Optimal Supply Networks III: Redistribution

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Principles of Complex Systems, Vols. 1, 2, & 3D
CSYS/MATH 6701, 6713, & a pretend number, 2023–2024| @pocsvox

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Computational Story Lab | Vermont Complex Systems Center
Santa Fe Institute | University of Vermont

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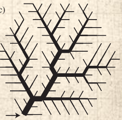

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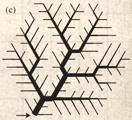

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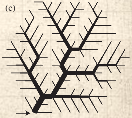

Outline

Distributed Sources

- Size-density law
- Cartograms
- A reasonable derivation
- Global redistribution
- Public versus Private

References

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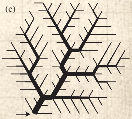
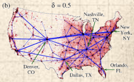


Many sources, many sinks

How do we distribute sources?



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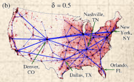

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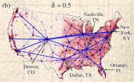

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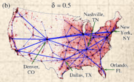

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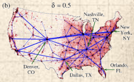
- Obvious: if density is uniform then sources are best distributed uniformly.



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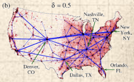

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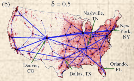
- Which lattice is optimal? The hexagonal lattice
- Q2: Given population density is uneven, what do we do?



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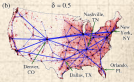
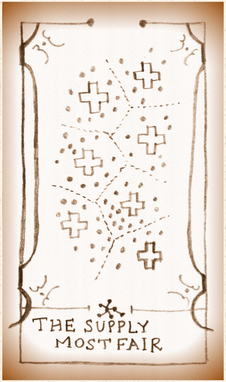
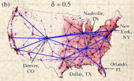
- Obvious: if density is uniform then sources are best distributed uniformly.
- Which lattice is optimal? The hexagonal lattice
- Q2: Given population density is uneven, what do we do?
- We'll follow work by Stephan (1977, 1984) [4, 5], Gastner and Newman (2006) [2], Um *et al.* (2009) [6], and work cited by them.



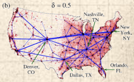

THE JOHN DORY
THE SUN
THE NETWORKS COMPLEX
"I am the Monarch of the Sea."
THE ARK OF THE RANDOM NETWORK
THE NETWORKS FREE OF SCALE
THE SMALLNESS OF WORLDS
THE THEORY OF SIX DEGREES
THE LANDSCAPES OF FORKING PATHS
THE NETWORKS OF BLOOD
THE TREES OF REALITY
THE ORDERS OF STREAMS
THE LAWS OF BRANCHING
THE UNKNOWN MECHANISM
THE BRAID OR THE BRANCH
THE LAW OF OPTIMAL FORKS
THE CHURCH OF QUARTEROLOGY
THE TRUTHICIDE
THE FAMILY OF PANCAKE COWS
THE FAMILY OF SPHERICAL COWS
THE IMPROBABLE PACHYDERM
THE ILL-SCALING OF ALLOMETRIC BODIES


THE SUPPLY MOST FAIR


(b)
δ = 0.5
Nashville, TN
New York, NY
Denver, CO
Dallas, TX
Orlando, FL

Optimal source allocation: Size-density law

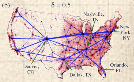
Solidifying the basic problem



Optimal source allocation: Size-density law

Solidifying the basic problem

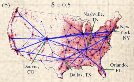
- Given a region with some population distribution ρ, most likely uneven.



Optimal source allocation: Size-density law

Solidifying the basic problem

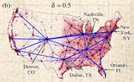
- Given a region with some population distribution ρ, most likely uneven.
- Given resources to build and maintain N facilities.



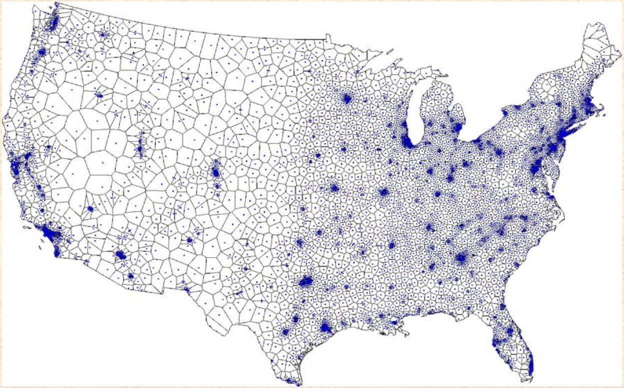
Optimal source allocation: Size-density law

Solidifying the basic problem

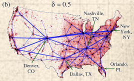
- Given a region with some population distribution ρ, most likely uneven.
- Given resources to build and maintain N facilities.
- Q: How do we locate these N facilities so as to minimize the average distance between an individual's residence and the nearest facility?



"Optimal design of spatial distribution networks"
Gastner and Newman,
Phys. Rev. E, **74**, 016117, 2006. [2]



- Approximately optimal location of 5000 facilities.
- Based on 2000 Census data.
- Simulated annealing + Voronoi tessellation.



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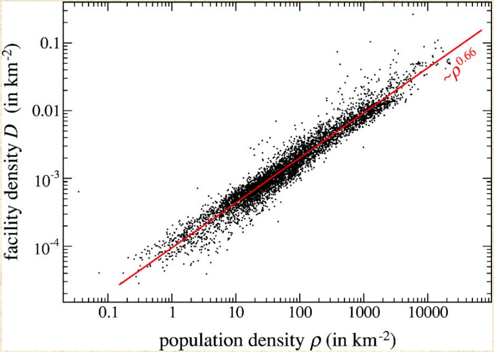
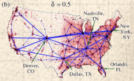


- Optimal facility density ρ_{fac} vs. population density ρ_{pop}.



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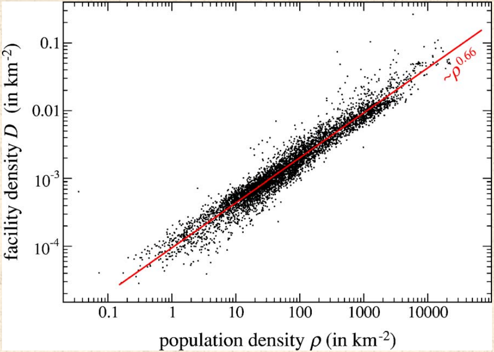

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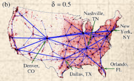
- Fit is $\rho_{\text{fac}} \propto \rho_{\text{pop}}^{0.66}$ with $r^2 = 0.94$.



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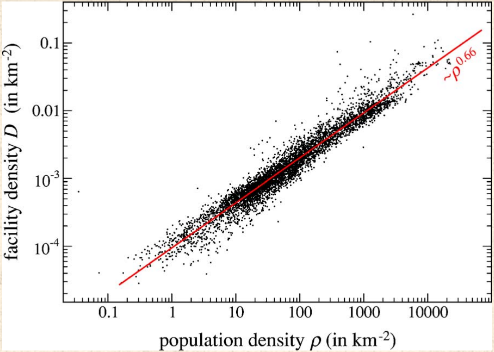

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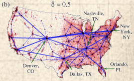
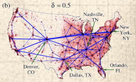
- Looking good for a 2/3 power ...



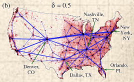
Outline



Optimal source allocation

Size-density law:

$$\rho_{\text{fac}} \propto \rho_{\text{pop}}^{2/3}$$



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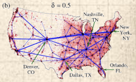
- Why?



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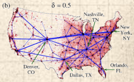
- Again: Different story to branching networks where there was either one source or one sink.



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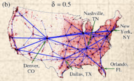
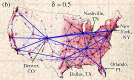
- Now sources & sinks are distributed throughout region.



Optimal source allocation

"Territorial division: The least-time constraint behind the formation of subnational boundaries"
G. Edward Stephan,
Science, **196**, 523–524, 1977. [4]

- We first examine Stephan's treatment (1977) [4, 5]



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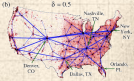

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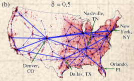
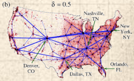
- Zipf-like approach: invokes principle of minimal effort.
- Also known as the Homer Simpson principle.



Optimal source allocation

- Consider a region of area A and population P with a single functional center that everyone needs to access every day.



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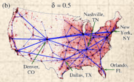

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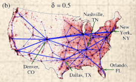

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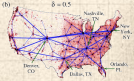

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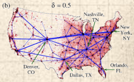
- Assume isometry: average travel distance $\langle d \rangle$ will be on the length scale of the region which is $\sim A^{1/2}$
- Average time expended per person in accessing facility is therefore

$$\langle d \rangle / \langle v \rangle = cA^{1/2} / \langle v \rangle$$

where c is an unimportant shape factor.



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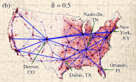

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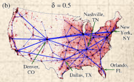

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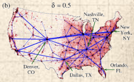
- If burden of mainenance is shared then average cost per person is τ/P where P = population.



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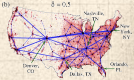
- Replace P by $\rho_{\text{pop}}A$ where ρ_{pop} is density.



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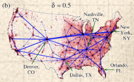

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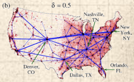
- Total average time cost per person:

$$T = \langle d \rangle / \langle v \rangle + \tau/(\rho_{\text{pop}}A)$$



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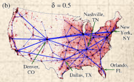

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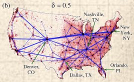
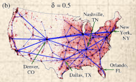
- Now Minimize with respect to A ...



Optimal source allocation

- Differentiating ...

$$\frac{\partial T}{\partial A} = \frac{\partial}{\partial A}\left(cA^{1/2}/\left\langle v \right\rangle + \tau/(\rho_{\mathsf{pop}}A)\right)$$



Optimal source allocation

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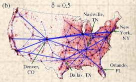

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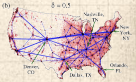

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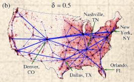

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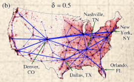

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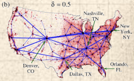

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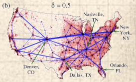
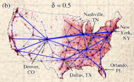


Optimal source allocation

An issue:

- Maintenance (τ) is assumed to be independent of population and area (P and A)



Optimal source allocation

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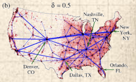
- Stephan's online book "The Division of Territory in Society" is here.
- (It used to be here.)
- The Readme is well worth reading (1995).



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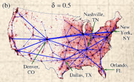
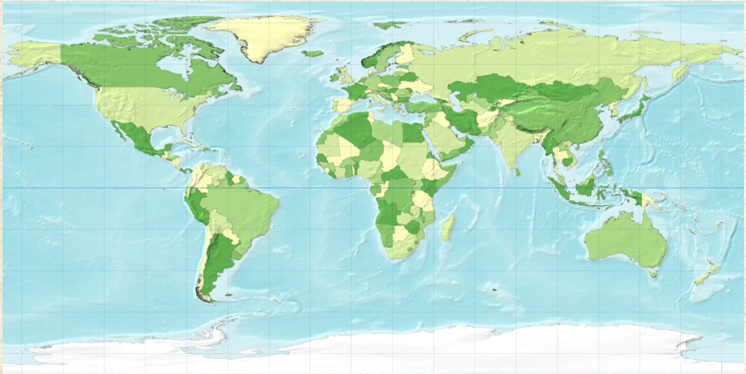
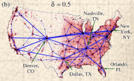


Cartograms

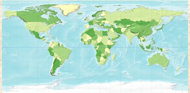
Standard world map:

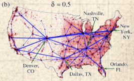




Cartograms

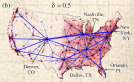
Cartogram of countries ‘rescaled’ by population:





Cartograms

Diffusion-based cartograms:



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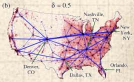

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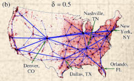

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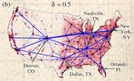

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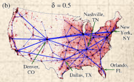

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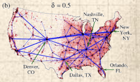
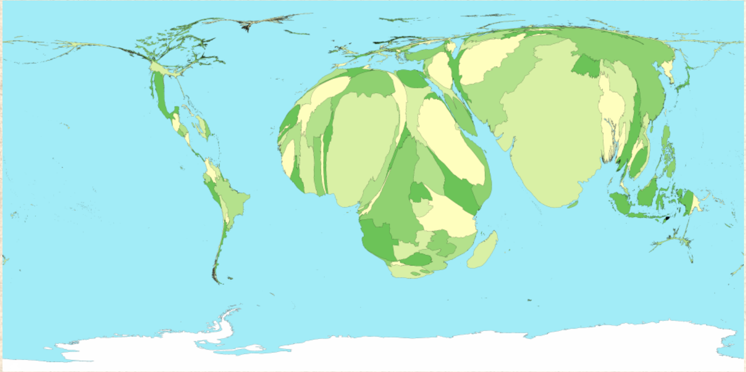
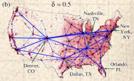


Cartograms

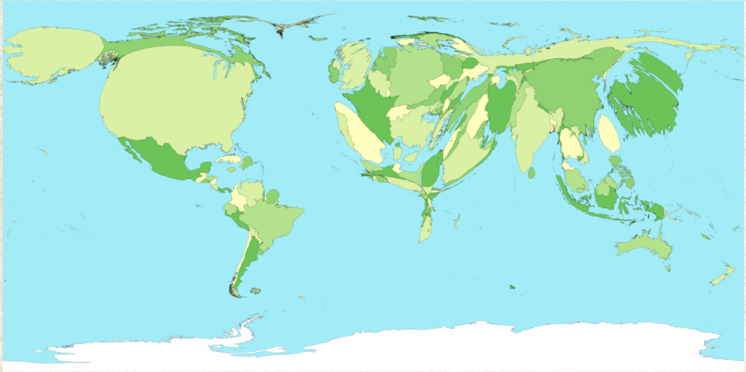
Child mortality:

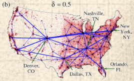




Cartograms

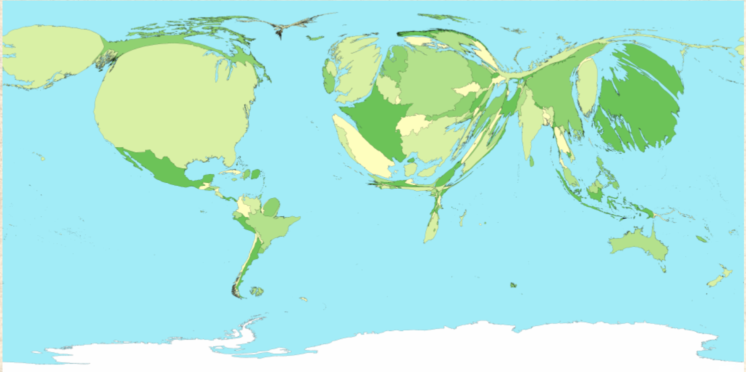
Energy consumption:

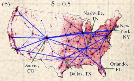




Cartograms

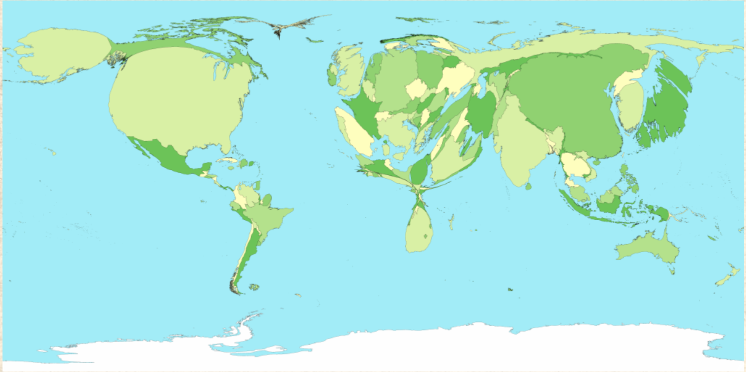
Gross domestic product:

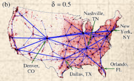




Cartograms

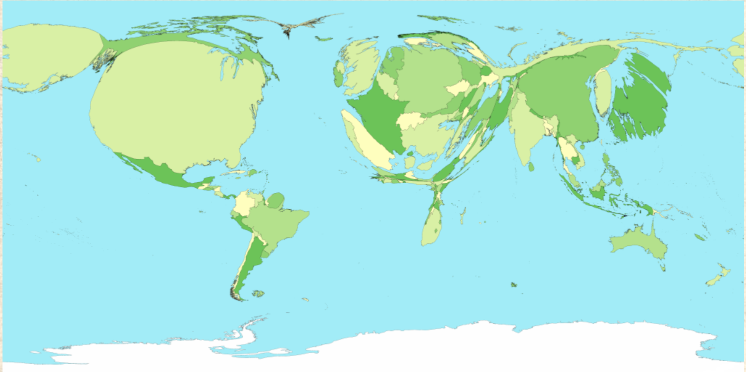
Greenhouse gas emissions:

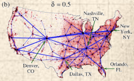




Cartograms

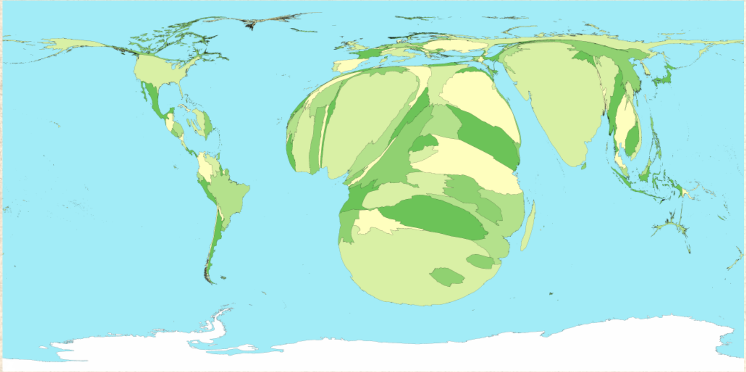
Spending on healthcare:

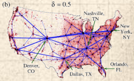




Cartograms

People living with HIV:

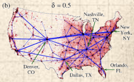




Cartograms

- The preceding sampling of Gastner & Newman's cartograms lives here.
- A larger collection can be found at worldmapper.org.

WORLDMAPPER The world as you've never seen it before



"Optimal design of spatial distribution networks"

Gastner and Newman,
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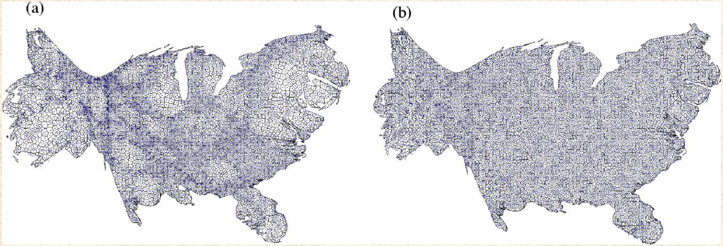
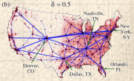


- Left: population density-equalized cartogram.



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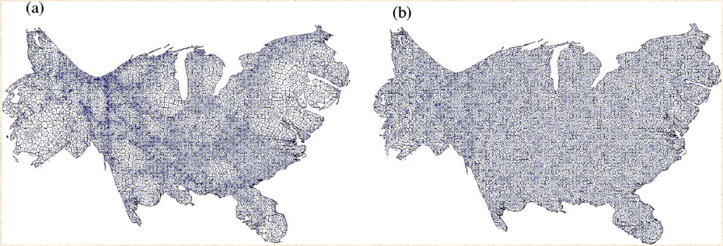
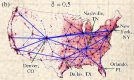


- Left: population density-equalized cartogram.
- Right: (population density)$^{2/3}$-equalized cartogram.



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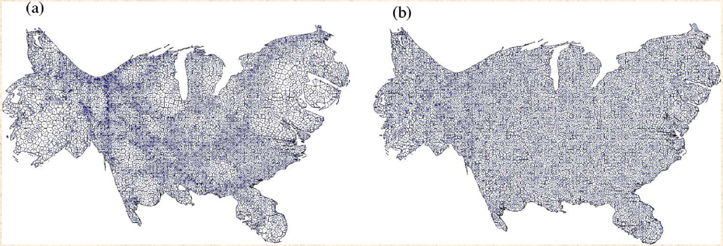

- Left: population density-equalized cartogram.
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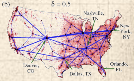
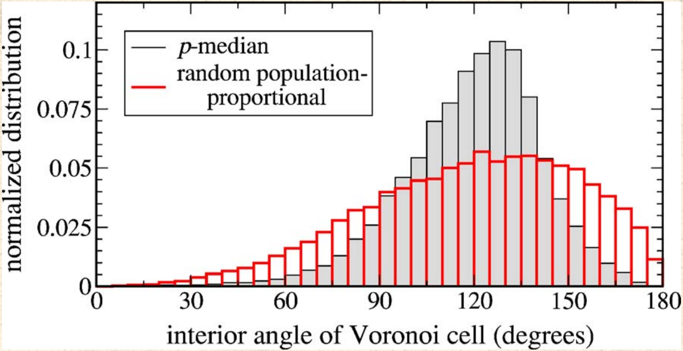
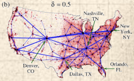
- Facility density is uniform for $\rho_{\text{pop}}^{2/3}$ cartogram.





From Gastner and Newman (2006) [2]

- Cartogram's Voronoi cells are somewhat hexagonal.



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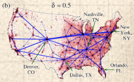
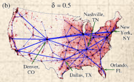


Deriving the optimal source distribution:



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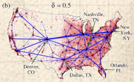

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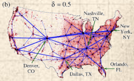

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- Formally, we want to find the locations of n sources $\{\vec{x}_1, \ldots, \vec{x}_n\}$ that minimizes the cost function

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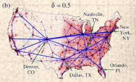

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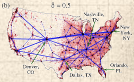
- Also known as the p-median problem, and connected to cluster analysis.



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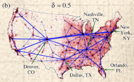

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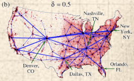
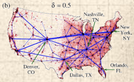


Size-density law

Approximations:

- For a given set of source placements $\{\vec{x}_1, \ldots, \vec{x}_n\}$, the region Ω is divided up into Voronoi cells, one per source.



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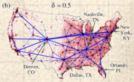

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- For a given set of source placements $\{\vec{x}_1, \ldots, \vec{x}_n\}$, the region Ω is divided up into Voronoi cells, one per source.
- Define $A(\vec{x})$ as the area of the Voronoi cell containing $\vec{x}$.
- As per Stephan's calculation, estimate typical distance from $\vec{x}$ to the nearest source (say i) as

$$c_i A(\vec{x})^{1/2}$$

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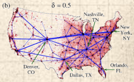

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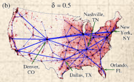
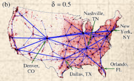


Size-density law

Carrying on:

- The cost function is now

$$F = c \int_{\Omega} \rho_{\text{pop}}(\vec{x}) A(\vec{x})^{1/2} \mathrm{d}\vec{x} \, .$$



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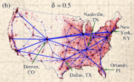

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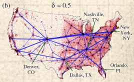

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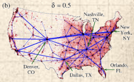

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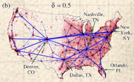

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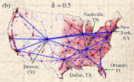
- By varying $\{\vec{x}_1, \ldots, \vec{x}_n\}$, minimize

$$G(A) = c\int_{\Omega} \rho_{\textrm{pop}}(\vec{x})A(\vec{x})^{1/2}\textrm{d}\vec{x} - \lambda\left(n - \int_{\Omega}\left[A(\vec{x})\right]^{-1}\textrm{d}\vec{x}\right)$$



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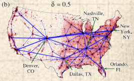

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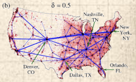

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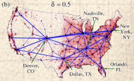

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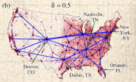

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Now a Lagrange multiplier story:

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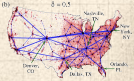

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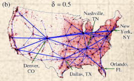

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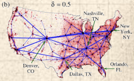

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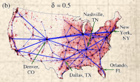
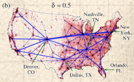
- Normalizing (or solving for λ):

$$\rho_{\text{fac}}(\vec{x}) = n \frac{[\rho_{\text{pop}}(\vec{x})]^{2/3}}{\int_{\Omega}[\rho_{\text{pop}}(\vec{x})]^{2/3}\mathrm{d}\vec{x}} \propto [\rho_{\text{pop}}(\vec{x})]^{2/3}.$$



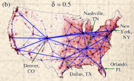
Outline



Global redistribution networks

One more thing:

- How do we supply these facilities?



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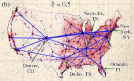

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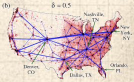

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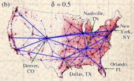

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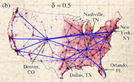

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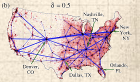

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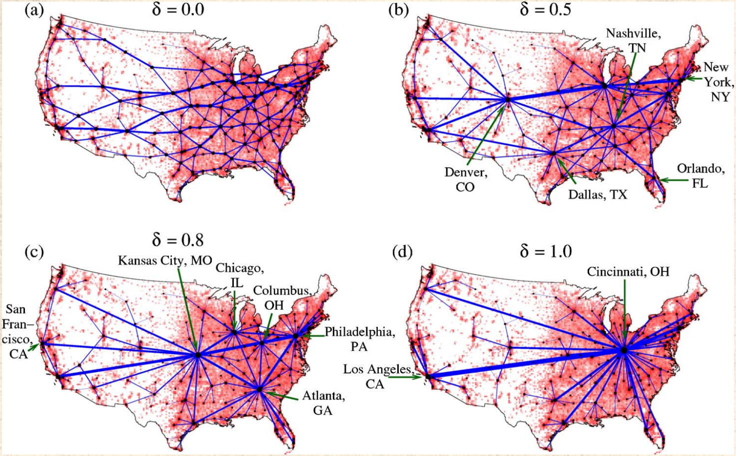
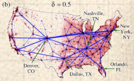
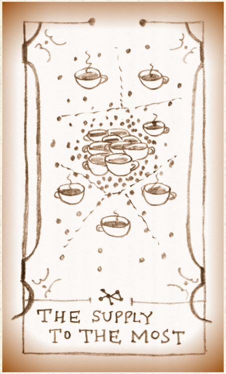


From Gastner and Newman (2006) [2]




THE JOHN DORY
THE SUN
THE NETWORKS COMPLEX
"I am the Monarch of the Sea."
THE ARK OF THE RANDOM NETWORK
THE NETWORKS FREE OF SCALE
THE SMALLNESS OF WORLDS
THE THEORY OF SIX DEGREES
THE LANDSCAPES OF FORKING PATHS
THE NETWORKS OF BLOOD
THE TREES OF REALITY
THE ORDERS OF STREAMS
THE LAWS OF BRANCHING
THE UNKNOWN MECHANISM
THE BRAID OR THE BRANCH
THE LAW OF OPTIMAL FORKS
THE CHURCH OF QUARTEROLOGY
D-1
D
THE TRUTHICIDE
THE FAMILY OF PANCAKE COWS
THE FAMILY OF SPHERICAL COWS
THE IMPROBABLE PACHYDERM
THE ILL-SCALING OF ALLOMETRIC BODIES
THE SUPPLY MOST FAIR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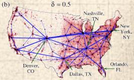
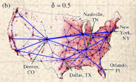

(b)
δ = 0.5
Nashville, TN
New York, NY
Denver, CO
Dallas, TX
Orlando, FL

Outline



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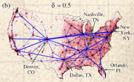
Public versus private facilities

Beyond minimizing distances:



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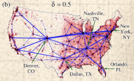

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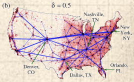

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- Two idealized limiting classes:
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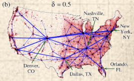

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- Two idealized limiting classes:
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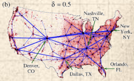
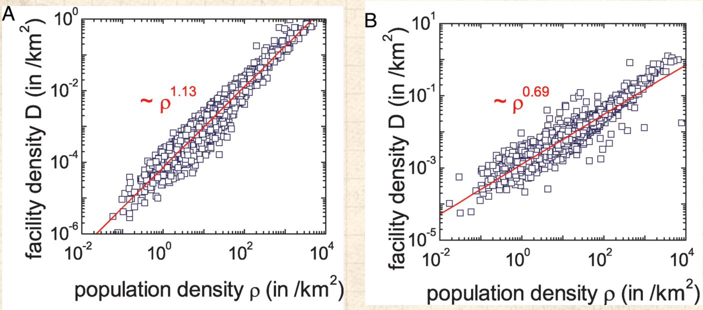
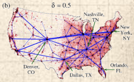
  2. Pro-social, public facilities: $\alpha = 2/3$.
- Um *et al.* investigate facility locations in the United States and South Korea.



Public versus private facilities: evidence



- Left plot: ambulatory hospitals in the U.S.
- Right plot: public schools in the U.S.



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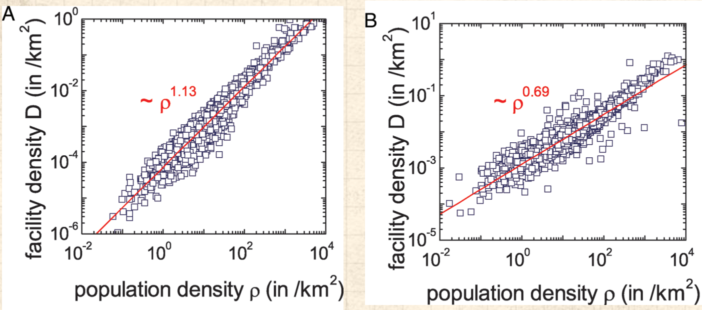
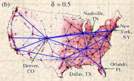


- Left plot: ambulatory hospitals in the U.S.
- Right plot: public schools in the U.S.
- Note: break in scaling for public schools. Transition from $\alpha \simeq 2/3$ to $\alpha = 1$ around $\rho_{\text{pop}} \simeq 100$.



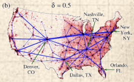
Public versus private facilities: evidence

US facility	α (SE)	R^2
Ambulatory hospital	1.13(1)	0.93
Beauty care	1.08(1)	0.86
Laundry	1.05(1)	0.90
Automotive repair	0.99(1)	0.92
Private school	0.95(1)	0.82
Restaurant	0.93(1)	0.89
Accommodation	0.89(1)	0.70
Bank	0.88(1)	0.89
Gas station	0.86(1)	0.94
Death care	0.79(1)	0.80
* Fire station	0.78(3)	0.93
* Police station	0.71(6)	0.75
Public school	0.69(1)	0.87

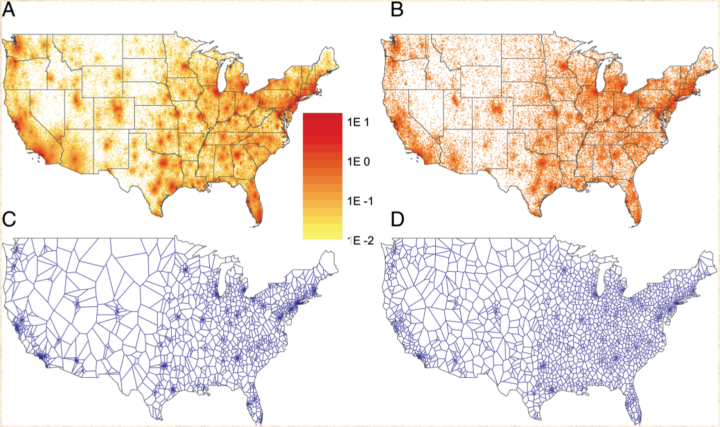
SK facility	α (SE)	R^2
Bank	1.18(2)	0.96
Parking place	1.13(2)	0.91
* Primary clinic	1.09(2)	1.00
* Hospital	0.96(5)	0.97
* University/college	0.93(9)	0.89
Market place	0.87(2)	0.90
* Secondary school	0.77(3)	0.98
* Primary school	0.77(3)	0.97
Social welfare org.	0.75(2)	0.84
* Police station	0.71(5)	0.94
Government office	0.70(1)	0.93
* Fire station	0.60(4)	0.93
* Public health center	0.09(5)	0.19

Rough transition between public and private at $\alpha \simeq 0.8$.

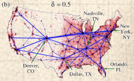
Note: * indicates analysis is at state/province level; otherwise county level.



Public versus private facilities: evidence



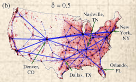
A, C: ambulatory hospitals in the U.S.; B, D: public schools in the U.S.; A, B: data; C, D: Voronoi diagram from model simulation.



Public versus private facilities: the story

So what's going on?

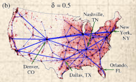
- Social institutions seek to minimize distance of travel.



Public versus private facilities: the story

So what's going on?

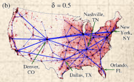
- Social institutions seek to minimize distance of travel.
- Commercial institutions seek to maximize the number of visitors.



Public versus private facilities: the story

So what's going on?

- Social institutions seek to minimize distance of travel.
- Commercial institutions seek to maximize the number of visitors.
- Defns: For the ith facility and its Voronoi cell V_i, define
  - n_i = population of the ith cell;
  - $\langle r_i \rangle$ = the average travel distance to the ith facility.
  - A_i = area of ith cell (s_i in Um *et al.* [6])



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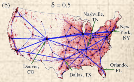
$$v_i = n_i \langle r_i \rangle^{\beta} \text{ with } 0 \leq \beta \leq 1.$$



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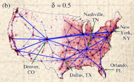
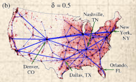
- Limits:
  - $\beta = 0$: purely commercial.
  - $\beta = 1$: purely social.



Public versus private facilities: the story

- Either proceeding as per the Gastner-Newman-Gusein-Zade calculation or, as Um *et al.* do, observing that the cost for each cell should be the same, we have:

$$\rho_{\mathrm{fac}}(\vec{x}) = n \frac{[\rho_{\mathrm{pop}}(\vec{x})]^{2/(\beta+2)}}{\int_{\Omega} [\rho_{\mathrm{pop}}(\vec{x})]^{2/(\beta+2)} \mathrm{d}\vec{x}} \propto [\rho_{\mathrm{pop}}(\vec{x})]^{2/(\beta+2)}.$$



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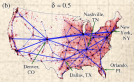
- For $\beta = 0$, $\alpha = 1$: commercial scaling is linear.



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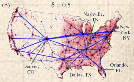
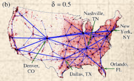
- For $\beta = 0$, $\alpha = 1$: commercial scaling is linear.
- For $\beta = 1$, $\alpha = 2/3$: social scaling is sublinear.



System type:	Dominant cost/benefit scaling:	Dominant constraint scaling:	Scaling of number of events per partition:	Density scaling:	Quantity equalized across partitions:
General form	$\rho_{\text{event}} V^{\alpha}$ $0 < \alpha \leq 1$	$V^{-\beta}$ $1 - \alpha \leq \beta \leq 1$	$N \propto V^{1-\alpha-\beta}$	$\rho_{\text{partition}} \propto \rho_{\text{event}}^{1/(\alpha+\beta)}$	$NV^{\alpha+\beta-1}$
I. Event rate equalizing with partition number constrained (for-profit)	$\sim \rho_{\text{event}} \ln V$	V^{-1}	$N \propto V^{0}$	$\rho_{\text{partition}} \propto \rho_{\text{event}}^{1}$	N
II. Minimizing average event access time with partition number constrained (p-median problem, pro-social)	$\rho_{\text{event}} V^{1/d}$	V^{-1}	$N \propto V^{-1/d}$	$\rho_{\text{partition}} \propto \rho_{\text{event}}^{d/(d+1)}$	$NV^{1/d}$
III. System under stochastic threat with partition boundary constrained (HOT model)	$\rho_{\text{event}} V^{1}$	$V^{-1/d}$	$N \propto V^{-1/d}$	$\rho_{\text{partition}} \propto \rho_{\text{event}}^{d/(d+1)}$	$NV^{1/d}$
IV. System under stochastic threat with partition number constrained	$\rho_{\text{event}} V^{1}$	V^{-1}	$N \propto V^{-1}$	$\rho_{\text{partition}} \propto \rho_{\text{event}}^{1/2}$	NV



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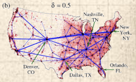

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