L27: Random Numbers

Or rather: not-so-random numbers

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Version: release

Lab 10 is due on April 7 at 12 pm (noon)

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► Pseudo-random numbers

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- ▶ Get some rest, have some fun, see friends and family
- ▶ Get a head start on lab 10 and on your E7 project

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After Spring break:

- Numerical differentiation (Chapter 17)
- Numerical integration (Chapter 18)

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- Numerical integration (Chapter 18)

Programming project:

- ▶ Instructions and support code will be released during Spring break
- ► See slides of lecture L26 (March 22nd) to get started on your project now
- "Random" components will be removed from the project

Pseudo-random numbers: introduction

Matlab can generate pseudo-random numbers. The numbers thus generated are not actually random, but are generated using an algorithm that produces predictable numbers with random-like properties

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```
>> % Create a 3 by 4 array of pseudo-random
>> % integers ranging from 1 to 100
>> randi([1, 100], [3, 4])
ans =
   82 92 28
                   97
   91
        64 55 16
   13
         10 96
                   98
>> % Try it again, and get different integers
>> randi([1, 100], [3, 4])
ans =
   96
         15
              80
   49
         43 96
                85
   81
         92
              66
                   94
```

Pseudo-random numbers: the seed

The starting point used by the pseudo-random number generator when generating random-like numbers is controlled by the seed Use rng(seed) to set the seed (seed: 0 or positive integer)

Pseudo-random numbers: the seed

The starting point used by the pseudo-random number generator when generating random-like numbers is controlled by the seed Use rng(seed) to set the seed (seed: 0 or positive integer)

```
>> % Use seed = 0
\gg rng(0)
>> randi([1, 100], [2, 5])
ans =
      13 64
   82
                   28
                        96
   91
     92 10
                   55
                        97
\gg % Use seed = 1
\gg rng(1)
>> randi([1, 100], [2, 5])
ans =
      1 15 19
   42
                       40
   73 31 10
                  35
                        54
>> % Use seed = 0 again
\gg rng(0)
>> randi([1, 100], [2, 5])
ans =
   82
      13 64
                   28
                         96
   91
         92 10
                   55
                         97
```

Pseudo-random numbers: the seed

Another example:

```
>> rnq(25)
>> randi([1, 1000000], [1, 1])
ans =
      870125
>> randi([1, 1000000], [1, 1])
ans =
      582277
\rightarrow rng(25)
>> randi([1, 1000000], [1, 1])
ans =
      870125
>> randi([1, 1000000], [1, 1])
ans =
      582277
```

Pseudo-random numbers: practice question

```
>> rnq(10)
>> randi([1, 20], [3, 7])
ans =
   16 15
                          13
            4
                               19
   1 10 16 14 11
                          15
                               15
   13
           4
                 20
                      17
                          6
                               11
```

What will the value of variable "v" be after executing the following code?

```
>> rng(10)
>> v = randi([1, 20], [1, 7]);
```

- (A) [16, 15, 4, 2, 1, 13, 19]
- (B) [16, 1, 13, 15, 10, 5, 4]
- (C) It is impossible to know

Pseudo-random numbers: practice question

```
>> rng(10)
>> randi([1, 20], [3, 7])
ans =
    16    15    4    2    1    13    19
    1    10    16    14    11    15    15
    13    5    4    20    17    6    11
```

What will the value of variable "v" be after executing the following code?

```
>> rng(10)
>> v = randi([1, 20], [1, 7]);
```

- (A) [16, 15, 4, 2, 1, 13, 19]
- (B) [16, 1, 13, 15, 10, 5, 4]
 - (C) It is impossible to know

Matlab creates arrays of pseudo-random numbers column-by-column

Matlab built-in functions for pseudo-random numbers

- ▶ randi([p, q], [m, n]): Generate a m × n array of pseudo-random integers between p and q* Numbers are drawn from the corresponding uniform distribution
- rand([m, n]): Generate a m × n array of pseudo-random floating-point numbers between 0 and 1** Numbers are drawn from the corresponding uniform distribution
- randn([m, n]): Generate a m × n array of pseudo-random floating-point numbers Numbers are drawn from the standard normal distribution
- rng(seed): Set the seed of the random number generator to seed (e.g., zero; a positive integer; 'shuffle' to set the seed to a "random" value)

^{*} both boundaries included

^{**} both boundaries excluded randi, rand, and randn all return variables of class double

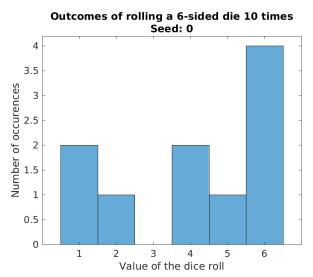
Uniform distribution: when generating pseudo-random numbers, all numbers have the same probability of being generated

Examples of application:

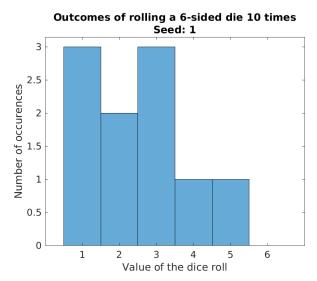
- Simulate flipping a well-balanced coin
- Simulate rolling a well-balanced die

For example: see the function my_dice_roller , which simulates n rolls of a p-sided die, and creates the histogram of outcomes

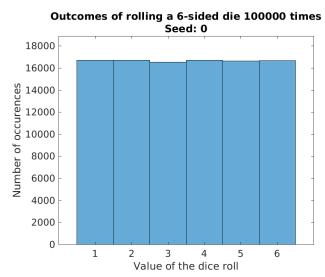
For a small number of dice rolls, the distribution of outcomes might not appear to be uniform



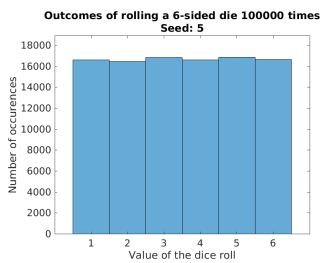
For a small number of dice rolls, the distribution of outcomes might not appear to be uniform



For a large number of dice rolls, the histogram of outcomes illustrates the uniformity of the distribution

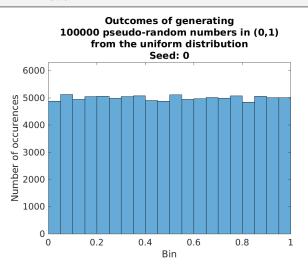


For a large number of dice rolls, the histogram of outcomes illustrates the uniformity of the distribution



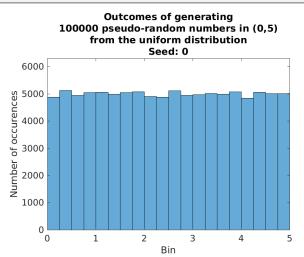
Generate 100000 pseudo-random floating-point numbers between 0 and 1 drawn from the uniform distribution:

```
rand([1, 100000]);
```



Generate 100000 pseudo-random floating-point numbers between 0 and 5 drawn from the uniform distribution:

```
rand([1, 100000]) * 5;
```



Normal distribution

Standard normal distribution: when generating a pseudo-random number using randn, the probability φ of a particular double x to be generated is proportional to:

$$\varphi(x) = \frac{1}{\sqrt{2\pi}} e^{-(x^2)/2} \tag{1}$$

The mean μ of this distribution is 0 and its standard deviation σ is 1

The **standard** normal distribution φ is a special case of the generic normal distribution f:

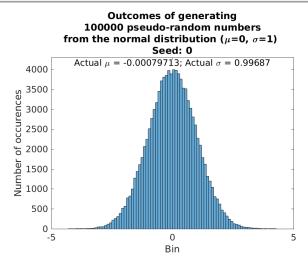
$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/(2\sigma^2)}$$
 (2)

Note: the number of different numbers that Matlab can generate is finite. The concept of probability distributions is slightly different for continuous variables

Normal distribution

Generate 100000 pseudo-random numbers drawn from the standard normal distribution:

```
randn([1, 100000]);
```



Normal distribution

Generate 100000 pseudo-random numbers drawn from the normal distribution with mean $\mu=$ 10 and standard deviation $\sigma=$ 5:

