

Probability

Why do we need probability?

Probability theory allows us to

- Analyze uncertain or random processes that takes place in real world.
- Make decisions in Uncertain scenarios.

Examples

- Gambling: How do we estimate the returns?
- Model building: How do we derive stochastic component of model?
- Lotteries
- Simulation

What do you mean by probability?

Frequentist - probabilities are the numbers that are derived from outcome of a large number of experiments or trials.

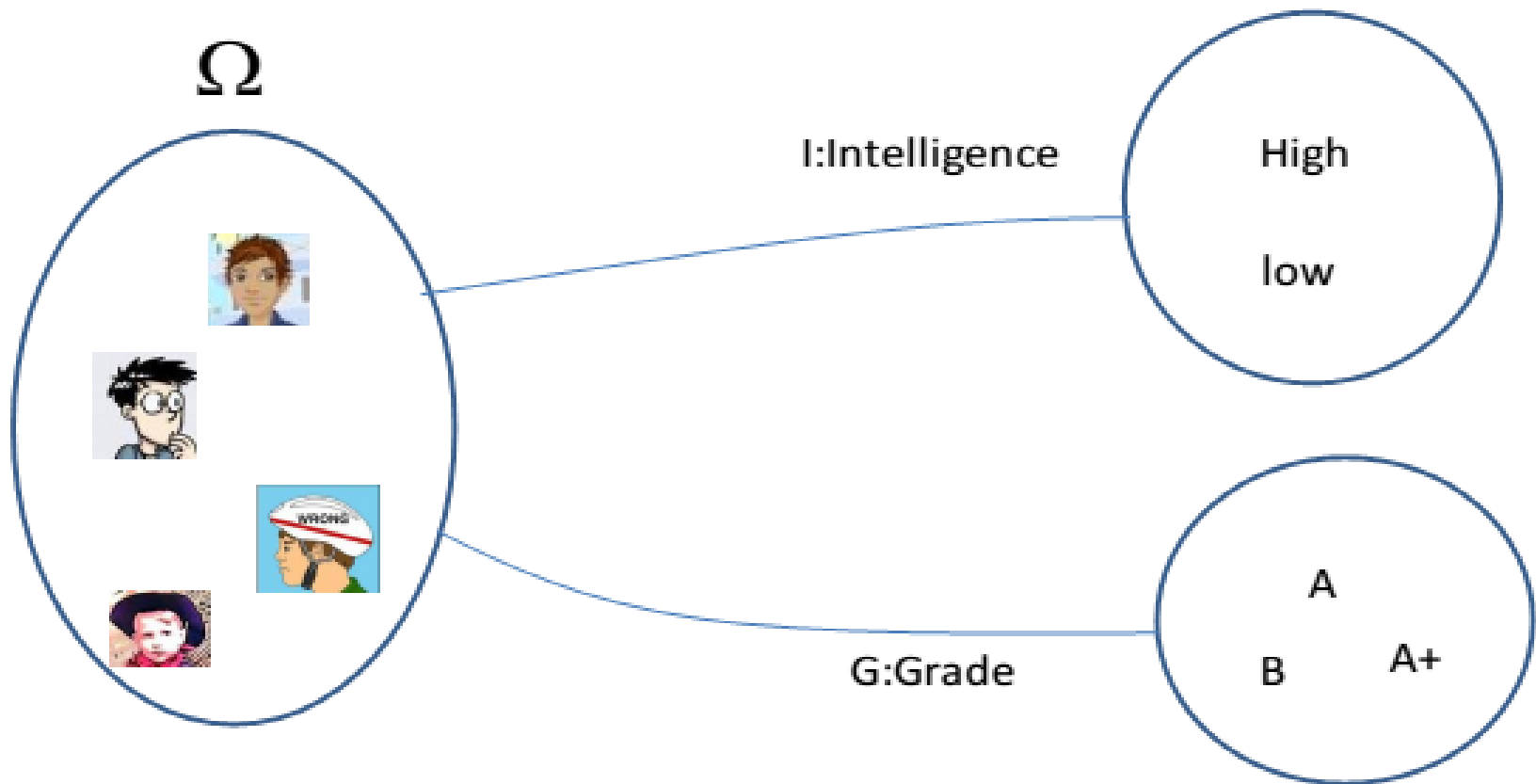
Bayesian - probabilities are numbers that express a degree of belief in something given any relevant evidence.

Random Experiments & Random Variables

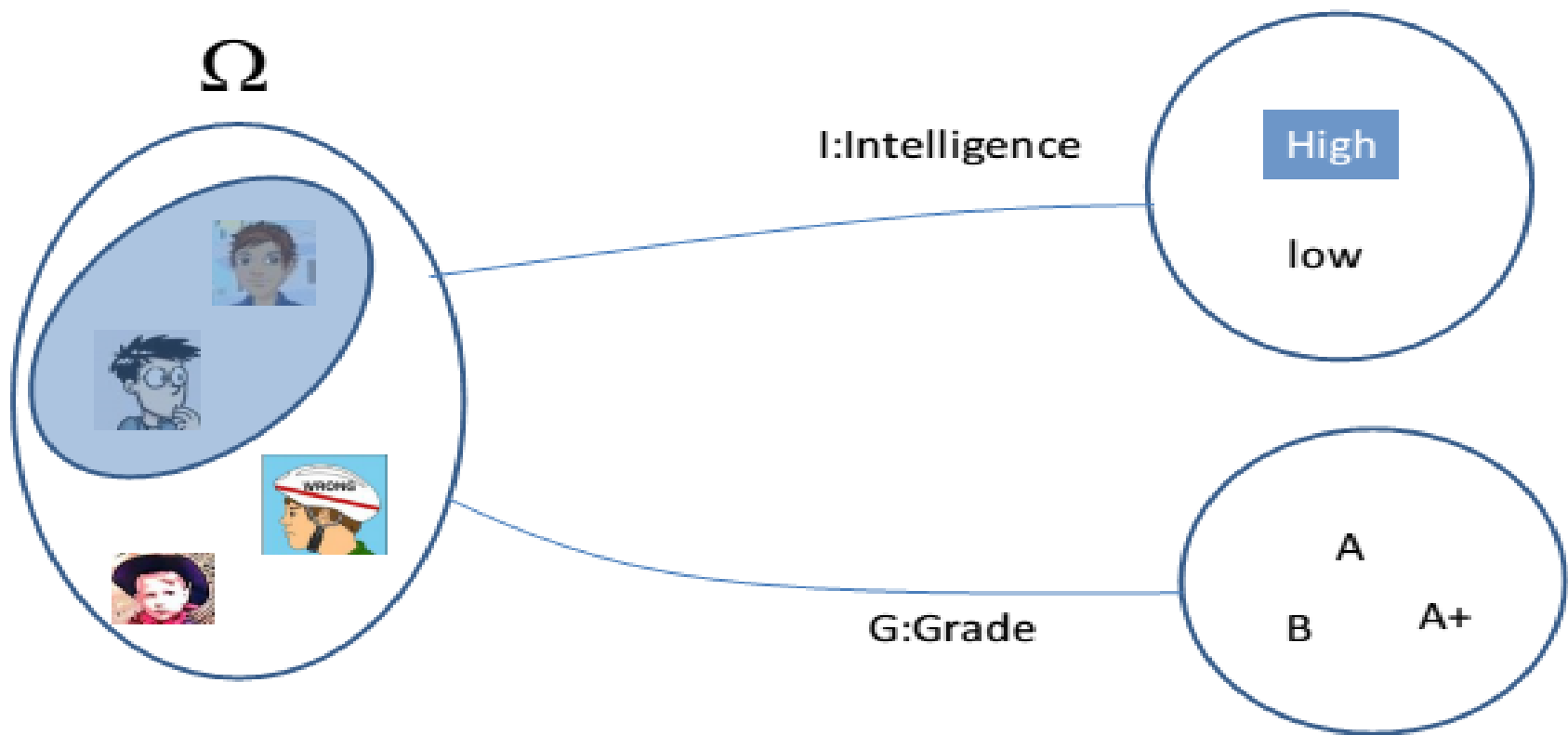
Random Variable

A variable is said to be a **random variable** when it comes from a random experiment

Random Variables



Random Variables



$P(I = \text{high}) = P(\{\text{all students whose intelligence is high}\})$

$P(G = A) = P(\{\text{all students whose grade is A}\})$

Types of random variables

Random Variables



discrete

continuous

Discrete RV Example

Experiment: toss a die and observe the value that appears on the top face

Random Variable: number on the top face



Continuous RV Example

Measurement

X : the height of an individual



Basic Probability

Probabilities

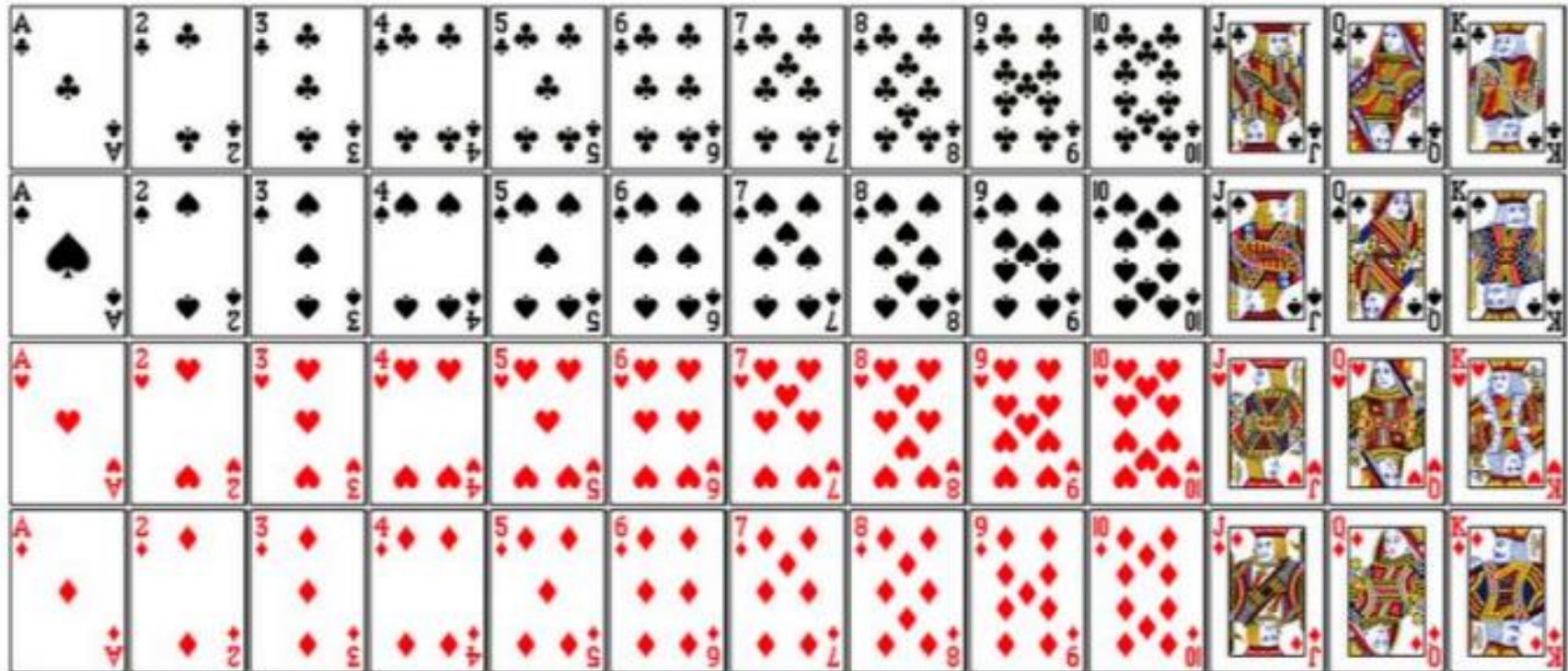
- What is the probability for getting heads if we toss a coin?
- What is the probability that we get 6 if we throw a dice?

Classical Approach

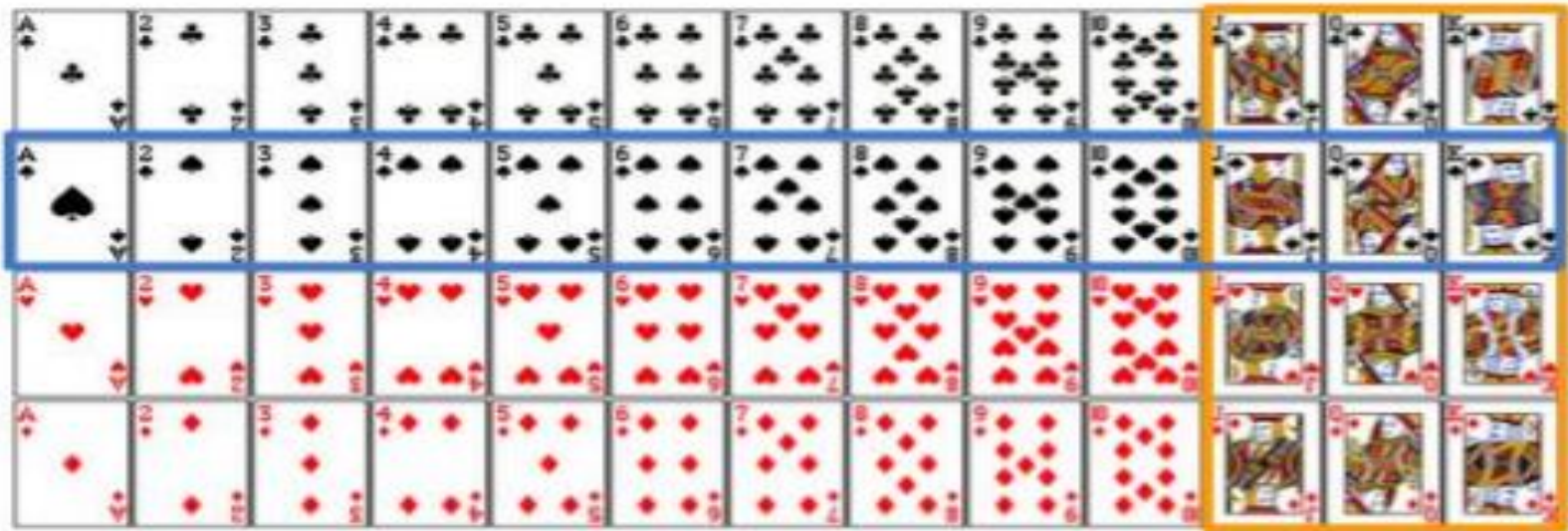
$$\text{Prob} = \frac{\text{number of favorable outcomes}}{\text{total number of outcomes}}$$

Addition Rule for Probabilities

Find the probability with which either spade or face card is selected?



Addition Rule



A = event a spade card is selected

B = event a face card is selected

$$P(\text{A}) = 13 / 52$$

$$P(\text{B}) = 12 / 52$$

$$P(\text{A and B}) = 3 / 52$$

$$P(\text{A or B}) = 13 / 52 + 12 / 52 - 3 / 52 = 22 / 52$$

Addition Rule

Probability of “A or B”

The probability of a composite event “A or B” is expressed as

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

Addition rule: disjoint events

If A and B are disjoint (i.e. mutually exclusive)

$$P(A \text{ or } B) = P(A) + P(B) - P(\text{A and B})$$

$$P(A \text{ or } B) = P(A) + P(B)$$

Conditional Probabilities

Conditional Probabilities

Age	Gender	
	Females	Males
Under 25 years	48	50
25 years & over	74	66

- a) If we pick random person, what is the probability for being male?
- b) If we pick random person, what is the probability for being under 25 years?
- c) If we pick random female, what is the probability for being above 25 years?
- d) If we pick random person under 25 years, what is the probability for being female?

Conditional Probability

- The probability that event A occurs **if we know for certain that event B** will occur is called **conditional probability**.
- The **conditional probability of A given B** is denoted: **$P(A/B)$** [Read as the probability of A given B]

Definition of Conditional Probability

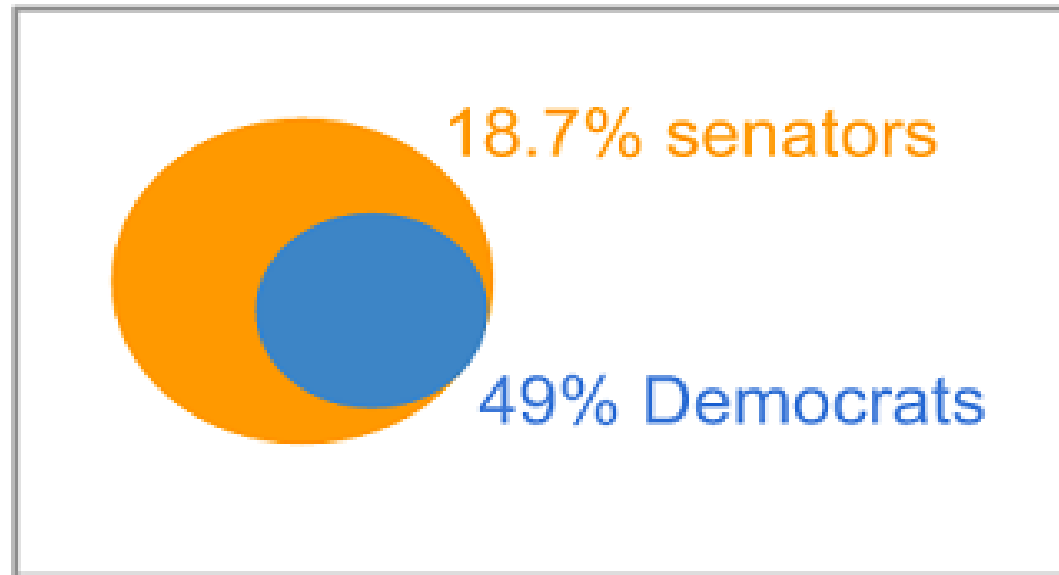
$$P(A|B) = \frac{P(A \text{ and } B)}{P(B)}$$

Corollary: The Chain Rule

$$P(A \text{ and } B) = P(A|B) P(B)$$

Multiplication Rule for Probability

110th Congress



Probability that a randomly selected member
is a **Democratic Senator**?

Multiplication Rule

$$P(\text{Dem and Sen}) = P(\text{Sen}) P(\text{Dem} \mid \text{Sen})$$

$$P(\text{D and S}) = 0.187 \times 0.49 = 0.092$$

Multiplication Rule (general case)

For any events A and B:

$$P(A \text{ and } B) = P(A) P(B | A)$$

or

$$P(B \text{ and } A) = P(B) P(A | B)$$

Chain Rule

For 2 variables:

$$P(A,B) = P(A | B) P(B)$$

For 3 variables:

$$\begin{aligned} P(A,B,C) &= P(A | B,C) P(B,C) \\ &= P(A | B,C) P(B | C) P(C) \end{aligned}$$

For n variables:

$$P(A_1, A_2, \dots, A_n) = P(A_1 | A_2, \dots, A_n) P(A_2 | A_3, \dots, A_n) P(A_{n-1} | A_n) P(A_n)$$

Multiplication Rule (independent events)

If events A and B are independent:

$$P(A \text{ and } B) = P(A) P(B)$$

or

$$P(B \text{ and } A) = P(B) P(A)$$

B is independent of A means that knowing B does not change our belief about A.

What is the probability of the following outcomes when a fair coin is independently tossed 6 times? (H: heads, T: tails)

- H H H H H H

Relating Conditional Probabilities

Problem1: Cancer Diagnosis

- Let's suppose we want to diagnose patients who have taken a test for cancer
- We know the test is 90% reliable (90% of people with the cancer get a positive test result, 10% of people without the cancer also get a positive result)
- 2% of the total population have this cancer
- Suppose we have a positive test result
- What is the probability our patient has cancer?

Problems2: Fair Coin or Not?

We have a coin that can be either fair or double headed. We had prior belief of it being fair with 0.9 probability.

- If we get heads after tossing coin, what is the probability for coin being fair?
- If we get heads again after second toss, what is the probability for coin being fair?
- Does the prior probability has any impact on our analysis?

Bayes' Rule

- The most important rule in all of probability
- Lets you “flip round” conditional probabilities
- Updates our belief about a in light of new information b



$$P(a|b) = \frac{P(b|a)P(a)}{P(b)}$$

A diagram illustrating Bayes' Theorem. The equation is
$$p(C | \mathbf{x}) = \frac{p(\mathbf{x} | C) p(C)}{p(\mathbf{x})}$$
. Four red labels with arrows point to specific parts of the equation: 'likelihood' points to $p(\mathbf{x} | C)$, 'prior probability' points to $p(C)$, 'evidence' points to $p(\mathbf{x})$, and 'posterior probability' points to $p(C | \mathbf{x})$.

likelihood

prior probability

posterior probability

evidence

$$p(C | \mathbf{x}) = \frac{p(\mathbf{x} | C) p(C)}{p(\mathbf{x})}$$

Probability Rules

Description (Rule)	Expression
Range	$0 \leq P(A) \leq 1$
Complementary probability	$P(A^c) = 1 - P(A)$
Decomposition	$P(A) = P(A \text{ and } B) + P(A \text{ and } B^c)$
Conditional probability	$P(B A) = P(A \text{ and } B) / P(A)$
Independence	$P(B A) = P(B) ; P(A B) = P(A)$
Multiplication rule	$P(A \text{ and } B) = P(B A) P(A)$
Addition rule	$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$
Baye's rule	$P(B A) = P(A B) P(B) / P(A)$

Expected Value, Variance & SD

A certain lottery works by picking 6 numbers from 1 to 49. It costs \$1.00 to play the lottery, and if you win, you win \$2 million after taxes. Answer the following questions:

- a. If you play the lottery once, what are your expected winnings or losses?
- b. If you play the lottery every week for 10 years, what are your expected winnings or losses?

Expectation or Expected Value

$$E(X) = \mu$$

Expected Value (μ)

The expected value of X is a **weighted average** of the possible values of X

Each value of X is weighted by the assigned probability

Expected Value Formula

$$\mu = p_1 x_1 + p_2 x_2 + \dots + p_n x_n$$

$$\mu = \sum_{i=1}^n p_i x_i$$

The probability function

$x\$$	$p(x)$
-1	.999999928
+ 2 million	7.2×10^{-8}

Calculate the probability of winning in 1 try:

$$\frac{1}{\binom{49}{6}} = \frac{1}{\frac{49!}{43!6!}} = \frac{1}{13,983,816} = 7.2 \times 10^{-8}$$

“49 choose 6”

Out of 49 numbers,
this is the number
of distinct
combinations of 6.

Expected Value

The probability function

$x\$$	$p(x)$
-1	.999999928
+ 2 million	7.2×10^{-8}

Expected Value

$$\begin{aligned} E(X) &= P(\text{win}) * \$2,000,000 + P(\text{lose}) * -\$1.00 \\ &= 2.0 \times 10^6 * 7.2 \times 10^{-8} + .999999928 (-1) \\ &= .144 - .999999928 = -\$.86 \end{aligned}$$

Conclusion for 1 time play:

Negative expected value is never good!

You shouldn't play if you expect to lose money!

Expected Value

Conclusion for playing lottery every week for 10years:

$$520 \times (-.86) = -\$447.20$$

Negative expected value is never good!

You shouldn't play if you expect to lose money!

Infomercial example

X: # of times buyers of a product watched a TV infomercial before purchasing it

# times watched TV infomercial	1	2	3	4	5
Percentage of buyers	27%	31%	18%	9%	15%

- Find the expected number of times buyers of a product watched TV commercial?
- Find the standard deviation of number of viewing behavior?

Variance of a random variable

$$\text{Var}(\mathbf{X}) = \sigma^2 \qquad \sigma^2 = \sum_{i=1}^n p_i (x_i - \mu)^2$$

Standard deviation of an RV

$$\sigma = \sqrt{\sum_{i=1}^n p_i (x_i - \mu)^2}$$