

# Probability Distributions

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## What is a Probability Distribution?

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Imagine you're analyzing real-world data — sales, weather, clicks, heights, etc. You'll notice:

- Some values are more **likely** than others.
- There's a **pattern** to how data is spread.

That pattern is described by a **probability distribution**.

**In short:**

A probability distribution tells you how likely different outcomes are.

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## What is a Probability Distribution?

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A probability distribution is a **mathematical function** or **table** that:

- Assigns **probabilities** to all possible outcomes of a random process.

There are two types:



### 1. Discrete Probability Distribution

- For outcomes you can **count** (e.g. number of heads in 3 coin tosses).
- Example: **Binomial Distribution**

Think: "What's the probability I get exactly 2 heads in 3 coin tosses?"

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## 2 Continuous Probability Distribution

- For outcomes that can be any number in a range (e.g. height, weight).
- Example: Normal Distribution ✓

✓ Think: "What's the probability someone's height is between 165 cm and 170 cm?"

height 175 cm  
178.2  
173.881

180.7721

weight → 63.7721 ✓  
75.2  
78.9

## A Simple Analogy

Let's say you roll a die:

- Outcomes = {1, 2, 3, 4, 5, 6} ✓
- Each has a probability of 1/6

✓ This is a uniform distribution (discrete).

$$\begin{aligned}P(1) &= \frac{1}{6} \\P(2) &= \frac{1}{6} \\&\vdots \\P(6) &= \frac{1}{6}\end{aligned}$$

Now imagine measuring people's heights:

- ✓ You don't get fixed values.
- ✓ Instead, you get a curve — most people around average height, fewer very short or very tall.
- ✓ That's a normal distribution (continuous).

## Why Are Distributions Useful in Data Science?

- ✓ 1. Model real-world randomness (user behavior, errors, arrivals, etc.)
2. Make predictions (how likely is a customer to buy?)
3. Run simulations (A/B testing, Monte Carlo)
4. Assume underlying patterns (e.g., linear regression assumes errors are normally distributed)

### Summary:

Type	Example	Used For
✓ Discrete	Binomial, Poisson	Count of events
✓ Continuous	Normal, Uniform	Measuring quantities

$$P(171.2) =$$
$$P(180.7) =$$
$$P(150.8) =$$

76kx  
↓