

On Spatial Dynamics

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Introduction

- Economists have long discussed the relationship between agglomeration and growth
 - ▶ Both are based on some form of weakly increasing returns
 - ★ Lucas (1988) and Krugman (1997)
- But no **Dynamic Spatial Theory** has emerged: Why?

The Literature

Literature can be divided in three main strands:

1. Dynamic extensions of the New Economic Geography framework

- Krugman (1991) meets Grossman and Helpman (1991). Nice survey in Baldwin and Martin (2004)
- Mostly 2 locations and no land
- Useful to think about the relationship between regional imbalances and growth
- But very stylized, not rich enough to capture other spatial characteristics within and across industries

The Literature

2. Urban dynamics and the size distribution of cities

- Built to match the size distribution of cities
- Use Gabaix (1999) insight that dynamic evolution key to determine the observed city heterogeneity
- Emphasis on obtaining a dynamic evolution of city size based on model's fundamentals: Black and Henderson (1999), Duranton (2007) and Rossi-Hansberg and Wright (2007)
- Establishes link from growth to the distribution of economic activity in cities
- But no space and no spatial links between cities (like transport costs)
 - ▶ Potential selection bias when looking only at cities

The Literature

3. Optimal evolution of capital in space with forward looking agents

- Full dynamic spatial framework with either diffusion or capital mobility
- The problem solved by Boucekkine et al. (2009) is given by:

$$\max_c \int_0^\infty \int_{\mathbb{R}} U(c(\ell, t)) L(\ell, t) e^{-\beta t} d\ell dt$$

subject to

$$\frac{\partial k(\ell, t)}{\partial t} - \frac{\partial^2 k(\ell, t)}{\partial \ell^2} = Z(\ell, t) f(k(\ell, t)) - \delta k(\ell, t) - c(\ell, t)$$

$$k(\ell, 0) = k_0(\ell) > 0 \text{ and } \lim_{\ell \rightarrow \pm\infty} \frac{\partial k(\ell, t)}{\partial \ell} = 0$$

- An 'ill-posed' problem so cannot be fully analyzed apart from special cases. Only necessary conditions can be advanced
- For examples with diffusion see Brock and Xepapadeas (2008)

The Importance of Space

Why is modeling geographically ordered space important?

- Land at a particular location is a **rival non-replicable input**
 - ▶ Economic density is endogenous, but non-replicability of land leads to decreasing returns: a dispersion force
- The ordering of economic activity in space determines outcomes
 - ▶ Ample evidence for patents (Jaffe et al., 1993), co-location of firms (Duranton and Overman 2005, 2008 and Ellison and Glaeser, 1997), and in general transport costs and mobility costs

The Importance of Space

Why is it hard?

- Adds another dimension
 - ▶ Forward looking agents need to understand the distribution of economic activity in all future dates for all feasible decisions
- Clearing markets is difficult
 - ▶ One possibility is to make probabilistic statements for a large number of locations as in trade (e.g. Eaton and Kortum, 2002)
 - ▶ Another is to clear market sequentially with compact continuous space (e.g. Rossi-Hansberg, 2005)
 - ★ Hard to do with non-symmetric two dimensional setups (like reality!)

An Alternative Model with Space

- Developed in Desmet and Rossi-Hansberg (2009) with two sectors
- Land is given by the unit interval $[0, 1]$, time is discrete, and total population is $\bar{L}$
- Agents solve

$$\max_{\{c(\ell, t)\}_0^\infty} E \sum_{t=0}^{\infty} \beta U(c(\ell, t)) \text{ s.t. } w(\ell, t) + \frac{\bar{R}(t)}{\bar{L}} = p(\ell, t) c(\ell, t)$$

- Free mobility implies that utilities equalize across regions each period
- Firm solve

$$\max_{L(\ell, t)} (1 - \tau(\ell, t)) (p(\ell, t) Z(\ell, t) L(\ell, t)^\mu - w(\ell, t) L(\ell, t))$$

Innovation

- The government of a county can decide to buy an opportunity to innovate by taxing local firms $\tau(\ell, t)$
- Buys a probability $\phi \leq 1$ of innovating at a cost $\psi(\phi)$ per unit of land
- If it innovates it draws a technology multiplier $z(\ell)$ from

$$\Pr[z < z_\ell] = \left(\frac{1}{z}\right)^a$$

such that TFP becomes $z_\ell Z_i(\ell, t)$.

- County G , with land measure I , will then maximize

$$\max_{\phi(\ell, t)} \int_G \frac{\phi(\ell, t)}{a-1} p(\ell, t) Z(\ell, t) L(\ell, t)^\mu d\ell - I\psi(\phi)$$

Diffusion, Transport Costs and Market Clearing

- Innovation diffuses spatially between time periods according to

$$Z_i(\ell, t+1) = \max_{r \in [0,1]} e^{-\delta|\ell-r|} Z(r, t)$$

- Transport costs such that if goods are produced in ℓ and consumed in r , $p(r, t) = e^{\kappa|\ell-r|} p(\ell, t)$
- Goods markets clear sequentially so define $H_i(\ell, t)$ by $H_i(0, t) = 0$ and by the differential equation

$$\frac{\partial H(\ell, t)}{\partial \ell} = \theta(\ell, t) x(\ell, t) - c(\ell, t) \left(\sum_i \theta(\ell, t) L(\ell, t) \right) - \kappa |H(\ell, t)|$$

then, the goods market clears if $H(1, t) = 0$.

- The labor market clearing condition is given by

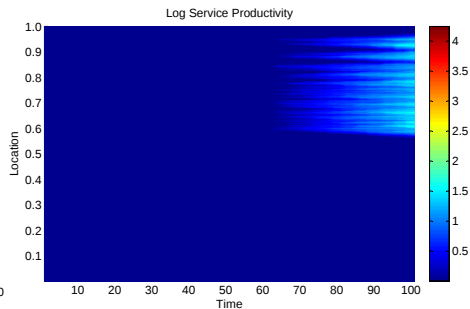
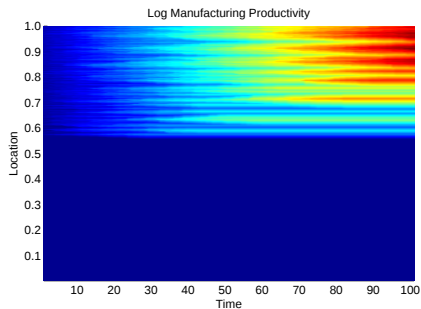
$$\int_0^1 L(\ell, t) d\ell = \bar{L}, \text{ all } t$$

Two Growth Effects

With two sectors:

1. **'Spillover' effect:** locations close to other locations in the same sector grow faster because they benefit from innovation investments close by. This in turn increases incentives to innovate in this locations
2. **'Trade' effect:** locations close to areas that import a particular good experience high prices for that good, thus providing incentives to innovate in that sector

Two Growth Effects



Some Evidence for Manufacturing

Decay Emp. Kernel:				0.1 (half life 7 km)				
Decay Imp. Kernel:	0.07	0.08	0.09	0.10	0.11	0.12	0.13	0.14
Half-Life Imp. Kernel (km):	9.9	8.7	7.7	6.9	6.3	5.8	5.3	5.0

Dependent variable: Log(Industry Employment 2000)-Log(Industry Employment 1990)

Log(Ind. Emp. 1990)	-0.053	-0.0526	-0.05222	-0.05191	-0.05166	-0.05162	-0.0514	-0.05125
	[12.70]***	[12.59]***	[12.49]***	[12.41]***	[12.34]***	[12.32]***	[12.26]***	[12.21]***
Log(Ind. Emp. Kernel 1990)	0.00624	0.00617	0.00607	0.00602	0.00605	0.00592	0.00584	0.00567
	[2.08]**	[2.05]**	[2.02]**	[2.00]**	[2.01]**	[1.97]**	[1.94]*	[1.88]*
Log(Ind. Imp. Kernel 1990)	0.00626	0.00624	0.00628	0.00638	0.00639	0.00632	0.00633	0.00626
	[10.71]***	[10.54]***	[10.49]***	[10.59]***	[10.53]***	[10.36]***	[10.36]***	[10.22]***
Constant	0.56213	0.55913	0.55678	0.55473	0.55276	0.55297	0.55164	0.55117
	[19.06]***	[18.92]***	[18.82]***	[18.74]***	[18.65]***	[18.64]***	[18.59]***	[18.55]***
Observations	2543	2543	2543	2543	2543	2543	2543	2543
R-squared	0.1221	0.121	0.1206	0.1213	0.1209	0.1197	0.1197	0.1187

Dependent variable: Log(Industry Employment 1990)-Log(Industry Employment 1980)

Log(Ind. Emp. 1980)	-0.03445	-0.03389	-0.03371	-0.03338	-0.03287	-0.03256	-0.0323	-0.03207
	[7.35]***	[7.22]***	[7.18]***	[7.10]***	[6.99]***	[6.92]***	[6.86]***	[6.81]***
Log(Ind. Emp. Kernel 1980)	0.03753	0.03786	0.03795	0.03806	0.0382	0.0382	0.03824	0.03828
	[10.77]***	[10.89]***	[10.92]***	[10.97]***	[11.03]***	[11.05]***	[11.08]***	[11.10]***
Log(Ind. Imp. Kernel 1980)	0.00176	0.00223	0.0024	0.00265	0.00297	0.00313	0.00332	0.00348
	[2.45]**	[3.06]***	[3.26]***	[3.59]***	[4.00]***	[4.21]***	[4.47]***	[4.68]***
Constant	0.1047	0.09882	0.09711	0.09398	0.08948	0.0871	0.08495	0.08312
	[3.21]***	[3.02]***	[2.97]***	[2.88]***	[2.73]***	[2.66]***	[2.60]***	[2.54]**
Observations	2857	2857	2857	2857	2857	2857	2857	2857
R-squared	0.0417	0.0428	0.0432	0.044	0.045	0.0456	0.0463	0.047

Absolute value of t statistics in brackets

* significant at 10%; ** significant at 5%; *** significant at 1%

Some Evidence for Services

Decay Emp. Kernel:				0.1 (half life 7 km)				
Decay Imp. Kernel:	0.07	0.08	0.09	0.10	0.11	0.12	0.13	0.14
Half-Life Imp. Kernel (km):	9.9	8.7	7.7	6.9	6.3	5.8	5.3	5.0

Dependent variable: Log(Service Employment 2000)-Log(Service Employment 1990)

Log(Serv. Emp. 1990)	0.00346	0.00383	0.00409	0.00426	0.0043	0.00443	0.00451	0.00458
	[1.29]	[1.43]	[1.53]	[1.59]	[1.61]	[1.66]*	[1.69]*	[1.72]*
Log(Serv. Emp. Kernel 1990)	0.00624	0.00603	0.00587	0.00576	0.00572	0.00563	0.00557	0.00552
	[4.55]***	[4.40]***	[4.28]***	[4.20]***	[4.17]***	[4.11]***	[4.07]***	[4.03]***
Log(Serv. Imp. Kernel 1990)	-0.00028	0.00014	0.00044	0.00065	0.00073	0.00089	0.00101	0.00113
	[0.74]	[0.36]	[1.15]	[1.67]*	[1.85]*	[2.26]**	[2.55]**	[2.87]***
Constant	0.18715	0.18406	0.18195	0.1805	0.18018	0.17918	0.17855	0.17789
	[8.08]***	[7.95]***	[7.86]***	[7.80]***	[7.79]***	[7.75]***	[7.73]***	[7.71]***
Observations	2277	2277	2277	2277	2277	2277	2277	2277
R-squared	0.0131	0.013	0.0135	0.0141	0.0144	0.0151	0.0157	0.0165

Dependent variable: Log(Service Employment 1990)-Log(Service Employment 1980)

Log(Serv. Emp. 1980)	0.04007	0.04012	0.04028	0.04034	0.04026	0.04024	0.04023	0.0402
	[13.89]***	[13.91]***	[13.99]***	[14.02]***	[14.00]***	[14.00]***	[14.00]***	[13.99]***
Log(Serv. Emp. Kernel 1980)	0.01013	0.01003	0.00987	0.00977	0.00977	0.00973	0.00975	0.00978
	[6.89]***	[6.82]***	[6.72]***	[6.67]***	[6.67]***	[6.65]***	[6.66]***	[6.68]***
Log(Serv. Imp. Kernel 1980)	0.00153	0.00176	0.00208	0.00232	0.00236	0.00249	0.00251	0.00245
	[3.72]***	[4.22]***	[4.96]***	[5.49]***	[5.58]***	[5.86]***	[5.90]***	[5.76]***
Constant	-0.19616	-0.19598	-0.19655	-0.19644	-0.19573	-0.19524	-0.19523	-0.19522
	[8.12]***	[8.11]***	[8.15]***	[8.15]***	[8.12]***	[8.11]***	[8.11]***	[8.11]***
Observations	2616	2616	2616	2616	2616	2616	2616	2616
R-squared	0.1191	0.1204	0.1227	0.1245	0.1248	0.1259	0.1261	0.1255

Absolute value of t statistics in brackets

* significant at 10%; ** significant at 5%; *** significant at 1%

Conclusion

- Frameworks in the literature either not rich enough, lack space, or only partially understood
- Need to develop new spatial dynamic framework that can be contrasted to the data
 - ▶ On what dimensions?
 - ▶ Should have both 'spillover' and 'trade' effects
- Need a structural way of relating to the data to be able to run counterfactual exercises
- These are mayor challenges for the next fifty years of regional science!