Cities as Complex Systems II

Individual and Collective Agents, Dynamics, Information

CSSS, Santa Fe, June 22, 2023

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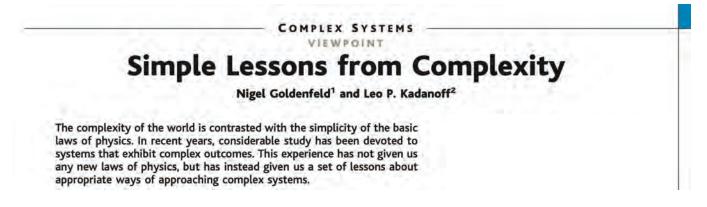
Mansueto Institute for Urban Innovation

The next few years are likely to lead to an increasing study of complexity in the context of statistical dynamics, with a view to better understanding physical, economic, social, and especially biological systems.

It will be an exciting time.

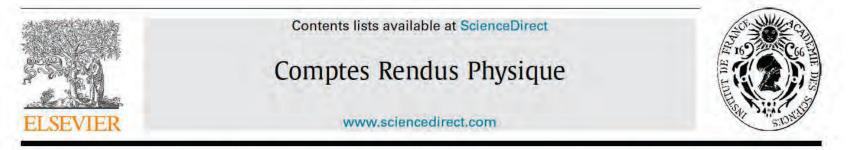
As science turns to complexity, one must realize that complexity demands attitudes quite different from those heretofore common in physics.

Goldenfeld & Kadanoff, Science, 1999



https://www.science.org/doi/10.1126/science.284.5411.87





From statistical physics to social sciences / De la physique statistique aux sciences sociales

Towards a statistical mechanics of cities

Vers une mécanique statistique des villes

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https://www.sciencedirect.com/science/article/pii/S1631070519300350

Canonical modeling in physics

$$\dot{x} = \frac{p}{m}, \quad \dot{p} = -\frac{dV(x)}{dx} \qquad \dot{x} \equiv \frac{dx}{dt}$$
position
energy
$$E(x, p) = \frac{p^2}{2m} + V(x)$$
Credit : wikipedia

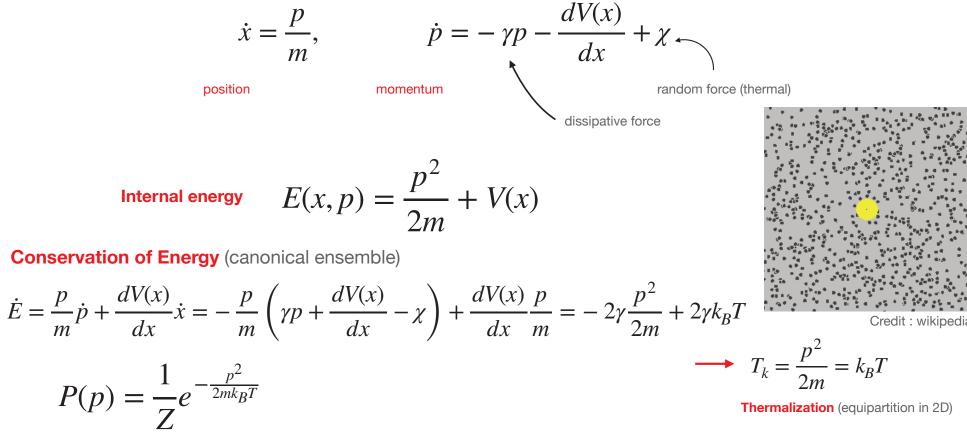
Conservation of Energy (micro-canonical ensemble)

Example: $V(x) = gx^2$

Harmonic oscillator

$$\dot{E} = \frac{p}{m}\dot{p} + \frac{dV(x)}{dx}\dot{x} = -\frac{p}{m}\frac{dV(x)}{dx} + \frac{dV(x)}{dx}\frac{p}{m} = 0.$$

Canonical modeling in physics, in contact with heat reservoir



Maxwell-Boltzmann distribution

Particles in Physics are identical

They are passive: lack 'behavior'

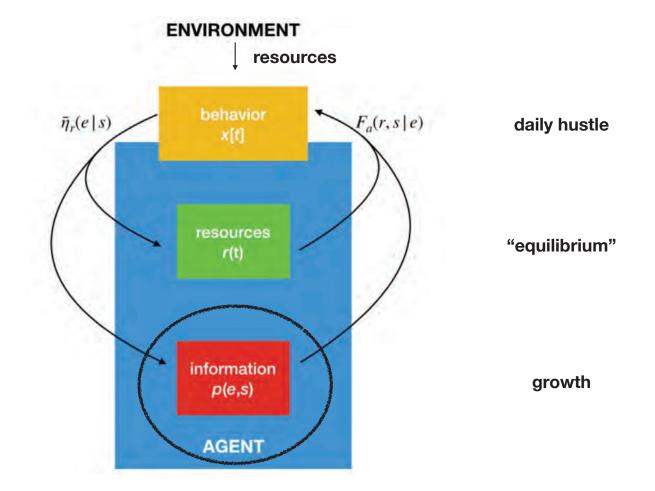
They are purposeless: lack intention

The situation is completely different for agents in complex systems

This requires ability to exert forces that underlie behavior, and make choices between courses of action

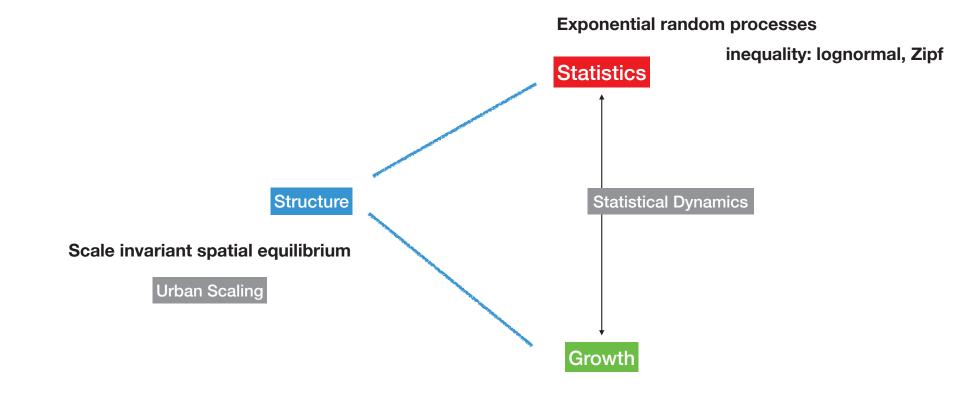
This requires reservoirs of both energy and information

The general agent in complex systems



Strategic active agents manage resources purposefully in stochastic environments

This has a macroscopic expression for complex systems, such as cities



energy and information management

learning, social dynamics

1. Behavior

$$\dot{x} = p/m,$$
 $\dot{p} = -\gamma p - \frac{dV(x)}{dx} + \chi + F_a$
 $[\gamma] = 1/t$ self-propelled force, selecting position and velocity

Conservation of energy:

$$\Delta E = -2(\gamma \,\Delta t) \langle \frac{p^2}{2m} \rangle + 2(\gamma \,\Delta t) k_{\rm B} T + \oint F_{\rm a} \cdot dx$$

$$\langle \frac{p^2}{2m} \rangle = k_{\rm B}T + \frac{1}{2(\gamma \Delta t)} W_{\rm a}$$

Work done by the self-propelling force over a cycle

$$W_{a} = \oint F_{a} \cdot dx$$

-

$$P(p_1, p_2) = \frac{1}{Z} e^{-\frac{p_1^2 + p_2^2}{2m\bar{w}_a}}$$

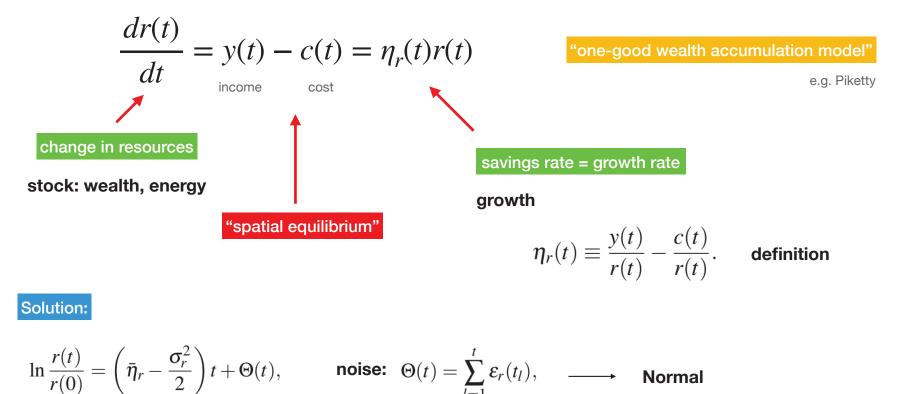
The temperature is the work per unit of dissipation depends on statistics of \bar{w}_a

$$\bar{E} = \langle \frac{p^2}{2m} \rangle \simeq \frac{1}{2(\gamma \Delta t)} W_a \equiv \bar{w}_a$$

Observed energy is now dominated by behavioral forces

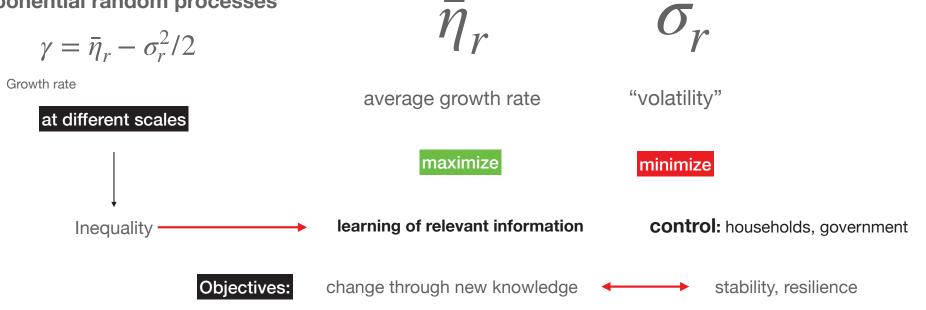
2. Energy dynamics

A Model of Resource Dynamics

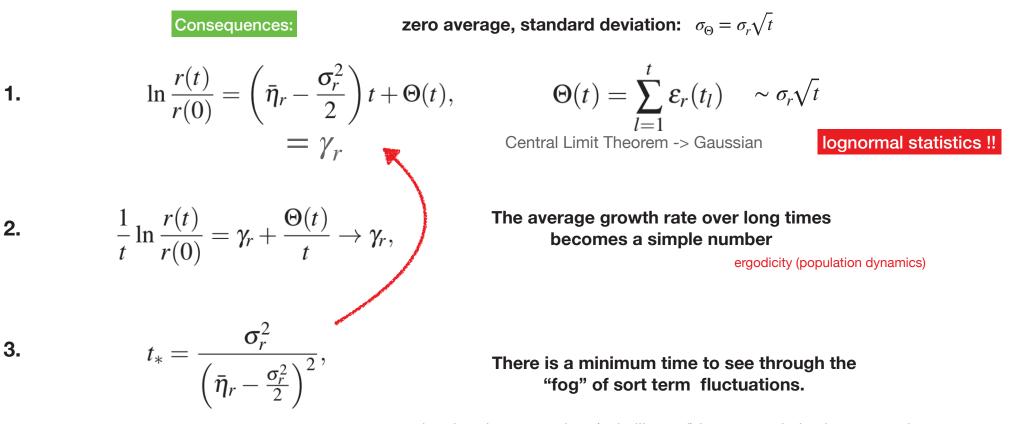


Two independent parameters:

Exponential random processes

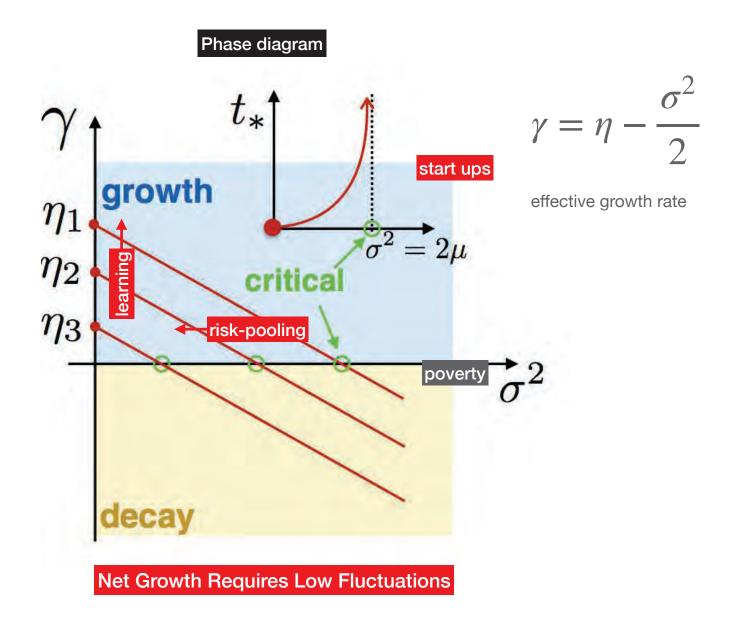


The counter intuitive consequences of stochastic geometric growth



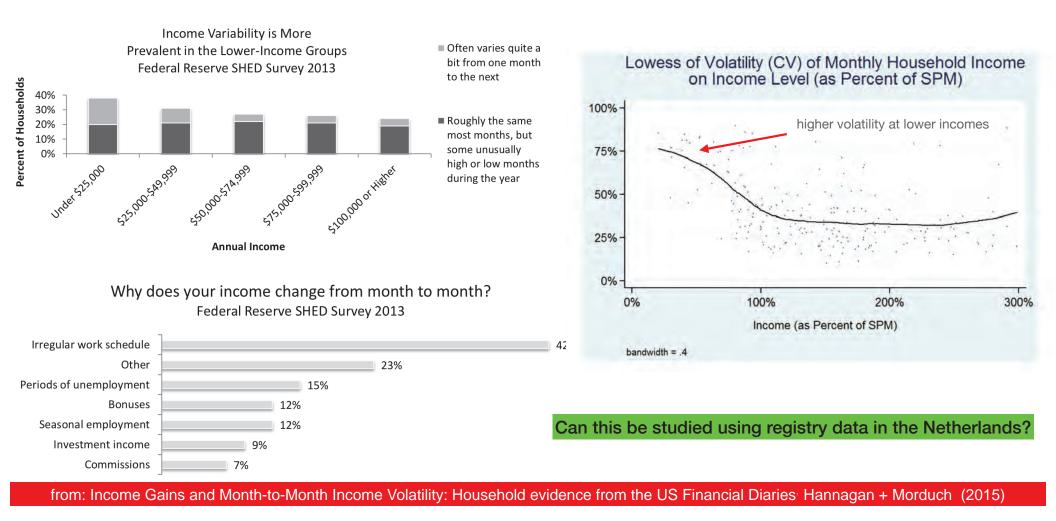
proportional to the uncertainty (volatility, std) between relative income and costs

people with a lot of uncertainty do not see growth as a possibility



Poverty is a low resources, high volatility state

lower effective growth rates, unclear horizons



How to lower volatility? How to derive lognormal and power law statistics?

Advances

SCIENCE ADVANCES | RESEARCH ARTICLE

SOCIAL SCIENCES

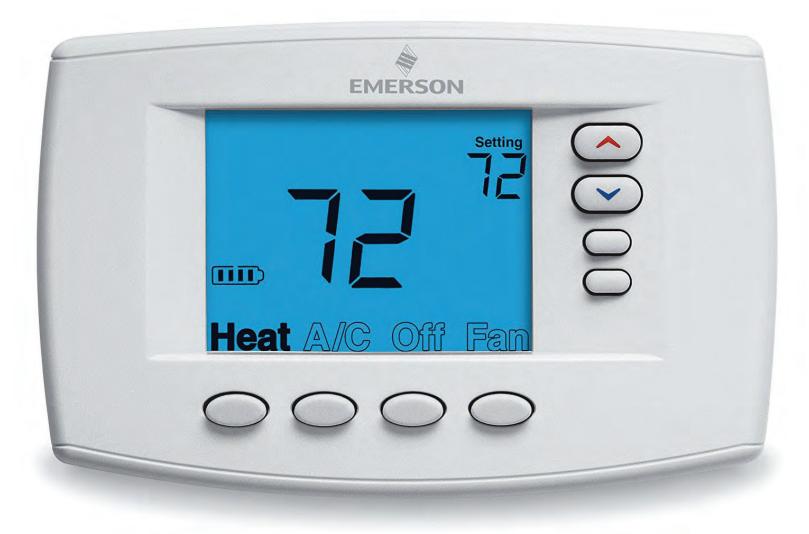
Urban growth and the emergent statistics of cities

Luis M. A. Bettencourt^{1,2}

Urban theory models cities as spatial equilibria to derive their aggregate properties as functions of extensive variables, such as population size. However, this assumption seems at odds with cities' most interesting properties as engines of fast and variable processes of growth and change. Here, we build a general statistical dynamics of cities across scales, from single agents to entire urban systems. We include agents' strategic behavior to produce predictable growth rates, which requires balancing relative incomes and costs over time. We implement these dynamics using stochastic differential equations and control theory to demonstrate a number of general emergent properties of cities deriving from limit theorems applied to growth rates. This framework establishes necessary conditions for scaling to be conserved by urban dynamics and shows how exponent corrections can be calculated. These ideas are tested using stochastic simulations and a long timeseries for 382 US Metropolitan Areas over nearly five decades.

https://www.science.org/doi/epdf/10.1126/sciadv.aat8812

Feedback Control is a big part of your life







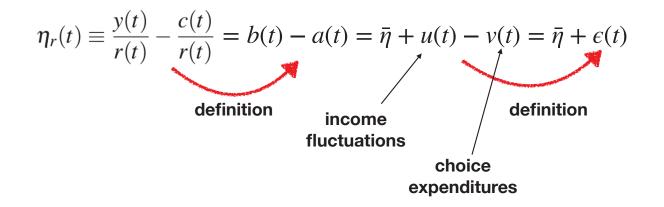




Balancing humans and robots

Self-driving cars

Want low fluctuations:



Consumption Smoothing

when I have more money can spend more and vice-versa

need to average (smoothen) expenditures and incomes over some time month, season, year How Control theory works:

PID Controller

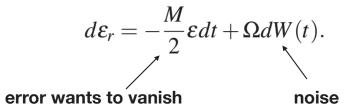
$$\overline{u(t)} = k_P \varepsilon_r(t) + k_I \int_0^t \varepsilon_r(t') dt' + k_D \frac{d\varepsilon_r}{dt},$$

Time costs to track income fluctuations

This gives a simple dynamics for errors

$$\frac{d^2 \varepsilon_r}{dt^2} + 2\zeta \omega \frac{d\varepsilon_r}{dt} + \omega^2 \varepsilon_r = F/m, \qquad m = k_D, \ \omega = \sqrt{k_I/k_D} \text{ and } \zeta = \frac{k_P + 1}{2\sqrt{k_D k_I}},$$

noisy spring



 $M = \omega/\zeta = k_I/[2(k_P+1)]$

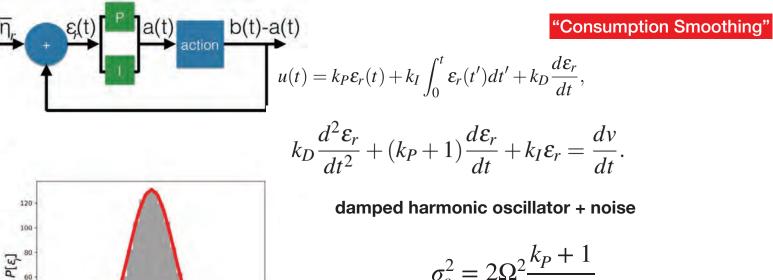
Statistics of error is "under control":

$$P[\varepsilon_r,t|\varepsilon_0] = \sqrt{\frac{M}{2\pi\Omega^2(1-e^{-Mt})}} e^{-\frac{M}{2\Omega^2} \left[\frac{(\varepsilon_r - \varepsilon_0 e^{-M/2t})^2}{1-e^{-Mt}}\right]} \to \sqrt{\frac{M}{2\pi\Omega^2}} e^{\frac{M}{2\Omega^2}}r^2,$$

quality of control

Why are relative incomes and costs correlated?

Control Scheme for Savings and Growth



$$\sigma_{\epsilon}^2 = 2\Omega^2 \frac{\kappa_P + 1}{k_I}$$

controlled by time-averaging of error

Er growth rates become Gaussian &

-0.005

0.000

0.005

-0.010

0.010

0.015

60

40

20

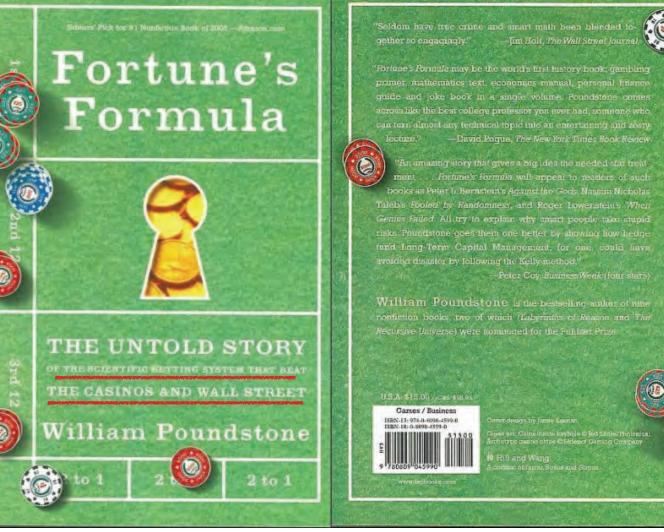
-0.015

growth is geometric and random

At larger (population) scales, there are many more mechanisms:

- insurance
- pension funds
- government social services
- corporate 'bail-outs'

How to increase growth rates?

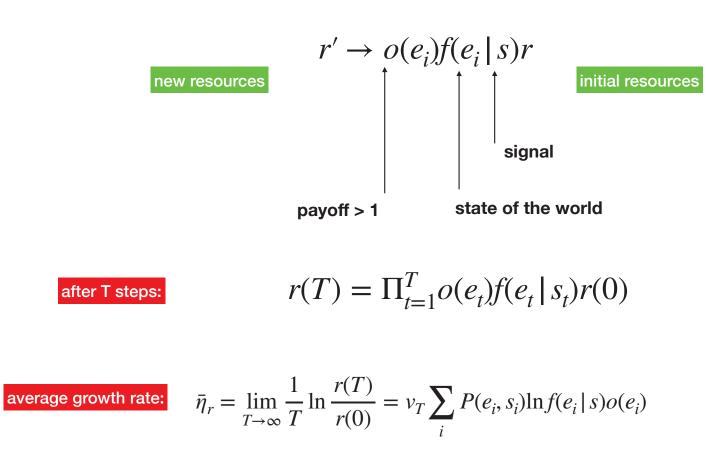


http://www.fortunesformula.com/

Investment and information

3. Information

Inter temporal optimization: sequential investment in stochastic environmental outcome



Maximal Growth Rate:

Simple intertemporal optimization without discounting

$$\bar{\eta}_r = \lim_{T \to \infty} \frac{1}{T} \ln \frac{r(T)}{r(0)} = v_T \sum_{e,s} P(e,s) \ln f(e \,|\, s) o(e)$$

best allocation:
$$f(e \mid s) \rightarrow P(e \mid s)$$

fair odds:
$$o(e) \rightarrow 1/P(e)$$

(worst case = "efficient markets")

$$\bar{\eta}_r^* = v_T \sum_{e,s} P(e,s) \ln \frac{P(e,s)}{P(e)P(s)} = v_T I[E;S]$$

Shannon's Mutual Information

This is also the Optimal Predictor of Environment |Signal ~ Bayes ~ Machine Learning

~maximal fitness

Ultimately, this involves developing a new theory of optimal choice:

- intertemporal (life course)

- deals with uncertainty
- social networks, power, distribution: organizations
- consistent across scales

better than 'rational choice theory', beyond behavioral economics



Current Opinion in Psychology Volume 33, June 2020, Pages 183-188



The effects of low socioeconomic status on decision-making processes

Jennifer Sheehy-Skeffington 🖂

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https://doi.org/10.1016/j.copsyc.2019.07.043 >

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Highlights

- Low income groups are criticised for making suboptimal decisions in domains such as health and finance.
- These reflect a psychological shift in response to socioecological cues prevalent in low socioeconomic status (SES) contexts.
- Low SES experiences present the cues of resource scarcity, environmental instability, and low subjective social status.
- These trigger a regulatory shift toward present (over future) goals and the up/down-regulation of specific cognitive skills.

https://www.sciencedirect.com/science/article/pii/S2352250X1930123X?via=ihub

Population dynamics of growth and learning



Free Access

The Rules of Information Aggregation and Emergence of Collective Intelligent Behavior

Luís M. A. Bettencourt

First published: 12 October 2009 | https://doi.org/10.1111/j.1756-8765.2009.01047.x | Citations: 20

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SECTIONS

IUS Ch 9



Abstract

Information is a peculiar quantity. Unlike matter and energy, which are conserved by the laws of physics, the aggregation of knowledge from many sources can in fact produce more information (synergy) or less (redundancy) than the sum of its parts. This feature

https://onlinelibrary.wiley.com/doi/10.1111/j.1756-8765.2009.01047.x



Physica A: Statistical Mechanics and its Applications Volume 607, 1 December 2022, 128180



Statistical dynamics of wealth inequality in stochastic models of growth

Jordan T. Kemp * 😤 🖾, Luis M.A. Bettencourt ^{b, c, d} 🖾

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Highlights

- · Agent growth rates and initial resources correlations affect aggregate growth rates.
- Negative correlated growth rate assignments reduce aggregate growth.
- · Volatility and growth rate, resource covariance dominate intermediate time dynamics.
- Variances in growth rates across the population dominate long time ٠ inequality.
- · The effects of heterogeneity on inequality outpace effects on growth.

https://www.sciencedirect.com/science/article/pii/S0378437122007385





April 2023

Article Contents

Abstract

Introduction Theory and modeling of information-based growth Population effects of information dynamics

Learning increases growth and reduces inequality in shared noisy environments a Jordan T Kemp 🖾, Luís M A Bettencourt 🛛 Author Notes

PNAS Nexus, Volume 2, Issue 4, April 2023, pgad093, https://doi.org/10.1093/pnasnexus/pgad093 Published: 22 March 2023 Article history +

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Abstract

Stochastic multiplicative dynamics characterize many complex natural phenomena such as selection and mutation in evolving populations, and the generation and distribution of wealth within social systems. Population heterogeneity in stochastic growth rates has been shown to be the critical driver of wealth inequality over long time scales. However, we still lack a general statistical theory that systematically explains the origins of these heterogeneities resulting from the dynamical adaptation of agents to their

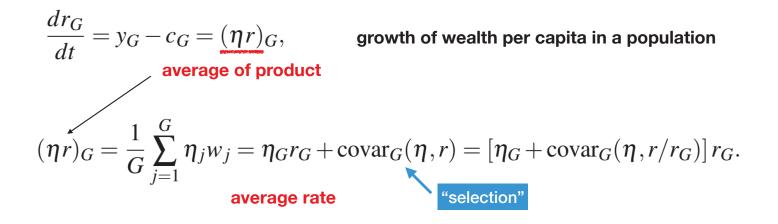
https://academic.oup.com/pnasnexus/article/2/4/pgad093/7083303



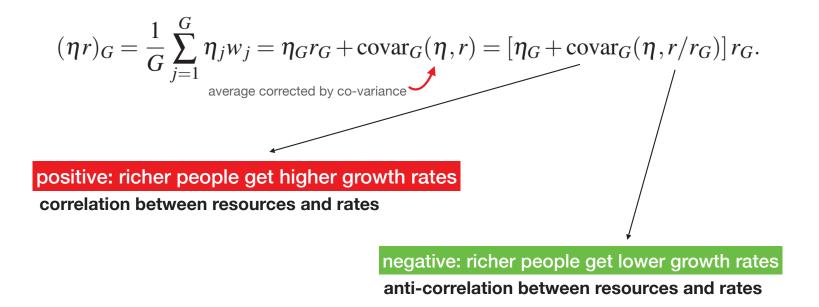
Growth rates for individuals vs collectives

Is the growth rate for a city the average of the growth rates for individuals?

Not in general:
$$r_G = \frac{1}{G} \sum_{j=1}^G r_j$$
 average wealth



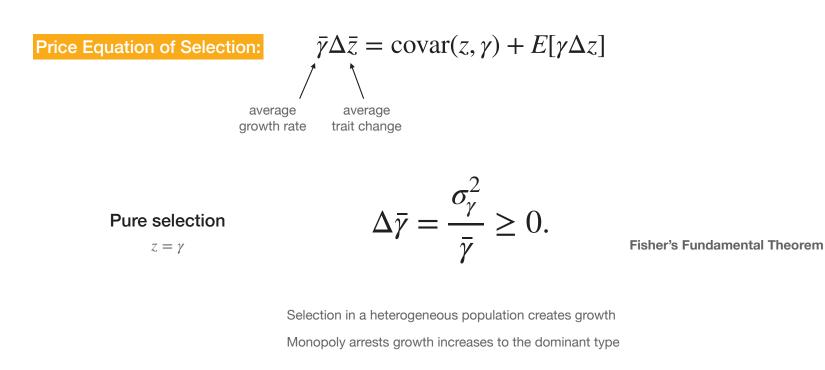
Consequences:



What situation gives the highest overall growth rates?

What situation gives the lowest inequality?

This is an instance of a very general result



Increasing growth in a population or maintaining it under obsolesce requires maintaining population diversity

This is a large-scale, long term result

For *N* agents we get:

$$I(e; \{s_i\}) = H[e] - H[e|\{s_i\}]$$

$$= -\sum_{i=1}^{N} \frac{\Delta H[e]}{\Delta s_i} - \sum_{i>j=1}^{N} \frac{\Delta^2 H[e]}{\Delta s_i \Delta s_j} - \cdots \frac{\Delta^N H[e]}{\Delta s_1 \dots \Delta s_N},$$

$$-\frac{\Delta H[e]}{\Delta s_i} = H[e] - H[e|s_i] = I(e, s_i);$$

For 2 agents we get:

 $I(e; \{s_1, s_2\}) = I(e; s_1) + I(e; s_2) - R_I(e; s_1, s_2); \quad R_I(e; s_1, s_2) = I(s_1; s_2) - I(s_1; s_2|e)$

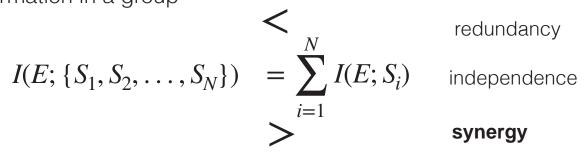
Coefficient of redundancy

Coordination given environment

Properties of Information and Institutions

Many problems need organizations: firms, NGOs, governments

Information in a group



more than the sum of the parts

The advantage of collective action requires synergy : diverse collective institutions

Bettencourt, Topics in Cognitive Science, 2009

to evaluate quality of an institution (group)

compare individual action to collective production

$$\Delta r_i = I(E; S_1) r_i(0) - c_i$$

Individual knowledge

 $\Delta r_i = B_i I(E; \{S_1, S_2, \dots, S_N\}) r_i(0) - C_i$

distribution of collective gains collective knowledge

collective costs

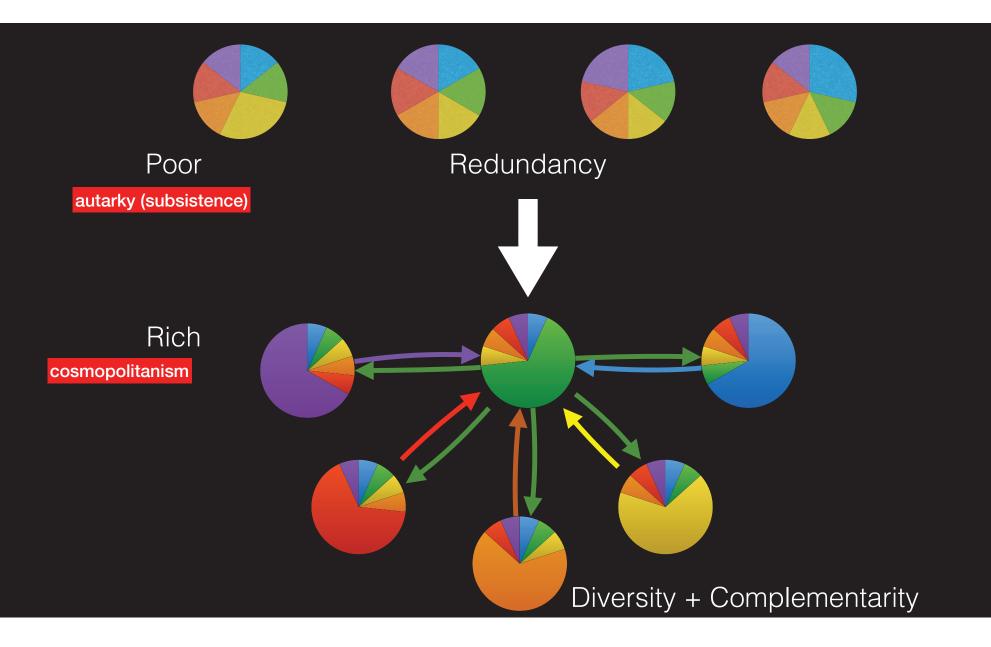
Collaborating in the groups is preferred when:

$$B_{i}I(E; \{S_{1}, S_{2}, \dots, S_{N}\}) > I(E; S_{i}) + \frac{C_{i} - c_{i}}{r_{i}(0)}$$
'fair' re-distribution of gains
$$f$$
Instance of the second strong synergy
Instance of the second synergy

osts of collective action vs individual cost

low individual information

cities / firms are good at strong synergies, lower collective costs; but not always fair distribution



Human development as a network process (in cities)

1. Measuring Human Development across scales



<u>nature</u> > <u>npj urban sustainability</u> > <u>articles</u> > article

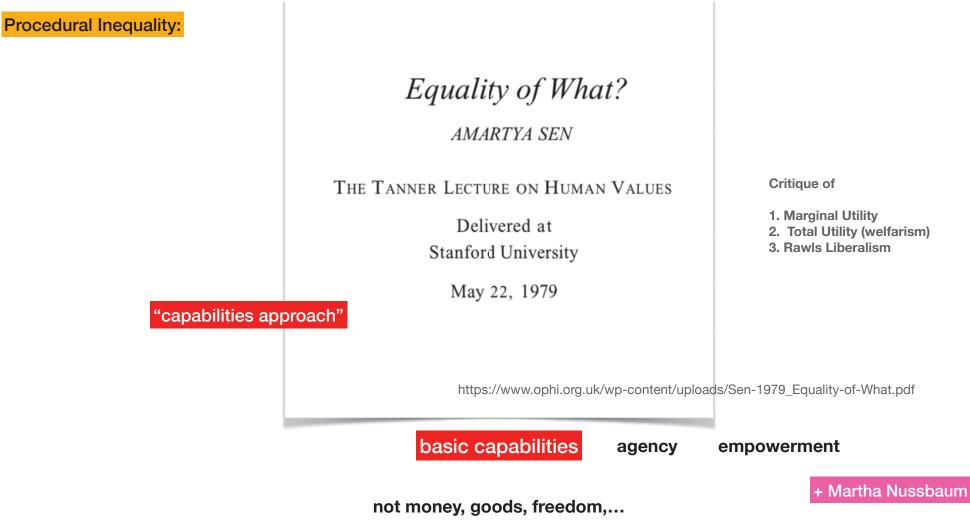
Article | Open Access | Published: 20 February 2023

Measuring health and human development in cities and neighborhoods in the United States

Suraj K. Sheth 🖾 & Luís M. A. Bettencourt

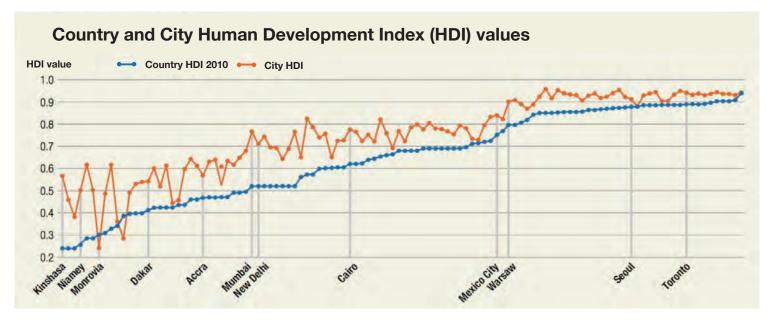
npj Urban Sustainability 3, Article number: 7 (2023) Cite this article

https://www.nature.com/articles/s42949-023-00088-y



https://ir.lawnet.fordham.edu/cgi/viewcontent.cgi?article=3391&context=flr

Broad Human Development is a Feature of Larger Cities !



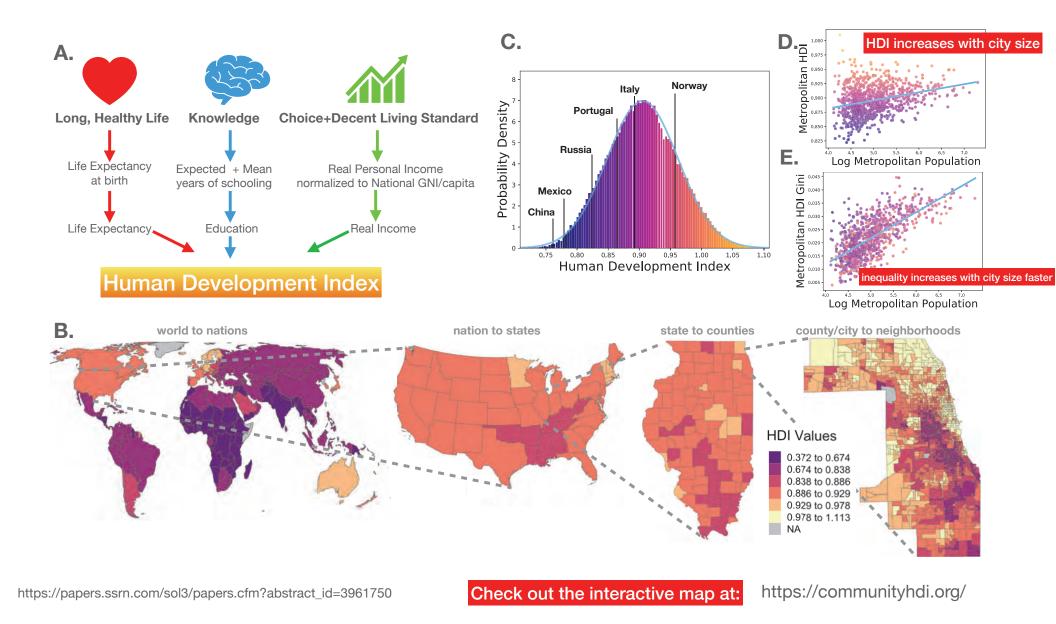
credit: UNDP Human Development Report 2013

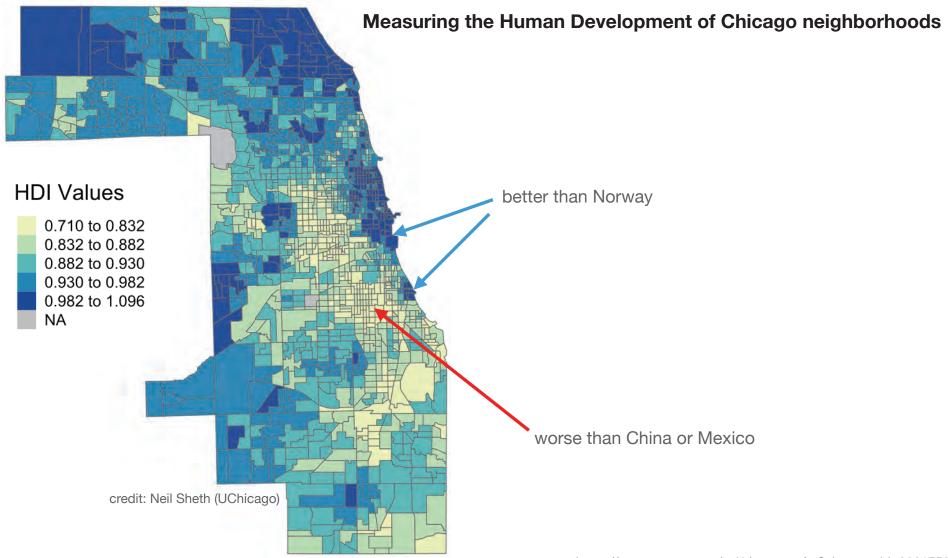
Human Development Index

"measures the ability of human populations to lead, long, healthy and fulfilled lives"

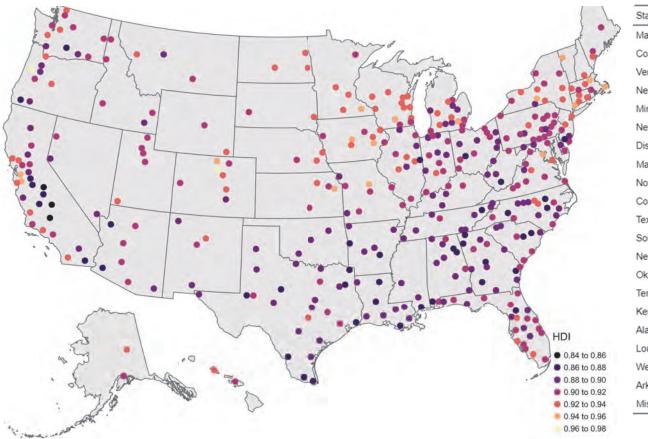
Rank HDI 2019 data (2020 🛊 Average annual **Country or Territory** \$ Change over 5 2019 data (2020 Mahbub ul Haq . HDI growth (2010- \$ years (2014)^[15] * report)[14] report)[14] 2019)[15] 1/3 $HDI_i = (education_i. life expectancy_i. real income_i)$ 1 H Norway 0.957 ▲ 0.20% 2 A (7) Ireland 0.955 ▲ 0.65% 2 -+ Switzerland 0.955 ▲ 0.16% 4 A (7) Hong Kong 0.949 ▲ 0.54% 4 **(**4) Iceland 0.949 ▲ 0.62% 6 **V**(3) 0.947 Resources (wealth, energy) Germany ▲ 0.24% 7 **V**(3) 0.945 A 0.41% Sweden 8 0.944 **V**(2) Australia ▲ 0.17% **Human Capabilities** Time 8 **V**(1) 0.944 Netherlands ▲ 0.32% 10 Denmark 0.940 7 (6) ▲ 0.28% 11 + Finland **v** (2) 0.938 ▲ 0.26% Information (knowledge) 11 Singapore 0.938 ▲ 0.35% -13 Stand Kingdom 0.932 ▲ 0.24% -14 **(1)** Belgium 0.931 ▲ 0.25% 14 A (3) 0.931 ▲ 0.30% New Zealand 16 **v**(1) **L** Canada 0.929 ▲ 0.34% The US is now #17 17 **V**(3) United States 0.926 ▲ 0.12%

because of health and educational inequality



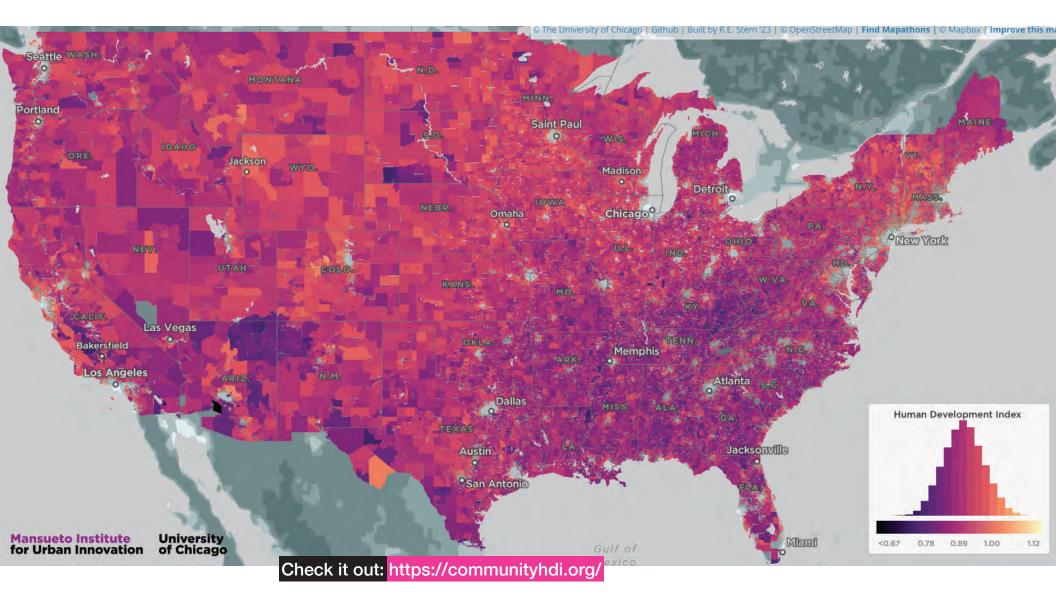


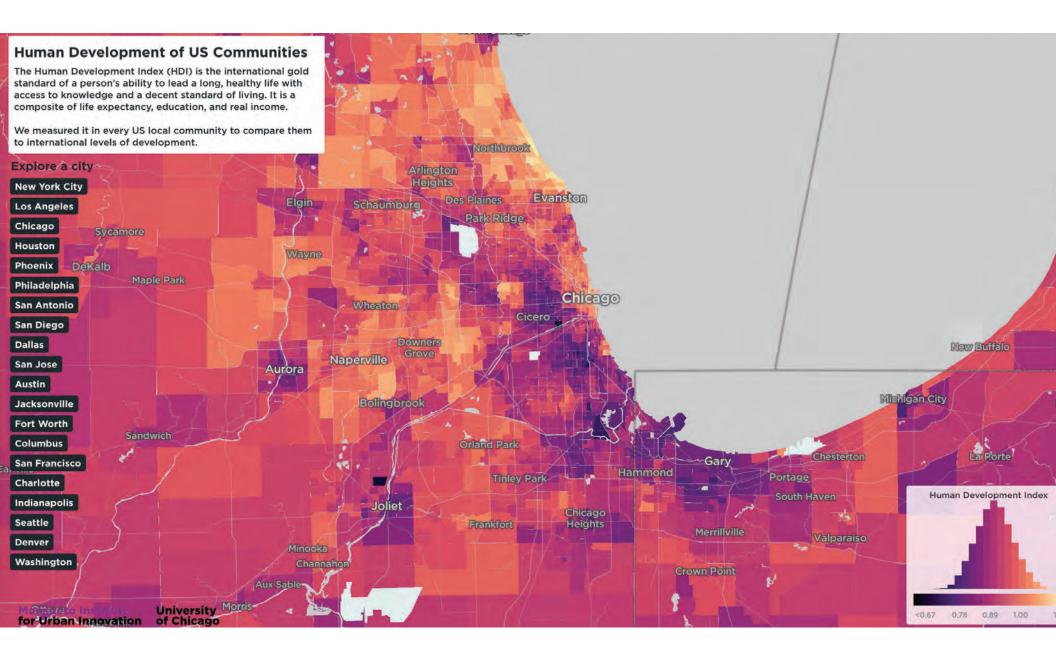
https://papers.ssrn.com/sol3/papers.cfm?abstract_id=3961750

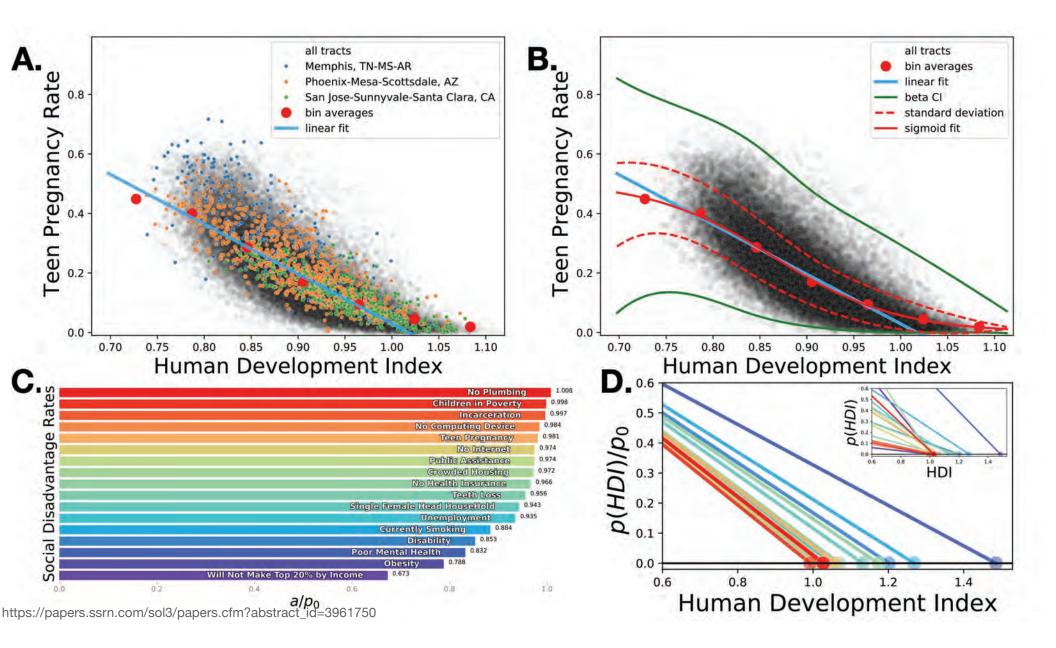


State/Federal District	HDI	Metropolitan Area	HDI
Massachusetts	0.9407	Boulder, CO	0.9624
Connecticut	0.9378	Corvallis, OR	0.9622
Vermont	0,9348	Ann Arbor, MI	0.9570
New Hampshire	0.9345	Iowa City, IA	0.9567
Minnesota	0.9330	San Jose-Sunnyvale-Santa Clara, CA	0.9557
New Jersey	0.9281	Bridgeport-Stamford-Norwalk, CT	0.9553
District of Columbia	0.9273	Ames, IA	0.9532
Maryland	0.9271	Lawrence, KS	0.9496
North Dakota	0.9244	Boston-Cambridge-Newton, MA-NH	0.9491
Colorado	0.9240	San Francisco-Oakland-Hayward, CA	0.9486
Texas	0.8964	Gadsden, AL	0.8647
South Carolina	0.8935	McAllen-Edinburg-Mission, TX	0.8641
Nevada	0.8919	Lake Havasu City-Kingman, AZ	0.8637
Oklahoma	0.8879	Laredo, TX	0.8614
Tennessee	0.8874	Dalton, GA	0.8614
Kentucky	0.8843	Brownsville-Harlingen, TX	0.8605
Alabama	0.8839	Yakima, WA	0.8604
Louisiana	0.8838	Pine Bluff, AR	0.8601
West Virginia	0.8832	Bakersfield, CA	0.8595
Arkansas	0.8798	Visalia-Porterville, CA	0.8579
Mississippi	0.8762	Madera, CA	0.8572

https://papers.ssrn.com/sol3/papers.cfm?abstract_id=3961750



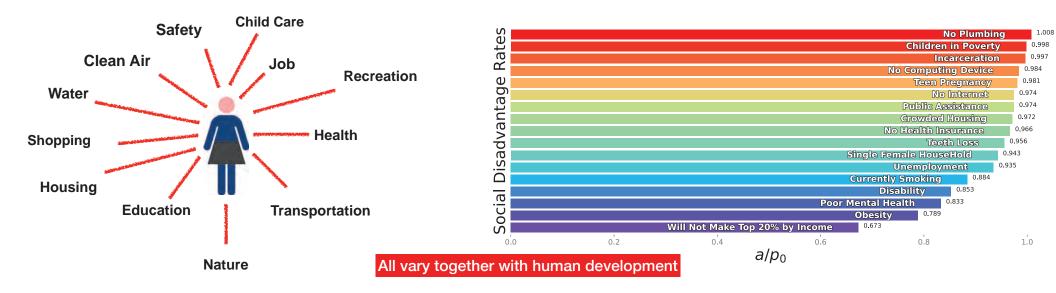




"Anna Karenina Principle" of Human Development

A deficiency in any one of a number of factors dooms an endeavor to failure. Consequently, a successful endeavor (subject to this principle) is one for which every possible deficiency has been avoided. https://en.wikipedia.org/wiki/Anna Karenina principle

All *high* human development communities are alike ("no problems"); but each *low* human development community is challenged in its own way



Solution: ensure systemic human capabilities = education and health and decent real income

Human development as a network process (in cities)

2. Incipient city networks and spatial network development

can we accelerate it, and make it better?

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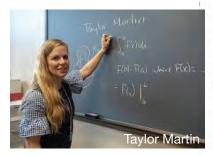
HOME > SCIENCE ADVANCES > VOL. 4, NO. 8 > TOWARD CITIES WITHOUT SLUMS: TOPOL.

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Abstract

ef.

Science Advances



The world is urbanizing quickly with nearly 4 billion people presently living in urban areas, about 1 billion of them in slums. Achieving sustainable development from rapid urbanization relies critically on creating cities without slums. We show that it is possible to diagnose systematically the central physical problem of slums -the lack of spatial accesses and related services-using a topological analysis of neighborhood maps and resolved by finding solutions to a sequence of constrained optimization problems. We set up the problem by showing that the built environment of any city can be decomposed into two types of networked spacesaccesses and places-and prove that these spaces display universal topological characteristics. We then show that while the neighborhoods of developed cities express the same common topology, urban slums fall into a different topological class. We demonstrate that it is always possible to find solutions that grow a street network in existing slums, providing universal accesses at minimal disruption and cost. We then show how elaborations of this procedure that include local preferences and reduce travel distances between places result from additional access construction. These methods are presently taking effect in neighborhoods in Cape Town (South Africa) and Mumbai (India), demonstrating their practical feasibility and emphasizing their role as a platform to enable communities and local governments to combine technical knowledge with local aspirations into contextually appropriate urban sustainable development solutions.







Open Access Editor's Choice Article

Worldwide Detection of Informal Settlements via Topological Analysis of Crowdsourced Digital Maps

by 😰 Satej Soman 1.* 🖾 🧟 🤮 Anni Beukes 1 🖂 🤤 😰 Cooper Nederhood 1 🖾 🧕 Nicholas Marchio 1 2 and C Luis M. A. Bettencourt 1.2,3 2 0

¹ Mansueto Institute for Urban Innovation, University of Chicago, Chicago, IL 60637, USA ² Department of Ecology and Evolution, University of Chicago, Chicago, IL 60637, USA

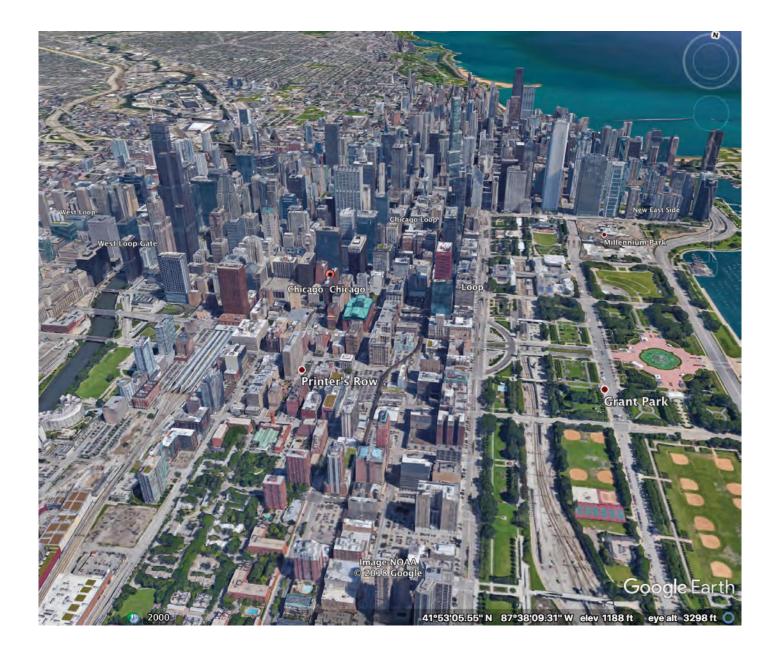
³ Department of Sociology, University of Chicago, Chicago, IL 60637, USA

* Author to whom correspondence should be addressed.

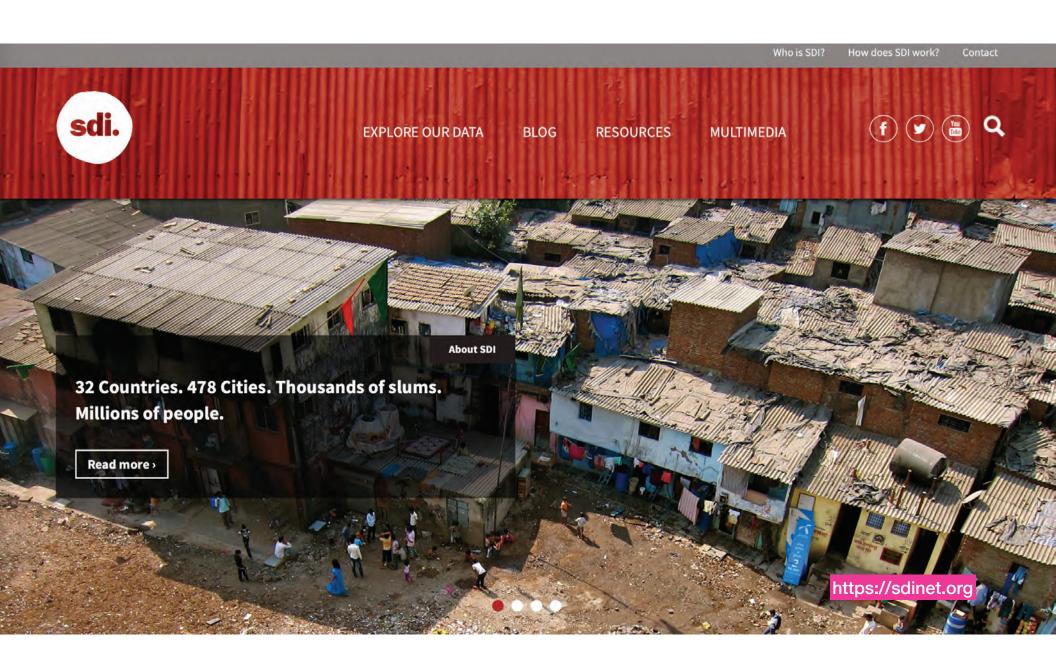
ISPRS Int. J. Geo-Inf. 2020, 9(11), 685; https://doi.org/10.3390/ijgi9110685



https://www.science.org/doi/10.1126/sciadv.aar4644









✓ Furniture shops

Communications

A 100

Other

Main means of transportation:

A CO

Mr. Hityaie

Train

* Walling

Bus

Motorbike

✓ Temples

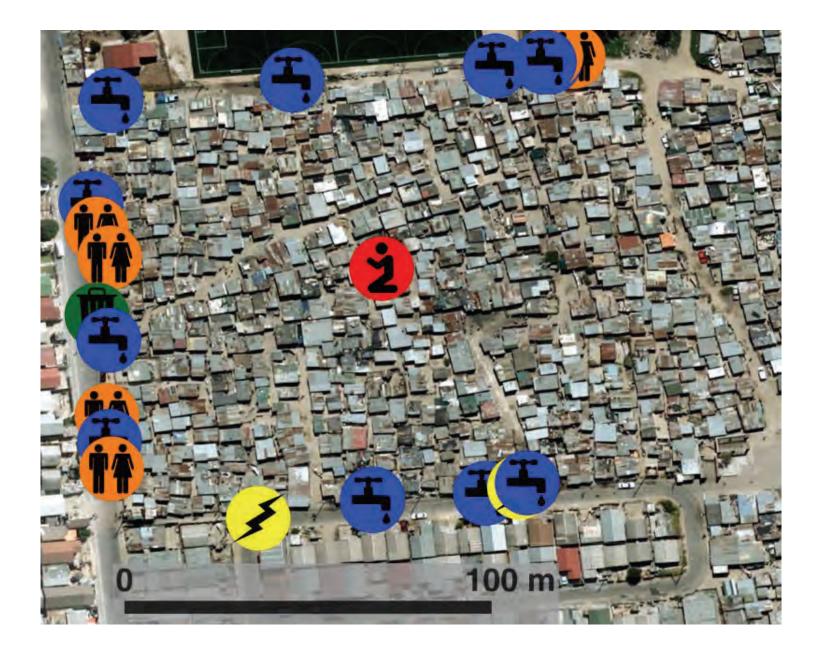
🗸 Car repair

✓ Churches

https://knowyourcity.info

Mosques





"This question is so banal, but seemed to me worthy of attention in that [neither] geometry, nor algebra, nor even the art of counting was sufficient to solve it."



The Königsberg Bridge problem asked whether it was possible for a person to walk through the city, crossing each bridge once and only once. Image credit: ScienceSource/SPL.

https://www.pnas.org/doi/10.1073/pnas.1904847116

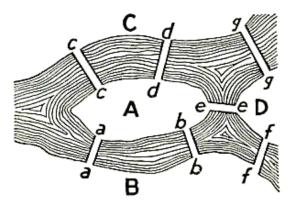


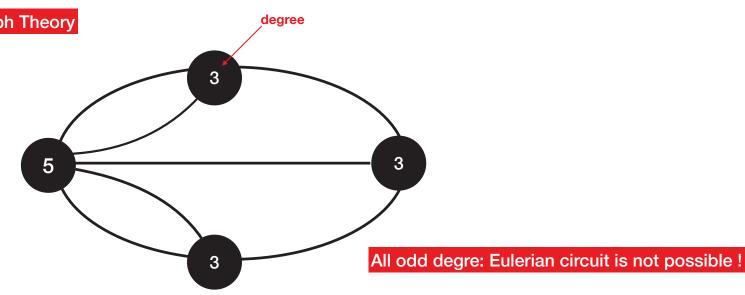
FIGURE 98. Geographic Map: The Königsberg Bridges.

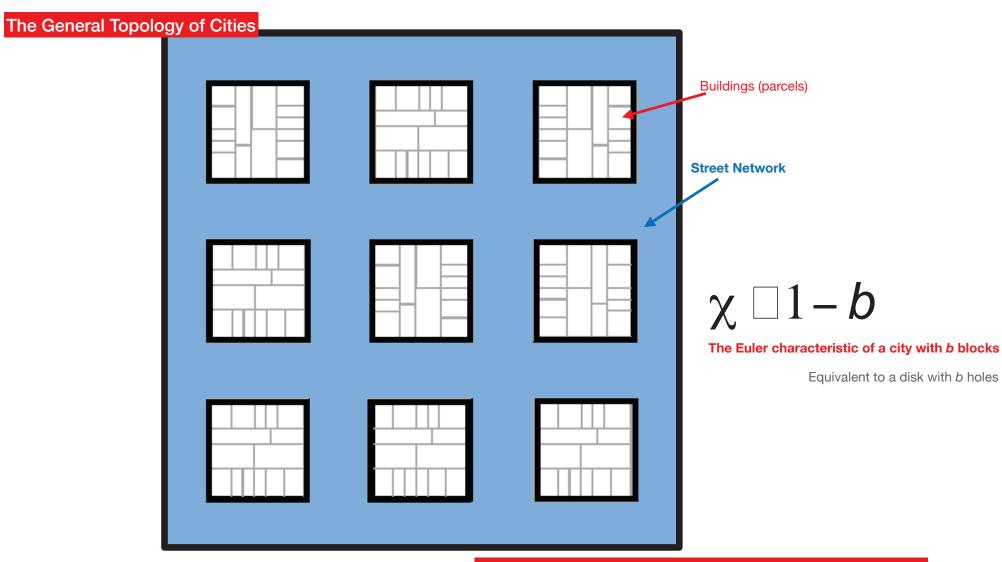
https://mathworld.wolfram.com/KoenigsbergBridgeProblem.html

Geometry —> Topology —> Graph Theory

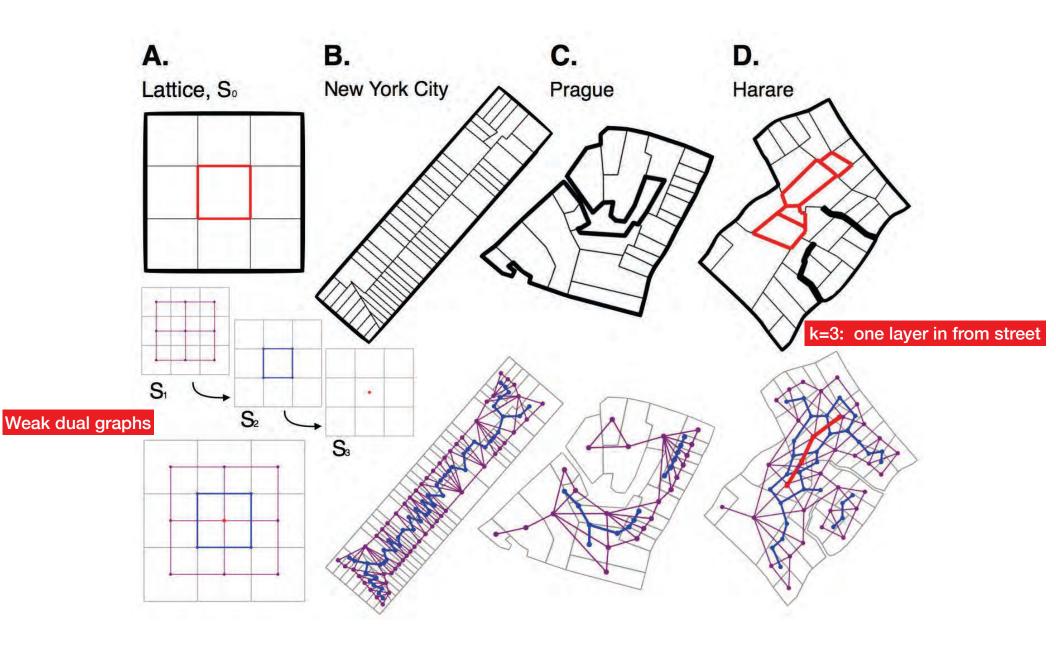


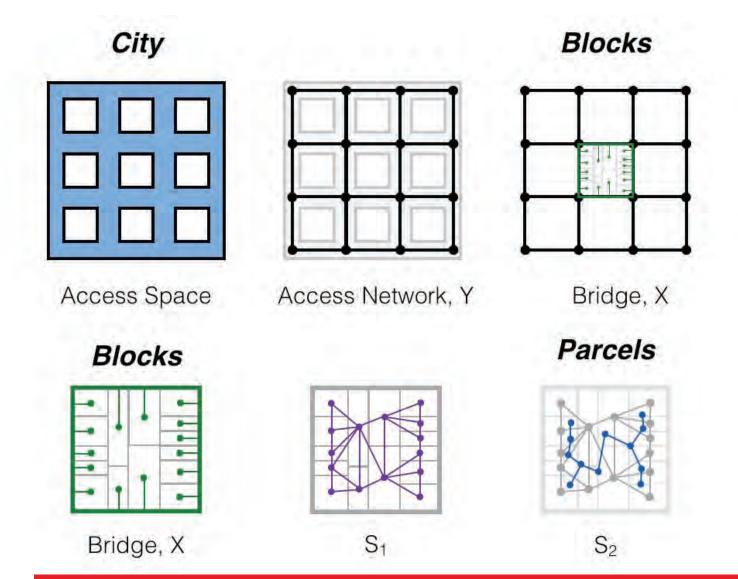
number of bridges must be even, expect at starting and finishing points





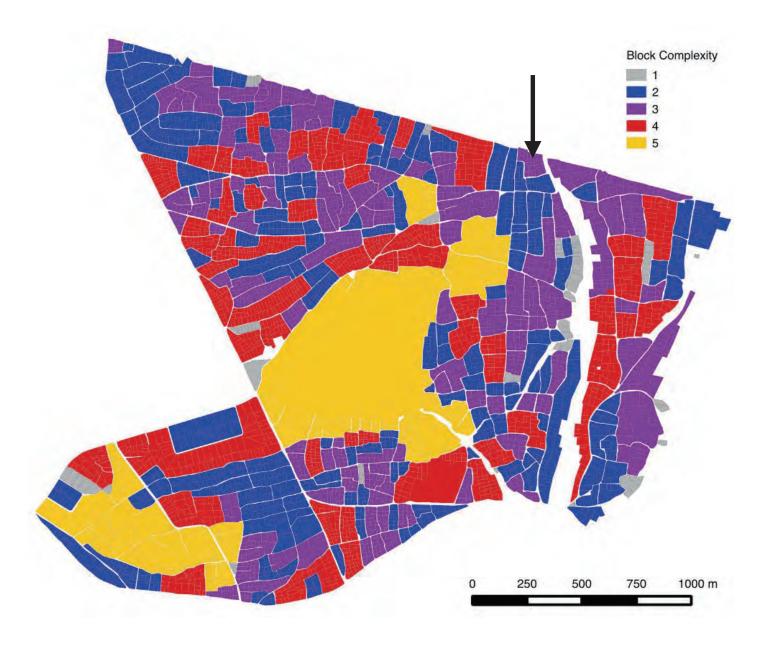
Every city is **topologically** like this, but geometry varies





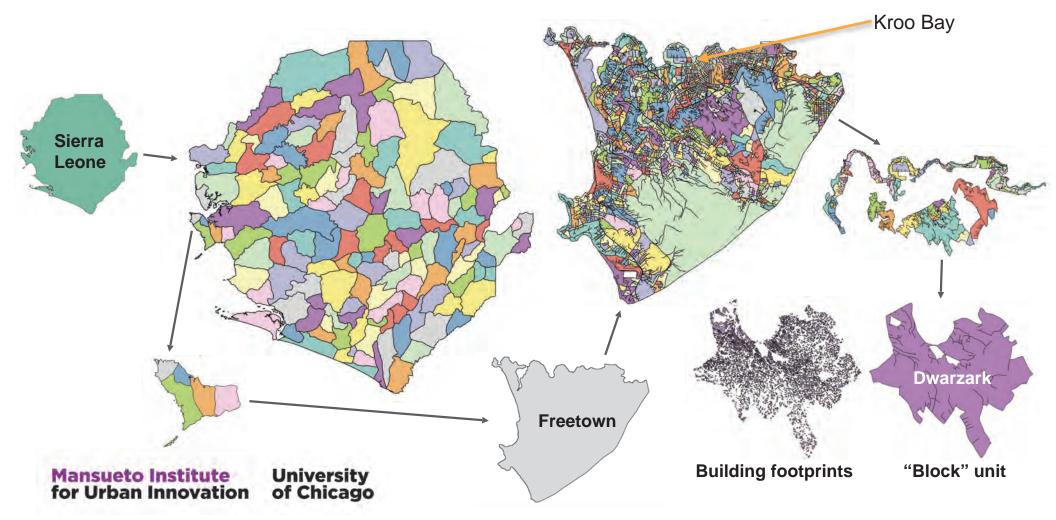
Algorithm for analyzing the topology of any city: And identify buildings without accesses



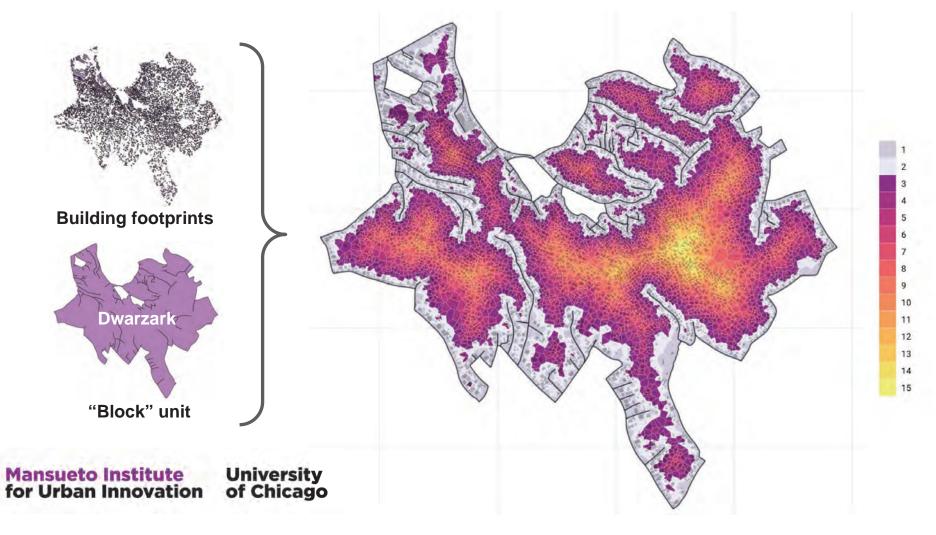


Understanding settlements from global to local scales

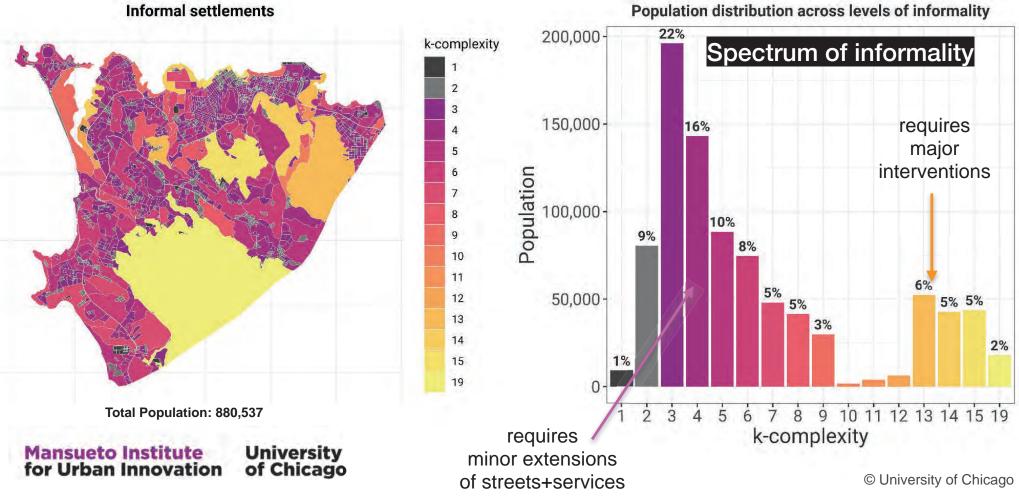
a global map (of cities) neighborhood by neighborhood



Detecting informality down to the street block

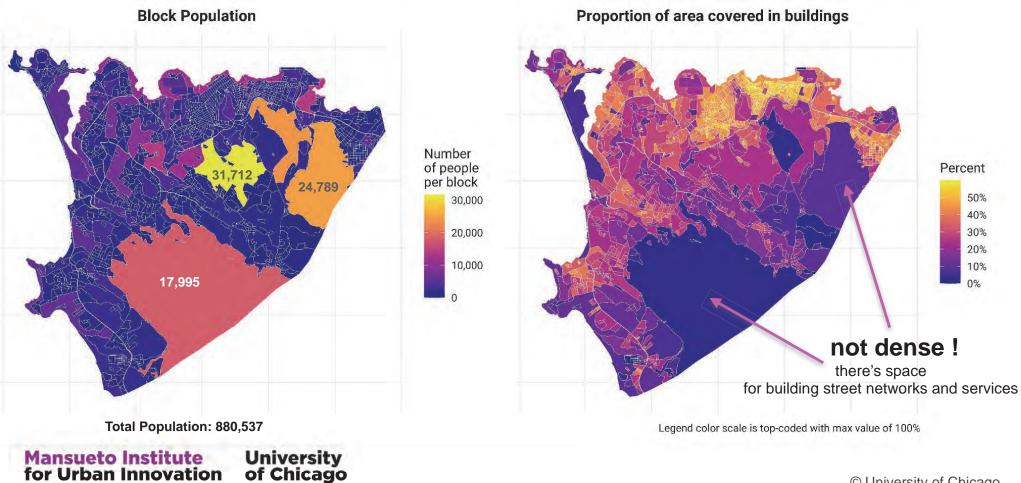


Mapping informality and population in Freetown



Population distribution across levels of informality

New indicators for measuring development in cities



© University of Chicago

Million Neighborhoods Data Explorer

Mansueto Institute for Urban Innovation :: University of Chicago

En

K Complexity

mapbox

Ata Explorer Chicago

To take up this project, young residents of Makoko were taught to pilot drones and populate the map with images from the community. Hounkpe is one of the residents on the project that is also teaching residents how to fly drones.

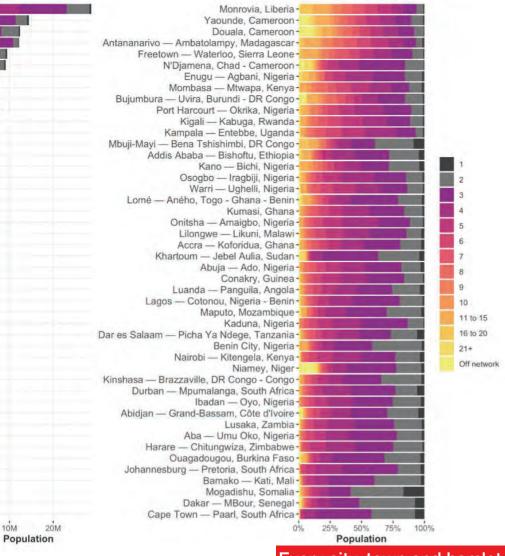


YASUYOSHI CHIBA/AFP/Getty Images

A classroom session at a private school in the Makoko on March 1, 2019.

"For years, Makoko has been ignored by governments of the state. This project is the first part of a conversation around inclusion. We are helping map out Makoko so that governments and other organizations can provide interventions and access to social services like electricity, healthcare, and education." Ottaviani told CNN.

https://www.cnn.com/2020/02/26/africa/nigeria-makoko-mapping-intl/index.html



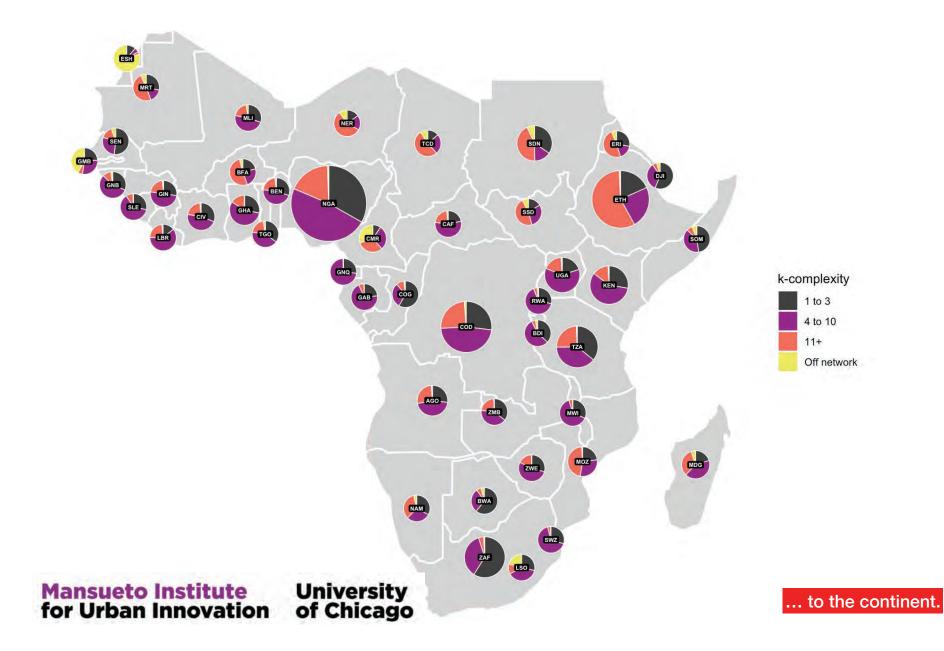
Lagos - Cotonou, Nigeria - Benin -Johannesburg - Pretoria, South Africa -Kinshasa - Brazzaville, DR Congo - Congo -Onitsha - Amaigbo, Nigeria -Addis Ababa - Bishoftu, Ethiopia -Luanda - Panguila, Angola -Nairobi - Kitengela, Kenya -Khartoum - Jebel Aulia, Sudan -Accra - Koforidua, Ghana-Kano - Bichi, Nigeria -Dar es Salaam - Picha Ya Ndege, Tanzania -Abidjan - Grand-Bassam, Côte d'Ivoire -Kampala - Entebbe, Uganda -Cape Town - Paarl, South Africa -Ibadan - Oyo, Nigeria -Dakar - MBour, Senegal -Durban - Mpumalanga, South Africa -Kumasi, Ghana -Abuja - Ado, Nigeria -Bamako — Kati, Mali-Port Harcourt - Okrika, Nigeria k-complexity Maputo, Mozambique -Lomé — Aného, Togo - Ghana - Benin -Antananarivo — Ambatolampy, Madagascar-Douala, Cameroon -Yaounde, Cameroon -Ouagadougou, Burkina Faso -Harare - Chitungwiza, Zimbabwe -Lusaka, Zambia -Conakry, Guinea -Mogadishu, Somalia -Mbuji-Mayi — Bena Tshishimbi, DR Congo -Kigali — Kabuga, Rwanda -N'Djamena, Chad - Cameroon -Benin City, Nigeria -Freetown - Waterloo, Sierra Leone -Kaduna, Nigeria Warri - Ughelli, Nigeria -Mombasa - Mtwapa, Kenya -Enugu - Agbani, Nigeria -Monrovia, Liberia -Bujumbura - Uvira, Burundi - DR Congo -Aba - Umu Oko, Nigeria -Osogbo - Iragbiji, Nigeria-Niamey, Niger-Lilongwe - Likuni, Malawi -OM

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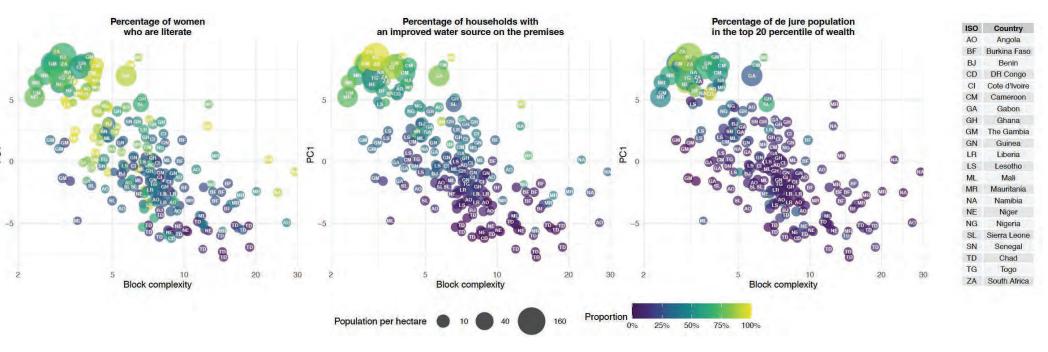
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Thank you !

Luís M. A. Bettencourt @BettencourtLuis Bettencourt@uchicago.edu }

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More on the complex systems, science of cities and lots of data:



Introduction to Urban Science

Evidence and Theory of Cities as Complex Systems

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A novel, integrative approach to cities as complex adaptive systems, applicable to issues ranging from innovation to economic prosperity to settlement patterns.

https://mitpress.mit.edu/books/introduction-urban-science