Mechanisms for Generating Power-Law Size Distributions, Part 2

Principles of Complex Systems | @pocsvox CSYS/MATH 300, Fall, 2016 | #FallPoCS2016

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Outline

Analysis

Catchphrases

Assumptions

Words

Optimization Minimal Cost

Model

Nutshell

References

Extra

Analysis

Rich-Get-Richer Mechanism Simon's Model

Mandelbrot vs. Simon

And the winner is ...?











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The deal:

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References





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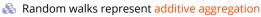
Mechanism

Optimization

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Aggregation:



- A Mechanism: Random addition and subtraction
- & Compare across realizations, no competition.
- Next: Random Additive/Copying Processes involving Competition.
- Nidespread: Words, Cities, the Web, Wealth, Productivity (Lotka), Popularity (Books, People, ...)

Pre-Zipf's law observations of Zipf's law

tachygraphy), Jean-Baptiste Estoup 🗗 [11].

Stenography (or shorthand or brachygraphy or

♣ 1910s: Felix Auerbach pointed out the Zipfitude

"Das Gesetz der Bevölkerungskonzentration"

("The Law of Population Concentration") [1].

Scientific papers per author (Lotka's law)

1910s: Word frequency examined re

of city sizes in

♣ 1926: Lotka [15]:

1924: G. Udny Yule [28]:

Species per Genus

& Competing mechanisms (trickiness)

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20 € 7 of 90

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Theoretical Work of Yore:

- 4 1953: Mandelbrot [17]: Optimality argument for Zipf's law; focus on language.
- 1955: Herbert Simon [24, 30]: Zipf's law for word frequency, city size, income, publications, and species per genus.
- 4 1965/1976: Derek de Solla Price [7, 8]: Network of Scientific Citations.
- A 1999: Barabasi and Albert [2]: The World Wide Web, networks-at-large.



- Political scientist (and much more)
- 🙈 Involved in Cognitive Psychology, Computer Science, Public Administration, Economics, Management, Sociology
- & Coined 'bounded rationality' and 'satisficing'
- Nearly 1000 publications (see Google Scholar ☑)
- An early leader in Artificial Intelligence, Information Processing, Decision-Making, Problem-Solving, Attention Economics, Organization Theory, Complex Systems, And Computer Simulation Of Scientific Discovery.
- 1978 Nobel Laureate in Economics (his Nobel bio is here \square).

Essential Extract of a Growth Model:

Random Competitive Replication (RCR):

- 1. Start with 1 elephant (or element) of a particular flavor at t=1
- 2. At time t = 2, 3, 4, ..., add a new elephant in one of two ways:
 - \triangleright With probability ρ , create a new elephant with a new flavor
 - = Mutation/Innovation
 - \bigcirc With probability 1ρ , randomly choose from all existing elephants, and make a copy.
 - = Replication/Imitation
 - Elephants of the same flavor form a group

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Random Competitive Replication:

Example: Words appearing in a language

- Consider words as they appear sequentially.
- \clubsuit With probability ρ , the next word has not previously appeared
 - = Mutation/Innovation
- $\mbox{\&}$ With probability $1-\rho$, randomly choose one word from all words that have come before, and reuse
 - = Replication/Imitation

Note: This is a terrible way to write a novel.

For example:



Fundamental Rich-get-Richer story;

& Competition for replication between individual

Competition for growth between groups of

Possible that no great knowledge of system

matching elephants is not random;

Selection on groups is biased by size;

Random selection sounds easy;

needed (but more later ...).

Your free set of tofu knives:

Some observations:

elephants is random;

- o 21 words used
- · next word 13 new with prob p
- · next word is a copy with prob 1-8 next word; prob: 6/21 ook 4/21 the 3/21 and penguin 2/21 Y21 library



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Sampling with super-duper replacement and sneaky sneaking in of new colors.

Related to Pólya's Urn Model , a special case of problems involving urns and colored balls .

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Random Competitive Replication:

Some observations:

- 🚓 Steady growth of system: +1 elephant per unit time.
- Steady growth of distinct flavors at rate ρ
- We can incorporate
 - 1. Elephant elimination

Ch. 3: An Urban Mystery, p. 46

distribution should be ..."1, 2

- 2. Elephants moving between groups
- 3. Variable innovation rate ρ
- 4. Different selection based on group size (But mechanism for selection is not as simple...)

"The Self-Organizing Economy" 3. 🗗

by Paul Krugman (1996). [14]

"...Simon showed—in a completely impenetrable

exposition!—that the exponent of the power law

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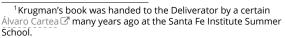
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²Let's use π for probability because π 's not special, right guys?

Random Competitive Replication:

 $P_k(t)$ = Probability of choosing an elephant that belongs to a group of size k:

- \aleph $N_{k,t}$ size k groups
- & t elephants overall

$$P_k(t) = \frac{kN_{k,t}}{t}.$$

Analysis

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Random Competitive Replication:

$N_{k,t}$, the number of groups with k elephants, changes at time t if

1. An elephant belonging to a group with k elephants is replicated:

$$\begin{split} N_{k,\,t+1} &= N_{k,\,t} - 1 \\ \text{Happens with probability } (1-\rho) k N_{k,\,t} / t \end{split}$$

2. An elephant belonging to a group with k-1elephants is replicated:

$$\begin{split} N_{k,\,t+1} &= N_{k,\,t} + 1 \\ \text{Happens with probability } & (1-\rho)(k-1)N_{k-1,\,t}/t \end{split}$$

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Random Competitive Replication:

Definitions:

PATT: KRIGNA

Self-Organizin

-Economy-

& k_i = size of a group i

 \aleph $N_{k,t}$ = # groups containing k elephants at time t.

Basic question: How does $N_{k,t}$ evolve with time?

First:
$$\sum_k kN_{k,\,t} = t = {\sf number} \ {\sf of} \ {\sf elephants} \ {\sf at} \ {\sf time} \ t$$

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Special case for $N_{1,t}$:

1. The new elephant is a new flavor: $N_{1,t+1} = N_{1,t} + 1$ Happens with probability ρ

Random Competitive Replication:

2. A unique elephant is replicated:

$$\begin{split} N_{1,\,t+1} &= N_{1,\,t} - 1 \\ \text{Happens with probability } (\mathbf{1} - \rho) N_1/t \end{split}$$

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• 9 9 € 20 of 90

Random Competitive Replication:

Putting everything together:

For k > 1:

$$\left\langle N_{k,\,t+1}-N_{k,\,t}\right\rangle = (1-\rho)\left(\frac{(+1)(k-1)\frac{N_{k-1,\,t}}{t}+(-1)k\frac{N_{k,\,t}}{t}}{t}\right) \text{primization finited Cost statements of the property of the propert$$

For k = 1:

$$\left< N_{1,\,t+1} - N_{1,\,t} \right> = (\textcolor{red}{+1})\rho + (\textcolor{red}{-1})(1-\rho)1 \cdot \frac{N_{1,\,t}}{t}$$

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Random Competitive Replication:

We have a simple recursion:

$$\frac{n_k}{n_{k-1}} = \frac{(k-1)(1-\rho)}{1+(1-\rho)k}$$

- \mathbb{A} Interested in k large (the tail of the distribution)
- & Can be solved exactly.

Insert question from assignment 4 2

For just the tail: Expand as a series of powers of 1/k

Insert question from assignment 4 🗷 We (okay, you) find

$$n_k \propto k^{-\frac{(2-\rho)}{(1-\rho)}} = k^{-\textcolor{red}{\gamma}}$$

$$\gamma = \frac{(2-\rho)}{(1-\rho)} = 1 + \frac{1}{(1-\rho)}$$

 \clubsuit Micro-to-Macro story with ρ and γ measurable.

$$\gamma = \frac{(2-\rho)}{(1-\rho)} = 1 + \frac{1}{(1-\rho)}$$

- $\mbox{\&}$ Observe $2 < \gamma < \infty$ for $0 < \rho < 1$.
- Solution For $\rho \simeq 0$ (low innovation rate):

$$\gamma \simeq 2$$

- 'Wild' power-law size distribution of group sizes, bordering on 'infinite' mean.
- & For $\rho \simeq 1$ (high innovation rate):

$\gamma \simeq \infty$

- All elephants have different flavors.
- Upshot: Tunable mechanism producing a family of universality classes.

\red{a} Recall Zipf's law: $s_r \sim r^{-\alpha}$ (s_r = size of the rth largest group of elephants)

 \clubsuit We found $\alpha = 1/(\gamma - 1)$ so:

$$\alpha = \frac{1}{\gamma - 1} = \frac{1}{1 + \frac{1}{(1 - \rho)} - 1} = 1 - \rho.$$

- & We (roughly) see Zipfian exponent [30] of $\alpha = 1$ for many real systems: city sizes, word distributions,
- & Corresponds to $\rho \to 0$, low innovation.
- & Krugman doesn't like it) [14] but it's all good.
- & Still, other quite different mechanisms are possible...
- Must look at the details to see if mechanism makes sense... more later.

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• 0 a ∩ 24 of 90

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Optimization

Extra





•9 q (~ 25 of 90

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References





• 9 9 € 26 of 90

$$\left\langle N_{1,\,t+1} - N_{1,\,t} \right\rangle = (\textcolor{red}{+1})\rho + (\textcolor{red}{-1})(1-\rho)1 \cdot \frac{N_{1,\,t}}{^t}$$



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Extra

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Random Competitive Replication:

Assume distribution stabilizes: $N_{k,t} = n_k t$ (Reasonable for t large)

- Drop expectations
- Numbers of elephants now fractional
- Okay over large time scales
- For later: the fraction of groups that have size k is n_k/ρ since

$$\frac{N_{k,t}}{\rho t} = \frac{n_k t}{\rho t} = \frac{n_k}{\rho}.$$





Random Competitive Replication:

Stochastic difference equation:

$$\left\langle N_{k,t+1}-N_{k,t}\right\rangle = (1-\rho)\left((k-1)\frac{N_{k-1,t}}{t}-k\frac{N_{k,t}}{t}\right)$$

becomes

$$n_k(t+1)-n_kt=(1-\rho)\left((k-1)\frac{n_{k-1}t}{t}-k\frac{n_kt}{t}\right)$$

$$n_k(\textcolor{red}{t} + 1 - \textcolor{red}{t}) = (1 - \rho) \left((k - 1) \frac{n_{k-1} \textcolor{red}{t}}{\textcolor{red}{t}} - k \frac{n_k \textcolor{red}{t}}{\textcolor{red}{t}} \right)$$

$$\Rightarrow n_k = (1-\rho)\left((k-1)n_{k-1} - kn_k\right)$$

$$\Rightarrow n_k (1 + (1 - \rho)k) = (1 - \rho)(k - 1)n_{k-1}$$

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What about small k?:

We had one other equation:



$$\left\langle N_{1,\,t+1}-N_{1,\,t}\right\rangle = \rho - (1-\rho)1\cdot\frac{N_{1,\,t}}{t}$$

 \Re As before, set $N_{1,t} = n_1 t$ and drop expectations



$$n_1(t+1)-n_1t=\rho-(1-\rho)1\cdot\frac{n_1t}{t}$$



$$n_1 = \rho - (1 - \rho)n_1$$

Rearrange:

$$n_1 + (1 - \rho)n_1 = \rho$$



$$n_1 = \frac{\rho}{2-\rho}$$

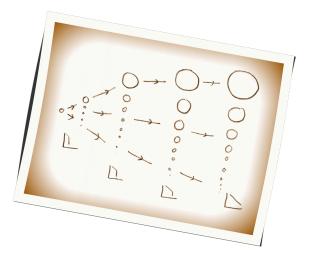
So...
$$N_{1,t} = n_1 t = \frac{\rho t}{2 - \rho}$$

- \Re Recall number of distinct elephants = ρt .
- Fraction of distinct elephants that are unique (belong to groups of size 1):

$$\frac{1}{\rho t} N_{1,\,t} = \frac{1}{\rho t} \frac{\rho \ell}{2-\rho} = \frac{1}{2-\rho}$$

(also = fraction of groups of size 1)

- \clubsuit For ρ small, fraction of unique elephants $\sim 1/2$
- Roughly observed for real distributions
- $\beta \rho$ increases, fraction increases
- Can show fraction of groups with two elephants $\sim 1/6$
- Model works well for large and small k #awesome



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Optimization

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Optimization

Extra





少 q (~ 28 of 90

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Optimization

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Words:

From Simon [24]:

Estimate $\rho_{\rm est} = \#$ unique words/# all words

For Joyce's Ulysses: $\rho_{\rm est} \simeq 0.115$

N_1 (real)	N_1 (est)	N_2 (real)	N_2 (est)
16,432	15,850	4,776	4,870

Evolution of catch phrases:

Yule's paper (1924) [28]:

"A mathematical theory of evolution, based on the conclusions of Dr J. C. Willis, F.R.S."

Simon's paper (1955) [24]:

"On a class of skew distribution functions" (snore)

From Simon's introduction:

It is the purpose of this paper to analyse a class of distribution functions that appear in a wide range of empirical data—particularly data describing sociological, biological and economic phenomena.

Its appearance is so frequent, and the phenomena so diverse, that one is led to conjecture that if these phenomena have any property in common it can only be a similarity in the structure of the underlying probability mechanisms.

Evolution of catch phrases:

Derek de Solla Price:

- First to study network evolution with these kinds of models.
- Citation network of scientific papers
- Price's term: Cumulative Advantage
- A Idea: papers receive new citations with probability proportional to their existing # of citations
- Directed network
- Two (surmountable) problems:
 - 1. New papers have no citations
 - 2. Selection mechanism is more complicated

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References







少 Q (~ 31 of 90

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Optimization

Nutshell

Extra







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Catchphrases

Optimization Assumptions Model

Analysis And the winner is...:

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Evolution of catch phrases:

Robert K. Merton: the Matthew Effect

Studied careers of scientists and found credit flowed disproportionately to the already famous

From the Gospel of Matthew:

"For to every one that hath shall be given... (Wait! There's more....)

but from him that hath not, that also which he seemeth to have shall be taken away. And cast the worthless servant into the outer darkness; there men will weep and gnash their

- (Hath = suggested unit of purchasing power.)
- Matilda effect:
 ✓ women's scientific achievements are often overlooked

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Another analytic approach: [9]

- & Focus on how the *n*th arriving group typically grows.
- Analysis gives:

$$S_{n,t} \sim \left\{ \begin{array}{l} \frac{1}{\Gamma(2-\rho)} \left[\frac{1}{t}\right]^{-(1-\rho)} \text{ for } n=1, \\ \rho^{1-\rho} \left[\frac{n-1}{t}\right]^{-(1-\rho)} \text{ for } n \geq 2. \end{array} \right.$$

- \clubsuit First mover is a factor $1/\rho$ greater than expected.
- \triangle Because ρ is usually close to 0, the first element is truly an elephant in the room.
- Appears that this has been missed for 60 years ...

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Evolution of catch phrases:

Merton was a catchphrase machine:

- 1. Self-fulfilling prophecy
- 2. Role model
- 3. Unintended (or unanticipated) consequences
- 4. Focused interview → focus group

And just to be clear...

Merton's son, Robert C. Merton, won the Nobel Prize for Economics in 1997.

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🗞 See visualization at paper's online app-endices 🗹

"Simon's fundamental rich-gets-richer model entails a dominant first-mover advantage" Dodds et al., Available online at http://arxiv.org/abs/0909.1104, 2016. ^[9] B. $\rho = 0.01$ C. $\rho=0.001$ D. $\rho = 0.0001$ $\rho = 0.01$ $\rho = 0.001$ \log_{10} group number n

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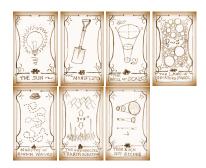


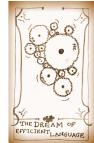


Evolution of catch phrases:

- Barabasi and Albert [2]—thinking about the Web
- Independent reinvention of a version of Simon and Price's theory for networks
- Another term: "Preferential Attachment"
- Considered undirected networks (not realistic but avoids 0 citation problem)
- Still have selection problem based on size (non-random)
- Solution: Randomly connect to a node (easy) ...
- ...and then randomly connect to the node's friends (also easy)
- "Scale-free networks" = food on the table for physicists

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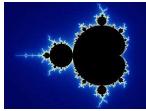
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•> < <> 40 of 90

Benoît Mandelbrot



- Mandelbrot = almond bread
- Bonus Mandelbrot set action: here .

Another approach:

Benoît Mandelbrot

- Communicate as much information as possible for
- \aleph Need measures of information (H) and average
- & Language evolves to maximize H/C, the amount of information per average cost.

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References





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Nutshell

Extra





•9 q (> 43 of 90

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We were born to be Princes of the Universe





Mandelbrot vs. Simon:

- A Mandelbrot (1953): "An Informational Theory of the Statistical Structure of Languages" [17]
- Simon (1955): "On a class of skew distribution functions" [24]
- A Mandelbrot (1959): "A note on a class of skew distribution functions: analysis and critique of a paper by H.A. Simon" [18]
- Simon (1960): "Some further notes on a class of skew distribution functions" [25]

I have no rival, No man can be my equal





Mandelbrot vs. Simon:

- Amandelbrot (1961): "Final note on a class of skew distribution functions: analysis and critique of a model due to H.A. Simon" [20]
- Simon (1961): "Reply to 'final note' by Benoit Mandelbrot" [27]
- Amandelbrot (1961): "Post scriptum to 'final
- 💫 Simon (1961): "Reply to Dr. Mandelbrot's post scriptum" [26]

I am immortal, I have inside me blood of kings

Mandelbrot:

"We shall restate in detail our 1959 objections to Simon's 1955 model for the Pareto-Yule-Zipf distribution. Our objections are valid quite irrespectively of the sign of p-1, so that most of Simon's (1960) reply was irrelevant." [19]

Simon:

"Dr. Mandelbrot has proposed a new set of objections to my 1955 models of the Yule distribution. Like his earlier objections, these are invalid." [27]

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Nutshell

References





•9 q (~ 47 of 90

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Optimization

Nutshell

Extra





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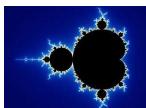
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- Mandelbrot = father of fractals

- Derived Zipf's law through optimization [17]
- & Idea: Language is efficient
- as little cost
- cost (C)...
- \Leftrightarrow Equivalently: minimize C/H.
- Recurring theme: what role does optimization play in complex systems?

The Quickening —Mandelbrot v. Simon:



- Things there should be only one of: Theory, Highlander Films.
- your head (funding remains tight).

Zipfarama via Optimization:

Mandelbrot's Assumptions:

- & Language contains n words: w_1, w_2, \dots, w_n .
- Words appear randomly according to this distribution (obviously not true...)
- Words = composition of letters is important
- Alphabet contains m letters
- Words are ordered by length (shortest first)

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Assumptions

Nutshell





Zipfarama via Optimization:

Total Cost C

- Cost of the *i*th word: $C_i \simeq 1 + \log_m i$
- & Cost of the *i*th word plus space: $C_i \simeq 1 + \log_m(i+1)$
- Simplify base of logarithm:

Zipfarama via Optimization:

Use Shannon's Entropy (or Uncertainty):

Information Measure

occurrence

$$C_i' \simeq \log_m(i+1) = \frac{\log_e(i+1)}{\log_e m} \propto \frac{\text{DD}(i+1)}{}$$

Total Cost:

$$C \sim \sum_{i=1}^n p_i C_i' \propto \sum_{i=1}^n p_i \, \Box \Box (i+1)$$

 $H = -\sum_{i=1}^{n} p_i \log_2 p_i$

🚓 (allegedly) von Neumann suggested 'entropy'...

 $-\log_2 p_i = \log_2 1/p_i$ = minimum number of bits

 \Leftrightarrow If $p_i = 1/2$, need only 1 bit (log₂1/ $p_i = 1$)

\$ If $p_i = 1/64$, need 6 bits ($\log_2 1/p_i = 6$)

needed to distinguish event i from all others

Proportional to average number of bits needed to encode each 'word' based on frequency of

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Rich-Get-Riche

Optimization

Model

Nutshell

References









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Rich-Get-Richer Mechanism Simon's Model Analysis

Optimization

Extra







Zipfarama via Optimization:

Word Cost

- Length of word (plus a space)
- Word length was irrelevant for Simon's method

Objection

Real words don't use all letter sequences

Objections to Objection

- Maybe real words roughly follow this pattern (?)
- Words can be encoded this way
- Na na na-na naaaaa...

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Optimization

Extra







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Zipfarama via Optimization:

Binary alphabet plus a space symbol

Zipfarama via Optimization:

i	1	2	3	4	5	6	7	8
word	1	10	11	100	101	110	111	1000
length	1	2	2	3	3	3	3	4
$1 + \log_{2} i$	1	2	2.58	3	3.32	3.58	3.81	4

- 3 Word length of 2^k th word: $= k + 1 = 1 + \log_2 2^k$
- \bowtie Word length of *i*th word $\simeq 1 + \log_2 i$
- \clubsuit For an alphabet with m letters, word length of *i*th word $\simeq 1 + \log_{m} i$.

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Optimization Assumptions

Nutshell

Extra





•9 q (~ 54 of 90

Information Measure

Use a slightly simpler form:

$$H = -\sum_{i=1}^n p_i \log_e p_i / \log_e 2 = -g \sum_{i=1}^n p_i \, \Box \Box \, p_i$$

where $q = 1/\square \square 2$

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Rich-Get-Riche Optimization

Model

Nutshell

References





少 Q (~ 58 of 90

Zipfarama via Optimization:

Minimize

$$F(p_1,p_2,\dots,p_n)=C/H$$

subject to constraint

$$\sum_{i=1}^{n} p_i = 1$$

A Tension:

- (1) Shorter words are cheaper
- (2) Longer words are more informative (rarer)

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Model

Optimization

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Power-Law Mechanisms, Pt. 2

Rich-Get-Richer

Nutshell

References





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Power-Law Mechanisms, Pt. 2

Rich-Get-Richer

Mechanism

Optimization

Analysis

Extra

Zipfarama via Optimization:

Finding the exponent

Now use the normalization constraint:

$$1 = \sum_{j=1}^n p_j = \sum_{j=1}^n (j+1)^{-H/gC} = \sum_{j=1}^n (j+1)^{-\alpha}$$

- \Re As $n \to \infty$, we end up with $\zeta(H/gC) = 2$ where ζ is the Riemann Zeta Function
- \Leftrightarrow Gives $\alpha \simeq 1.73$ (> 1, too high) or $\gamma = 1 + \frac{1}{\alpha} \simeq 1.58$ (very wild)
- \clubsuit If cost function changes $(j+1 \rightarrow j+a)$ then exponent is tunable
- & Increase a, decrease α

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Rich-Get-Riche

Optimization

Analysis

Nutshell

References





•9 q (~ 63 of 90

Zipfarama via Optimization:

Time for Lagrange Multipliers:

Minimize

$$\Psi(p_1, p_2, \dots, p_n) =$$

$$F(p_1, p_2, \dots, p_n) + \lambda G(p_1, p_2, \dots, p_n)$$

where

$$F(p_1,p_2,\ldots,p_n) = \frac{C}{H} = \frac{\sum_{i=1}^n p_i \, \text{dd}(i+1)}{-g \sum_{i=1}^n p_i \, \text{dd}\, p_i}$$

and the constraint function is

$$G(p_1, p_2, \dots, p_n) = \sum_{i=1}^{n} p_i - 1 (=0)$$

Insert question from assignment 3 2

Zipfarama via Optimization:

All told:

- Reasonable approach: Optimization is at work in evolutionary processes
- But optimization can involve many incommensurate elephants: monetary cost, robustness, happiness,...
- Mandelbrot's argument is not super convincing
- Exponent depends too much on a loose definition of cost

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Rich-Get-Richer Mechanism Simon's Model Analysis

Optimization Analysis And the u

Extra







Zipfarama via Optimization:

Some mild suffering leads to:



$$p_{j} = e^{-1-\lambda H^{2}/gC}(j+1)^{-H/gC} \propto (j+1)^{-H/gC}$$

- $\mbox{\&}$ A power law appears [applause]: $\alpha = H/gC$
- \aleph Next: sneakily deduce λ in terms of g, C, and H.
- Find

$$p_{i} = (j+1)^{-H/gC}$$

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• ୨ ଏ (~ 61 of 90

Rich-Get-Richer

Optimization Analysis

Nutshell

Extra





•9 q (~ 62 of 90

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Rich-Get-Riche

Optimization

Analysis

Nutshell Extra







少 q (~ 65 of 90

From the discussion at the end of Mandelbrot's paper:

- & A. S. C. Ross: "M. Mandelbrot states that 'the actual direction of evolution (sc. of language) is, in fact, towards fuller and fuller utilization of places'. We are, in fact, completely without evidence as to the existence of any 'direction of evolution' in language, and it is axiomatic that we shall remain so. Many philologists would deny that a 'direction of evolution' could be theoretically possible; thus I myself take the view that a language develops in what is essentially a purely random manner."
- A Mandelbrot: "As to the 'fundamental linguistic units being the least possible differences between pairs of utterances' this is a logical consequence of the fact that two is the least integer greater than

Reconciling Mandelbrot and Simon

Numerous efforts...

3. D'Souza et al., 2007:

Scale-free networks [10]

1. Carlson and Doyle, 1999:

Highly Optimized Tolerance

2. Ferrer i Cancho and Solé, 2002:

Zipf's Principle of Least Effort [13]

Mixture of local optimization and randomness

(HOT)—Evolved/Engineered Robustness [5, 6]

More:

PoCS | @pocsvox What Shannon said about meaning in his Power-Law Mechanisms, Pt. 2 1948 paper "A mathematical theory of communication": [23]

The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point. Frequently the messages have meaning; that is they refer to or are correlated according to some system with certain physical or conceptual entities. These semantic aspects of communication are irrelevant to the engineering problem. The significant aspect is that the actual message is one selected from a set of possible messages. The system must be designed to operate for each possible selection, not just the one which will actually be chosen since this is unknown at the time of design.

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Rich-Get-Riche

Optimization

Analysis

Nutshell







•2 0 € 69 of 90

INTRODUCTION

The Psycho-Biology of Language is not calculated to please every taste. Zipf was the kind of man who would take roses apart to count their petals; if it violates your sense of values to tabulate the different words in a Shakespearean sonnet, this is not a book for you. Zipf took a scientist's view of language - and for him that meant the statistical analysis of language — and for him that meant the statistical analysis of language as a biological, sychological, social process. If such analysis repels you, then leave your language alone and avoid George Kingsley Zipf like the plague. You will be much happier reading Mark Twain: "There are liars, damned liars, and statisticians." Or W. H. Auden: "Thou shalt not sit with statisticians nor commit a

However, for those who do not flinch to see beauty murthought that in the very heart of all the freedom language

as the law of gravitation?

social science.'

dered in a good cause, Zipf's scientific exertions yielded some wonderfully unexpected results to boggle the mind and tease the imagination. Language is - among other things — a biological, psychological, social process; to apply sta-tistics to it merely acknowledges its essential unpredictability, without which it would be useless. But who would have allows us Zipf would find an invariant as solid and reliable

More Other mechanisms:

Much argument about whether or not monkeys. typing could produce Zipf's law... (Miller, 1957) [21]

A Miller gets to slap Zipf rather rudely in an introduction to a 1965 reprint of Zipf's "Psycho-biology of Language" [22, 29]

Let us now slap Miller around by simply reading his words out (see next slides):



Side note: Miller mentions "Genes of Language."

Still fighting: "Random Texts Do Not Exhibit the Real Zipf's Law-Like Rank Distribution" [12] by Ferrer-i-Cancho and Elvevåg, 2010.

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•9 a (~ 67 of 90

Rich-Get-Richer

Optimization

Analysis

Nutshell

References

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ჟ q ભ 66 of 90

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Power-Law Mechanisms, Pt. 2

Rich-Get-Richer

Mechanism

Optimization

Analysis

Extra

Rich-Get-Richer

Optimization Analysis

Nutshell Extra

References





•∩ a ∩ 68 of 90

Put it this way. Suppose that we acquired a dozen monkeys and chained them to typewriters until they had produced some very long and random sequence of characters. Suppose further that we defined a "word" in this monkeytext as any sequence of letters occurring between successive spaces. And suppose finally that we counted the occurrences of these "words" in just the way Zipf and others counted the occurrences of real words in meaningful texts. When we plot our results in the same manner, we will find exactly the same "Zipf curves" for the monkeys as for the human authors. Since we are not likely to argue that the poor monkeys were searching for some equilibrium between uniformity and diversity in expressing their ideas, such explanations seem equally inappropriate for human authors.

A mathematical rationalization for this result has been provided by Benoit Mandelbrot. The crux of it is that if we assume that word-boundary markers (spaces) are scattered randomly through a text, then there will necessarily be more occurrences of short than long words. Add to this fact the further observation that the variety of different words available increases exponentially with their length and the phenomenon Zipf reported becomes inescapable: a few short words will be used an enormous number of times while a vast number of longer words will occur infrequently or not at all.

So Zipf was wrong. His facts were right enough, but not

while a vast number of longer words will occur infrequently or not at all.

So Zipf was wrong. His facts were right enough, but not his explanations. In a broader sense he was right, however, for he called attention to a stochastic process that is frequently seen in the social sciences, and by accumulating statistical data that cried out for some better explanation be challenged his colleagues and his successors to explore an important new type of probability distribution. Zipf belongs among those rare but stimulating men whose failures are more profitable than most men's successes.

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Rich-Get-Richer Mechanism

Optimization

Analysis

Extra





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Rich-Get-Riche Mechanisn Simon's Model Analysis Words

Optimization

Analysis

Nutshell Extra









•9 q ← 71 of 90

So who's right?

Bornholdt and Ebel (PRE), 2001:

"World Wide Web scaling exponent from Simon's 1955 model" [4].

- Show Simon's model fares well.
- \aleph Recall ρ = probability new flavor appears.
- 🚓 Alta Vista 🗹 crawls in approximately 6 month period in 1999 give $\rho \simeq 0.10$
- \Leftrightarrow Leads to $\gamma=1+\frac{1}{1-\rho}\simeq 2.1$ for in-link distribution.
- \clubsuit Cite direct measurement of γ at the time: 2.1 ± 0.1 and 2.09 in two studies.

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Rich-Get-Richer

Optimization

Analysis Nutshell

References





•9 q (~ 72 of 90

So who's right?

So who's right?

apparent.

Replication models.

Others are also not happy:

Krugman and Simon

Nutshell:

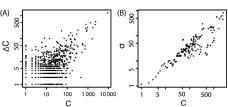


FIG. 2. Left panel: Plots of ΔC versus C from the Etch release (15.08.2007) to the latest Lenny version (05.05.2008) in double logarithmic scale. Only positive values are displayed. The linear regression $\Delta C = R \times C + C_0$ is significant at the 95% confidence level, with a small value $C_0 = 0.3$ at the origin and R =0.09. Right panel: same as left panel for the standard deviation of

Rough, approximately linear relationship between C number of in-links and ΔC .

Simonish random 'rich-get-richer' models agree in

Power-lawfulness: Mandelbrot's optimality is still

Optimality arises for free in Random Competitive

"The Self-Organizing Economy" (Paul Krugman,

Substantial work done by Urban Geographers

🚳 "Déjà vu, Mr. Krugman" (Berry, 1999)

& Krugman touts Zipf's law for cities, Simon's model

detail with empirical observations.

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Power-Law Mechanisms, Pt. 2

Rich-Get-Riche

Ш

Optimization

Nutshell

References







2 9 0 76 of 90

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Rich-Get-Richer Mechanism Simon's Model Analysis

Optimization

Nutshell

Extra







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Rich-Get-Riche

Optimization

Nutshell

Extra

References

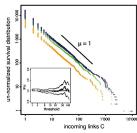




少 Q (~ 79 of 90

So who's right?

Recent evidence for Zipf's law...



So who's right?

incoming links C

Source Linux Distribution" [16]

"Empirical Tests of Zipf's Law Mechanism in Open

Maillart et al., PRL, 2008:

FIG. 1 (color online). (Color Online) Log-log plot of the number of packages in four Debian Linux Distributions with bibiting the color of the color of the color of the color of the Substributions are Woody (1907.2002) (corange diamonats), Sarge (06.06.2005) (green crosses). Eich (15.08.2007) (blue circles), Lenny (15.22.2007) (black-5). The intest shows the maximum likelihood estimate (MLE) of the exponent µ together with two boundaries defining in 95% confidence interval (ap-with two boundaries defining in 95% confidence interval (ap-

FIG. 1 (color online). (Color Online) Log-log plot of the number of packages in four Deban Llnux Distributions with more than C indirected links. The four Deban Llnux for the color of the color of the color of the color Sage (06,06.205) (green crosses). Each (15,08.2007) (but circles), Lenny (15,12.2007) (black-5). The intest shows the maximum likelihood estimate (MLE) of the exponent µ together with two boundaries defining its 95% confidence interval (ap-with two boundaries defining its 95% confidence interval (ap-

Maillart et al., PRL, 2008: "Empirical Tests of Zipf's Law Mechanism in Open Source Linux Distribution" [16]

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Rich-Get-Richer Mechanism

Optimization

Nutshell

Extra





•9 q (> 74 of 90

Power-Law Mechanisms, Pt. 2

Optimization

Nutshell





少 Q (~ 75 of 90

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Rich-Get-Richer

Extra References

Who needs a hug?

From Berry [3]

- 🚵 Déjà vu, Mr. Krugman. Been there, done that. The Simon-Ijiri model was introduced to geographers in 1958 as an explanation of city size distributions, the first of many such contributions dealing with the steady states of random growth processes, ...
- But then, I suppose, even if Krugman had known about these studies, they would have been discounted because they were not written by professional economists or published in one of the top five journals in economics!

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Rich-Get-Richer

Optimization

Nutshell

Extra References





References II

- [5] J. M. Carlson and J. Doyle. Highly optimized tolerance: A mechanism for power laws in designed systems. Phys. Rev. E, 60(2):1412-1427, 1999. pdf
- [6] I. M. Carlson and I. Doyle. Complexity and robustness. Proc. Natl. Acad. Sci., 99:2538-2545, 2002. pdf
- D. J. de Solla Price. Networks of scientific papers. Science, 149:510-515, 1965. pdf
- D. J. de Solla Price. A general theory of bibliometric and other cumulative advantage processes. J. Amer. Soc. Inform. Sci., 27:292–306, 1976. pdf

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Rich-Get-Riche

Optimization

Nutshell

References







少 q (~ 83 of 90

Who needs a hug?

From Berry [3]

- ... [Krugman] needs to exercise some humility, for his world view is circumscribed by folkways that militate against recognition and acknowledgment of scholarship beyond his disciplinary frontier.
- Urban geographers, thank heavens, are not so afflicted.

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Rich-Get-Richer Mechanism

Optimization

Nutshell

Extra References





References III

- [9] P. S. Dodds, D. R. Dewhurst, F. F. Hazlehurst, C. M. Van Oort, L. Mitchell, A. J. Reagan, J. R. Williams, and C. M. Danforth. Simon's fundamental rich-gets-richer model entails a dominant first-mover advantage, 2016. Available online at http://arxiv.org/abs/0909.1104.pdf
- [10] R. M. D'Souza, C. Borgs, J. T. Chayes, N. Berger, and R. D. Kleinberg. Emergence of tempered preferential attachment from optimization. Proc. Natl. Acad. Sci., 104:6112–6117, 2007. pdf
- [11] J.-B. Estoup. Gammes sténographiques: méthode et exercices pour l'acquisition de la vitesse.

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Rich-Get-Richer Mechanism Simon's Model Analysis Catchphrase

Optimization

Nutshell

Extra References

PoCS





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References I

- [1] F. Auerbach. Das gesetz der bevölkerungskonzentration. Petermanns Geogr. Mitteilungen, 59:73–76, 1913.
- A.-L. Barabási and R. Albert. Emergence of scaling in random networks. Science, 286:509–511, 1999. pdf <a>C
- B. J. L. Berry. [3] Déjà vu, Mr. Krugman. Urban Geography, 20:1–2, 1999. pdf <a>C̄
- [4] S. Bornholdt and H. Ebel. World Wide Web scaling exponent from Simon's 1955 model. Phys. Rev. E, 64:035104(R), 2001. pdf

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Rich-Get-Richer

Optimization Mandelbrot vs. Simo

Nutshell

Extra References





•9 q (~ 82 of 90

References IV

[12] R. Ferrer-i Cancho and B. Elvevåg. Random texts do not exhibit the real Zipf's law-like rank distribution. PLoS ONE, 5:e9411, 03 2010.

Institut Sténographique, 1916.

- [13] R. Ferrer-i Cancho and R. V. Solé. Zipf's law and random texts. Advances in Complex Systems, 5(1):1-6, 2002.
- [14] P. Krugman. The Self-Organizing Economy. Blackwell Publishers, Cambridge, Massachusetts, 1996.

Power-Law Mechanisms, Pt. 2

Rich-Get-Riche

Optimization

Nutshell

Extra

References





少 q (~ 85 of 90

References V

[15] A. J. Lotka.

The frequency distribution of scientific productivity.

Journal of the Washington Academy of Science, 16:317–323, 1926.

[16] T. Maillart, D. Sornette, S. Spaeth, and G. von Krogh.

Empirical tests of Zipf's law mechanism in open source Linux distribution.

Phys. Rev. Lett., 101(21):218701, 2008. pdf

[17] B. B. Mandelbrot.

An informational theory of the statistical structure of languages.

In W. Jackson, editor, <u>Communication Theory</u>, pages 486–502. Butterworth, Woburn, MA, 1953. pdf 🗗

References VI

[18] B. B. Mandelbrot.

A note on a class of skew distribution function. Analysis and critique of a paper by H. A. Simon. Information and Control, 2:90–99, 1959.

[19] B. B. Mandelbrot.

Final note on a class of skew distribution functions: analysis and critique of a model due to H. A. Simon.

Information and Control, 4:198–216, 1961.

[20] B. B. Mandelbrot.

Post scriptum to 'final note'.

Information and Control, 4:300–304, 1961.

References VII

[21] G. A. Miller.

Some effects of intermittent silence.

American Journal of Psychology, 70:311–314, 1957. pdf 2

[22] G. A. Miller.

Introduction to reprint of G. K. Zipf's "The Psycho-Biology of Language." MIT Press, Cambridge MA, 1965. pdf

[23] C. E. Shannon.

A mathematical theory of communication. The Bell System Tech. J., 27:379–423,623–656, 1948. pdf

[24] H. A. Simon.

On a class of skew distribution functions. Biometrika, 42:425–440, 1955. pdf 🗗

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Rich-Get-Richer Mechanism Simon's Model Analysis

Optimization Minimal Cost

Assumptions Model Analysis

And the winner is...
Nutshell

Evtra

References





•9 q (~ 86 of 90

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Rich-Get-Richer Mechanism Simon's Model Analysis

Optimization
Minimal Cost
Mandelbrot vs. Simon
Assumptions
Model

Nutshell

Extra

References





少∢(~ 87 of 90

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Rich-Get-Richer Mechanism Simon's Model Analysis Words

Optimization Minimal Cost

Assumptions Model Analysis

Nutshell Extra

References





少∢(~ 88 of 90

References VIII

[25] H. A. Simon.

Some further notes on a class of skew distribution functions.

Information and Control, 3:80-88, 1960.

[26] H. A. Simon.

Reply to Dr. Mandelbrot's post scriptum. Information and Control, 4:305–308, 1961.

[27] H. A. Simon.

Reply to 'final note' by Benoît Mandelbrot. Information and Control, 4:217–223, 1961.

[28] G. U. Yule.

A mathematical theory of evolution, based on the conclusions of Dr J. C. Willis, F.R.S.

Phil. Trans. B, 213:21-87, 1925. pdf 2

References IX

[29] G. K. Zipf. The Psycho-Biology of Language. Houghton-Mifflin, New York, NY, 1935.

[30] G. K. Zipf.

Human Behaviour and the Principle of Least-Effort.

Addison-Wesley, Cambridge, MA, 1949.

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Rich-Get-Richer Mechanism Simon's Model Analysis

Catchphrases
Optimization

Mandelbrot vs. Simo Assumptions Model

Analysis And the winner is...?

Nutshell

EXTra

References







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Rich-Get-Richer Mechanism Simon's Model Analysis Words

Optimization Minimal Cost Mandelbrot vs. Simon Assumptions Model

Nutshell

Extra

References



