

Master(of)equations

In honor of Tamás S. Bíró on the occasion of his 70th birthday

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Wigner RCP

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Genesis 1:2:

“The earth was a formless chaos,
and darkness covered the face of the deep...”
and still it is ...



1984 Characterization paper: the prototype (Glanzel W. , T A , Schubert A)

A distribution can be characterized by relations by truncated moments,

$$e(x) = \mathbb{E}[X | X > x],$$

For suitable functions this identifies Pearson-type continuous distributions and their discrete analogues. In particular

$$e(x) = ax + b$$

characterizes the Warring distribution for $a + b - 1 \geq 2\sqrt{b(a - 1)}$ and the uniform discrete distribution when $a = 0.5$ and $2b$ is a natural number.

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Based on that today the characterized distribution families are:

Pearson, beta, beta-prime/infinite beta, finite beta, gamma, exponential, Pareto, inverse-gamma/Pearson-V type, Irwin, hypergeometric, inverse Pólya-Eggenberger, Waring, binomial, Poisson, negative binomial, negative hypergeometric, discrete uniform, Cauchy, normal, uniform, power-function, modified Weibull, log-modified Weibull, Weibull-geometric, beta Weibull-geometric, generalized exponential geometric, gamma exponentiated Weibull, beta exponential, gamma exponentiated, gamma geometric, generalized exponential, exponentiated Pareto, exponentiated gamma, SKS, generalized gamma, Amoroso, New Weibull-Pareto, exponentiated transmuted generalized Rayleigh, transmuted Weibull Lomax, McDonald generalized-K, beta generalized, Kumaraswamy generalized, McDonald normal, Kumaraswamy-inverse Weibull, beta extended Weibull, gamma-G, gamma extended Weibull, generalized beta extended Pareto, Poisson Birnbaum-Saunders, skew normal, ratio of Rayleigh, ratio of gamma/exponential, ratio of Weibull, Zografos-Balakrishnan log-logistic, Lindley variants.

2020 Bíró line: replace truncated moments by tail cumulatives

For a nonnegative variable X with density P ,

$$C(x) = \int_x^{\infty} P(y) dy, \quad F(x) = \int_x^{\infty} \frac{y}{\langle X \rangle} P(y) dy.$$

These are the population tail and the value-weighted tail. In scientometrics, X becomes citations or productivity.

Characterization view

- Condition on the tail $X > x$.
- Compare different truncated expectations.
- Distribution class follows from simple tail relations.

$$\mathbb{E}[g(X) \mid X > x] = \lambda(x)\mathbb{E}[h(X) \mid X > x].$$

The continuity is methodological: infer global distributional structure from tail-conditioned quantities.

Gintropic view

- Keep two canonical tail summaries.
- Plot F against C as a Lorenz map.
- Study the gap from equality.

$$\sigma(x) = F(x) - C(x), \quad G = 2 \int_0^1 \sigma(C) dC.$$

Matthew (25:29), "For to everyone who has, more will be given, and he will have an abundance; but from the one who has not, even what he has will be taken away. "

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2018,19 Bíró TS. Nédá Z growth-reset master equation supplies a dynamical route to the Tsallis–Pareto family.

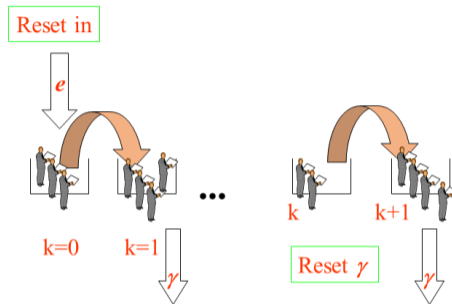
Linear growth with constant reset

$$\dot{P}(k, t) = \mu(k-1)P(k-1, t) - [\mu(k) + \gamma(k)]P(k, t),$$

with a reset boundary equation at $k = 0$.

$$\mu(k) = ak + b, \quad \gamma(k) = \gamma.$$

This mechanism produces a stationary Tsallis–Pareto / cut power-law distribution.



- Preferential growth: cumulative advantage, “rich get richer” – Matthew effect.
- Reset: retirement, failure, dropout, field exit, loss of visibility.
- Gini dynamics tracks the emergence of inequality in the evolving distribution.

The population density $\rho(x, t)$ evolves through transition rates $W(y | x, \rho)$.

$$\frac{\partial \rho(x, t)}{\partial t} = \int W(x | y, \rho) \rho(y, t) dy - \rho(x, t) \int W(y | x, \rho) dy.$$

- The first term is inflow into state x from other states y .
- The second term is outflow from state x to other states y .

Let X be citation count per publication for a researcher or institution.

$$\langle X \rangle = \frac{N_{\text{cit}}}{N_{\text{pub}}}, \quad h = N_{\text{pub}} C(h).$$

For scaling distributions,

$$P(x) = \frac{1}{\langle X \rangle} p\left(\frac{x}{\langle X \rangle}\right),$$

so C, F, σ depend on $z = x/\langle X \rangle$. Do a little calculation...

$$\frac{h^2}{N_{\text{cit}}} = z \left(1 + \frac{z}{b-1}\right)^{-b}.$$

At $z = 1$ when $b = 2$

$$\frac{h^2}{N_{\text{cit}}} \simeq \frac{1}{4}.$$

Consequence

The Hirsch index is not independent: it is constrained by N_{pub} , N_{cit} , and the tail parameter; the rule $\sqrt{N_{\text{cit}}} \simeq 2h$ is a bulk scaling trend, not an identity.

All who take the citation will judged by the citation.



Tamás S. Biró

Wigner Research Centre for Physics

Field: Physical Sciences

Discipline: Physics

Total publications: 236

Total predicted citations: 6,738

Overall (All Fields) predicted h-index: 45

	Scholars ranked	Productivity			Impact			Quality			ScholarGPS® Ranks ¹		
		Publications	Rank	Top Percentage Rank	Citations	Rank	Top Percentage Rank	h-index	Rank	Top Percentage Rank	Rank	Z Score ¹	Top Percentage Rank
Overall (All Fields)	29,161,103	96.08	45,743	0.16%	1,358.12	180,491	0.62%	20	107,399	0.4%	82,985	1.86	0.28%
Physical Sciences	3,705,091	96.08	9,632	0.26%	1,358.12	38,041	1.03%	20	22,069	0.65%	17,638	1.81	0.48%
Physics	998,186	96.08	3,153	0.32%	1,358.12	11,535	1.16%	20	6,745	0.73%	5,598	1.79	0.56%
Specialties													
Nuclear physics	54,086	10.95	1,174	2.17%	248.75	1,502	2.78%	8	959	2.07%	1,136	1.78	2.1%
Quark	16,088	10.78	331	2.05%	189.98	570	3.54%	7	345	2.64%	378	1.8	2.35%
Gauge theory	12,224	9.42	189	1.55%	44.67	1,645	13.46%	4	567	6.06%	521	1.74	4.26%
Collision	108,844	7.27	948	0.87%	352.65	483	0.44%	10	204	0.21%	353	1.95	0.32%
QCD matter	1,653	6.87	20	1.18%	94.15	78	4.69%	4	48	4.02%	43	1.8	2.57%
Particle physics	44,978	5.75	2,431	5.4%	57.3	4,931	10.96%	4	3,558	9.41%	3,430	1.57	7.62%
Energy	1,669,067	5.7	27,224	1.63%	82.96	70,795	4.24%	6	22,774	1.71%	31,262	1.85	1.87%
Statistical mechanics	38,762	5.15	1,015	2.62%	68.77	1,957	5.05%	5	600	2.1%	961	1.84	2.48%
Thermodynamics	190,811	4	3,524	1.89%	109.42	3,743	1.96%	4	3,252	2.37%	3,098	1.89	1.62%
Lattice gauge theory	1,636	3.42	112	6.82%	12.5	542	33.1%	3	122	10.45%	204	1.41	12.44%
Plasma (physics)	229,182	2.67	13,295	5.83%	63.83	8,794	3.84%	3	10,444	6.6%	10,525	1.71	4.59%
Power (physics)	119,741	2.45	1,654	1.39%	71.7	1,463	1.22%	4	567	0.76%	794	1.94	0.66%
Radiation	653,917	2.33	34,148	5.27%	6.33	168,572	25.86%	2	41,237	10.74%	60,442	1.49	9.24%

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For nonnegative X with finite mean,

$$C(x) = \Pr\{X \geq x\}, \quad F(x) = \frac{\mathbb{E}[X \mathbf{1}_{X \geq x}]}{\mathbb{E}X}, \quad \sigma(x) = F(x) - C(x).$$

- $F \geq C$: the value-share tail is above the population-share tail.
- σ is nonnegative and reaches its maximum at $x = \langle X \rangle$.
- The Gini index is the area under the gintropy curve:

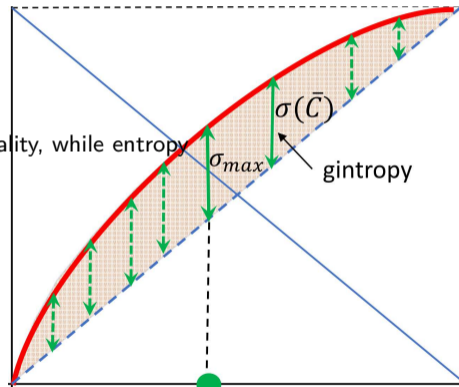
$$G = 2 \int_0^1 \sigma(C) dC.$$

Interpretation

Gintropy is not merely another index. It is the local Lorenz-gap density whose integral is the Gini index.

2023 Biro, T, Jozsa Néda - The Gini index measures inequality, while entropy measures disorder or uncertainty.

The idea of *gintropy* is to connect these two viewpoints through the Lorenz curve.



2020 Bíró Néda The Gini index can be written as the average of the gintropy density:

$$G = 2 \int_0^1 \sigma(C) dC = 2 \int_0^\infty \sigma(x)\rho(x) dx.$$

Thus σ plays the role of an “entropy-like density”, but its integral gives an inequality measure rather than a thermodynamic entropy.

Interpretation

$$\sigma(C) = 0 \iff \text{perfect equality at that cumulative level.}$$

Large $\sigma(C)$ means that the rich-end population share and the rich-end output share strongly separate.

The maximum of σ occurs at the mean output level:

$$\sigma_{\max} = \sigma(\langle x \rangle).$$

For a Tsallis–Pareto tail

$$C(x) = (1 + ax)^{-b}, \quad P(x) = ab(1 + ax)^{-b-1}, \quad b > 1,$$

with $q = 1 - 1/b$,

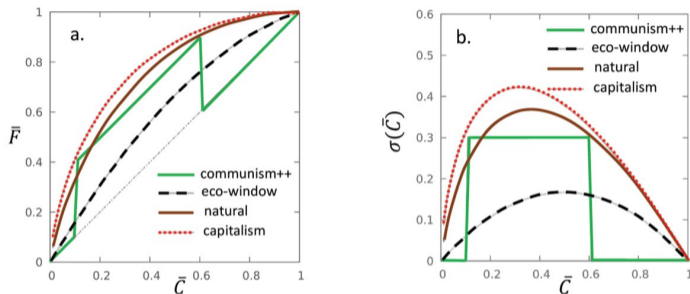
$$\sigma(C) = \frac{C^q - C}{1 - q}, \quad \lim_{q \rightarrow 1} \sigma(C) = -C \log C, \quad G = \frac{1}{q + 1}.$$

- Pareto tails generate Tsallis-type gintropy.
- The exponential limit returns the Shannon kernel.
- The same formulas apply to wealth, citations, and other valued outputs if the tail law is comparable.

Caricature	Distribution type	Gintropy shape
Communism	Single value	$\sigma(C) = 0$
Communism+++	Two classes	Step-like plateau
Eco-window	Uniform window	$\sigma(C) = 3G C(1 - C)$
Natural	Exponential	$\sigma(C) = -C \ln C$
Capitalism	Pareto / Tsallis–Pareto	$\sigma(C) = \frac{C^q - C}{1 - q}$

Message

Different social caricatures have different entropy-like densities. The Gini index is not only a number; through $\sigma(C)$ it has a whole “spectral profile” over cumulative social rank.



The labels are intentionally schematic.

- **Communism:** everyone has the same output; inequality vanishes.
- **Communism++:** a two-class society; inequality appears as a discrete jump.
- **Eco-window:** outputs are uniformly spread over an interval.
- **Natural:** exponential output distribution; the Shannon form appears.
- **Capitalism:** heavy-tailed output distribution; the Tsallis form appears.

Bis Hundert und Zwanzig Tamás!