

# Uniform distribution

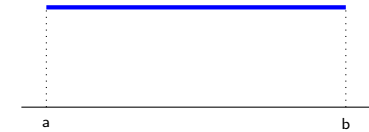
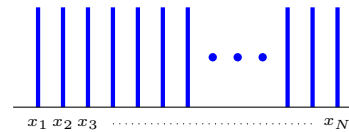
Statistics for Business

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# Uniform distribution

The **uniform distribution** is the probability distribution that gives the same probability to all possible values. There are two types of uniform distributions:



Discrete uniform distribution

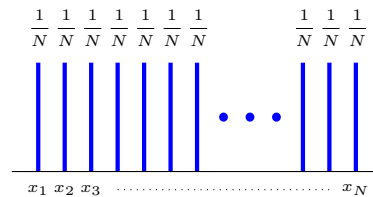
Continuous uniform distribution

We apply uniform distributions when there is **absolute uncertainty** about what will occur. We also use them **when we take a random sample from a population**, because in such cases all the elements have the same probability of being drawn. Finally, they are also used **to create random numbers**, as random numbers are those that have the same probability of being drawn.

# Discrete uniform distribution

## Probability function

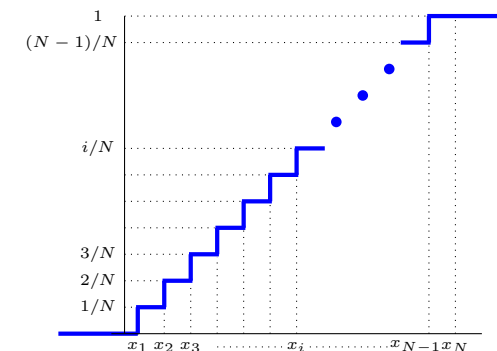
$$P[X = x] = \frac{1}{N}; x = x_1, x_2, \dots, x_N$$



# Discrete uniform distribution

## Distribution function

$$F(x) = P[X \leq x_i] = \frac{i}{N}; x_i = x_1, x_2, \dots, x_N$$



## Discrete uniform distribution

### Notation, mean and variance

$$X \sim U(x_1, x_2, \dots, x_N) \begin{cases} \mu = \frac{x_1 + x_N}{2} \\ \sigma^2 = \frac{(x_N - x_1 + 2)(x_N - x_1)}{12} \\ = \frac{(x_N - x_1 + 1)^2 - 1}{12} \end{cases}$$

## Discrete uniform distribution

### Distribution of the maximum

We draw a value from a uniform distribution  $n$  times. How is distributed the maximum among those  $n$  values?

- Among  $n$  values, **the maximum will be less than  $x_i$**  when all of them are less than  $x_i$ :

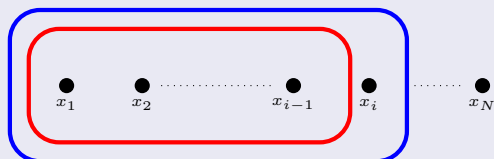
$$P[X_{max} \leq x_i] = \frac{i}{N} \times \frac{i}{N} \times \dots \times \frac{i}{N} = \left(\frac{i}{N}\right)^n$$

- Among  $n$  values, **the maximum will be less than  $x_{i-1}$**  when all of them are less than  $x_{i-1}$ :

$$P[X_{max} \leq x_{i-1}] = \left(\frac{i-1}{N}\right)^n$$

## Discrete uniform distribution

### Distribution of the maximum



Thus, among  $n$  values the probability of **the maximum being  $x_i$**  is:

$$P[X_{max} = x_i] = \left(\frac{i}{N}\right)^n - \left(\frac{i-1}{N}\right)^n$$

$$\text{Blue oval} - \text{Red oval}$$

## Discrete uniform distribution

### Distribution of the maximum: example

We throw a dice 4 times. We assume (logically) that the number of points follow a discrete uniform distribution. Calculate the probability of the maximum being 1, 2, 3, 4, 5 or 6:

- $P[X_{max} = 6] = \left(\frac{6}{6}\right)^4 - \left(\frac{5}{6}\right)^4 = 0.5177$
- $P[X_{max} = 5] = \left(\frac{5}{6}\right)^4 - \left(\frac{4}{6}\right)^4 = 0.2847$
- $P[X_{max} = 4] = \left(\frac{4}{6}\right)^4 - \left(\frac{3}{6}\right)^4 = 0.1350$
- $P[X_{max} = 3] = \left(\frac{3}{6}\right)^4 - \left(\frac{2}{6}\right)^4 = 0.0501$
- $P[X_{max} = 2] = \left(\frac{2}{6}\right)^4 - \left(\frac{1}{6}\right)^4 = 0.0115$
- $P[X_{max} = 1] = \left(\frac{1}{6}\right)^4 - \left(\frac{0}{6}\right)^4 = 0.0007$
- Probability decreases (logically)**

## Discrete uniform distribution

### Distribution of the minimum

Likewise,  
among  $n$  values the probability of the smallest being  $x_i$  is:

$$\begin{aligned} P[X_{min} = x_i] &= P[X_{min} \geq x_i] - P[X_{min} \geq x_{i+1}] \\ &= \left(\frac{N - (i - 1)}{N}\right)^n - \left(\frac{N - i}{N}\right)^n \end{aligned}$$

## Discrete uniform distribution

### Applications: sampling in finite populations

- When we draw a random sample from a finite population, all the elements of the population have the same probability. Thus, the model for the sampling should be the uniform discrete distribution.
- When random sampling is made with devolution, the probability of drawing a given sample of size  $n$  is:

$$P[x_1, x_2, \dots, x_n] = \frac{1}{N} \times \frac{1}{N} \times \dots \times \frac{1}{N} = \left(\frac{1}{N}\right)^n$$

- When random sampling is made without devolution, the probability of drawing a given sample of size  $n$  is given by the hypergeometric distribution:

$$P[x_1, x_2, \dots, x_n] = \frac{1}{\binom{N}{n}} = \frac{1}{N} \times \frac{1}{N-1} \times \dots \times \frac{1}{N-(n-1)} \times n!$$

## Discrete uniform distribution

### Application: German tank problem

- It's a classical probability problem, posed in World War II
- German tanks have a number from 1 to an unknown  $N$  number.
- Allies wanted to know that number. For that purpose, they collect the numbers of destroyed tanks.
- The model for numbers is the discrete uniform distribution:  $1, 2, \dots, N$  numbered tanks have all the same probability of being destroyed.
- Allies collect  $n$  random values from that distribution, they calculate how distributes the maximum of them and calculate the expected value, without devolution, and missing the  $N$  value:

$$E[X_{max}] = \frac{n(N+1)}{n+1}$$

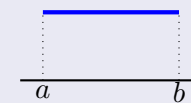
- They take the  $x_{max}$  maximum and equal to the expected value, in order to estimate the unknown  $N$  value

$$x_{max} = \frac{n(N+1)}{n+1} \rightarrow \hat{N} = x_{max} + \frac{x_{max} - n}{n}$$

- E.g., destroyed tank numbers: 82,123,345,614.
- The estimate for the number of tanks is:  $\hat{N} = 614 + \frac{614 - 4}{4} = 766.5$ .

## Continuous uniform distribution

### Density function



$$f(x) = \frac{1}{b-a}; a < x < b$$

$$F(x) = P[X < x] = \frac{x-a}{b-a}; a \leq x \leq b$$

## Continuous uniform distribution

### Notation, mean and variance

$$X \sim U(a, b)$$

$$X \sim U(a, b) \begin{cases} \mu = \frac{a+b}{2} \\ \sigma^2 = \frac{(b-a)^2}{12} \end{cases}$$

- The mean value is rather intuitive: as the possible values in the support have all the same probability, the mean value will be in the middle.
- The wider is the interval of the support, the bigger is the dispersion, and so the variance.

## Continuous uniform distribution

### Distribution of the maximum

Taken  $n$  values from a  $U(a, b)$  distribution, which is the distribution of the maximum  $M$ ?

- Distribution function:

$$F(M = x) = P[M < x] = \left(\frac{x-a}{b-a}\right)^n ; a \leq x \leq b$$

- Mean value:  $E[M] = a + \frac{n(b-a)}{n+1}$

- E.g., if we want to estimate the maximum of a  $U(0, 10)$  distribution (that is, assuming we don't know the maximum is 10) with 4 values, we should expect that with the maximum of those 4 values we would reach on average  $0 + \frac{4 \times 10}{5} = 8$ , that is, 80% of the true value. With 9 data we would reach 90%.

## Continuous uniform distribution

### Distribution of the minimum

Taken  $n$  values from a  $U(a, b)$  distribution, which is the distribution of the minimum  $m$ ?

- Distribution function:

$$F(m = x) = P[m < x] = 1 - \left(\frac{b-x}{b-a}\right)^n ; a \leq x \leq b$$

- Mean value:  $E[m] = a + \frac{b-a}{n+1}$
- E.g., if we want to estimate the minimum of a  $U(0, 10)$  distribution (that is, assuming we don't know the maximum is 0) with 4 values, we should expect that with the minimum of those 4 values we would reach on average  $0 + \frac{10}{5} = 2$ . With 10 data we would get 1 on average.

## Continuous uniform distribution

### Distribution of the range

Taken  $n$  values from a  $U(a, b)$  distribution, which is the distribution of the range  $R$  (maximum - minimum)?

- Distribution function (for  $a \leq x \leq b$ ):

$$F(R = x) = P[R < x] = n\left(\frac{x}{b-a}\right)^{n-1} \left(\frac{b-a-x}{b-a}\right) + \left(\frac{x}{b-a}\right)^n$$

- $E[R] = (b-a) \frac{n-1}{n+1}$

- E.g., if we want to estimate the range of a  $U(0, 10)$  distribution (that is, assuming we don't know the maximum is 10-0=10) with 4 values, we should expect that with the maximum of those 4 values we would reach on average  $0 + 10 \times \frac{3}{5} = 6$ , 60% from the true value. With 10 data we would get  $9/11=8.1$  on average.

## Continuous uniform distribution

### Standard uniform distribution

$$X \sim U(0,1)$$

Random numbers from 0 to 1 come from this distribution. We can create (better, simulate) them, by typing SHIFT+RAN# in the calculator.

### Stochastic simulation for uniform distributions

Stochastic simulation is artificially creating data, following a given distribution.

Simulating a continuous uniform distribution is very easy compared to other distributions: random numbers follow the  $U(0,1)$  distribution, and to simulate  $U(a,b)$  we just have to make this linear transform:  $U(a,b) = a + (b-a)U(0,1)$

So, naming *sim* the simulated data:  $sim_{U(a,b)} = a + (b-a)sim_{U(0,1)}$