

Borders and Distance in Knowledge Spillovers: Dying over Time or Dying with Age? - Evidence from Patent Citations

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Overview

Empirical Specification
and Data

Main Results

Conclusion

Ongoing Work

Research Questions

- How localized is intranational and international knowledge flow? How do:
 - national and subnational borders affect diffusion
 - distance and internal distance affect diffusion
- Does the pattern of knowledge diffusion change?
 - Time trend?
 - Age profile?
- What are the sources of border effects in knowledge flows?
 - (Assignee) Self-citation
 - Aggregation bias

Motivation

- Contentious debate about localization of intranational knowledge flows.
 - *Jaffe, Trajtenberge and Henderson (1993, QJE), HJT (2005, AER), Thompson and Fox-Kean (2005, AER).*
- Black box of localization of knowledge flows.
 - Most studies only examine the localization effect, e.g., *JTH (1993, 2005), TFK (2005), Thompson (2006, REStat), Griffith, Lee and Reenen (2007, NBER).*
 - Do not explicitly decompose the contribution from distance and borders.
 - New and old knowledge may be different.

What I Do

- Look at cross-patent citation database from NBER.
 - Patents embody ideas/knowledge.
 - Region i 's patents cite region j 's patents = knowledge flows from region j to i .
 - Use patent citations to track knowledge flows.
- Assign patents to MSA (Metropolitan Statistical Areas), state and national level.
- Characterize age distribution of knowledge diffusion.
- Estimate border and distance effects.
- Analyze the changing pattern (age profile and time trend) of knowledge diffusion.

Main Findings

- Borders and distance matter for knowledge flows:
 - Excluding self-citations:
 - halving distance \uparrow citations by 5.5%;
 - 85% (of initial) knowledge lost crossing national border;
 - 78% lost crossing MSA border; 12% lost crossing state border.
 - Including self-citations \uparrow border and distance effects.
(Existing literature did not look at self-citations.)
- On average, national borders effect larger than subnational.
 - Size of border and distance effects \downarrow with patent age.
 - Size of border and distance effects \uparrow over time.
- Self-citation accounts for 50% border effects.
 - Disaggregated data \downarrow border effects.

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Contribution

- Novel age profiles for border and distance effects.
Consistent with knowledge diffusion process.
- New findings on time trend of border and distance effects.
(not extensively studied in literature)
- Newly constructed finer data (matched at MSA level) helps to explore subnational localization and sources of border effect.
- Part of the resolutions proposed to border puzzle in knowledge flows might be extended and linked to trade flows in future.

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- NBER Patent Citation Database:
More than **3 million** patents granted by US patent office.
All citations (more than **16 million**) made by each patent since 1975.
More than 40% from foreigners (outside of the U.S.)
Use patent citations between 1980 and 1997.
- Sample contains **357** regions:
270 MSAs in the U.S. + 49 phantom MSAs (non-metro area for each state)
Outside of the U.S.: 38 countries (main patent cited nations)
- Cover more than **99.9%** patents and citations in NBER data.
More than 93% can be matched to MSA.
- Sample size: $357 \times 357 \times 18 = \mathbf{2294082}$ region pairs.

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- Empirical gravity equation motivated by theoretical gravity equation of knowledge flows (See Appendix of paper for derivation)
- Fixed effects: to control for unobserved multilateral resistance terms.

Empirical Gravity Equation (Baseline Regression)

$$\ln\left(\frac{c_{ij}}{y_i y_j}\right) = k + \alpha \ln d_{ij} + \beta_1 B_{ij}^{sn} + \beta_2 B_{ij}^n + r_1^i C I^i + r_2^j C E^j + (1 - \sigma) \varepsilon_{ij}$$

c_{ij} : how many citations region j receives from region i (i.e., region i cites region j 's knowledge; knowledge flows from j to i).

y_j : total number of citations region j receives.

$C I^i$: 1 if i is the citing region, 0 o.w.;

$C E^j$: 1 if j is the cited region, 0 o.w.

Question 1

■ How localized is knowledge diffusion?

- Halving distance \uparrow knowledge flows by 6.5% (5.5% if without self-citations).
- Excluding self-citations, aggregate knowledge flows: 85% (of initial) knowledge lost crossing national border; 78% lost crossing MSA border; 12% lost crossing state border.
- National border effect always larger than subnational.
- Self-citations (SC) partly exaggerate border and distance effects.

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Specification:	(1)	(2)	(3)	(4)	(5)	(6)
	With self-citation			Without self-citation		
lnd_{ij}	-0.131** (0.002)	-0.211** (0.002)	-0.154** (0.002)	-0.116** (0.002)	-0.167** (0.002)	-0.128** (0.002)
B_{ij}^m	-2.134** (0.014)		-2.245** (0.013)	-1.509** (0.015)		-1.573** (0.014)
B_{ij}^s	-0.224** (0.009)	-0.655** (0.009)		-0.124** (0.009)	-0.433** (0.009)	
B_{ij}^n	-2.589** (0.018)	-0.858** (0.015)	-2.433** (0.017)	-1.903** (0.019)	-0.695** (0.015)	-1.821** (0.018)
B_{ij}^m effect	8.449** (0.119)		9.440** (0.126)	4.524** (0.067)		4.823** (0.067)
B_{ij}^s effect	1.252** (0.011)	1.925** (0.017)		1.132** (0.011)	1.542** (0.014)	
B_{ij}^n effect	13.316** (0.243)	2.360** (0.034)	11.390** (0.196)	6.707** (0.126)	2.003** (0.029)	6.178** (0.109)
Citing-region effect	yes	yes	yes	yes	yes	yes
Cited-region effect	yes	yes	yes	yes	yes	yes
Year dummies	yes	yes	yes	yes	yes	yes
F-statistics	1826	1714	1825	1721	1672	1723
Adjusted R^2	0.74	0.73	0.74	0.73	0.72	0.73

Notes: ** Significant at 1% level.

Question 2

■ Does the pattern of knowledge diffusion change?

Conjecture in literature: "... given that we know that localization effects are likely to fade over time ... " (HJT, 2005, AER)

- Age is defined as a citation lag between cited and citing patent.
- Use proportion of citation received in total citation to characterize age distribution of knowledge diffusion.

Main Results

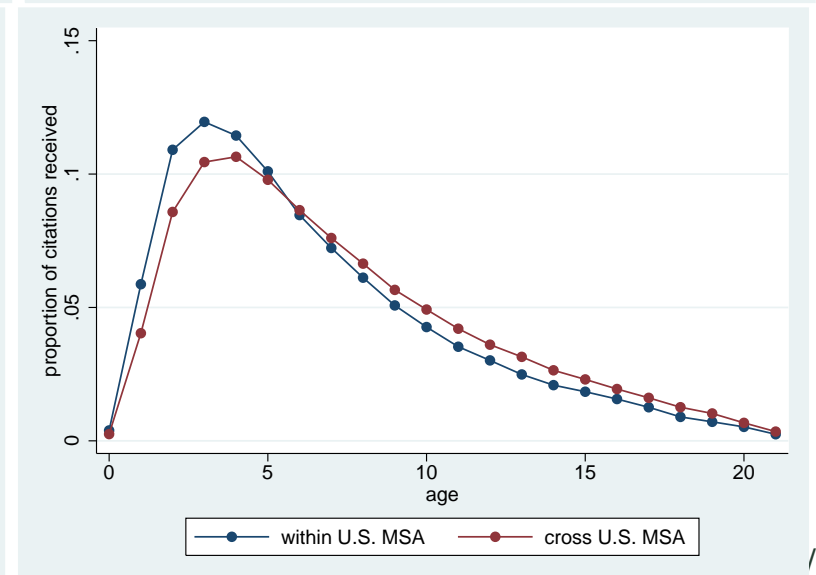
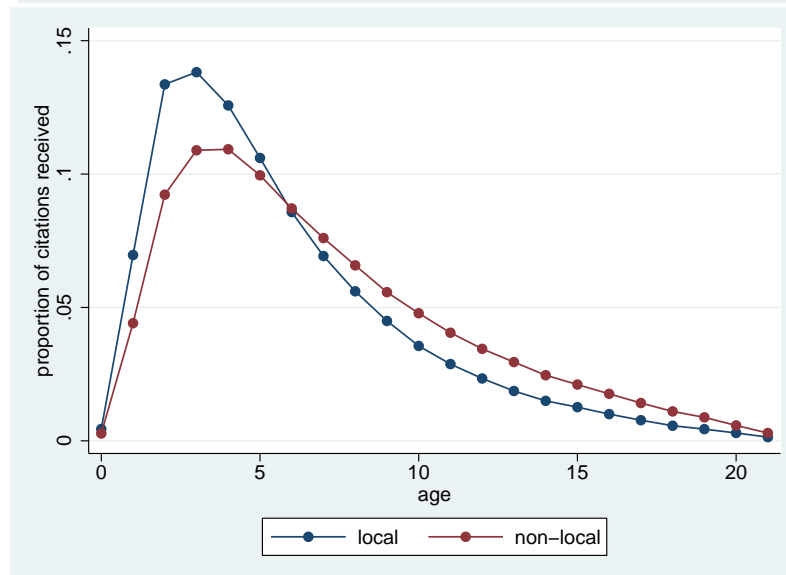
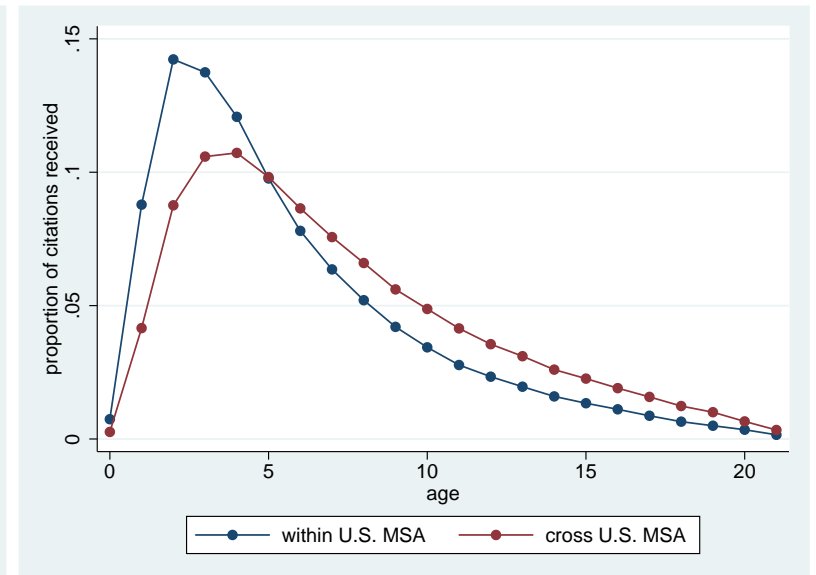
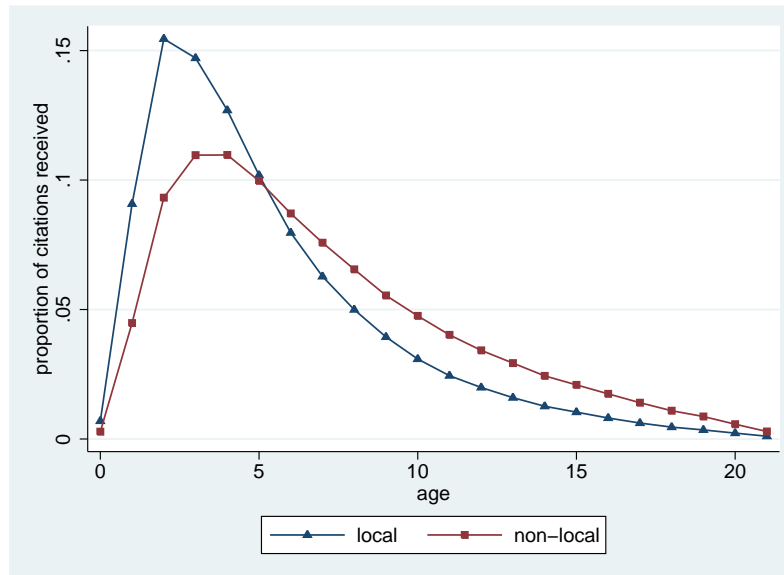
Age distribution of knowledge flows (above: w/ SC; below: w/o SC)

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Distance and border effects decrease with age of knowledge.

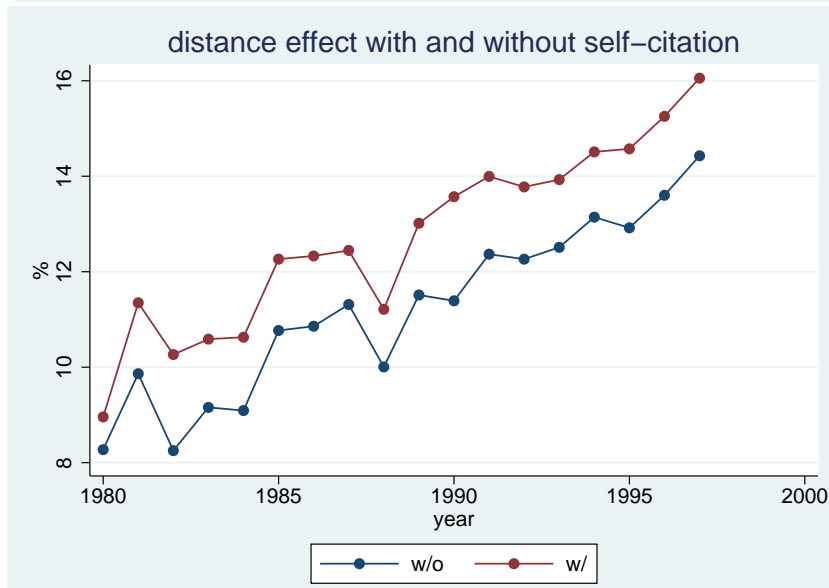
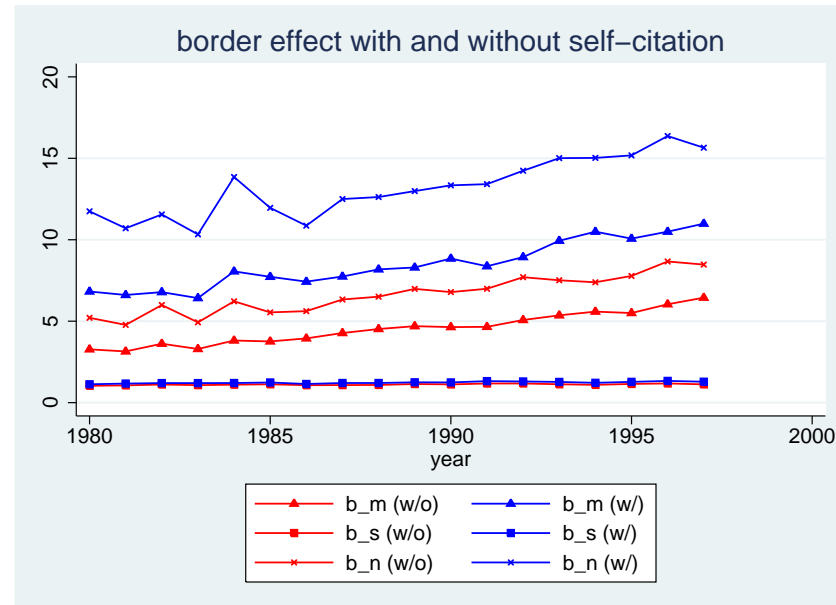
Estimates by Age of Knowledge (without Self-citation)

Specification:	age [0,5)	age [5,10)	age [10,15)	age [15,20)	age [20,more)
$\ln d_{ij}$	-0.092** (0.003)	-0.091** (0.004)	-0.079** (0.005)	-0.065** (0.008)	-0.059* (0.028)
MSA border effect	3.713** (0.074)	3.214** (0.077)	2.694** (0.092)	2.379** (0.136)	1.996** (0.382)
state border effect	1.114** (0.016)	1.099** (0.018)	1.071** (0.025)	1.074† (0.041)	1.068 (0.148)
national border effect	5.863** (0.151)	4.618** (0.140)	3.726** (0.159)	3.289** (0.229)	2.492** (0.570)
Citing-region effect	yes	yes	yes	yes	yes
Cited-region effect	yes	yes	yes	yes	yes
Year dummies	yes	yes	yes	yes	yes
F-statistics	824	710	399	232	46
Adjusted R^2	0.68	0.65	0.63	0.66	0.69

Notes: ** Significant at 1% level. * Significant at 5% level. † Significant at 10% level.

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Time trend



Question 3: Sources of border effect?

- 11% of citations are self-citation.
They account for approximately 50% MSA and national border effects.
- Aggregation bias:
 - Overestimate aggregate border effect.
 - Evidence:
 - 1) State to MSA level decomposing ↓ border effects.
 - 2) By age group decomposing ↓ border effects.
 - 3) By category decomposing ↓ border effects.

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Border and Distance Effects by Category (without Self-citation)

	Whole sample	Cat 1 Chemical	Cat 2 C.&C.	Cat 3 D.&M.	Cat 4 E.&E.	Cat 5 Mechanical	Cat 6 Others
lnd_{ij}	-0.116** (0.002)	-0.055* (0.009)	-0.056** (0.015)	-0.020 (0.014)	-0.047** (0.009)	-0.057** (0.006)	-0.084** (0.004)
B_{ij}^m Effect	4.524** (0.067)	2.323** (0.131)	1.530** (0.153)	1.693** (0.149)	2.162** (0.132)	2.467** (0.086)	2.461** (0.068)
B_{ij}^s Effect	1.132** (0.011)	1.154** (0.050)	1.048 (0.077)	1.101 (0.077)	1.096* (0.050)	1.098** (0.028)	1.094** (0.021)
B_{ij}^n Effect	6.707** (0.126)	3.753** (0.263)	2.338** (0.273)	2.402** (0.247)	3.164** (0.231)	4.142** (0.184)	3.781** (0.132)
Citing effect	yes	yes	yes	yes	yes	yes	yes
Cited effect	yes	yes	yes	yes	yes	yes	yes
Year effect	yes	yes	yes	yes	yes	yes	yes
F-statistics	1721	214	177	174	233	413	554
Adjusted R^2	0.73	0.55	0.59	0.57	0.58	0.64	0.65

At finer category/industry level, distance less important; border effects ↓ but still significant.

National border effects always larger than subnational.

Robustness to Alternative Specifications

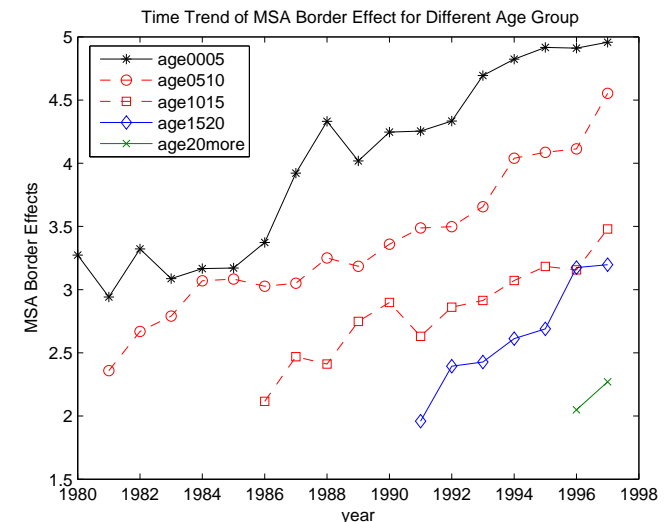
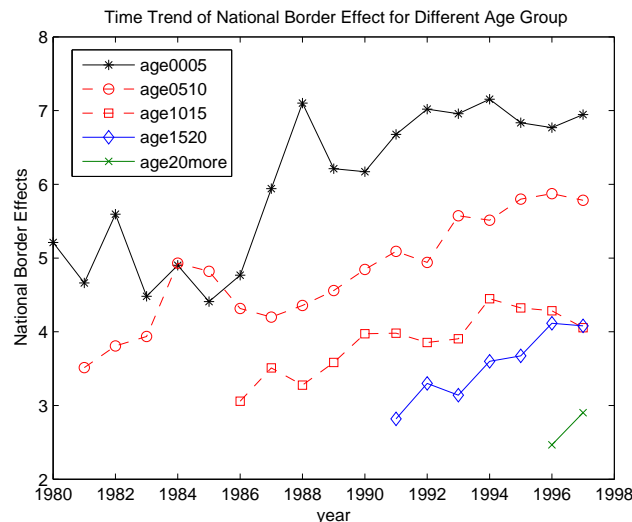
Test of interaction between time effect and age effect.

Time trend for border effects holds even for a subsample with similar age.

Among each category, age profiles are robust.

Among each category and each age group, time trend for border and distance effects \uparrow .

Exceptions of distance effect for very old patents.



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- Intranational and international knowledge diffusion are both strongly localized.
Subnational border, national border and distance contribute to localization.
Most subnational border effect comes from MSA level, rather than state.
- New knowledge faces the largest impediment to the diffusion.
Time trend of border and distance effects is increasing.
- Sorting out self-citation substantially reduces border effects.
Decomposing data contributes to the reduction of border effects.

Trade, Technology Diffusion and Welfare

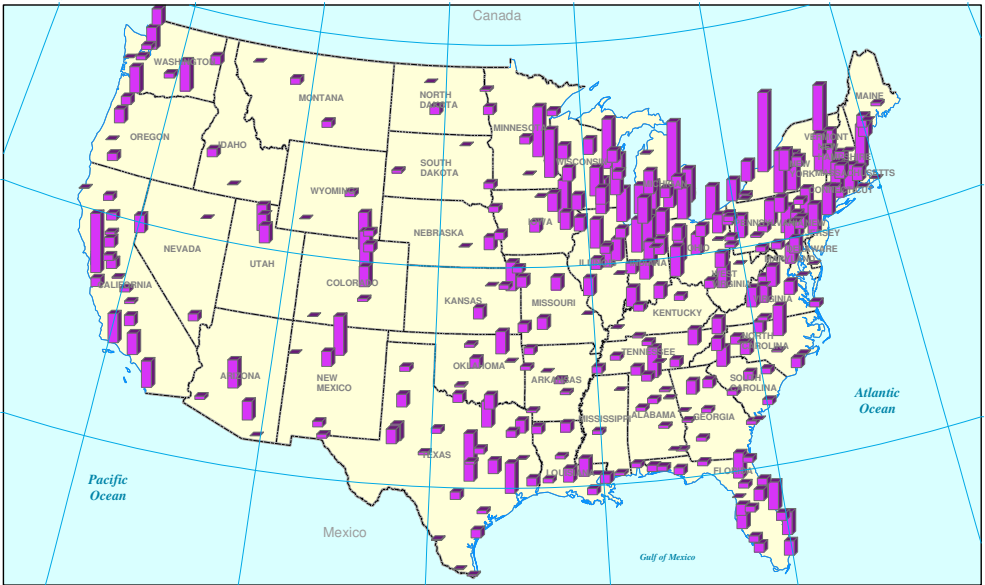
- Introducing country-pair specific barriers to bilateral technology diffusion in an Eaton-Kortum (Alvarez and Lucas) model of trade
- To investigate and quantify the impact of technology diffusion on trade pattern and welfare gains evaluation

Knowledge Agglomeration and Border Effects

- To explain why the size of border effect is inversely related to the degree of concentration in knowledge flows.
- Consistent with stylized facts in trade flows.

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Thank you !