart Diaspora 2023

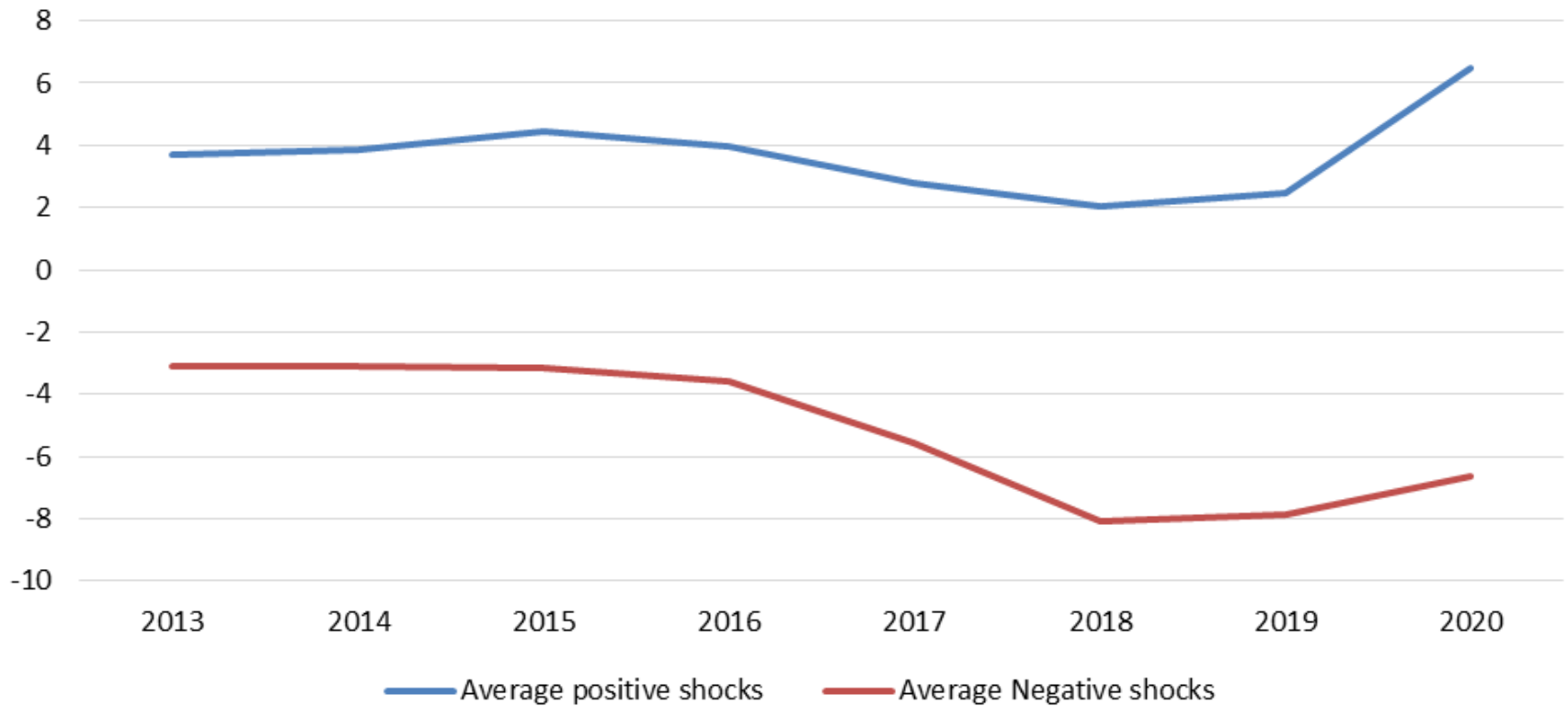
10 - 13 Aprilie 2023, Timișoara



Thank you for your attention!

Appendix A

Credit activity shocks



Appendix B

➤ Summary statistics

Variables	Mean	Std. Dev.	Min	Max
GII	37.05	11.77	13.60	68.40
GII_IN	44.32	11.95	19.90	74.20
GII_OUT	29.77	12.45	6.400	68.60
CREDIT_SD	4.811	5.206	0.127	56.21
CREDIT_PS	3.706	6.038	0.000	53.55
CREDIT_NS	-5.156	16.47	-170.3	0.000
GDP_CAP	9.618	1.100	6.957	12.00

➤ Panel unit root tests

Tests	Fisher-type panel unit root tests			Levin, Lin & Chu
	W-stat	ADF	PP	t*
GII	4.922	148.9	235.7	-8.236***
GII_IN	-0.035	239.4	371.3***	-9.559***
GII_OUT	4.848	169.0	282.4**	-2.227**
CREDIT_SD	-8.767***	442.2***	222.8	-49.37***
CREDIT_PS	-3.206***	239.3***	167.2	-10.96***
CREDIT_NS	-3.210***	191.2***	114.1	-27.89***
GDP_CAP	7.946	141.3	169.1	-0.730