

**KOLMOGOROV COMPLEXITY AS ENERGY:
COMPUTER SCIENCE vs PHYSICS**

Yuri I. Manin

PLAN

PART I: ZIPF'S LAW

PART II: ERROR-CORRECTING CODES AND THEIR
ASYMPTOTIC BOUNDS

PART III: KOLMOGOROV COMPLEXITY AS ENERGY

PART IV: COMPUTABILITY AND MATHEMATICS
OF KOLMOGOROV COMPLEXITY

PART V: CAN WE COMPUTE ASYMPTOTIC BOUNDS?

PART I: ZIPF'S LAW

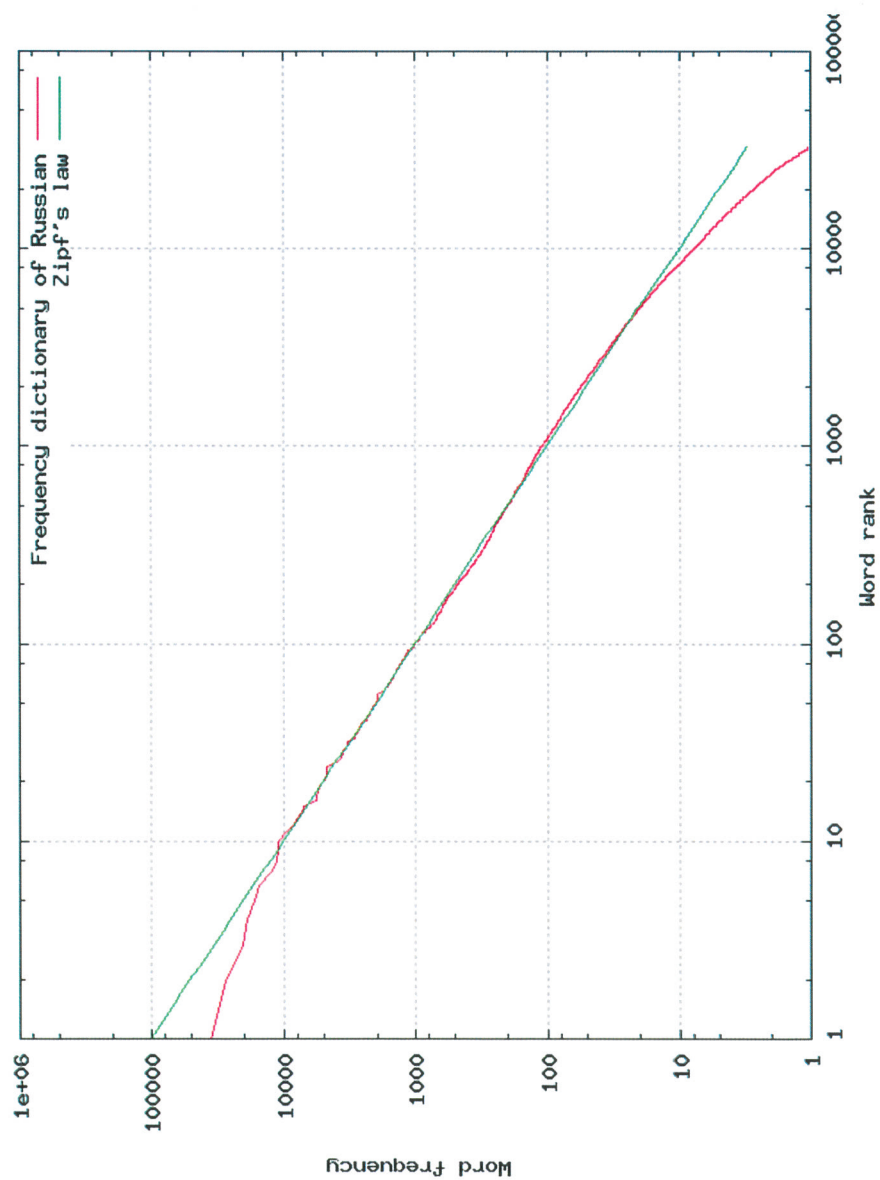
- Consider a corpus of texts in a given language, make the list of words occurring in them and the numbers of occurrences. Range these words in the order of *diminishing* frequencies. Define the Zipf rank of a word as its number in this ordering.

- Zipf's Law :

FREQUENCY

IS INVERSELY PROPORTIONAL TO THE RANK

PICTURE:
Zipf's distribution of Russian words(logarithmic scale)



- Universality of Zipf's law : the law is empirically observed in very different databases, that allow one to calculate frequency of occurrence of certain *patterns* ("words") in certain *massifs of data*.
- Example on the next page: patterns in *financial audit data*.



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An investigation of Zipf's Law for fraud detection (DSS#06-10-1826R(2))

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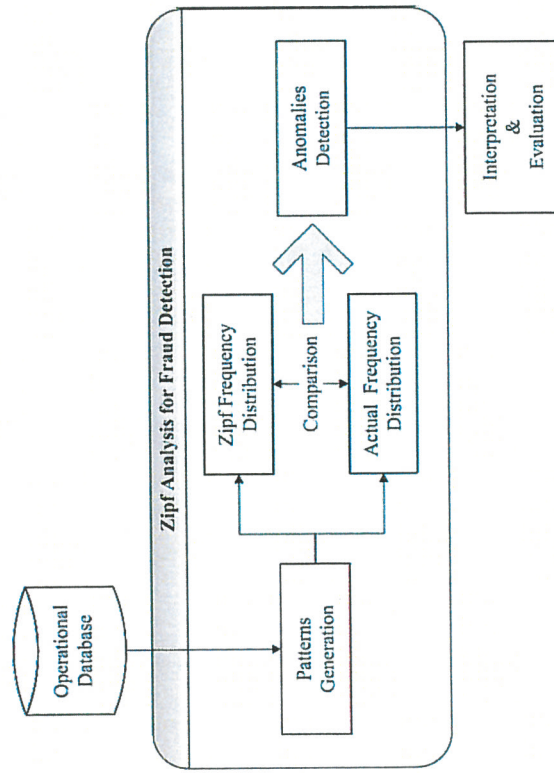


Fig. 1. Fraud detection model of Zipf Analysis.

PART II

ERROR-CORRECTING CODES
AND THEIR ASYMPTOTIC BOUNDS

• BASIC NOTATION:

Alphabet A := a finite set of cardinality $q \geq 2$.

Code $C \subset A^n$:= a subset of words of length n .

Hamming distance between two words:

$$d((a_i), (b_i)) := \text{card}\{i \in (1, \dots, n) \mid a_i \neq b_i\}.$$

Code parameters: cardinality of the alphabet q and

$$n(C) := n, \quad k(C) := k := \lfloor \log_q \text{card}(C) \rfloor,$$

$$d(C) := d = \min \{d(a, b) \mid a, b \in C, a \neq b\}.$$

Relative distance and Transmission rate :

$$\delta(C) := \frac{d(C)}{n(C)}, \quad R(C) = \frac{k(C)}{n(C)}.$$

Briefly, C is an $[n, k, d]_q$ -code.

SOURCE DATA*Encoding:*

A sequence
of code words

Noisy channel:

Sequence of
(corrupted) code words

Error correction:

(Ideally) sequence of
initial code words

Decoding:**TRANSMITTED DATA**

Examples: Morse code and Barcodes

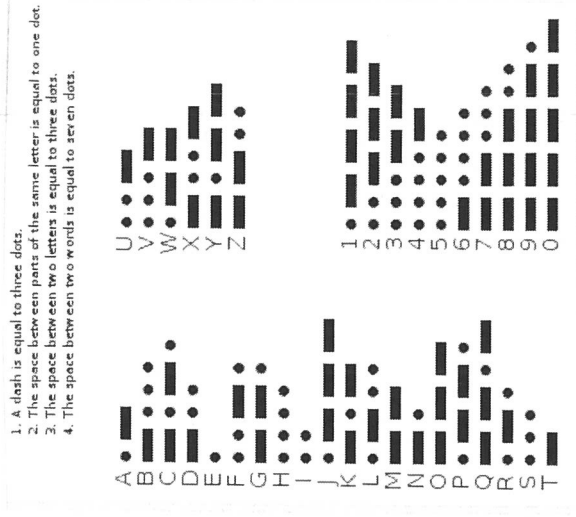
Samuel F. B. Morse
(from 1836 on)

Alphabet: {dash, dot, space},

$q = 3$.

Block length: $n = 7$

$d = ?$: (*Exercise*)



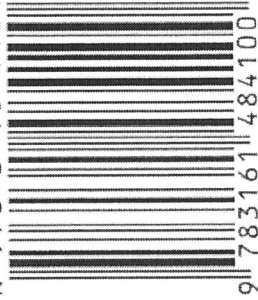
Norman J. Woodland

(from 1949 on):

“His [...]inspiration came from Morse code, and he formed his first barcode from sand on the beach.

I just extended the dots and dashes downwards and made narrow lines and wide lines out of them”

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A 13-digit ISBN, 978-3-16-148410-0, as represented by an EAN-13 bar code.

- Explaining terms:

Minimal relative distance and Transmission rate :

$$\delta(C) := \frac{d(C)}{n(C)}, \quad R(C) = \frac{[k(C)]}{n(C)}.$$

Minimal relative distance must match channel's noisiness: probability of corruption of one letter.

Transmission rate is the share of meaningful (code) words; their number must be maximized for any given relative distance.

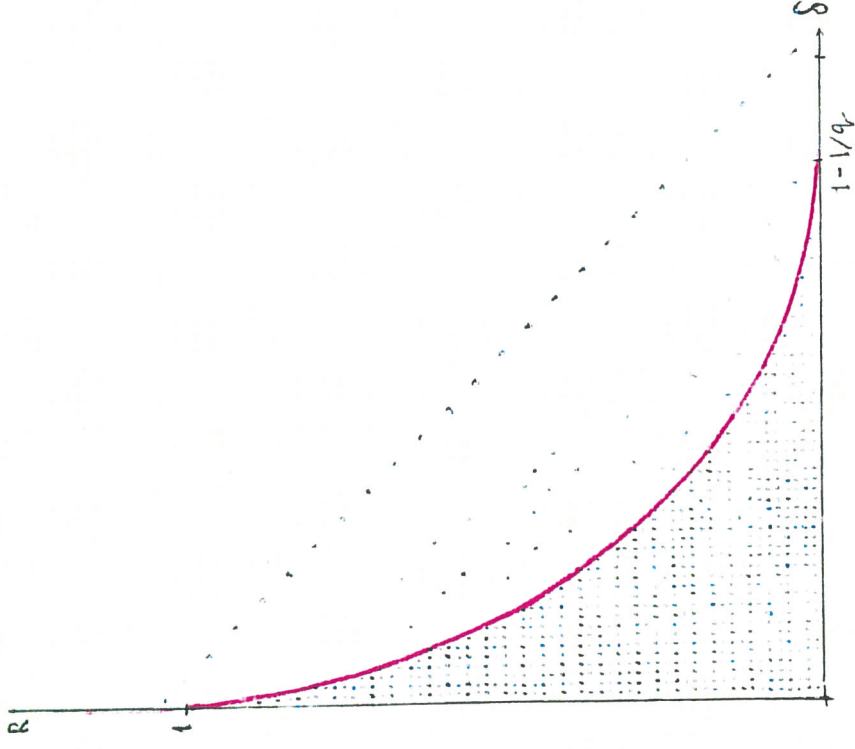
- A good code must maximize minimal relative distance when the transmission rate is chosen.
- One more property of good codes: they must admit *efficient algorithms of encoding, decoding and error-correction*.

How this can be achieved: consider structured codes.
Typical choice:

- Linear codes $:=$ linear subspaces of $\mathbb{F}_q^n$.

- Code points:

$$[n, k, d]_q - \text{code } C \mapsto P_C := (R(C), \delta(C)) = \left(\frac{[k(C)]}{n(C)}, \frac{d(C)}{n(C)} \right)$$



How a finite pixel plot of all code points
might look (q fixed)

Explanations to the picture :

- **DEFINITION.** *Multiplicity* of a code point is the number of codes that project onto it.
- **THEOREM** (Yu.M., 1981 + 2011). There exists such a continuous function $\alpha_q(\delta)$, $\delta \in [0, 1]$, that
 - (i) The set of code points of infinite multiplicity is exactly the set of rational points $(R, \delta) \in [0, 1]^2$ satisfying $R \leq \alpha_q(\delta)$.
 The curve $R = \alpha_q(\delta)$ is called the asymptotic bound.

- (ii) Code points x of finite multiplicity all lie above the asymptotic bound and are called isolated ones:
for each such point there is an open neighborhood containing x as the only code point.
- (iii) The same statements are true for linear codes, with, a possibly, different asymptotic bound $R = \alpha_q^{lin}(\delta)$.

PART III: KOLMOGOROV COMPLEXITY AS ENERGY

IIIa. AN INTUITIVE DESCRIPTION

- Logarithmic Kolmogorov complexity of a combinatorial object ω is defined as

*the measure of compressibility of ω :=
the length of the shortest program that can generate ω*

- **A representative example:**

Combinatorial object = an integer $N > 0$, represented by N written dashes.

Then the length of its compressed description is $\leq c \log N$, and its exponential Kolmogorov complexity is $\leq C N$.

IIIb. ZIPIF'S RANK AND ZIPIF'S LAW FROM COMPLEXITY

• I suggest that (at least in some situations) Zipf's law emerges as the combined effect of two factors:

- (A) Rank ordering coincides with the ordering with respect to the growing (exponential) Kolmogorov complexity $K(w)$ up to a factor $\exp(O(1))$.
- (B) The probability distribution producing Zipf's law is (an approximation to) the L . Levin a priori distribution.

If we accept (A) and (B), then Zipf's law follows from two basic properties of Kolmogorov complexity:

- (a) rank of w defined according to (A) is $\exp(O(1)) \cdot K(w)$.
- (b) L. Levin's a priori distribution assigns to an object w probability $\sim KP(w)^{-1}$ where KP is the exponentiated prefix Kolmogorov complexity, and we have, up to $\exp(O(1))$ -factors,

$$K(w) \preceq KP(w) \preceq K(w) \cdot \log^{1+\varepsilon} K(w)$$

with arbitrary $\varepsilon > 0$.

- NB A probability distribution on infinity of objects cannot be constructed directly from K : the series $\sum_m K(m)^{-1}$ diverges. However, on finite sets of data the small discrepancy is additionally masked by the dependence of both K and KP on the choice of an optimal encoding.

- **COMPLEXITY AS EFFORT.** The picture described above agrees with Zipf's motto "minimization of effort", but reinterprets the notion of effort: its role is now played by the logarithm of the Kolmogorov complexity that is, by the length of the maximally compressed description of an object.

Such a picture makes sense especially if the objects satisfying Zipf's distribution, are *generated* rather than simply *observed*.

IIIc. ASYMPTOTIC BOUNDS FROM COMPLEXITY

• Oracle assisted approximate computation of the asymptotic bound.

– The set $Codes_q$ of all q -ary codes in a fixed alphabet A is a constructive world.

– CLAIM. If an oracle produces for us elements of $Codes_q$ in their Kolmogorov order, then we can write an oracle assisted algorithm that for each “pixel size” N^{-1} enumerates all code points of the form

$$(k/N, d/N), \quad a, d \in \mathbf{Z}_+$$

CF. PICTURE ON PAGE 13

• Partition function for codes involving complexity.

– The function $\alpha_q(\delta)$ is continuous and strictly decreasing for $\delta \in [1, 1 - q^{-1})$.

Hence the limit points domain $R \leq \alpha_q(\delta)$ can be equally well described by the inequality $\delta \leq \beta_q(R)$ where β_q is the function inverse to α_q .

– Fix an $R \in \mathbf{Q} \cap (0, 1)$. For $\Delta \in \mathbf{Q} \cap (0, 1)$, put

$$Z(R, \Delta; \beta) := \sum_{C: R(C)=R, 1-\Delta \leq \delta(C) \leq 1} K_u(C)^{-\beta + \delta(C) - 1},$$

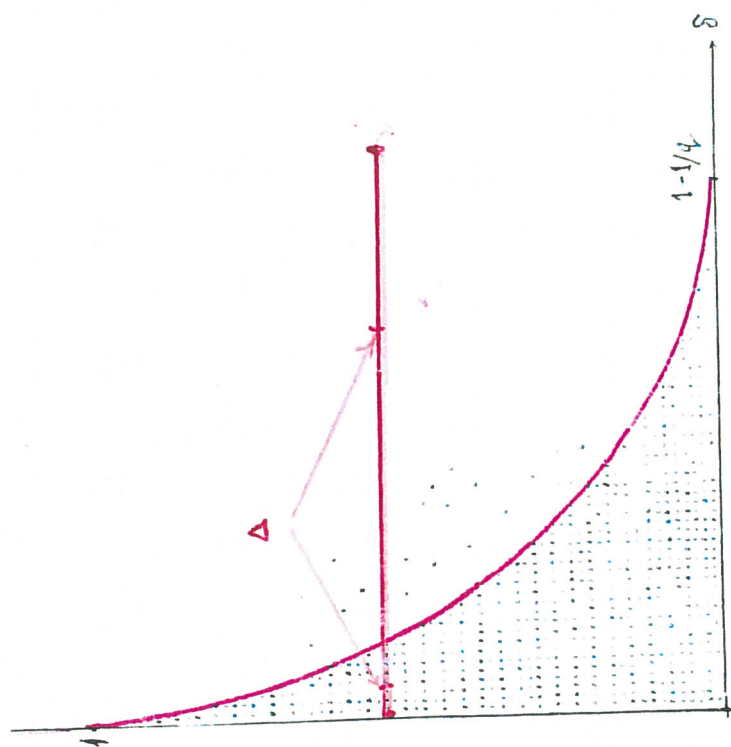
where K_u is an exponential Kolmogorov complexity on $Codes_q$.

• Theorem. (i) If $\Delta > \beta_q(R)$, then $Z(R, \Delta; \beta)$ is a real analytic function of β .

(ii) If $\Delta < \beta_q(R)$, then $Z(R, \Delta; \beta)$ is a real analytic function of β for $\beta > \beta_q(R)$ such that its limit for $\beta - \beta_q(R) \rightarrow +0$ does not exist.

• Thermodynamical analogies.

- The argument β of the partition function corresponds to the inverse temperature.
- The transmission rate R corresponds to the density ρ .
- Our asymptotic bound transported into $(T = \beta^{-1}, R)$ -plane as $T = \beta_q(R)^{-1}$ becomes the phase transition boundary in the (temperature, density)-plane.



PART IV: COMPUTABILITY AND MATHEMATICS OF KOLMOGOROV COMPLEXITY

IVa. A BRIEF GUIDE TO COMPUTABILITY

- Three descriptions of partial recursive functions.

NB A “function”, say, $f: X \rightarrow Y$, below always means a pair $(f, D(f))$, where $D(f) \subset X$ and $f: D(f) \rightarrow Y$. The definition domain is not always mentioned explicitly. If $D(f) = X$, the function will be called “total”; generally it may be called “partial” one.

• Intuitive description. A function $f : Z_+^m \rightarrow Z_+^n$ is (partial) recursive iff it is “semi-computable” in the following sense:

there exists an algorithm F accepting as inputs vectors $x = (x_1, \dots, x_m) \in Z_+$ with the following properties:

- if $x \in D(f)$, F produces as output $f(x)$.
- if $x \notin D(f)$, F either produces answers “NO”, or works indefinitely long without producing any output.

- Formal description (sketch). It starts with two lists:
 - An explicit list of “obviously” semi-computable elementary functions such as constant functions, projections onto i -th coordinate etc.
 - An explicit list of elementary operations, such as an inductive definition, that can be applied to several semi-computable functions and “obviously” produces from them a new semi-computable function.
 - After that, the set of partial recursive functions is defined as the minimal set of functions containing all elementary functions and closed wrt all elementary operations.

- Diophantine description (A DIFFICULT THEOREM).
A function $f : \mathbf{Z}_+^m \rightarrow \mathbf{Z}_+^n$ is partial recursive iff there is a polynomial

$$P(x_1, \dots, x_m; y_1, \dots, y_n; t_1, \dots, t_q) \in \mathbf{Z}[x, y, t]$$

such that the graph

$$\Gamma_f := \{(x, f(x))\} \subset \mathbf{Z}_+^m \times \mathbf{Z}_+^n$$

is the projection of the subset $P = 0$ in $\mathbf{Z}_+^m \times \mathbf{Z}_+^n \times \mathbf{Z}_+^q$.

IVb. CONSTRUCTIVE WORLDS

- An (infinite) constructive world is a countable set X (usually of some finite Bourbaki structures) given together with a class of structural numberings : intuitively computable bijections $\nu : \mathbb{Z}_+ \rightarrow X$ which form a principal homogeneous space over the group of totally recursive permutations of $\mathbb{Z}_+$.

• CHURCH'S THESIS: Let X , Y be two constructive worlds, $\nu_X : \mathbf{Z}_+ \rightarrow X$ $\nu_Y : \mathbf{Z}_+ \rightarrow X$ their structural numberings, and F an (intuitive) algorithm that takes as input an object $x \in X$ and produces an object $F(x) \in Y$ whenever x lies in the domain of definition of F .

Then $f := \nu_Y^{-1} \circ F \circ \nu_X : \mathbf{Z}_+ \rightarrow \mathbf{Z}_+$ is a partial recursive function.

KOLMOGOROV COMPLEXITY

- Kolmogorov complexity and Kolmogorov order.
 - Let X be a constructive world. For any (semi)–computable function $u : \mathbf{Z}_+ \rightarrow X$, the (exponential) complexity of an object $x \in X$ relative to u is

$$K_u(x) := \min \{m \in \mathbf{Z}_+ \mid u(m) = x\}.$$

If such m does not exist, we put $K_u(x) = \infty$.

- **CLAIM:** there exists such u (“an optimal Kolmogorov numbering”, or “decompressor”) that for each other v , some constant $c_{u,v} > 0$, and all $x \in X$,

$$K_u(x) \leq c_{u,v} K_v(x).$$

This $K_u(x)$ is called Kolmogorov complexity of x .

- A Kolmogorov order of a constructive world X is a bijection $K = K_u : X \rightarrow \mathbb{Z}$ arranging elements of X in the increasing order of their complexities K_u .

• WARNINGS :

- Any optimal numbering is only partial function, and its definition domain is not decidable.
- Kolmogorov complexity K_u itself is not computable. It is the lower bound of a sequence of computable functions.
- Kolmogorov order of $\mathbb{Z}_+$ cardinally differs from the natural order in the following sense: it puts in the initial segments very large numbers that can be at the same time Kolmogorov simple.

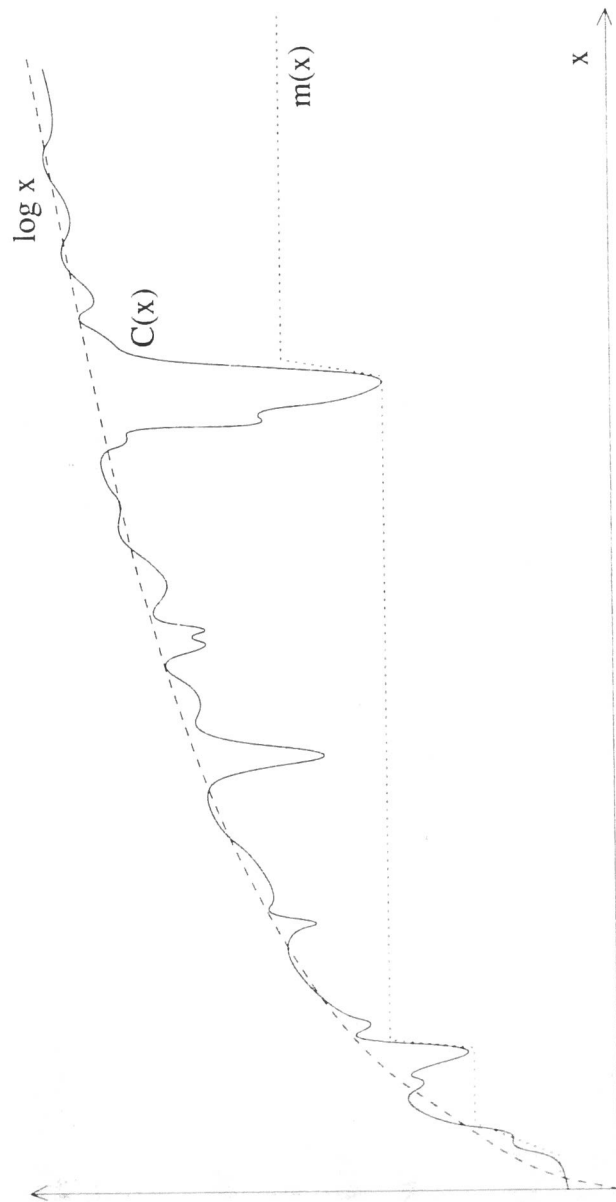
– Example: let $a_n := n^{n^{\dots^n}}$ (n times).

Then $K_u(a_n) \leq cn$ for some $c > 0$.

- FRACTALITY

- The (graph of) logarithmic complexity has very rich self-similarity properties.

Arguably, they explain the universality of Zipf's Law coming from complexity.



THE GRAPH OF THE KOLMOGOROV COMPLEXITY

PART V:
CAN WE COMPUTE ASYMPTOTIC BOUNDS?

- What is known :
 - Many upper and lower bounds for asymptotic bounds α_q and $\alpha_q^{lin} \leq \alpha_q$.
 - Some isolated codes.

- What is NOT known :
 - any explicit formula for α_q or α_q^{lin} .
 - any single value of $\alpha_q(\delta)$ or $\alpha_q^{lin}(\delta)$ for $0 < \delta < 1 - q^{-1}$.
 - any method of approximate computation of $\alpha_q(\delta)$ or $\alpha_q^{lin}(\delta)$.
 - Clearly, $\alpha_q^{lin} \leq \alpha_q$. Is this inequality strict somewhere?

SOME ESTIMATES

- Statistical lower estimates : Gilbert–Varshamov bounds.
 - Most unstructured q -ary codes lie lower or only slightly above the curve

$$R = \frac{1}{2}(1 - H_q(\delta)),$$

$$H_q(\delta) = \delta \log_q(q - 1) - \delta \log_q \delta - (1 - \delta) \log_q(1 - \delta).$$

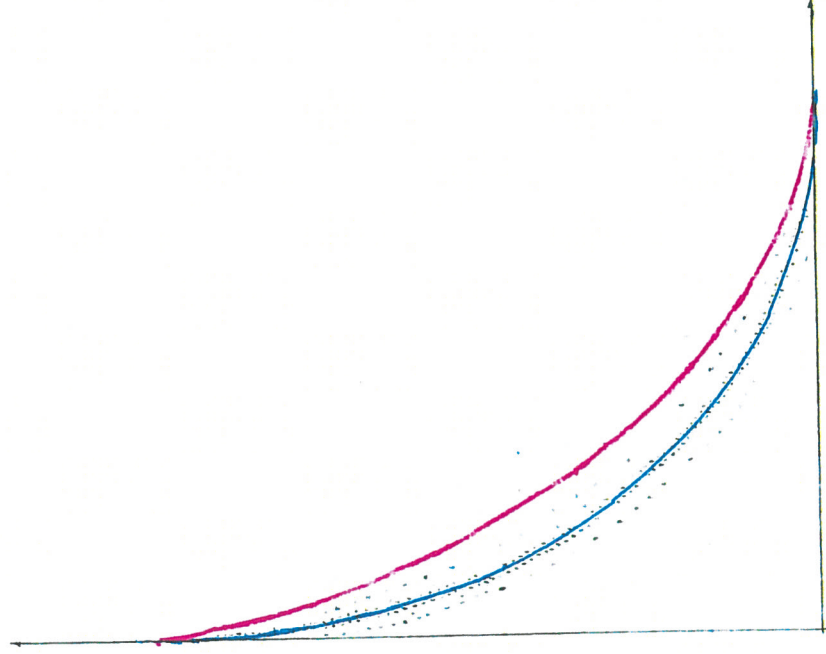
- Most linear q -ary codes lie lower or only slightly above the Gilbert–Varshamov bound

$$R = 1 - H_q(\delta).$$

- In particular,

$$\alpha_q(R) \geq \alpha_q^{lin}(R) \geq 1 - H_q(\delta)$$

Can we see the asymptotic bound
plotting the set of (linear) code points of bounded size?



NO, we will see a cloud of points
concentrating near the Varshamov–Gilbert bound

- Combinatorial upper estimate : Singleton bound:

$$R(C) + \delta(C) \leq 1 + \frac{1}{n(C)}$$

- Corollary:

$$\alpha_q(\delta) \leq 1 - \delta.$$

- Some isolated (linear) codes (Reed – Solomon) :
 - **Parameters:** $1 \leq k \leq n \leq q, d = n + 1 - k$
 - **Code** $C \in \mathbf{F}_q^n$:
choose pairwise distinct $x_1, \dots, x_n \in \mathbf{F}_q$,
embed the space of polynomials $f(x) \in \mathbf{F}_q[x]$ of degree
 $\leq k - 1$ into $\mathbf{F}_q^n$ by

$$f \mapsto (f(x_1), \dots, f(x_n)) \in \mathbf{F}_q^n.$$

• Generalization : algebraic geometric codes.

– Input data: a complete smooth algebraic curve X over $\mathbf{F}_q$, a line bundle L on this curve, and a choice of generating vector in each geometric fiber $L(x)$, $x \in X(\mathbf{F}_q)$.

– The resulting linear code C : image of $\Gamma(X, L)$ wrt the map

$$\Gamma(X, L) \rightarrow \bigoplus_{x \in X(\mathbf{F}_q)} L(x) \cong \mathbf{F}_q^{\text{card} X(\mathbf{F}_q)}$$

– Example : Reed–Solomon codes essentially correspond to $X = \mathbf{P}^1$, $L = \mathcal{O}(k-1)$.

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THANK YOU FOR YOUR ATTENTION!