

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:

$$P(X=5)=(30/5) (1/6)^5 (56)^{25}$$

Using Excel (as hinted):

$$\begin{aligned} &= \text{BINOM.DIST}(5, 30, 1/6, \text{FALSE}) \\ &= 0.192 \end{aligned}$$

Result:

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

Probability ≈ 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.

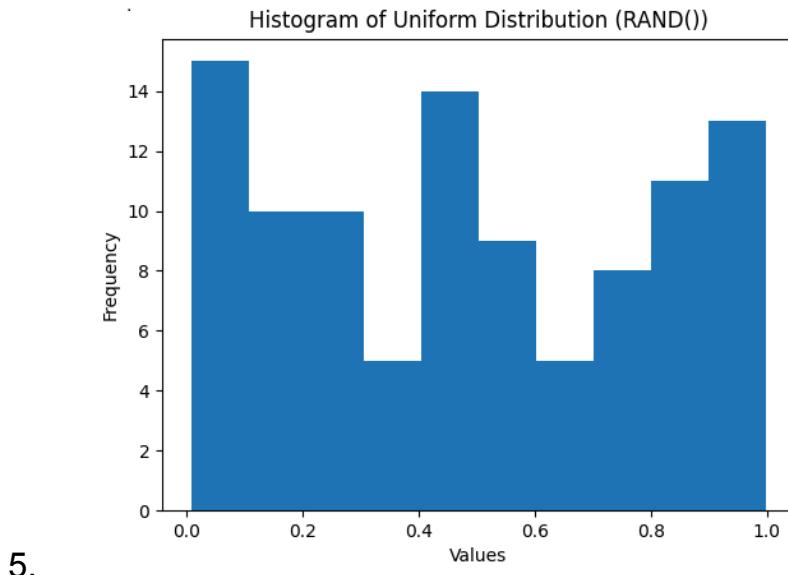
Answer: Step 1: Generate 100 Uniform(0,1) values

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]**.

Distribution

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 (μ) = 50
- Standard deviation (σ) = 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**.

Distribution

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
 - (μ) = mean of the dataset
 - (σ) = standard deviation of the dataset
 - (z) = standardized value (z-score)
-

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	≈ 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}

Distribution

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** → ($P(X \geq 8)$)

Excel Formula

In Excel, use:

=1 - BINOM.DIST(7, 15, 0.5, TRUE)

Distribution

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:

Distribution

$Y = \log(X)$ (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.