

too much information !



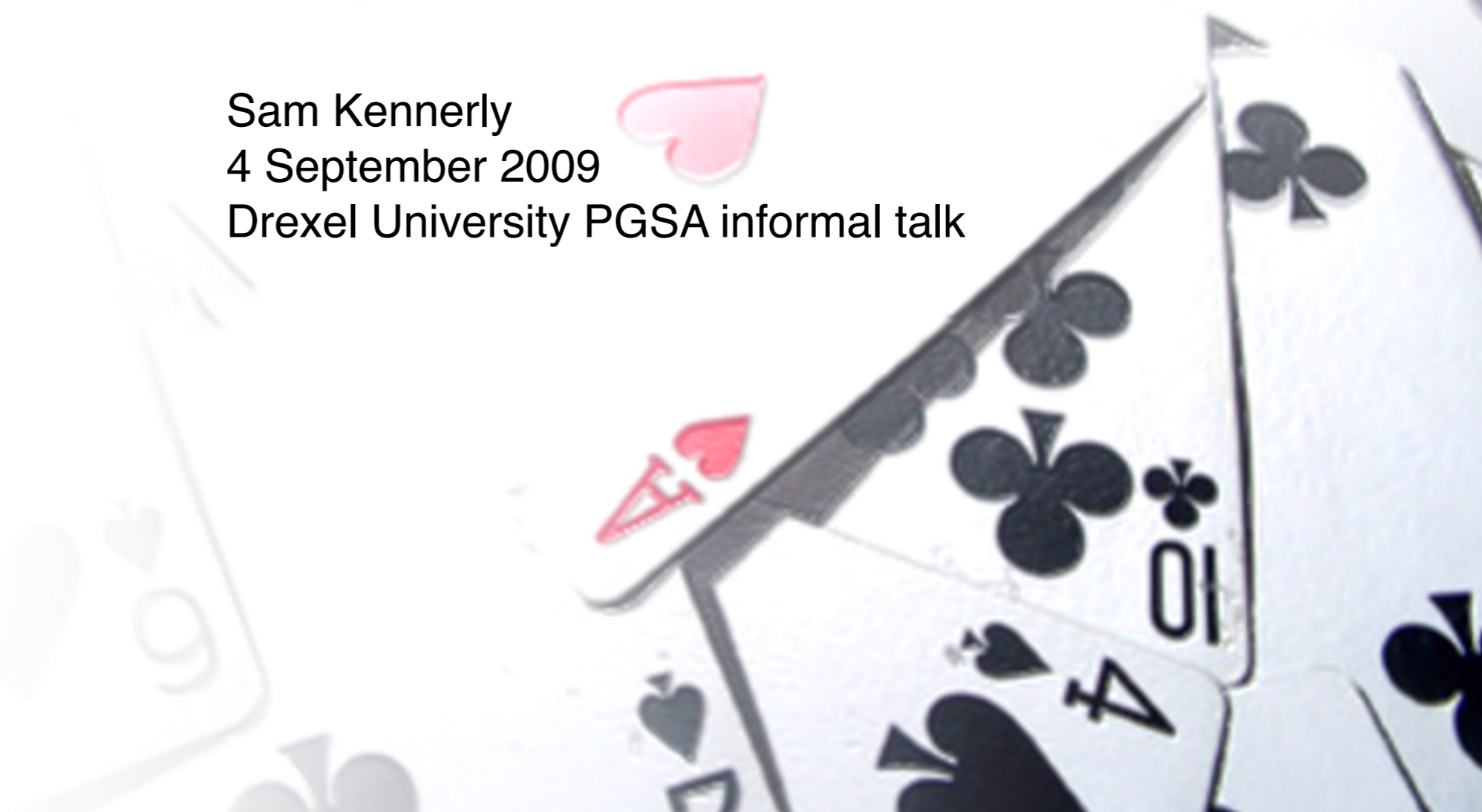
Introduction to Classical and Quantum Information Theory

and other random topics from probability and statistics

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Drexel University PGSA informal talk





0.0 DNA and Beethoven's 9th Symphony

- ◇ In my last presentation, I said the information content of the human genome is about equal to a recording of Beethoven's 9th.
- ◇ **3 billion base pairs** in human DNA, each occupied by **1 of 4 bases**.
Representing each base by two binary digits, we need $(2 \text{ bits}) \times (3 \text{ billion})$
= **6 gigabits** = 750 MB of disk space to sequence a genome.
- ◇ An audio CD records two 16-bit samples every 44,100th of a second. The 9th is about 72 minutes long, so it needs $(2)(16)(44,100)(72)(60)$ bits = **6 Gb**.
- ◇ Question: Do we really need **all** those bits? Can't we .zip them or something?





0.1 DNA and Beethoven's 9th Symphony

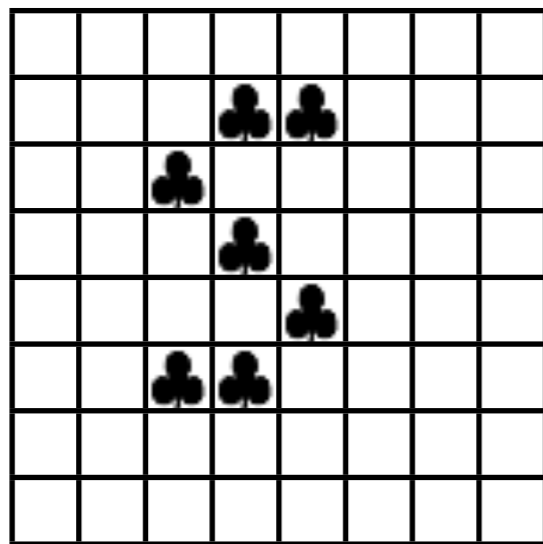
- ◇ DNA answer: The *entropy rate* of DNA is about 1.7 bits per base, about 85% of the maximum 2 bits/base. Shannon's **source coding theorem** says that no algorithm can compress the genome to less than $(0.85)(750\text{MB}) = 637.5 \text{ MB}$.
- ◇ Real-life compression is imperfect; source-coding theorem gives a **lower bound** on file size. Compression schemes designed for one type of data may work poorly for others. (ZIP is notoriously bad for audio encoding.)
- ◇ Beethoven answer: The entropy rate depends on the recording, but existing *Golomb-Rice* encoders compress to about 50-60% original size.
- ◇ **Lossy compression** can make files smaller, but **information is destroyed!** Examples: mp3/aac/ogg (audio), jpg/gif (graphics), DivX/qt/wmv (video)
- ◇ Experiments suggest VBR-mp3 at 18% is good enough to trick listeners.
- ◇ How much of DNA info is "junk" is debated; 95% is a popular estimate.



1.0 What is entropy?

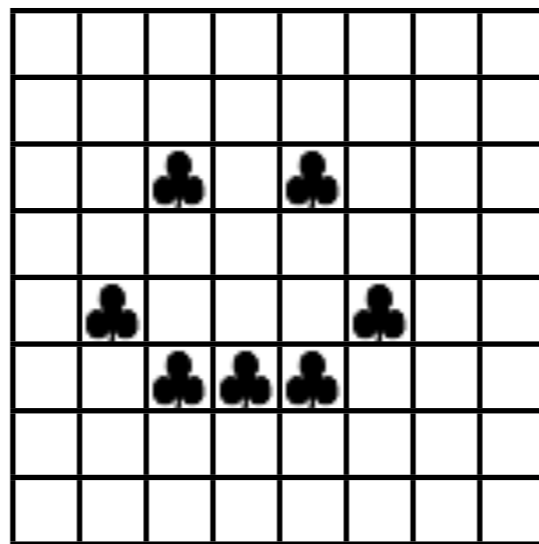
- ◇ Old-fashioned answer: Entropy is a measure of how disordered a system is.
- ◇ Dilemma: How do we define disorder? A broken egg is more disordered than a not-broken egg... but which of the following pictures is least disordered?

system 1



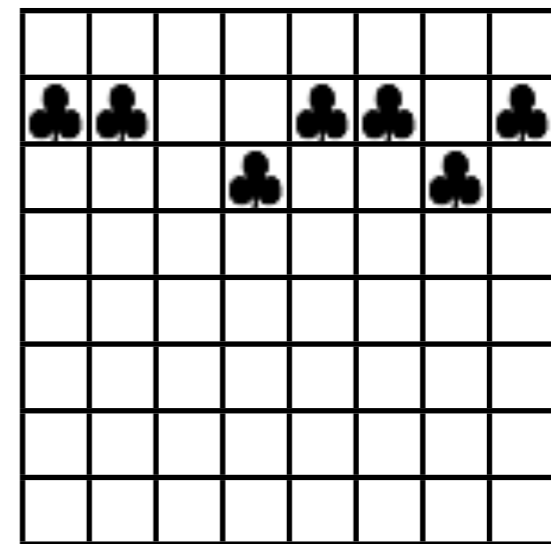
Letter "S"

system 2



Smiley Face

system 3



Sicilian Dragon

- ◇ Moral of story: **Disorder is in the eye of the beholder.**

2.0 Boltzmann's entropy

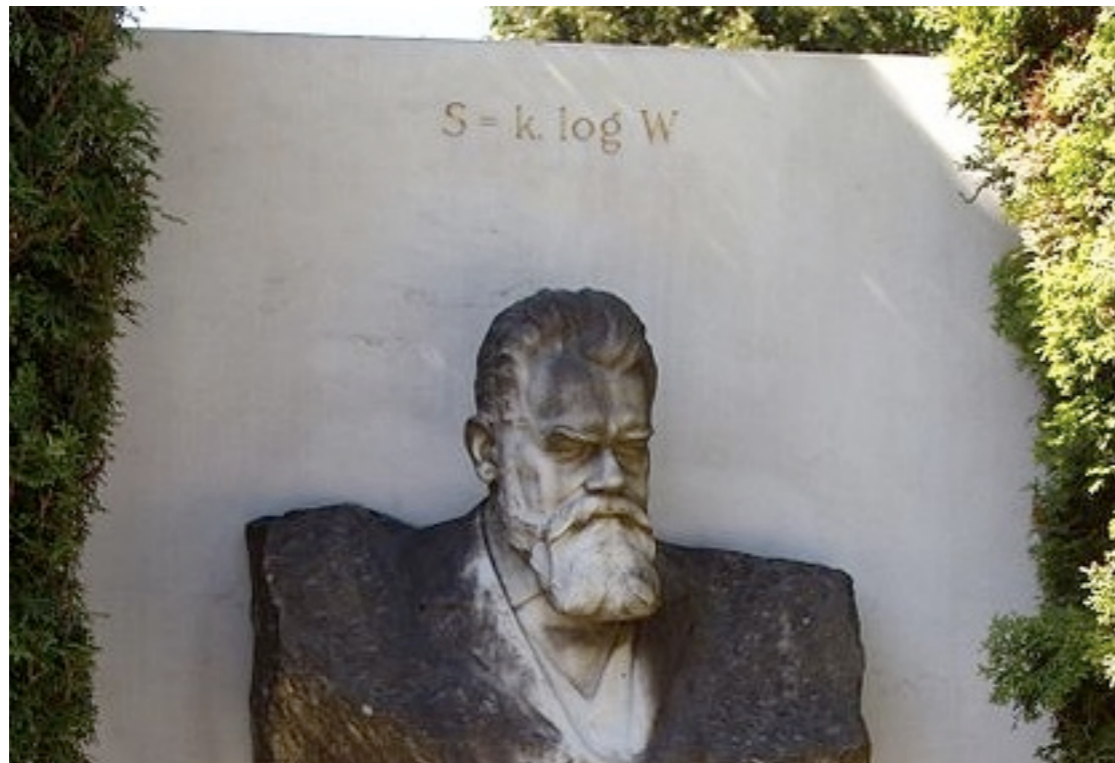
- ◇ Question: How do we model the behavior of gases in a steam engine?
- ◇ 1 L of ideal gas at STP has $2.7 \cdot 10^{22}$ molecules. If each has 3 position and 3 momentum coordinates, differential eqn. of motion has $\sim 10^{23}$ variables. (Actual gases are much more complicated, of course.)
- ◇ Solving this equation is an impractical way to build locomotives.
- ◇ Answer: Call each configuration the system a **microstate**. If two different microstates have the same Energy, Volume, and Number of particles, call them equivalent. A **macrostate** a set of microstates with the same (E,V,N) values.
- ◇ **Multiplicity** $\Omega(E,V,N)$ is the number of microstates for a given macrostate.
- ◇ Ω is measure of how much information we are ignoring in our model of the system. For this reason, I like to call it the **ignorance** of a macrostate.

2.1 Boltzmann's entropy

- ◇ This method of counting microstates per macrostate is called **microcanonical ensemble theory**. Boltzmann defined the entropy of a macrostate like so:

$$S(E, V, N) = k \ln(\Omega)$$

- ◇ This entropy is the **logarithm of ignorance** times a constant $k \approx 1.38 \cdot 10^{23} \text{ J/K}$.
- ◇ To help us remember this formula, Boltzmann had it carved into his tombstone.



- ♣ This is Ludwig Boltzmann's tomb in Vienna.
(Apparently he was one of those people who prefer "log" to "ln." Also he used W for multiplicity, but you get the idea.)
- ♣ Boltzmann's kinetic theory of gases caused some controversy because it apparently requires systems to be inherently discrete.
- ♣ Quantum-mechanical systems with discrete energy levels fit nicely into this theory!



3.0 Shannon's entropy

- ◇ In 1937, Claude Shannon wrote a famous Master's thesis about using Boolean algebra to write computer programs. During WWII he worked with Alan Turing on cryptography and electronic control theory for Bell Labs.
- ◇ Shannon later published his **source-coding** and **noisy-channel** theorems. These placed limits on file compression and the data capacity of a medium subject to noise and errors. Both theorems use this definition of entropy:

$$S[p_n] = - \sum_n p_n \log(p_n)$$

for discrete probability distributions

$$S[p(x)] = - \int p \log(p) dx$$

for continuous probability distributions

- ◇ **Gibbs' entropy** from thermodynamics is Shannon's entropy times k ,* though Shannon's entropy is defined for probability distributions, not physical states. S is a measure of **how much information is revealed by a random event**.

* Prof. Goldberg and I opine that temperatures should be written in Joules, in which case $k = 1$.

3.1 Shannon's entropy

- ◇ For a random variable X , a continuous probability distribution $p(x)$ is defined:

$$P[a \leq X \leq b] = \int_a^b p(x) dx$$

- ◇ A probability distribution $p(x)$ is also called a **probability density function** or **PDF**. (Technically $p(x)$ doesn't have to be a function as long as it can be integrated. For example, Dirac's $\delta(x)$ is a valid PDF but not a function.)
- ◇ From the definition it follows that $p(x) \geq 0$ and $\int_{-\infty}^{+\infty} p(x) dx = 1$.
- ◇ Example: Cryptographers perform **frequency analysis** on ciphertexts by writing a discrete PDF for how often each letter appears. For a plaintext, this PDF has non-maximal entropy; the letter "E" is more probable than "Q."
- ◇ Example: Password entropy is maximized by using uniformly-chosen random letters instead of English words. Including numbers and symbols increases S .

3.2 Shannon's entropy

- ◇ To better understand Shannon's entropy, first define a **surprisal** $I_n = \log(p_n^{-1})$ for each possible random outcome p_n .
- ◇ Example: Alice rolls two dice at the same time. Bob bets her \$1 that she will not roll "boxcars" (two 6's). If Alice wins, Bob's surprisal will be $\log(36)$.
- ◇ Example: The table below shows how surprised we should be when dealt certain types of Texas Hold 'Em hands preflop.

hand	AA	AA/KK	99 or better	any pair	any suited	the hammer
surprisal	$\log(221)$	$\log(111)$	$\log(37)$	$\log(17)$	$\log(4.25)$	$\log(111)$

- ◇ Shannon's entropy for a PDF is **the expectation value of surprisal**.

$$\left\langle \log \left(\frac{1}{p_n} \right) \right\rangle = - \left\langle \log(p_n) \right\rangle = - \sum_n p_n \log(p_n)$$

IMPORTANT TECHNICALITY: $0 \log(0) = 0$. Use l'Hôpital's rule and $\lim_{x \rightarrow 0} [x \log(x)] = \lim_{y \rightarrow \infty} [\log(y)/y]$.



3.3 Shannon's entropy

- ◇ Question: What base to use for log?
- ◇ Answer: Any number! Information entropy comes in dimensionless units.

base	2	e	10
unit name	bit	nat	hartley (or ban)

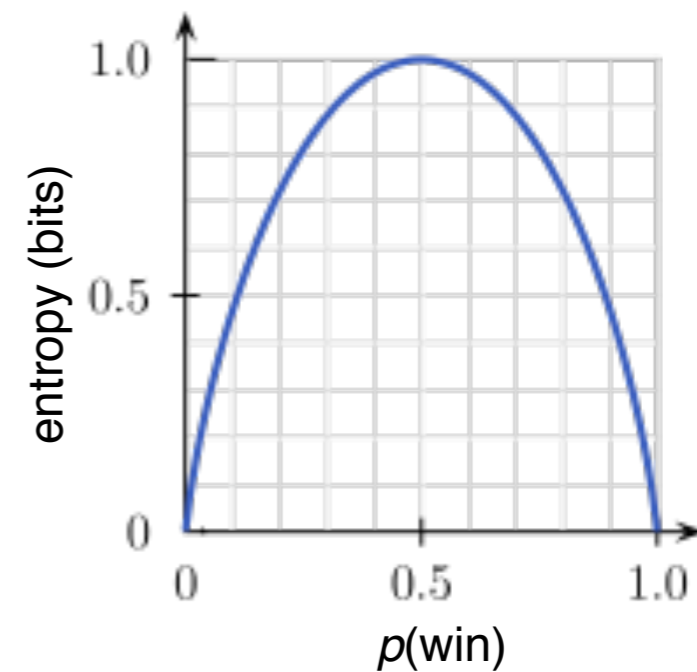
- ♣ Shannon is credited with inventing the term “bit” for the entropy of a single fair coin toss.
- ♣ Ralph Hartley was a Bell Labs information-theorist working with Turing and Shannon.

- ◇ Question: Why use a logarithm in the definition of entropy?
- ◇ Answer: Observing N outcomes of a random process should give us N times as much information as one outcome. Information is an **extensive quantity**.
- ◇ Example: Rolling a die once has 6 possible outcomes and rolling it twice has 6^2 outcomes. The entropy of two die rolls is $\log(6) + \log(6) = \log(6^2) \approx 5.17$.

3.4 Shannon's entropy

- ◇ The entropy of a fair coin toss is $(.5)(\log 2) + (.5)(\log 2)$. In base 2, that's **1 bit**.
- ◇ The entropy of an *unfair* coin toss is given by the **binary entropy function**.
- ◇ 2-player Hold 'Em preflop all-in hands are examples of unfair coin tosses:

hand	p(win)	surprisal	entropy
AA vs AKs	87%	2.9	0.557 bit
AKo vs 89s	59%	1.3	0.976 bit
89s vs 44	52%	1.1	0.999 bit
44 vs AKo	54%	1.1	0.996 bit
KK vs 88	80%	2.3	0.722 bit



Here best **hand** is written first, **p(win)** is prob. best hand wins, and **surprisal** is $\log_2 ([1-p(\text{win})]^{-1})$.

3.5 Shannon's entropy

- ◇ For an N -sided fair die, each outcome has surprisal N . The entropy is

$$-\sum_1^N \frac{1}{N} \log\left(\frac{1}{N}\right) = \sum_1^N \frac{1}{N} \log(N) = \log(N)$$

so Boltzmann's entropy is just Shannon's entropy for a uniform discrete PDF.

- ◇ If $p(x)$ is zero outside a certain range, S is maximal for a **uniform distribution**. (Of course! A fair die (or coin) is inherently less predictable than an unfair one.)
- ◇ For a given standard deviation σ , S is maximal if $p(x)$ is a **normal distribution**. In this sense, bell curves are "maximally random" - but be *very careful* interpreting this claim! Some PDFs (e.g. Lorentzians) have no well-defined σ .
- ◇ For multivariate PDFs, *Bayes' theorem* is used to define **conditional entropy**:

$$p(x|y) = \frac{p(x)}{p(y)} p(y|x) \quad \Rightarrow \quad S[X|Y] = -\sum_{x,y} p(x,y) \log\left(\frac{p(x,y)}{p(y)}\right)$$

4.1 Thermodynamics

- ◇ Recall how temperature is defined in thermodynamics: $\frac{1}{T} \equiv \left(\frac{\partial S}{\partial U} \right)_{N,V}$
- ◇ Define **coldness*** $\beta = 1 / T$. Given a system with fixed particle number and volume, find the probability of each state as a function of internal energy U .
- ◇ Find Shannon's entropy for each PDF, then find $\beta = (\partial S / \partial U)$. The result is an information-theoretical definition of temperature in Joules per nat!
- ◇ In other words, coldness is a measure of **how much entropy a system gains when its energy is increased**. Equivalently, T is a measure of how much energy is needed to increase the entropy of a system.
- ◇ It is energetically "cheap" to increase the entropy of a cold system. If a hot system gives energy to a cold one, the total entropy of both systems increases. The observation that heat flows from hot things to cold leads to the 2nd Law...

* Coldness is more intuitive when dealing with negative temperatures, which are *hotter* than ∞ Kelvins!



4.2 Thermodynamics

- ◇ There have been many attempts to clearly state the 2nd Law of Thermo:
- ◇ Statistical: The entropy of a closed* system at thermal equilibrium is more likely to increase than decrease as time passes.
- ◇ Clausius: “Heat generally cannot flow spontaneously from a material at lower temperature to a material at higher temperature.”
- ◇ Kelvin: “It is impossible to convert heat completely into work in a cyclic process.”
- ◇ Murphy: “If there’s more than one way to do a job, and one of those ways will result in disaster, then somebody will do it that way.”
- ◇ My attempt: “Any system tends to acquire information from its environment.”

* **Loschmidt’s paradox** points out that if a system is truly “closed,” i.e. it does not interact with its environment in any way, then the statistical version of the 2nd Law violates time-reversal symmetry!



5.0 Von Neumann's entropy

- ◇ Despite his knowledge of probability, Von Neumann was reportedly a terrible poker player, so he invented **game theory**.
- ◇ Imagine playing 10,000 games of rock-paper-scissors for \$1 per game. **Pure strategies** can be exploited: if your opponent throws only scissors, you should throw only rocks, etc. The best option is a **mixed strategy** in which you randomly choose rock, paper, or scissors with equal probability.
- ◇ Assume your opponent knows the probability of each of your actions. The entropy of a pure strategy is 0. The entropy of $1/3$ rock + $1/3$ paper + $1/3$ scissors is $\log(3) \approx 1.58$ bits, which is the maximum possible for this game.
- ◇ Von Neumann's poker models (and all modern ones) favor mixed strategies. But unlike rock-paper-scissors, the best strategy is *not* the one that maximizes entropy. The best poker players **balance** their strategies by mixing profitable plays with occasional entropy-increasing bluffs and slowplays.

5.1 Von Neumann's entropy

- ◇ Von Neumann (and possibly also Felix Bloch and Lev Landau) developed an alternate way to write quantum mechanics in terms of **density operators**.
- ◇ Density operators are useful for describing *mixed states* and systems in thermal equilibrium. The related *von Neumann entropy* is also used to describe entanglement in quantum computing research.
- ◇ Density operators are defined as real combinations of **projection operators**. A projection P is a linear operator such that $P = P^2$ ($= P^3 = P^4 = \dots$)
- ◇ For any vector Ψ , there is a projection P_Ψ . In Dirac notation, $\hat{P}_\Psi = |\Psi\rangle\langle\Psi|$. This notation says, "Give P_Ψ a vector. It will take the inner product of that vector with Ψ to produce a number, and it will output Ψ times that number."

$$|a\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \hat{P}_a = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad |b\rangle = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \Rightarrow \hat{P}_b = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

5.2 Von Neumann's entropy

- ◇ A **pure** quantum state can be represented by some state vector ψ in some complex vector space. Its density operator is defined $\rho = P_\psi$.
- ◇ **Mixed** quantum states represent uncertain preparation procedures. For example, Alice prepares a spin-1/2 particle in the S_z eigenstate $|\uparrow\rangle$. Chuck then performs an S_x measurement but *doesn't* tell Bob the result. Bob knows the state is now either $\frac{1}{\sqrt{2}}(|\uparrow\rangle + |\downarrow\rangle)$ or $\frac{1}{\sqrt{2}}(|\uparrow\rangle - |\downarrow\rangle)$, but he doesn't know which!
- ◇ Bob can still write a density operator for this mixture of states. He constructs a projection operator for each possible state, then multiplies each operator by 50% and adds the two operators together:

$$\hat{P}_1 = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \quad \hat{P}_2 = \frac{1}{2} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \quad \Rightarrow \quad \hat{\rho} = \frac{1}{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

- ◇ In general, a density operator is defined $(p_1)\hat{P}_1 + (p_2)\hat{P}_2 + (p_3)\hat{P}_3 \cdots$ where each P is the projection of a state and each p the probability the system is in that state.

5.3 Von Neumann's entropy

- ◇ If Bob measures the z-spin of his mixed-state particle, the expectation value of his measurement is the *trace* of the operator $[S_z][\rho]$. (For a matrix, trace is the sum of diagonal elements. In this case, that would be 0.)
- ◇ The diagonal elements of ρ are the probability of Bob finding S_z to be $+1/2$ or $-1/2$. If Bob wants to know the probability of finding the result of some other measurement, he rewrites ρ using the eigenstates of that operator as his basis.
- ◇ The time-evolution of ρ follows the **Von Neumann equation**, the density-operator version of the Schrödinger equation: $i\hbar \partial_t \hat{\rho} = [\hat{H}, \hat{\rho}]$
- ◇ **Von Neumann's entropy** is defined by putting ρ into Shannon's entropy:

$$S = - \sum_n p_n \log(p_n) = -Tr[\hat{\rho} \log(\hat{\rho})]$$

- ◇ Performing an observation changes ρ in such a way that S *always* increases!

5.4 Von Neumann's entropy

- ◇ Question: How do you find the log of an *operator* ?
- ◇ Answer: If the operator is Hermitian, it can be diagonalized by a unitary transformation $H = U^{-1}DU$. Since $\text{Exp}[U^{-1}DU] = U^{-1} \text{Exp}[D] U$, we can “log” an operator by finding the log of its eigenvalues and then similarity transforming.
- ◇ A projection P_ψ made from a vector ψ is always Hermitian. A real combination of Hermitian operators is also Hermitian, so ρ is **Hermitian**. In fact, all its eigenvalues are in the interval $[0, 1]$ (Remember, zero eigenvalues can be ignored in the entropy formula because $0 \log(0) = 0$.)
- ◇ The definition of ρ can be used to prove that its trace $\text{Tr}[\rho] = 1$ always.
- ◇ The quantum version of **canonical ensemble** thermodynamics uses density operators. The **partition function** Z and density operator ρ are given by:

$$Z = \text{Tr}[\text{EXP}(-\beta\hat{H})] \quad \hat{\rho} = \frac{1}{Z} \text{EXP}(-\beta\hat{H})$$

6.0 Quantum information paradoxes

- ◇ According to the Schrödinger, Heisenberg, and Von Neumann equations, quantum time evolution is unitary. Unitary transformations are *always* invertible, which means they can *never* destroy information about a state.
- ◇ The Copenhagen interpretation, however, says that measuring a system “collapses” it into an eigenstate. This time evolution is a projection onto a vector, so it is singular. Singular transformations *always* destroy information. Schrödinger thought this “damned quantum jumping” was absurd.
- ◇ Von Neumann’s entropy is *increased* by projective measurements. Does this help solve Schrödinger’s objection? If entropy is the amount of random information in a system, perhaps measurements only scramble information.
- ◇ Hawking, ‘t Hooft, Susskind, and Bekenstein claim that black holes maximize entropy for a given surface area, and if one of two entangled particles is sucked into the horizon, Hawking radiation is emitted as a *mixed* state. This is not unitary time-evolution either! Do black holes count as observers?

Something Completely Different

- ◇ Humans seem to be naturally inept at understanding certain concepts from probability and statistics. Some notorious examples are below:
- ◇ 1. Fighter pilots at a particular airbase are each shot down with probability 1% on each mission. What are the odds that a pilot completes 200 missions?
- ◇ 2. Betting on a number in roulette pays 35:1. There are 38 numbers on an American roulette wheel. What is the expectation value of 100 bets on red 7?
- ◇ 3. You are offered 3 doors to choose from on a game show. Behind one is a car; the other two contain goats. Your host, Monty, chose the winning door before the show by throwing a fair 3-sided die. After you choose a door, Monty will open another door. This door will always reveal a goat, and Monty will ask if you want to change your answer. (If your first choice is the car, he will reveal either goat at random 50% of the time.) Should you change your answer?

too much information !



The End

Answers:

1) 13.4%

2) -5.26 bets

3) Yes!

