Source: www.keithschwarz.com/darts-dice-coins/

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Why alias method?

You are given an *n*-sided die where side *i* has probability p_i of being rolled.

How do you efficiently simulate S rolls of the die (for large S)?

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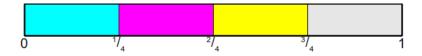
- Vanilla inverse CDF method: O(nS)
 - each sample costs O(n)

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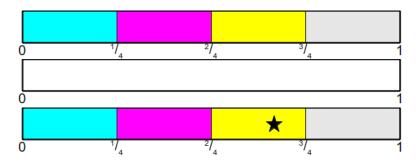
How do you efficiently simulate S rolls of the die (for large S)?

- Vanilla inverse CDF method: O(nS)
 - each sample costs O(n)
- Vose's alias method: O(n) pre-processing + O(S)
 - each sample costs O(1)!

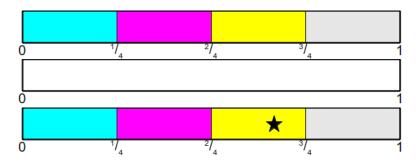




1. Sample *x* uniformly from [0, 1)

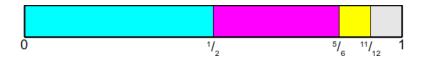


- 1. Sample *x* uniformly from [0, 1)
- 2. Return floor(xn)

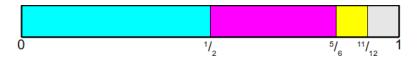


- 1. Sample *x* uniformly from [0, 1)
- 2. Return floor(xn)
 - Generate uniform r. v. in O(1)
 - Find bin in O(1)

Simulating a loaded die using inverse cdf method



Simulating a loaded die using inverse cdf method



Pre-processing: compute $c_i = \sum_{j=0}^i p_j$

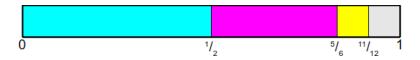
c = cumsum(p)

For each of the *S* samples:

- 1. Sample x uniformly from [0, 1)
- 2. Find minimum *i* such that $x < c_i$

 $x \in [\sum_{j=0}^{i-1} p_j, \sum_{j=0}^{i} p_j)$

Simulating a loaded die using inverse cdf method



Pre-processing: compute $c_i = \sum_{j=0}^i p_j$

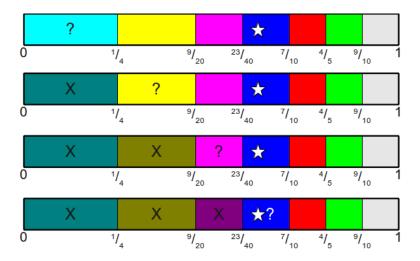
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For each of the S samples:

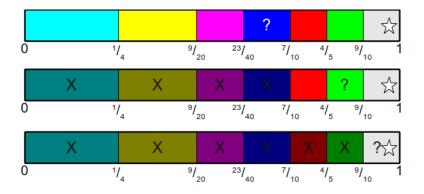
- 1. Sample x uniformly from [0, 1)
- 2. Find minimum *i* such that $x < c_i$
 - Generate uniform r. v. in O(1)
 - Find bin not O(1) anymore

$$x \in [\sum_{j=0}^{i-1} p_j, \sum_{j=0}^{i} p_j)$$

Inverse CDF method: Linear search O(n)

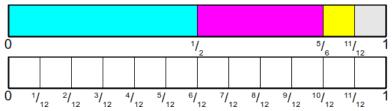


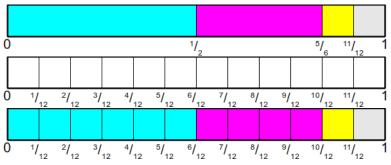
Inverse CDF method: Binary search $O(\log n)$

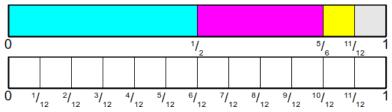


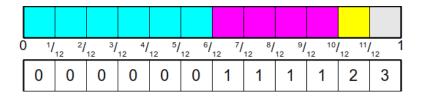


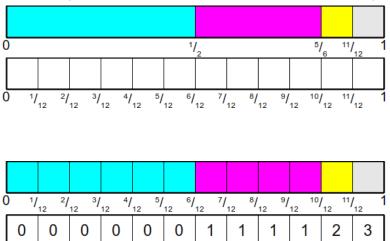










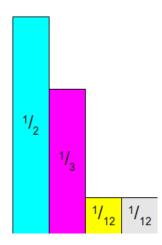


- Can sample in O(1)!
- Memory depends on LCM of the denominators: best O(n), worst case O(∏ⁿ_{i=1} d_i)

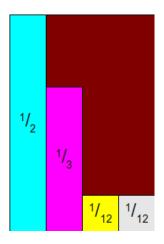
And now for something completely different ...

From 1 dimension to 2 dimensions

- p = [1/2, 1/3, 1/12, 1/12]
- say width = w and height p_i

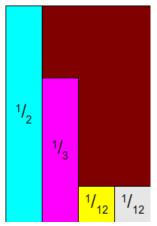


Darts



- 1. Throw dart
- 2. Return *i* if dart hits *i*th rectangle, else go to step 1

Darts



 $Pr(\text{hit rectangle } i|\text{hit some rectangle}) = \frac{\text{area of rectangle } i}{\text{total area of valid rectangle}} \\ = \frac{wp_i}{w\sum_j p_j} = p_i$

width w and height h don't matter

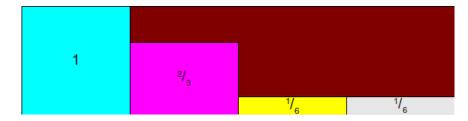
Say height is $h \cdot p_i$

 $Pr(\text{hit rectangle } i|\text{hit some rectangle}) = \frac{\text{area of rectangle } i}{\text{total area of valid rectangle}} \\ = \frac{hwp_i}{hw} = p_i$

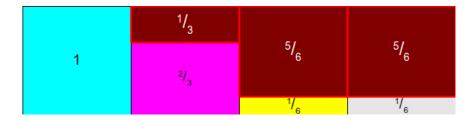
width *w* and height *h* don't matter Say height is $h \cdot p_i$

 $Pr(\text{hit rectangle } i|\text{hit some rectangle}) = \frac{\text{area of rectangle } i}{\text{total area of valid rectangle}}$ $= \frac{hwp_i}{hw} = p_i$

Set
$$h = \frac{1}{p_{\text{max}}}$$
 and $w = 1$ for convenience



A different way of simulating a loaded die



To generate a sample:

- 1. Choose *i* uniformly from one of *n* rectangles
- 2. Sample x uniformly from [0, 1)

3. If
$$x \leq \frac{p_i}{p_{\max}}$$
, return *i*, else go to step 1

Computational Complexity

- For each sample:
 - 1. Choose *i* uniformly from one of *n* rectangles
 - 2. Sample x uniformly from [0, 1)

3. If
$$x \le \frac{p_i}{p_{\max}}$$
, return *i*, else go to step 1

Pr(some side is chosen)

$$=\sum_{i=0}^{n-1} \left(\frac{1}{n} \frac{p_i}{p_{max}}\right) = \frac{1}{n} \sum_{i=0}^{n-1} \frac{p_i}{p_{max}} = \frac{1}{n \cdot p_{max}} \sum_{i=0}^{n-1} p_i = \frac{1}{n \cdot p_{max}}$$

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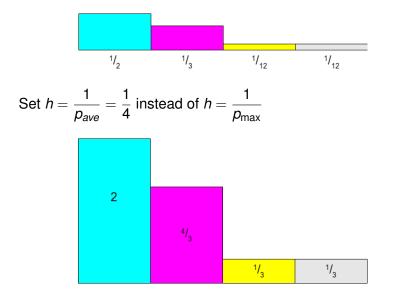
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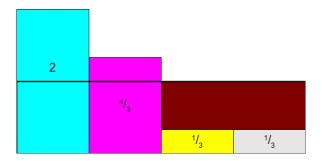
- Expected number of flips = $n \cdot p_{max}$
- Best case: O(1) for $p_{max} = 1/n$
- Worst case: O(n) for $p_{max} = 1$

How do we improve on the worst case performance?

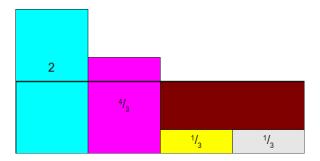




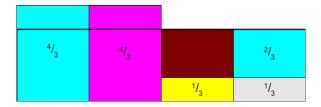
Draw a horizontal line at height 1 and mark invalid regions in red



Draw a horizontal line at height 1 and mark invalid regions in red



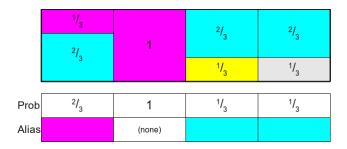
Key idea: Eliminate the wasteful red region such that each rectangle contains at most 2 valid colors.







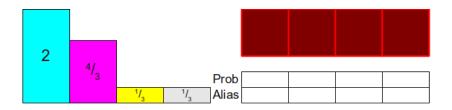
Alias method and corresponding alias table



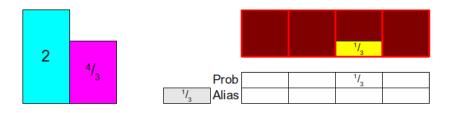
- Prob table contains height of *i* (probability of coin)
- · Alias contains id of alternative color

Constructing alias tables

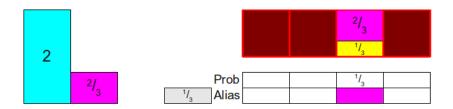
- An alias table can be constructed for any p
- *O*(1) for sampling cost once alias table has been constructed
- Pre-processing cost for alias table construction:
 - Naive alias method: $O(n^2)$
 - Alias method: $O(n \log n)$
 - Vose's alias method: O(n)

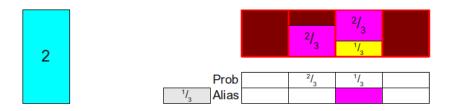


- Find some rectangle that has height at most 1 and place it into its own column, setting the Prob table to the height of that rectangle.
- Find some rectangle that has height at least 1 and use it to top off the column, setting the Alias table to correspond to the side of the die represented by the rectangle.

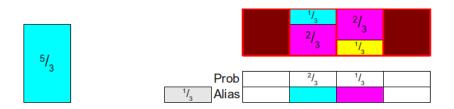


Choose column 2 as alias





Choose column 1 as alias



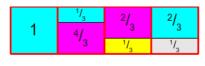




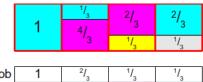
Choose column 1 as alias

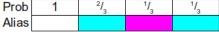






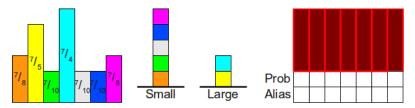
Prob	1	² / ₃	1/ ₃	1/ ₃
Alias				





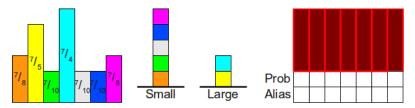
- $O(n^2)$ using unsorted arrays
- $O(n \log n)$ using binary search tree
- O(n) using Vose's method

Vose's method - 1 Consider $p = (\frac{1}{8}, \frac{1}{5}, \frac{1}{10}, \frac{1}{4}, \frac{1}{10}, \frac{1}{10}, \frac{1}{8})$

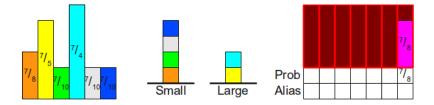


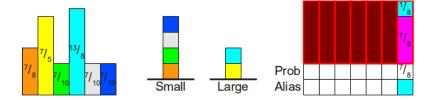
- Maintain two (unordered) stacks for small (height \leq 1) and large (height > 1)

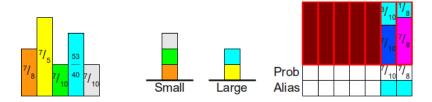
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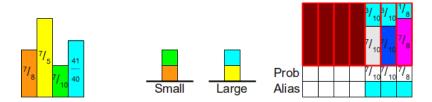


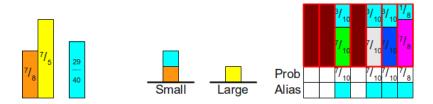
- Maintain two (unordered) stacks for small (height \leq 1) and large (height > 1)
- Pseudocode:
 - 1. Pop top of small stack, say *s*, and fill in the corresponding prob column
 - 2. Pop top of large stack, say ℓ , to fill in the remaining 1 s
 - 3. If $\ell (1 s) \le 1$, move $\ell (1 s)$ to top of the small stack
- Use of stack allows *O*(*n*) construction

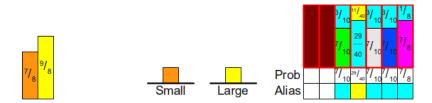


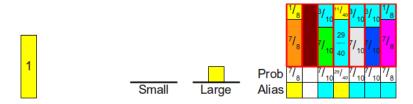












Comparison of different methods

Algorithm	Initialization Time Best Worst	Generation Time Best Worst	Memory Usage Best Worst
Loaded Die from Fair Die	$\Theta(n) \qquad O(\prod_{i=0}^n d_i)$	$\Theta(1)$	$\Theta(n) \qquad O(\prod_{i=0}^n d_i)$
Loaded Die from Biased Coins	$\Theta(n)$	$\Theta(1)$ $\Theta(n)$	$\Theta(n)$
Roulette Wheel Selection	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$
Optimal Roulette Wheel Selection	$O(n^2)$	$\Theta(1)$ $O(\log n)$	$\Theta(n)$
Fair Die/Biased Coin Loaded Die	$\Theta(n)$	$\Theta(1)$ $\Theta(n)$ (expected)	$\Theta(n)$
Naive Alias Method	$O(n^2)$	$\Theta(1)$	$\Theta(n)$
Alias Method	$O(n \log n)$	$\Theta(1)$	$\Theta(n)$
Vose's Alias Method	$\Theta(n)$	$\Theta(1)$	$\Theta(n)$

Thank you!