Experimental Computation and Visual Theorems: The Computer as Collaborator

Jonathan Borwein FRSC FAAS FAA FBAS

(With Aragón, Bailey, P. Borwein, Skerritt, Straub, Tam, Wan, Zudilin, ...)





Centre for Computer Assisted Research Mathematics and its Applications The University of Newcastle, Australia





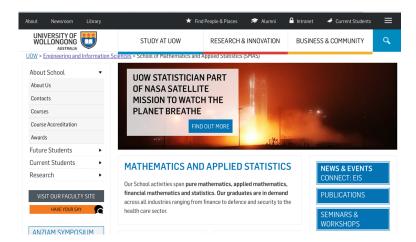




http://carma.newcastle.edu.au/meetings/evims/ http://www.carma.newcastle.edu.au/jon/visuals-ext-abst.pdf

For 2015 Presentations

Prepared for Wollongong



Dedicated to Jacques Hadamard, A Universal Mathematician (1998)



"The object of mathematical rigor is to sanction and legitimize the conquests of intuition, and there was never any other object for it."-JSH (1865-1963)

last dozen of the first hundred of his year", said at the celebration of Hadamard's centenary:

The taupin who saw Jacques Hadamard enter the lecture theatre, found a teacher who was active, alive, whose reasoning combined exactness and dynamism. Thus the lecture became a struggle and an adventure. Without rigour suffering, the importance of intuition was restored to us, and the better students were delighted. For the others, the intellectual life was less comfortable, but so exciting... And then, above all, we knew quite well that with such a guide we never risked going under [II.5. p. 8].

Mandelbrojt recalled at the same jubilee:

For several years, Hadamard also gave lectures at the Collège de France: lectures which were long, hard, infinitely interesting. He never tried to hide the difficulties, on the contrary he brought them out. The audience thought together with him; these lectures provoked creativity. The day after a lecture by Hadamard was rich, full and all day long one thought about the ideas.

It was in these lectures that I learnt the secrets of the function $\zeta(s)$ of Riemann, it was there that I understood the significance of analytic continuation, of quasi-analyticity, of Dirichlet series, of the role of functional calculus in the calculus of variations [II.5, p. 25-27].

Digital Assistance PART II. Case Studies PART I: Visual Theorems

EXTENDED ABSTRACT

Long before current graphic, visualisation and geometric tools were available, John E. Littlewood (1885-1977) wrote in his delightful Miscellanv :

A heavy warning used to be given [by lecturers] that pictures are not rigorous; this has never had its bluff called and has permanently frightened its victims into playing for safety. Some pictures, of course, are not rigorous, but I should say most are (and I use them whenever possible myself). [p. 53]

¹J.E. Littlewood, A mathematician's miscellany, London: Methuen (1953); Littlewood, J. E. and Bollobás, Béla, ed., Littlewood's miscellany, Cambridge University Press, 1986.

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Over the past decade, the role of visual computing in my own research has expanded dramatically.

In part this was made possible by the increasing speed and storage capabilities—and the growing ease of programming—of modern multi-core computing environments [BMC].

¹J.E. Littlewood, *A mathematician's miscellany*, London: Methuen (1953); Littlewood, J. E. and Bollobás, Béla, ed., Littlewood's miscellany, Cambridge University Press, 1986.

But, at least as much, it has been driven by my group's paying more active attention to the possibilities for graphing, animating or simulating most mathematical research activities.

²See http://www.carma.newcastle.edu.au/jon/Completion.pdf and http://www.carma.newcastle.edu.au/jon/dr-fields11.pptx.

PART I: Visual Theorems

But, at least as much, it has been driven by my group's paying more active attention to the possibilities for graphing, animating or simulating most mathematical research activities.

- I first briefly discuss both visual theorems and experimental computation.
- I then turn to dynamic geometry (iterative reflection methods [AB]) and matrix completion problems (applied to protein conformation [ABT]).2 (Case studies I)

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- After an algorithmic interlude (Case studies II), I end with description of work from my group in probability (behaviour of short random walks [BS, BSWZ]) and transcendental number theory (normality of real numbers [AB3]). (Case studies III)

²See http://www.carma.newcastle.edu.au/jon/Completion.pdf and http://www.carma.newcastle.edu.au/jon/dr-fields11.pptx.

My plans



PART I: Visual Theorems PART III: Randomness Random-ish

My plans



While all this work involved significant, often threaded [BSC], numerical-symbolic computation, I shall focus on the visual components.

PART I: Visual Theorems Digital Assistance PART II. Case Studies PART III: Randomness

My plans



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- What we have seen and heard so far
- My inclinations on the day
- How I manage my time

Digital Assistance PART II. Case Studies PART I: Visual Theorems PART III: Randomness

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JMB was among roughly 60 new 2015 Fellows of the American Mathematical Society. He was cited "For contributions to nonsmooth analysis and classical analysis as well as experimental mathematics and visualization of mathematics."

Key References and URLS

PART I: Visual Theorems

- AB F. Aragon and J.M. Borwein, "Global convergence of a non-convex Douglas-Rachford iteration." J. Global Optim. **57**(3) (2013), 753–769.
- AB3 F. Aragon, D. H. Bailey, J.M. Borwein and P.B. Borwein, "Walking on real numbers." *Mathematical Intelligencer.* **35**(1) (2013), 42–60.
- ABT F. Aragon, J. M.Borwein, and M. Tam, "Douglas-Rachford feasibility methods for matrix completion problems. *ANZIAM Journal*. Accepted March 2014. Available at http://arxiv.org/abs/1308.4243.
 - BS J.M. Borwein and A. Straub, "Mahler measures, short walks and logsine integrals." Theoretical Computer Science. Special issue on Symbolic and Numeric Computation. 479 (1) (2013), 4-21. DOI: http://link.springer.com/article/10.1016/j.tcs.2012.10.025.
- BSC J.M. Borwein, M. Skerritt and C. Maitland, "Computation of a lower bound to Giuga's primality conjecture." *Integers* **13** (2013). Online Sept 2013 at #A67,
 - http://www.westga.edu/~integers/cgi-bin/get.cgi.
- BSWZ J.M. Borwein, A. Straub, J. Wan and W. Zudilin (with an Appendix by Don Zagier), "Densities of short uniform random walks." *Can. J. Math.* **64**(5), (2012), 961-990.

http://dx.doi.org/10.4153/CJM-2011-079-2

Digital Assistance PART II. Case Studies

...and 3D?



NAMS 2005. KnotPlot in a Cave

Considerable obstacles generally present themselves to the beginner, in studying the elements of Solid Geometry, from the practice which has hitherto uniformly prevailed in this country, of never submitting to the eye of the student, the figures on whose properties he is reasoning, but of drawing perspective representations of them upon a plane.

I hope that I shall never be obliged to have recourse to a perspective drawing of any figure whose parts are not in the same plane.—Augustus De Morgan

In Adrian Rice, "What Makes a Great Mathematics Teacher?" MAA Monthly, 1999.

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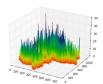
Visual Theorems:

Animation, Simulation and Stereo ...

See http://vis.carma.newcastle.edu.au/: Stoneham movie











Cinderella, 3.14 min of Pi, Catalan's constant and Passive 3D

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Animation, Simulation and Stereo ...

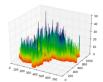
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The latest developments in computer and video technology have provided a multiplicity of computational and symbolic tools that have rejuvenated mathematics and mathematics education. Two important examples of this revitalization are experimental mathematics and visual theorems

— ICMI Study **19** (2012)









Cinderella, 3.14 min of Pi, Catalan's constant and Passive 3D

Visualising large matrices

Large matrices often have structure that pictures will reveal but which numeric data may obscure.

• The picture shows a 25×25 Hilbert matrix on the left and on the right a matrix required to have 50% sparsity and non-zero entries random in [0,1].

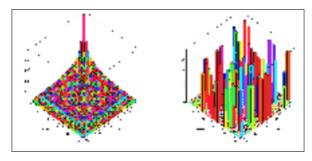


Figure: The Hilbert matrix (L) and a sparse random matrix (R)

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Visualising large matrices

The 4 × 4 Hilbert matrix is

$$\begin{bmatrix} 1 & 1/2 & 1/3 & 1/4 \\ 1/2 & 1/3 & 1/4 & 1/5 \\ 1/3 & 1/4 & 1/5 & 1/6 \\ 1/4 & 1/5 & 1/6 & 1/7 \end{bmatrix}$$

Random-ish

Visualising large matrices

The 4×4 Hilbert matrix is

PART I: Visual Theorems

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Hilbert matrices are notoriously unstable numerically. The left of the Figure shows the inverse of the 20×20 Hilbert matrix computed symbolically exactly. The middle shows enormous numerical errors if one uses 10 digit precision, and the right even if one uses 20 digits.

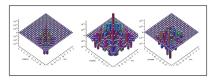
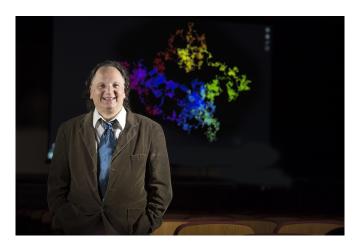


Figure: Inverse 20×20 Hilbert matrix (L) and 2 numerical inverses (R)

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Me and my collaborators



MAA 3.14

http://www.carma.newcastle.edu.au/jon/pi-monthly.pdf

2012 walk on π (went *viral*)

Biggest mathematics picture ever?

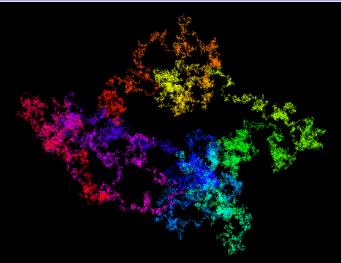


Figure: Walk on first 100 billion base-4 digits of π (normal?).

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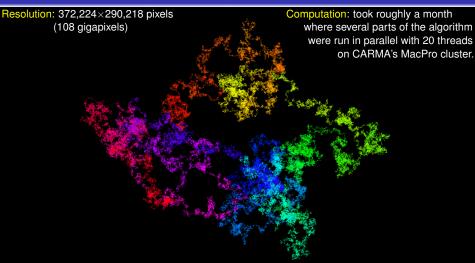


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http://gigapan.org/gigapans/106803

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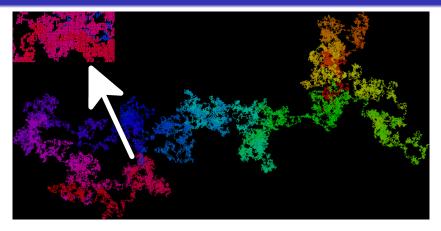
Outreach: images and animations led to high-level research which went viral



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Outreach:

images and animations led to high-level research which went viral



- 100 billion base four digits of π on Gigapan
- Really big pictures are often better than movies (NASA and AMS)

My number-walk collaborators



My short-walk collaborators



James Wan



Armin Straub



Wadim Zudilin

My short-walk collaborators



James Wan



Armin Straub



Wadim Zudilin





and Don Zagier, ...

Dedication: To my friend

Richard E. Crandall (1947-2012)









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- A remarkable man and a brilliant (physical and computational) scientist and inventor, from Reed College
 - Chief scientist for NeXT
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 - and High Performance Computing head
- Developer of the Pixar compression format
 - and the iPod shuffle

http://en.wikipedia.org/wiki/Richard_Crandall

PART I: Visual Theorems Digital Assistance PART II. Case Studies PART III: Randomness

Some early conclusions:

So I am sure they get made

Key ideas: randomness, normality of numbers, planar walks, and fractals



How not to experiment

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How not to experiment

Maths can be done *experimentally* (it is fun)

- using computer algebra, numerical computation and graphics: SNaG
- computations, tables and pictures are experimental data
- but you can not stop thinking

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- keep your eyes open (conquer fear)

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- and what you know you can usually use
- you do not need to know much before you start research (as we shall see)

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DHB and JMB, Exploratory Experimentation in Mathematics (2011), www.ams.org/notices/201110/rtx111001410p.pdf

It is not knowledge, but the act of learning, not possession but the act of getting there, which grants the greatest enjoyment.

PART I: Visual Theorems

When I have clarified and exhausted a subject, then I turn away from it, in order to go into darkness again; the never-satisfied man is so strange if he has completed a structure, then it is not in order to dwell in it peacefully, but in order to begin another.

I imagine the world conqueror must feel thus, who, after one kingdom is scarcely conquered, stretches out his arms for others.



Carl Friedrich Gauss (1777-1855)

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Carl Friedrich Gauss (1777-1855)

- In an 1808 letter to his friend Farkas (father of Janos Bolyai)
- Archimedes, Euler, Gauss are the big three

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Walking on Real Numbers A Multiple Media Mathematics Project

Visit our extensive WALKS gallery

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PUBLICATIONS View our article from the Mathematical Intelligencer, as well as related publications, in this section.

DRESENTATIONS This section contains presentations related to our research.

We have received coverage in the popular has grown from there.

press for our world it all. started with the original Wired' article and news CALLEGY Our extensive gallery of research images.

GIGADAN IMAGES (external link) Clicking here will take you to our very hi-nes. research images of number walks.

TIMES Our page of link are associated w project.

A TABLE OF SLIGHTLY WRONG FOUATIONS AND IDENTITIES USEFUL FOR **APPROXIMATIONS** TROLLING TEACHERS

(FIXIND USING A MIX OF TRIAL AND FRRDR. MATERIALIZA, AND ROBERT MUNERO'S FREST TOOL.) ALL UNITS ARE SI MAS UNLESS OFFERLASE NOTED.

KCTHIKN:		DWEN
ONE LIGHT-YEAR(H)	998	ONE PART IN 40
ENRTH SURFACE(**)	698	ONE PART N 130
OCERNS VOLUME(+1)	919	ONE PART N 70
SECONDS NA YEAR	754	ONE PART N 400
SECONDS NA YEAR (AZAV MENOS)	525,600-60	ONE PART IN 1400
AGE OF THE UNINERSE (60000)	15"	DNE PRRT N 70
PLANCK'S CONSTRUCT	30 ^{11 e}	DNE PRATE
FINE STRUCTURE CONSTRUCT	140	
FUNDAMENTAL CHARGE	3 Hπ ^{eF}	ONE PRINT N 500
WHITE HOUSE SWITCHBOARD	e ^{VI+*V8}	
JENNYS CONSTRUCT	(7 ^{#-à} -9)π²	

TRYLING TEACHERS COORD VOLUMENS SECONOS NA VINC STEEDING MAYOR DANKER INWE 308 OK PRO PAC STRUCTURE FINDHENIA. OHKE 05 RM 711-78 (7º4-9)9° UNIO SIGNA SINY CHIEFE 00 t 00 mg OK MAD 15.25 DE HAS 7/14/23/5/275 RETHANCES PRODN-ELECTRON LITERS N & GRACIN 6-1440

> 5º6e

SUGHTLY WRONG

APPROXIMATIONS

MOTIVATED by the desire to visualize large mathematical data sets, especially in number theory, we offer various tools for refloating point numbers as planar (or three dimensional) walks and for quantitatively measuring their "randomness". This is ou homepage that discusses and showcases our research. Come back regularly for updates.

RESEARCH TEAM: Francisco J. Aragón Artacho, David H. Bailey, Jonathan M. Borwein, Peter B. Borwein with the assistance of Ja Fountain and Matt Skerritt.

CONTACT: Fran Aragon

Almost all I mention is accessible at http://carma.newcastle.edu.au/walks/

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Computer Assisted Research Maths: what it is?

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Like contemporary chemists — and before them the alchemists of old—who mix various substances together in a crucible and heat them to a high temperature to see what happens, today's experimental mathematicians put a hopefully potent mix of numbers, formulas, and algorithms into a computer in the hope that something of interest emerges. (JMB-Devlin, Crucible 2008, p. 1)

 Quoted in International Council on Mathematical Instruction Study 19: On Proof and Proving, 2012

Experimental Mathematics: Integer Relation Methods

Secure Knowledge without Proof. Given real numbers $\beta, \alpha_1, \alpha_2, \dots, \alpha_n$, Helaman Ferguson's integer relation method (PSLQ), finds a nontrivial linear relation of the form

$$a_0\beta + a_1\alpha_1 + a_2\alpha_2 + \dots + a_n\alpha_n = 0, \tag{1}$$

where a_i are integers—if one exists and provides an exclusion bound otherwise.



Carving His Own Unique Niche. In Symbols and Stone

CMS D. Borwein Prize: Madelung



2013 Lattice Sums book (CUP)

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• If $a_0 \neq 0$ then (1) assures β is in rational vector space generated by $\{\alpha_1, \alpha_2, \dots, \alpha_n\}$.



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- $\beta = 1, \alpha_i = \alpha^i$ means α is algebraic of degree n



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- 2000 Computing in Science & Engineering: PSLQ one of top 10 algorithms of 20th century

(2001 CISE article on Grand Challenges (JB-PB))



Carving His Own Unique Niche, In Symbols and Stone

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PSLQ in action

In all serious computations of π from 1700 (by John Machin) until 1980 some version of a *Machin formula* was used. These write

$$\arctan(1) = a_1 \cdot \arctan\left(\frac{1}{p_1}\right) + a_2 \cdot \arctan\left(\frac{1}{p_2}\right) + \dots + a_n \cdot \arctan\left(\frac{1}{p_n}\right)$$
 (2)

for rationals a_1, a_2, \ldots, a_n and integers $p_1, p_2, \ldots, p_n > 1$.

Recall the Taylor series $\arctan(x) = \sum_{n=0}^{\infty} \frac{(-1)^n}{2n+1} x^{2n+1}$. Combined with (2) this computes $\pi = 4\arctan(1)$ efficiently, especially if the p_n are not too small.

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Recall the Taylor series $\arctan(x) = \sum_{n=0}^{\infty} \frac{(-1)^n}{2^{n+1}} x^{2n+1}$. Combined with (2) this computes $\pi = 4\arctan(1)$ efficiently, especially if the p_n are not too small. For instance, Machin found

$$\pi = 16 \arctan\left(\frac{1}{5}\right) - 4 \arctan\left(\frac{1}{239}\right)$$

while Euler discovered

$$\arctan(1) = \arctan\left(\frac{1}{2}\right) + \arctan\left(\frac{1}{5}\right) + \arctan\left(\frac{1}{8}\right).$$
 (3)

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 (2)

for rationals a_1, a_2, \dots, a_n and integers $p_1, p_2, \dots, p_n > 1$.

Recall the Taylor series $\arctan(x) = \sum_{n=0}^{\infty} \frac{(-1)^n}{2n+1} x^{2n+1}$. Combined with (2) this computes $\pi = 4\arctan(1)$ efficiently, especially if the p_n are not too small. For instance. Machin found

$$\pi = 16 \arctan\left(\frac{1}{5}\right) - 4 \arctan\left(\frac{1}{239}\right)$$

while Euler discovered

$$\arctan(1) = \arctan\left(\frac{1}{2}\right) + \arctan\left(\frac{1}{5}\right) + \arctan\left(\frac{1}{8}\right).$$
 (3)

- I have a function 'pslq' in Maple. When input data for PSLQ it predicts
 an answer to the precision requested. And checks it to ten digits more
 (or some other precision).
- This makes the code a real experimental tool as it predicts and confirms.

```
pslq(arctan(1), [arctan(1/2), arctan(1/5), arctan(1/8)],20);;
                             [1, 1, 1, 1], "Error is", 0., "checking to", 30, places
                             \frac{1}{4}\pi = \arctan\left(\frac{1}{2}\right) + \arctan\left(\frac{1}{5}\right) + \arctan\left(\frac{1}{8}\right)
> pslq(arctan(1), [arctan(1/2), arctan(1/3), arctan(1/8)],20);
                         [1, 1, 1, 0], "Error is", -1. 10<sup>-30</sup>, "checking to", 30, places
                                       \frac{1}{4}\pi = \arctan\left(\frac{1}{2}\right) + \arctan\left(\frac{1}{2}\right)
> pslq(arctan(1),[arctan(1/2),arctan(1/5), arctan(1/9)],20);
    [42613, 72375, 22013, -40066], "Error is", 2.31604649037 10<sup>-15</sup>, "checking to", 30, places
             \frac{1}{4} \pi = \frac{72375}{42613} \arctan\left(\frac{1}{2}\right) + \frac{22013}{42613} \arctan\left(\frac{1}{5}\right) - \frac{40066}{42613} \arctan\left(\frac{1}{9}\right)
> pslq(Pi,[arctan(1/5), arctan(1/239)],20);
                         [1, 16, -4], "Error is", 2.8 10<sup>-30</sup>, "checking to", 30, places
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The third shows that when no relation exists the code may find a good approximation but using very large rationals.

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- The third shows that when no relation exists the code may find a good approximation but using very large rationals.
- So it diagnoses failure because it uses large coefficients and because it is not true to the requested 30 places.

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- Specialized Packages or General Purpose Languages such as Fortran, C++, Python, CPLEX, PARI, SnapPea, and MAGMA.

- Web Applications such as: Sloane's Encyclopedia of Integer Sequences, the Inverse Symbolic Calculator, Fractal Explorer, Jeff Weeks' Topological Games, or Euclid in Java.3
 - Most of the functionality of the ISC is built into the "identify" function *Maple* starting with version 9.5. For example, identify (4.45033263602792) returns $\sqrt{3} + e$. As always, the experienced will extract more than the novice.

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- Web Databases including Google, MathSciNet, ArXiv, GitHub, Wikipedia, MathWorld, MacTutor, Amazon, Wolfram Alpha, the DLMF (all formulas of which are accessible in MathML, as bitmaps, and in TEX) and many more that are not always so viewed.

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PART I: Visual Theorems Digital Assistance PART II. Case Studies

Digital Assistance

All entail data-mining. Franklin argues "exploratory experimentation" facilitated by "widening technology", as in finance, pharmacology, astrophysics, medicine, and biotechnology, is leading to a reassessment of what legitimates experiment; in that a "local model" is not now prerequisite. Sørenson says experimental mathematics is following similar tracks.

These aspects of exploratory experimentation and wide instrumentation originate from the philosophy of (natural) science and have not been much developed in the context of experimental mathematics. However, I claim that e.g. the importance of wide instrumentation for an exploratory approach to experiments that includes concept formation also pertain to mathematics.

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Digital Assistance PART II. Case Studies PART I: Visual Theorems

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In consequence, boundaries between mathematics and natural sciences and between inductive and deductive reasoning are blurred and getting more so.

I leave the philosophically-vexing if mathematically-minor question as to if genuine mathematical experiments exist even if one embraces a fully idealist notion of mathematical existence. They sure feel like they do.

Top Ten Algorithms (20C):

all but one well used in CARMA

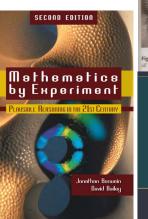
Algorithms for the Ages

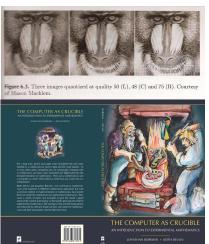
"Great algorithms are the poetry of computation," says Francis Sullivan of the Institute for Defense Analyses' Center for Computing Sciences in Bowie, Maryland. He and Jack Dongarra of the University of Tennessee and Oak Ridge National Laboratory have put together a sampling that might have made Robert Frost beam with pride--had the poet been a computer jock. Their list of 10 algorithms having "the greatest influence on the development and practice of science and engineering in the 20th century" appears in the January/February issue of Computing in Science & Engineering. If you use a computer, some of these algorithms are no doubt crunching your data as you read this. The drum roll, please:

- 1. 1946: The Metropolis Algorithm for Monte Carlo. Through the use of random processes, this algorithm offers an efficient way to stumble toward answers to problems that are too complicated to solve exactly.
- 2. 1947: Simplex Method for Linear Programming. An elegant solution to a common problem in planning and decision-making.
- 3. 1950: Krylov Subspace Iteration Method. A technique for rapidly solving the linear equations that abound in scientific computation.
- 4. 1951: The Decompositional Approach to Matrix Computations. A suite of techniques for numerical linear algebra.
- 5. 1957: The Fortran Optimizing Compiler. Turns high-level code into efficient computer-readable
- 6. 1959: OR Algorithm for Computing Eigenvalues. Another crucial matrix operation made swift and practical.
- 7. 1962: Quicksort Algorithms for Sorting. For the efficient handling of large databases.
- 8. 1965: Fast Fourier Transform. Perhaps the most ubiquitous algorithm in use today, it breaks down waveforms (like sound) into periodic components.
- 9. 1977: Integer Relation Detection. A fast method for spotting simple equations satisfied by collections of seemingly unrelated numbers.
- 10. 1987: Fast Multipole Method. A breakthrough in dealing with the complexity of n-body calculations, applied in problems ranging from celestial mechanics to protein folding.

From Random Samples, Science page 799, February 4, 2000.

Experimental Mathematics: PSLQ is core to CARMA







Experimental Mathematics (2004-08, 2009, 2010)

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It is given for complex numbers a and b by

$$\mathcal{R}(a,b) = \frac{a}{1 + \frac{b^2}{1 + \frac{4a^2}{1 + \frac{9b^2}{1 + \dots}}}}.$$
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We eventually determined from highly sophisticated arguments that:

PART I: Visual Theorems

Simulation in pure mathematics

Theorem (Six formulae for $\Re(a,a), a > 0$)

$$\Re(a,a) = \int_0^\infty \frac{\operatorname{sech}\left(\frac{\pi x}{2a}\right)}{1+x^2} dx$$

$$= 2a \sum_{k=1}^\infty \frac{(-1)^{k+1}}{1+(2k-1)a}$$

$$= \frac{1}{2} \left(\psi \left(\frac{3}{4} + \frac{1}{4a} \right) - \psi \left(\frac{1}{4} + \frac{1}{4a} \right) \right)$$

$$= \frac{2a}{1+a} {}_2F_1 \left(\frac{\frac{1}{2a} + \frac{1}{2}, 1}{\frac{1}{2a} + \frac{3}{2}} \right| - 1 \right)$$

$$= 2 \int_0^1 \frac{t^{1/a}}{1+t^2} dt$$

$$= \int_0^\infty e^{-x/a} \operatorname{sech}(x) dx.$$

Here ${}_{2}F_{1}$ is the hypergeometric function. If you do not know ψ ('psi'), you can easily look it up once you can say 'psi'. Notice that

$$\mathcal{R}(a,a) = 2 \int_0^1 \frac{t^{1/a}}{1+t^2} dt$$

so that $R(1,1) = \log 2$.

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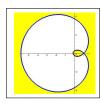
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- After making no progress analytically, Crandall and I decided in 2003, taking a somewhat arbitrary criterion for convergence, to colour yellow points for which the fraction seemed to converge.
- We sampled one million points and reasoned a few thousand mis-categorisations would not damage the experiment.



Simulation in pure mathematics



Simulation in pure mathematics

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Since for positive a, b the fraction satisfies

$$\mathscr{R}(\frac{a+b}{2}, \sqrt{ab}) = \frac{\mathscr{R}(a,b) + \mathscr{R}(b,a)}{2}$$

this gave us enormous impetus to continue our eventually successful hunt for a proof.

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Let $S \subseteq \mathbb{R}^m$. The (nearest point or metric) projection onto S is the (set-valued) mapping,

$$P_S x := \underset{s \in S}{\operatorname{arg\,min}} \|s - x\|.$$

The reflection w.r.t. S is the (set-valued) mapping,

$$R_S := 2P_S - I$$
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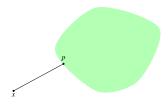


x

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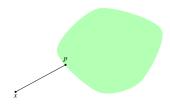


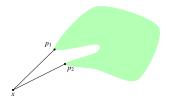


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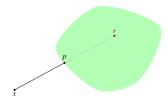


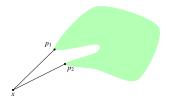


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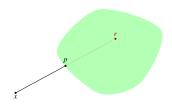


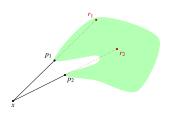


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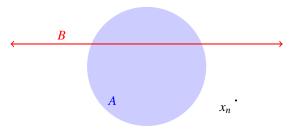
The Douglas-Rachford Algorithm (1956–1979–)

Theorem (Douglas-Rachford in finite dimensions)

Suppose $A, B \subseteq \mathbb{R}^m$ are closed and convex. For any $x_0 \in \mathbb{R}^m$ define

$$x_{n+1} := T_{A,B}x_n$$
 where $T_{A,B} := \frac{I + R_BR_A}{2}$.

If $A \cap B \neq \emptyset$, then $x_n \to x$ such that $P_A x \in A \cap B$. Else $||x_n|| \to +\infty$.

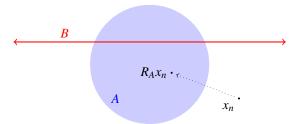


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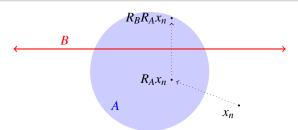
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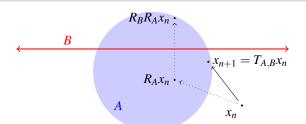


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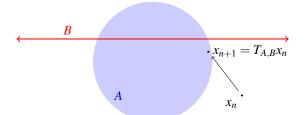


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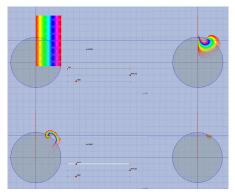
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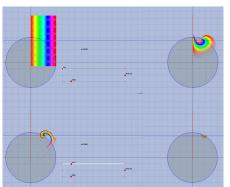
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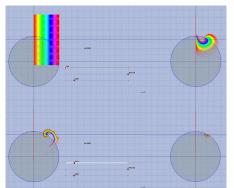


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PART I: Visual Theorems

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 - http://carma.newcastle.edu.au/jon/expansion.html

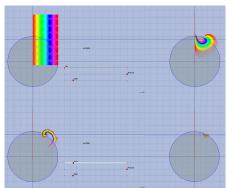


- 20000 starting points coloured by distance from y-axis
- after 0,7,14,21 steps
- a "generic visual theorem"?

In this case we have:

PART I: Visual Theorems

- Some local and fewer global convergence results.
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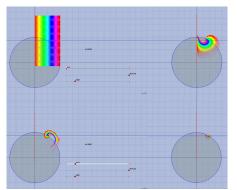
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 - showing global convergence off the (chaotic) v-axis?

ANIMATION

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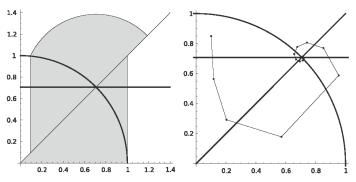
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- 20000 starting points coloured by distance from y-axis
- after 0,7,14,21 steps
- a "generic visual theorem"?
 - showing global convergence off the (chaotic) y-axis?
- note the error from using only 14 digit computation.

What we could *prove* (L) and what we could *see* (R)



2012 Proven region of convergence in grey 2014 Lyapunov function based proof of global convergence (Benoist)

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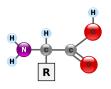
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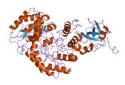
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Protein conformation determination Case study I:

Proteins: large biomolecules comprising multiple amino acid chains.⁴







RuBisCO



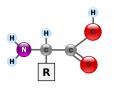
Matt Tam

⁴RuBisCO (responsible for photosynthesis) has 550 amino acids (smallish).

⁵A coupling which occurs through space, rather than chemical bonds.

Case study I: Protein conformation determination

Proteins: large biomolecules comprising multiple amino acid chains.⁴







Generic amino acid

RuBisCO

Matt Tam

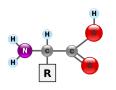
- Proteins participate in virtually every cellular process!
- Protein structure → predicts how functions are performed.
- NMR spectroscopy (Nuclear Overhauser effect⁵) can determine a subset of interatomic distances without damage (under 6Å).

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Case study I: Protein conformation determination

Proteins: large biomolecules comprising multiple amino acid chains.4







Generic amino acid

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A low-rank Euclidean distance matrix completion problem.

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PART I: Visual Theorems

Numerics if reconstructed using reflection methods

We use only interatomic distances below 6Å typically constituting less than 8% of the total nonzero entries of the distance matrix.

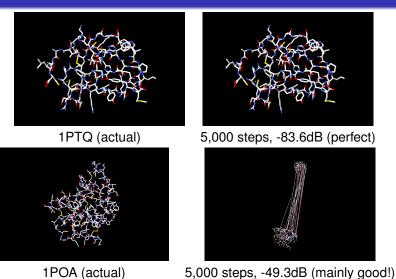
Table. Six Proteins: average (maximum) errors from five replications.

Protein	# Atoms	Rel. Error (dB)	RMSE	Max Error
1PTQ	404	-83.6 (-83.7)	0.0200 (0.0219)	0.0802 (0.0923)
1HOE	581	-72.7 (-69.3)	0.191 (0.257)	2.88 (5.49)
1LFB	641	-47.6 (-45.3)	3.24 (3.53)	21.7 (24.0)
1PHT	988	-60.5 (-58.1)	1.03 (1.18)	12.7 (13.8)
1POA	1067	-49.3 (-48.1)	34.1 (34.3)	81.9 (87.6)
1AX8	1074	-46.7 (-43.5)	9.69 (10.36)	58.6 (62.6)

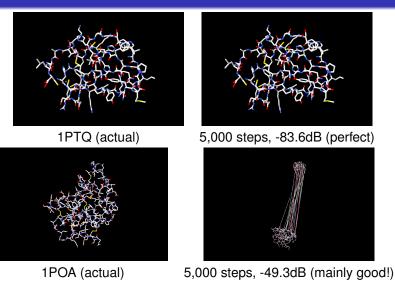
$$\begin{split} & \text{Rel. error}(dB) := 10 \log_{10} \left(\frac{\|P_{C_2} P_{C_1} X_N - P_{C_1} X_N\|^2}{\|P_{C_1} X_N\|^2} \right), \\ & \text{RMSE} := \sqrt{\frac{\sum_{i=1}^m \|\hat{p}_i - p_i^{nue}\|_2^2}{\# \text{ of atoms}}}, \qquad & \text{Max} := \max_{1 \leq i \leq m} \|\hat{p}_i - p_i^{nue}\|_2. \end{split}$$

• The points $\hat{p}_1, \hat{p}_2, \dots, \hat{p}_n$ denote the best fitting of p_1, p_2, \dots, p_n when rotation, translation and reflection is allowed

What do the reconstructions look like?

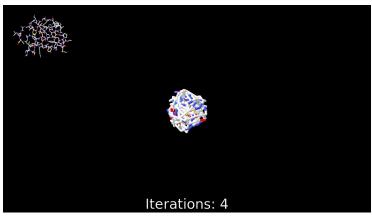


What do the reconstructions look like?



• The picture of 'failure' suggests many strategies

What do reconstructions look like?



Video: First 3,000 steps of the 1PTQ reconstruction.

At http://carma.newcastle.edu.au/DRmethods/1PTQ.html

What do the Reconstructions Look Like?

An optimised implementation gave a ten-fold speed-up.

What do the Reconstructions Look Like?

An optimised implementation gave a ten-fold speed-up. This allowed for the following experiment to be performed:

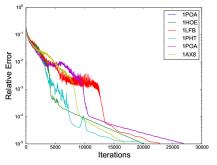


Figure: Relative error by iterations (vertical axis logarithmic).

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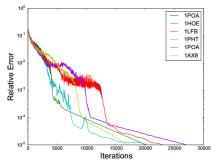


Figure: Relative error by iterations (vertical axis logarithmic).

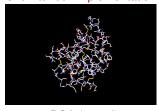
- For < 5,000 iterations, the error exhibits non-monotone oscillatory behaviour. It then decreases sharply. Beyond this progress is slower.
- Is early termination to blame? Terminate when error < −100dB.

 Jonathan Borwein (University of Newcastle, Australia)

 Wisual Theorems www.carma.newcastle.edu.au/walks

A More Robust Stopping Criterion

The "un-tuned" implementation (from previous slide):



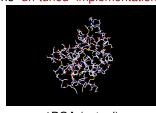
1POA (actual)



5,000 steps (\sim 2d), -49.3dB

A More Robust Stopping Criterion

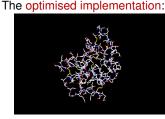
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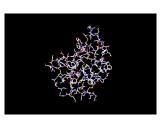
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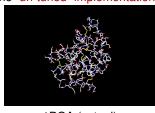
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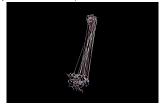
28,500 steps (\sim 1d), -100dB (perfect!)

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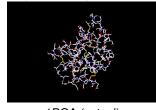


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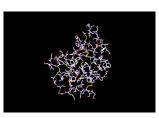


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The optimised implementation:



1POA (actual)



28,500 steps (\sim 1d), -100dB (perfect!)

Similar results observed for the other test proteins

What do reconstructions look like?

There are many projection methods, so why use Douglas-Rachford?

Douglas-Rachford reflection method reconstruction:









500 steps. -25 dB.

1.000 steps. -30 dB.

2.000 steps. -51 dB.

5.000 steps. -84 dB.

Alternating projection method reconstruction:









500 steps. -22 dB.

1.000 steps. -24 dB.

2.000 steps. -25 dB.

5.000 steps. -28 dB.

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500 steps. -22 dB.

1.000 steps. -24 dB.

2.000 steps. -25 dB.

 Yet MAP works very well for optical abberation correction (Hubble, amateur telescopes).

Whv?

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How the mathematical software world has changed

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- To his surprise, a total of 94 teams, representing 25 different nations, submitted results. Twenty of these teams received a full 100 points (10 correct digits for each problem).
- Bailey, Fee and I guit at 85 digits!

The hundred digit challenge

The problems and solutions are dissected most entertainingly in

[1] F. Bornemann, D. Laurie, S. Wagon, and J. Waldvogel (2004). "The Siam 100-Digit Challenge: A Study In High-accuracy Numerical Computing", SIAM, Philadelphia.



▶ SKIP

The hundred digit challenge

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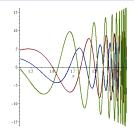
Success in solving these problems required a broad knowledge of mathematics and numerical analysis, together with significant computational effort, to obtain solutions and ensure correctness of the results. As described in [1] the strengths and limitations of Maple, Mathematica, MATLAB (The 3Ms), and other software tools such as PARI or GAP, were strikingly revealed in these ventures.

Almost all of the solvers relied in large part on one or more of these three packages, and while most solvers attempted to confirm their results, there was no explicit requirement for proofs to be provided.

The integral

$$I(\alpha) = \int_0^2 [2 + \sin(10\alpha)] x^{\alpha} \sin\left(\frac{\alpha}{2 - x}\right) dx$$

depends on the parameter α . What is the value $\alpha \in [0,5]$ at which $I(\alpha)$ achieves its maximum?

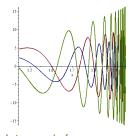


Integrands for some α

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Integrands for some α

• $I(\alpha)$ is expressible in terms of a Meijer-G function —a special function with a solid history that we use below.

$$I(\alpha) = 4\sqrt{\pi} \Gamma(\alpha) G_{2A}^{3,0} \left(\frac{\alpha^2}{16} \middle| \frac{\frac{\alpha+2}{2}, \frac{\alpha+3}{2}}{\frac{1}{2}, \frac{1}{2}, 1, 0} \right) [\sin(10\alpha) + 2].$$

- Unlike most contestants, Mathematica and Maple will figure this out; help files or a web search then inform the scientist.
- This is another measure of the changing environment. It is usually a good idea—and not at all immoral—to data-mine.

A particle at the center of a 10×1 rectangle undergoes Brownian motion (i.e., 2-D random walk with infinitesimal step lengths) till it hits the boundary. What is the probability that it hits at one of the ends rather than at one of the sides?

ANIMATION

Walking in a 10 × 5 box

ANIMATION

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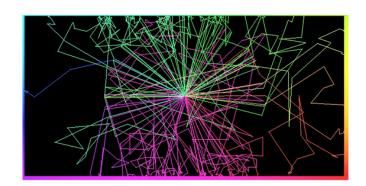


Hitting the Ends. Bornemann [1] starts his remarkable solution by exploring *Monte-Carlo methods*, which are shown to be impracticable.

- He reformulates the problem deterministically as the value at the center of a 10 × 1 rectangle of an appropriate harmonic measure of the ends, arising from a 5-point discretization of Laplace's equation with Dirichlet boundary conditions.
- This is then solved by a well chosen sparse Cholesky solver. A reliable numerical value of 3.837587979 · 10⁻⁷ is obtained and the problem is solved *numerically* to the requisite ten places.
- This is the warm up....

Walking in a $b \times a$ box

ANIMATION



We may proceed to develop two analytic solutions, the *first* using *separation of variables* on the underlying PDE on a general $2a \times 2b$ rectangle. We learn that with $\rho := a/b$

$$p(a,b) = \frac{4}{\pi} \sum_{n=0}^{\infty} \frac{(-1)^n}{2n+1} \operatorname{sech}\left(\frac{\pi(2n+1)}{2}\rho\right).$$
 (5)

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- Three terms yields 50 correct digits: p(10,1) = 0.00000038375879792512261034071331862048391007930055940724...
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A second method using conformal mappings, yields

$$\operatorname{arccot} \rho = p(a,b) \frac{\pi}{2} + \operatorname{arg} K\left(e^{ip(a,b)\pi}\right)$$
 (6)

where K is the *complete elliptic integral* of the first kind.

• We have entered the wonderful world of modular functions

We have entered the wonderful world of modular functions

Bornemann et al ultimately show that the answer is

$$p = \frac{2}{\pi}\arcsin(k_{100})\tag{7}$$

where

$$k_{100} := \left(\left(3 - 2\sqrt{2} \right) \left(2 + \sqrt{5} \right) \left(-3 + \sqrt{10} \right) \left(-\sqrt{2} + \sqrt[4]{5} \right)^2 \right)^2,$$

is a *singular value*. [In general $p(a,b) = \frac{2}{\pi} \arcsin \left(k_{(a/b)^2} \right)$.]

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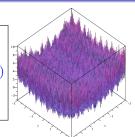
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is a *singular value*. [In general $p(a,b) = \frac{2}{\pi} \arcsin(k_{(a/b)^2})$.]

- No one (except harmonic analysts perhaps) anticipated a closed form—let alone one like this.
- Can be done for some other shapes (perhaps, convex with piecewise smooth boundaries, starting at barycentre), and for self-avoiding walks.

What is the global minimum of the function

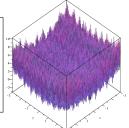
$$\exp(\sin(50x)) + \sin(60e^y) + \sin(70\sin x) + \sin(\sin(80y))$$
$$-\sin(10(x+y)) + (x^2 + y^2)/4?$$



... zooming

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$$\exp(\sin(50x)) + \sin(60e^y) + \sin(70\sin x) + \sin(\sin(80y))$$
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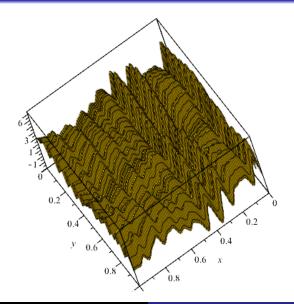


- Can be solved in a global optimization package or by a damped Newton method
- In Mathematica by NMinimize[f[x, y], x, y, Method -> "RandomSearch", "SearchPoints" -> 250, WorkingPrecision -> 201
- In Maple by NLPSolve (f(x,y), x = -4 .. 4, y = -4 .. 4, initial point = $\{x = -.4, y = -.1\}$);
- or by 'zooming' on $[-3,3] \times [-3,3]$.

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Trefethen's problem #4

... zooming on [0,1]



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Algorithm performance

a simulated interlude

Proposition (Polylogarithm computation)

(a) For s = n a positive integer,

$$\operatorname{Li}_{n}(z) = \sum_{m=0}^{\infty} \zeta(n-m) \frac{\log^{m} z}{m!} + \frac{\log^{n-1} z}{(n-1)!} (H_{n-1} - \log(-\log z)).$$
 (8)

(b) For any complex order s not a positive integer.

$$Li_{s}(z) = \sum_{m \ge 0} \zeta(s - m) \frac{\log^{m} z}{m!} + \Gamma(1 - s)(-\log z)^{s - 1}.$$
 (9)

Here $\zeta(s) := \sum_{n=0}^{\infty} z_n$ and continuations, $H_n := 1 + \frac{1}{2} + \frac{1}{3} + \cdots + \frac{1}{n}$, and $\sum_{n=0}^{\infty} z_n$ avoids the singularity at $\zeta(1)$.

In (8), $|\log z| < 2\pi$ precludes use when $|z| < e^{-2\pi} \approx 0.0018674$. For small |z|, however, it suffices to use the definition

$$\operatorname{Li}_{s}(z) = \sum_{k=1}^{\infty} \frac{z^{k}}{k^{s}}.$$
 (10)

Algorithm performance

a simulated interlude

• We found (10) faster than (8) whenever |z| < 1/4, for precision from 100 to 4000 digits. We illustrate for Li₂ in the Figure.

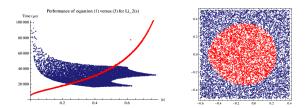


Figure: (L) Timing (8) (blue) and (10) (red).(R) blue region where (8) is faster.

PART I: Visual Theorems

- We found (10) faster than (8) whenever |z| < 1/4, for precision from 100 to 4000 digits. We illustrate for Li₂ in the Figure.
- Timings show microseconds required for 1,000 digit accuracy as the modulus goes from 0 to 1 with blue showing superior performance of (8). The region records 10,000 trials of random z, such that $-0.6 < \Re(z) < 0.4, -0.5 < \Im(z) < 0.5$.

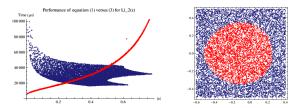


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We shall explore things like:

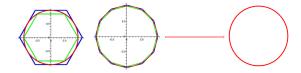
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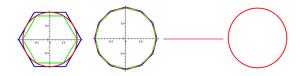
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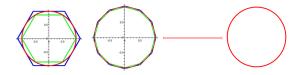


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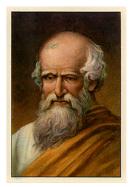
 $6 \mapsto 12 \mapsto 24 \mapsto 48 \mapsto 96$ to obtain the estimate

$$3\frac{10}{71} < \pi < 3\frac{10}{70}$$
.

PART I: Visual Theorems PART III: Randomness

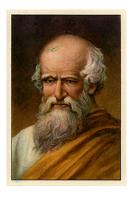
Archimedes' "Method of Mechanical Theorems"

Pi movie below



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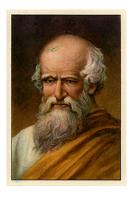


... certain things first became clear to me by a mechanical method (Codex C), although they had to be proved by geometry afterwards because their investigation by the said method did not furnish an actual proof.

But it is of course easier, when we have previously acquired, by the method, some knowledge of the questions, to supply the proof than it is to find it without any previous knowledge.

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But it is of course easier, when we have previously acquired, by the method, some knowledge of the questions, to supply the proof than it is to find it without any previous knowledge.

• Only recently rediscovered and even more recently reconstructed ...

Proving π is not $\frac{22}{7}$

PART I: Visual Theorems

Even *Maple or Mathematica* 'knows' this since

$$0 < \int_0^1 \frac{(1-x)^4 x^4}{1+x^2} dx = \frac{22}{7} - \pi, \tag{11}$$

though it would be prudent to ask 'why' it can perform the integral and 'whether' to trust it?

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 Accidentally, 22/7 is one of the early continued fraction approximation to π . These commence:

$$3, \frac{22}{7}, \frac{333}{106}, \frac{355}{113}, \dots$$

Proving π is not $\frac{22}{7}$

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Proving π is not $\frac{22}{7}$

PART I: Visual Theorems

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$$\int_0^{\mathbf{t}} \frac{x^4 (1-x)^4}{1+x^2} dx = \frac{1}{7} t^7 - \frac{2}{3} t^6 + t^5 - \frac{4}{3} t^3 + 4t - 4 \arctan(t)$$

as differentiation easily confirms, and the fundamental theorem of calculus proves (11). QED

> An opinion without 3.14 is an onion. You'll understand.

Randomness

PART I: Visual Theorems

• The digits expansions of $\pi, e, \sqrt{2}$ appear to be "random":

```
\pi = 3.141592653589793238462643383279502884197169399375...
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Digital Assistance PART II. Case Studies PART I: Visual Theorems PART III: Randomness

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 1949 ENIAC (Electronic Numerical Integrator and Calculator) computed of π to 2,037 decimals (in 70 hours)—proposed by polymath John von Neumann (1903-1957) to shed light on distribution of π (and of e).





Gauss map. Remove the integer, invert the fraction and repeat: for 3.1415926 and 2.7182818 to get the fractions below.

$$\pi = 3 + \cfrac{1}{7 + \cfrac{1}{15 + \cfrac{1}{1 + \cfrac{1}{292 + \cfrac{1}{1 + \cfrac{1}{1 + \dots}}}}}}$$



Two continued fractions

Change representations often

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Two continued fractions

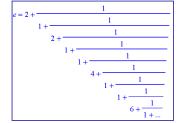
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Leonhard Euler (1707 -**1783**) named e and π .

"Lisez Euler, lisez Euler, c'est notre maître à tous." Simon Laplace (1749-1827)

Digit	Ocurrences
0	99,993,942
1	99,997,334
2	100,002,410
3	99,986,911
4	100,011 ,958
5	99,998 ,885
6	100,010,387
7	99,996,061
8	100,001,839
9	100,000,273
Total	1,000,000,000

Table: Counts of first billion digits of π . Second half is 'right' for law of large numbers.

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 - Fuler found the 292
 - e has a fine continued fraction
- There are infinitely many sevens in the decimal expansion of π
- There are infinitely many ones in the ternary expansion of π
- There are equally many zeroes and ones in the binary expansion of π
- Or pretty much anything else...

A hard question







A hard question







It might be:

- Unpredictable (fair dice or coin-flips)?
- Without structure (noise)?
- Algorithmically random (π is not)?
- Quantum random (radiation)?
- Incompressible ('zip' does not help)?

A hard question







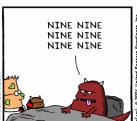
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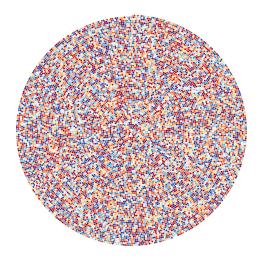
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Best Theorem [BBCP, 04] (Feeble but hard) Asymptotically all degree d algebraics have at least $n^{1/d}$ ones in binary (should be n/2

PART I: Visual Theorems PART III: Randomness Random-ish

Randomness in Pi?

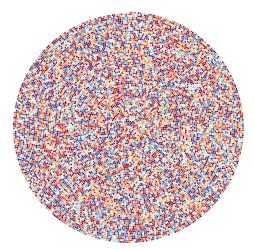
http://mkweb.bcgsc.ca/pi/art/



PART I: Visual Theorems PART III: Randomness

Randomness in Pi?

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• a better color palette for art if not for science

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Normality

A property random numbers must possess

Definition

A real constant α is b-normal if, given the positive integer b > 2 (the base), every m-long string of base-b digits appears in the base-b expansion of α with precisely the expected limiting frequency $1/b^m$.

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• Given an integer b > 2, almost all real numbers, with probability one, are b-normal (Borel).







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- Indeed, almost all real numbers are b-normal simultaneously for all positive integer bases ("absolute normality").







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- Indeed, almost all real numbers are b-normal simultaneously for all positive integer bases ("absolute normality").
- Unfortunately, it has been very difficult to prove normality for any number in a given base b, much less all bases simultaneously.







Normal numbers

concatenation numbers

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C_{10} := 0.123456789101112131415161718...
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- Copeland–Erdős constant
- Normality proofs are not known for π , e, $\log 2$, $\sqrt{2}$ etc.

A real constant α is b-normal if, given the positive integer b > 2 (the base), every m-long string of base-b digits appears in the base-b expansion of α with precisely the expected limiting frequency $1/b^m$.

Theorem (Davenport-Erdös (1952))

Let p be any polynomial positive on the natural numbers. Then the concatenation number

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- Includes Champernowne's number and 0.1491625... (Besicovich)
- See H. Davenport and P. Erdös, "Note on normal numbers." Can. J. Math., 4 (1952), 58-63.

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8	100,001,839	08	10,000,565	800	999,962
9	100,000,273	09	9,999,276	009	999,059
		10	9,997,289	010	998,884
		11	9,997,964	011	1,001,188
		:	:	:	:
		99	10,003,709	099	999,201
				:	:
				999	1,000,905
TOTAL	1,000,000,000	TOTAL	1,000,000,000	TOTAL	1,000,000,000

Table: Counts for the first billion digits of π .

← Counts of first trillion hex digits

otal	1.000.000.000.000
F	62499937801
E	62499875079
D	62499613666
С	62500188610
В	62499955595
A	62500266095
9	62500120671
8	<u>62500</u> 216752
7	62499878794
6	62499925426
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 - They are 353*CB*3*F*7*F*0*C*9*ACCFA*9*AA*215*F*2

See www.karrels.org/pi/index.html



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Modern π Calculation Records:

Name	Year	Correct Digits
Miyoshi and Kanada	1981	2,000,036
Kanada-Yoshino-Tamura	1982	16,777,206
Gosper	1985	17,526,200
Bailey	Jan. 1986	29,360,111
Kanada and Tamura	Sep. 1986	33,554,414
Kanada and Tamura	Oct. 1986	67,108,839
Kanada et. al	Jan. 1987	134,217,700
Kanada and Tamura	Jan. 1988	201,326,551
Chudnovskys	May 1989	480,000,000
Kanada and Tamura	Jul. 1989	536,870,898
Kanada and Tamura	Nov. 1989	1,073,741,799
Chudnovskys	Aug. 1991	2,260,000,000
Chudnovskys	May 1994	4,044,000,000
Kanada and Takahashi	Oct. 1995	6,442,450,938
Kanada and Takahashi	Jul. 1997	51,539,600,000
Kanada and Takahashi	Sep. 1999	206,158,430,000
Kanada-Ushiro-Kuroda	Dec. 2002	1,241,100,000,000
Takahashi	Jan. 2009	1,649,000,000,000
Takahashi	April 2009	2,576,980,377,524
Bellard	Dec. 2009	2,699,999,990,000
Kondo and Yee	Aug. 2010	5,000,000,000,000

Oct. 2011

Dec. 2013



and IBM Blue Gene/L at LBL

Kondo and Yee

Kondo and Yee

10,000,000,000,000

12,100,000,000,000

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Prior to **1996**, most folks thought to compute the d-th digit of π , you had to generate the (order of) the entire first *d* digits. **This is not true**:

• at least for hex (base 16) or binary (base 2) digits of π .

PART I: Visual Theorems

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 - a computational cost growing only slightly faster than the digit position.
- An algorithm found by computer—now used to check record π computations and in some compilers.

Random-ish

This is based on the following then new formula for π :

$$\pi = \sum_{i=0}^{\infty} \frac{1}{16^i} \left(\frac{4}{8i+1} - \frac{2}{8i+4} - \frac{1}{8i+5} - \frac{1}{8i+6} \right) \tag{12}$$

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$$\pi = 4_2 F_1 \left(1, \frac{1}{4}; \frac{5}{4}, -\frac{1}{4} \right) + 2 \tan^{-1} \left(\frac{1}{2} \right) - \log 5$$

where $_{2}F_{1}(1,1/4;5/4,-1/4) = 0.955933837...$ is a Gaussian hypergeometric function.

Reverse Engineered Mathematics

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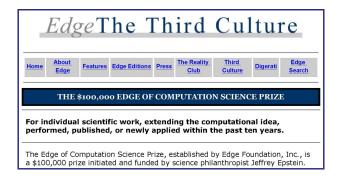
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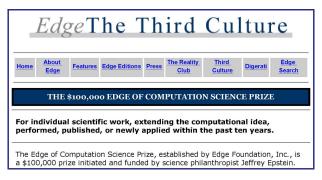
Bailey-Crandall (220) link BBP and normality.

Edge of Computation Prize Finalist (2005)



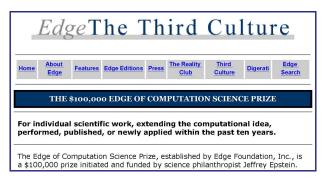
Digital Assistance PART II. Case Studies PART I: Visual Theorems PART III: Randomness Random-ish

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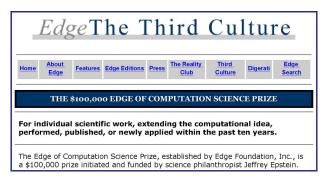


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Edge of Computation Prize Finalist (2005)



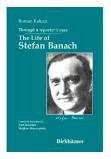
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- Won by David Deutsch discoverer of Quantum Computing.

Another Nazi casuality

A mathematician is a person who can find analogies between theorems: a better mathematician is one who can see analogies between proofs and the best mathematician can notice analogies between theories. 6





⁶Only the best get stamps. Quoted in

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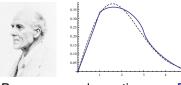


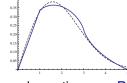
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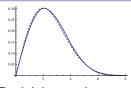
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From a vast literature





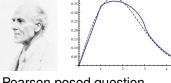


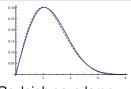


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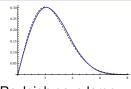
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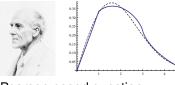
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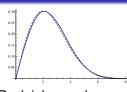
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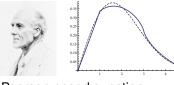
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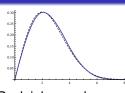
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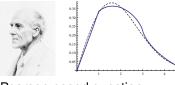
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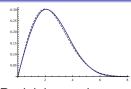
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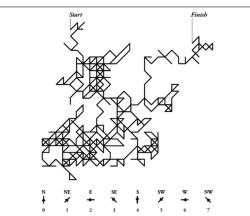
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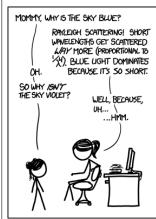
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- appear in graph theory, quantum chemistry, in quantum physics as hexagonal and diamond lattice integers, etc...

Random-ish

The first walk (Venn)



The first person to visualize the random nature of pi's decimal digits was the Victorian nathematician John Venn. In The Logic of Chance (1888), he suggested that the digits 0 to 7 epresent eight compass directions, and he followed the path tracked by these digits in pi. He nisses out the initial 3, and starts 14159. Venn's image was the first "random walk", an idea now ised frequently in probability and statistics. (The illustration is taken from my book, Alex's Adventures in Numberland)



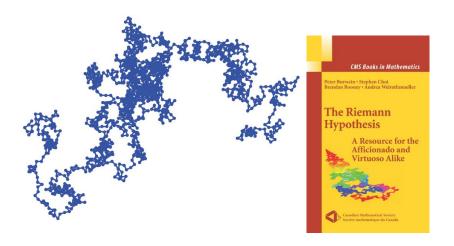
MY HOBBY: TEACHING TRICKY QUESTIONS TO THE CHILDREN OF MY SCIENTIST FRIENDS.

PART I: Visual Theorems PART III: Randomness

One 1500-step ramble: a familiar picture

Liouville function

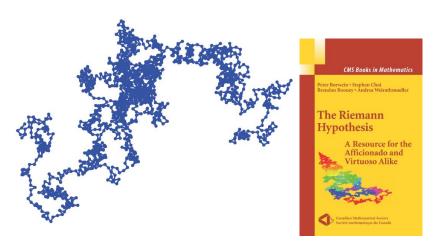
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Digital Assistance PART II. Case Studies PART I: Visual Theorems PART III: Randomness Random-ish

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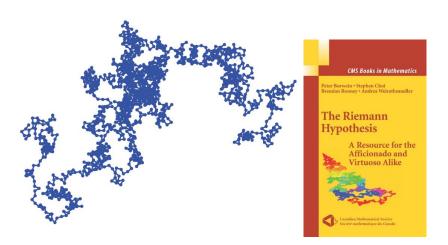
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Digital Assistance PART II. Case Studies PART III: Randomness PART I: Visual Theorems

One 1500-step ramble: a familiar picture

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Random-ish



- 1D (and 3D) easy. Expectation of RMS distance is easy (\sqrt{n}) .
- 1D or 2D *lattice*: probability one of returning to the origin.

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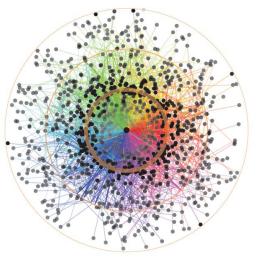
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PART I: Visual Theorems PART III: Randomness Random-ish

Case study II: short rambles

a less familiar picture?





1000 three-step uniform planar walks

The moments of an *n*-step planar walk: $W_n := W_n(1)$

Second simplest case:

$$W_2 = \int_0^1 \int_0^1 \left| e^{2\pi i x} + e^{2\pi i y} \right| dx dy = ?$$

⁷Quadrature was our first interest

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$$W_n(s) = \int_{[0,1]^n} \left| \sum_{k=1}^n e^{2\pi x_k i} \right|^s d(x_1, \dots, x_{n-1}, x_n)$$

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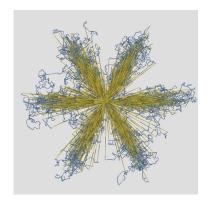
• So $W_2 = 4 \int_0^{1/4} \cos(\pi x) dx = \frac{4}{\pi}$.

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PART I: Visual Theorems Random-ish

Art meets science

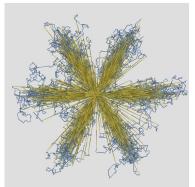
AAAS & Bridges conference



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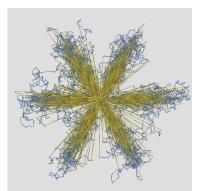
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A visualization of six routes that 1000 ants took after leaving their nest in search of food. The jagged blue lines represent the breaking off of random ants in search of seeds.

(Nadia Whitehead 2014-03-25 16:15)

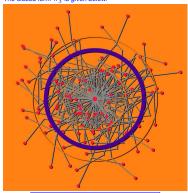


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(JonFest 2011 Logo) Three-step random walks. The (purple) expected distance travelled is 1.57459 ...

The closed form W_3 is given below.

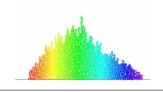


$$W_3 = \frac{16\sqrt[3]{4}\pi^2}{\Gamma(\frac{1}{3})^6} + \frac{3\Gamma(\frac{1}{3})^6}{8\sqrt[3]{4}\pi^4}$$

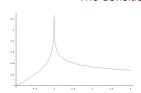
Simulating the densities for n = 3,4

ANIMATION

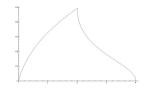




The densities p_3 (L)



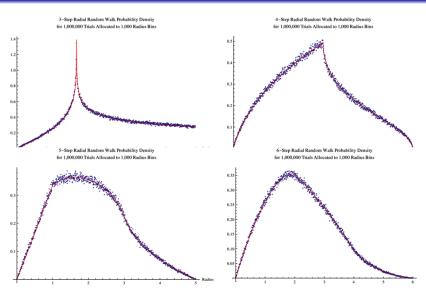
 p_4 (R) and



Simulation thanks to Cam Rogers

The radial densities for $3 \le n \le 6$





Pearson's original full question

and comment on p_5

A man starts from a point O and walks l yards in a straight line; he then turns through any angle whatever and walks another *l* yards in a second straight line. He repeats this process n times. I require the probability that after these nstretches he is at a distance between r and $r + \delta r$ from his starting point, O.

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"the graphical construction, however carefully reinvestigated, did not permit of our considering the curve to be anything but a straight line... Even if it is not absolutely true, it exemplifies the extraordinary power of such integrals of J products to give extremely close approximations to such simple forms as horizontal lines."

 2015. Our analysis of short walks extends interestingly to arbitrary dimensions ...

Let $\sigma(x) := \frac{3-x}{1+x}$. Then σ is an involution on [0,3] sending [0,1] to [1,3]:

$$p_3(x) = \frac{4x}{(3-x)(x+1)} p_3(\sigma(x)). \tag{13}$$

So
$$\frac{3}{4}p_3'(0) = p_3(3) = \frac{\sqrt{3}}{2\pi}, \ p(1) = \infty.$$

The radial densities for n = 3.4 are modular functions

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So $\frac{3}{4}p_3'(0) = p_3(3) = \frac{\sqrt{3}}{2\pi}$, $p(1) = \infty$. We found:

$$p_3(\alpha) = \frac{2\sqrt{3}\alpha}{\pi (3+\alpha^2)} {}_2F_1\left(\begin{array}{c} \frac{1}{3}, \frac{2}{3} \\ 1 \end{array} \middle| \frac{\alpha^2 (9-\alpha^2)^2}{(3+\alpha^2)^3}\right) = \frac{2\sqrt{3}}{\pi} \frac{\alpha}{AG_3(3+\alpha^2, 3(1-\alpha^2)^{2/3})}$$

where AG₃ is the *cubically convergent* mean iteration (1991):

$$AG_3(a,b) := \frac{a+2b}{3} \bigotimes \left(b \cdot \frac{a^2 + ab + b^2}{3} \right)^{1/3}.$$

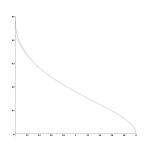


We ultimately deduce on $2 \le \alpha \le 4$ a hyper-closed form:

$$p_4(\alpha) = \frac{2}{\pi^2} \frac{\sqrt{16 - \alpha^2}}{\alpha} {}_3F_2\left(\frac{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}}{\frac{5}{6}, \frac{7}{6}} \left| \frac{\left(16 - \alpha^2\right)^3}{108 \,\alpha^4} \right). \tag{15}$$

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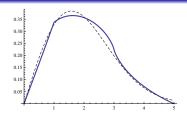
- $\leftarrow p_4$ from (15) vs 18-terms of empirical power series
- ✓ Proves $p_4(2) = \frac{2^{7/3}\pi}{3\sqrt{3}} \Gamma(\frac{2}{3})^{-6} =$ $\frac{\sqrt{3}}{\pi}W_3(-1)\approx 0.494233<\frac{1}{2}$
- Empirically, quite marvelously, we found — and proved by a subtle use of distributional Mellin transforms — that on [0,2] as well:

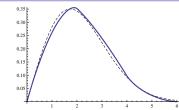
$$p_4(\alpha) \stackrel{?}{=} \frac{2}{\pi^2} \frac{\sqrt{16 - \alpha^2}}{\alpha} \Re_3 F_2 \left(\frac{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}}{\frac{5}{6}, \frac{7}{6}} \middle| \frac{\left(16 - \alpha^2\right)^3}{108 \alpha^4} \right)$$
 (16)

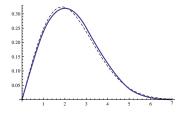
(Discovering this \Re brought us full circle.)

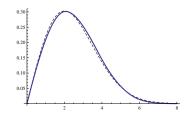
The radial densities for $5 \le n \le 8$

(and large n approximation)



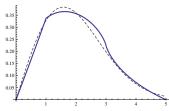


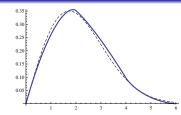




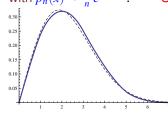
The radial densities for 5 < n < 8

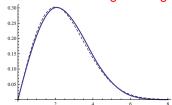
(and large *n* approximation)





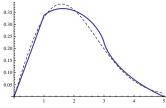
Both p_{2n+4}, p_{2n+5} are *n*-times continuously differentiable for x > 0with $p_n(x) \sim \frac{2x}{n} e^{-x^2/n}$. So "four is small" but "eight is large."

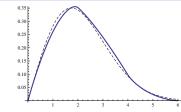




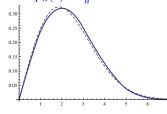
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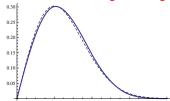
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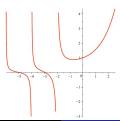
• Pearson wondered if p_5 was linear on [0,1]. Only disproven in sixties.

and graph on real line

Theorem (Meijer-G form for W_3)

For s not an odd integer

$$W_3(s) = \frac{\Gamma(1+\frac{s}{2})}{\sqrt{\pi} \Gamma(-\frac{s}{2})} G_{33}^{21} \begin{pmatrix} 1,1,1 \\ \frac{1}{2},-\frac{s}{2},-\frac{s}{2} \end{pmatrix} \frac{1}{4}.$$

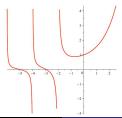


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- First found by Crandall via CAS.
- Proved using residue calculus methods.



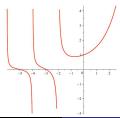
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- First found by Crandall via CAS.
- Proved using residue calculus methods.
- $W_3(s)$ is among the first non-trivial higher order Meijer-G function to be placed in closed form.



Theorem (Meijer form for W_4)

For $\Re s > -2$ and s not an odd integer

$$W_4(s) = \frac{2^s}{\pi} \frac{\Gamma(1+\frac{s}{2})}{\Gamma(-\frac{s}{2})} G_{44}^{22} \begin{pmatrix} 1, \frac{1-s}{2}, 1, 1\\ \frac{1}{2} - \frac{s}{2}, -\frac{s}{2}, -\frac{s}{2} \end{pmatrix} | 1$$
 (17)

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He [Gauss (or Mathematica)] is like the fox, who effaces his tracks in the sand with his tail.— Niels Abel (1802-1829)

Random-ish

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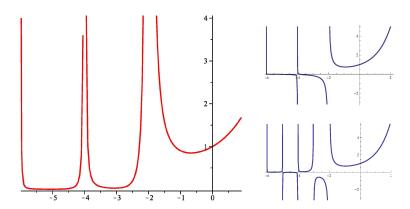
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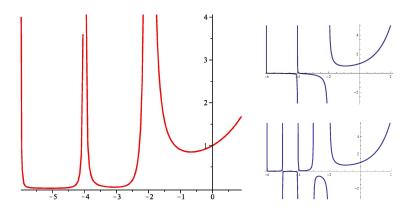
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But we really need a formula with s = 1, that is an **integer**.

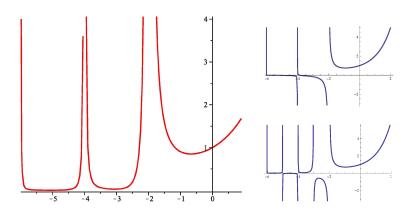
Visualizing W_4, W_5 , and W_6 on the real line



Visualizing W_4, W_5 , and W_6 on the real line



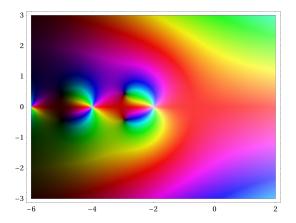
• Use recursion from s > 1



- Use recursion from s > 1
- Nonnegativity of W₄ was hard to prove (Wan)

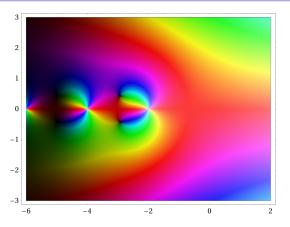
PART I: Visual Theorems PART III: Randomness Random-ish

Visualizing W_4 in the complex plane



PART I: Visual Theorems Digital Assistance PART II. Case Studies PART III: Randomness

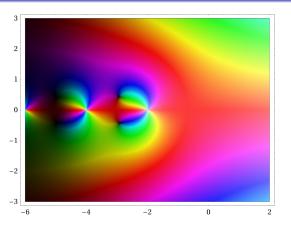
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 Easily drawn now in Mathematica from the Meijer-G representation

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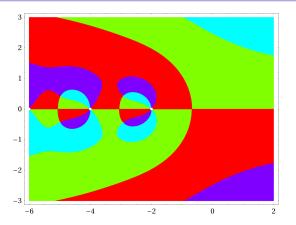


- Easily drawn now in Mathematica from the Meijer-G representation
- Each point is coloured differently (black is zero and white infinity). Note the poles and zeros.

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Visualizing W_4 in the complex plane:

sometimes less is more

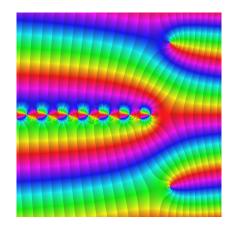


- Easily drawn now in Mathematica from the Meijer-G representation.
- Each quadrant is coloured differently (black is zero and white infinity). Note the poles and zeros.

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Visualizing W_4 in the complex plane:

sometimes less is more



- Less easily drawn now from the Meijer-G representation.
- As prepared for Springer's Mathematical Beauties (2016).

Simplifying the Meijer integrals for W_3 and W_4

• We (humans and/or computers) now obtained:

Simplifying the Meijer integrals for W_3 and W_4

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Corollary (Hypergeometric forms for non-integer s > -2)

$$\textbf{\textit{W}}_{3}(\textbf{\textit{s}}) = \frac{\tan\left(\frac{\pi s}{2}\right)}{2^{2s+1}} \binom{s}{\frac{s-1}{2}}^{2} {}_{3}F_{2} \left(\frac{\frac{1}{2},\frac{1}{2},\frac{1}{2}}{\frac{s+3}{2},\frac{s+3}{2}} \left| \frac{1}{4} \right. \right) + \binom{s}{\frac{s}{2}} {}_{3}F_{2} \left(\frac{-\frac{s}{2},-\frac{s}{2},-\frac{s}{2}}{1,-\frac{s-1}{2}} \left| \frac{1}{4} \right. \right),$$

and

$$W_4(s) = \frac{\tan\left(\frac{\pi s}{2}\right)}{2^{2s}} {s \choose \frac{s-1}{2}}^3 {}_4F_3 \left(\frac{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{s}{2} + 1}{\frac{s+3}{2}, \frac{s+3}{2}} \right| 1 + {s \choose \frac{s}{2}} {}_4F_3 \left(\frac{\frac{1}{2}, -\frac{s}{2}, -\frac{s}{2}, -\frac{s}{2}}{1, 1, -\frac{s-1}{2}} \right| 1 \right).$$

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• We (humans) were able to provably take the limit at ± 1 : e.g.,

$$W_{4}(-1) = \frac{\pi}{4} {}_{7}F_{6}\left(\begin{array}{c} \frac{5}{4}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \\ \frac{1}{4}, 1, 1, 1, 1, 1 \end{array} \middle| 1\right) = \frac{\pi}{4} \sum_{n=0}^{\infty} \frac{(4n+1) {\binom{2n}{n}}^{6}}{4^{6n}}$$
$$= \frac{\pi}{4} {}_{6}F_{5}\left(\begin{array}{c} \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \\ 1, 1, 1, 1, 1, 1 \end{array} \middle| 1\right) + \frac{\pi}{64} {}_{6}F_{5}\left(\begin{array}{c} \frac{3}{2}, \frac{3}{2}, \frac{3}{2}, \frac{3}{2}, \frac{3}{2}, \frac{3}{2} \\ 2, 2, 2, 2, 2, 2 \end{array} \middle| 1\right).$$

from Meijer-G values.

With much work involving moments of elliptic integrals we obtain:

Theorem (Tractable hypergeometric form for W_3)

(a) For $s \neq -3, -5, -7, ...$, we have

$$W_3(s) = \frac{3^{s+3/2}}{2\pi} \beta \left(s + \frac{1}{2}, s + \frac{1}{2} \right) {}_{3}F_2 \left(\frac{s+2}{2}, \frac{s+2}{2}, \frac{s+2}{2} \left| \frac{1}{4} \right. \right). \tag{18}$$

(b) For every natural number k = 1, 2, ...,

$$W_3(-2k-1) = \frac{\sqrt{3} {2k \choose k}^2}{2^{4k+1} 3^{2k}} {}_3F_2\left(\frac{\frac{1}{2}, \frac{1}{2}, \frac{1}{2}}{k+1, k+1} \middle| \frac{1}{4}\right).$$

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 The following formula hints at role played by Bessel functions (Kluywer 1906 and http:

//www.carma.newcastle.edu.au/jon/walks-anu.pdf):

$$W_n = n \int_0^\infty J_1(x) J_0(x)^{n-1} \frac{\mathrm{d}x}{x} \approx \frac{\sqrt{n\pi}}{2}.$$

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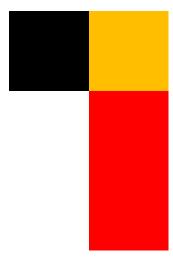




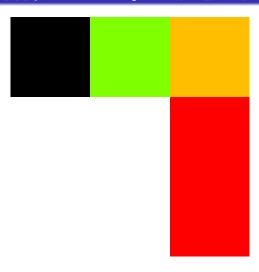


$$2 = \leftarrow$$

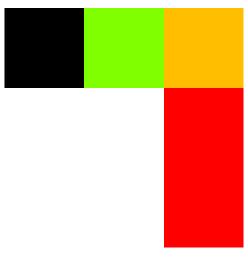




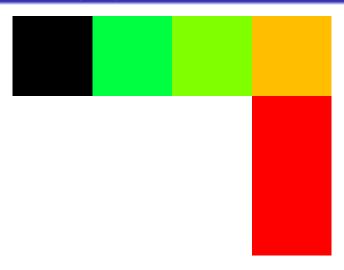
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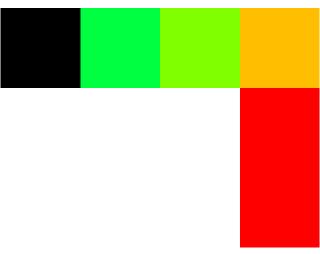


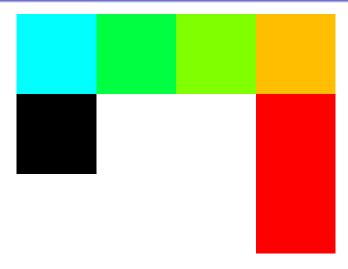
Pick a random number in $\{0,1,2,3\}$ and move according to $0=\rightarrow$, $1=\uparrow$, $2=\leftarrow$, $3=\downarrow$

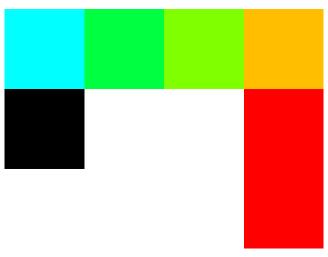


 $2 = \leftarrow$

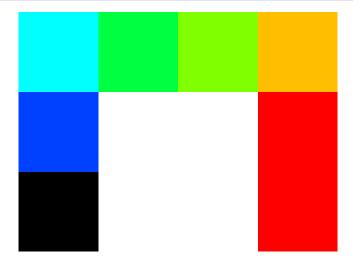




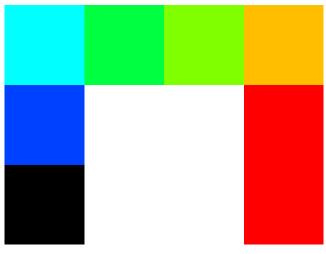




$$3 = \downarrow$$



Pick a random number in $\{0,1,2,3\}$ and move according to $0 = \rightarrow$, $1 = \uparrow$, $2 = \leftarrow$, $3 = \downarrow$

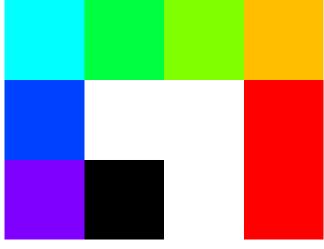


 $0 = \rightarrow$

PART I: Visual Theorems PART III: Randomness Random-ish

What is a (base four) random walk?

Pick a random number in $\{0,1,2,3\}$ and move according to $0=\rightarrow$, $1=\uparrow$, $2=\leftarrow$, $3=\downarrow$



11222330

What is a random walk (base 4)?

Pick a random number in $\{0,1,2,3\}$ and move $0 = \rightarrow$, $1 = \uparrow$, $2 = \leftarrow$, $3 = \downarrow$

ANIMATION

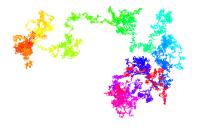


Figure: A million step base-4 pseudorandom walk. We use the spectrum to show when we visited each point (ROYGBIV and R).

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Random walks look similarish

Chaos theory (order in disorder)

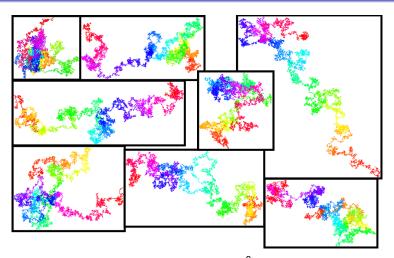


Figure: Eight different base-4 (pseudo)random⁸ walks of one million steps.

⁸ Python uses the *Mersenne Twister* as the core generator. It has a period of $2^{19937} - 1 \approx 10^{6002}$.

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Base-b random walks:

Our direction choice

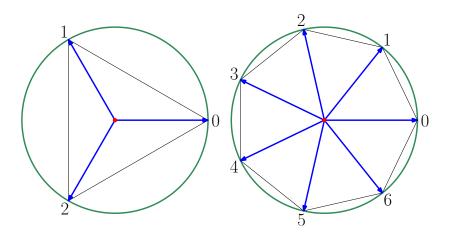


Figure: Directions for base-3 and base-7 random walks.

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III: Two rational numbers

ANIMATION

The base-4 digit expansion of Q1 and Q2:

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III: Two rational numbers

ANIMATION



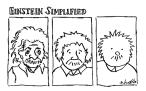


Figure: Self-referent walks on the rational numbers Q1 (top) and Q2 (bottom).

Two more rationals

Hard to tell from their decimal expansions

The following relatively small rational numbers [G. Marsaglia, 2010]

$$\label{eq:Q3} \textit{Q3} = \frac{3624360069}{7000000001} \quad \text{and} \quad \textit{Q4} = \frac{123456789012}{1000000000061},$$

have base-10 periods with huge length of 1,750,000,000 digits and **1,000,000,000,060** digits, respectively.

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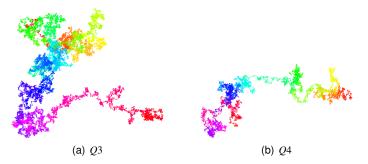


Figure: Walks on the first million base-10 digits of the rationals *Q*3 and *Q*4.

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Walks on the digits of numbers

ANIMATION

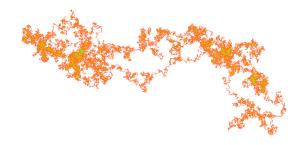


Figure: A walk on the first 10 million base-4 digits of π .

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Walks on the digits of numbers

Coloured by hits (more pink is more hits)

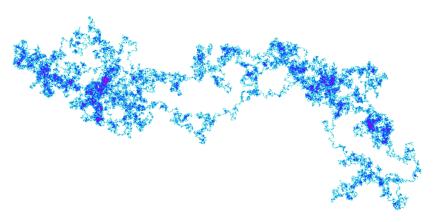


Figure: 100 million base-4 digits of π coloured by number of returns to points.

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The Stoneham numbers

$$\alpha_{b,c} = \sum_{n=1}^{\infty} \frac{1}{c^n b^{c^n}}$$

1973 Richard Stoneham proved some of the following (nearly 'natural') constants are b-normal for relatively prime integers b, c:

$$\alpha_{b,c} := \frac{1}{cb^c} + \frac{1}{c^2b^{c^2}} + \frac{1}{c^3b^{c^3}} + \dots$$

Such super-geometric sums are Stoneham constants. To 10 places

$$\alpha_{2,3} = \frac{1}{24} + \frac{1}{3608} + \frac{1}{3623878656} + \dots$$

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Theorem (Normality of Stoneham constants, Bailey-Crandall '02)

For every coprime pair of integers $b \ge 2$ and $c \ge 2$, the constant $\alpha_{b,c}$ is b-normal.

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Given coprime $b \ge 2$ and $c \ge 2$, such that $c < b^{c-1}$, the constant $\alpha_{b,c}$ is bc-nonnormal.

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Given coprime $b \ge 2$ and $c \ge 2$, such that $c < b^{c-1}$, the constant $\alpha_{b,c}$ is bc-nonnormal.

• Since $3 < 2^{3-1} = 4$, $\alpha_{2,3}$ is 2-normal and 6-nonnormal!

$$lpha_{b,c} = \sum_{n=1}^{\infty} \frac{1}{c^n b^{c^n}}$$

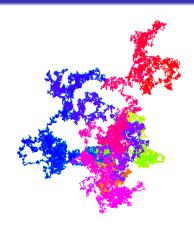


Figure: $\alpha_{2,3}$ is 2-normal (top) but 6-nonnormal (bottom). Is seeing believing?

$$\alpha_{b,c} = \sum_{n=1}^{\infty} \frac{1}{c^n b^{c^n}}$$

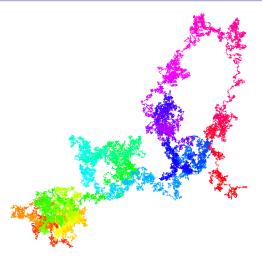


Figure: Is $\alpha_{2,3}$ 3-normal or not?

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The expected distance to the origin



Theorem

The expected distance d_N to the origin of a base-b random walk of N steps behaves like to $\sqrt{\pi N}/2$.

The expected distance to the origin



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Number	Base	Steps	Average normalized dist. to the origin: $\frac{1}{\text{Steps}} \sum_{N=2}^{\text{Steps}} \frac{\text{dist}_N}{\sqrt{\pi N}}$	Normal
Mean of 10,000 random walks	4	1,000,000	1.00315	Yes
Mean of 10,000 walks on the digits of π	4	1,000,000	1.00083	?
$\alpha_{2,3}$	3	1,000,000	0.89275	?
$\alpha_{2,3}$	4	1,000,000	0.25901	Yes
$\alpha_{2,3}$	6	1,000,000	108.02218	No
π	4	1,000,000	0.84366	?
π	6	1,000,000	0.96458	?
π	10	1,000,000	0.82167	?
π	10	1,000,000,000	0.59824	?
$\sqrt{2}$	4	1,000,000	0.72260	?
Champernowne C ₁₀	10	1,000,000	59.91143	Yes

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Number of points visited

For a 2D lattice

• The expected number of distinct points visited by an N-step random walk on a two-dimensional lattice behaves for large N like $\pi N/\log(N)$ (Dvoretzky–Erdős, **1951**).

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• For example, for $N = 10^6$ these bounds are (199256.1,203059.5), while $\pi N/\log(N) = 227396$, which overestimates the expectation. PART I: Visual Theorems PART III: Randomness Random-ish

Catalan's constant

$G = 1 - 1/4 + 1/9 - 1/16 + \cdots$

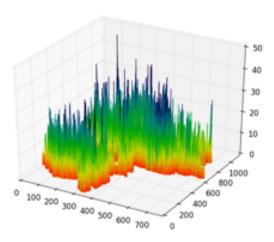


Figure: A walk on one million quad-bits of G with height showing frequency

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Paul Erdős (1913-1996)

"My brain is open"





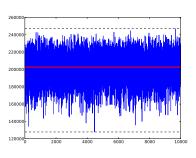
(a) Paul Erdős (Banff 1981. I was there)

(b) Émile Borel (1871–1956)

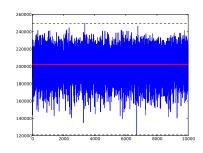
Figure: Two of my favourites. Consult MacTutor.

Number of points visited:

Again π looks random



(a) (Pseudo)random walks.



(b) Walks built by chopping up 10 billion digits of π .

Figure: Number of points visited by 10,000 million-steps base-4 walks.

Points visited by various base-4 walks

Number	Steps	Sites visited	Bounds on the expectation of sites visited by a random walk	
			Lower bound	Upper bound
Mean of 10,000 random walks	1,000,000	202,684	199,256	203,060
Mean of 10,000 walks on the digits of π	1,000,000	202,385	199,256	203,060
$\alpha_{2,3}$	1,000,000	95,817	199,256	203,060
$\alpha_{3,2}$	1,000,000	195,585	199,256	203,060
π	1,000,000	204,148	199,256	203,060
π	10,000,000	1,933,903	1,738,645	1,767,533
π	100,000,000	16,109,429	15,421,296	15,648,132
π	1,000,000,000	138,107,050	138,552,612	140,380,926
e	1,000,000	176,350	199,256	203,060
$\sqrt{2}$	1,000,000	200,733	199,256	203,060
log 2	1,000,000	214,508	199,256	203,060
Champernowne C_4	1,000,000	548,746	199,256	203,060
Rational number Q_1	1,000,000	378	199,256	203,060
Rational number Q_2	1,000,000	939,322	199,256	203,060

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Normal numbers need not be so "random" ...

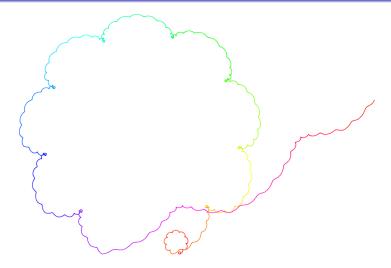


Figure: Champernowne $C_{10} = 0.123456789101112...$ (normal). Normalized distance to the origin: 15.9 (50,000 steps).

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Normal numbers need not be so "random" ...

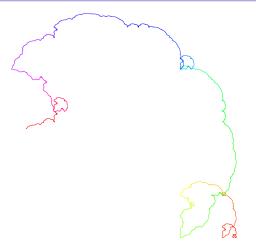


Figure: Champernowne $C_4 = 0.123101112132021...$ (normal). Normalized distance to the origin: 18.1 (100,000 steps). Points visited: 52760. Expectation: (23333, 23857).

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Normal numbers need not be so "random" ...

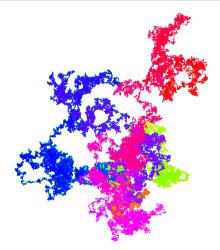


Figure: Stoneham $\alpha_{2,3} = 0.0022232032...4$ (normal base 4). Normalized distance to the origin: 0.26 (1,000,000 steps). Points visited: 95817. Expectation: (199256, 203060).

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Normal numbers need not be so "random" ...

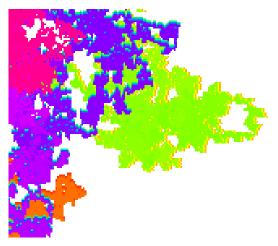
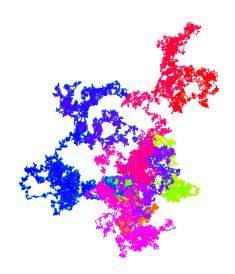


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$\alpha_{2,3}$ is 4-normal but not so "random"

ANIMATION



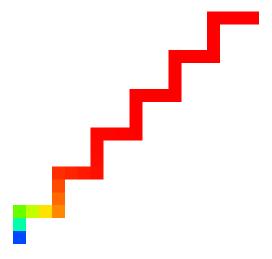


Figure: A pattern in the digits of $\alpha_{2,3}$ base 4. We show only positions of the walk after $\frac{3}{2}(3^n+1), \frac{3}{2}(3^n+1)+3^n$ and $\frac{3}{2}(3^n+1)+2\cdot 3^n$ steps, $n=0,1,\ldots,11$.

Theorem (Base-4 structure of Stoneham $\alpha_{2,3}$)

Denote by a_k the k^{th} digit of $\alpha_{2,3}$ in its base 4 expansion: $\alpha_{2,3} = \sum_{k=1}^{\infty} a_k/4^k$, with $a_k \in \{0,1,2,3\}$ for all k. Then, for all n = 0,1,2,...one has:

(i)
$$\sum_{k=\frac{3}{2}(3^n+1)}^{\frac{3}{2}(3^n+1)+3^n} e^{a_k\pi i/2} = \left\{ \begin{array}{ll} -i, & \text{n odd} \\ -1, & \text{n even} \end{array} \right.;$$

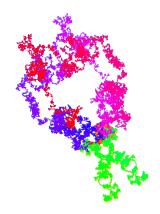
(ii)
$$a_k = a_{k+3^n} = a_{k+2\cdot 3^n}$$
 if $k = \frac{3(3^n + 1)}{2}, \frac{3(3^n + 1)}{2} + 1, \dots, \frac{3(3^n + 1)}{2} + 3^n - 1.$





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Likewise, $\alpha_{3.5}$ is 3-normal ... but not very "random" ANIMATION



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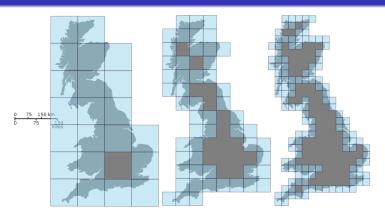
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Box-dimension:

Tends to '2' for a planar random walk

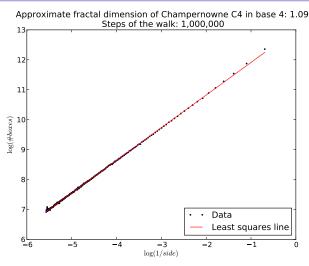


$$\label{eq:box-dimension} \begin{aligned} & \text{Box-dimension} = \lim_{\text{side} \to 0} \frac{\log(\text{\# boxes})}{\log(1/\text{side})} \end{aligned}$$

Norway is "frillier" — Hitchhiker's Guide to the Galaxy

Box-dimension:

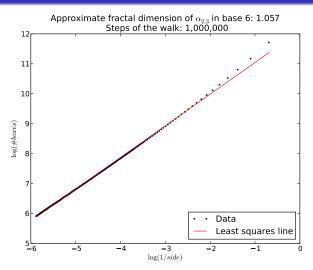
Tends to '2' for a planar random walk



Fractals: self-similar (zoom invariant) partly space-filling shapes (clouds & ferns not buildings & cars). Curves have dimension 1, squares dimension 2 PART I: Visual Theorems Digital Assistance PART II. Case Studies PART III: Randomness

Box-dimension:

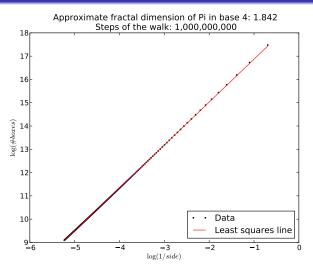
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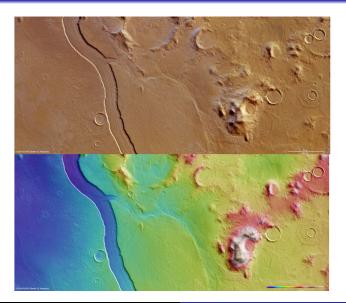
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Fractals everywhere

From Mars



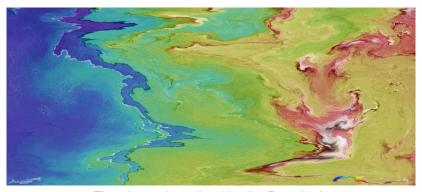


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Fractals everywhere

From Mars



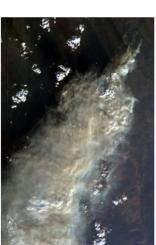


The picture fractalized by the Barnsley's http://frangostudio.com/frangocamera.html

Fractals everywhere

From Space







Fractals everywhere

 $1 \mapsto 3 \text{ or } 1 \mapsto 8 \text{ or } \dots$

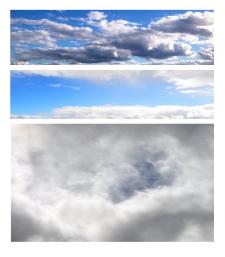




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Fractals everywhere

 $1 \mapsto 3 \text{ or } 1 \mapsto 8 \text{ or } \dots$







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Fractals everywhere

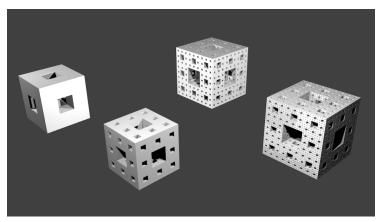
 $1 \mapsto 3 \text{ or } 1 \mapsto 8 \text{ or } \dots$



Pascal triangle modulo two [1] [1,1] [1,2,1] [1,3,3,1,] [1,4,6,4,1] [1,510,10,5,1] ... PART I: Visual Theorems Digital Assistance PART II. Case Studies PART III: Randomness Random-ish

Fractals everywhere

 $1 \mapsto 3 \text{ or } 1 \mapsto 8 \text{ or } \dots$



Steps to construction of a Sierpinski cube

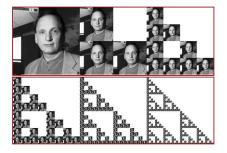
The Sierpinski Triangle

$$1 \mapsto 3 \mapsto 9$$



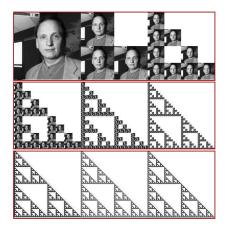
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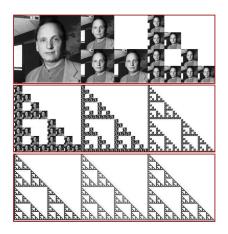
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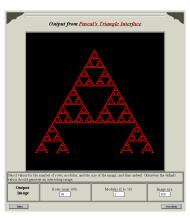
 $1 \mapsto 3 \mapsto 9$



The Sierpinski Triangle

 $1 \mapsto 3 \mapsto 9$





http:

//oldweb.cecm.sfu.ca/cgi-bin/organics/pascalform

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Three dimensional walks:

Using base six — soon on 3D screen

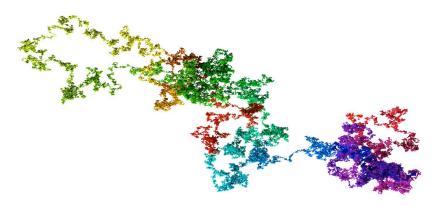


Figure: Matt Skerritt's 3D walk on π (base 6), showing one million steps. But 3D random walks are not recurrent.

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Three dimensional walks:

Using base six — soon on 3D screen

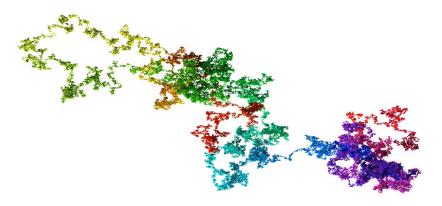


Figure: Matt Skerritt's 3D walk on π (base 6), showing one million steps. But 3D random walks are not recurrent.

"A drunken man will find his way home, a drunken bird will get lost forever." (Kakutani)

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Three dimensional printing:

3D everywhere



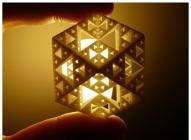


Figure: The future is here ...

www.digitaltrends.com/cool-tech/the-worlds-first-plane-created-entirely-by-3d-printing-takes-flight/ www.shapeways.com/shops/3Dfractals

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Chaos games:

Move half-way to a (random) corner

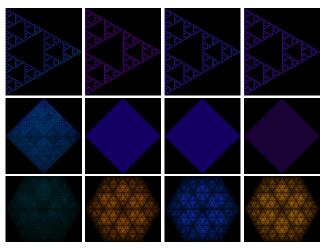


Figure: Coloured by frequency — leads to random fractals.

Row 1: Champernowne C_3 , $\alpha_{3,5}$, random, $\alpha_{2,3}$. Row 2: Champernowne C_4 ,

 π , random, $\alpha_{2,3}$. Row 3: Champernowne C_6 , $\alpha_{3,2}$, random, $\alpha_{2,3}$.

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Automatic numbers: Thue-Morse and Paper-folding

Automatic numbers are never normal. They are given by simple but fascinating rules...giving structured/boring walks:



Figure: Paper folding. The sequence of left and right folds along a strip of paper that is folded repeatedly in half in the same direction. Unfold and read 'right' as '1' and 'left' as '0': 1 0 1 1 0 0 1 1 1 0 0 1 0 0

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Thue-Morse constant (transcendental; 2-automatic, hence nonnormal):

$$TM_2 = \sum_{n=1}^{\infty} \frac{1}{2^{t(n)}}$$
 where $t(0) = 0$, while $t(2n) = t(n)$ and $t(2n+1) = 1 - t(n)$

 $0.01101001100101101001011001101001\dots$

Automatic numbers: Thue-Morse and Paper-folding

Automatic numbers are never normal. They are given by simple but fascinating rules...giving structured/boring walks:

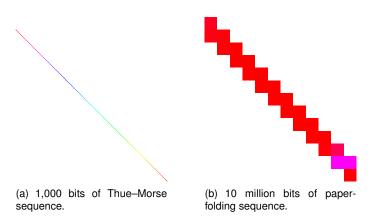


Figure: Walks on two automatic and so nonnormal numbers.

Automatic numbers:

Turtle plots look great!



(a) Ten million digits of the paperfolding sequence, rotating 60°.



(b) One million digits of the paperfolding sequence, rotating 120° (a dragon curve).



(c) 100,000 digits of the Thue-Morse sequence, rotating 60° (a Koch snowflake).



(d) One million digits of π , rotating 60°.

Figure: Turtle plots on various constants with different rotating angles in base 2—where '0' yields forward motion and '1' rotation by a fixed angle.

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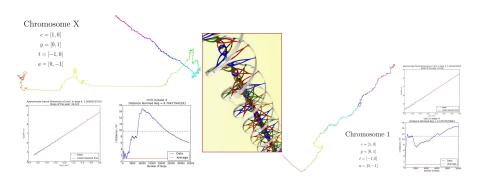
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Genomes as walks:

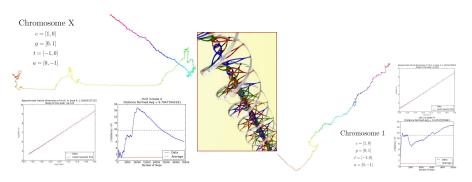
We are all base 4 numbers (ACGT/U)



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Genomes as walks:

We are all base 4 numbers (ACGT/U)

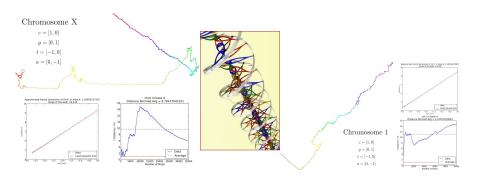


The X Chromosome (34K) and Chromosome One (10K).

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Genomes as walks:

We are all base 4 numbers (ACGT/U)



The X Chromosome (34K) and Chromosome One (10K).

Chromosomes look less like π and more like concatenation numbers?

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DNA for Storage:

We are all base 4 numbers (ACGT/U)

News Science Biochemistry and molecular biology

Shakespeare and Martin Luther King demonstrate potential of DNA storage

All 154 Shakespeare sonnets have been spelled out in DNA to demonstrate the vast potential of genetic data storage

lan Sample, science correspondent The Guardian, Thursday 24 January 2013 Jump to comments (...)



When written in DNA, one of Shakespeare's sonnets weighs 0.3 millionths of a millionth of a gram. Photograph: Oli Scarff/Getty

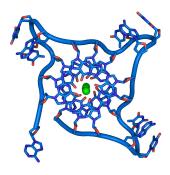


Figure: The potential for DNA storage (L) and the quadruple helix (R)

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The end

with some fractal dessert



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The end

with some fractal dessert



Thank you

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