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Using Asymmetry to Estimate Potential

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# Abstract

Network analysis is often based on matrices of connections. An asymmetric square array of this type can be decomposed into two components. The anti-symmetric part is especially interesting because this can be used to view influence patterns. A journal to journal table is examined as an example. Another application is to the geographic migration of people.

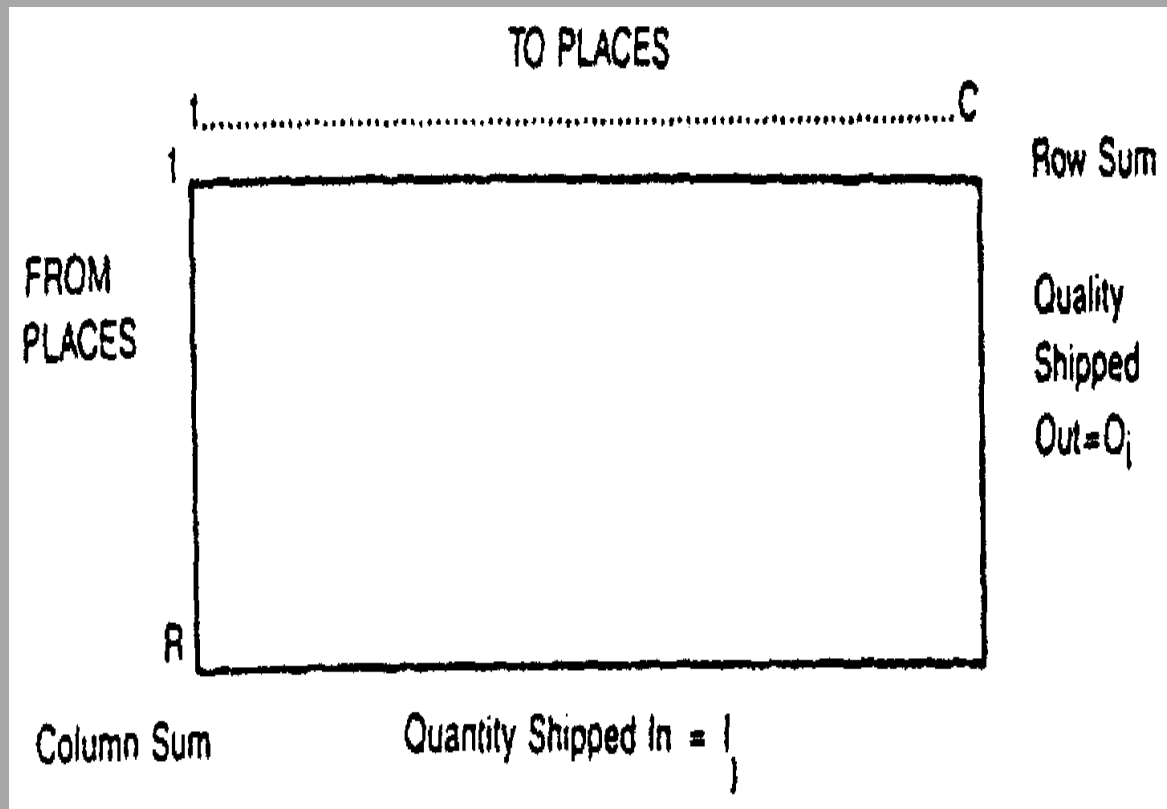
The concern is with complete, square, asymmetric, valued tables, though the procedure may also work with two mode tables.

For this demonstration I have used only small examples.

One example is based on geographic data, the other on journal-to-journal citations.

# The form of a movement table $M_{ij}$

In a non-geographic, network, environment the 'places' are sometimes called 'actors'.



Let  $M_{ij}$  represent the movement table, with  $i$  rows and  $j$  columns. It can be separated into two parts, as follows.

$$M_{ij} = M^+ + M^-$$

where

$$M^+ = (M_{ij} + M_{ji})/2 \quad \text{symmetric}$$

$$M^- = (M_{ij} - M_{ji})/2 \quad \text{skew symmetric}$$

The variance can also be computed for each component,

and the degree of asymmetry can be computed.

## How the two parts are used

I consider the symmetric component as a type of background.

The real interest is in the asymmetric part.

In the geographic case the position of the places is known.

But if locations are not given then the symmetric part may be used to make an estimate of these positions.

This estimate is made using an ordination, trilateration, or multidimensional scaling algorithm.

The first example uses a 33 by 33 matrix of commuting in the vicinity of Munich, Germany.

The matrix is shown next.

A map of the regions is given in:

D. Fliedner, 1962, “Zyklonale Tendenzen bei Bevölkerungs und Verkehrsbewegungen in Städtischen Bereichen untersucht am Beispiel der Städte Göttingen, München, und Osnabrück”, *Neues Archiv für Niedersachsen*, 10:15 (April 4): 277-294, (following p. 285).





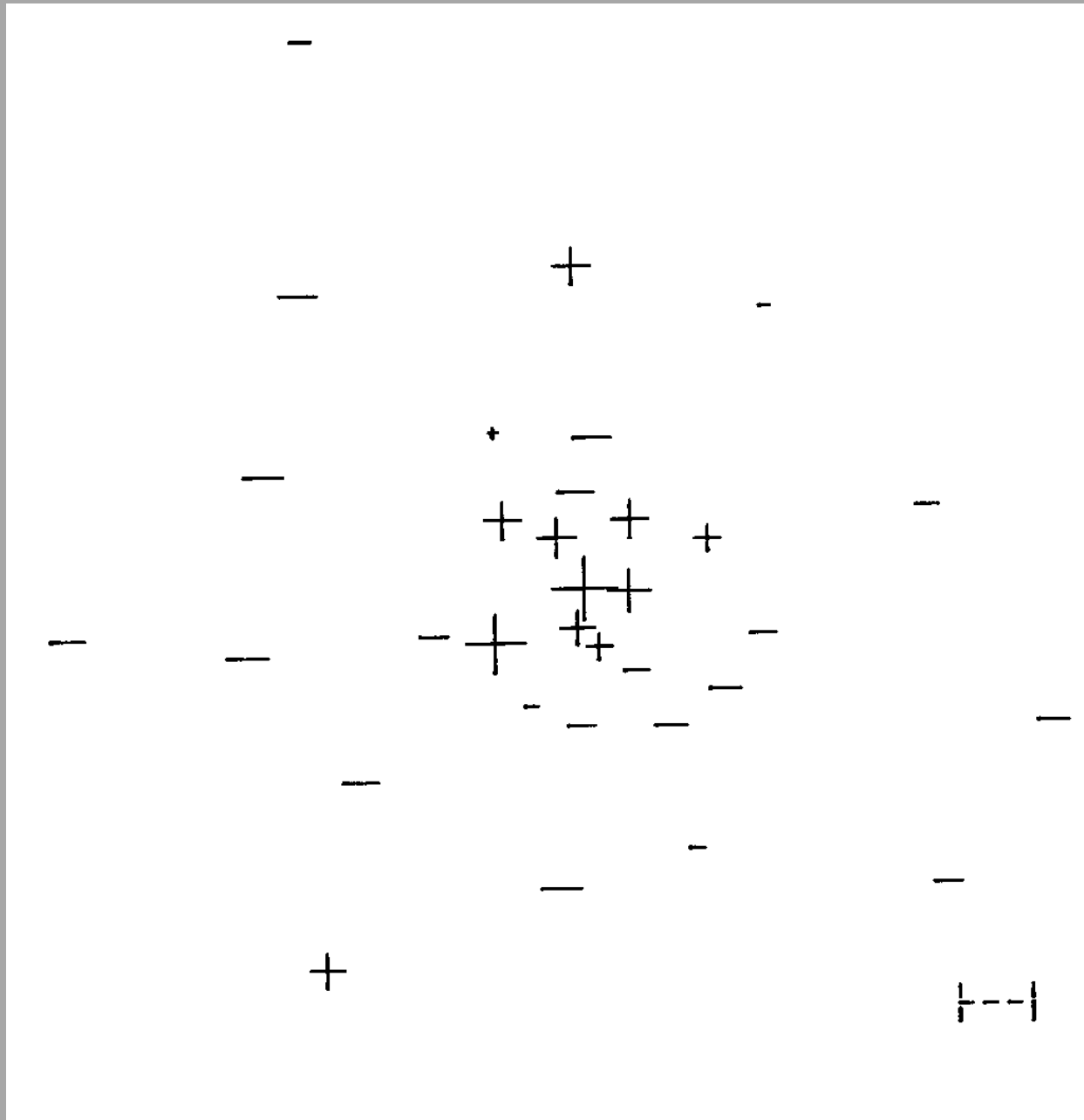
Adding across the table, the column marginals give the outsums (a.k.a. outdegree). Summing down the rows gives the insums (a.k.a indegree).

The ‘sending’ places (rows) are known as ‘sources’, and are shown on the map as negative signs.

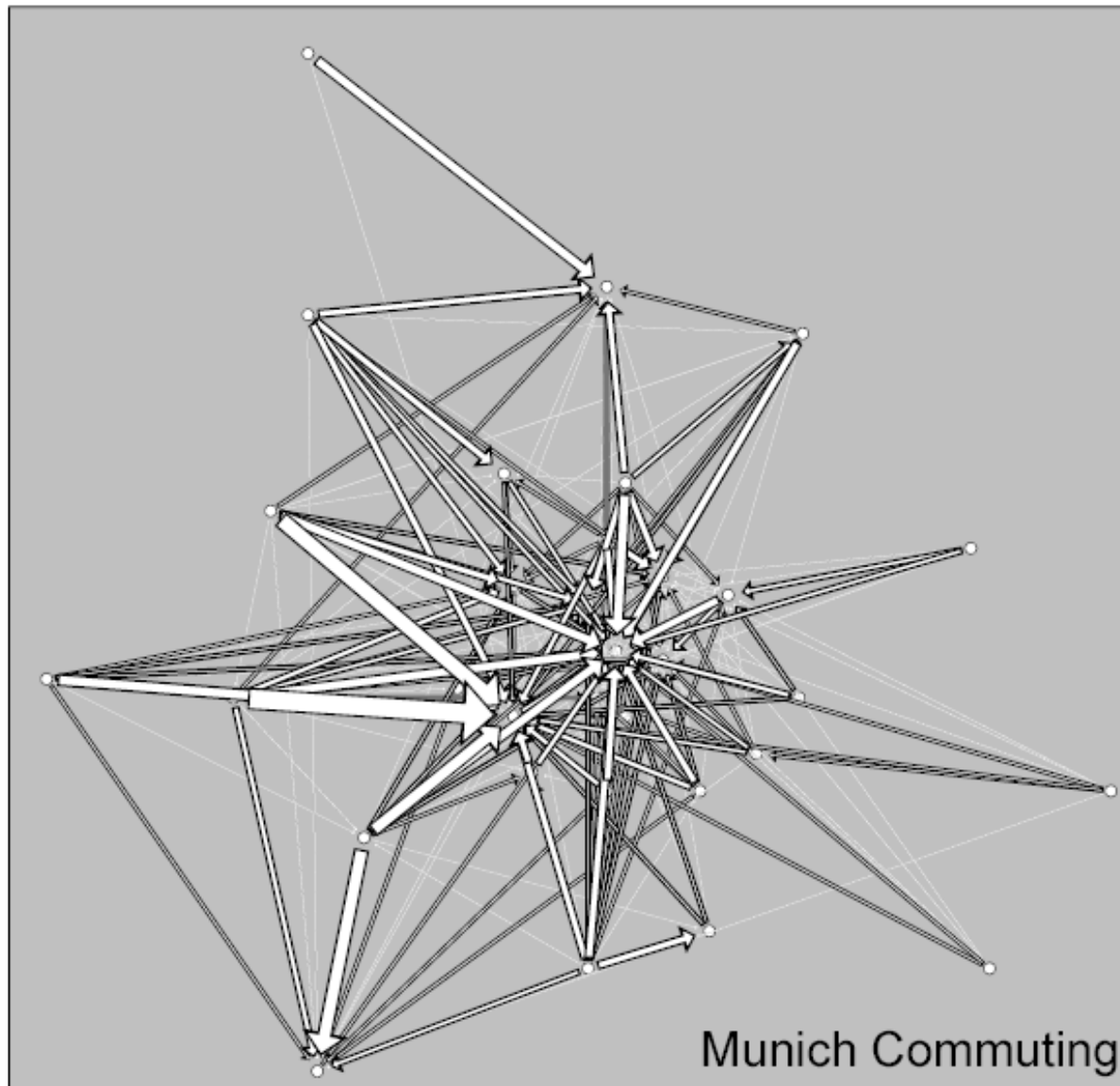
The ‘receiving’ places (columns) are the ‘sinks’ and are shown as plus signs.

The size of the symbol represents the magnitude of the movement volume.

# Munich Commuting (1939)



# 1939 Net Communting



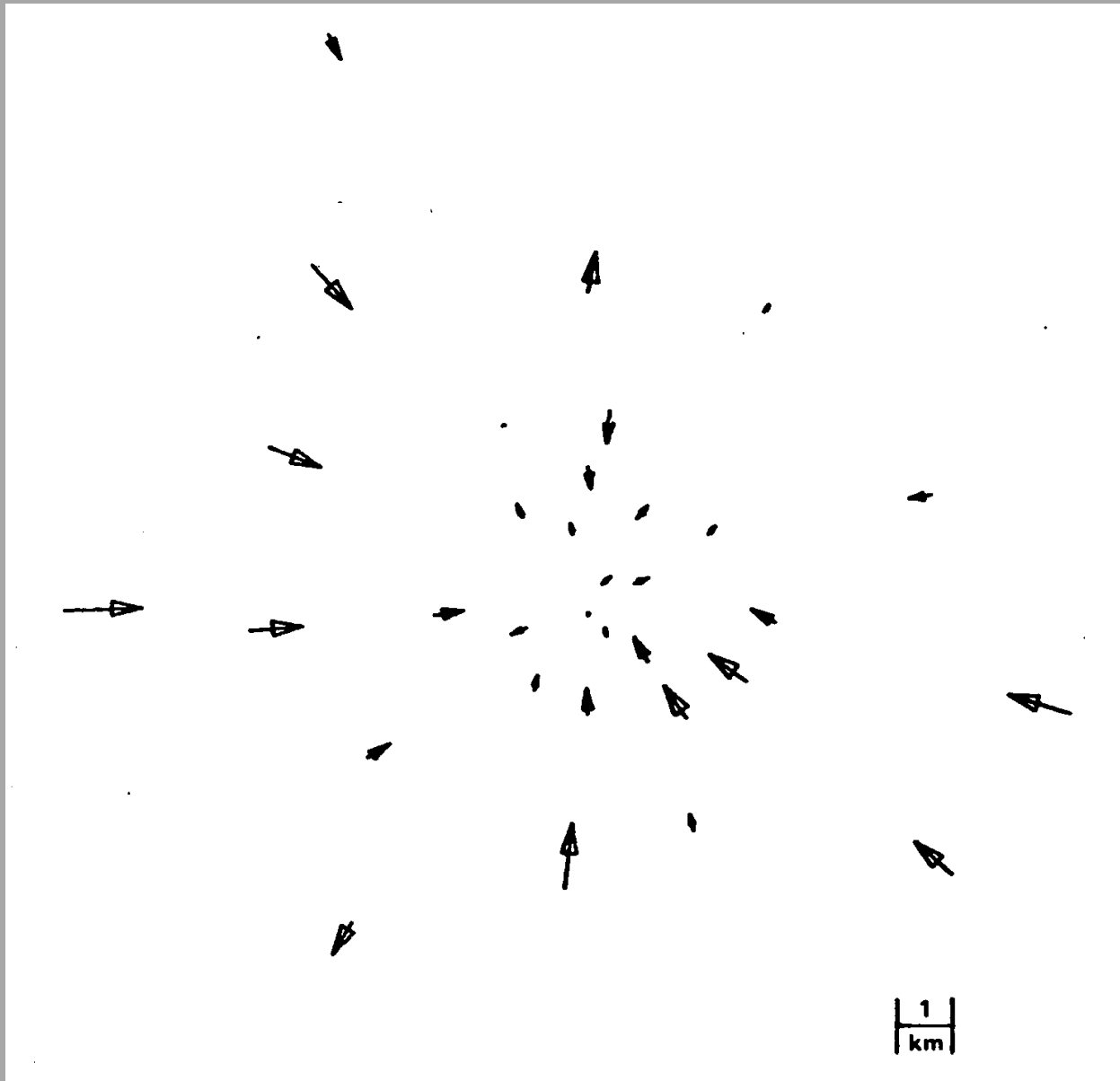
An alternative is a local vector to indicate the movement from each source location, showing the direction and magnitude of the net movement.

The computation is based on the asymmetry of the movement table.

Small directed vectors represent this movement on the next map.

# Munich Commuting

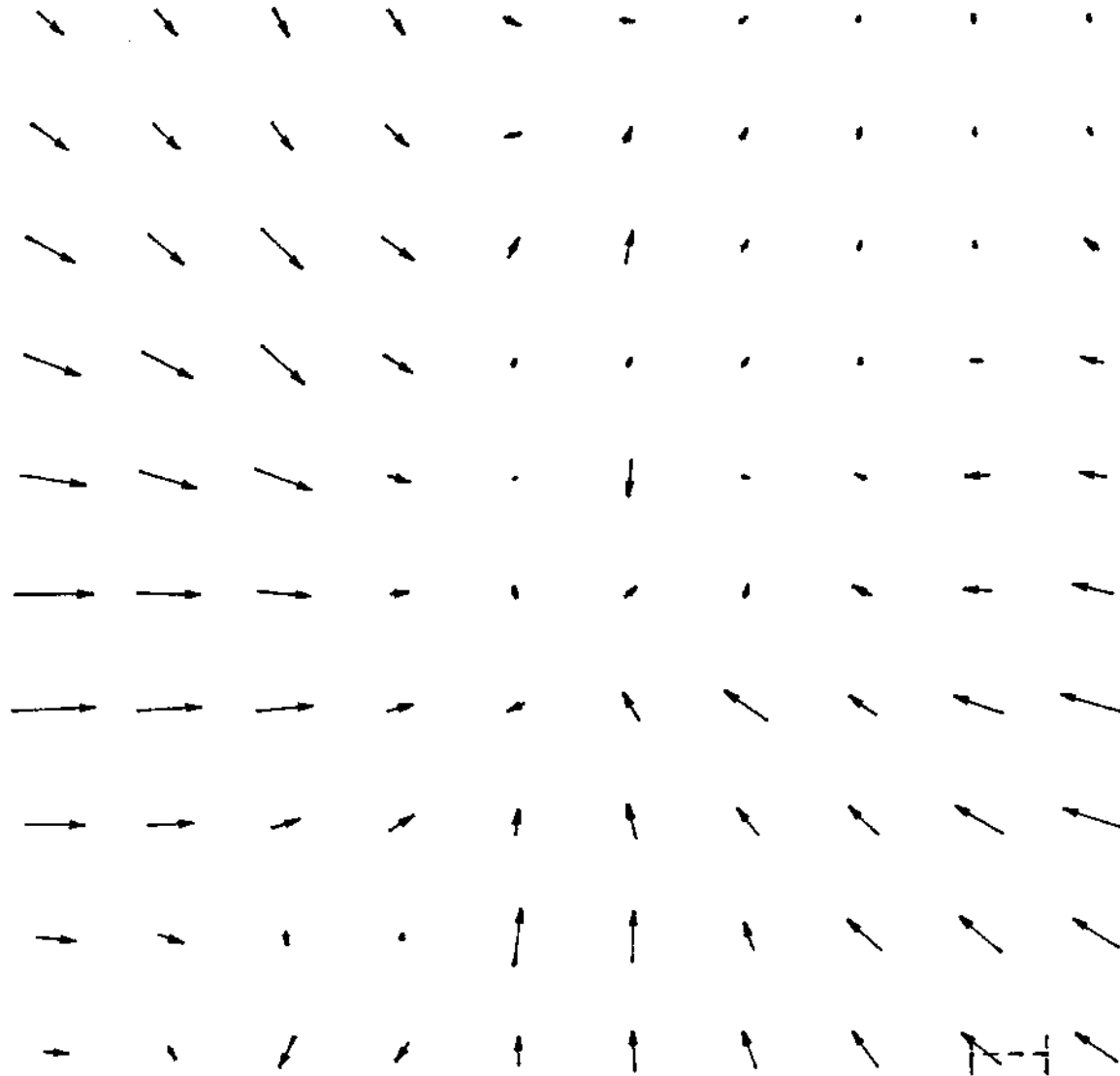
Displacement vectors



An interpolation is then performed to obtain a vector field from the isolated individual vectors.

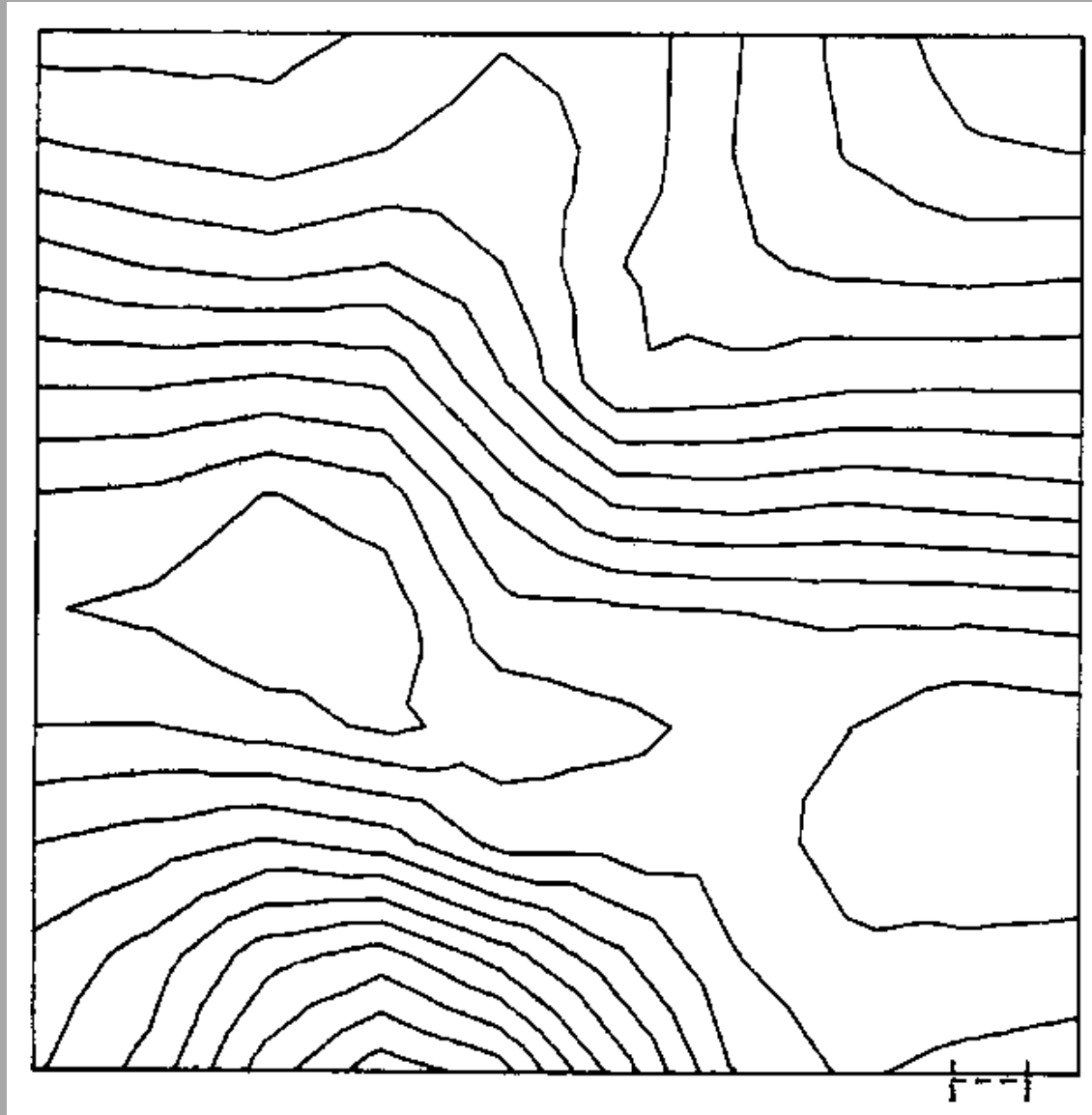
This is done to simplify the mathematical integration needed to obtain the forcing function.

# Interpolated Field of displacement vectors



# Computed Potential

based on the displacement vectors





The computed potential should have the vector field as its gradient.

This is a hypothesis that can be tested.

The base level of the potential is determined only up to a constant of integration.

The vector field, to be a gradient field, must be curl free. This can also be tested.

The attempt is now made to apply these ideas  
in a social space.

This can be considered a development of Lewin's  
*Topological Psychology* or his *Field Theory in  
the Social Sciences*.

The data represent citations between a small set of  
psychological journals. Larger citation tables are  
now also available.

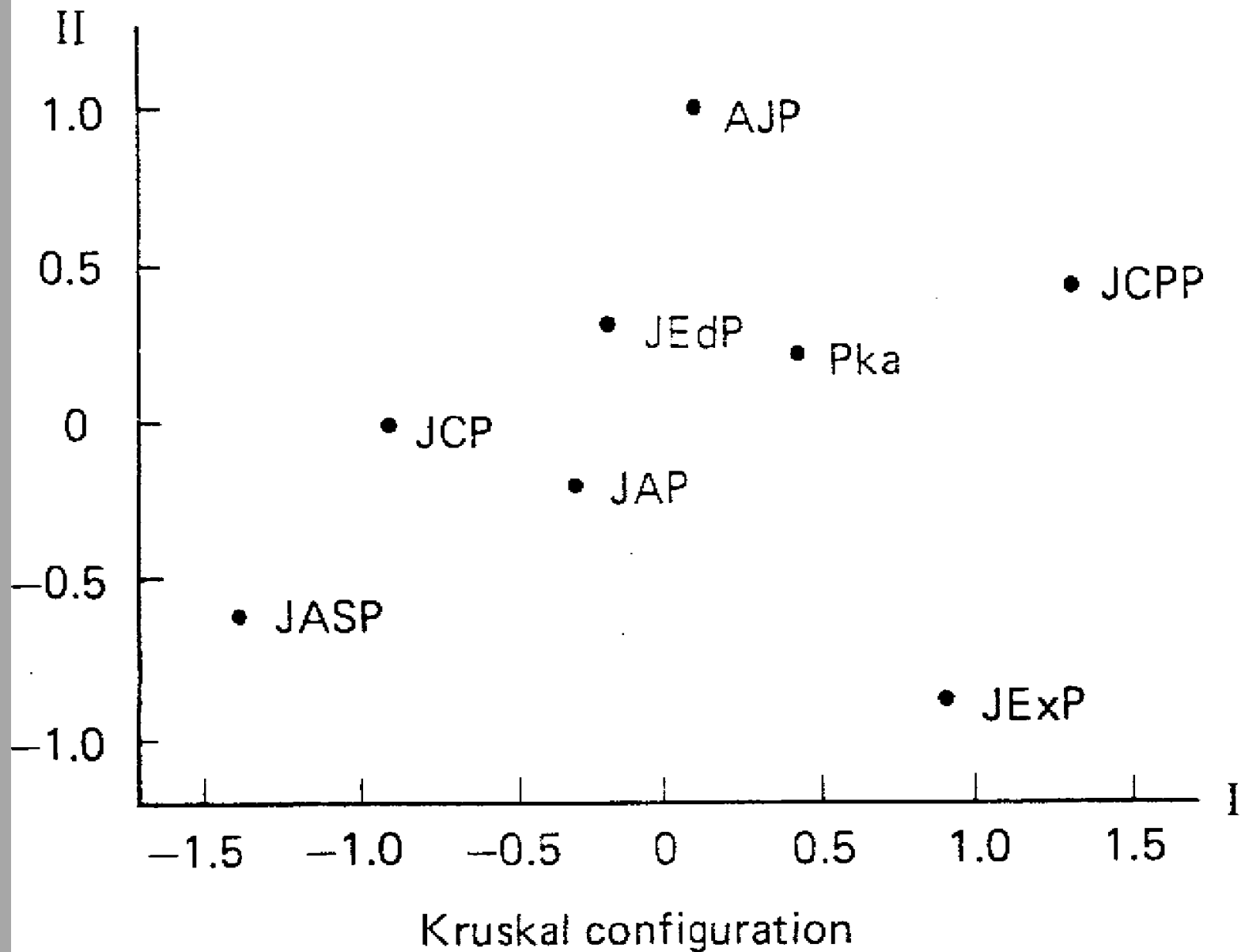
# Citations among psychology journals

Coombs et al 1970

Data from 1964

	<i>AJP</i>	<i>JASP</i>	<i>JAP</i>	<i>JCPP</i>	<i>JCP</i>	<i>JEdP</i>	<i>JExp</i>	<i>Pka</i>	<i>Total</i>
<i>American Journal of Psychology</i>	119	8	4	21	0	1	85	2	240
<i>Journal of Abnormal and Social Psychology</i>	32	510	16	11	73	9	119	4	774
<i>Journal of Applied Psychology</i>	2	8	84	1	7	8	16	10	136
<i>Journal of Comparative and Physiological Psychology</i>	35	8	0	533	0	1	126	1	704
<i>Journal of Consulting Psychology</i>	6	116	11	1	225	7	12	7	385
<i>Journal of Educational Psychology</i>	4	9	7	0	3	52	27	5	107
<i>Journal of Experimental Psychology</i>	125	19	6	70	0	0	586	15	821
<i>Psychometrika</i>	2	5	5	0	13	2	13	58	98
<i>Total</i>	325	683	133	637	321	80	984	102	3,265

# In Journal Space



To	Journal to Journal Citations									Net	
From									X	Y	Flow
AJP	119	8	4	21	0	1	85	2	125	910	-85
JASP	32	510	16	11	73	9	19	4	-1382	-644	91
JAP	2	8	84	1	7	8	16	10	-261	-237	3
JCPP	35	8	0	533	0	1	126	1	1302	366	67
JCP	6	116	11	1	225	7	12	7	-924	-2	64
JEdP	4	9	7	0	3	52	27	5	-180	324	27
JExP	125	19	6	70	0	0	586	15	904	-924	-163
Pka	2	5	5	0	13	2	13	58	416	207	-4

AJP Am J of Psychology

JASP J of Abnormal & Social Psychology

JAP J of Applied Psychology

JCPP J of Comparative & Physiological Psychology

JCP J of Consulting Psychology

JEdP J of Educational Psychology

JexP J of Experimental Psychology

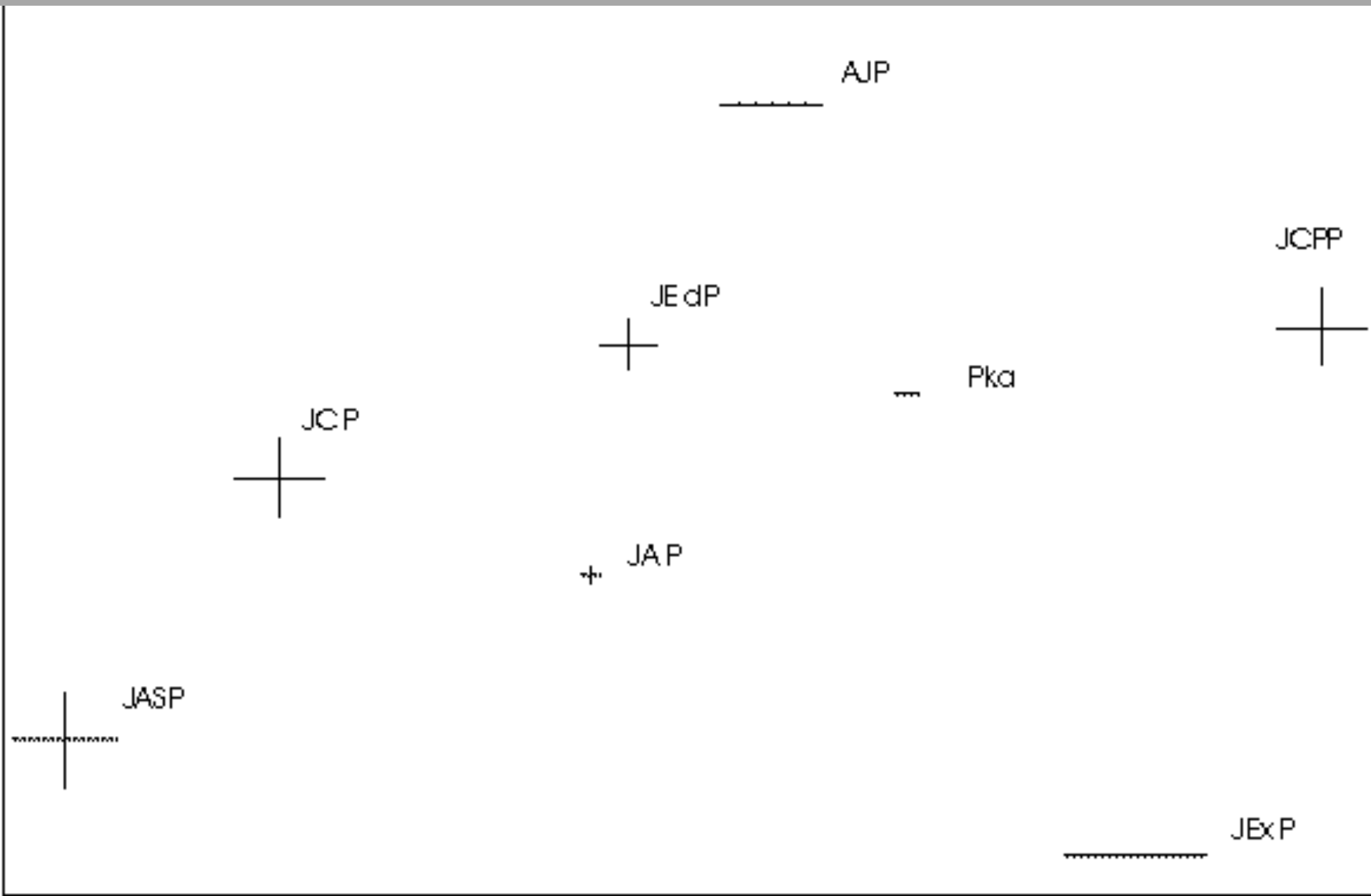
Pka Pyschometrika

C. Coombs, J. Dawes, A Twersky, 1970, *Mathematical Psychology*, Prentice Hall, Engelwood Cliffs, NY, Pages 73-75

The table gives the being-cited journal across the columns. But the information can be considered to move from that journal to the citing journal.

Therefore the transpose is used to produce the source to sink map.

# Journal Sources and Sinks



We now have an assignment problem. How to get 163 citations from JExp, 85 from AJP, & 4 from Pka to the 5 receiving journals, using only the marginals. There are obviously many possibilities

One solution is to use the “Transportation Problem” (Koopmans, Kantorovich, ~1949): Minimize  $M..d..$ , subject to  $M..J = O_I$ ,  $M_I.. = I_J$ ,  $M_{IJ} \geq 0$ , given the distances computed from the coordinates and using the simplex method for the solution.

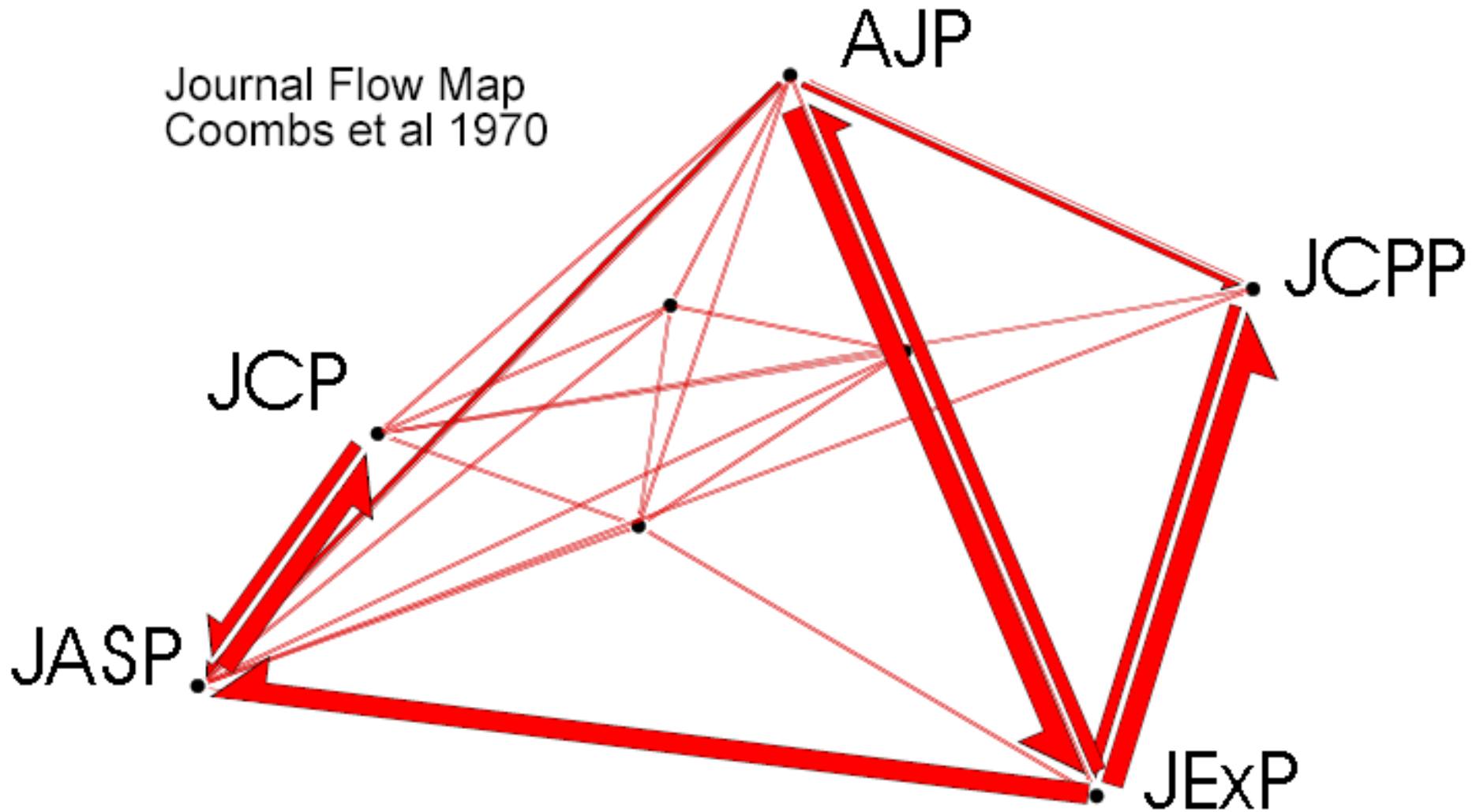
A more realistic solution is given by the quadratic transportation problem: Minimize  $M^2..d..$ , subject to the same constraints.

Both of these solutions result in discrete answers, and ‘shadow prices’. We are looking for a spatially continuous solution that allows vectors and streamlines, in order to determine spatial flow fields and a continuous potential.



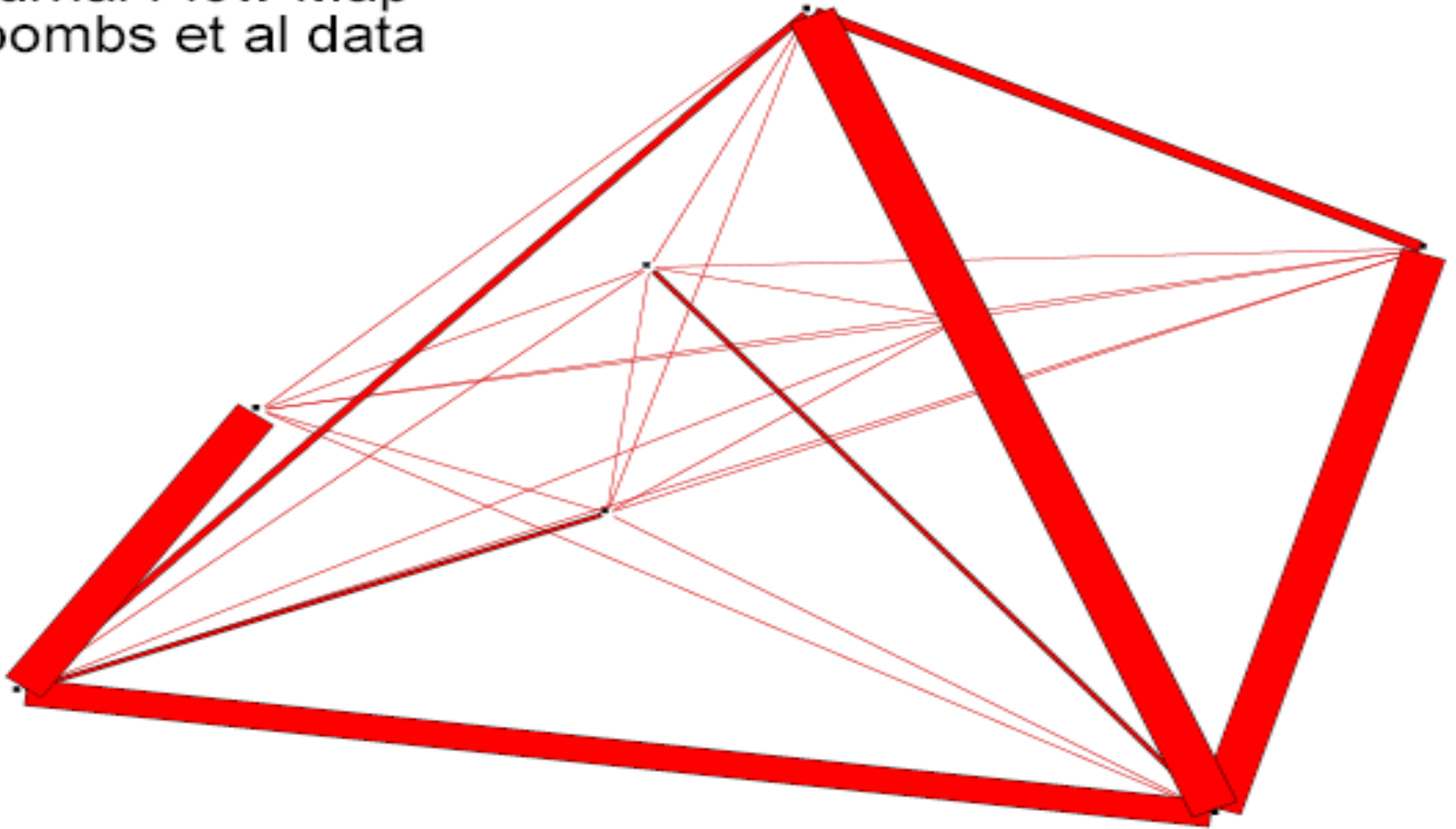
# The observed two-way flow between the journals

Journal Flow Map  
Coombs et al 1970



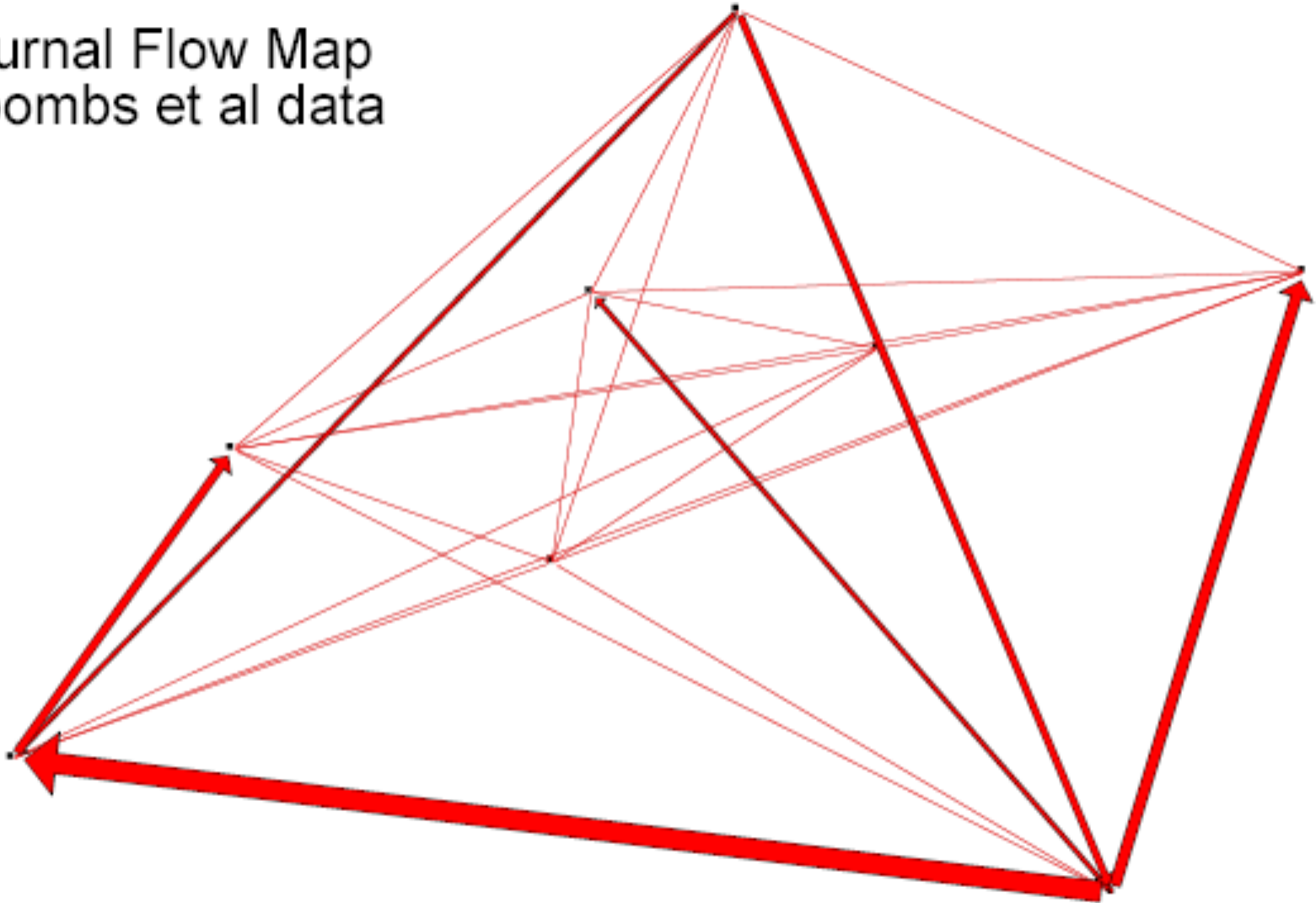
# The total flow between the journals

Journal Flow Map  
Coombs et al data



# The net flow between the journals

Journal Flow Map  
Coombs et al data



The next step is to compute the displacements between the cited journals.

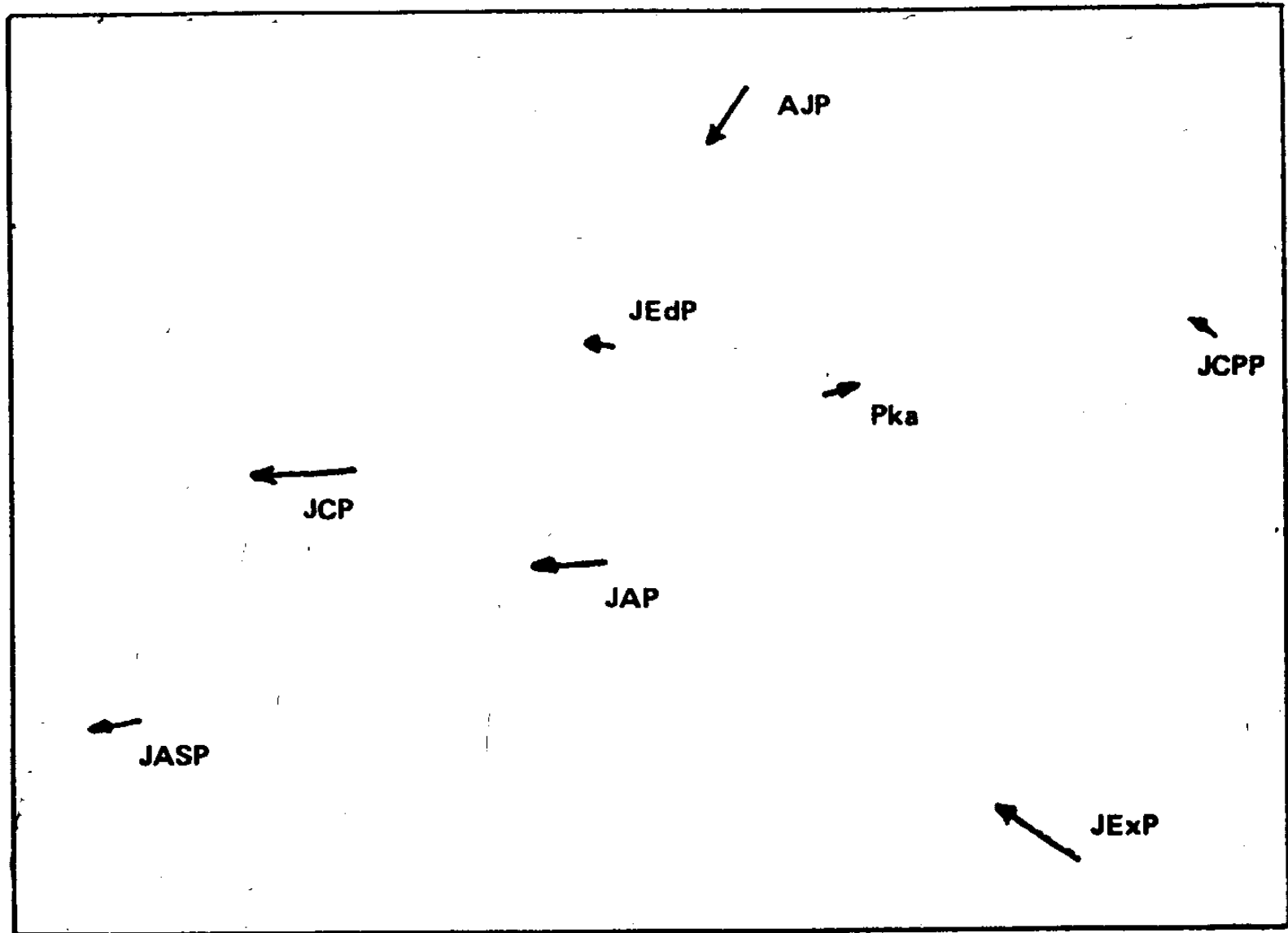
This is based on the asymmetry of the citations table.

The fundamental idea being that there exists a ‘wind’ making movement easier in some directions.

The mathematical details are given in a published paper.

W. Tobler, 1976, “Spatial Interaction Patterns”, *J. of Environmental Systems*, VI(4):271-301

# Displacement between Journal Citations

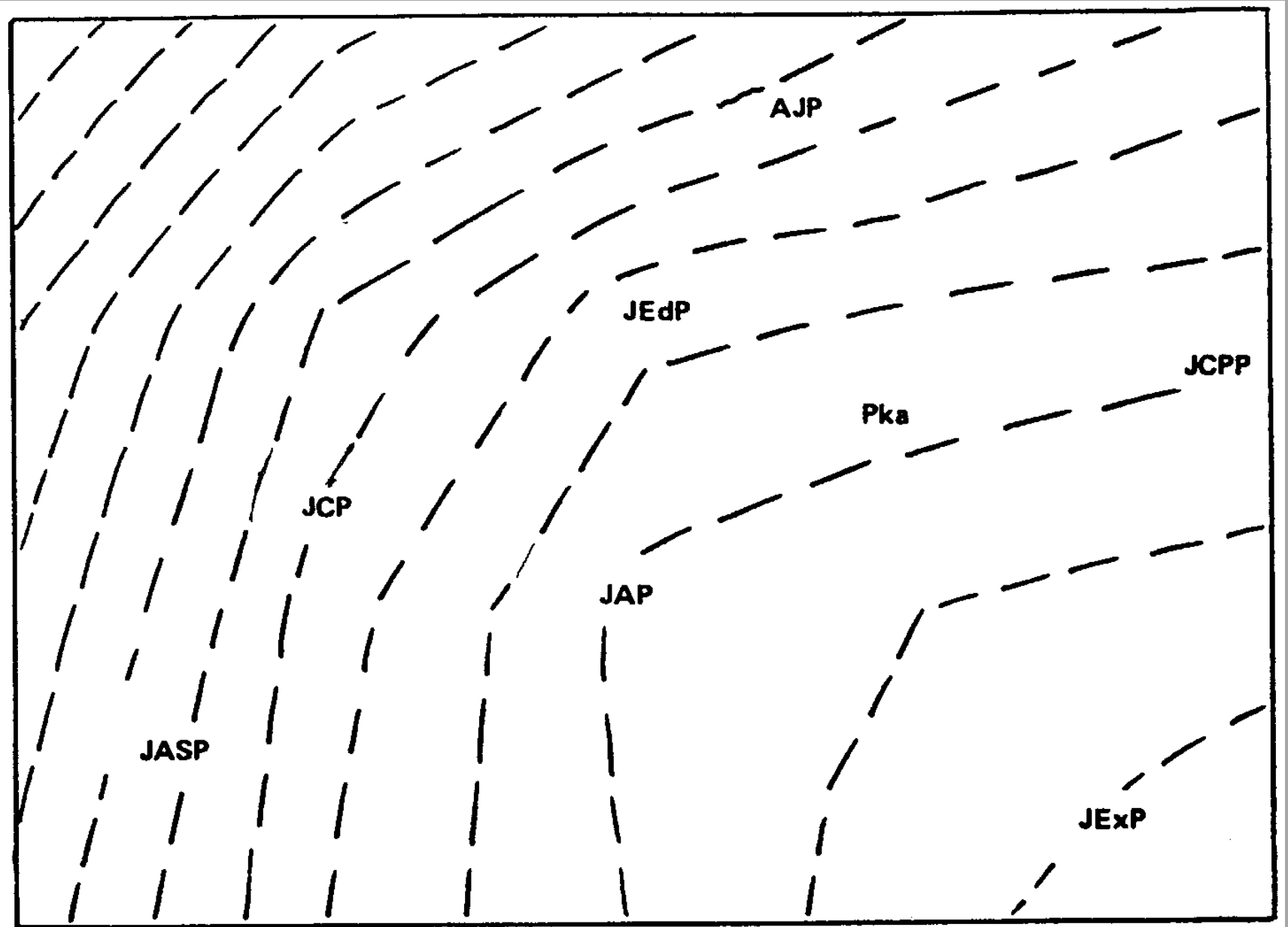


Then the potential is computed by integration.

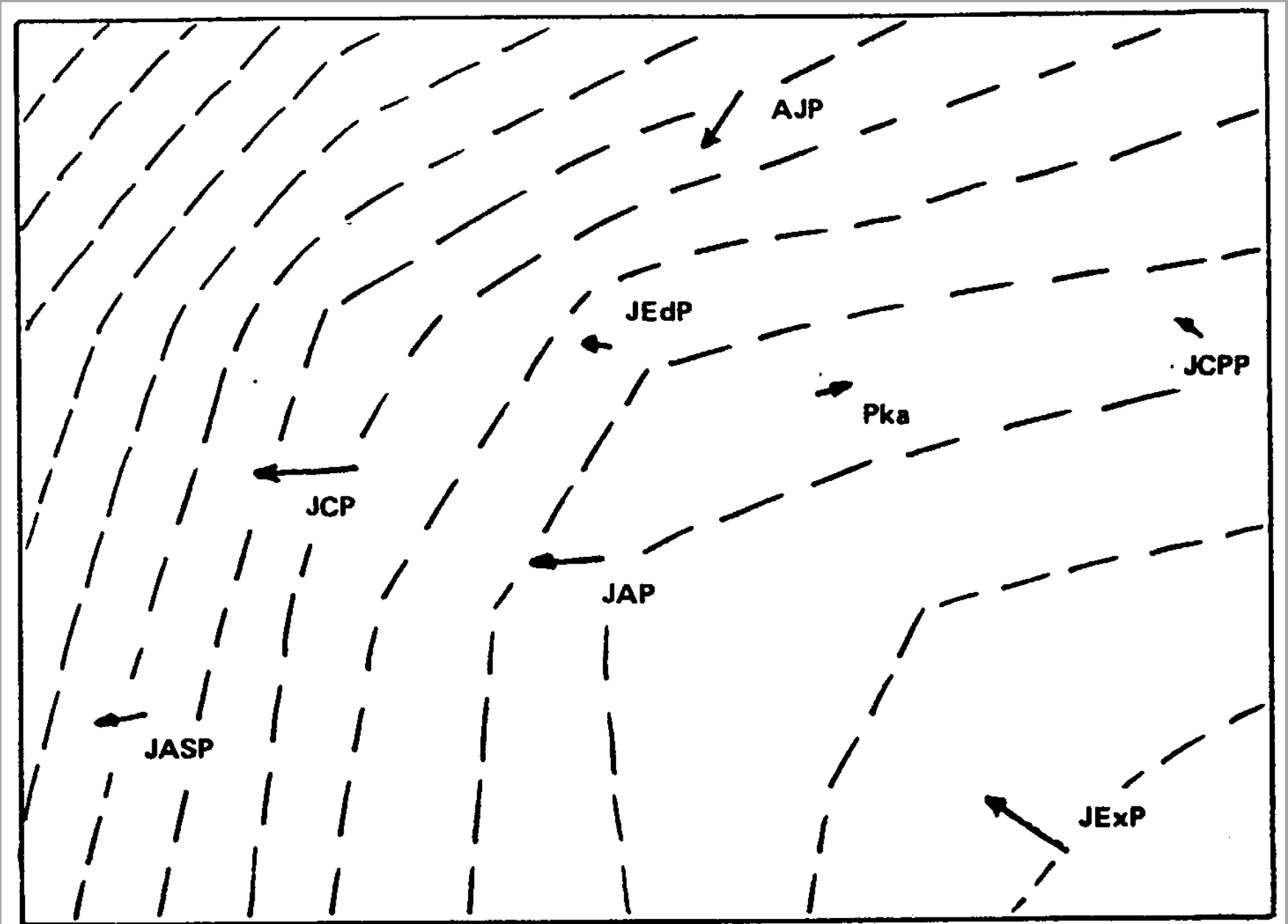
This potential should be such that its gradient coincides with the displacement vectors.

It may be necessary to use an iteration to obtain this result.

# Journal Potential Function



# Flow and Potential between Psychological Journals





## Some questions

Suppose a new psychological journal were started.

Where should it be inserted into in this space?

Does it make sense to treat journal citations as being located in a continuous two-dimensional social space?

Can other social data be treated in a similar fashion, for example social mobility tables?

And more general network data?

# CONCLUSION

I have given some speculative thoughts on how one might represent network relations with vectors, fields, and scalar potentials in a continuous social space.

Still needed are error estimates.

Your comments are desired.

Thank you for your attention.

<http://www.geog.ucsb.edu/~tobler>

## References

K. Boyack, 2004, XXX, *Proceedings, National Academy of the United States*, 101, Supplement 1, (April 6): 5192-5199.

C. Coombs, J. Dawes, A. Twersky, 1970, *Mathematical Psychology*, Prentice Hall, Englewood Cliffs, NY.

D. Fliedner, 1962, "Zyklonale Tendenzen bei Bevölkerungs und Verkehrsbewegungen in Städtischen Bereichen untersucht am Beispiel der Städte Göttingen, München, und Osnabrück", *Neues Archiv für Niedersachsen*, 10:15 (April 4): 277-294 (Table 2, p. 281, map following p. 285).

K. Lewin, 1936, *Principles of Topological Psychology*, McGraw Hill, New York

K. Lewin, 1951, *Field Theory in the Social Sciences*, Harper, New York.

W. Tobler, 1976, "Spatial Interaction Patterns", *J. of Environmental Systems*, VI (4) 1976/77, pp. 271-301.

W. Tobler, 1981, "A Model of Geographic Movement", *Geographical Analysis*, 13 (1): 1-20.

W. Tobler, 1996, "A Graphical Introduction to Surveying Adjustment", *Cartographica*, 33-42.

S. Wasserman, Faust, K., 1994, *Social Network Analysis: Methods and Application*, Cambridge University Press, Cambridge.

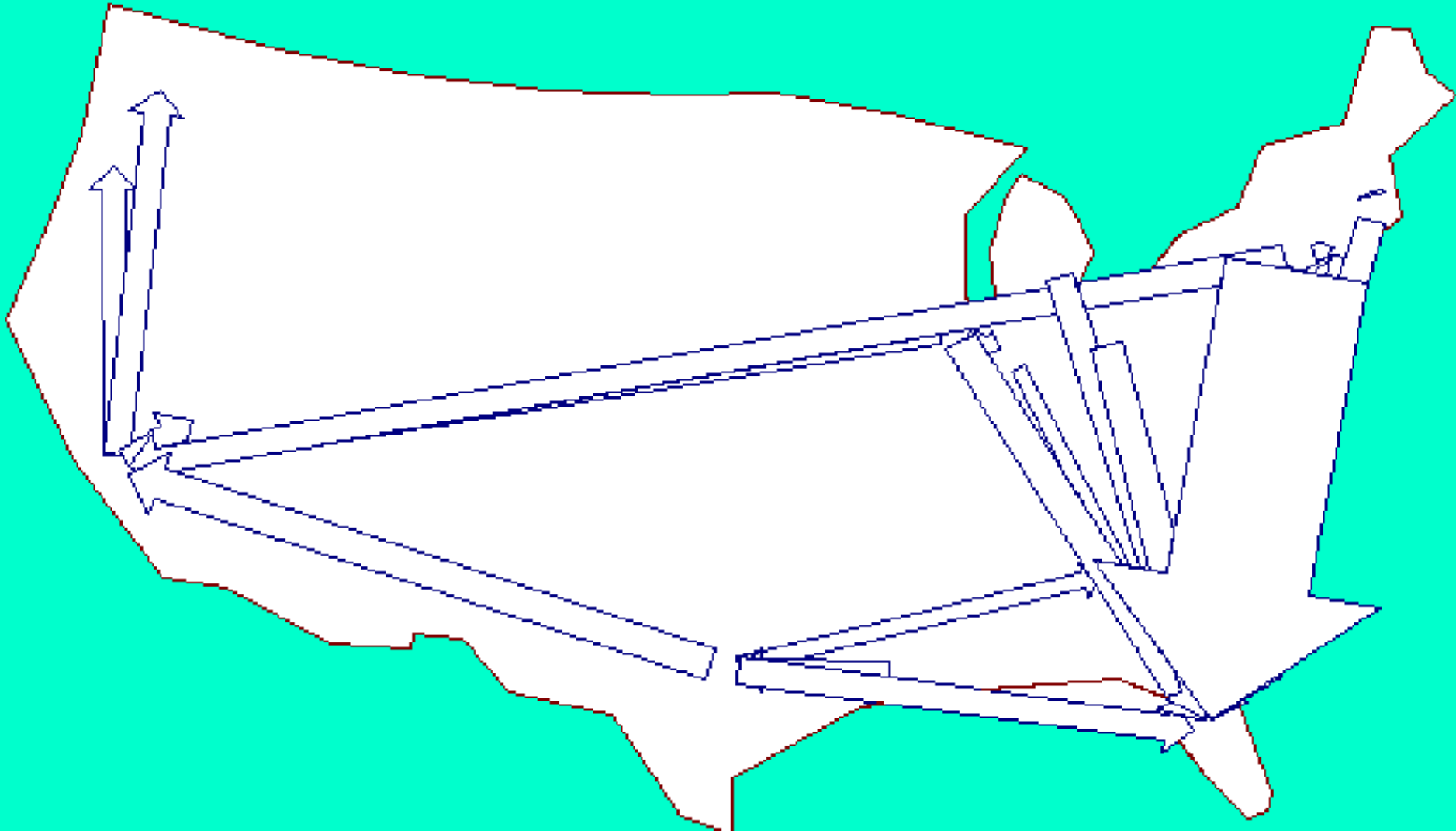
# Another geographic example as motivation

This was not presented at Sunbelt XXV

# The conventional net movement map

Based on movement between state centroids

(Computer sketch. Optimum deletion: values below mean ignored)

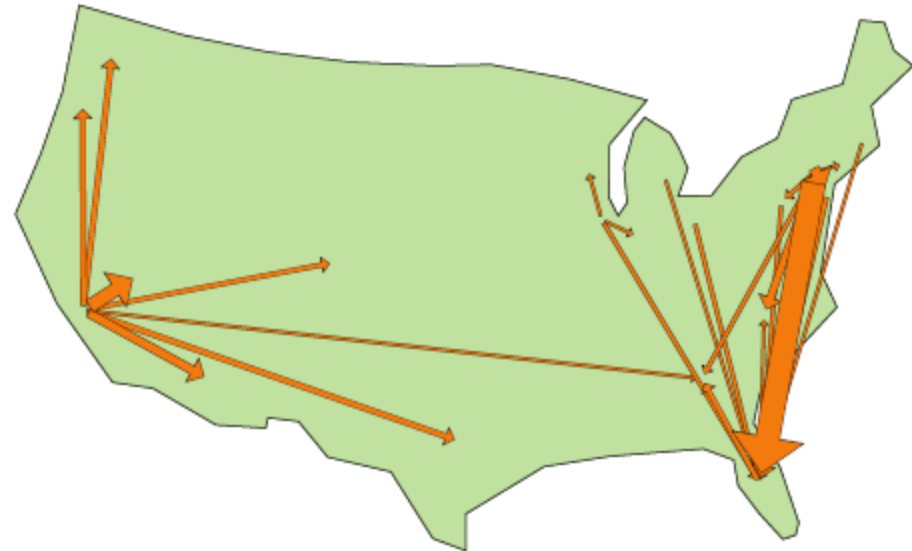
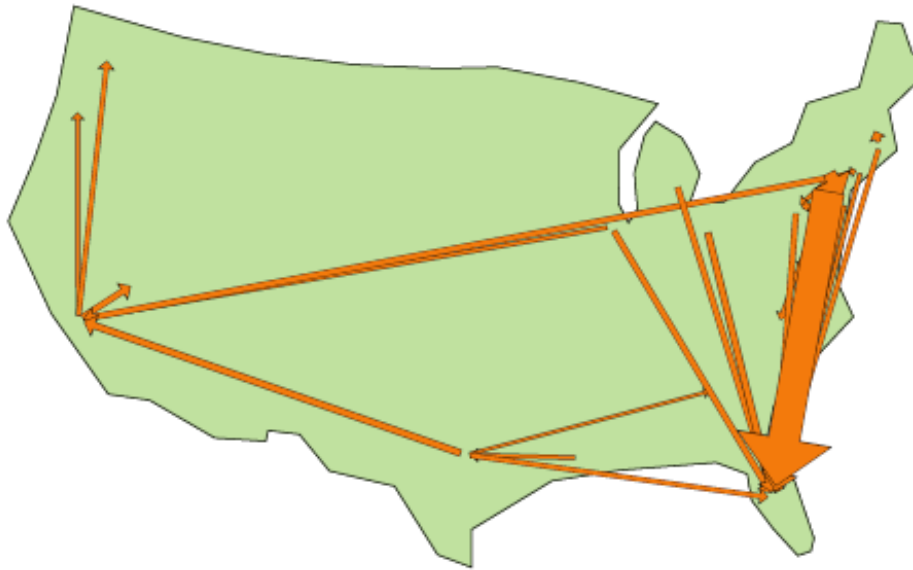


# Net Migration in the United States

These patterns persist for a long time

1985-1990

1995-2000



# Several ways of representing geographic space in a continuous fashion for migration (or other) studies are:

Writing scalar values as continuous functions of latitude and longitude (or rectangular or polar plane coordinates), perhaps estimated by least squares, as two dimensional algebraic or trigonometric polynomials, splines, eigenfunctions, or spherical harmonics or wavelets. This can be considered as an elaboration of spatial trend analysis. See:

W. Tobler, 1969, Geographic filters and their inverses, *Geog. Anal.*, 1:234-253

W. Tobler, 1992, Preliminary representation of world population by spherical harmonics, *Proc. Natl. Acad. Sci USA*, 89: 6262-6264.

Writing vector fields, or interaction data, in a similar fashion as a four dimensional spline or polynomial function of the origin & destination location coordinates. See:

P. Slater, 1993, "International Migration & Air Travel: Smoothing & Estimation" *Appl. Math. & Comp.*, 53: 225-234

Expanding regression coefficients in a geographically weighted manner.  
See

J. Jones, E. Casetti, 1992, *Applications of the expansion method*, Routledge, London

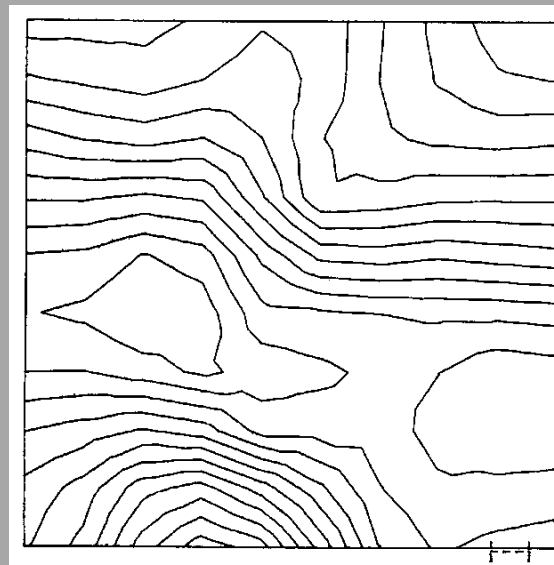
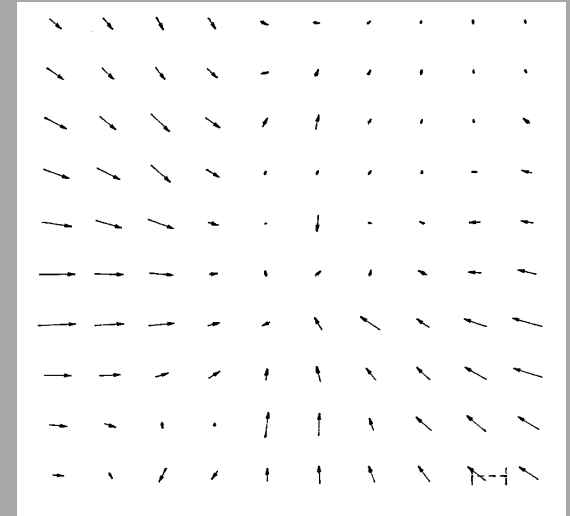
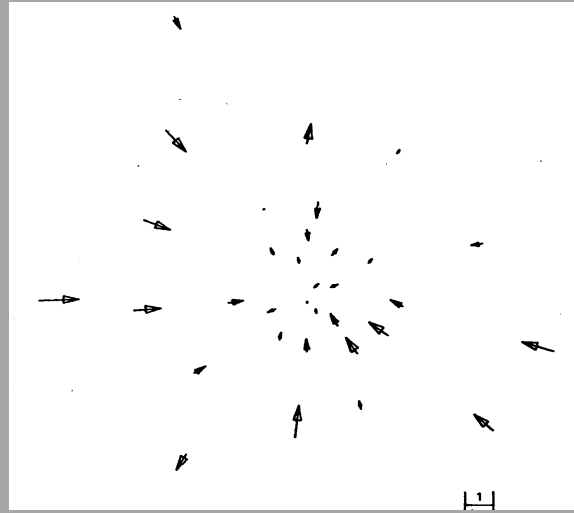
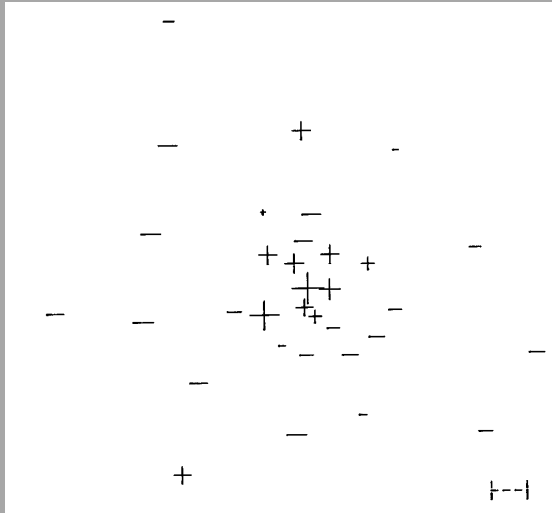
S. Fotheringham, et al, 2002, *Geographically weighted regression*, Wiley, Chichester.

Approximation by a two dimensional lattice, as in the present study.

You have just seen this simple example

# Commuting in Munich 1939

Left to right: from places (sources -) and to places (sinks +),  
vectors, and interpolated vectors, and the implied potential field





In the previous slide the source places (origins) were used to make a set of vectors pointing towards the sinks (destinations). These were then interpolated to obtain a field of vectors. Integration (in the mathematical sense) was then used to construct a potential field, shown by contours. The magnitude and direction of the vectors correspond to the gradient of the potential surface.

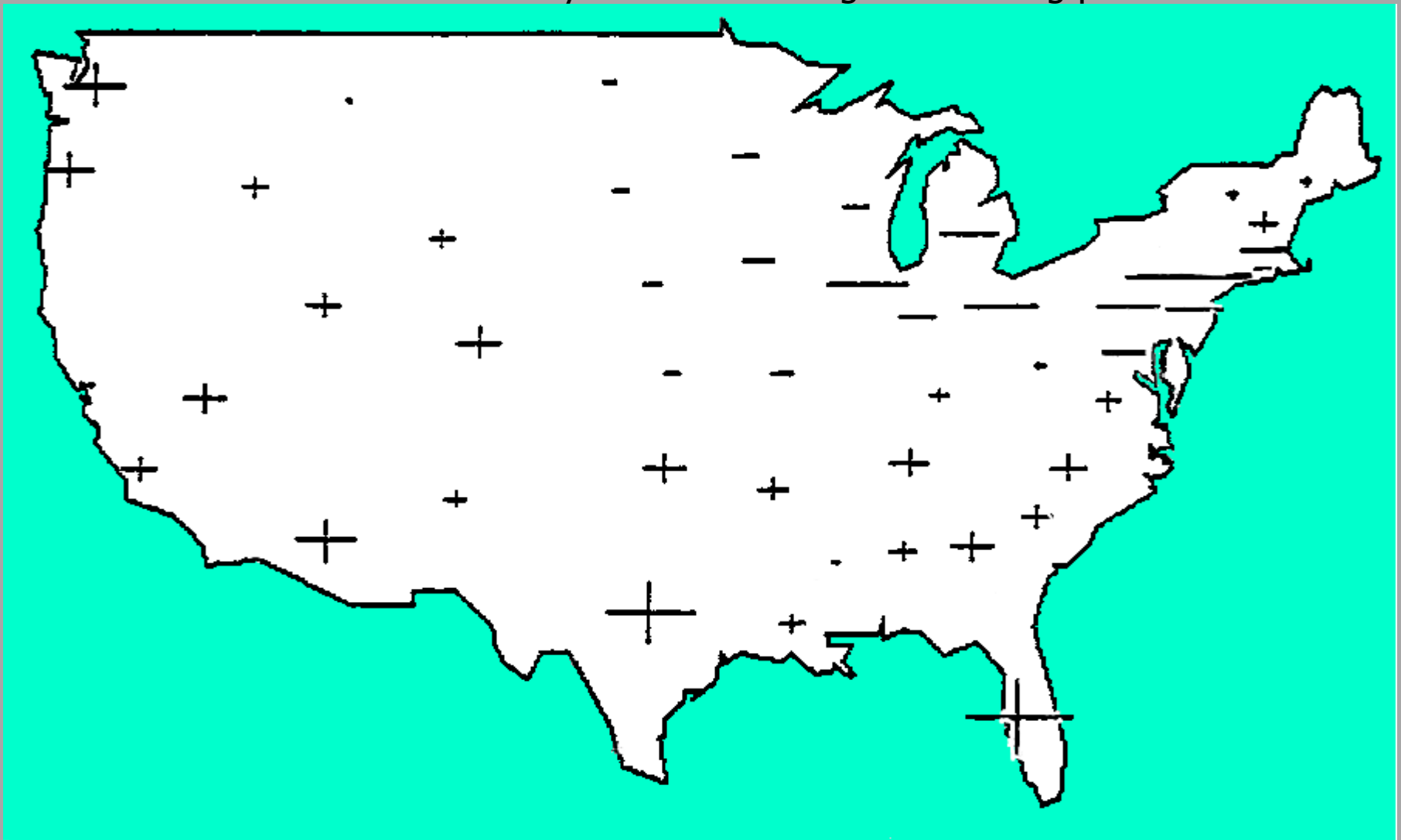
In the next example something similar is done. Except that the model is set up on the basis of interpolating the sources (out-flows) and sinks (in-flows) for the contiguous US. Then the potential is computed directly. The gradient vector field is obtained from this potential field.

To carry out this operation first assign the in-migration and out-migration totals to each state. Then 'rasterize' the region of interest into a large set of equally spaced nodes and spread the population change over the nodes in each state. This allows the treatment to approximate a **continuous** migration surface and is illustrated on the next slides. This is one of several ways to treat geographic space in a continuous fashion.

# Gaining and Losing States

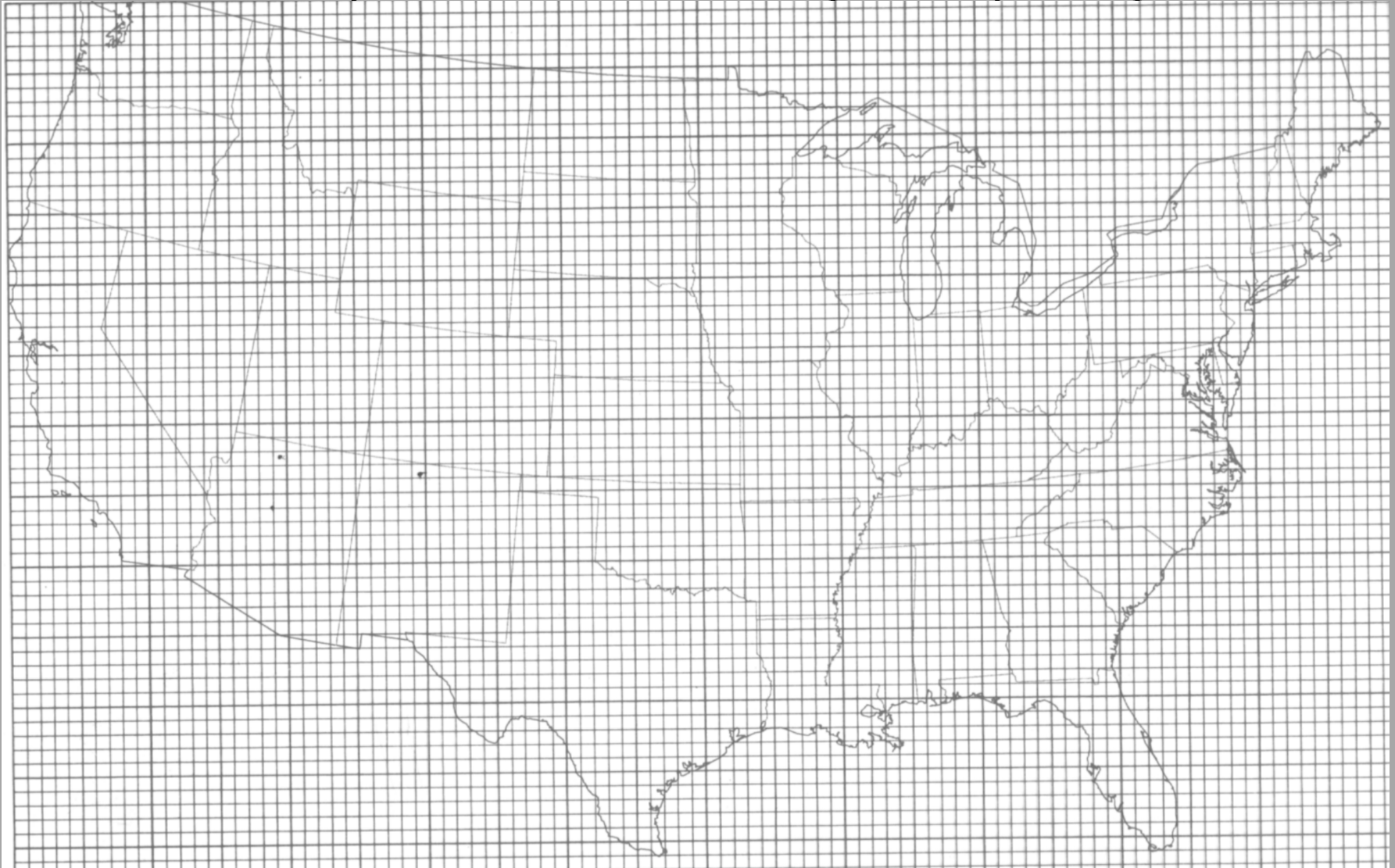
Based on the marginals of a 48 by 48 migration table,  
1965-1970 data.

Sketch in the boundary between leaving and arriving places.



# “Rasterize” the USA to form a lattice.

Use a point-in-polygon program to assign nodes to individual states. Then assign in and out values to these nodes. There will be one equation for each node on this raster. Then solve the system of  $\sim 6000$  simultaneous equations to yield the potential.



In the U.S. example both the in-migration and the out-migration amounts were spread over all of the nodes making up each of the individual states.

Pycnophylactic reallocation was used to do this.

As a related item, world population estimates are now available by fine geographic (lat/lon) quadrangles.

Why does the census not release migration data in this format, by latitude and longitude quadrangles?

If that were done then the spherical version of the model to be described could be used directly.

Studies of urban commuting can also benefit from data recorded in a raster format instead of irregularly shaped traffic zones.

W. Tobler, 1997, "Movement Modeling on the Sphere", *Geographic and Environmental Modeling*, 1(1): 97-103.

# Now we need to derive the continuous version of the Push-Pull model.

In the discrete case there is one equation for every pair of places:

$$M_{ij} = (R_i + E_j) / D_{ij}$$

obtained by solving the simultaneous pair for the Lagrangians:

$$\sum_{j=1}^c R_i / D_{ij} + \sum_{j=1}^c E_j / D_{ij} = 2 O_i$$

$$\sum_{i=1}^r R_i / D_{ij} + E_i \sum_{i=1}^r 1 / D_{ij} = 2 I_j$$

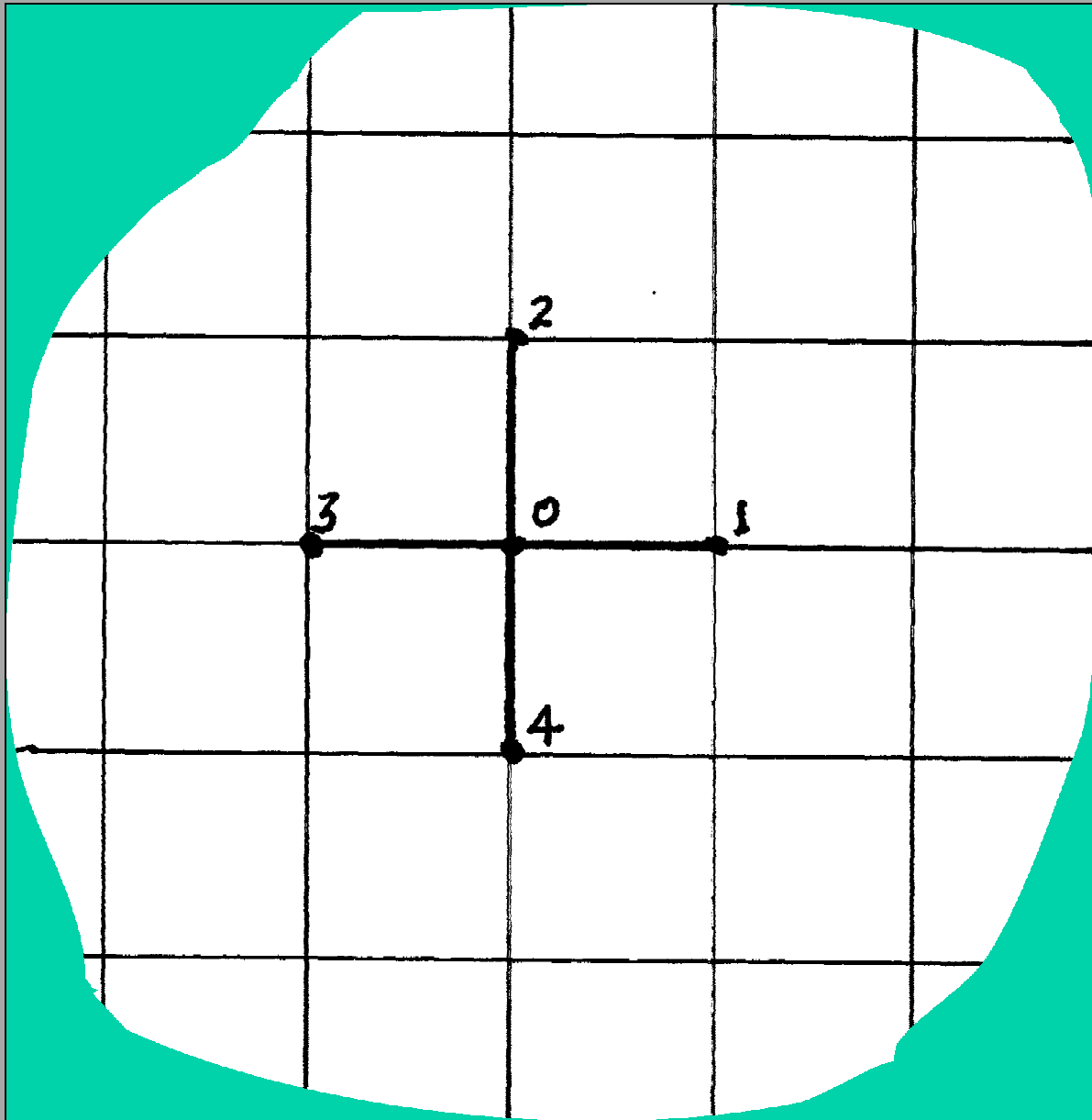
The  $E$  (‘pulls’) and  $R$  (‘pushes’) are the Lagrangians.

These simultaneous equations are solved for the pushes and pulls.

Also obtained were the ‘Attractivity’  $A = E - R$  and the ‘Turnover’  $T = E + R$ .

# In the raster look at one node and its neighbors

A raster is a special kind of network where movement takes place between neighboring nodes



# Derivation of a continuous model for the grid

In the push-pull model  $M_{ij} = (R_i + E_j) / D_{ij}$

For the square mesh take all  $D_{ij}$  to be the same. Set them equal to 1.

Use the subscript 0 for the center node, and index the neighbor nodes from 1 to 4

Then the moves **from** the center to the neighbors is

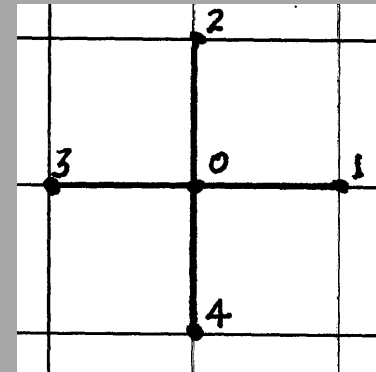
$$M_{01} = R_0 + E_1$$

$$M_{02} = R_0 + E_2$$

$$M_{03} = R_0 + E_3$$

$$M_{04} = R_0 + E_4$$

$$M_{0j} = 4 R_0 + E_1 + E_2 + E_3 + E_4$$



But  $M_{0j}$  are the moves out of node 0, and this is  $O_j$  the outsum.

In the same way  $M_{10} = R_1 + E_0$ , etc for  $M_{20}$ ,  $M_{30}$ ,  $M_{40}$ .

These are the moves **into** node 0 from the neighbors, and this is  $I_i$ .

Thus the pair of equations become

$$O_j = 4 R + E_1 + E_2 + E_3 + E_4$$

$$I_i = 4 E + R_1 + R_2 + R_3 + R_4$$

after dropping the subscript for the central node.

There is one pair of equations for each node.

(A non-regular set of neighbors could also be used)..

An aside:

Incorporating differential transport disutilities into the model.

From the previous slide we can insert a differential transport weight factor into the movement, as follows:

$M_{01} = R_0 + E_1$  becomes  $= (R_0 + E_1)/W_{01}$  where  $W_{01}$  is the equivalent of  $d_{01}$  but more realistic (for example road distance, or travel time or cost). Then similarly for all  $M_{0j}$ .

Now do the same for  $M_{10}$  inserting a  $W_{10}$ , etc. Recognize that  $W_{01}$  is not the same as  $W_{10}$  and that the weights will be different across every edge, and that they may change rapidly with time. Adjacent cells will naturally have two common, but differentially directed, link values.

It might be helpful to draw and label weights for a system of nine cells.

Doing this naturally leads to a rather more complicated system of equations.



(aside continued)

As a result:

$$R_0 = [O_J - (\sum E_k/w_{0k})] / \sum 1/w_{0k}$$

$$E_0 = [I_I - (\sum R_k/w_{k0})] / \sum 1/w_{k0}$$

The summations are over  $k = 1$  to  $4$

All  $w_{pq}$  and  $I_i$  and  $O_j$  are assumed known.

The same set of equations hold for all cells except those on the borders of the region.

Known are  $2 w'$  s per edge +  $2 * p * q - 1$  in and outsums ( $I'$  s and  $O'$  s) minus  $4 * (p + q)$  (Dirichelet or Neumann) values at the edges.

Unknown are  $2 * p * q$  pushes ( $R'$  s) and pulls ( $E'$  s).

Can this system be solved for all  $R'$  s and  $E'$  s?

The distance values  $D_{ij}$ , as constants, have been dropped in the square mesh, for pedagogic purposes, but not a mathematical necessity.

Each place, except along the margins of the region, will have four neighbors.

Just derived were the two equations at each node:

$$4E = I - (R_1 + R_2 + R_3 + R_4), \quad 4R = O - (E_1 + E_2 + E_3 + E_4).$$

The central E and R require no subscript; their neighboring locations are indexed from one to four - or if you wish - North, South, East, and West directions.

Now add  $-4R$  to both sides of the first equation and  $-4E$  to both sides of the second, rearrange slightly, and using  $T = E + R$ , to obtain

$$R_1 + R_2 + R_3 + R_4 - 4R = I - 4T, \quad E_1 + E_2 + E_3 + E_4 - 4E = O - 4T,$$

**The left-hand sides are recognized as finite difference versions of the Laplacian.**

Thus we can write, approximately and for a limiting uniform fine mesh, the pair

$$\partial^2 R / \partial u^2 + \partial^2 R / \partial v^2 = I(u,v) - 4T(u,v),$$

$$\partial^2 E / \partial u^2 + \partial^2 E / \partial v^2 = O(u,v) - 4T(u,v),$$

assuming that R and E are differentiable spatial functions and that I and O are continuous densities given as functions of the Cartesian coordinates u and v.

## Now, making use of the continuous movement model.

<http://www.geog.ucsb.edu/~tobler/presentations/A Flow Talk.pps>

In this continuous model, we have a coupled system of two simultaneous partial differential equations covering the entire region. These equations can be combined to yield either gross movements or net movements.

For the simultaneous movement in both directions at each pair of places **add** the two equations to get the 'turnover' potential.

For the net movement we need only the difference between the 'in' and 'out' at each node for the 'attractivity' potential, as follows:

By subtraction from the two previous equations we have the single partial differential equation

$$\partial^2 A / \partial u^2 + \partial^2 A / \partial v^2 = I(u,v) - O(u,v),$$

where A can be thought of as the attractivity of each location. This is the well-known Poisson equation for which numerical solutions are easily obtained. Once  $A(u,v)$  - the potential - has been found from this equation, the net movement pattern is given by the vector field

$$\mathbf{V} = \text{grad } A,$$

or by the difference in potential between each pair of mesh nodes.

# The result is a system of linear partial differential equations

The number of simultaneous equations depends of the mesh size

These are solved by a finite difference iteration to obtain the potential field (after specifying a boundary condition).

This potential can be contoured and its gradient computed and drawn on a map.

In other words a map is computed using a continuous movement model.

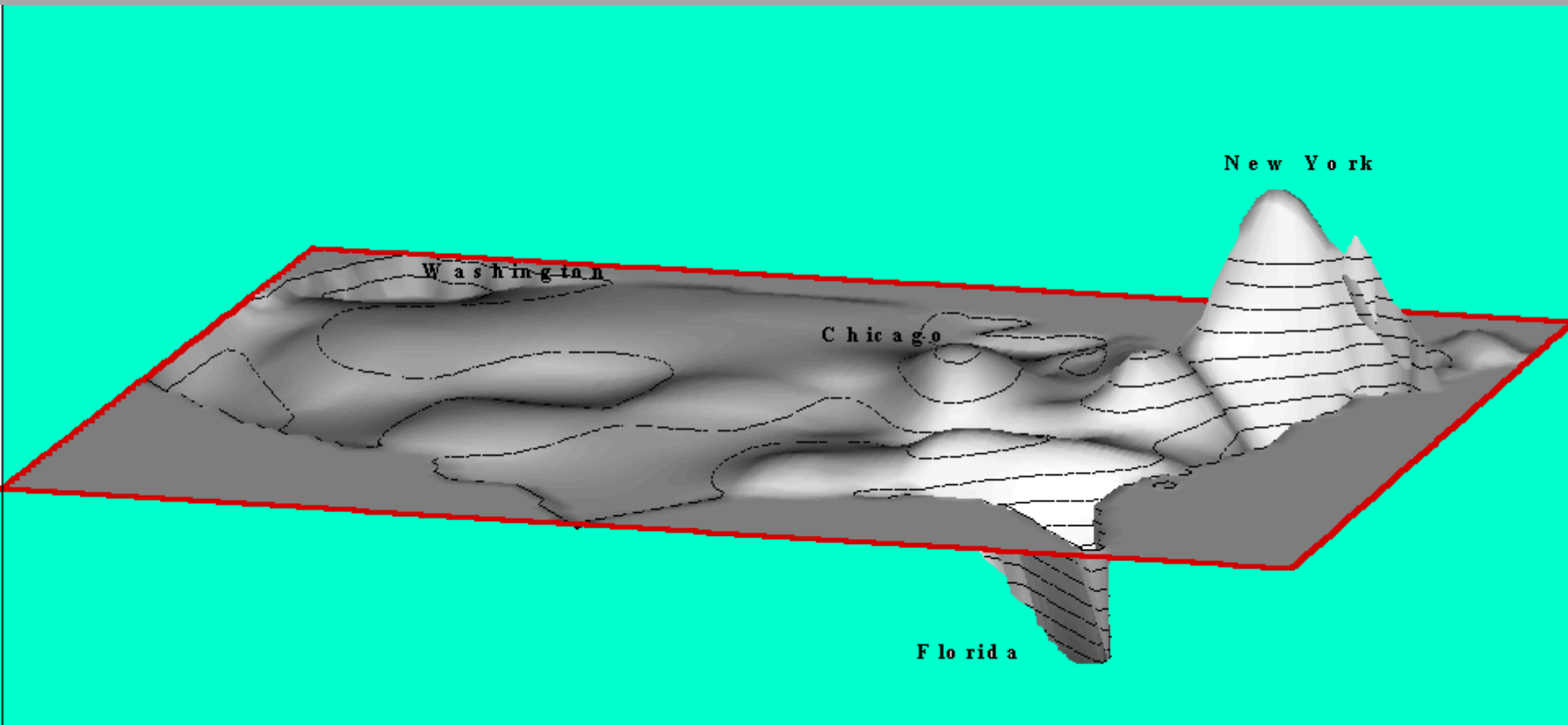
Estimates of the potentials for two different populations (male & female for example) can be added to get the correct potential for the sum.

W. Tobler, 1981, "A Model of Geographic Movement", *Geogr. Analysis*, 13 (1): 1-20  
G. Dorigo, & Tobler, W., 1983, "Push Pull Migration Laws", *Annals, AAG*, 73 (1): 1-17.

The potential gives

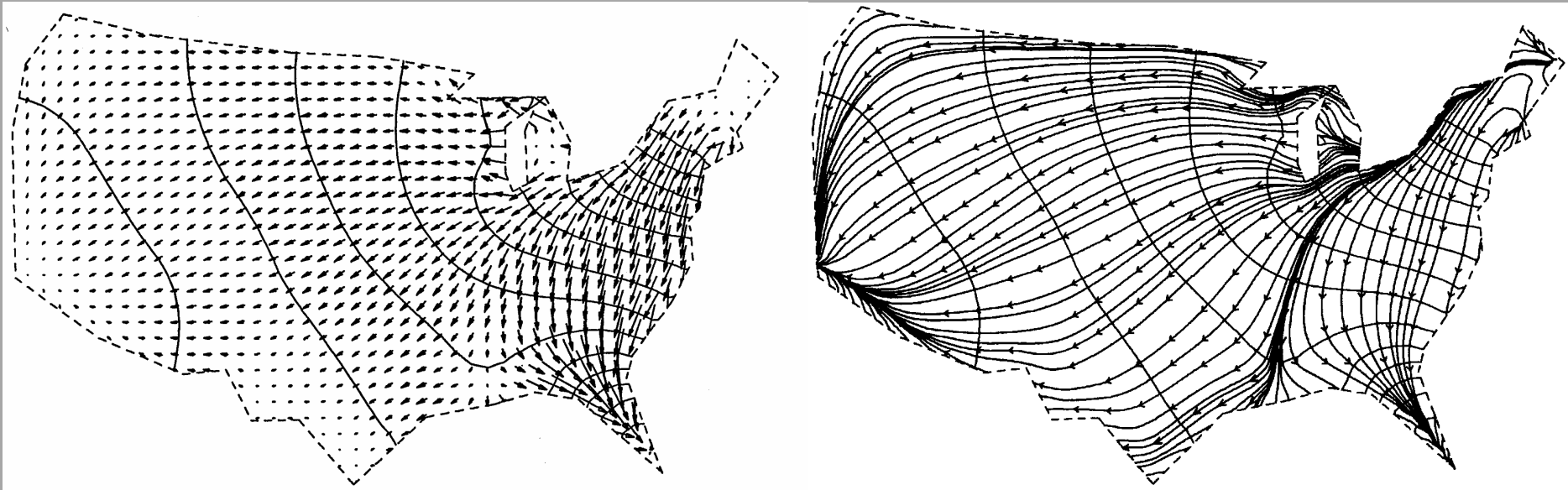
# The Pressure to Move in the US

Based on the continuous spatial model  
Using state data



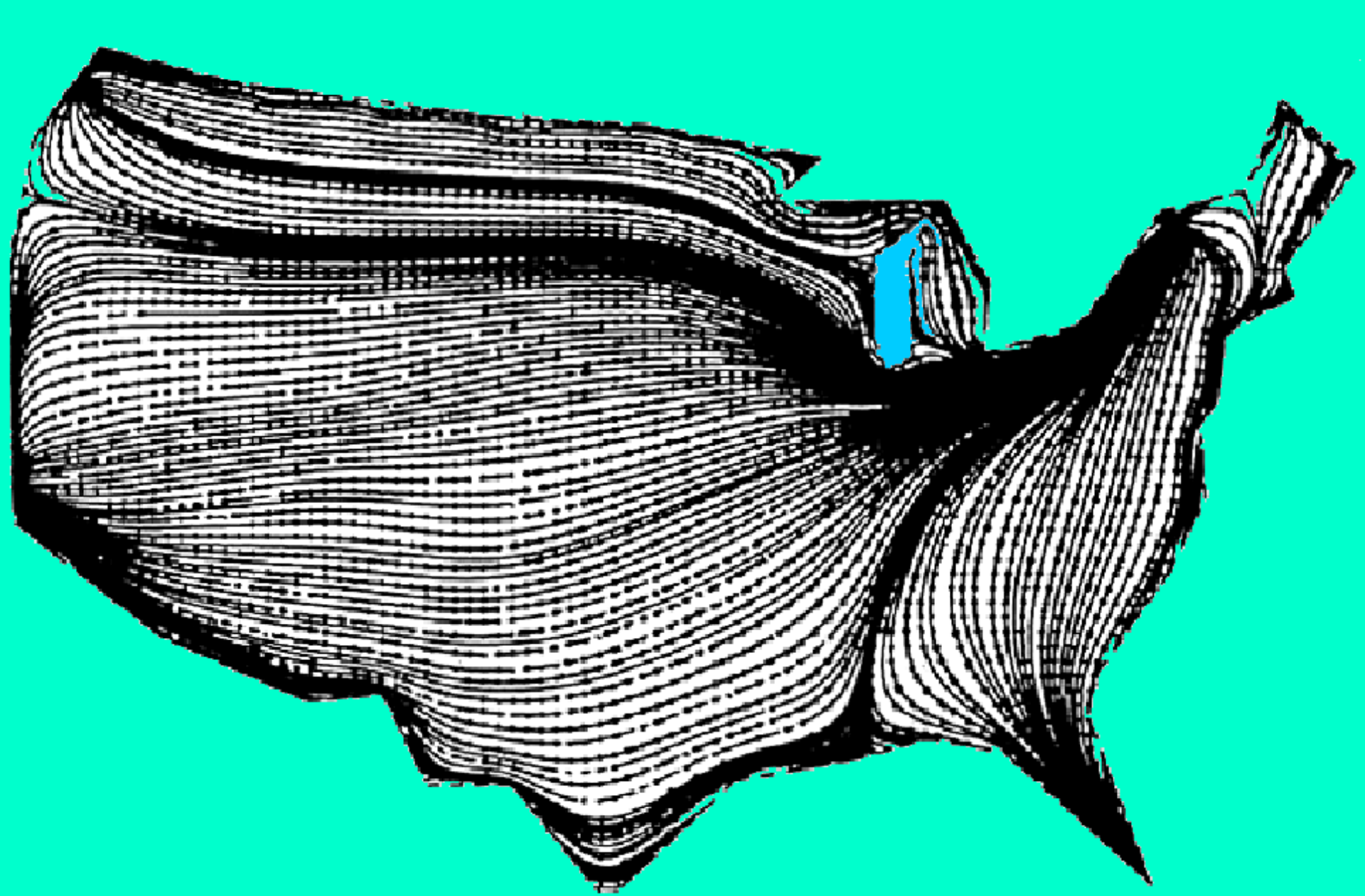
Another view

# The migration potentials shown as contours and with gradient vectors connected to give streaklines



# 16 Million People Migrating

An ensemble average. Note the distinct migration domains.



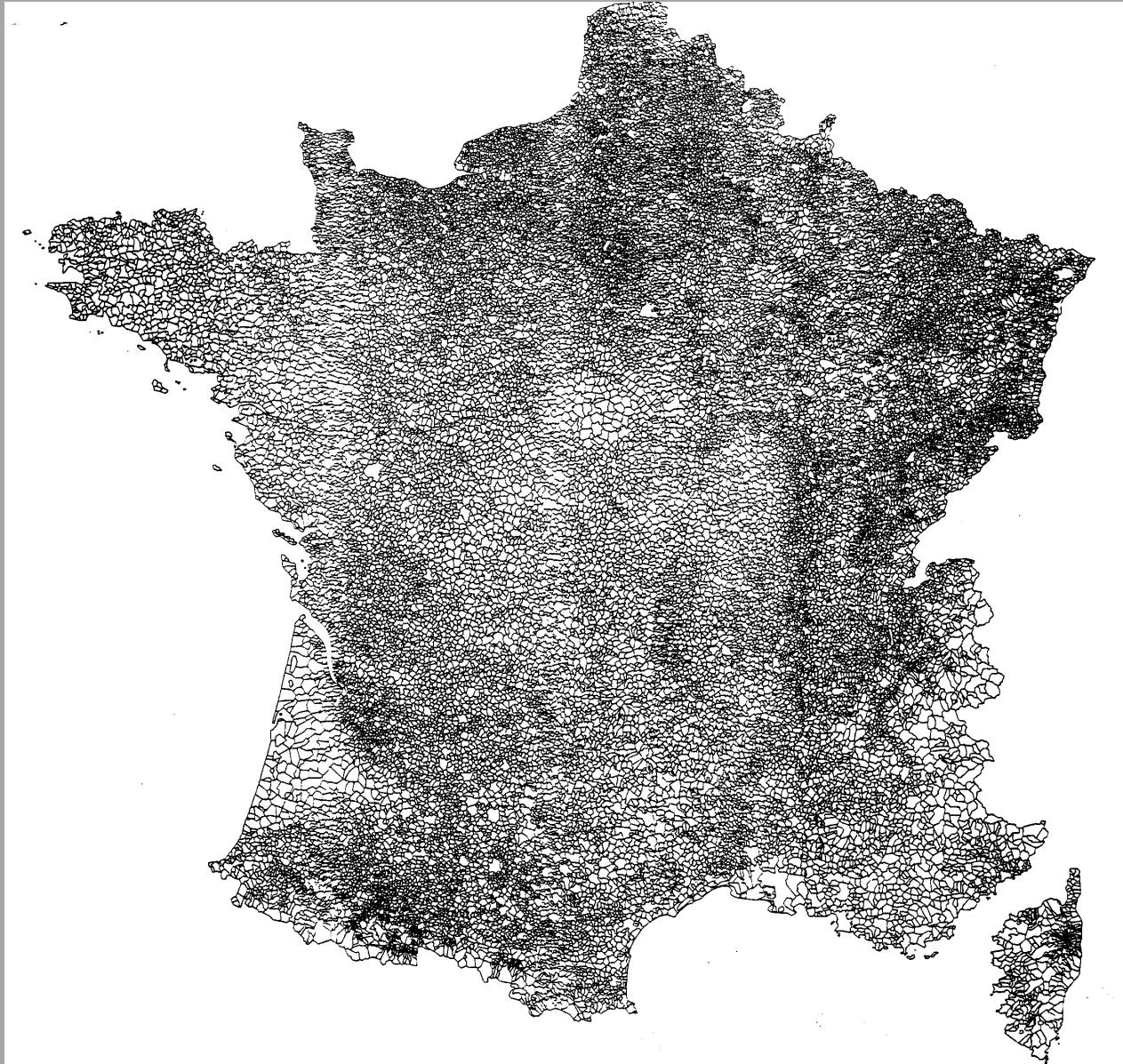
That these migration maps resemble maps of wind or ocean currents is not surprising given that we in fact speak of migration flows and backwaters, and use many such hydrodynamic terms when discussing migration and movement phenomena.

The foregoing equations have captured some of this effect in a realistic manner.

One advantage of the continuous potential model is in the clarity that it provides of the overall pattern and domains.



# France's 36,545 Communes



# Think Big! Think High Resolution!

The 36,545 communes of France could yield a migration or interaction table with as many as 1,335,537,025 entries. (3 km average resolution)

My assertion is:

Looking at a conventional flow map in this amount of detail would not be useful, but a vector field could show divergences, convergences, and reveal interesting domain patterns. And the potential surface would yield further insight.

Thank You For Your Attention

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