

Monte Carlo Simulations

September 10, 2026

Section 2.3: Monte Carlo Simulation

- With *Empirical Probability*, we perform an experiment many times n and count the number of occurrences n_a of an event $\mathcal{A}$
 - The *relative frequency* of occurrence of event $\mathcal{A}$ is n_a/n
 - The *frequency theory of probability* asserts that the relative frequency converges as $n \rightarrow \infty$

$$\Pr(\mathcal{A}) = \lim_{n \rightarrow \infty} \frac{n_a}{n}$$

- *Axiomatic Probability* is a formal, set-theoretic approach
 - Mathematically construct the sample space and calculate the number of events $\mathcal{A}$
- The two are complementary!

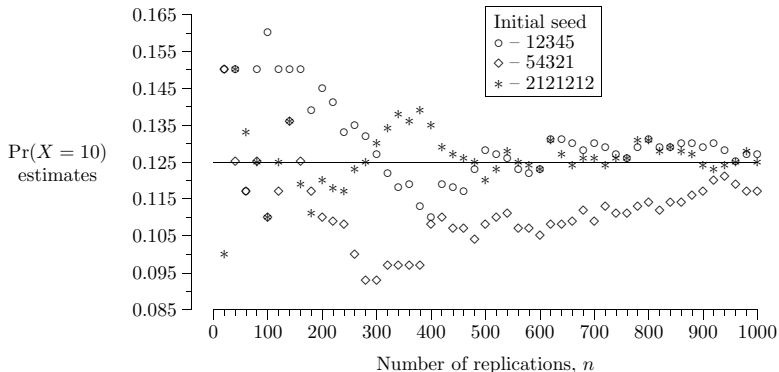
- If three fair dice are rolled, which sum is more likely, a 9 or a 10?
 - There are $6^3 = 216$ possible outcomes

$$\Pr(X = 9) = \frac{25}{216} \cong 0.116 \quad \text{and} \quad \Pr(X = 10) = \frac{27}{216} = 0.125$$

- Program `galileo` calculates the probability of each possible sum between 3 and 18
- The drawback of Monte Carlo simulation is that it only produces an estimate
 - Larger n does not guarantee a more accurate estimate

Example 2.3.6

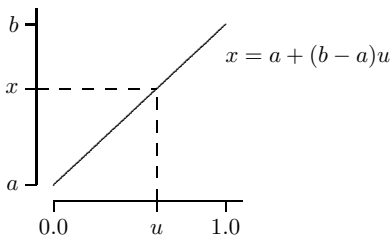
- Frequency probability estimates converge slowly and somewhat erratically



- You should *always* run a Monte Carlo simulation with multiple initial seeds

Random Variates

- A *Random Variate* is an algorithmically generated realization of a random variable
- $u = \text{Random}()$ generates a *Uniform*(0, 1) random variate
- How can we generate a *Uniform*(a, b) variate?



Generating a Uniform Random Variate

```
double Uniform(double a, double b)    /* use  $a < b$  */ {  
    return (a + (b - a) * Random());  
}
```

Equilikely Random Variates

- *Uniform*(0, 1) random variates can also be used to generate an *Equilikely*(*a*, *b*) random variate

$$\begin{aligned}0 < u < 1 &\iff 0 < (b - a + 1)u < b - a + 1 \\&\iff 0 \leq \lfloor (b - a + 1)u \rfloor \leq b - a \\&\iff a \leq a + \lfloor (b - a + 1)u \rfloor \leq b \\&\iff a \leq x \leq b\end{aligned}$$

- Specifically, $x = a + \lfloor (b - a + 1) u \rfloor$

Generating an Equilikely Random Variate

```
long Equilikely(long a, long b)    /* use a < b */ {  
    return (a + (long) ((b - a + 1) * Random()));  
}
```

- **Example 2.3.3** To generate a random variate x that simulates rolling two fair dice and summing the resulting up faces, use

$$x = \text{Equilikely}(1, 6) + \text{Equilikely}(1, 6);$$

Note that this is *not* equivalent to

$$x = \text{Equilikely}(2, 12);$$

- **Example 2.3.4** To select an element x at random from the array $a[0], a[1], \dots, a[n-1]$ use

$$i = \text{Equilikely}(0, n - 1);$$
$$x = a[i];$$

Equilikely(a, b) Geometry and *Uniform*($a, b + 1$) Composition

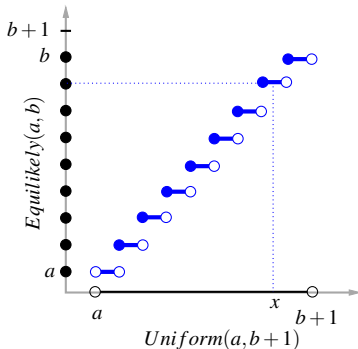
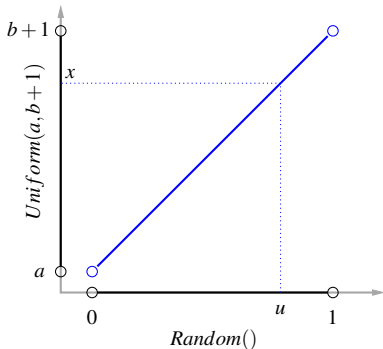
Our definition of

$$\text{Equilikely}(a, b) \rightarrow a + \lfloor ((b - a + 1) \cdot \text{Random}()) \rfloor$$

could also be written using *Uniform*($\cdot$):

$$\text{Equilikely}(a, b) \rightarrow \lfloor \text{Uniform}(a, b + 1) \rfloor$$

which hints at the geometry happening with the pRNG values...



Whew! You're ready!

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