# Class 08

Random Number Generation & An Intro to Monte Carlo Methods

## Outline

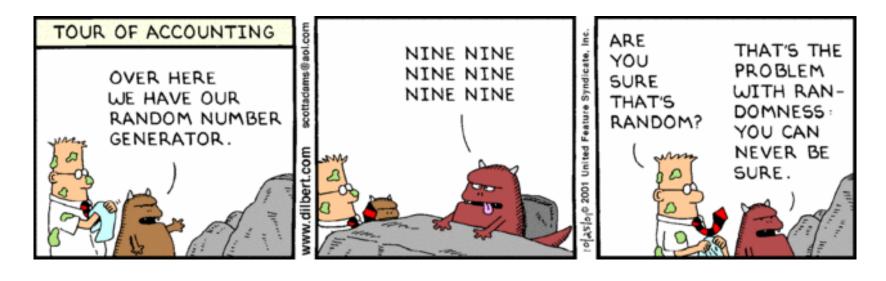
- Random Number Generators
- Engines, Distributions, and Seeds
- Intro to Monte Carlo Methods

### Random Number Generation

- A random number generator (RNG) is one that produces a stream of numbers in an undeterminable pattern.
- Many applications require the use of a random number.
  - Statistics & analysis
  - Simulations
  - Cryptography
  - Games
- Random Number Generation (RNG) can be based on hardware or on software.

### Random Number Generation

The quality of a random number generator that we are concerned about is its period, or how long it goes before repeating a number.



## RNG - Reproducibility & Seeds

- When using random numbers we need to consider the reproducibility of those random numbers.
- We want systems to be reproducible to help facilitate debugging and for improving quality control.
  - Imagine your system crashes only *some* of the time. If you have unpredictable random inputs, tracing the root cause becomes immensely difficult!
- Engines use *seeds* to control the pattern that it creates. Use the same seed and the same exact pattern emerges from the engine. Not so random...

### RNG - Deterministic vs. Indeterminism

- Engines are initialized with a seed; this seed may be set by the user or set with some form of entropy.
- Allowing the user to set the seed allows the system to be run with that same seed again to produce the same results. This is called *deterministic RNG*.
- Entropy is some characteristic about the (computing) environment at that moment in time. E.g. available RAM, system clock, etc. or some combination that is (nearly) impossible to replicate. Using entropy is called *indeterministic RNG*.

## RNG - Engines & Distributions

- We can produce random and pseudo random numbers for our programs using a combination of generators (engines) and distributions.
  - ▶ All are provided by the C++ header *random*.
- Engines are tasked with producing a random number.
- Distributions are tasked with taking the number generated by the engine and transforming it according to some set of rules.
- C++ provides a handful of engines and many different distributions.
  - Common engine: std::mt19937
  - Common distributions: std::uniform\_real\_distribution, std::uniform\_int\_distribution, std::normal distribution
  - These are all data types!

## RNG - Engines & Distributions

- The engines and distributions in C++ are called functors.
  - They are variables that look and act like functions.
- Engines are called like functions without parameters:

```
auto r = eng();
```

Distributions are called like functions, taking in an engine as a parameter:

```
auto r = dis(eng);
```

## **Engines - Random Device**

- A random device is an indeterministic engine that usually uses hardware entropy as its seed.
  - Sensors on chip. e.g. temperature, sound, etc.
  - User inputs. e.g. mouse movement, key presses, etc.
  - ► CPU cycling/throttling, memory available, etc.
- Not consistent across platforms, and heavily dependent on the type of hardware entropy in use.
  - Poor entropy availability will result in poor randomness.
  - Not all systems can provide hardware entropy.
- Typically used to seed another engine.

## **Engines - Random Device**

Example

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto rd = std::random_device{};
    for (int i = 0; i < 10; ++i)
    {
        fmt::print("{}\n", rd()); // produce a random number and print it
    }
}</pre>
```

- This method generates a random number stream that has a period length equal to a Mersenne Prime.
  - Is considered a pseudorandom number generator
- $\triangleright$  A very common implementation uses the *Mersenne Prime*  $2^{19937}$  1 as its period.
  - ▶ The period of a random number generator is the number of samples that occur before a number is repeated.
- While this method is not cryptographically secure, it is perfectly well suited for many scientific methods.
  - Incredibly long period is useful
  - It is memory intensive compared to other engines, which may be more problematic on small devices with limited memory.

396296801920114486480902661414318443276980300066728104984095451588176077132969843762134621790396391341285205627619600513106646376648615994236675486537480241964350 1221191641747710902480633491091704827441228281186632445907145787138351234842261380074621914004818152386666043133344875067903582838283562688083236575482068479639 5454903864111585429504569920905636741539030968041471

Here we are creating a *std*::*mt19937* engine. Whenever we want to generate a new random number, we *call* the engine like it is a function.

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto engine = std::mt19937{};
    fmt::print("{}\n", engine());
}
```

► This sample below uses a new seed to change the pattern produced by the engine. Here our seed is 111.

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto engine = std::mt19937{111};
    fmt::print("{}\n", engine());
}
```

You will commonly see the seed 1337 in code for this course. It has absolutely zero scientifically significant meaning. It is just a good number.

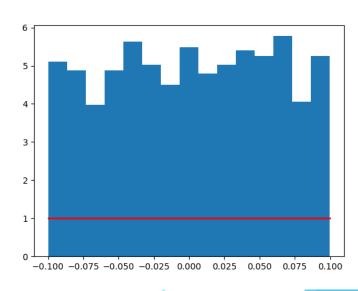
This sample below uses a special objected *random\_device* to generate a random seed. This allows our program to generate something new every time we run it.

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto rd = std::random_device{};
    auto engine = std::mt19937{rd()};
    fmt::print("{}\n", engine());
}
```

### **Distributions - Uniform**

- Uniform Distributions provide an equal probability of sampling to every value within the specified range.
  - ▶ i.e. all possibilities are equal! If there are 10 possible outcomes, each of those outcomes would have a 10% chance of occurring (100 / 10 = 10)
  - Useful for representing a random event that shows no bias
- The diagram here shows binned counts of 1000 random samples taken uniformly between -0.1 and 0.1.
- The red line represents the *probability density* function, or the likelihood of selecting a random sample at that value. Here it is a flat line, and so all samples have an equal chance of being selected!



### **Distributions - Uniform**

- C++ distinguishes between uniform distributions for integers and floatingpoint numbers.
  - std::uniform\_int\_distribution is used only for integer types
  - std::uniform\_real\_distribution is used only for floating-point types
- C++ however has more than one integer type, and more than one floatingpoint type, and so we need to explicitly tell these distributions what data type they are going to manage (like std::vector!).

```
auto d1 = std::uniform_int_distribution<int>{0, 100};
auto d2 = std::uniform_real_distribution<double>{0.0, 100.0};
```

## **Uniform Integer Distribution**

#### Example

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto dis = std::uniform_int_distribution<>{1, 100};
    auto engine = std::mt19937{};
    fmt::print("{}\n", dis(engine));
}
```

### Uniform Real Distribution

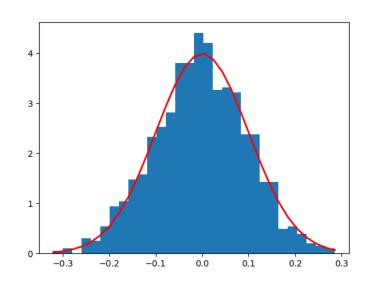
#### Example

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto dis = std::uniform_real_distribution<>{1.0, 100.0};
    auto engine = std::mt19937{};
    fmt::print("{}\n", dis(engine));
}
```

### Distributions - Normal

- Normal Distributions provide a biased probability of sampling values within some range of a center value (mean); values closer to the mean are more likely to occur.
  - More realistic for modeling real events by being able to capture likely and unlikely events.
- The diagram here shows binned counts of 1000 random samples taken from a mean of 0.0 and a standard deviation of ±0.1
- The red line represents the *probability density* function, or the likelihood of selecting a random sample at that value. Here the curve indicates that values closer to the mean are more likely!



### Distributions - Normal

- C++ only allows normal distributions to be used with floating-point numbers!
- Like std::uniform\_real\_distribution though we still need to explicitly tell it what sort of floating-point it will manage!

### Normal Distribution

#### Example

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto dis = std::normal_distribution<double>{50.0, 25.0};
    auto engine = std::mt19937{};
    fmt::print("{}\n", dis(engine));
}
```

### Improper Use of Random Numbers

What's wrong here?

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto dis = std::uniform_real_distribution<int>{0, 10};
    auto engine = std::mt19937{};
    fmt::print("{}\n", dis(engine));
}
```

## Improper Use of Random Numbers

```
#include <random>
#include <fmt/format.h>

auto main() -> int
{
    auto dis = std::uniform_real_distribution<int>{0, 10};
    auto engine = std::mt19937{};
    fmt::print("{}\n", dis(engine));
}
```

Uniform real distributions only work with floating point types!

### RNG - Common Tasks

- generating random points in 2D or 3D space
  - generate 2 or 3 numbers, and together they represent a point!
- adding noise to data
  - sometimes we want to fuzz data to make it appear more realistic, like when simulating signal processing
- strategy patterns
  - predictive analytics understand the depth of your system and where it can go
  - Al design use randomness to represent the choices made by Al

## Baking a Cake

- A recipe for a cake may specify the following:
  - Ingredients:
    - ▶ 3 cups flour
    - ▶ 1 tbsp baking soda
    - ▶ 1 tsp salt
    - ▶ 1 cup butter
    - ▶ 1 cup milk
    - > 5 eggs
  - Directions
    - mix
    - ▶ bake for 30 min at 350F

## Baking a Cake

- A recipe for a cake may specify the following (not this simply...):
  - Ingredients:
    - ▶ 3 cups flour
    - ▶ 1 tbsp baking soda
    - ▶ 1 tsp salt
    - ▶ 1 cup butter
    - ▶ 1 cup milk
    - 5 eggs

How perfectly are you measuring each of these quantities?

What if you miscounted 4 eggs?

- Directions
  - mix
  - ▶ bake for 30 min at 350F

How well did you mix?
Was the oven really at 350F?
Did it bake for exactly 30 minutes?
How is the heat distributed in the oven?

### Monte Carlo Methods

- Monte Carlo Methods are a class of techniques that utilize random number generation to aid in representing uncertainty.
- In theory algorithms and measurements are precise and without error. The simulation is a perfect world.
- In practice apparatus are imperfect, human elements are unpredictable, and details are unaccounted for.
- Monte Carlo Methods allow an algorithm to run with randomly imperfect data so that one execution to the next is different, which over time will paint a better picture.

### Monte Carlo Methods

- Various aspects of the system are represented with varying distributions that align reasonably well with their realistic behavior.
  - This is usually dictated by real-world measurements and/or subject matter experts that understand the system exceptionally well.
- We run a Monte Carlo method many times over until the resulting data converges.
  - This can take tens, hundreds, and even thousands of executions to observe a convergence in the data.
- When checking for a numerical convergence, we utilize some error threshold (epsilon).
  - This will typically be  $ε = 1e^{-8}$  (this is a really small number!)
  - If our data is off from the expect result by a difference less than  $\epsilon$  then we say our result is close enough.

## List of Engines & Distributions

https://en.cppreference.com/w/cpp/header/random