Fundamentals

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Principles of Complex Systems, Vol. 1 | @pocsvox CSYS/MATH 300, Fall, 2020

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Computational Story Lab | Vermont Complex Systems Center Vermont Advanced Computing Core | University of Vermont























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Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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On Instagram at pratchett the cat

PoCS, Vol. 1 Fundamentals 3 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell References



Outline

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References

PoCS, Vol. 1 Fundamentals 4 of 74

Data

Measurement

Emergence

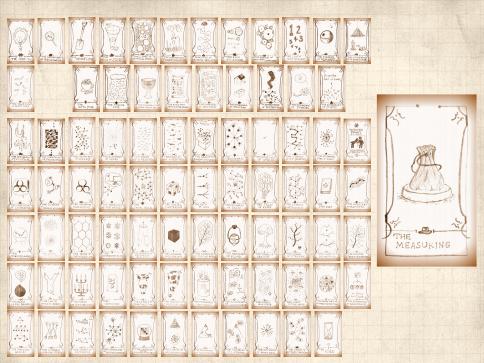
Self-Organization

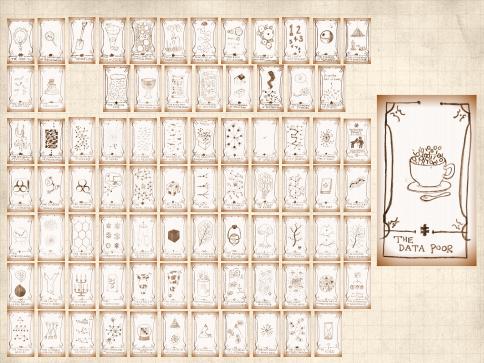
Modeling

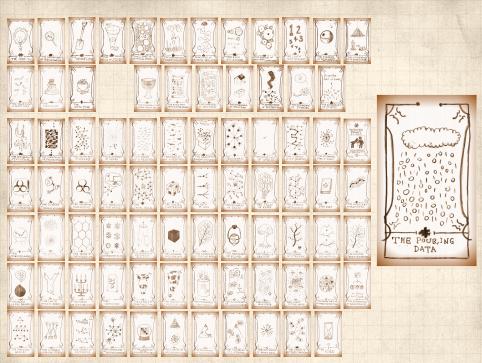
Statistical Mechanics

Nutshell









Data, Data, Everywhere—the Economist, Feb 25, 2010



 \Leftrightarrow Exponential growth: \sim 60% per year.

Big Data Science:

- 2013: year traffic on Internet estimate to reach 2/3 Zettabytes (1ZB = 10³EB = 10⁶PB = 10⁹TB)
- Large Hadron Collider: 40 TB/second.
- 2016—Large Synoptic Survey Telescope: 140 TB every 5 days.
- Arr Facebook: \sim 250 billion photos (mid 2013)
- Twitter: ~ 500 billion tweets (mid 2013)

PoCS, Vol. 1 Fundamentals 8 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



No really, that's a lot of data

Data inflation

2

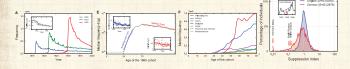
Unit	Size	What it means
Bit (b)	1 or 0	Short for "binary digit", after the binary code (1 or 0) computers use to store and process data
Byte (B)	8 bits	Enough information to create an English letter or number in computer code. It is the basic unit of computing
Kilobyte (KB)	1,000, or 2 ¹⁰ , bytes	From "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 1-2GE
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 The Economist
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

Yotta and Zetta were added in 1991; terms for larger amounts have yet to be established.

Big Data—Culturomics:

"Quantitative analysis of culture using millions of digitized books" by Michel et al., Science, 2011 [9]



http://www.culturomics.org/ and Google Books ngram viewer d

PoCS, Vol. 1 Fundamentals 10 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

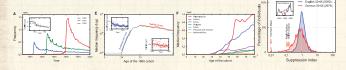
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http://www.culturomics.org/ and Google Books ngram viewer

Barney Rubble:



"Characterizing the Google Books corpus: Strong limits to inferences of socio-cultural and linguistic evolution"

Pechenick, Danforth, and Dodds, PLoS ONE, **10**, e0137041, 2015. [10] PoCS, Vol. 1 Fundamentals 10 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





Lord Kelvin (possibly):

"To measure is to know."

PoCS, Vol. 1 Fundamentals 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





Lord Kelvin (possibly):



"To measure is to know."



"If you cannot measure it, you cannot improve it."

PoCS, Vol. 1 **Fundamentals** 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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Bonus:

PoCS, Vol. 1 **Fundamentals** 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

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PoCS, Vol. 1 **Fundamentals** 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

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"X-rays will prove to be a hoax."

PoCS, Vol. 1 **Fundamentals** 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

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PoCS, Vol. 1 **Fundamentals** 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

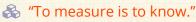
Statistical Mechanics

Nutshell



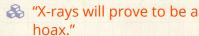


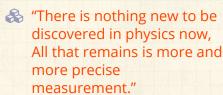
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PoCS, Vol. 1 Fundamentals 11 of 74

Data

Measurement

Emergence

Self-Organization

Modeling Statistical

Mechanics

Nutshell



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" **3** C by Daya Sobel (2007). [16]

Billionths of a second accuracy: Atomic clocks (Lord Kelvin, 1879) PoCS, Vol. 1 Fundamentals 12 of 74

Data

Measurement

Emergence

Self-Organization

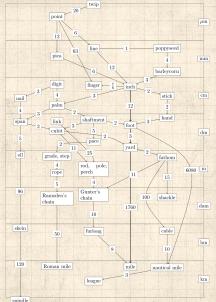
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Statistical Mechanics

Nutshell



Our struggle to sensibly measure anything at all:



By 42CrMo4, Christoph Päper – English length units graph (PNG), CC BY-SA 4.0 https://commons.wikimedia.org/w/index.php?curid=61338012 From https://en.wikipedia.org/wiki/Barleycorn_(unit) 7

PoCS, Vol. 1 Fundamentals 13 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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 leasily 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 [8] per @SilverVVulpes : Also: Inventing Temperature, Hasok Chang, 2004 [3]

PoCS, Vol. 1 Fundamentals 14 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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

From "Bias in Mental Testing", Arthur Jensen, 1980 [8] per @SilverVVulpes : Also: Inventing Temperature, Hasok Chang, 2004 [3]

PoCS, Vol. 1 Fundamentals 15 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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MRLOVENSTEIN.COM

PoCS, Vol. 1 Fundamentals 16 of 74

Data

Measurement

Emergence

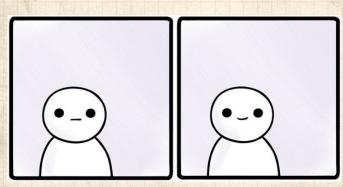
Self-Organization

Modeling

Statistical Mechanics

Nutshell





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Data

Measurement

Emergence

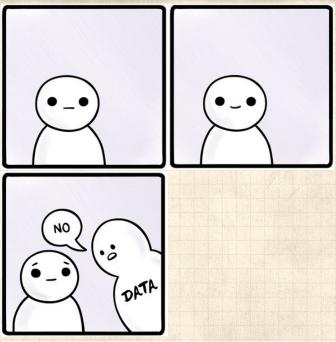
Self-Organization

Modeling

Statistical Mechanics

Nutshell





PoCS, Vol. 1 Fundamentals 16 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





PoCS, Vol. 1 Fundamentals 16 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Data Angry. Data Smash.

PoCS, Vol. 1 Fundamentals 17 of 74

Data

Measurement

Emergence

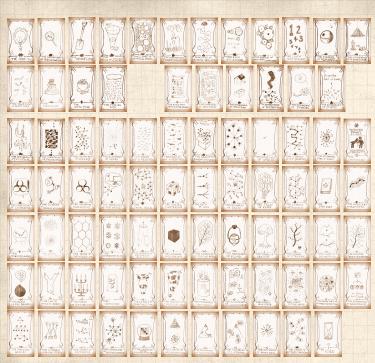
Self-Organization

Modeling

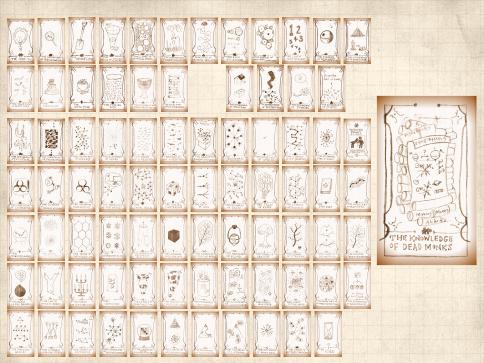
Statistical Mechanics

Nutshell











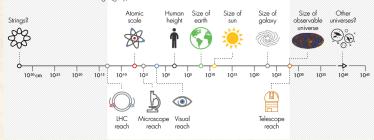


Limits of testability and happiness in Science:

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).



PoCS, Vol. 1 Fundamentals 21 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



The Newness of being a Scientist (1833 on):

Google books Ngram Viewer



PoCS, Vol. 1 Fundamentals 22 of 74

Data

Measurement

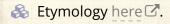
Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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PoCS, Vol. 1 **Fundamentals** 22 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



Etymology here ...

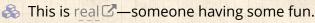


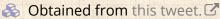
"Scientists are the people who ask a question about a phenomenon and proceed to systematically go about answering the question themselves. They are by nature curious, creative and well organized."

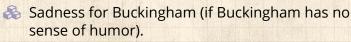


Please do not measure complex systems with one number:









PoCS, Vol. 1 Fundamentals 23 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



The conceptual trapping pit of a single scale:



Lure of simplicity: Comparisons and rankings are easy.

PoCS, Vol. 1 **Fundamentals** 24 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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 ...

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Lure of simplicity: Comparisons and rankings are easy.

A single scale measure is very appealing, very hard to resist and hard to push back against when widely adopted.

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell References



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Examples:

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling Statistical

Mechanics Nutshell



Lure of simplicity: Comparisons and rankings are easy.

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Examples:

Grade point average (GPA)

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



- Lure of simplicity: Comparisons and rankings are easy.
- A single scale measure is very appealing, very hard to resist and hard to push back against when widely adopted.
- Examples:
 - Grade point average (GPA)
 - College rankings, City rankings, Country rankings, Wine scores, Michelin Guide ☑, Yelp scores, Amazon ratings ☑, ...

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

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 - Body Mass Index (BMI)
 - Intelligence Quotient (IQ)¹
 - Effective temperature

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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 - Body Mass Index (BMI)

 - Effective temperature
 - Price for all things: One dimension of belief

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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 - Salary!

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

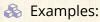
Statistical Mechanics

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Grade point average (GPA)

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PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

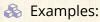
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PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

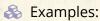
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- Effective temperature
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- Salary!
- stock market valuation for corporations
- Complexity of civilizations [17]
- A 1-d axis for political ideologies (a spatial metaphor trap, thanks France! ☑)

PoCS, Vol. 1 Fundamentals 24 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Personality distributions:



"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. [11]

Five Factor Model (FFM):

Openness [O]

Conscientiousness [C]

Extraversion [E]

Agreeableness [A]

Neuroticism [N]

"...a robust and widely accepted framework for conceptualizing the structure of personality... Although the FFM is not universally accepted in the field..." [11]

PoCS, Vol. 1 Fundamentals 25 of 74 Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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339-369, 2008, [11]



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A concern: self-reported data.

PoCS, Vol. 1 Fundamentals 25 of 74 Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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"...a robust and widely accepted framework for conceptualizing the structure of personality... Although the FFM is not universally accepted in the field..." [11]

A concern: self-reported data. Bigger concern: mass manipulation.

PoCS, Vol. 1 Fundamentals 25 of 74 Data

Measurement

Emergence

Self-Organization

Modeling

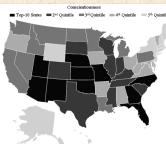
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Nutshell









PoCS, Vol. 1 **Fundamentals** 26 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

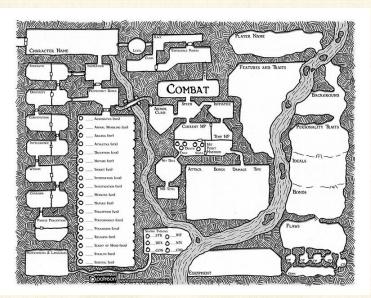








Dungeons & Dragons' full embrace of complexity:



PoCS, Vol. 1 Fundamentals 27 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



From here .

Dungeons & Dragons—Two alignment axes for character:



Law-Chaos (vertical) and Good-Evil (horizontal). PoCS, Vol. 1 Fundamentals 28 of 74

Data

Measurement Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



²From this Reddit thread ☑, where, naturally, the choices are enthusiastically debated.

Please just don't do it:

PoCS, Vol. 1 Fundamentals 29 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Dopeness, etc.:

PoCS, Vol. 1 Fundamentals 30 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

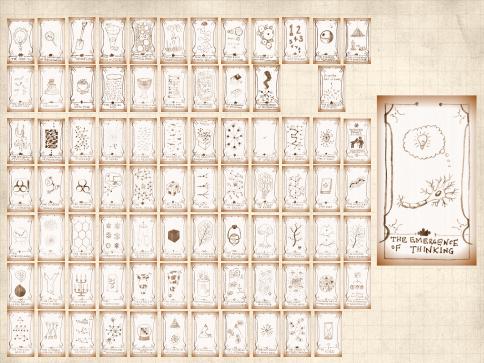


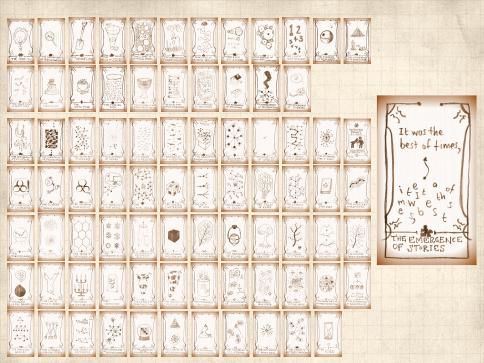




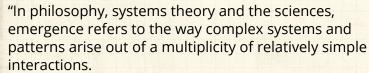








The Wikipedia on Emergence (2006):





Data

Measurement

Emergence

Self-Organization

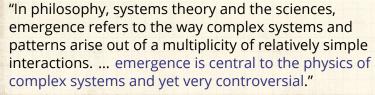
Modeling

Statistical Mechanics

Nutshell



The Wikipedia on Emergence (2006):





Data

Measurement

Emergence

Self-Organization

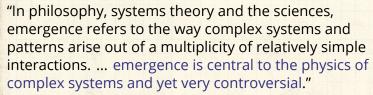
Modeling

Statistical Mechanics

Nutshell



The Wikipedia on Emergence (2006):



Wikipedia, 2016:

In philosophy, systems theory, science, and art, emergence is a process whereby larger entities arise through interactions among smaller or simpler entities such that the larger entities exhibit properties the smaller/simpler entities do not exhibit.

PoCS, Vol. 1 Fundamentals 35 of 74

Data

Measurement

Emergence

Self-Organization

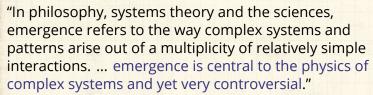
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Nutshell



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The philosopher G. H. Lewes first used the word explicity in 1875.

Fundamentals

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



PoCS, Vol. 1 35 of 74

Fireflies ⇒ Synchronized Flashes:

PoCS, Vol. 1 Fundamentals 36 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References

Film: Sir David Attenborough, BBC.

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



Tornadoes, financial collapses, human emotion aren't found in water molecules, dollar bills, or carbon atoms.

PoCS, Vol. 1 Fundamentals 37 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Tornadoes, financial collapses, human emotion aren't found in water molecules, dollar bills, or carbon atoms.

Examples:

- Fundamental particles ⇒ Life, the Universe, and Everything
- Genes ⇒ Organisms
- Neurons etc. \Rightarrow Brain \Rightarrow Thoughts
- Religion, Collective behaviour
- Reople ⇒ The Web
- Reople ⇒ Language, and rules of language

PoCS, Vol. 1 Fundamentals 37 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Examples:

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- Genes ⇒ Organisms
- Neurons etc. \Rightarrow Brain \Rightarrow Thoughts
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- People \Rightarrow The Web
- Reople ⇒ Language, and rules of language
- $\ref{eq:special}$? \Rightarrow time; ? \Rightarrow gravity; ? \Rightarrow reality.

PoCS, Vol. 1 Fundamentals 37 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



"The whole is more than the sum of its parts" -Aristotle

Friedrich Hayek (Conomist/Philospher/Nobelist):

Markets, legal systems, political systems are emergent and not designed. PoCS, Vol. 1 Fundamentals 38 of 74

Data

Measurement

Emergence

Self-Organization

Modeling Statistical

Mechanics Nutshell



Friedrich Hayek (Conomist/Philospher/Nobelist):

- Markets, legal systems, political systems are emergent and not designed.
- 'Taxis' = made order (by God, Sovereign, Government, ...)

PoCS, Vol. 1 Fundamentals 38 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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- & 'Cosmos' = grown order

PoCS, Vol. 1 Fundamentals 38 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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- Archetypal limits of hierarchical and decentralized structures.

PoCS, Vol. 1 Fundamentals 38 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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- Hierarchies arise once problems are solved. [5]

PoCS, Vol. 1 Fundamentals 38 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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- & Hierarchies arise once problems are solved. [5]
- Decentralized structures help solve problems.

PoCS, Vol. 1 Fundamentals 38 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Friedrich Hayek (Conomist/Philospher/Nobelist):

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- Hierarchies arise once problems are solved. [5]
- Decentralized structures help solve problems.
- Dewey Decimal System versus tagging.

PoCS, Vol. 1 Fundamentals 38 of 74

Data

Measurement Emergence

Self-Organization

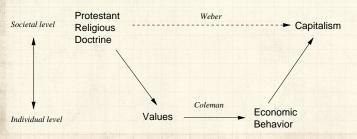
Modeling

Statistical Mechanics

Nutshell



James Coleman I in Foundations of Social Theory:



Understand macrophenomena arises from microbehavior which in turn depends on macrophenomena. [4] PoCS, Vol. 1 Fundamentals 39 of 74

Data

Measurement

Emergence

Self-Organization

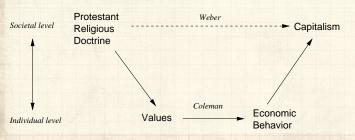
Modeling

Statistical Mechanics

Nutshell



James Coleman in Foundations of Social Theory:



Understand macrophenomena arises from microbehavior which in turn depends on



More on Coleman here .

macrophenomena. [4]

PoCS, Vol. 1 **Fundamentals** 39 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



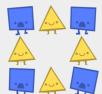
Thomas Schelling ♂ (Economist/Nobelist):



"Micromotives and Macrobehavior" [14]

- Segregation [12, 15]
- Wearing hockey helmets [13]
- Seating choices

Vi Hart and Nicky Case's Polygonthemed visualization ☑:



PoCS, Vol. 1 Fundamentals 40 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



The emergence of taste:



Molecules ⇒ Ingredients ⇒ Taste

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



nvtimes.com 2

PoCS, Vol. 1 **Fundamentals** 41 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Reductionism and food:

PoCS, Vol. 1 Fundamentals 42 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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..."

PoCS, Vol. 1 **Fundamentals** 42 of 74

Measurement

Emergence

Self-Organization

Modeling Statistical

Mechanics Nutshell



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."

PoCS, Vol. 1 Fundamentals 42 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





🚓 "people don't eat nutrients, they eat foods, and foods can behave very differently than the nutrients they contain."

PoCS, Vol. 1 **Fundamentals** 43 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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. PoCS, Vol. 1 Fundamentals 43 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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 PoCS, Vol. 1 Fundamentals 43 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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." PoCS, Vol. 1 Fundamentals 43 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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-coumoric acid, p-hydroxy-benzoic acid, palmitic acid, rosmarinic acid, selenium, tannin, thymol, tryptophan, ursolic acid, vanillic acid.

PoCS, Vol. 1 Fundamentals 44 of 74 Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





"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."

PoCS, Vol. 1 Fundamentals 45 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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."

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

PoCS, Vol. 1 Fundamentals 45 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



This is a Collateralized Debt Obligation:



PoCS, Vol. 1 Fundamentals 46 of 74

Dat

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"The Universe is made of stories, not of atoms."



From "The Speed of Darkness" (1968) by Muriel Rukeyser 2

PoCS, Vol. 1 **Fundamentals** 47 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"The Universe is made of stories, not of atoms."



PoCS, Vol. 1 **Fundamentals** 47 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



From "The Speed of Darkness" (1968) by Muriel Rukeyser 2



Quoted by Metatron in Supernatural, Meta Fiction, S9E18.

(Sir Terry) Pratchett's ☑ Narrativium ☑:



PoCS, Vol. 1 Fundamentals 48 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



(Sir Terry) Pratchett's ☑ Narrativium ☑:



"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." PoCS, Vol. 1 Fundamentals 48 of 74

Data

Measurement

Emergence

Self-Organization

Modeling Statistical

Mechanics Nutshell



(Sir Terry) Pratchett's ☑ Narrativium ☑:



"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." PoCS, Vol. 1 Fundamentals 48 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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

PoCS, Vol. 1 **Fundamentals** 49 of 74

Measurement

Emergence

Self-Organization

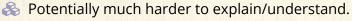
Modeling

Statistical Mechanics

Nutshell



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



PoCS, Vol. 1 Fundamentals 49 of 74

Data

Measurement

Emergence

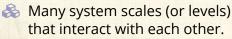
Self-Organization

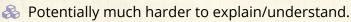
Modeling

Statistical Mechanics

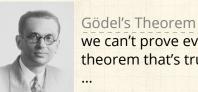
Nutshell







Even mathematics: [6]



Gödel's Theorem 2: we can't prove every theorem that's true



"Gödel, Escher, Bach" [7]

PoCS, Vol. 1 **Fundamentals** 49 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

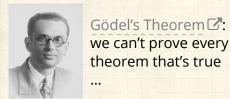
Nutshell



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

Potentially much harder to explain/understand.

Even mathematics: [6]





"Gödel, Escher, Bach" [7]

PoCS, Vol. 1 Fundamentals 49 of 74

Data

Measurement

Emergence

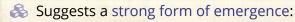
Self-Organization

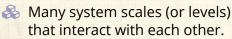
Modeling

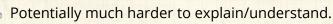
Statistical Mechanics

Nutshell

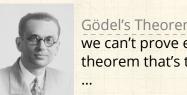








Even mathematics: [6]



Gödel's Theorem 2: we can't prove every theorem that's true



"Gödel, Escher, Bach" [7]

PoCS, Vol. 1 **Fundamentals** 49 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References



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

Roughly speaking, there are two types of emergence:

PoCS, Vol. 1 Fundamentals 50 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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.

PoCS, Vol. 1 Fundamentals 50 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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.

PoCS, Vol. 1 Fundamentals 50 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Reductionist techniques can explain weak emergence.

PoCS, Vol. 1 **Fundamentals** 51 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Reductionist techniques can explain weak emergence.

Magic explains strong emergence. [2]

PoCS, Vol. 1 Fundamentals 51 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Fundamentals 51 of 74

PoCS, Vol. 1

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

References

Reductionist techniques can explain weak emergence.

Magic explains strong emergence. [2]

But: maybe magic should be interpreted as an inscrutable yet real mechanism that cannot ever be simply described.

PoCS, Vol. 1 **Fundamentals** 51 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell References

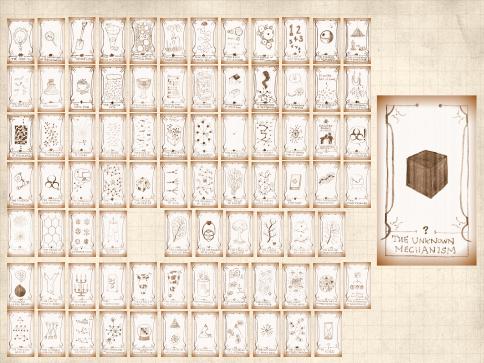
Reductionist techniques can explain weak emergence.

Magic explains strong emergence. [2]

But: maybe magic should be interpreted as an inscrutable yet real mechanism that cannot ever be simply described.

Gulp.





Limits of Science | Radiolab



Listen to Steve Strogatz, Hod Lipson, and Michael Schmidt (Cornell) in the last piece (11:16) on Radiolab's show 'Limits' (April 5, 2010).



(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 Eureqa at UVM's 2011 TEDx event "Big Data, Big Stories."

PoCS, Vol. 1 Fundamentals 53 of 74

Dat

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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."

PoCS, Vol. 1 Fundamentals 54 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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." (also: Self-assembly)

PoCS, Vol. 1 Fundamentals 54 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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." (also: Self-assembly)

Examples:

 $\ \, \& \,$ Molecules/Atoms liking each other \rightarrow $\,$ Gases, liquids, and solids.

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Protein folding.

& Imitation \rightarrow Herding, flocking, mobs, ...

PoCS, Vol. 1 Fundamentals 54 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



"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." (also: Self-assembly)

Examples:

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 \clubsuit Spin alignment \to Magnetization.

Protein folding.

 \mathbb{A} Imitation \rightarrow Herding, flocking, mobs, ...

Fundamental question: how likely is 'complexification'?

PoCS, Vol. 1 Fundamentals 54 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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

PoCS, Vol. 1 **Fundamentals** 55 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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

Statistical techniques for comparisons and descriptions.

PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling Statistical

Mechanics Nutshell

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

Statistical techniques for comparisons and descriptions.

Methods from statistical mechanics and computer science. PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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

Statistical techniques for comparisons and descriptions.

Methods from statistical mechanics and computer science.

Machine learning (but beware the black box).

PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



- Differential equations, difference equations, linear algebra, stochastic models.
- Statistical techniques for comparisons and descriptions.
- Methods from statistical mechanics and computer science.
- Machine learning (but beware the black box).
- Computer modeling, everything from
 - Artisanal toy models
 - to kitchen sink models.

PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Key advance (more soon):

Representation of complex interaction patterns as complex networks.

PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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

Statistical techniques for comparisons and descriptions.

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Computer modeling, everything from

Artisanal toy models

to kitchen sink models.

Key advance (more soon):

Representation of complex interaction patterns as complex networks.

The driver: Massive amounts of Data

PoCS, Vol. 1 Fundamentals 55 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Rather silly but great 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

Amusing interview here

PoCS, Vol. 1 **Fundamentals** 56 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

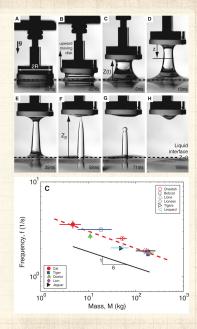




Another great, great moment in scaling:

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

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 allomety 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.



PoCS, Vol. 1 **Fundamentals** 57 of 74

Measurement

Emergence

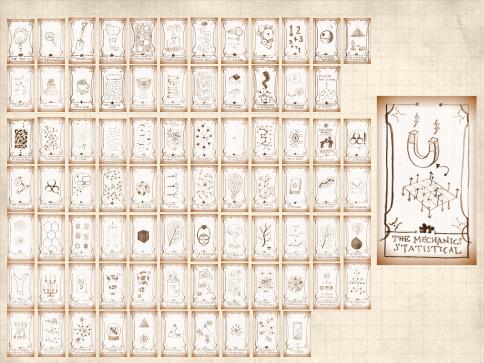
Self-Organization

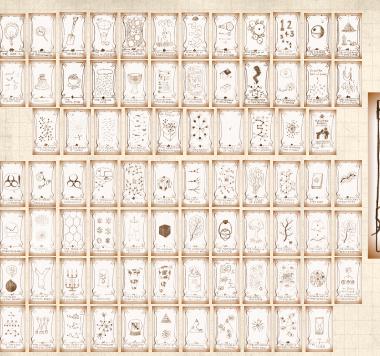
Modeling

Statistical Mechanics

Nutshell









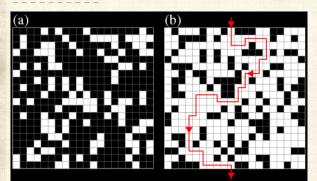


Statistical Mechanics is "a science of collective behavior."



Simple rules give rise to collective phenomena.

Percolation:



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

PoCS, Vol. 1 **Fundamentals** 60 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell







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

PoCS, Vol. 1 **Fundamentals** 61 of 74

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Emergence

Self-Organization

Modeling

Statistical Mechanics

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PoCS, Vol. 1 Fundamentals 61 of 74

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Measurement

Emergence

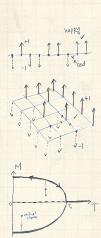
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PoCS, Vol. 1 Fundamentals 61 of 74

Data

Measurement

Emergence

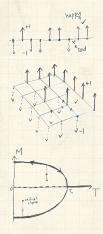
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PoCS, Vol. 1 Fundamentals 61 of 74

Data

Measurement

Emergence

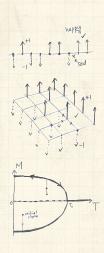
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PoCS, Vol. 1 Fundamentals 61 of 74

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Measurement

Emergence

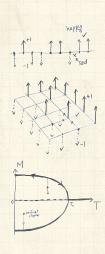
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PoCS, Vol. 1 Fundamentals 61 of 74

Data

Measurement

Emergence

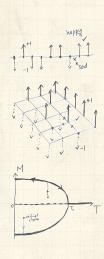
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Modeling

Statistical Mechanics

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PoCS, Vol. 1 Fundamentals 61 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

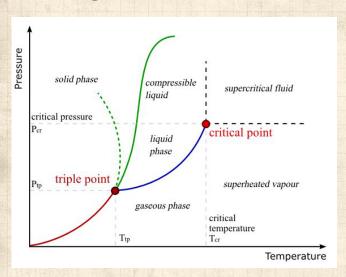
Nutshell

References



Example 2-d Ising model simulation:

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



Qualitatively distinct macro states.

PoCS, Vol. 1 Fundamentals 62 of 74

Data

Measurement

Emergence

Self-Organization

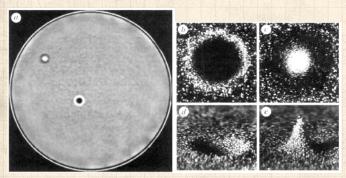
Modeling

Statistical Mechanics

Nutshell



Oscillons, bacteria, traffic, snowflakes, ...



Umbanhowar et al., *Nature*, 1996 [18]

PoCS, Vol. 1 Fundamentals 63 of 74

Data

Measurement

Emergence

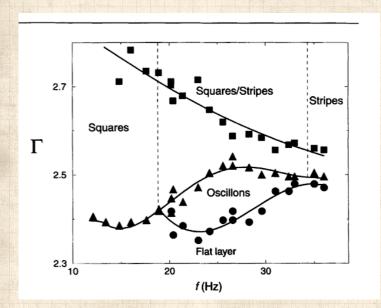
Self-Organization

Modeling

Statistical Mechanics

Nutshell





PoCS, Vol. 1 Fundamentals 64 of 74

Data

Measurement

Emergence

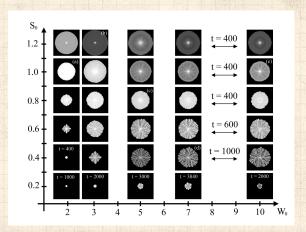
Self-Organization

Modeling

Statistical Mechanics

Nutshell





 W_0 = initial wetness, S_0 = initial nutrient supply http://math.arizona.edu/~lega/HydroBact.html

PoCS, Vol. 1 Fundamentals 65 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



Analytic issues:



4 1-d: simple (Ising & Lenz, 1925)



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Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 66 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 66 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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<page-header> 3-d: extremely hard...

4-d and up: simple.

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8

PoCS, Vol. 1 Fundamentals 66 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Historical surprise:



Origins of Statistical Mechanics are in the studies of people... (Maxwell and co.)

PoCS, Vol. 1 **Fundamentals** 67 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Now physicists are using their techniques to study everything else including people... PoCS, Vol. 1 Fundamentals 67 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 67 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Beyond Statistical Mechanics:

PoCS, Vol. 1 Fundamentals 67 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 67 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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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.

PoCS, Vol. 1 Fundamentals 67 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





The central concepts Complexity and Emergence are reasonably well defined.

PoCS, Vol. 1 **Fundamentals** 68 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





The central concepts Complexity and Emergence are reasonably well defined.

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PoCS, Vol. 1 **Fundamentals** 68 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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There is no general theory of Complex Systems.



But the problems exist... Complex (Adaptive) Systems abound... PoCS, Vol. 1 **Fundamentals** 68 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





There is no general theory of Complex Systems.

But the problems exist... Complex (Adaptive) Systems abound...

And the observation of Universality of dynamical systems, statistical mechanics, and other quantitative areas means not everything is special and different.

PoCS, Vol. 1 Fundamentals 68 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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We use whatever tools we need.

PoCS, Vol. 1 Fundamentals 68 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell





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Science ≃ Describe + Explain.

PoCS, Vol. 1 Fundamentals 68 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 72 of 74

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell



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PoCS, Vol. 1 Fundamentals 74 of 74

Data

Measurement

Emergence

Self-Organization

Modeling

Statistical Mechanics

Nutshell

