Lesson 11: Spatial Interaction Models

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15 Mar 2023

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What Spatial Interaction Models are?

Spatial interaction or "gravity models" estimate the flow of people, material, or information between locations in geographical space.



(i) Note

Spatial interaction models seek to explain existing spatial flows. As such it is possible to measure flows and predict the consequences of changes in the conditions generating them. When such attributes are known, it is possible to better allocate transport resources such as conveyances, infrastructure, and terminals.

Conditions for Spatial Flows

• Three interdependent conditions are necessary for a spatial interaction to occur:



Representation of a Movement as a Spatial Interaction

Representing mobility as a spatial interaction involves several considerations:

- Locations: A movement is occurring between a location of origin and a location of destination. i generally denotes an origin while j is a destination.
- Centroid: An abstraction of the attributes of a zone at a point.
- Flows: Flows are generally expressed by a valued vector Tij representing an interaction between locations i and j.
- Vectors: A vector Tij links two centroids and has a value assigned to it (50) which can represents movements.



Constructing an O/D Matrix

- The construction of an origin / destination matrix requires directional flow information between a series of locations.
- Figure below represents movements (O/D pairs) between five locations (A, B, C, D and E). From this graph, an O/D matrix can be built where each O/D pair becomes a cell. A value of 0 is assigned for each O/D pair that does not have an observed flow.



O/D Matrix

	Α	В	С	D	Е	Ti
Α	0	0	50	0	0	50
В	0	0	60	0	30	90
С	0	0	0	30	0	30
D	20	0	80	0	20	120
E	0	0	90	10	0	100
Tj	20	0	280	40	50	390

Three Basic Types of Interaction Models

- The general formulation of the spatial interaction model is stated as **Tij**, which is the interaction between location i (origin) and location j (destination). Vi are the attributes of the location of origin i, Wj are the attributes of the location of destination j, and Sij are the attributes of separation between the location of origin i and the location of destination j.
- From this general formulation, three basic types of interaction models can be derived:



Gravity Models

The general formula (also known as unconstrained):

$$T_{ij} = k \frac{V_i^{\mu} W_j^{\alpha}}{d_{ij}^{\beta}}$$

- Tij is the transition/trip or flow, T, between origin i (always the rows in a matrix) and destination *j* (always the columns in a matrix). If you are not overly familiar with matrix notation, the *i* and *j* are just generic indexes to allow us to refer to any cell in the matrix.
- V is a vector (a 1 dimensional matrix or, if you like, a single line of numbers) of origin attributes which relate to the emissivity of all origins in the dataset, *i* – this could be any of the origin-related variables.

- variables.
- *d* is a matrix of costs (frequently distances hence, d) relating to the flows between *i* and *j*.
- to decrease.

• W is a vector of destination attributes relating to the attractiveness of all destinations in the dataset, *j* – similarly, this could be any of the destination related

• k, μ, α and β are all model parameters to be estimated. β is assumed to be negative, as with an increase in cost/distance we would expect interaction

Unconstrained (Totally constrained) case

The O-D Matrix

The estimated O-D matrix:



and distance matrix:

3 21 5 $\mathbf{2}$ 151 $\mathbf{2}$ 15 $\mathbf{2}$ 103 5 102

and the calculation T11

 $0.004944 \times 160 \times 200$ $\mathbf{2}$

Suppose $\lambda = 1$, $\alpha = 1$, and $\beta = -1$ for this system.

est	inatio	ns (j)		$\sum_{i} \hat{T}_{ij}$
C	1	2	3	J
	79	19	35	133
	29	412	49	490
	36	33	98	167
	144	464	182	790
				$\sum_{i} \sum_{j} \hat{T}_{ij}$

- = 79

The Origin (Production) Constrained Model

In the Origin Constrained Model,

- *Oi* does not have a parameter as it is a known constraint.
- Ai is known as a **balancing factor** and is a vector of values which relate to each origin *i* which do the equivalent job as k in the unconstrained/total constrained model but ensure that flow estimates from each origin sum to the know totals *Oi* rather than just the overall total.

 $T_{ij} = A_i O_i W_j^{\alpha} d_{ij}^{-\beta}$

where:

$$O_i = \sum_j T_{ij}$$

and:

$$A_i = \frac{1}{\sum_j W_j^{\alpha} d_{ij}^{-} \beta}$$

Oringin (Production) constrained case

The O-D Matrix

The estimated O-D matrix:

	Destinations (j)		estinations (j) Total Outlflows Destination		tinatio	tions (j) \sum_{i}					
	from/to	1	2	3	$\left(\sum_{j} T_{ij} \text{ or } O_{i}\right)$		from/to	1	2	3	Ĵ
	1	100	20	40	160		1	95	23	42	160
Origins (i)	2	60 40	300	90	450	Origins (i)	$\frac{1}{2}$	27	378	45	450
	J Total Inflows	40	$\overline{00}$	90	Crand Total	OIIGIND(0)	2	38	36	106	180
	$(\sum T_{ii} \text{ or } D_i)$	200	370	220	790	$\sum \hat{m}$	0	100	497	100	700
	$(\sum_{i} ij)$ or Dj	200	010	220		$\sum_{i} T_{ij}$		160	437	193	790
					$\left(\sum_{i}\sum_{j}T_{ij} \text{ or } T\right)$	ι					
						A1 is calcula	ated as sh	nown	belov	N:	
and dista	ance matrix	7 •				-		1			
						$A_1 = \left[\frac{200}{2} + \right]$	$\frac{370}{15} + \frac{200}{5}$	=	[168.6]	$7]^{-1} =$	0.00592
	1 2		3				10 0				
1	2 15		5								
2	15 2		10			Hence, T11	is calcula	ted a	s sho	wn be	elow:
3	5 10		2								
			-			$\hat{T}_{11} = \frac{0.005929}{0.005929}$	$\frac{9 \times 160 \times 20}{2}$	$\frac{00}{0} = 98$	5		

Suppose $\lambda = 1$, $\alpha = 1$, and $\beta = -1$ for this system.

$$\left[\frac{200}{5}\right]^{-1} = [168.67]^{-1} = 0.005929$$

The Destination (Attraction) Constrained Model

$$T_{ij} = D_j B_j V_i^{\mu} d_{ij}^{-} \beta$$

where:

$$D_j = \sum_i T_{ij}$$

and:

$$B_j = \frac{1}{\sum_i V_i^{\mu} d_{ij}^{-\beta}}$$

Destination (Attraction) constrained case

The O-D Matrix

The estimated O-D matrix:

	Destinations (j)				Total Outlflows		
	from/to	1	2	3	$(\sum_{i} T_{ij} \text{ or } O_i)$		
	1	100	20	40	$^{\jmath}$ 160		
Origins (i)	2	60	300	90	450		
	3	40	50	90	180		
	Total Inflows				Grand Total		
	$\left(\sum_{i} T_{ij} \text{ or } D_{j}\right)$	200	370	220	790		
	i				$\left(\sum_{i}\sum_{j}T_{ij} \text{ or } T\right)$		



B1 is calculated as shown below:

$$B_1 = \left[\frac{160}{2} + \frac{450}{15} + \frac{180}{5}\right]^{-1} = [146]^{-1} = 0.006849$$

Hence, T11 is calculated as shown below:

$$\hat{T}_{11} = \frac{0.006849 \times 160}{2}$$

and distance matrix:

	1	2	3
1	2	15	5
2	15	2	10
3	5	10	2

Suppose $\lambda = 1$, $\alpha = 1$, and $\beta = -1$ for this system.

)est	tinatio	ns (j)		$\sum_{i} \hat{T}_{ij}$
0	1	2	3	J
	110	16	42	168
	41	328	59	428
	49	26	119	194
	200	370	220	790

 $0 \times 200 = 110$

The Doubly Constrained Model

$$T_{ij} = A_i O_i B_j D_j d_{ij}^- \beta$$

where:

$$O_i = \sum_j T_{ij}$$

 $D_j = \sum_i T_{ij}$

and:

$$A_{i} = \frac{1}{\sum_{j} B_{j} D_{j} d_{ij}^{-} \beta}$$
$$B_{j} = \frac{1}{\sum_{i} A_{i} O_{i} d_{ij}^{-} \beta}$$

(i) Note

Note that the calculation of *Ai* relies on knowing *Bj* and the calculation of *Bj* relies on knowing *Ai* – something of a conundrum to which the solution is elegantly described by Senior (1979), who sketches out a very useful algorithm for iteratively arriving at values for *Ai* and *Bj* by setting each to equal 1 initially and then continuing to calculate each in turn until the difference between successive iterations of the *Ai* and *Bj* values is small enough not to matter.

Doubly constrained case

The O-D Matrix

and distance matrix:

1

 $\mathbf{2}$

15

5

1

 $\mathbf{2}$

3

The estimated O-D matrix:

	Destinations (j)				Total Outlflows
	from/to	1	2	3	$\left(\sum_{i} T_{ij} \text{ or } O_i\right)$
	1	100	20	40	$^{\jmath}$ 160
Origins (i)	2	60	300	90	450
	3	40	50	90	180
	Total Inflows				Grand Total
	$(\sum T_{ij} \text{ or } D_j)$	200	370	220	790
	i				$\left(\sum_{i}\sum_{j}T_{ij} \text{ or } T\right)$

3

5

10

2



 $\hat{T}_{11} = \frac{0.0046 \times 160 \times 1.45 \times 200}{2} = 107$

computer.

Suppose $\lambda = 1$, $\alpha = 1$, and $\beta = -1$ for this system.

2

 $\mathbf{2}$

10

15

Dest	tinatio	ns (j)		$\sum_{i} \hat{T}_{ij}$
/to	1	2	3	J
	107	13	40	160
	47	334	69	450
	46	23	111	180
	200	370	220	790

Hence, T11 is calculated as shown below:

Notice that A1 and B1 are computed by using