## Doing Physics with Random Numbers

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## Concepts

- Random numbers can be used to measure things that aren't so random
- Uncertainty in averages can be estimated from the measurements themselves
- A Markov process can be used to sample from a probability distribution
- Physical properties can be computed using a Markov process



## Monte Carlo Simulation: Buffon's Needle

- Consider a grid of equally spaced lines, separated by a distance *d*
- Take a needle of length *l*, and repeatedly toss it at random on the grid
- Record the number of "hits", times that the needle touches a line, and "misses", times that it doesn't
  - OK, do it! (Buffon's toothpick)



## Monte Carlo Simulation: Buffon's Needle

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- Take a needle of length *l*, and repeatedly toss it at random on the grid
- Record the number of "hits", times that the needle touches a line, and "misses", times that it doesn't
- Buffon showed that the probability of a "hit" is

$$P = \frac{2l}{\pi d}$$

• This experiment provides a means to evaluate  $\pi$ 

$$\pi = \frac{2l}{Pd}$$

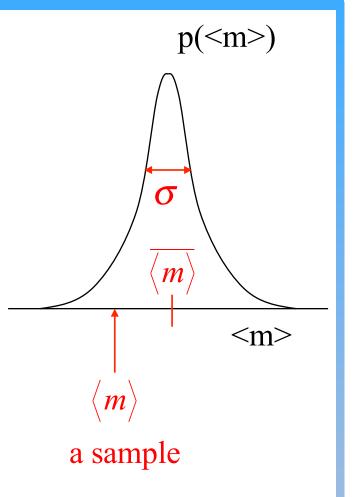


# Quantifying Uncertainty

- Averages <m> obtained by a stochastic process are known to follow a Gaussian distribution
- Any given <m> will represent a sample from this distribution
- The width of the distribution is related to the standard deviation of the same set of numbers used to compute <m>

$$\sigma_{\langle m \rangle}^2 = \frac{1}{n} \sigma_m^2$$

• Use this to quantify uncertainty in <m>





## The Weather

- Model the weather as having three states:
  - Sunny, Cloudy, Rainy
- The weather tomorrow is related to the weather today
  - For example, Cloudy today:
    - 10% chance it will be Cloudy again tomorrow
    - 50% chance it will be Sunny tomorrow
    - 40% chance it will be Rainy tomorrow
- *Transition probabilities* define the likelihood of the state of the weather tomorrow given the state of the weather today
- We might ask, knowing all (9) transition probabilities, what are the fraction of days that are Sunny, Cloudy, and Rainy?



#### Markov Processes

#### Random walk

 movement through a series of well-defined states in a way that involves some element of randomness ("stochastic")

#### Markov process

- random walk that has no "memory"
  - selection of next state depends only on current state, and not on prior states
- process is fully defined by a set of <u>transition probabilities</u>  $p_{ij}$
- $p_{ij}$  = probability of selecting state j next, given that presently in state i.



# Transition-Probability Matrix

# • Example - system with three states If Cloudy, will stay Cloudy tomorrow, with probability 0.1 I = Cloudy 2 = Sunny 3 = Rainy $\Pi \equiv \begin{pmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{pmatrix} = \begin{pmatrix} 0.1 & 0.5 & 0.4 \\ 0.9 & 0.1 & 0.0 \\ 0.3 & 0.3 & 0.4 \end{pmatrix}$ Never Rainy tomorrow if Sunny today

- Requirements of transition-probability matrix
  - all probabilities non-negative, and no greater than unity
  - sum of each row is unity
  - probability of staying in present state may be non-zero



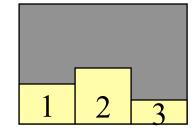
## Distribution of State Occupancies

• Consider process of repeatedly moving from one state to the next, choosing each subsequent state according to  $\Pi$ 

$$-1 \rightarrow 2 \rightarrow 2 \rightarrow 1 \rightarrow 3 \rightarrow 2 \rightarrow 2 \rightarrow 3 \rightarrow 3 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow \text{etc.}$$

Histogram the occupancy number for each state

$$- n_1 = 3$$
  
 $- n_2 = 5$   
 $- n_3 = 4$ 
 $p_1 = 0.33 \text{ (Cloudy)}$   
 $p_2 = 0.42 \text{ (Sunny)}$   
 $p_3 = 0.25 \text{ (Rainy)}$ 



- After very many steps, a limiting distribution emerges
- <u>Click here</u> for an applet that demonstrates a Markov process and its approach to a limiting distribution



## Some Uses of Markov Processes

- Physics (molecular simulation)
- Chemistry (modeling kinetics)
- Designing tests
- Speech recognition
- Information science
- Queuing theory
- Google PageRank
- Economics and Finance
- Social sciences
- Mathematical biology
- Genetics
- Games
- Algorithmic music composition
- Text generators



## Designer Transition Probabilities

- Say we want to sample states with a desired probability distribution ( $p_i$  are given), using a Markov process
- How do we design transition probabilities  $p_{ij}$ ?
- Many choices are possible for a given distribution
- Metropolis algorithm provides a good choice

$$p_{ij} = \min\left(\frac{p_j}{p_i}, 1\right)$$

$$p_{ii} = 1 - \sum_{i \neq i} p_{ij}$$



#### In-class Markov Process

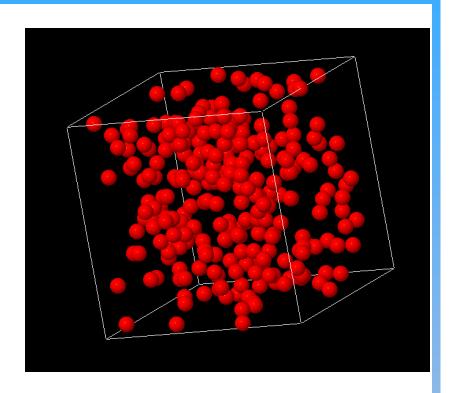
#### Three groups

- Group 1:  $p_1 = 0.3$
- Group 2:  $p_2 = 0.1$
- Group 3:  $p_3 = 0.6$
- For state at group *n*, perform trial:
  - Select one of the other states (m) with equal probability
  - Compare state probabilities
    - If  $p_m > p_n$ , let new state be m; done with trial
    - Otherwise, select a random number *r* in (0, 1)
    - If  $p_m / p_n > r$ , let new state be m; done with trial
    - Otherwise, let new state be *n* (again); done with trial
- Repeat



# Calculating Physical Properties

- Molecular simulation
- Generate box of atoms/ molecules
- Postulate a model for how they interact
- Generate configurations appropriate to postulated model
- Record averages over generated configurations





# Generating Configurations

#### Monte Carlo

- Sample configurations using a Markov process
- Atoms move around randomly, but in a controlled way
- Movements aren't physically meaningful, but the sampled distribution of configurations are

#### • Molecular dynamics

- Sample configurations according to Newton's laws
  - F = ma
- Move/accelerate all atoms at once
- Direction and amount depends on current velocities and forces
- Movements looks realistic, like a movie
- Let's see some demos...



# An Application of Molecular Simulation



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