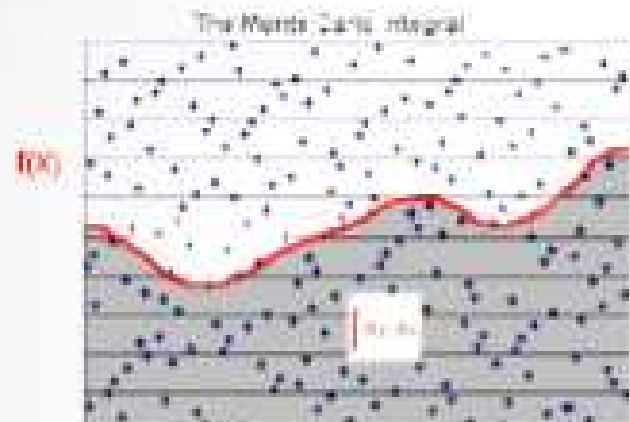


# Simulations in Python

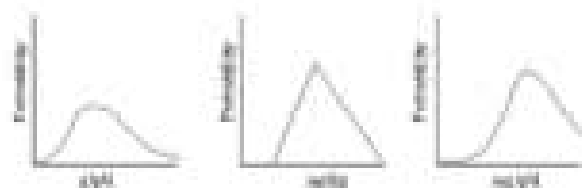
# Some Applications



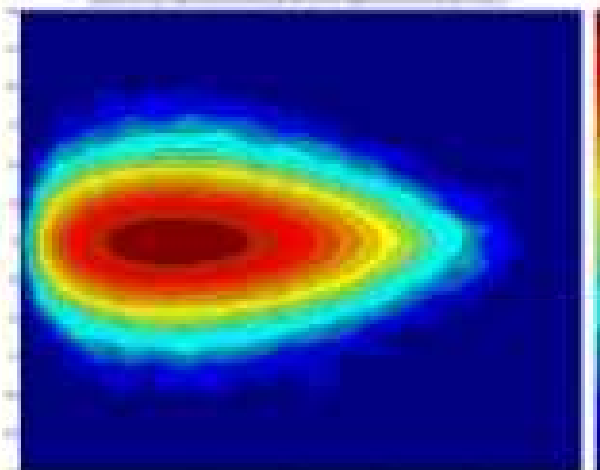
<http://marcoagel.usuarios.nrc.puc-rio.br/quasimodo.html>

## Monte Carlo Simulation

Probability Density Function (PDF) of the Sum of Two Independent Random Variables



US Food and Drug Administration



Dr.-Ing. Matthias Westhäuser, Statistical Analysis of Fiber Optical Systems using Multicanonical Monte Carlo Methods (<http://www.ftt.e-technik.fu-dortmund.de/forschungsprojekt.php?id=18&lang=en>)

# What is a Monte Carlo method?

- An algorithm that uses a source of (pseudo) random numbers
- Repeats an “experiment” many times and calculates a statistic, often an average
- Estimates a value (often a probability)
- ... usually a value that is hard or *impossible* to calculate analytically

# Simple example: dice statistics

- We can **analyze** throwing a pair of dice and get the following probabilities for the sum of the two dice:

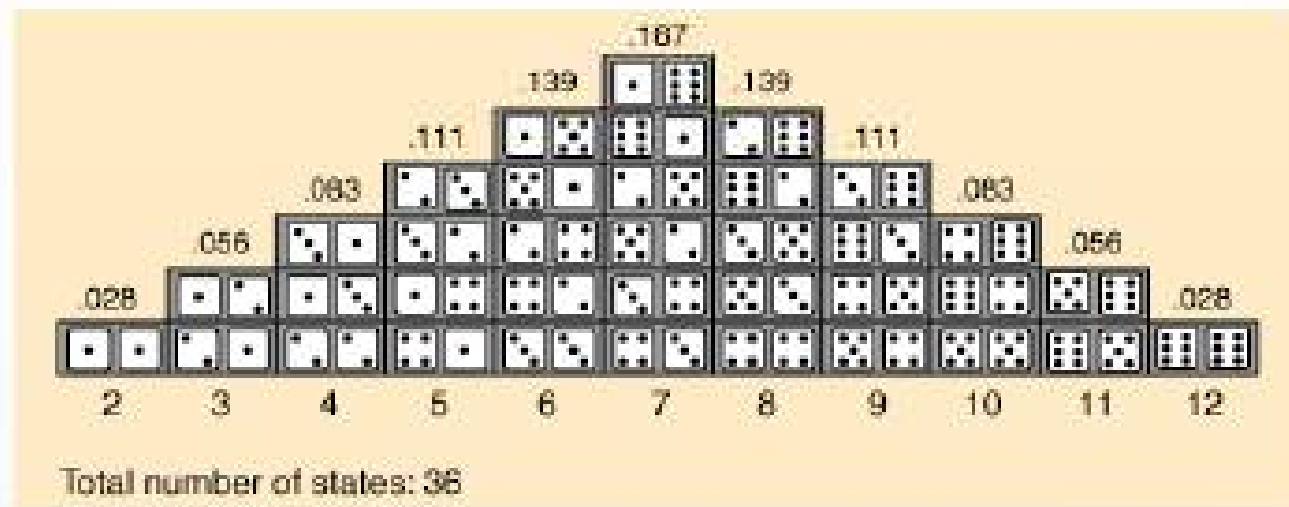


Image source:

<http://www.goldsim.com/Web/Introduction/Probabilistic/MonteCarlo/> via <http://www.goldsim.com/Web/Introduction/Probabilistic/MonteCarlo/>

## Simple example: dice statistics

- ... **or** we can throw a pair of dice 100 times and record what happens, or 10000 times for a more accurate estimate.

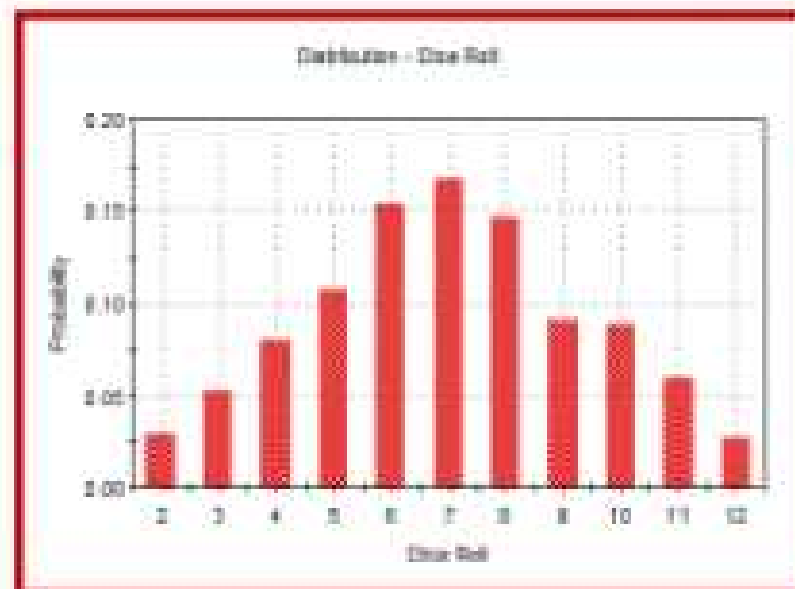
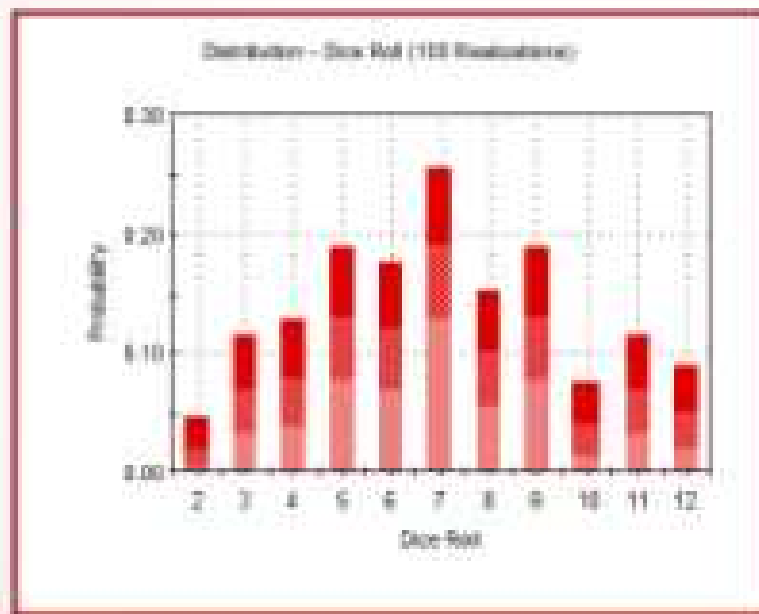


Image source:

<http://www.goldsim.com/Web/Introduction/Probabilistic/MonteCarlo/> via

# A game of dice

```
def dice_game() :  
    strikes = 0  
    winnings = 0  
    while strikes < 3 : # 3 strikes and you're out  
        die1 = roll() # a random number 1...6  
        die2 = roll()  
        if die1 == die2 :  
            strikes = strikes + 1  
        else :  
            winnings = winnings + die1 + die2  
    return winnings    # in cents
```

## The Hungry Dice Player

- In our simple game of dice:  
*Can I expect to make enough money playing it to buy lunch?*
- That is, what is the expected (average) value won in the game?
- We could figure it out by applying laws of probability
- ...or use a Monte Carlo method

## Monte Carlo method for the hungry dice player

```
def average_winnings(runs) :  
    # runs is the number of experiments to run  
    total = 0  
    for n in range(runs) :  
        total = total + dice_game()  
    return total/runs  
  
>>> [round(average_winnings(10),2) for i in range(5)]  
[85.8, 94.8, 120.7, 123.3, 90.0]  
>>> [round(average_winnings(100),2) for i in range(5)]  
[105.97, 102.95, 107.74, 134.4, 114.54]  
>>> [round(average_winnings(1000),2) for i in range(5)]  
[106.84, 107.11, 105.59, 104.28, 106.41]  
>>> [round(average_winnings(10000),2) for i in range(5)]  
[104.94, 105.71, 105.81, 105.74, 104.62]
```

# The Clueless Student

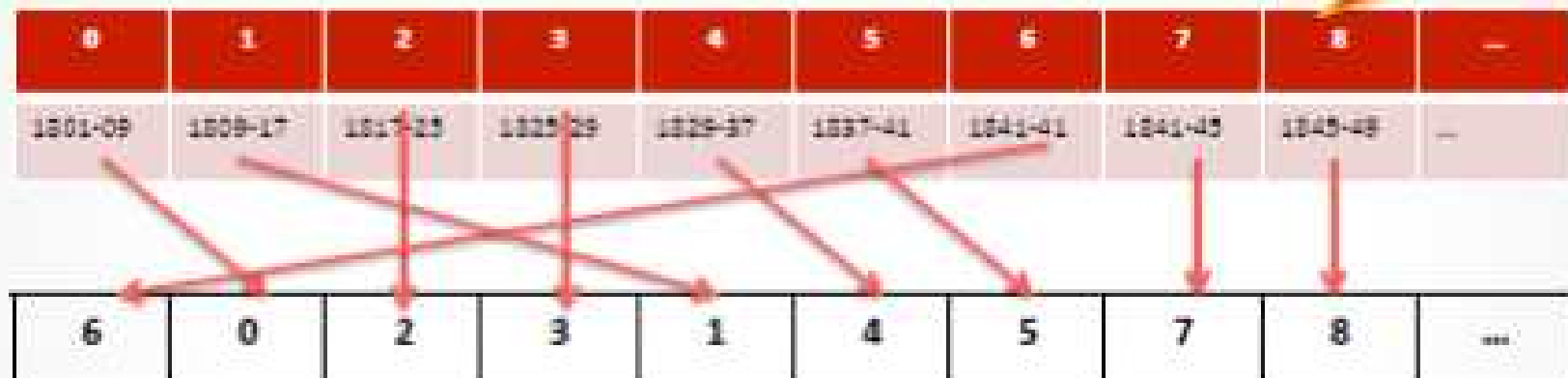
*A clueless student faced a pop quiz: a list of the 24 Presidents of the 19<sup>th</sup> century and another list of their terms in office, but scrambled. The object was to match the President with the term. If the student guesses a random one-to-one matching, how many matches will be right out of the 24, on average?*

## The quiz

1. Monroe	a. 1801-1809
2. Jackson	b. 1869-1877
3. Arthur	c. 1885-1889
4. Madison	d. 1850-1853
5. Cleveland	e. 1889-1893
6. Jefferson	f. 1845-1849
7. Lincoln	g. 1837-1841
8. Van Buren	h. 1859-1857
9. Adams	i. 1809-1817
etc.	etc.



# Representing a guess



0	1	2	3	4	5	6	7	8	—
Jefferson	Madison	Monroe	Adams	Jackson	Van Buren	Harrison	Tyler	Polk	—

indexes

# Representing a guess

- Representing a guess – examples:
  - [ 0, 1, 2, 3, 4, 5, ..., 23 ] represents a completely correct guess
  - [ 1, 0, 2, 3, 4, 5, ..., 23 ] represents a guess that is correct except that it gets the first two presidents wrong.
    - A guess is just a permutation (shuffling) of the numbers 0 ... 23.
- Let's define a *match* in a guess to be any number  $k$  that occurs in position  $k$ . (E.g., 0 in position 0, 10 in position 10)
- With this representation, our question becomes: *if I pick a random shuffling of the numbers 0...23, how many (on average) matches occur?*

# Randomly permuting a list

To get a random shuffling of the numbers 0 to 23 we use the `shuffle` function from module `random`:

```
>>> nums = list(range(10))
```

```
>>> nums
```

```
[0, 1, 2, 3, 4, 5, 6, 7, 8, 9]
```

```
>>> shuffle(nums)
```

```
>>> nums
```

```
[4, 5, 3, 2, 0, 9, 6, 1, 8, 7]
```

```
>>> shuffle(nums)
```

```
>>> nums
```

```
[3, 6, 1, 4, 5, 8, 2, 9, 0, 7]
```

We will solve a more general problem

# Algorithm

- Input: *pairs* (number of things to be matched), *samples* (number of samples to test)
- Output: average number of correct matches per sample
- Method
  1. Set *num\_correct* = 0
  2. Do the following *samples* times:
    - a. Set *matching* to a random permutation of the numbers  $0 \dots pairs-1$
    - b. For  $i$  in  $0 \dots pairs$ , if  $matching[i] = i$  add one to *num\_correct*
  3. The result is  $num\_correct / samples$

# Code for the clueless student

```
from random import shuffle
# pairs is the number of pairs to be guessed
# samples is the number of samples to take
def student(pairs, samples) :
    num_correct = 0
    matching = list(range(pairs))
    for i in range(samples) :
        shuffle(matching)          # generate a guess
        for j in range(pairs) :
            if matching[j] == j :
                num_correct = num_correct + 1
    return num_correct / samples
```

## Running the code

- The mathematical analysis says the expected value is exactly 1 (**no matter how many matches are to be guessed**).

```
>>> student(24, 10000)
0.9924
>>> student(24, 10000)
1.0071
>>> student(10, 10000)
1.0224
>>> student(10, 10000)
0.9999
>>> student(5, 10000)
1.0039
>>> student(5, 10000)
0.9826
```

## More samples – smaller error

```
>>> 1 - student(5, 1000)
0.036000000000000003
>>> 1 - student(5, 10000)
0.0059000000000000016
>>> 1 - student(5, 100000)
0.00141000000000000223
>>> 1 - student(5, 1000000)
-0.00066799999999998909
```

# The Umbrella Quandary

- Mr. X walks between home and work every day
- He likes to keep an umbrella at each location
- But he always forgets to carry one if it's not raining
- If the probability of rain is  $p$ , how many trips can he expect to make before he gets caught in the rain? (Assuming that if it's not raining when he starts a trip, it doesn't rain during the trip.)



# The trivial cases

- What if it always rains?
- What if it never rains (ok, that was too easy)
- So we only need to think about a probability of rain greater than zero and less than one

# Solving the umbrella quandary

- Analysis of the problem can be done with Markov chains
- But we're just humble programmers, we'll simulate and measure

# Simulating an event with a given probability

- In contrast to the clueless student problem we're given a probability of an event
- We want to simulate that the event happens, with the given probability  $p$  (where  $p$  is a number between 0 and 1)
- Technique: get a random float between 0 and 1; if it's less than  $p$  simulate that the event happened

```
if random() < p :  
    raining = True
```

## Representing home, work, and umbrellas

- Use 0 for home, 1 for work, and a two-element list for the number of umbrellas at each location
- How should we initialize?
- `location = 0`  
`umbrellas = [1, 1]`

## Figuring out when to stop

- We want to count the number of trips before Mr. X gets wet, so we want to keep simulating trips until he does.
- To keep track:
- `wet = False`  
`trips = 0`  
`while (not wet) :`  
`...`

# Changing locations

- Mr. X walks between home (0) and work (1)
  - To keep track of where he is:  
`location = 0 # start at home`
  - To move to the other location:  
`location = 1 - location`
  - To find how many umbrellas at current location:  
`umbrellas[location]`

## Putting it together

```
from random import random

def umbrella(p) :           # p is the probability of rain
    wet = False
    trips = 0
    location = 0
    umbrellas = [1, 1]     # index 0 stands for home, 1 stands for work
    while (not wet) :
        if random() < p :   # it's raining
            if umbrellas[location] == 0 : # no umbrella
                wet = True
            else :
                trips = trips + 1
                umbrellas[location] -= 1      # take an umbrella
                location = 1 - location        # switch locations
                umbrellas[location] += 1      # put umbrella
        else :               # it's not raining, leave umbrellas where they are
            trips = trips + 1
            location = 1 - location
    return trips
```

# Running simulations

```
>>> umbrella(.5)
22
>>> umbrella(.5)
4
>>> umbrella(.5)
13
>>> umbrella(.5)
2
>>> umbrella(.5)
2
```

## Great, but we want averages

- One experiment doesn't tell us much—we want to know, **on average**, if the probability of rain is  $p$ , how many trips can Mr. X make without getting wet?
- We add code to run `umbrella(p)` 10,000 times for different probabilities of rain, from  $p = .01$  to  $.99$  in increments of  $.01$
- We accumulate the results in a list that will show us how the average number of trips is related to the probability of rain.

# Running the experiments

```
# 10,000 experiments for each probability from .01
to .99
# Accumulate averages in a list
def test() :
    results = [None]*99
    p = .01
    for i in range(99) :
        trips = 0
        for k in range(10000) :
            trips = trips + umbrellas(p)
        results[i] = trips/10000
        p = p + .01
    return results
```

# Crude plot of results

