

re·cur·sion (ri'kərZHən)

(noun) See recursion.

# Atul Vyas Memorial Lecture

## How to roll a five-sided die

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What  
is  
Recursion?

## *Breaking up the problem*

**Recursion** is an essential mathematical tool that breaks a problem into versions of the same problem.

## *A classic problem*

How many ways are there to arrange  $n$  objects?

Example:  $n = 3$

123

231

132

312

213

321

Each arrangement is called a **permutation**

# Factorials

## *Definition*

The number of ways to arrange  $n$  objects in a line is called  $n$  factorial, and written  $n!$ .

So for example

$$3! = 6.$$

## *A classic problem*

Think about which object gets put into position 1

1 □ □    2 □ □    3 □ □

- ▶ There are 3 choices for first position
- ▶ The remaining two boxes arrange 2 objects
- ▶ So  $3! = 3 \cdot 2!$

## *More generally*

Same idea with  $n$  objects gives a **recursive formula** for factorials

$$n! = n \cdot (n - 1)!$$

Need also a base case:

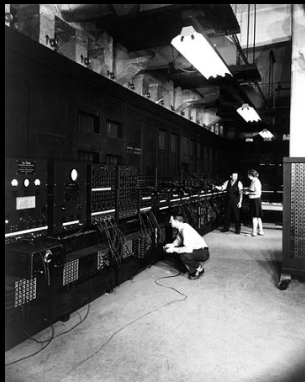
$$1! = 1$$

Can use this recursive formula for calculation

$$5! = 5 \cdot 4! = 5 \cdot 4 \cdot 3! = 5 \cdot 4 \cdot 3 \cdot 2! = 5 \cdot 4 \cdot 3 \cdot 2 \cdot 1 = 120.$$



*For a computer...*



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Factorial

*Input:  $n$*

*Output:  $n!$*

---

- 1) If  $n = 1$  then output 1 and quit
  - 2) Else output  $n \cdot \text{Factorial}(n - 1)$
-



# Simulation via recursion

## Problem

Suppose I want to generate uniformly from the set of permutations of  $n$  objects. So

$$\mathbb{P}(X = (x_1, \dots, x_n)) = \frac{1}{n!}.$$

## Notation

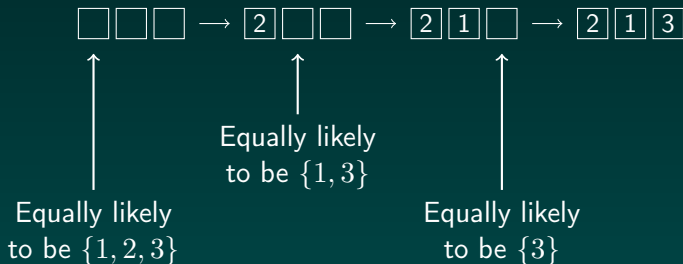
- ▶  $\mathbb{P}$  means probability of
- ▶  $(x_1, \dots, x_n)$  is an arbitrary permutation (ex: (1, 2, 3))
- ▶  $X$  is a random variable that is the random permutation

## Using recursion and symmetry

Using recursion to generate a random permutation:

- ▶ The first position is equally likely to be  $1, 2, \dots, 1/n$ .
- ▶ Then generate the rest of permutation from items that are left.

For example:



## *In pseudocode*

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Uniform\_Permutation

*Input:* item set  $I = \{i_1, \dots, i_n\}$

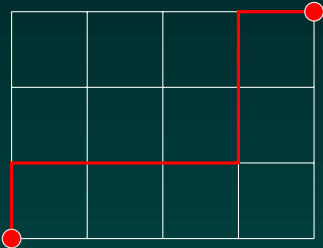
*Output:* permutation  $(x_1, \dots, x_n)$

---

- 1) If  $n = 0$  then output  $\emptyset$  and quit
  - 2) Else
  - 3)     Choose  $i$  uniformly at random from  $\{i_1, \dots, i_n\}$
  - 4)     Remove  $i$  from the set  $I$
  - 5)     Output  $(i, \text{Uniform\_Permutation}(I))$
-

# Path counting

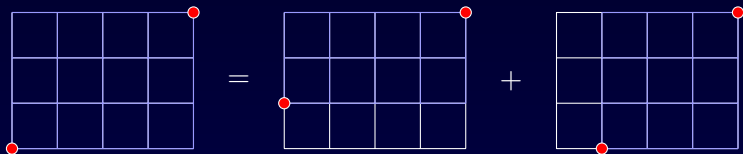
How many paths are there from  $(0,0)$  to  $(4,3)$  that use only right moves or up moves?



## Answer: use recursion!

Each path to  $(x, y)$  either begins with...

- ▶ ...an up move. # of paths from  $(1, 0)$  to  $(x, y)$  same as paths from  $(0, 0)$  to  $(x - 1, y)$ , or...
- ▶ ...a right move. # of paths from  $(0, 1)$  to  $(x, y)$  same as paths from  $(0, 0)$  to  $(x, y - 1)$ .



$$P(x, y) = P(x - 1, y) + P(x, y - 1), \quad P(0, 1) = P(1, 0) = 1.$$

## Table of paths

Using the recursive formula gives:

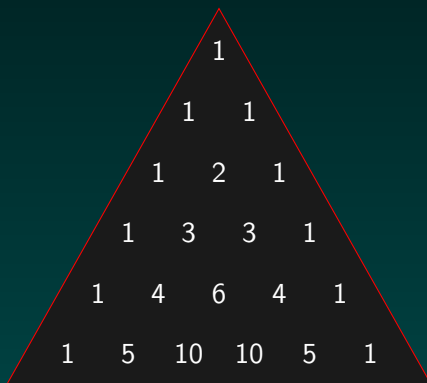
		<i>y</i>				
		0	1	2	3	4
<i>x</i>	0	1	1	1	1	1
	1	1	2	3	4	5
	2	1	3	6	10	15
	3	1	4	10	20	35
	4	1	5	15	35	70

$$P(4, 3) = 35$$



# *Pascal's Triangle*

Turned 45 degrees, gives Pascal's Triangle:



Uses: # of subsets of a set, Binomial coefficients

## Simulation question

Sampling uniformly from grid paths



Need 4 right moves and 3 up moves

Randomly permuted

First move has a  $4/(4 + 3)$  chance of being to the right

(Otherwise move up)

Recursively draw the rest of the path

# The pseudocode

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Up\_Right\_Grid\_Path

*Input:*  $x, y$

*Output:* path consisting of up and right moves

---

- 1) If  $x = y = 0$  return  $\emptyset$
  - 2) With probability  $x/(x + y)$
  - 3)      $P \leftarrow \text{Up\_Right\_Grid\_Path}(x - 1, y)$
  - 4)     Output (right,  $P$ )
  - 5) Otherwise
  - 6)      $P \leftarrow \text{Up\_Right\_Grid\_Path}(x, y - 1)$
  - 7)     Output (up,  $P$ )
-

# Structure

So far, our recursions have two elements:

- ▶ A recursive call to the same algorithm
- ▶ A base case

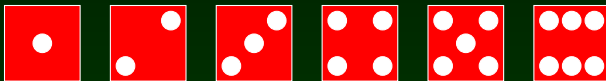
What if we don't have a base case!

# Infinite Recursion



# *A fair six sided die*

Suppose I have a die with six sides



I am equally likely on a roll to get one of  $\{1, 2, 3, 4, 5, 6\}$

I can roll the die as many times as I want

How can I use this to roll a five sided die?



# *Use infinite recursion*

Intuitive answer:

1. Roll a six sided die. If the answer is in  $\{1, 2, 3, 4, 5\}$ , output answer and stop.
2. Otherwise, recursively roll a five sided die.

## *In pseudocode*

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Fair\_five\_sided\_die

*Output:*  $X$  uniform draw from  $\{1, 2, 3, 4, 5\}$

---

- 1) Let  $X$  the roll of a fair six sided die
  - 2) If  $X \in \{1, 2, 3, 4, 5\}$ , output  $X$  and quit
  - 3) Else
  - 4)      $X \leftarrow \text{Fair\_five\_sided\_die}$
  - 5)     Output  $X$  and quit
-



## Remember our earlier code for permutations

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Uniform\_Permutation

*Input:* item set  $I = \{i_1, \dots, i_n\}$

*Output:* permutation  $(x_1, \dots, x_n)$

---

- 1) If  $n = 0$  then output  $\emptyset$  and quit
  - 2) Else
  - 3)     Choose  $i$  uniformly at random from  $\{i_1, \dots, i_n\}$
  - 4)     Remove  $i$  from the set  $I$
  - 5)     Output  $(i, \text{Uniform\_Permutation}(I))$
- 

The first line is the base case

# *Infinite recursion*

Fair\_five\_sided\_die does not have a base case!

- ▶ Could call recursives twice...
- ▶ ...or a million times...
- ▶ ...or a billion times.
- ▶ Very unlikely to do so, but it could happen!
- ▶ There is no upper bound on the number of recursions

# *Our intuition is that this works*

## *Theorem (H. 2015)*

*The Fundamental Theorem of Perfect Simulation [FTPS] says essentially that when proving a recursive probabilistic algorithm works, as long as the algorithm terminates with probability 1, you can assume that the recursive call generates from the correct distribution.*

## Applying the FTSP to Fair\_five\_sided\_die

### Fact

The output of Fair\_five\_sided\_die is equally likely to be any of  $\{1, 2, 3, 4, 5, 6\}$

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Fair\_five\_sided\_die

Output:  $X$  uniform draw from  $\{1, 2, 3, 4, 5\}$

---

- 1) Let  $X$  the roll of a fair six sided die
  - 2) If  $X \in \{1, 2, 3, 4, 5\}$ , output  $X$  and quit
  - 3) Else
  - 4)      $X \leftarrow$  Fair\_five\_sided\_die
  - 5)     Output  $X$  and quit
-

## Applying the FTSP to Fair\_five\_sided\_die

### Fact

The output of Fair\_five\_sided\_die is equally likely to be any of  $\{1, 2, 3, 4, 5\}$

### Proof

The only way the algorithm never terminates is if we roll an infinite number of 6's in a row:

6, 6, 6, 6, 6, . . . .

The chance of this happening is

$$\frac{1}{6} \cdot \frac{1}{6} \cdot \frac{1}{6} \cdots = \lim_{n \rightarrow \infty} \left(\frac{1}{6}\right)^n = 0.$$

So the FTSP can be applied!

## *Applying the FTPS to Fair\_five\_sided\_die*

### Proof (continued)

Now consider, what is  $\mathbb{P}(X = 3)$ ? Well,  $X$  could equal three in line 1, which happens with probability  $1/6$ . Or  $X = 6$  (which happens with probability  $1/6$ ), and we roll  $X$  recursively at line 4. The FTPS says that we can assume the line 4 call has the correct probability, so

$$\mathbb{P}(X = 3) = \frac{1}{6} + \frac{1}{6} \cdot \frac{1}{5} = \frac{5 + 1}{30} = \frac{1}{5}$$

The same argument gives  $\mathbb{P}(X = i) = 1/5$  for  $i \in \{1, 2, 3, 4, 5\}$ , so we're done!  $\square$

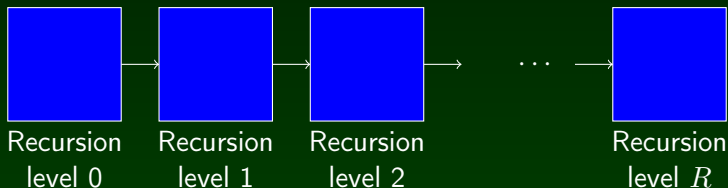
## *More about the FTSP*

### The Fundamental Theorem of Perfect Simulation

- ▶ Very useful in proofs for probabilistic recursive algorithms
- ▶ Another term for probabilistic recursive algorithm is perfect simulation
- ▶ Hence the name
- ▶ Why does it work?

# *Fundamental Theorem of Perfect Simulation proof*

Original algorithm uses recursion a random number of times  $R$ :



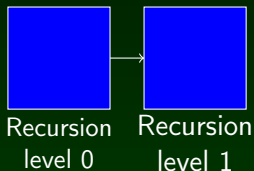
Call the output of this algorithm  $X$



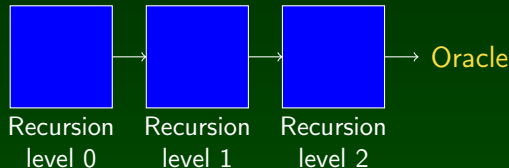
# Fundamental Theorem of Perfect Simulation proof

Set  $n$ , and if  $R > n$ , use magic (oracle) instead of recursion.

Example:  $n = 2$  and  $R = 1$



Example:  $n = 2$  and  $R = 4$



Call the output of this algorithm  $Y_n$

# *Fundamental Theorem of Perfect Simulation proof*

## Outline of proof of FTSP

- ▶ Because we used the oracle,  $Y_n$  has the target distribution
- ▶ Now consider as  $n$  becomes larger and larger...
- ▶ Because  $R$  is finite with probability 1,

$$\lim_{n \rightarrow \infty} Y_n = X \text{ (with probability 1)}$$

- ▶ So  $X$  also must have the correct distribution!

Flipping a biased coin

## *Biased coin*

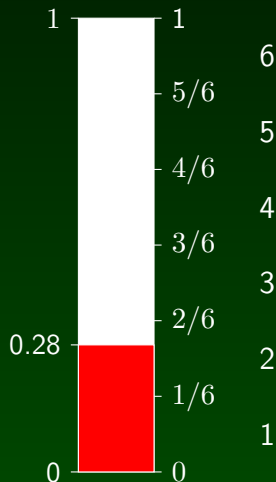
Suppose that I want to simulate an event

- ▶ The event happens 28% of the time.
- ▶ Represent by a picture:



- ▶ All I have is my trusty six sided die

## Divide probability using the die



Let  $X$  be a roll of the die

If  $X = 1$  definitely in red region

If  $X \geq 3$  definitely in white region

If  $X = 2$  could be either red or white

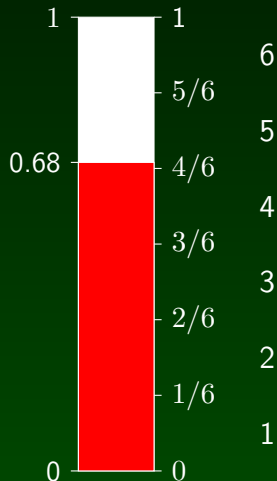
## What to do when $X = 2$ ? Recursion!

When  $X = 2$ , the bar looks like:



- ▶ The percentage of red area is  $(0.28 - 1/6)/(2/6 - 1/6) = 0.68$
- ▶ Flip a new coin recursively with probability 68%

## *Divide probability using the die*



Let  $X$  be a roll of the die

If  $X \leq 4$  definitely in red region

If  $X = 6$  definitely in white region

If  $X = 5$  could be either red or white

## *How to find regions*

Floor function rounds down to nearest integer:

$$\lfloor 4.2 \rfloor = 4, \quad \lfloor 3 \rfloor = 3$$

Ceiling function rounds up to nearest integer:

$$\lceil 4.2 \rceil = 5, \quad \lceil 3 \rceil = 3.$$

For  $p = 0.68$ ,

- ▶  $X \leq \lfloor 6 \cdot 0.68 \rfloor = \lfloor 4.08 \rfloor = 4$  in red region
- ▶  $X \geq \lceil 6 \cdot 0.68 \rceil + 1 = 6$  in white region



# Pseudocode

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Event\_decision

*Input:*  $p$  (probability event occurs)

*Output:* S or F (Success or Failure)

---

- 1) Randomly draw  $X$  uniformly from  $\{1, 2, 3, 4, 5, 6\}$
  - 2) If  $X \leq \lfloor 6p \rfloor$
  - 3)     Output S and quit
  - 4) Elseif  $X \geq \lceil 6p \rceil + 1$
  - 5)     Output F and quit
  - 6)     Output Event\_decision( $6p - \lfloor 6p \rfloor$ )
-

## *Notes on Event\_decision*

The input  $p$  can change at every level of recursion:

$$0.28 \rightarrow 0.68 \rightarrow 0.08 \rightarrow 0.48 \rightarrow 0.88 \rightarrow 0.28 \rightarrow$$

Sequence cycles if and only if original  $p$  is a rational number.

Can use FTSPS to prove that algorithm is correct

# Independent Sets of a Graph

# Graphs

## *Definition*

A **graph** consists of **nodes** and **edges** which connect pairs of nodes



A line graph with 5 nodes

# Independent sets

## Definition

An **independent set** of a graph is a subset of nodes, no two of which are adjacent.



Not an independent set



An independent set

*Question: how many independent sets are there in a line graph with  $n$  nodes?*

The node on one end can be out of the ind. set, leaving  $n - 1$  nodes



The node on the end is in the ind. set, so the node next is out, leaving  $n - 2$  nodes



The recursive formula is

$$F(n) = F(n - 1) + F(n - 2).$$

# Fibonacci

Number of independent sets in a line graph with  $n$  nodes:

$$F(n) = F(n - 1) + F(n - 2), \quad F(0) = F(-1) = 1.$$

This gives rise to the famous **Fibonacci** sequence

$$1, 1, 2, 3, 5, 8, 13, 21, 34, 55, \dots$$

## *Leonardo of Pisa aka Fibonacci*



In the **Liber Abaci** (Book of Calculation), he introduced Hindu-Arabic numerals to Europe, and introduced the Fibonacci sequence as an example.

1170-1250



# Simulating uniformly over the independent sets

Can try acceptance rejection:

- ▶ For each node, flip a fair coin
- ▶ If heads, try to put the node in the independent set
- ▶ If result is actually an independent set accept, otherwise reject and start again

Draw a sample



reject and try again



accept as independent set!

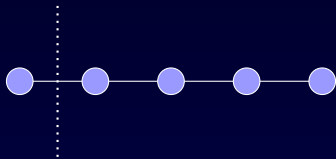
*One tiny problem...*

This is very, very, slow!

## *Speeding things up....*

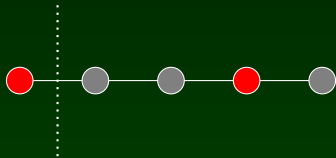
Suppose that I break the line graph into two pieces

This is called a **cut** of the graph



## *Piece by piece AR*

Draw independent sets for the two sides of the cut



## *Bringing the two halves together*



Accept!



Reject!



Accept!



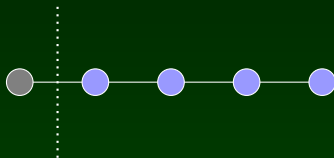
Accept!

If together they form an independent set of original graph accept

Otherwise start over

## *Something to note*

If the left side is not in the independent set, accept for any right side!



We don't even need to draw the right side in order to accept!

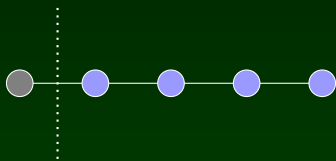
## *New algorithm!*

Generate first node of independent set of a line graph with  $n$  nodes

1. Uniformly choose node 1 in or out of the ind set. If it is out of ind set, accept and return
2. Else recursively draw the value for node 2 of ind set on nodes  $\{2, \dots, n\}$
3. If node 2 is in the ind set, reject both sides and start over.
4. Otherwise, accept and return

## *New algorithm in action*

First random choice puts node 1 out of independent set...

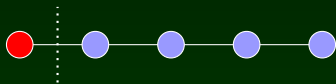


Accept and return

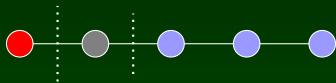


## *New algorithm in action*

First random choice puts node 1 in independent set...



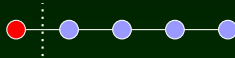
Recursively draw value for node 2. Say it is out of ind set



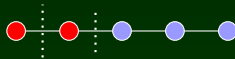
Then accept node 2, which means also accept node 1!

# *New algorithm in action*

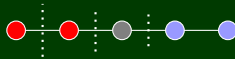
First random choice puts node 1 in independent set...



Recursively draw value for node 2. Say node 2 is in the ind set



Recursively draw value for node 3. Say it is out of the ind set



Then node 2 gets accepted as in the ind set, which means node 1 gets rejected because it was next to a node already in the independent set! Reset and start over!

## *What is the chance that left most node is in the independent set*

Suppose we have a line graph with  $n$  nodes...

...let  $p_n$  be chance that left most node is in independent set

- ▶ To be in the independent set, we have to try to put the node in the independent set. This happens with probability  $1/2$ .
- ▶ Then either 1) the node next to it is not in the independent set or 2) the node next to it is in the independent set and we reject, start over, and eventually put the first node in.

$$p_n = \frac{1}{2} [(1 - p_{n-1}) + p_{n-1}p_n]$$

## *The sequence of probabilities*

So we know:

$$p_n = \frac{1}{2} [(1 - p_{n-1}) + p_{n-1}p_n]$$

Doing some algebra:

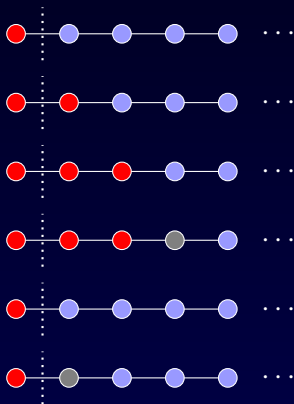
$$p_n = \frac{1 - p_{n-1}}{2 - p_{n-1}}$$

Use this and  $p_1 = 1/2$  to find the first few values:

$$p_1 = \frac{1}{2}, p_2 = \frac{1}{3}, p_3 = \frac{2}{5}, p_4 = \frac{3}{8}, p_5 = \frac{5}{13}, \dots$$

# *Infinite line graph*

This algorithm works even if you have an infinite number of nodes!



Accept!

## *What is the chance that the first node is in the set*

Instead of

$$p_n = \frac{1 - p_{n-1}}{2 - p_{n-1}}$$

you get

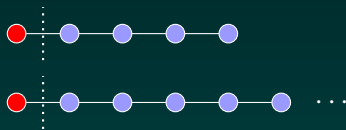
$$p = \frac{1 - p}{2 - p}$$

Unique solution:

$$p = (1/2)(3 - \sqrt{5}) = 0.3857\dots$$

# Moving from finite graph to infinite graph

As the graph gets larger,  $p_n \rightarrow p$ :



## *Related to a well known constant*

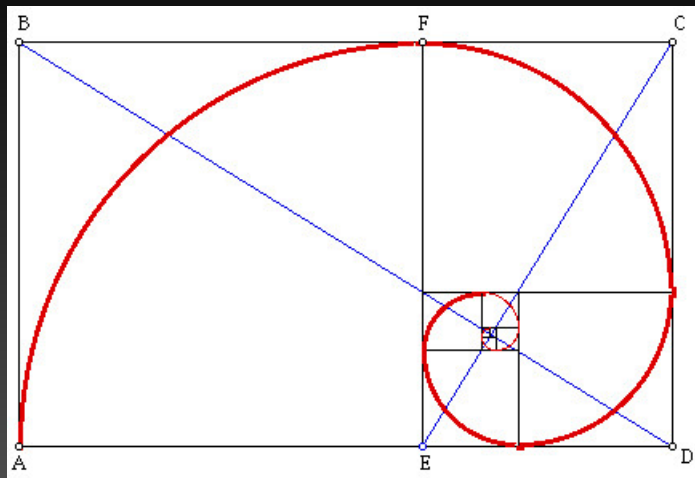
The constant can be rewritten as

$$p = \frac{1}{2}(3 - \sqrt{5}) = 1 - \frac{1}{\phi},$$

where  $\phi$  is the **Golden Ratio**



# *Geometric view of Golden Ratio*

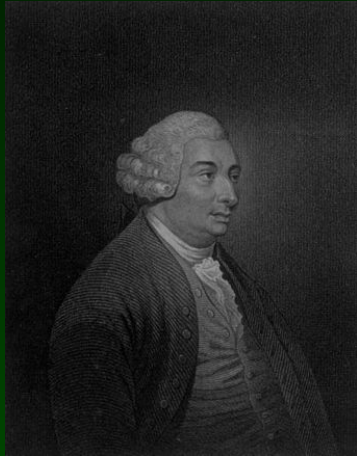


# *The Fundamental Theorem of Perfect Simulation*

Gives us multiple ways to simulate from tough examples:

- ▶ Acceptance/rejection
- ▶ Coupling from the past
- ▶ Randomness Recycler
- ▶ Popping algorithms
- ▶ Bernoulli factory

# *The wonderful idea of infinity*



*How can we satisfy ourselves  
without going on in infinitum?  
And, after all, what satisfaction  
is there in that infinite  
progression?*

David Hume, 1779

Thank you!