



# Random Variables and Probability

## Learning outcomes:

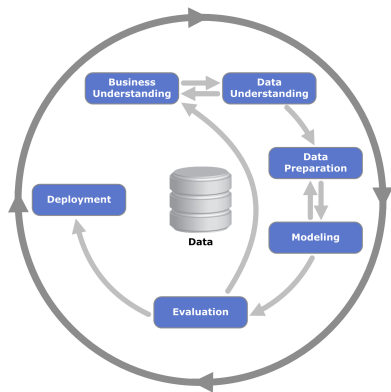


Figure: The [CRISP-DM](#) process.

- ▶ Talk about a little bit of math;
- ▶ Introduce the ideas of random variables and probability;
- ▶ Discuss Bernoulli, Binomial, and Normal random variables;

## Motivation – reasoning about loot boxes...

- ▶ Imagine a player in CounterStrike claiming that opening loot boxes gives them a legendary item 30% of the time.
- ▶ The player insists this isn't just luck—it's consistent.
- ▶ How can we assess this claim?  
Conduct an experiment:
  - ▶ **Sample** (open)  $n$  loot boxes in the game;
  - ▶ Record the number of legendary items obtained.



# What are random variables?

- ▶ **Sample space** for an experiment or random trial = all possible outcomes or results of that experiment;
- ▶ **Random variable:**
  - ▶ A variable whose values depend on outcomes of a random phenomenon;
  - ▶ A variable whose values have a relative likelihood governed by a probability distribution;
- ▶ **Probability distribution:** a mathematical function that gives the probabilities that the different possible outcomes occur. Must satisfy three rules:
  1. Probabilities are  $> 0$ ;
  2. If you add up the probabilities for all possible outcomes they must equal 1;
  3. To find the probability of two mutually exclusive events happening you add up their individual probabilities.

## What are random variables: examples.

- ▶ The outcome of a FAIR coin toss:
  - ▶ Sample space is:  $\{\text{Heads}, \text{Tails}\}$ ;
  - ▶ Probability distribution is:  $\{0.5, 0.5\}$ ;
- ▶ The outcome of rolling a six sided dice:
  - ▶ Sample space is:  $\{1, 2, 3, 4, 5, 6\}$ ;
  - ▶ Probability distribution is:  $\{\frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}, \frac{1}{6}\}$ ;
- ▶ The number of heads you get if you toss a coin 10 times:
  - ▶ Sample space is:  $\{0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}$ ;
  - ▶ Probability distribution is?
- ▶ The number of legendary items obtained from loot boxes:
  - ▶ If  $N$  loot boxes are opened, the sample space is:  $\{0, 1, 2, 3, \dots, N\}$ ;
  - ▶ Probability distribution is?

# Bernoulli Random Variables

- ▶ A **Bernoulli random variable** is used to model the outcome of an experiment that can be answered as a **boolean** – typical answers could be:
  - ▶ Yes or No;
  - ▶ Success or Failure;
  - ▶ True or False;
  - ▶ 1 or 0;
- ▶ Has a probability distribution with a single parameter:  $\pi$ , the probability of Yes/Success/True/1;
- ▶ Let's take the outcomes to be 0 and 1. Then we can write this as a function:

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$$\text{or: } p(k, \pi) = \pi^k (1 - \pi)^{1-k} \text{ for } k = 0 \text{ or } k = 1.$$

# Binomial Random Variables

- ▶ A **binomial random variable** is used to model the sum of  $n$  Bernoulli experiments – how many successes after  $n$  tries?
- ▶ Has a probability distribution with three parameters:
  1.  $n$ , the number of tries;
  2.  $k$ , the number of successes;
  3.  $\pi$ , the probability of each success;
- ▶ If we had a single try we could have either  $k = 1$  success or  $k = 0$  successes:

$$p(k = 1, n = 1, \pi) = \pi$$

$$p(k = 0, n = 1, \pi) = 1 - \pi$$

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- ▶ If we had two tries we could have  $k = 2, 1$  or  $0$  successes:

$$p(k = 2, n = 2, \pi) = \pi^2$$

$$p(k = 1, n = 2, \pi) = \pi(1 - \pi) + (1 - \pi)\pi$$

$$p(k = 0, n = 2, \pi) = (1 - \pi)^2$$

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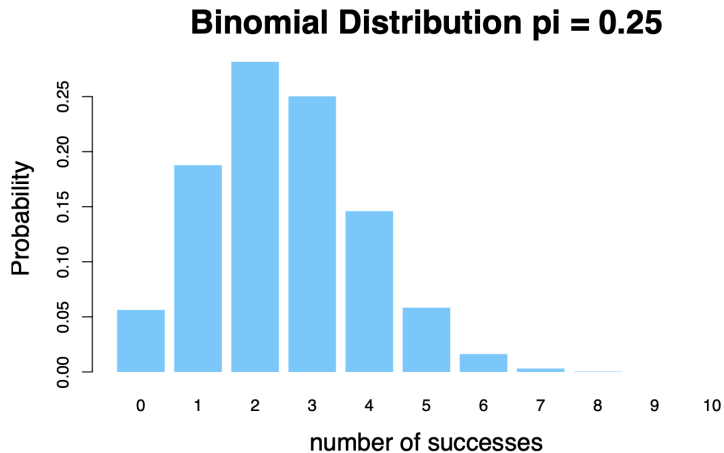
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# Binomial Random Variables

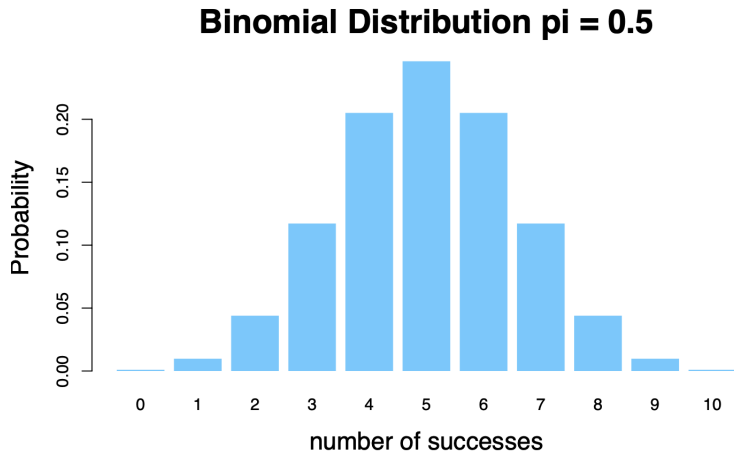
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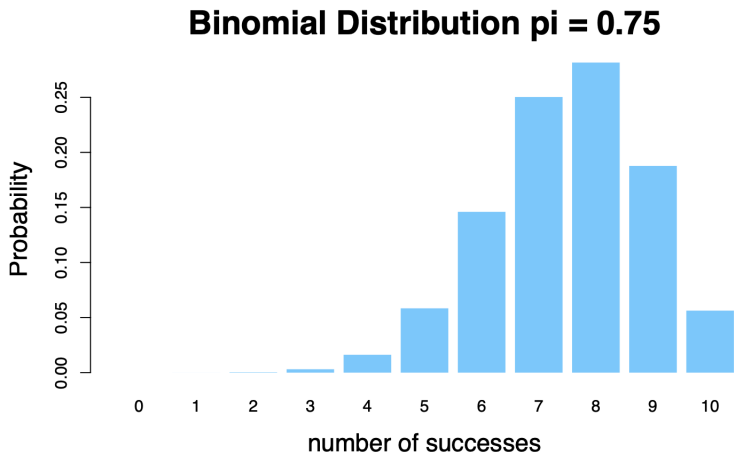
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- ▶ Has a probability distribution with two parameters:  $\mu$ , the mean and  $\sigma$  the standard deviation:

$$N(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{(x - \mu)^2}{2\sigma^2} \right\};$$

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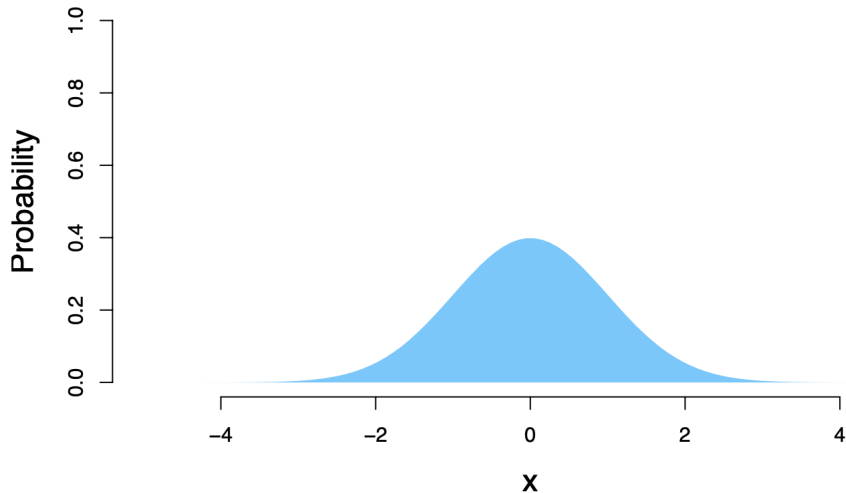
$$N(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp \left\{ -\frac{(x - \mu)^2}{2\sigma^2} \right\};$$

- ▶ When  $\mu = 0$  and  $\sigma = 1$  we call it the **standard normal**:

$$N(x, \mu = 0, \sigma = 1) = \frac{1}{\sqrt{2\pi}} \exp \left\{ -\frac{x^2}{2} \right\}.$$

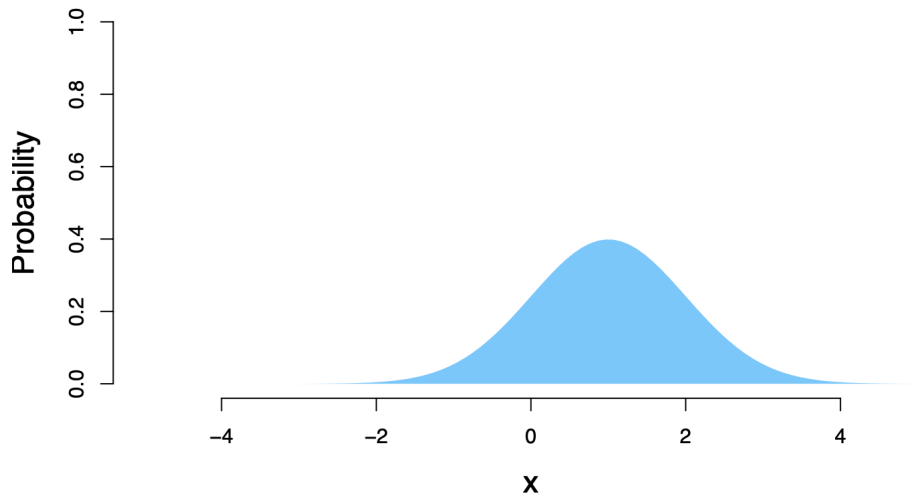
Normal – what happens as mean changes?

**Normal Distribution, mean = 0 sd = 1**



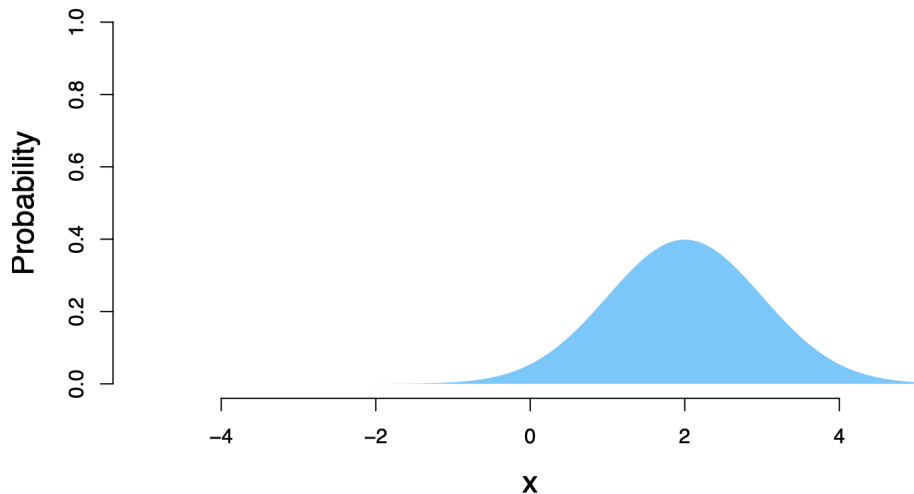
Normal – what happens as mean changes?

**Normal Distribution, mean = 1 sd = 1**



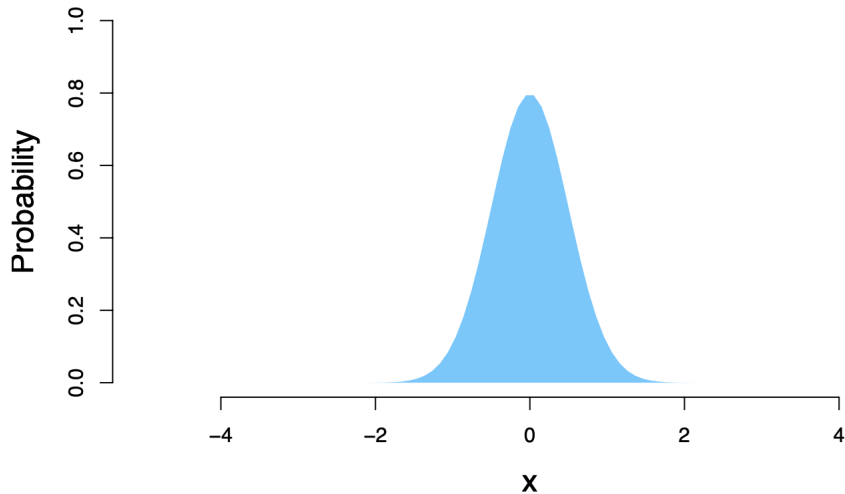
Normal – what happens as mean changes?

**Normal Distribution, mean = 2 sd = 1**



Normal – what happens as standard deviation changes?

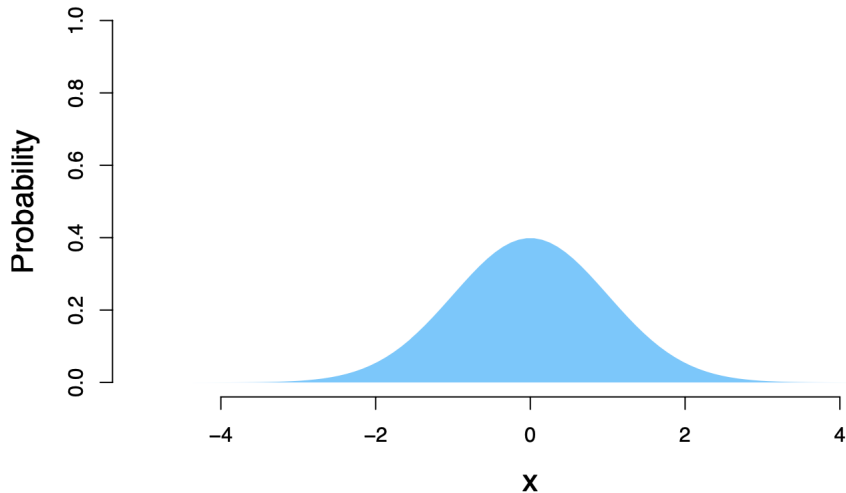
**Normal Distribution, mean = 0 sd = 0.5**





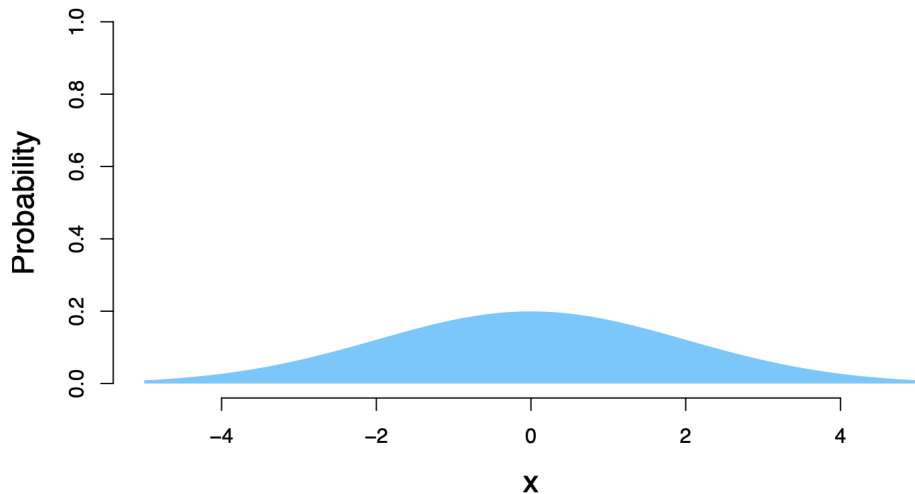
Normal – what happens as standard deviation changes?

**Normal Distribution, mean = 0 sd = 1**



Normal – what happens as standard deviation changes?

**Normal Distribution, mean = 0 sd = 2**



# What can we do with probability?

## Motivation – reasoning about voters...

- ▶ Cavalier Johnson: a political candidate running a campaign for Mayor of Milwaukee;
- ▶ Johnson's campaign claims that he will win the election – in this case that means getting more than 50% of votes from Milwaukee voters;
- ▶ How to assess the Johnson campaign claim? Conduct an experiment:
  - ▶ **Sample**  $n$  voters from Milwaukee;
  - ▶ Record the number of voters who will vote for Johnson.



If we see that  $k$  out of  $n$  loot boxes contain legendary items, what is the probability the player's claim about the drop rate is correct?