## STATS 100A: Probability

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### Plan

Part 1: basic concepts and rules through examples

Part 2: random variables --- one at a time

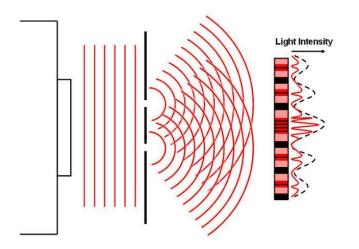
Part 3: two or more random variables

### **Emphases**

- Concepts, intuitions
- Calculations in precise notation
- Scientific applications

## Why do we need probability theory?

In the following, all the pictures are taken from internet and all the credits belong to the authors. The content of this lecture will not be in the homework assignments and final exam







### At the most fundamental level, the physical laws are probabilistic

- Two-slit experiment: an electron goes through two slits simultaneously like water waves that can be superposed
- 1. The evolution of wave function is governed by Schrodinger equation
- 2. Born: |wave function| $^2$  = probability density function
- 3. Observer's measurement, wave function collapses
- Heisenberg uncertainty principle
- Bohr: wave-particle duality
- Einstein (against probabilistic interpretation): God does not play dice
- Schrodinger (also against) cat paradox: superposition of alive and dead

### **Two-slit experiment**

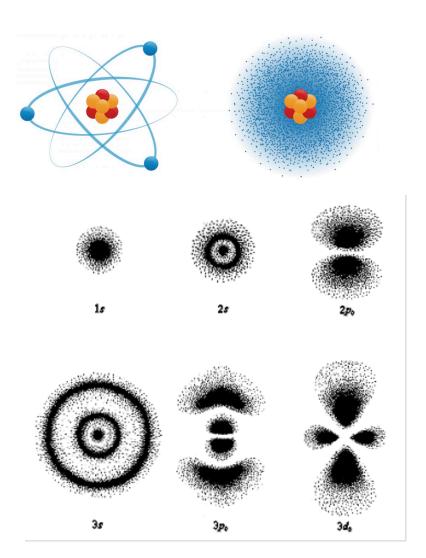


### Wave function → probability density

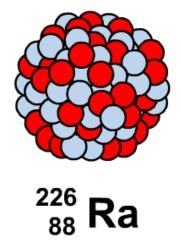


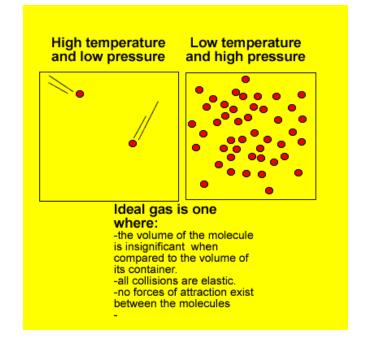
# Electron cloud: population of possibilities probability density

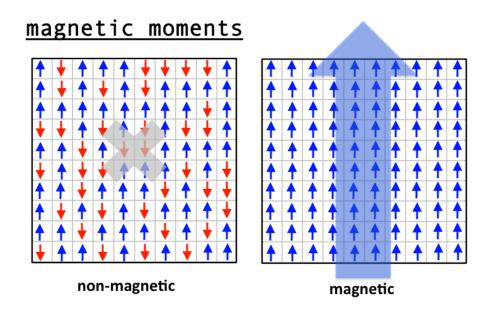
$$p(x) = |\Psi(x)|^2$$



### Particle decay

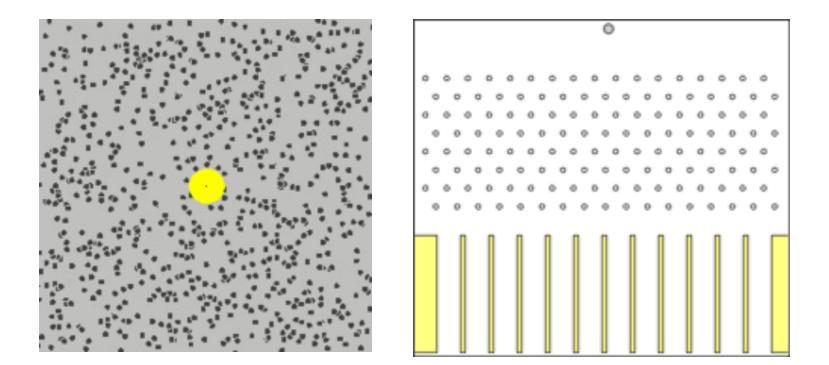




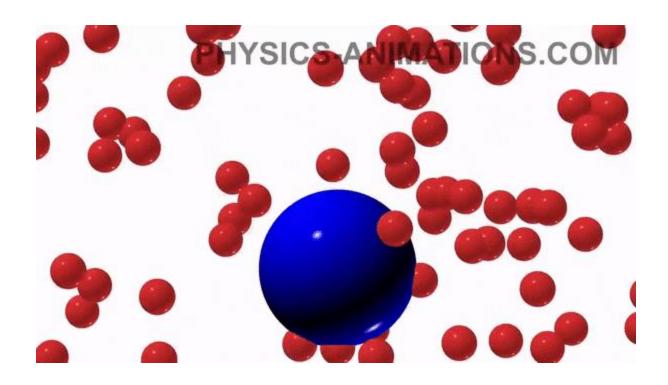


# Even if physical laws are deterministic If we study the motions of 10<sup>23</sup> elements, we need statistical models

- Configuration X in Omega =  $\{X: \text{energy}(X) = \text{fixed}\}\$ X(t) evolves in Omega and traverses every configuration with equal frequency at a random time t,  $X(t) \sim \text{uniform}(\text{Omega}) \rightarrow \text{subsystem } x(t)$  of X(t)
- Ideal gas: temperature, pressure etc. are statistical properties
- Ferromagnetism: Ising model, magnetism is also statistical property
- Phase transition: water/ice/gas, same model  $\rightarrow$  different distributions
- Arrow of time: increase in entropy, again a statistical property



- Brownian motion: dust particles in water show zig-zag paths
- Einstein: caused by bombardments from invisible water molecules
- established existence of molecules and atoms
- Left: 2D; Right: 1D simplified, Galton board/quincunx, Gaussian, CLT
- Stochastic differential equation
- Finance, Black-Scholes equation
- Diffusion models for generative AI







# Artificial intelligence, machine learning, deep learning, computer vision, natural language processing, robotics ...

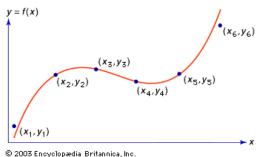
All based on probabilistic framework

• Training data come from a probability distribution

$$(x_i, y_i, i = 1, ..., n) \sim p(x, y)$$

• Learn about this probability distribution

$$\max_{\theta} \frac{1}{n} \sum_{i=1}^{n} \log p_{\theta}(y_i|x_i) \to p_{\hat{\theta}}(y|x)$$



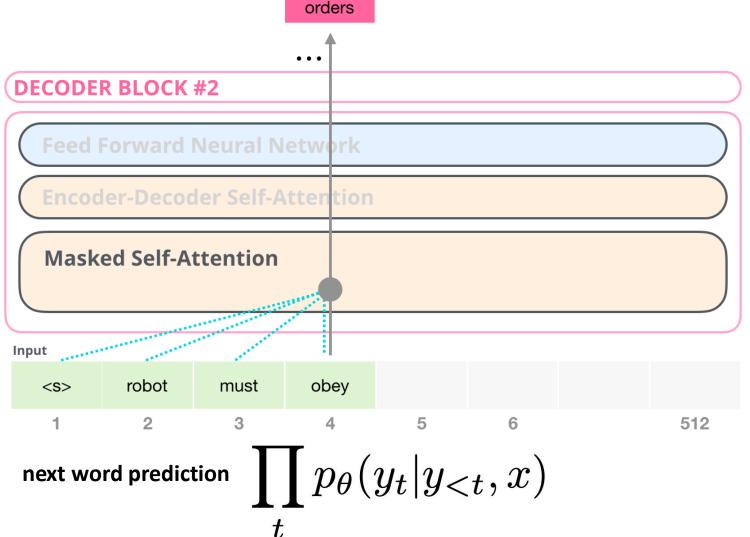
- Generalize to testing data: x: instruction, y: completion
- Memorization and generalization

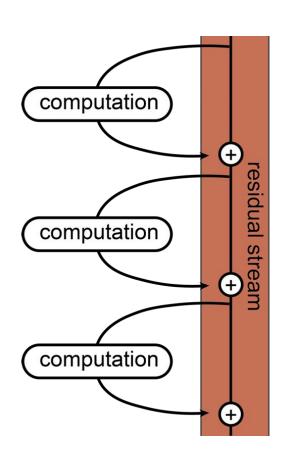
Generative Pre-Trained Transformer

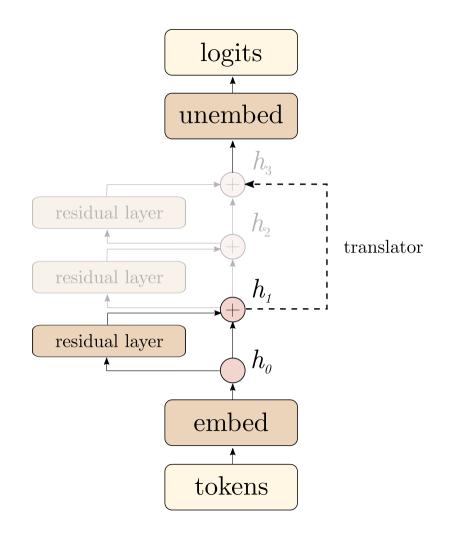
GPT (2018)













Diffusion denoising probability model 
$$\prod_t p_{ heta}(y_{t-1}|y_t,x)$$



Alone astronaut on Mars, mysterious, colorful, hyper realistic



Pyramid shaped mountain above a still lake, covered with snow





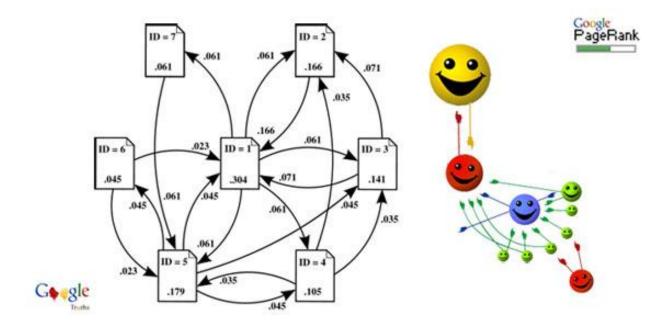
Prompt: A stylish woman walks down a Tokyo street filled with warm glowing neon and animated city signage. She wears a black leather jacket, a long red dress, and black boots, and carries a black purse. She wears sunglasses and red lipstick. She walks confidently and casually. The street is damp and reflective, creating a mirror effect of the colorful lights. Many pedestrians walk about.











### Google page-rank, random surfing model

- 85% probability, randomly click a link to the next page
- 15% probability, randomly go to another page
- Markov chain → stationary distribution
- Imagine 1 billion people follow the same random walk
- Eventually they will be distributed over all the webpages
- Popularity of a page is based on its stationary distribution



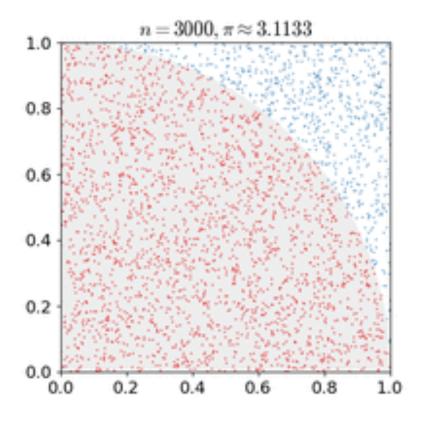


#### **Monte Carlo algorithm**

- Monte Carlo European Las Vegas
- Stan Ullman: uncle going to Monte Carlo
- Metropolis: Monte Carlo algorithm
- High-dimensional integration or optimization
- Ironically, the most accurate algorithms may be based on randomness
- Avoid the curse of dimensionality
- population (e.g., 300 million)  $\rightarrow$  sample (e.g., 1000 examples)

### Metropolis algorithm: ranked #1 algorithm in scientific computing

- The first computers: MANIAC
- Manhattan project: atomic bomb, hydrogen bomb
- Markov chain Monte Carlo



Foundation for statistics, machine learning, AI, etc.

Applications in many other areas and disciplines

Importance only comes after calculus and linear algebra