

# Distribution

**Question 1:** Simulate 30 rolls with `=RANDBETWEEN(1,6)`. What is the probability of rolling a 3 exactly 5 times? (Hint: Use `BINOM.DIST`)

**Answer:** Each roll of a fair die has probability

$$p = P(\text{rolling a 3}) = 1/6$$

We want the probability of getting exactly 5 threes in 30 rolls, which follows a binomial distribution:

$$P(X=5) = \binom{30}{5} (1/6)^5 (5/6)^{25}$$

Using Excel (as hinted):

`=BINOM.DIST(5, 30, 1/6, FALSE)`

`=0.192`

**Result:**

$$P(X=5) \approx 0.19$$

**Probability  $\approx$  0.19 (or 19%)**

So, there is about a **19% chance** that a 3 appears **exactly 5 times in 30 rolls**.

**Question 2:** Generate 100 values in Excel using the continuous uniform distribution `RAND()` and plot a histogram. Describe the shape of the distribution.

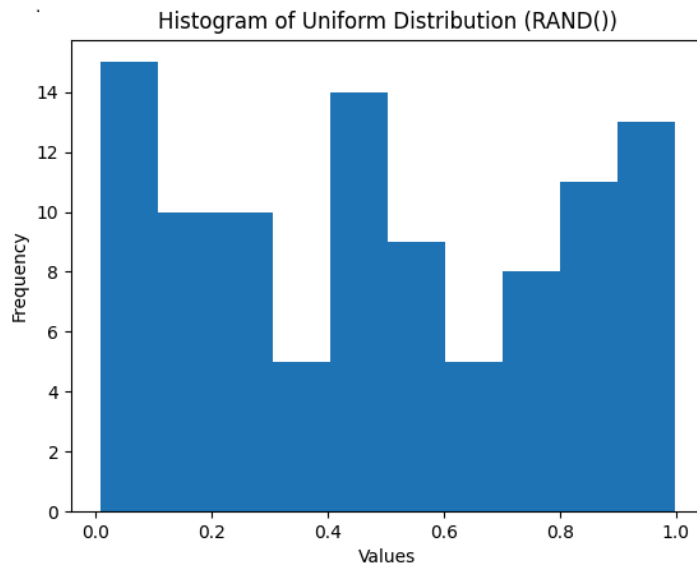
**Answer: Step 1: Generate 100 Uniform(0,1) values**

1. Open Excel.
2. In cell **A1**, type:  
`=RAND ()`
3. Press **Enter**.
4. Click on cell A1 and drag the fill handle down to **A100**.

This generates 100 random values from a **continuous uniform distribution on [0, 1]**.

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## Step 2: Create a histogram



5.

## Step 3: Describe the shape of the distribution

- The histogram should appear **approximately flat (rectangular)**.
- The bars should have **roughly equal heights** across the interval from 0 to 1.
- Small irregularities are normal due to **random sampling** and the relatively small sample size (100 values).

**Question 3: A dataset has a mean of 50 and a standard deviation of 5. What percentage of values lie between 45 and 55 if the data follows a normal distribution?**

**Answer:** Since the data follows a **normal distribution**:

- Mean ( $\mu$ ) = 50
- Standard deviation ( $\sigma$ ) = 5
- Range 45 to 55 =  $\mu \pm 1\sigma$

According to the **Empirical Rule (68–95–99.7 rule)**:

- About **68%** of the values lie within **1 standard deviation** of the mean.

**Answer: Approximately 68% of the values lie between 45 and 55.**

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**Question 4: What is the concept of standardization (z-score), and why is it important in data analysis? Explain the formula and how standardization transforms a dataset.**

**Answer: Concept of Standardization (Z-score)**

Standardization is a statistical technique used to convert data values into a common scale called z-scores. A z-score tells us how many standard deviations a data point is away from the mean.

## **Z-score Formula**

$$z = (X - \mu) / \sigma$$

**Where:**

- ( X ) = original data value
  - (  $\mu$  ) = mean of the dataset
  - (  $\sigma$  ) = standard deviation of the dataset
  - ( z ) = standardized value (z-score)
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## **How Standardization Transforms a Dataset**

After standardization:

- The **mean becomes 0**
- The **standard deviation becomes 1**
- Original values are converted into **unit-free** z-scores

**Example:** If a value has a z-score of +2, it means the value is 2 standard deviations above the mean.

## **Why Standardization Is Important in Data Analysis**

### **1. Comparison Across Different Scales**

Allows comparison of variables measured in different units (e.g., height vs salary).

### **2. Identifying Outliers**

Very high or low z-scores indicate potential outliers.

# Distribution

## 3. Improves Model Performance

Many machine learning algorithms (e.g., k-means, SVM, linear regression) perform better with standardized data.

## 4. Probability & Normal Distribution Analysis

Helps in finding probabilities using the **standard normal distribution table**.

## 5. Feature Scaling

Prevents variables with larger scales from dominating analysis.

## Question 5: What is Kurtosis and their type?

**Answer:** Kurtosis is a statistical measure that describes the shape of a data distribution, specifically how peaked or flat it is compared to a normal distribution. It also indicates the weight of the tails (presence of extreme values).

### Types of Kurtosis →

There are **three main types of kurtosis**:

Type	Shape	Tails	Kurtosis Value
Mesokurtic	Normal	Moderate	$\approx 3$
Leptokurtic	Sharp peak	Heavy	$> 3$
Platykurtic	Flat	Light	$< 3$

Question 6: Explain why the uniform distribution is a good model for the outcome of rolling a fair die.

Answer: The **uniform distribution** is a good model for the outcome of rolling a **fair die** because **all possible outcomes are equally likely**.

### Explanation

When you roll a fair six-sided die, the possible outcomes are:  
{1, 2, 3, 4, 5, 6}

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For a die to be *fair*, each number has the **same probability** of occurring:

$$P(1) = P(2) = P(3) = P(4) = P(5) = P(6) = 1/6$$

This exactly matches the definition of a **uniform distribution**, where:

- Every outcome in the sample space has **equal probability**
- No outcome is favored over another

## Key Reasons

### 1. Equal Likelihood

Each face of a fair die has the same chance of appearing.

### 2. Discrete and Finite Outcomes

The die has a fixed number of outcomes (6), which suits a **discrete uniform distribution**.

### 3. No Bias or Weighting

A fair die is designed so that shape, weight, and balance do not influence results.

## Conclusion

Because all six outcomes are equally probable and independent, the **discrete uniform distribution** is the most appropriate and accurate model for rolling a fair die.

**Question 7: Use Excel to compute the probability of getting at least 8 successes in 15 trials with success probability 0.5**

**Answer:** To compute this in **Excel**, we use the **binomial distribution**.

**Given:**

- Number of trials ( $n$ ) = **15**
- Probability of success ( $p$ ) = **0.5**
- We want **at least 8 successes**  $\rightarrow (P(X \geq 8))$

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## Excel Formula

In Excel, use:

$$=1 - \text{BINOM.DIST}(7, 15, 0.5, \text{TRUE})$$

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## Explanation:

- `BINOM.DIST(7, 15, 0.5, TRUE)` calculates  $P(X \leq 7)$
- Since  $P(X \geq 8) = 1 - P(X \leq 7)$  we subtract from 1.

## Result (Approximate)

The probability is:

$$P(X \geq 8) \approx 0.5$$

## Alternative Method (Direct Sum)

You could also calculate:

$$\begin{aligned} &= \text{BINOM.DIST}(8, 15, 0.5, \text{FALSE}) \\ &+ \text{BINOM.DIST}(9, 15, 0.5, \text{FALSE}) \\ &+ \dots \\ &+ \text{BINOM.DIST}(15, 15, 0.5, \text{FALSE}) \end{aligned}$$

But the **first method is faster and recommended.**

## Final Answer:

Using Excel, the probability of getting **at least 8 successes in 15 trials** with  $p = 0.5$  is approximately **0.5**.

**Question 8: How does log transformation help in stabilizing variance and making data more normally distributed?**

**Answer: Log transformation** is a common data-preprocessing technique used to **reduce skewness**, **stabilize variance**, and make data **closer to a normal distribution**, which is important for many statistical methods.

## 1. What is Log Transformation?

In log transformation, each data value (X) is replaced by:

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$Y = \log(X)$  (or  $\log(X+c)$  if zeros exist)

Common logs used:

- Natural log ( $\ln$ )
- Log base 10
- Log base 2

## 2. How Log Transformation Stabilizes Variance

**Problem: Heteroscedasticity**

In many datasets, variance **increases with the mean** (e.g., income, sales, population).

**How Log Helps**

- Log **compresses large values** more than small values
- Reduces the spread of high-value observations
- Makes variance more constant across different levels of data

**Example:**

Raw data: 10, 100, 1000

Log data: 1, 2, 3

Large gaps become smaller → **variance stabilizes**

## 3. How Log Transformation Makes Data More Normal

**Problem: Right-Skewed Data**

Many real-world datasets are **positively skewed** (long right tail).

**How Log Helps**

- Pulls in extreme high values
- Reduces right skewness
- Makes the distribution more **symmetric and bell-shaped**

**Example:**

Income or salary data often become more normally distributed after log transformation.

## 4. Additional Benefits

# Distribution

## 1. Reduces effect of outliers

Extreme values have less influence after log transformation.

## 2. Improves model assumptions

Helps meet assumptions of:

- Linear regression
- ANOVA
- Correlation analysis

## 3. Linearizes relationships

Turns exponential relationships into linear ones.

# 5. When to Use Log Transformation

✓ Data is **right-skewed**

✓ Variance increases with mean

✓ Values are **positive**

✗ Not suitable for zero or negative values (unless adjusted)

## Conclusion:

Log transformation compresses large values, stabilizes variance, reduces skewness, and helps data better approximate a normal distribution—making statistical analysis more reliable and accurate.