

Fundamentals

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Principles of Complex Systems, Vols. 1, 2, 3D, 4 Fourever, V for Vendetta

Prof. Peter Sheridan Dodds

Computational Story Lab | Vermont Complex Systems Institute
University of Vermont | Santa Fe Institute



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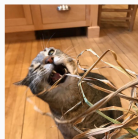
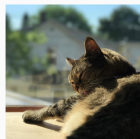
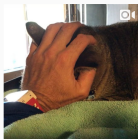
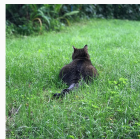
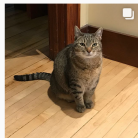
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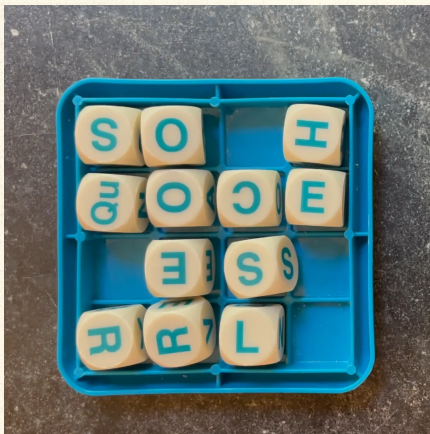
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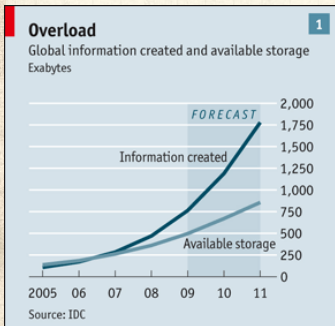
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The Boggoracle Speaks:



Around 2010: Data becomes Big Data because it's about us:



Exponential growth: \sim 60% per year.

Big Data Science over time:

Vera C. Rubin Observatory: 200,000 photos, 1.28 petabytes/year.

Large Hadron Collider, 2024: 1 petabyte/second.

\sim 2 trillion photos taken in 2024

Over 10 billion photos shared each day

Large language models, 2024: Up to 10^{13} tokens.



No really, that's a lot of data

Data inflation

2

Unit	Size	What it means
Bit (b)	1 or 0	Short for "binary digit", after the binary code (1 or 0) computers use to store and process data
Byte (B)	8 bits	Enough information to create an English letter or number in computer code. It is the basic unit of computing
Kilobyte (KB)	1,000, or 2^{10} , bytes	From "thousand" in Greek. One page of typed text is 2KB
Megabyte (MB)	1,000KB; 2^{20} bytes	From "large" in Greek. The complete works of Shakespeare total 5MB. A typical pop song is about 4MB
Gigabyte (GB)	1,000MB; 2^{30} bytes	From "giant" in Greek. A two-hour film can be compressed into 1-2GB
Terabyte (TB)	1,000GB; 2^{40} bytes	From "monster" in Greek. All the catalogued books in America's Library of Congress total 15TB
Petabyte (PB)	1,000TB; 2^{50} bytes	All letters delivered by America's postal service this year will amount to around 5PB. Google processes around 1PB every hour
Exabyte (EB)	1,000PB; 2^{60} bytes	Equivalent to 10 billion copies of <i>The Economist</i>
Zettabyte (ZB)	1,000EB; 2^{70} bytes	The total amount of information in existence this year is forecast to be around 1.2ZB
Yottabyte (YB)	1,000ZB; 2^{80} bytes	Currently too big to imagine

Source: *The Economist* The prefixes are set by an intergovernmental group, the International Bureau of Weights and Measures. Yotta and Zetta were added in 1991; terms for larger amounts have yet to be established.

Estimates of internet data, past, present, future:¹

Created:



2010: ~ 2 zettabytes

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¹<https://www.statista.com/statistics/871513/worldwide-data-created/>

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-  2023: ~ 120 zettabytes
-  2024: ~ 150 zettabytes



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
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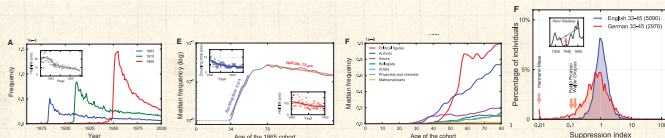
 Video accounts for about 50% of all data.



¹<https://www.statista.com/statistics/871513/worldwide-data-created/> 

Big Data—Culturomics:

“Quantitative analysis of culture using millions of digitized books” by Michel et al., Science, 2011 ^[14]



 Google Books ngram viewer 

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Deceased: <http://www.culturomics.org/> 





Internet-scale data sets can be overly **exciting**.




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Witness:


 The End of Theory: The Data Deluge Makes the Scientific Theory Obsolete (Anderson, Wired) 




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 “The Unreasonable Effectiveness of Data,”
Halevy et al. ^[10].


 c.f. Wigner’s “The Unreasonable Effectiveness of Mathematics in the Natural Sciences” ^[25]




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But:


 For scientists, description is only part of the battle.




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But:

 For scientists, description is only part of the battle.

 We still need to **understand**.



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Lord Kelvin (possibly):



“To measure is to know.”



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Bonus:



“X-rays will prove to be a hoax.”



“There is nothing new to be discovered in physics now, All that remains is more and more precise measurement.”



“My beard will always be cool.”



A brief history of measuring time:



Megaliths for Big Time



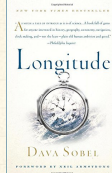
Sundials, 1500 BC, Egypt (solid for over 2000 years)





Escapements (200s), Hourglasses (1300s?), Pendulum clocks (Galileo, 1500s)



Chronometers, 1700s:





“Longitude: The True Story of a Lone Genius Who Solved the Greatest Scientific Problem of His Time”  
by Dava Sobel (2007). ^[22]




Billionths of a second accuracy: Atomic clocks (Lord Kelvin, 1879)





Some very, very silly units of measurement courtesy of the
Imperial system :

 22 yards in a chain = 1 cricket pitch, 100 links in a chain, 10
chains in a furlong, 80 chains in a mile.




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Also:





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 14 pounds in a stone.

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Also:

 Fahrenheit, Celcius, and Kelvin.

 The entire metric system.



References



From [https://en.wikipedia.org/wiki/Barleycorn_\(unit\)](https://en.wikipedia.org/wiki/Barleycorn_(unit))

Measuring temperature was thought impossible:

The properties measured by our instruments usually begin as subjective judgments. Temperature is a good example. People were aware of variations in temperature long before there were any objective measurements of temperature. Judgments of temperature are imperfectly correlated among different persons, or even the same person at different times, depending on the humidity, the person's activity level and age, surrounding air currents, and so on. The idea that anything as subtle and complex as all the manifestations of changes in temperature could be measured and quantified on a single numerical scale was scoffed at as impossible, even by the leading philosophers of the sixteenth century.

The first thermometer invented by Galileo in 1592 did not go far in dispelling the notion that temperature was inherently unmeasurable, because the earliest thermometers, for about their first hundred years, were so imperfect as to make it possible for those who wished to do so to argue that no one could ever succeed in measuring temperature. Temperature was then confounded with all the subtleties of subjective judgment, which easily seem incompatible with a single numerical scale of measurement. How could the height of a column of mercury in a glass tube possibly reflect the rich varieties of temperature—damp cold, dank cold, frosty cold, crisp cold, humid heat, searing heat, scalding heat, dry heat, feverish heat, prickly heat, and so on?

From “Bias in Mental Testing”, Arthur Jensen, 1980^[12]
per @SilverVVulpes[↗]: Also: Inventing Temperature, Hasok Chang,
2004^[5]



Measuring temperature was thought impossible:

The early thermometers were inconsistent, both with themselves and with each other. Because they consisted of open-ended glass tubes, they were sensitive to changes in barometric pressure as well as to temperature. And there were problems of calibration, such as where to locate the zero point and how to divide the column of mercury into units. It was believed, incorrectly, that all caves had the same temperature, so thermometers were calibrated in caves. The freezing and boiling points of water were also used in calibration, but, as these vary with impurities in the water and the barometric pressure, the calibration of different thermometers at different times and places resulted in thermometers that failed to correlate perfectly with one another in any given instance. They lacked reliability, as we now would say.

All the while, no one knew what temperature is in a theoretical or scientific sense. There was no theory of thermodynamics that could explain temperature phenomena and provide a complete scientific rationale for the construction and calibration of thermometers. Yet quite adequate and accurate thermometers, hardly differing from those we use today, were eventually developed by the middle of the eighteenth century. Thus the objective measurement of temperature considerably preceded the development of an adequate theory of temperature and heat, and necessarily so, as the science of thermodynamics could not possibly have developed without first having been able to quantify or measure the temperatures of liquids, gasses, and other substances independently of

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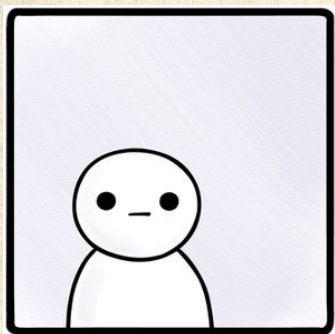
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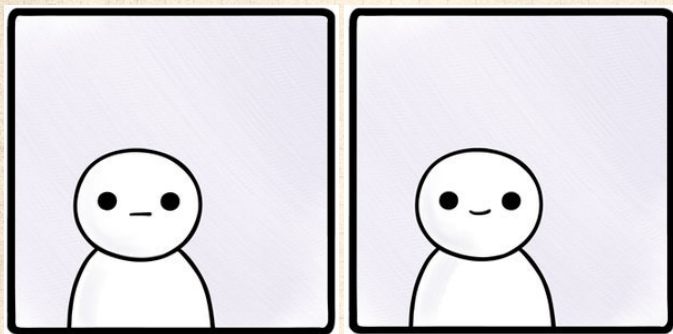
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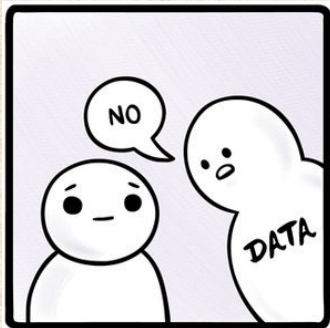
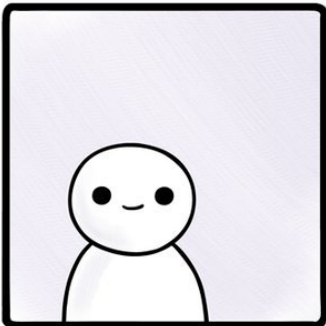
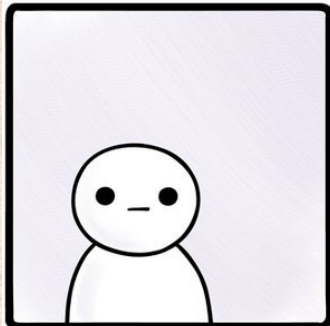
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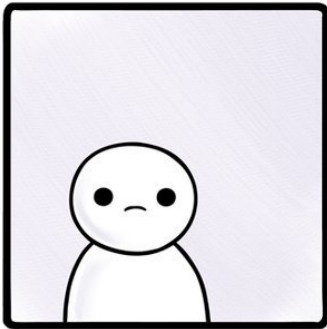
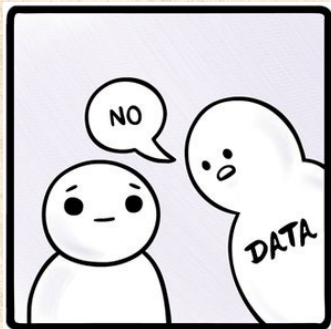
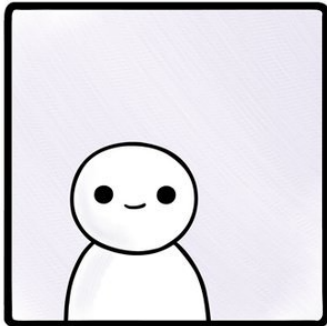
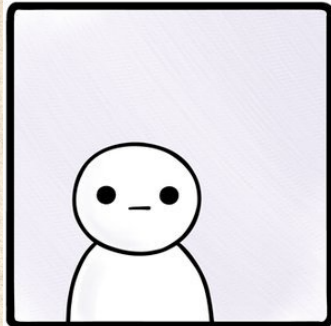
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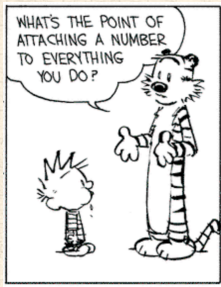
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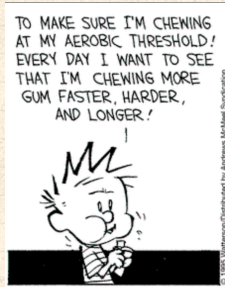
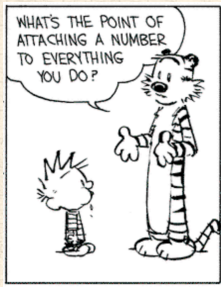
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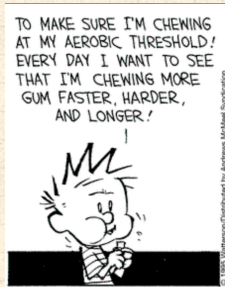
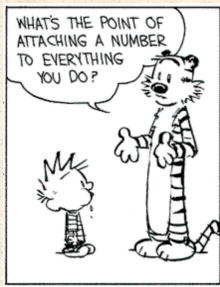
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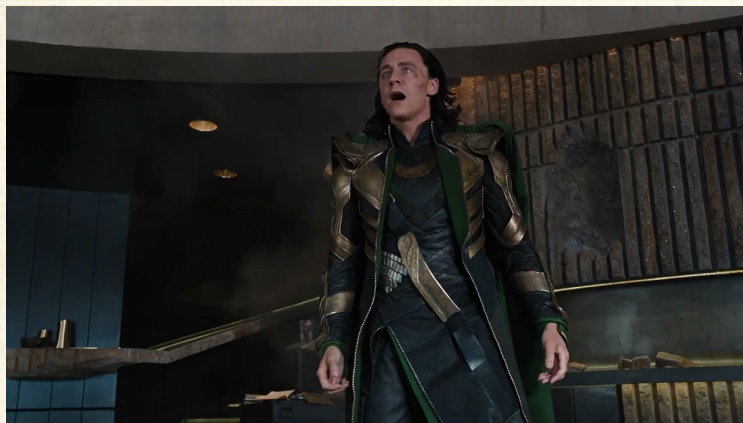
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Data Angry. Data Smash.  



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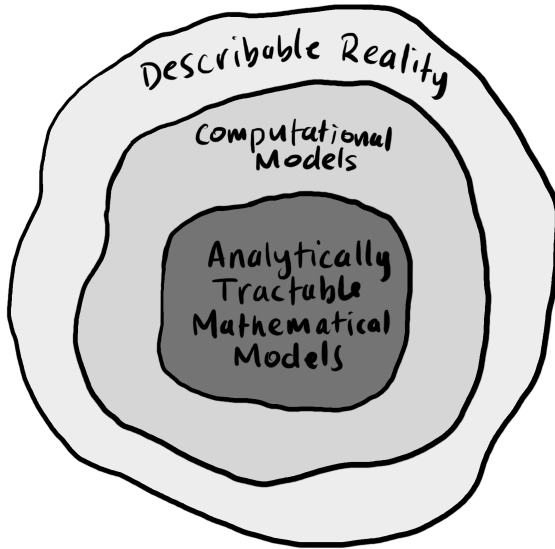
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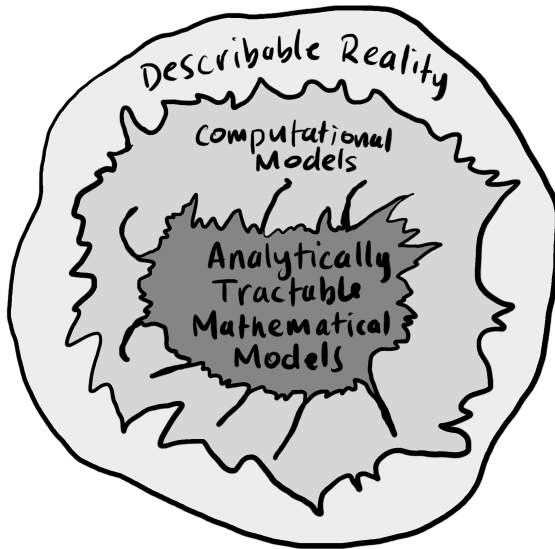
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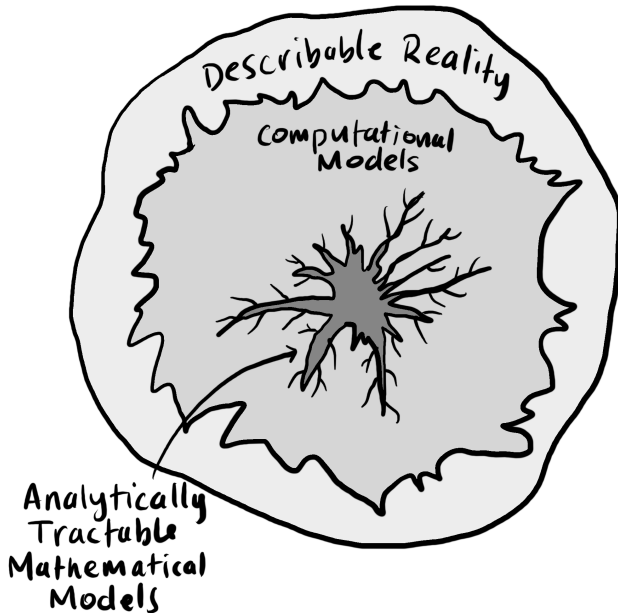
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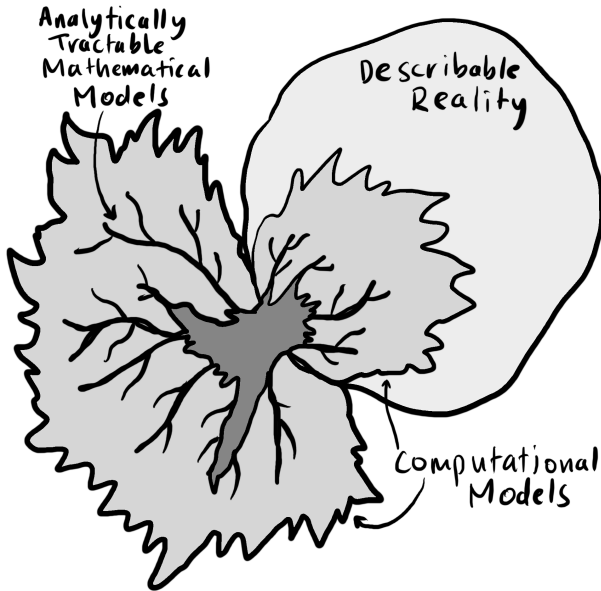
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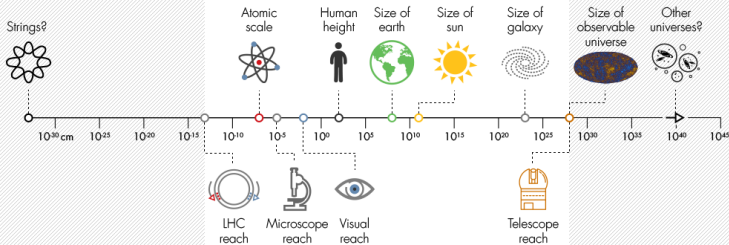


Limits of testability and happiness in Science:

From A Fight for the soul of Science [↗](#) in Quanta Magazine
(2016/02):

The Ends of Evidence

Humans can probe the universe over a vast range of scales (white area), but many modern physics theories involve scales outside of this range (grey).



The Newness of being a Scientist (1833 on):

Google books Ngram Viewer

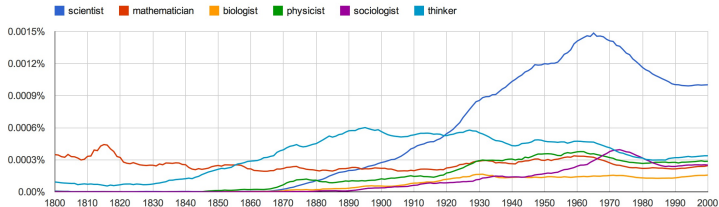
Graph these [case-sensitive](#) comma-separated phrases: scientist, mathematician, biologist, physicist, sociologist



between 1800 and 2000 from the corpus English with smoothing of 3

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Consilience:

<https://www.npr.org/2010/05/21/127037417/how-the-word-scientist-came-to-be> 

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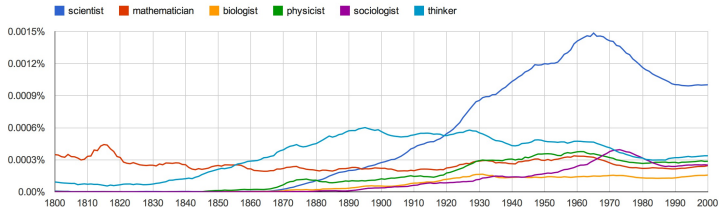
Graph these [case-sensitive](#) comma-separated phrases:



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 “Scientists are the people who ask a question about a phenomenon and proceed to **systematically** go about answering the question themselves. They are by nature curious, creative and well organized.”

Consilience:

<https://www.npr.org/2010/05/21/127037417/how-the-word-scientist-came-to-be> 

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Please do not measure complex systems with one number:



This is real —someone having some fun.



Obtained from this tweet. 



Sadness for Buckingham (if Buckingham has no sense of humor).



The conceptual trapping pit of a single scale:



Lure of simplicity: Comparisons and rankings are easy.

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²Phrenology  was at least a 2-d map; see also palmistry 

The conceptual trapping pit of a single scale:



Lure of simplicity: Comparisons and rankings are easy.



A single scale measure is very appealing, very hard to resist ...

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-  Salary!
-  stock market valuation for corporations














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
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-  stock market valuation for corporations
-  Complexity of civilizations ^[23]
-  A 1-d axis for political ideologies (a spatial metaphor trap, thanks France! )



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Personality distributions:

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“A Theory of the Emergence, Persistence, and Expression of Geographic Variation in Psychological Characteristics” ↗

Rentfrow, Gosling, and Potter,
Perspectives on Psychological Science, 3, 339–369,
2008. ^[17]

Five Factor Model (FFM):



Extraversion [E]



Agreeableness [A]



Conscientiousness [C]



Neuroticism [N]



Openness [O]

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Bigger concern: mass manipulation.



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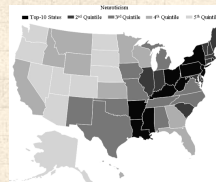
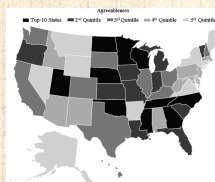
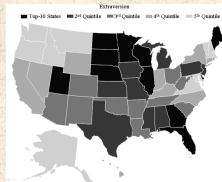
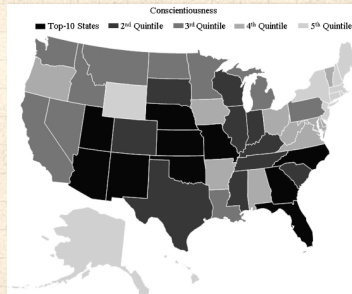
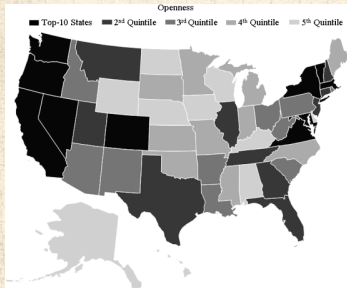
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Dungeons & Dragons' full embrace of complexity:

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A detailed D&D character sheet template with a tree-like layout. The sheet is divided into several main sections:

- CHARACTER NAME**: A large box at the top left.
- LEVEL**: A circle with a line through it, next to the character name.
- CLASS**: A box below the level.
- EXPERIENCE POINTS**: A box to the right of the class.
- COMBAT**: A central box with a large 'X' through it.
- ARMOR CLASS**: A box below the combat section.
- SPEED**: A box to the right of the armor class.
- INITIATIVE**: A box to the right of the speed.
- CURRENT HP**: A box below the initiative.
- TEMP HP**: A box to the right of the current HP.
- HIT POINT MAXIMUM**: A box to the right of the temp HP.
- DEATH SAVED**: A box below the hit point maximum.
- HIT DICE**: A box to the left of the death saved.
- ATTACK BONUS**: A box below the hit dice.
- DAMAGE TYPE**: A box to the right of the attack bonus.
- IDEALS**: A box to the right of the damage type.
- BONDS**: A box below the ideals.
- FLAWS**: A box to the right of the bonds.
- EQUIPMENT**: A box at the bottom left.
- CHARACTERISTICS**: A vertical column of boxes on the left side, including Strength, Dexterity, Constitution, Intelligence, Wisdom, and Charisma.
- PROFICIENCY BONUS**: A box to the right of the characteristics.
- PROFICIENCIES & LANGUAGES**: A box at the bottom left.
- SAVING THROWS**: A box at the bottom center, including STR, DEX, CON, INT, WIS, and CHA.
- ACROBATICS (DEX)**, **ANIMAL HANDLING (WIS)**, **ARCANA (INT)**, **ATHLETICS (STR)**, **DECEPTION (CHA)**, **DISGUISE (DEX)**, **INSIGHT (WIS)**, **INTIMIDATION (CHA)**, **INVESTIGATION (INT)**, **MEDICINE (WIS)**, **NATURE (WIS)**, **PERCEPTION (WIS)**, **PERFORMANCE (STR)**, **PERMISSION (CHA)**, **RELIGION (INT)**, **SIGHT OF HEAVEN (WIS)**, **STEALTH (DEX)**, and **SURVIVAL (WIS)** are listed in a column on the right side.

From [here](#)



Dungeons & Dragons—Two alignment axes for character:



Law–Chaos
(vertical) and
Good–Evil
(horizontal).

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³From this Reddit thread, where, naturally, the choices are enthusiastically debated.

Please just don't do it.  

[door slams]

The Office
S7E15: The Search
February 3, 2011



The wisdom of Jason Mendoza :

The Good Place
S2E04: Team Cockroach
October 5, 2017



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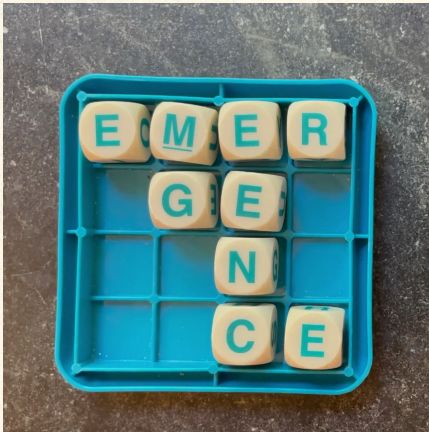
Modeling

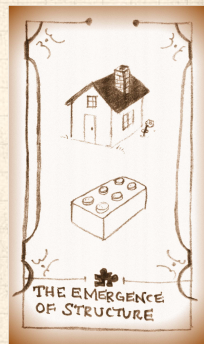
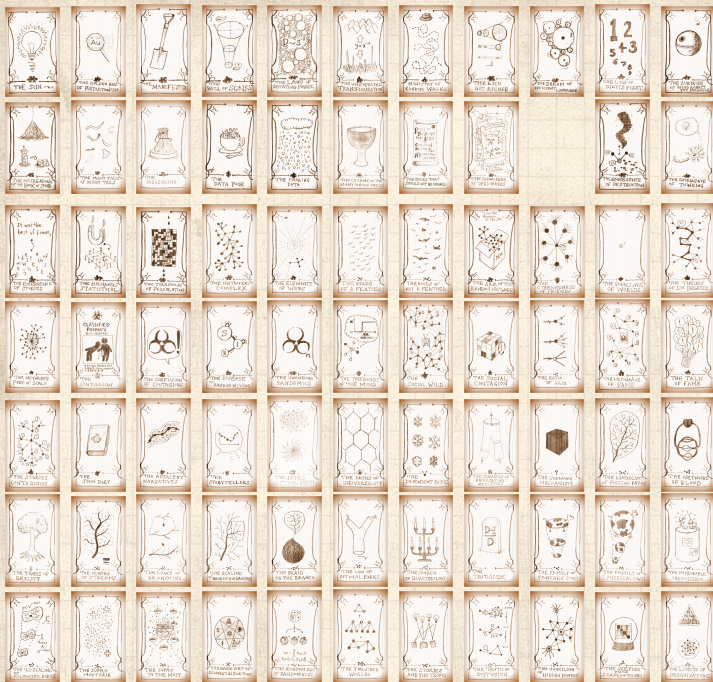
Statistical Mechanics

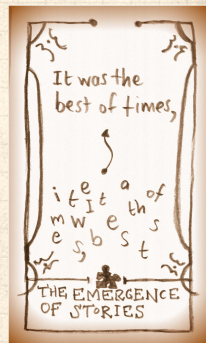
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The Boggoracle Speaks:







Emergence:

The Wikipedia on Emergence (2006):

“In philosophy, systems theory and the sciences, emergence refers to the way complex systems and patterns arise out of a multiplicity of relatively simple interactions.



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The philosopher G. H. Lewes^[13] is thought to have first described emergence in 1875.



Fireflies \Rightarrow Synchronized Flashes:



Film: Sir David Attenborough, BBC.

Voiceover: Steve Strogatz on Radiolab's Emergence, S1E3 .



Emergence:

There's no tornado in a water molecule,
no financial collapse in a dollar bill,
no Dungeon Crawler Carl in the letters of the alphabet,
no love in a carbon atom.

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Examples:


 Fundamental particles \Rightarrow Life, the Universe, and Everything



 Genes \Rightarrow Organisms


 Neurons etc. \Rightarrow Brain \Rightarrow Thoughts


 People \Rightarrow Language, and rules of language


 People \Rightarrow Religion, Society, Collective behavior

 People \Rightarrow Internet

  ? \Rightarrow space ^[?]

 ? \Rightarrow gravity

 ? \Rightarrow time ^[15, 2, 1]


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Emergence:

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Friedrich Hayek  (Economist/Philosopher/Nobelist):

 Markets, legal systems, political systems are emergent and not designed.



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 **Hierarchies** arise once problems are solved. ^[8]




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 Dewey Decimal System versus tagging.



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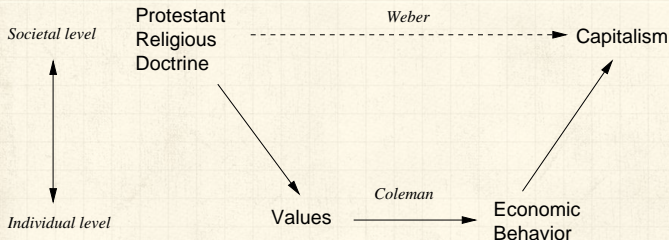
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James Coleman  in *Foundations of Social Theory*:



Understand macrophenomena arises from microbehavior which in turn depends on macrophenomena. ^[6]



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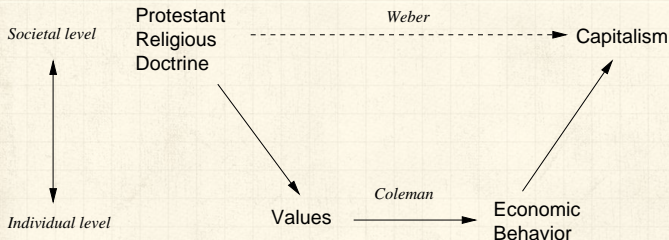
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More on Coleman [here](#) .



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Thomas Schelling  (Economist/Nobelist):



“Micromotives and Macrobehavior” [20]




Segregation [18, 21]

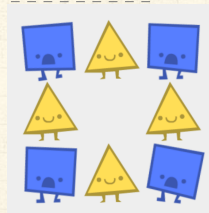


Wearing hockey helmets [19]



Seating choices

Vi Hart and Nicky Case’s
Polygon-themed
visualization :



The emergence of taste:

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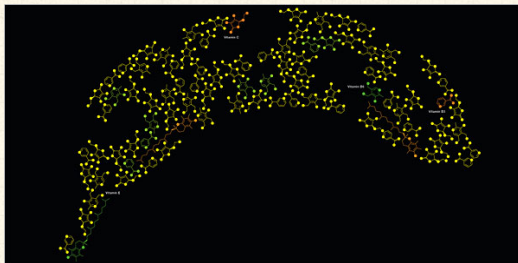
References



Molecules \Rightarrow Ingredients \Rightarrow Taste



See Michael Pollan's article on nutritionism  in the New York Times, January 28, 2007.



nytimes.com 



Reductionism

Reductionism and food:

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Reductionism and food:

 Pollan: “even the simplest food is a hopelessly complex thing to study, a virtual wilderness of chemical compounds, many of which exist in complex and dynamic relation to one another...”



Reductionism and food:

 Pollan: “even the simplest food is a hopelessly complex thing to study, a virtual wilderness of chemical compounds, many of which exist in complex and dynamic relation to one another...”

 “So ... break the thing down into its component parts and study those one by one, even if that means ignoring complex interactions and contexts, as well as the fact that the whole may be more than, or just different from, the sum of its parts. This is what we mean by reductionist science.”





“people don’t eat nutrients, they eat foods, and foods can behave very differently than the nutrients they contain.”



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“people don’t eat nutrients, they eat foods, and foods can behave very differently than the nutrients they contain.”



Studies suggest diets high in fruits and vegetables help prevent cancer.



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- Studies suggest diets high in fruits and vegetables help prevent cancer.
- So... find the nutrients responsible and eat more of them



- “people don’t eat nutrients, they eat foods, and foods can behave very differently than the nutrients they contain.”
- Studies suggest diets high in fruits and vegetables help prevent cancer.
- So... find the nutrients responsible and eat more of them
- But “in the case of beta carotene ingested as a supplement, scientists have discovered that it actually increases the risk of certain cancers. Oops.”



Thyme's known antioxidants:

4-Terpineol, alanine, anethole, apigenin, ascorbic acid, beta carotene, caffeic acid, camphene, carvacrol, chlorogenic acid, chrysoeriol, eriodictyol, eugenol, ferulic acid, gallic acid, gamma-terpinene isochlorogenic acid, isoeugenol, isothymonin, kaempferol, labiatic acid, lauric acid, linalyl acetate, luteolin, methionine, myrcene, myristic acid, naringenin, oleanolic acid, p-coumaric acid, p-hydroxy-benzoic acid, palmitic acid, rosmarinic acid, selenium, tannin, thymol, tryptophan, ursolic acid, vanillic acid.



[cnn.com]



“It would be great to know how this all works, but **in the meantime** we can enjoy thyme in the knowledge that it probably doesn’t do any harm (since people have been eating it forever) and that it may actually do some good (since people have been eating it forever) and that even if it does nothing, we like the way it tastes.”



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Gulf between theory and practice (see baseball and bumblebees).



This is a Collateralized Debt Obligation:

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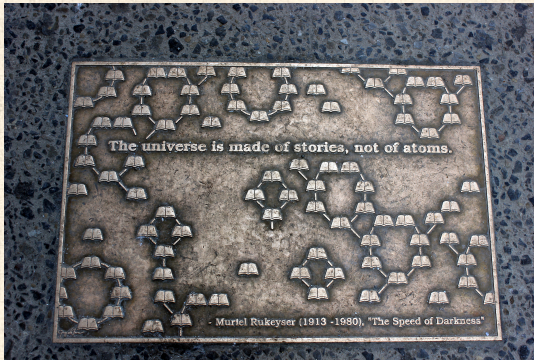
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“The Universe is made of stories, not of atoms.”

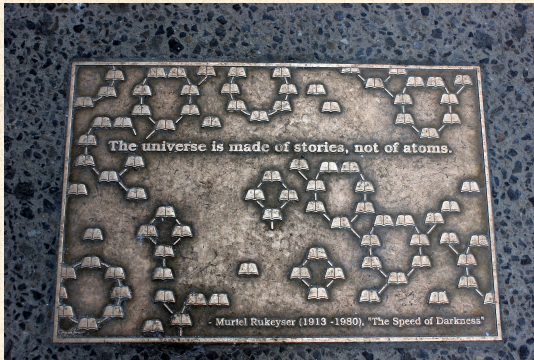


From “The Speed of Darkness” (1968) by Muriel Rukeyser [↗](#)





“The Universe is made of stories, not of atoms.”



From “The Speed of Darkness” (1968) by [Muriel Rukeyser](#)



Quoted by Metatron in Supernatural, Meta Fiction, S9E18.



(Sir Terry) Pratchett's Narrativium .

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Emergence

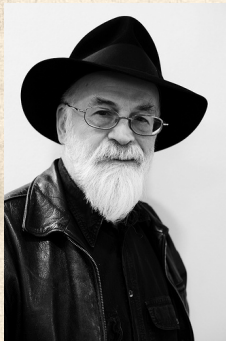
Self-Organization

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(Sir Terry) Pratchett's Narrativium :

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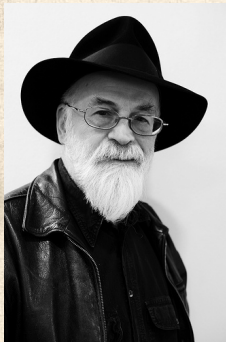
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“The most common element on the disc, although not included in the list of the standard five: earth, fire, air, water and surprise. It ensures that everything runs properly as a story.”



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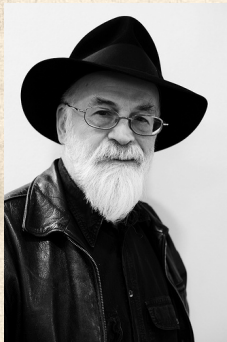
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“The most common element on the disc, although not included in the list of the standard five: earth, fire, air, water and surprise. It ensures that everything runs properly as a story.”



“A little narrativium goes a long way: the simpler the story, the better you understand it. Storytelling is the opposite of reductionism: 26 letters and some rules of grammar are no story at all.”



Higher complexity:



Many system scales (or levels)
that interact with each other.

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Potentially much harder to explain/understand.



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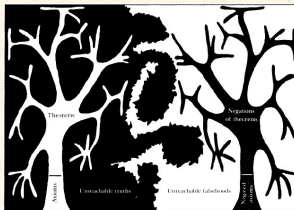


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Even mathematics: [9]



Gödel's Theorem ↗:
we can't prove every
theorem that's true ...



“Gödel, Escher, Bach” [11]



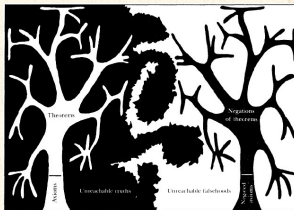
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- Suggests a **strong form of emergence**:



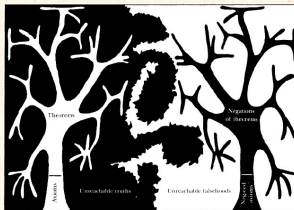
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- Suggests a **strong form of emergence**: Some phenomena cannot be analytically deduced from elementary aspects of a system.



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Roughly speaking, there are **two types** of **emergence**:



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Roughly speaking, there are **two types** of **emergence**:

I. Weak emergence:

System-level phenomena is different from that of its constituent parts yet can be connected theoretically.



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Roughly speaking, there are **two types** of **emergence**:

I. Weak emergence:

System-level phenomena is different from that of its constituent parts yet can be connected theoretically.

II. Strong emergence:

System-level phenomena fundamentally cannot be deduced from how parts interact.



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Reductionist techniques can explain weak emergence.



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 Reductionist techniques can explain weak emergence.

 Magic explains strong emergence. ^[4]



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- Reductionist techniques can explain weak emergence.
- Magic explains strong emergence. ^[4]
- But: maybe **magic** should be interpreted as an **inscrutable yet real mechanism** that cannot ever be **simply described**.



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- Reductionist techniques can explain weak emergence.
- Magic explains strong emergence. ^[4]
- But: maybe magic should be interpreted as an inscrutable yet real mechanism that cannot ever be simply described.
- Gulp.





Listen to Steve Strogatz, Hod Lipson, and Michael Schmidt (Cornell) in the last piece on Radiolab's show 'Limits' [↗](#) (April 5, 2010). Starts at 50:30.



(El Bibliomata/flickr)

Dr. Steve Strogatz wonders if we've reached the limits of human scientific understanding, and should soon turn the reins of research over to robots. Cold, calculating robots. Then, Dr. Hod Lipson and Michael Schmidt walk us through the workings of a revolutionary computer program that they developed—a program that can deduce mathematical relationships in nature, through simple observation. The catch? As Dr. Gurol Suel explains, the program gives answers to complex biological questions that we humans have yet to ask, or even to understand.

TAGS: mind bending


Pair with some slow TV [↗](#)

Bonus: Mike Schmidt's talk on Eureka [↗](#) at

VCSC's increasingly ancient 2011 TEDx event "Big Data, Big Stories." [↗](#)



Definitions

“Self-organization  is a process in which the internal organization of a system, normally an open system, increases in complexity without being guided or managed by an outside source.”

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Definitions

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Definitions


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Examples:


 Molecules/Atoms liking each other →
Gases, liquids, and solids.

 Spin alignment → Magnetization.


 Protein folding.


 Imitation → Herding, flocking, mobs, ...





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Fundamental question: how likely is ‘complexification’?



Tools and techniques:

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Tools and techniques:




Differential equations, difference equations, linear algebra, stochastic models.



Tools and techniques:

 Differential equations, difference equations, linear algebra, stochastic models.

 Statistical techniques for comparisons and descriptions.







Tools and techniques:

- 🧱 Differential equations, difference equations, linear algebra, stochastic models.
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 -  Artisanal toy models
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Key advance (more soon):



-  Representation of **complex interaction patterns** as **complex networks**.



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Key advance (more soon):

-  Representation of complex interaction patterns as complex networks.
-  The driver: Massive amounts of Data



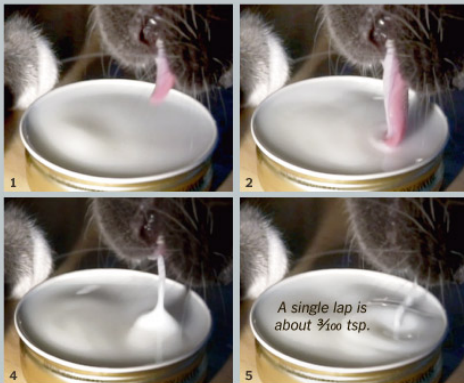
A whimsical-powerful example of real science:

“How Cats Lap: Water Uptake by *Felis catus*” 

Reis et al., *Science*, 2010.


A Study of Cat Lapping

Adult cats and dogs are unable to create suction in their mouths and must use their tongues to drink. A dog will scoop up liquid with the back of its tongue, but a cat will only touch the surface with the smooth tip of its tongue and pull a column of liquid into its mouth.



Source: Science

THE NEW YORK TIMES: IMAGES FROM VIDEO BY ROMAN STOCKER, SUNGHWAN JUNG, JEFFREY M. ARISTOFF AND PEDRO M. REIS

Amusing interview 

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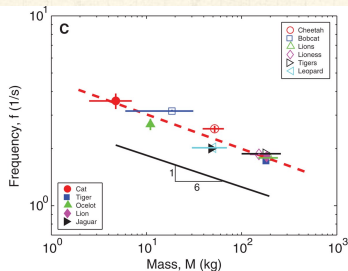
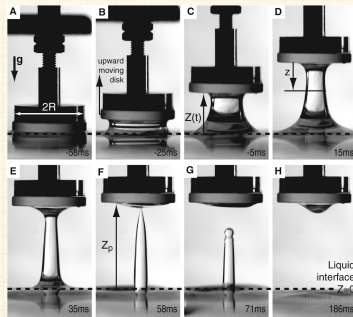


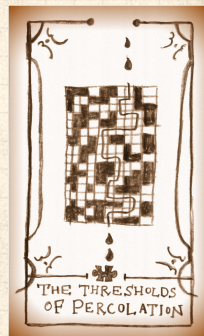


Another great, great moment in scaling:

$$f \sim M^{-1/6}$$

The balance of inertia and gravity yields a prediction for the lapping frequency of other felines. Assuming isometry within the Felidae family (i.e., that lapping height H scales linearly with tongue width R and animal mass M scales as R^3), the finding that Fr^* is of order one translates to the prediction $f \sim R^{-1/2} \sim M^{-1/6}$. Isometry or marginally positive allometry among the Felidae has been demonstrated for skull (20, 21) and limb bones (22). Although variability by function can lead to departures from isometry in interspecific scalings (23), reported variations within the Felidae (23, 24) only minimally affect the predicted scaling $f \sim M^{-1/6}$. We tested this $-1/6$ power-law dependence by measuring the lapping frequency for eight species of felines, from videos acquired at the Zoo New England or available on YouTube (16). The lapping frequency was observed to decrease with animal mass as $f = 4.6 M^{-0.181 \pm 0.024}$ (f in s^{-1} , M in kg) (Fig. 4C), close to the predicted $M^{-1/6}$. This close agreement suggests that the domestic cat's inertia- and gravity-controlled lapping mechanism is conserved among felines.





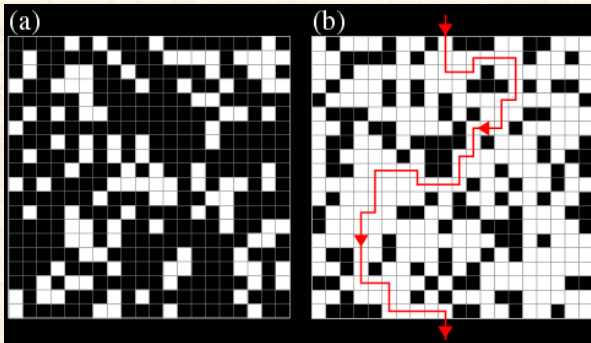


Statistical Mechanics is “a science of collective behavior.”



Simple rules give rise to collective phenomena.

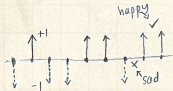
Percolation: 



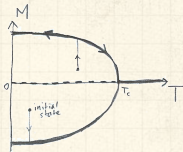
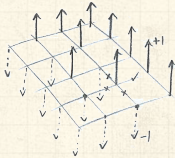
Snares from Michael Gastner's page on percolation [no longer online]



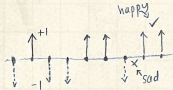
The Ising Model of a ferromagnet:



Each atom is assumed to have a local spin that can be **up** or **down**: $S_i = \pm 1$.



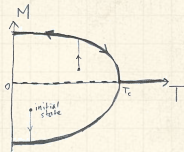
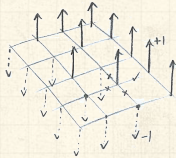
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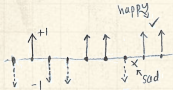
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Spins are assumed to be arranged on a lattice.



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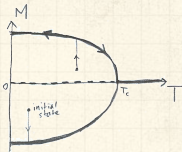
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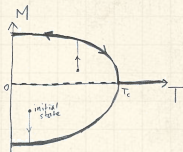
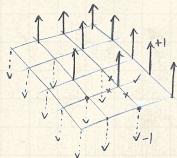
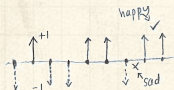
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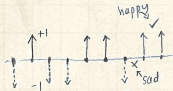
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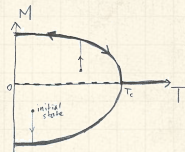
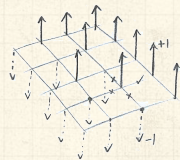
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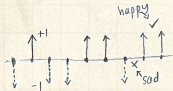
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The drosophila of statistical mechanics.



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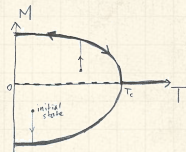
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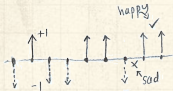
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Criticality: Power-law distributions at critical points.



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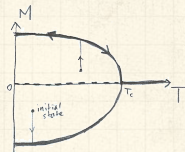
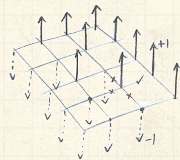
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Example 2-d Ising model simulation:

<https://mattbierbaum.github.io/ising.js/>



Phase diagrams

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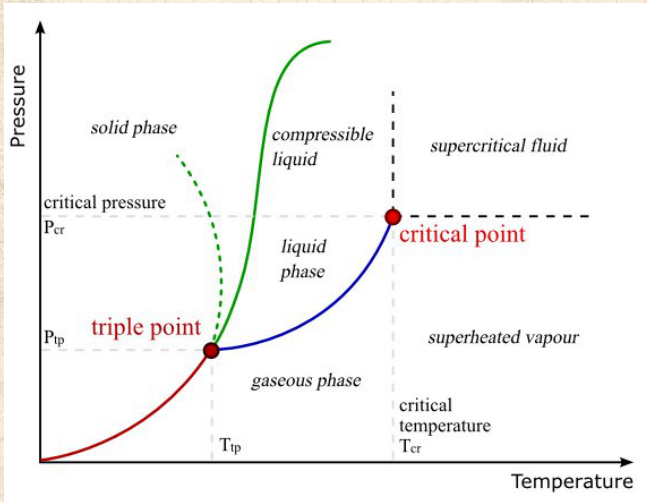
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Qualitatively distinct macro states.



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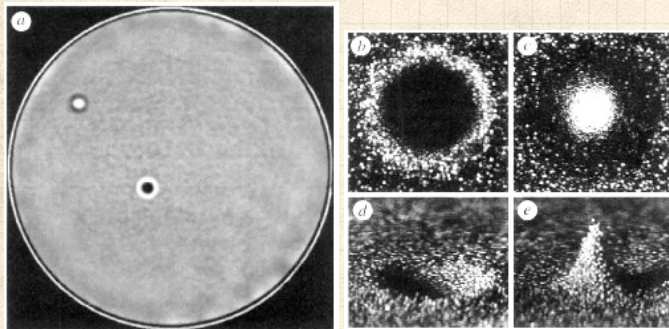
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Oscillons, bacteria, traffic, snowflakes, ...



Umbanhowar et al., *Nature*, 1996 ^[24]



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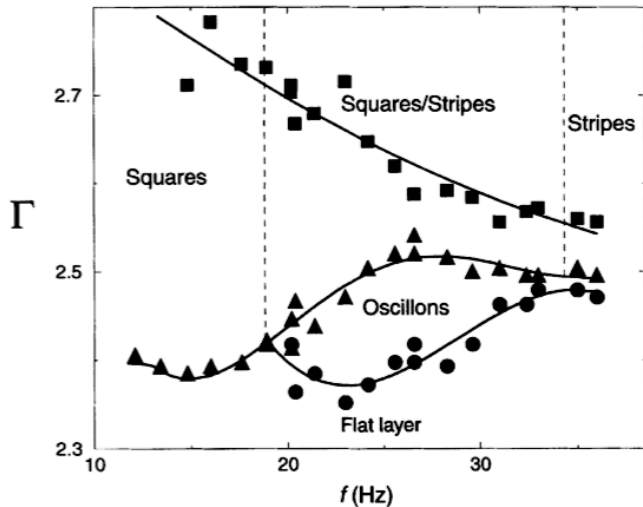
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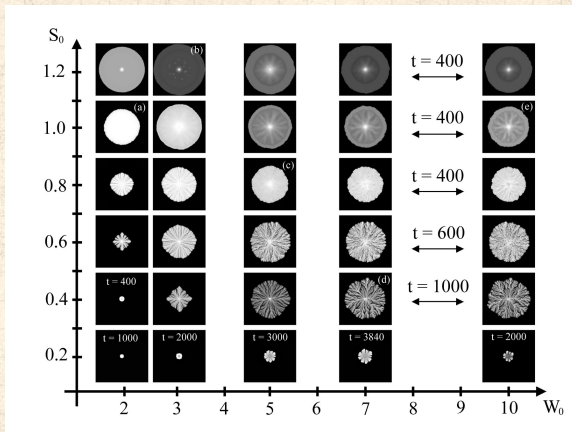
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W_0 = initial wetness, S_0 = initial nutrient supply

<http://math.arizona.edu/~lega/HydroBact.html>



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Analytic issues:



1-d: simple (Ising & Lenz, 1925)



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 4-d and up: simple.



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-  See lower and upper critical dimension  in statistical physics.
-  Also: Curse and Blessing of Dimensionality 




Historical surprise:




Origins of Statistical Mechanics are in the studies of people...
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Analytic approaches have their limits, especially in
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Beyond Statistical Mechanics:

- Analytic approaches have their limits, especially in evolutionary, algorithm-rich systems.
- Algorithmic methods and simulation techniques will continue to rise in importance.



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Complex systems are everywhere.





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The central concepts **Complexity** and **Emergence** are reasonably well defined.



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- We use whatever tools we need.



- Complex systems are everywhere.
- The central concepts **Complexity** and **Emergence** are reasonably well defined.
- There is no single unified **general theory of Complex Systems** but there are big stories that cover giant spaces (fluids, rich-get-richer).
- And the observation of Universality ↗ of dynamical systems, statistical mechanics, and other quantitative areas means not everything is special and different.
- Framing from the Manifesto: Science's focus is moving to Complex Systems **because it finally can.** ^[7]
- We use whatever tools we need.
- Science \simeq Describe + Explain.





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