

Proximity as a Source of Comparative Advantage

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Introduction

- Classic Ricardo: specialize in good in which relatively more productive
- Many-good many-country Ricardo [*Costinot(2009)*]: export relatively more in sector in which relatively more productive
- Ranking of relative technology stocks establishes ranking of relative sectoral exports [*CDK(2012)*]
- BUT: increasingly international production unbundling
- Contribution of input cost channel to define comparative advantage?

Ricardo in the data

- What's inside the black box of technology? [*Chor*(2010)]
- Complementarity specific country-sector characteristics
- Welfare analysis: trade in inputs magnifies gains from trade [*EK*(2002), *CP*[2012]]
- BUT: what about role of inputs in determining pattern of trade?
- Country-sector complementarity in different dimension:
 - Cost of inputs matters more in certain sectors
 - Countries can be ranked in terms of proximity to suppliers

What this paper does

- Uses stylized model to spell out mechanism through which inputs may become source of comparative advantage
- Derives theoretically grounded measure of proximity *to suppliers*
- Shows that this proximity characteristic creates wedge in the cost of inputs across countries
- Verifies in data that input cost channel co-determines intersectoral specialization
- Quantifies contribution of the input cost channel relatively to technology



Production function

- Finite number of sectors k
- Within sector: infinite countable number of varieties
 $\alpha \in A \equiv \{1, \dots, \infty\}$
- Variety production function Cobb-Douglas (inputs & labor)

$$\omega_i^k = \nu_i^{1-\zeta^k} P_i^{\zeta^k} \epsilon^k$$

where ζ^k is 'input intensity' characteristic of sector

- Landed cost given by

$$c_{ij}^k(\alpha) = \frac{\omega_i^k \tau_{ij}^k}{z_i^k(\alpha)}$$

- z drawn from Frechet: $Prob[Z > z] = 1 - \exp\left[-(z/z_i^k)^{-\theta}\right]$

Price indices

- Perfect competition: least cost variety bought

$$p_j^k(\alpha) = \min_i [c_{ij}^k(\alpha)]$$

- Sectoral price index in the destination across all exporters

$$E [p_j^k(\alpha)^{1-\sigma}] = (P_j^k)^{1-\sigma} = \Gamma [\Phi_j^k]^{-(1-\sigma)/\theta}$$

1. $\Gamma = \Gamma [(\theta + 1 - \sigma)/\theta]$
 2. $\Phi_j^k = \sum_{i \in I} [c_{ij}^k]^{-\theta}$
 3. $c_{ij}^k = \omega_i^k \tau_{ij}^k / z_i^k$, with z_i^k fundamental sectoral productivity
- Overall price index (cost of input bundle):

$$P_i = \prod_{k=1}^K P_i^k \gamma^k$$

- Sectoral trade share: $\pi_{ij}^k = [c_{ij}^k]^{-\theta} / \Phi_j^k$



Proximity characteristic

- Use definition of sectoral price index

$$P_j^k = \kappa \left[\Phi_j^k \right]^{-1/\theta}$$

- To write:

$$P_j^k = \kappa \left[\bar{\Phi}^k \right]^{-1/\theta} \left\{ \sum_{n=1}^N \tau_{nj}^\theta \pi_{nj}^k \right\}^{1/\theta}$$

- Use definition of overall price index: $P_j = \prod_{k=1}^K \left[P_j^k \right]^{\gamma^k}$

- To write:

$$P_j = \kappa \prod_{k=1}^K \left[\bar{\Phi}^k \right]^{-\gamma^k/\theta} \prod_{k=1}^K \left\{ \sum_{n=1}^N \tau_{nj}^\theta \pi_{nj}^k \right\}^{\gamma^k/\theta}$$

Industry-specific cost component

- Cost of input bundle consists of:
 - world's best practice across sectors
 - destination-specific proximity to suppliers:
→ trade costs weighed by probability this supplier is least cost
- Industry-specific cost component ω^k :

$$\underbrace{\omega_j^k = \epsilon^k \kappa^{\zeta^k} \left\{ \prod_{s=1}^S [\bar{\Phi}^s]^{-\gamma^s/\theta} \right\}^{\zeta^k}}_{\text{sector-specific}}$$

$$\underbrace{\left[\nu_j^k \right]^{1-\zeta^k} \left\{ \prod_{s=1}^S \left[\sum_{n=1}^N \tau_{nj}^\theta \pi_{nj}^s \right]^{\gamma^s/\theta} \right\}^{\zeta^k}}_{\text{exporter-sector-specific}}$$

Pattern of RCA

- Relative sectoral exports to market j

$$\ln \left\{ X_{ij}^k / X_{i'j}^k \right\} = \theta \left[\ln \frac{z_i^k}{z_{i'}^k} - (1 - \zeta^k) \ln \frac{\nu_i^k}{\nu_{i'}^k} - \ln \frac{\tau_{ij} \tau_i^{E,k}}{\tau_{i'j} \tau_{i'}^{E,k}} \right] \\ + \theta \left[-\zeta^k \ln \left\{ \frac{\prod_{s=1}^S \left[\sum_{n=1}^N \tau_{ni}^\theta \pi_{ni}^s \right]^{\gamma^s / \theta}}{\prod_{s=1}^S \left[\sum_{n=1}^N \tau_{ni'}^\theta \pi_{ni'}^s \right]^{\gamma^s / \theta}} \right\} \right]$$

- Proximity: $\overline{PROX}_i^M = 1 / \prod_{s=1}^S \left\{ \sum_{n=1}^N \pi_{ni}^s \tau_{ni}^\theta \right\}^{\gamma^s / \theta}$.
- Four exporter-sector cost components: technology, wages, proximity, export costs
- Retrieved in estimation relatively benchmark country and sector: exporter-sector dummy

Estimation: Three-step procedure

- **First step:** retrieve exporter-sector dummies (cross-section)

$$X_{ij,t}^k = \exp \left\{ fe_{ij,t} + fe_{j,t}^k + fe_{i,t}^k + \zeta_{ij,t}^k \right\}$$

- Dummy contains cost components specific to exporter-sector:

$$\widehat{fe}_{i,t}^k = \theta \ln(z_{i,t}^k) - \theta(1 - \zeta^k) \ln \nu_{i,t}^k - \theta \zeta^k \ln(P_{i,t}) - \theta \ln(\tau_{i,t}^{E,k})$$

- **Second step:** estimate model parameters (all years pooled)

$$\widehat{fe}_{i,t}^k = \theta \left[\ln \widehat{z}_{i,t}^k - (1 - \zeta^k) \ln \widehat{\nu}_{i,t}^k \right] + fe_t + \lambda_{it}^k$$

$\widehat{z}_{i,t}^k$: TFP; $\widehat{\nu}_{i,t}^k$: wages (instrumented)

Three-step procedure (contd.)

- Residual of second step $\widehat{\lambda}_{it}^k$ contains:
 - index of trade frictions incurred in sourcing inputs (proximity)
 - trade cost paid to get domestic varieties to world markets
- **Third step:** proximity mechanism in residual component?
- Split sample by proximity & form pairwise sectoral residuals
- Interact relative proximity with sectoral input intensity
- Look at sign and significance of β_1 (pooled data)

$$\frac{1}{\widehat{\theta}} \left[\widehat{\lambda}_{i,t}^k - \widehat{\lambda}_{i',t}^k \right] = \beta_0 + \beta_1 \ln \left\{ \left(\frac{\widehat{PROX}_{i,t}^M}{\widehat{PROX}_{i',t}^M} \right)^{\zeta^k} \right\} \\ + fe_{i,t} - fe_{i',t} + \eta_{ii',t}^k$$



Data: 1995-2009

- WIOD: ISIC Rev.3 2-digit sectors
output, inputs, labor expenditure, workforce, capital expenditure, investment, capital stocks
- COMEXT: CN8 digit data aggregated to 2-digit
bilateral imports by EU-15 from main partners
- COMTRADE: total imports and exports by sector
- ANBERD: R&D data (nominal expenditure, research personnel)
- Statistical Yearbooks China: R&D data

Technology and wages

- TFP: fit Cobb-Douglas production function

$$\ln(\bar{z}_i^k) = \ln Y_i^k - \beta_{L,i}^k \ln I_i^k - \beta_{H,i}^k \ln H_i^k - \beta_{K,i}^k \ln K_i^k$$

Y, I real output (inputs), H hours worked, K capital use

- Wages

- hourly wage ν_i^k reported in WIOD
- hourly wage adjusted for worker efficiency

$$\bar{\nu}_i^k = \sum_{edu} \frac{\omega_{edu,i}^k}{\bar{\omega}_{edu,i}^k} \bar{\nu}_{edu,i}^k$$

$edu = \{l, m, h\}$ is skill; ω ($\bar{\omega}$) is cost (hour) share by skill;
 $\bar{\nu}_{edu,i}^k$ is efficiency-adjusted wage by skill

$$\bar{\nu}_{edu,i}^k = \nu_{edu,i}^k e^{-g S_{edu}}$$

where $g = .06$: return to education; S_{edu} : average nb years schooling

Instruments

- TFP instrumented with R&D
 1. R&D personnel and real capital stocks: [(I), (II)]
 2. Deflated R&D expenditure: [(III), (IV)]
- Wages instrumented with workforce
 1. Number persons engaged: [(II), (IV)]
 2. Efficiency-adjusted workforce: [(I), (III)]

$$\bar{L}_i^k = \sum_{edu} \bar{L}_{edu,i}^k$$

where $\bar{L}_{edu,i}^k$ is efficiency adjusted nb workers

$$\bar{L}_{edu,i}^k = L_{edu,i}^k e^{g^S S_{edu}}$$

- Bottleneck: R&D data
- Drops: Russia, Bulgaria, Brazil, India, Indonesia, Lithuania, Latvia

Sample of countries

Table: Sample of countries: from 42 to 26

ID	Country	Type	ID	Country	Type
AT	Austria	intra-eu15	PL	Poland	ceec
BE	Belgium-Luxembourg	intra-eu15	RO	Romania	ceec
DK	Denmark	intra-eu15	SK	Slovakia	ceec
FI	Finland	intra-eu15	SI	Slovenia	ceec
FR	France	intra-eu15	TR	Turkey	ceec
DE	Germany	intra-eu15	CA	Canada	other devpd
GR	Greece	intra-eu15	JP	Japan	other devpd
IE	Ireland	intra-eu15	KR	Korea	other devpd
IT	Italy	intra-eu15	NO	<i>Norway</i>	other devpd
NL	Netherlands	intra-eu15	CH	<i>Switzerland</i>	other devpd
PT	Portugal	intra-eu15	US	USA	other devpd
ES	Spain	intra-eu15	BR	<i>Brazil</i>	other emerging
SW	Sweden	intra-eu15	CN	China	other emerging
GB	United Kingdom	intra-eu15	IN	<i>India</i>	other emerging
BG	<i>Bulgaria</i>	ceec	ID	<i>Indonesia</i>	other emerging
HR	<i>Croatia</i>	ceec	MY	<i>Malaysia</i>	other emerging
CZ	Czech Republic	ceec	MX	Mexico	other emerging
EE	Estonia	ceec	RU	<i>Russia</i>	other emerging
HU	Hungary	ceec	SG	<i>Singapore</i>	other emerging
LV	<i>Latvia</i>	ceec	TW	Taiwan	other emerging
LT	<i>Lithuania</i>	ceec	TH	<i>Thailand</i>	other emerging

- Sample: focus on main EU15 trading partners
- In blue: dropped b/c absent from WIOD database
- In red: R&D bottleneck



Estimated parameters

- Estimated heterogeneity $\hat{\theta}$ (EK: 8.3; CDK: 6.5; SW: 4.5):
 1. overidentified: 7.28(.51), 6.72(.43)
 2. identified: 7.84(.52), 7.28(.45)
 3. NB: 4.5 in one-sector economy
- Precisely estimated coefficient on hourly wage: $-\theta(1 - \zeta_k)$
- Estimated sectoral input intensity $\hat{\zeta}_k$:
 1. one sector economy: $\zeta_k = \zeta = .69$ (matches data)
 2. Sector-specific: strongly correlated ζ_k in WIOD

Sectoral input intensity

Table: Sectoral factor share of inputs

	DATA	(I)	(II)	(III)	(IV)
17-18	0.68	0.82	0.79	0.79	0.78
19	0.72	0.97	0.88	0.95	0.87
20	0.67	0.64	0.68	0.65	0.68
21-22	0.63	0.61	0.65	0.62	0.66
24	0.69	0.74	0.74	0.75	0.75
25	0.65	0.74	0.74	0.74	0.74
26	0.62	0.70	0.71	0.70	0.71
27-28	0.66	0.78	0.77	0.78	0.78
29	0.64	0.62	0.66	0.62	0.66
30-33	0.66	0.66	0.68	0.66	0.69
34-35	0.76	0.75	0.74	0.76	0.75
36-37	0.65	0.62	0.67	0.63	0.68

- estimated parameters higher in levels
- higher variability in estimated parameters
- strongly correlated with income share of inputs in data



Proximity ranking

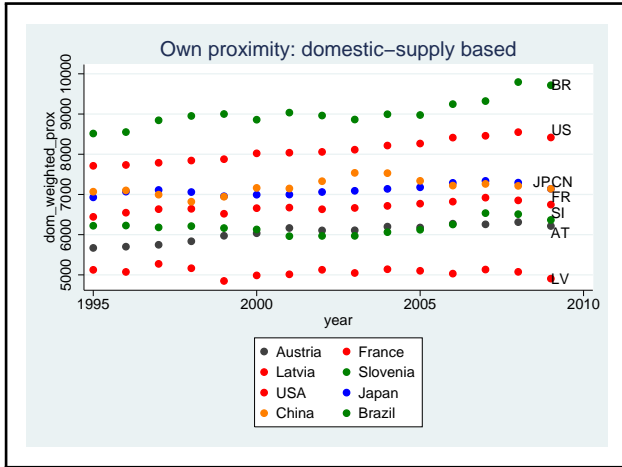
- Compute proximity characteristic in each year

$$\left[\overline{PROX}_{i,t}^M \right]^{-1} = \prod_{s=1}^S \left\{ \sum_{n=1}^N \pi_{ni,t}^s \tau_{ni}^\theta \right\}^{\gamma^s / \theta}$$

- distance as proxy of bilateral trade frictions
 - observed market shares as weights (incl. domestic)
 - estimated θ , expenditure shares γ^k from data
- Instrument with proximity endowment: unweighted norm of distance vector

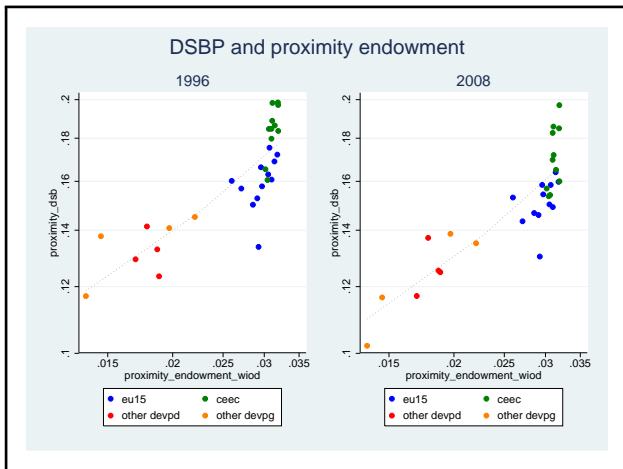
$$\left[PROX_i^M \right]^{-1} = \left[\sum_{n=1}^N dist_{in}^2 \right]^{0.5}$$

Persistence of proximity characteristic



- plots reciprocal of proximity for subset of countries
- illustrates variability across countries and persistence overtime

Microfounded proximity & proximity endowment



- persistent characteristic
- $> 2/3$ total variance picked up by proximity endowment



Proximity mechanism

- Group countries according to proximity characteristic
- Compute pairwise sectoral residuals rescaled by $\widehat{\theta}$
- Compute relative proximity rescaled by $\widehat{\zeta}^k$
- Focus on intersectoral variation: include exporter-year fixed effects

$$\frac{1}{\widehat{\theta}} \left[\widehat{\lambda}_{i,t}^k - \widehat{\lambda}_{i',t}^k \right] = \beta_0 + \beta_1 \ln \left\{ \left(\frac{\widehat{PROX}_{i,t}^M}{\widehat{PROX}_{i',t}^M} \right)^{\widehat{\zeta}^k} \right\} + fe_{i,t} - fe_{i',t} + \eta_{ii',t}^k$$

- Proximity mechanism determines residual ranking of relative sectoral exports if β_1 positive, significant

Results for the full sample

Table: Proximity mechanism in the residual component of RCA rankings

	<i>all</i> (I)	<i>all</i> (I)	<i>all</i> (IV)	<i>all</i> (IV)	<i>devd</i> (I)	<i>devg</i> (I)
<i>relprox * inpint</i>	0.689*** (0.064)	0.375*** (0.093)	1.255*** (0.100)	0.658*** (0.152)	1.288*** (0.101)	0.176** (0.078)
<i>recent</i>		0.585*** (0.126)		1.033*** (0.200)		
Obs	17748	17748	20097	20097	8883	8865
R^2	0.674	0.674	0.665	0.665	0.541	0.776
Recent FE		YES		YES		

- results robust to instrumenting procedure
- proximity matters more in recent period (2001-2009)

Results by sub-group

Table: Proximity mechanism in residual component of RCA rankings

	(I)	(II)	(III)	(IV)
<i>eu15-to-devpd</i> <i>nb-obs</i>	1.379*** 5541	2.359*** 5541	1.344*** 6399	2.263*** 6399
<i>ceec-to-devpd</i> <i>nb-obs</i>	1.151*** 3342	2.242*** 3342	0.890*** 3894	1.712*** 3894
<i>eu15-to-devpg</i> <i>nb-obs</i>	0.165 5529	0.356** 5529	0.254** 6100	0.520*** 6100
<i>ceec-to-devpg</i> <i>nb-obs</i>	0.191* 3336	0.623*** 3336	0.127 3704	0.489*** 3704



Variance decomposition

- Quantify contribution of input cost channel to RCA
- Work with relative exporter-sector dummies
- Split sample by proximity & form pairwise combinations
- Calculate total explained variance by TFP, wages, proximity
- Focus on share uniquely attributable to relative proximity

$$\frac{1}{\widehat{\theta}} \left(\widehat{fe}_{i,t}^k - \widehat{fe}_{i',t}^k \right) = \alpha_0 + \alpha_1 \ln \left[\frac{\widehat{z}_{i,t}^k}{\widehat{z}_{i',t}^k} \right] + \alpha_2 \ln \left\{ \left[\frac{\widehat{v}_{i,t}^k}{\widehat{v}_{i',t}^k} \right]^{-(1-\widehat{\zeta}^k)} \right\} +$$

$$\alpha_3 \ln \left\{ \left[\frac{\widehat{PROX}_{i,t}^M}{\widehat{PROX}_{i',t}^M} \right]^{\widehat{\zeta}^k} \right\} + fe_{i,t} + fe_{i',t} + \xi_{ii',t}^k$$

Unexplained variance attributable to proximity

Table: Fraction of residual variance attributable to proximity

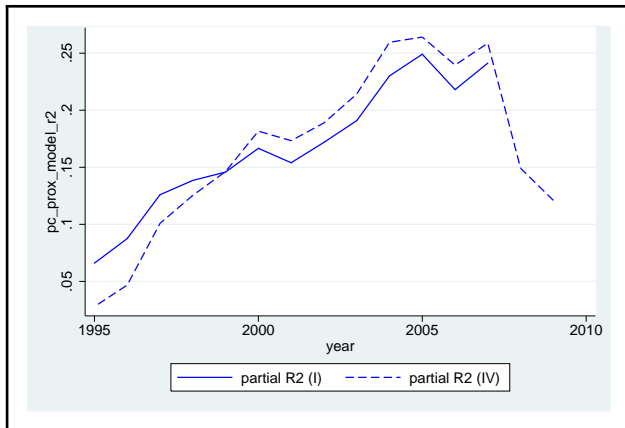
	<i>all</i> (I)	<i>all</i> (II)	<i>all</i> (III)	<i>all</i> (IV)
<i>relprox * inprox</i>	2.777*** (0.282)	3.381*** (0.336)	2.583*** (0.255)	3.043*** (0.297)
R^2	0.178	0.200	0.181	0.196
Obs	17,748	17,748	20,097	20,097

Table: Coefficient of partial determination (proximity, all years)

	<i>all</i> (I)	<i>all</i> (II)	<i>all</i> (III)	<i>all</i> (IV)
<i>resid - relprox</i>	2.601*** (0.305)	3.180*** (0.363)	2.446*** (0.283)	2.907*** (0.330)
R^2	0.154	0.173	0.154	0.169
Obs	17,748	17,748	20,097	20,097

Increasing importance overtime

Figure: Coefficient of partial determination (proximity, annual)



Focus on intersectoral variation

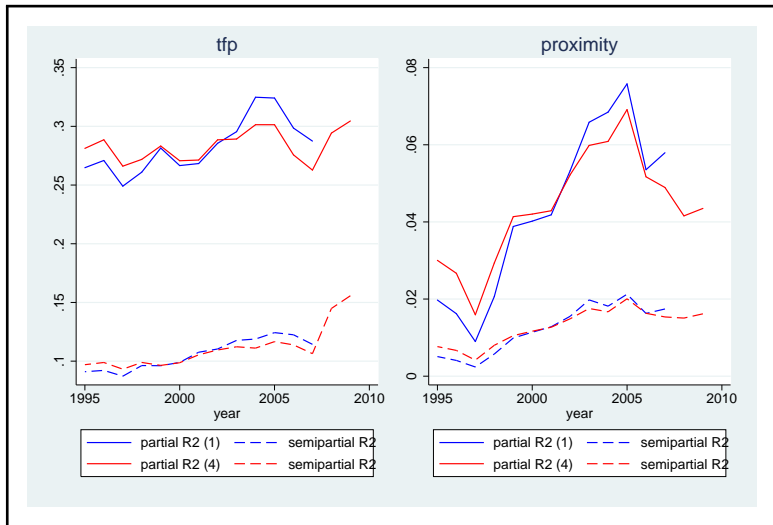
Table: The intersectoral component of RCA rankings

	<i>all</i> (I)	<i>all</i> (II)	<i>all</i> (III)	<i>all</i> (IV)	β -coef (I)
<i>tfp</i>	2.143*** (0.110)	2.105*** (0.107)	2.124*** (0.111)	1.994*** (0.107)	2.50
<i>wage</i>	1.981*** (0.112)	1.919*** (0.109)	2.291*** (0.120)	2.178*** (0.117)	2.32
<i>proximity</i>	1.668*** (0.160)	2.964*** (0.274)	1.642*** (0.156)	2.861*** (0.265)	0.24
R^2	0.731	0.731	0.731	0.726	
Obs	17,748	17,748	20,097	20,097	

- Proximity matters at the intersectoral level
- BUT contribution much lower (see standardized coef. col.5)

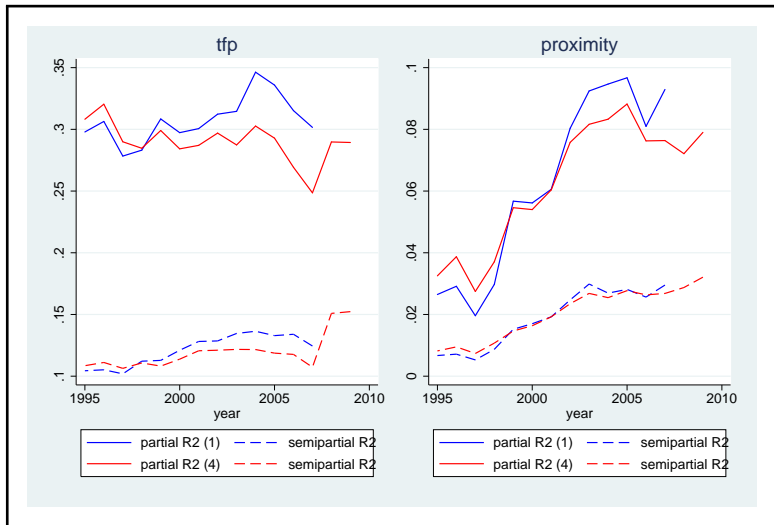
Increasing importance overtime

Figure: Partial and semipartial r^2 in cross section: full sample



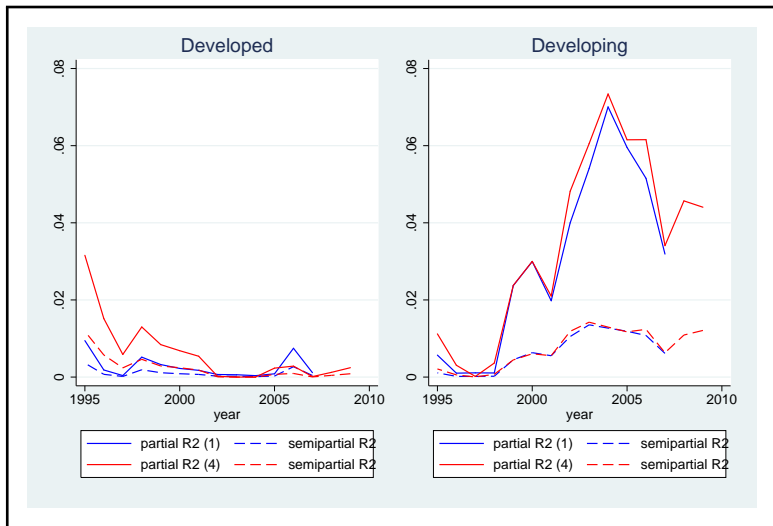
Results by subgroup: EU-15

Figure: Partial and semipartial r^2 : EU15



Results by subgroup: CEECs

Figure: Partial and semipartial r^2 : proximity (CEECs)



Does proximity constitute a source of comparative advantage?

- Determines wedge in relative cost of the input bundle which matters more in input-intensive sectors
- Input cost channel contributes to shape pattern of RCA across partners which differ in proximity to suppliers
- This mechanism has growing importance overtime
- BUT: intersectoral specialization still determined by ranking of relative technology stocks

Robustness & Further Work

- Use model structure to compute price indices: do results on the role of proximity stand?
- Use explicit IO structure: do results in this paper establish a lower bound on the importance of the input cost channel?
- Look into persistence: which type (magnitude) of shocks on structure of trade costs (technology stocks) needed to inflect pattern of specialization?