

# Probability and Randomness

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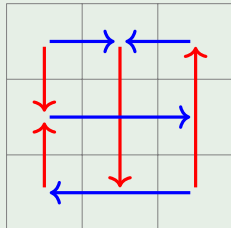
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# The Dilemma

- If there's a saddle point, we should pick that
- If there's no saddle point, there's no clear best choice
- But if we settle on one strategy, we can be exploited.

## Example (Rock Paper Scissors)

|          | Rock | Paper | Scissors |
|----------|------|-------|----------|
| Rock     | 0    | -1    | 1        |
| Paper    | 1    | 0     | -1       |
| Scissors | -1   | 1     | 0        |



## Discussion Question

How do we pick a strategy?

# Probability

- Want to talk about randomness.

## Definition

- The possible results of a random process are **outcomes**
- The set of all possible outcomes is the **sample space**.
- Term comes from statistics—taking a random sample.

## Example

- Tossing a coin
  - Possible outcomes: “heads” and “tails”.
  - Sample space:  $\{h, t\}$ .
- Rolling a six-sided die
  - Six possible outcomes.
  - Sample space:  $\{1, 2, 3, 4, 5, 6\}$ .

## Definition

- Assign each outcome a **probability** between 0 and 1
- For  $k$  outcomes, write  $p_1, p_2, \dots, p_n$
- $0 \leq p_k \leq 1$ 
  - $p_k = 0$  means outcome  $k$  can't happen
  - $p_k = 1$  means outcome  $k$  must happen
- $p_1 + p_2 + \dots + p_n = 1$ 
  - I.e. exactly one thing will happen
- The list of numbers  $P = (p_1, \dots, p_n)$  is a **probability distribution**

# Probability and Chance

## Remark

- In English, often refer to probabilities as **chances**
- Usually given as a percentage
- A 50% chance is the same as  $p = 1/2$ .

## Example (Coin toss)

- Chance of heads: 50%
- Chance of tails: 50%
- $P = (1/2, 1/2)$ .

## Example (Die roll)

- Six faces, each is equally likely
- $P = (1/6, 1/6, 1/6, 1/6, 1/6, 1/6)$ .

## Remark

- Our definition assumes we have listed every possible outcome.
- In real life that's never quite true.
  - What if a coin lands on edge?
- Useful assumption
- Matches assumption that our matrix contains every possible strategy.

# Probabilities are not always equal

- Sometimes one outcome is more likely than another
- Sometimes obvious, sometimes subtle
- In English, “choose at random” usually means every option is equally likely
- In this class, it usually does not.

## Definition

- The **uniform distribution** on  $n$  outcomes gives every outcome the same probability
- Get the probability distribution  $P = (1/n, 1/n, \dots, 1/n)$ .

# Non-Uniform Probability Distributions

- Can just give a non-uniform distribution
- $P = (1/3, 2/3)$  means first option is half as likely as the second one.

## Example (Weather forecasting)

- Might say 50% chance of sun, 30% chance of clouds, 20% chance of rain.
- Interpret this as a probability distribution:  $P = (0.5, 0.3, 0.2)$ .

# Non-Uniform Probability Distributions

- Many non-uniform distributions come from uniform distributions

## Example (Deck of cards)

- 12 face cards
- 40 non-face cards
- Each *card* equally likely
- $P = (12/52, 40/52) = (3/13, 10/13) \approx (0.23, 0.77)$

# Non-Uniform Probability Distributions

- We can usually build a non-uniform distribution out of a uniform one.

## Example

- Say we want the distribution  $P = (1/6, 1/4, 1/3, 1/4)$ .
- Check they sum to 1?
  - $\frac{1}{6} + \frac{1}{4} + \frac{1}{3} + \frac{1}{4} = \frac{2}{12} + \frac{3}{12} + \frac{4}{12} + \frac{3}{12} = \frac{2+3+4+3}{12} = 1.$
- Can generate with a uniform distribution over 12 possibilities.
- E.g. take 12 cards, and write  $A$  on 2,  $B$  on 3,  $C$  on 4,  $D$  on 3
  - Draw a card uniformly at random: get distribution  $P$
- Or divide a spinner into 12 equal sections
- Or look at seconds on your watch. First 10 seconds gives  $A$ , next 15 gives  $B$ , next 20 gives  $C$ , last 15 gives  $D$ .

# Random Variables

- Want to talk about probability in context of strategic choices
- Need to think about how *valuable* outcomes are
- The standard terminology is really bad.

## Definition

- Suppose we have a sample space with  $n$  outcomes
- and a probability distribution  $P = (p_1, p_2, \dots, p_n)$
- A **random variable**  $X$  on this sample space is a function that assigns a real number to each of the  $n$  possible outcomes.
- Can write as  $X = (x_1, x_2, \dots, x_n)$ .

# Random Variables

## Remark

- We think of the random variable as the value of an outcome—how much we like it.
- Similar to difference between “outcome” and “payoff”.

## Example

- Rolling a six-sided die:
  - $x_1 = 1, x_2 = 2, x_3 = 3, x_4 = 4, x_5 = 5, x_6 = 6$
  - $X = (1, 2, 3, 4, 5, 6)$

## Example

- Chance of getting a penny, dime, or quarter
  - $x_1 = 1, x_2 = 10, x_3 = 25$
  - $X = (1, 10, 25)$ .

## Definition

- Given a sample space with  $n$  outcomes
- With probability distribution  $P = (p_1, p_2, \dots, p_n)$ .
- Let  $X$  be a random variable that assigns the payoff  $x_k$  to outcome  $k$ .
- We define the **expected value** of  $X$  to be

$$E = E(X) = p_1x_1 + p_2x_2 + \dots + p_nx_n.$$

- Idea: average payoff per play if you play a bunch of times.
- If you run your process a hundred times, you will get about  $100 \cdot E(X)$ .

## Example

- What is the expected value of rolling a six-sided die?

$$\begin{aligned} E &= \frac{1}{6} \cdot 1 + \frac{1}{6} \cdot 2 + \frac{1}{6} \cdot 3 + \frac{1}{6} \cdot 4 + \frac{1}{6} \cdot 5 + \frac{1}{6} \cdot 6 \\ &= \frac{1}{6} (1 + 2 + 3 + 4 + 5 + 6) = \frac{1}{6} \cdot 21 = 3.5. \end{aligned}$$

## Example (Lottery)

- Grand prize: \$100,000,000 with probability 1 in 150,000,000
  - Second prize: \$200,000 with probability 1 in 3,000,000
  - Third prize: \$10,000 with probability 1 in 150,000
  - Fourth prize: \$10 with probability 1 in 300
  - What is the expected value of playing?
- 
- How many possible outcomes? **Five**
  - How to compute?

# Expected Value

| Prize:       | Grand         | Second      | Third     | Fourth | Lose  |
|--------------|---------------|-------------|-----------|--------|-------|
| Payoff:      | \$100,000,000 | \$200,000   | \$10,000  | \$10   | \$0   |
| Probability: | 1/150,000,000 | 1/3,000,000 | 1/150,000 | 1/300  | 0.997 |

$$\begin{aligned} E &= \frac{100,000,000}{150,000,000} + \frac{200,000}{3,000,000} + \frac{10,000}{150,000} + \frac{10}{300} + 0 \cdot 0.997 \\ &\approx 0.667 + 0.067 + 0.067 + 0.033 + 0 \\ &\approx 0.833. \end{aligned}$$

- Average payoff: about 83 cents.

# Expected Value

|              |               |             |           |        |       |
|--------------|---------------|-------------|-----------|--------|-------|
| Prize:       | Grand         | Second      | Third     | Fourth | Lose  |
| Payoff:      | \$100,000,000 | \$200,000   | \$10,000  | \$10   | \$0   |
| Probability: | 1/150,000,000 | 1/3,000,000 | 1/150,000 | 1/300  | 0.997 |

- Average payoff: about 83 cents.

## Discussion Question

- Would you play if you got a ticket for free?
- Would you pay a dollar for a ticket?
- What's the most you'd be willing to pay for a ticket?

# Expected Value

- Suppose you pay \$1 for a ticket
- New payoffs for each outcome:

| Prize:       | Grand         | Second      | Third     | Fourth | Lose  |
|--------------|---------------|-------------|-----------|--------|-------|
| Payoff:      | \$99,999,999  | \$199,999   | \$9,999   | \$9    | -\$1  |
| Probability: | 1/150,000,000 | 1/3,000,000 | 1/150,000 | 1/300  | 0.997 |

$$E = \frac{99,999,999}{150,000,000} + \frac{199,999}{3,000,000} + \frac{9,999}{150,000} + \frac{9}{300} + (-1) \cdot 0.997$$
$$\approx 0.667 + 0.067 + 0.067 + 0.033 + -1 \approx -0.167.$$

- Lose about 17 cents per ticket.

# Expected Value

## Discussion Question

- When rolling a six-sided die, do you expect to get a 3.5?
- When entering that lottery, do you expect to lose seventeen cents?

## Remark

- “Expected value” is a dangerous phrasing
- You do not (usually) expect to get that exact value
- Computes the *average* result over many attempts
- Useful for analyzing repeated decisions
- Sort of useful for analyzing one-off decisions

## Definition

The **expected value principle** says a rational player will choose the option with the largest expected value.

- Can be seen as a definition of “rational”
- Often taken as prescriptive: a player *should* choose
- Has many caveats! Just one model of decision-making.

# Limitations to Expected Value Principle

- Useful framework for analyzing decisions
- Most useful when making repeated decisions
- Most useful when all payoffs are on the same scale

## Discussion Question

Should you pay \$1 for a 1 in 100,000 chance of getting \$100,001?

- Expected value:  $\frac{100,001}{100,000} = 1.00001$
- More than a dollar, so expected value principle says yes
- But should you?

# Limitations to Expected Value Principle

## Discussion Question

Should you pay \$1,000,000 for a 1 in 1,000 chance of getting \$2,000,000,000?

- Expected value:  $\frac{2,000,000,000}{1,000} = 2,000,000$
- Expected value principle says yes
- Is this a good idea?
- Do you have a million dollars?
- What would you do with 2 billion dollars?
- What will you do if you lose?

# Limitations to Expected Value Principle

## Diminishing marginal utility of money

- First dollar is more valuable than the millionth dollar
  - People worry about “risk of ruin”—chance of going broke
  - This is the point of insurance—negative expected value, but reduces risk of ruin.
- 
- Can avoid some problems by giving “utility” instead of dollars
    - Kind of fake, not clear what we’re measuring
    - But useful mathematical abstraction
    - (In the real-life applications we care about, the numbers are kind of made up anyway.)
  - This is just one way to analyze decisions—but it’s often useful, and it’s the framework we’ll use in this course.

# Mixed Strategies in Games

- Now have the tools to analyze games like Rock Paper Scissors.

|          | Rock | Paper | Scissors |
|----------|------|-------|----------|
| Rock     | 0    | -1    | 1        |
| Paper    | 1    | 0     | -1       |
| Scissors | -1   | 1     | 0        |

- No saddle points
- Every strategy has a counter-strategy
- Any predictable strategy will lose.
- Need unpredictability—which means randomness.

## Definition

- Consider an  $m \times n$  zero-sum two-player game
- The strategies corresponding to rows and columns are now called **pure strategies**
- A **mixed strategy** for Row is a probability distribution  $P = (p_1, \dots, p_m)$  on their set of pure strategies
  - Row chooses a pure strategy at random, choosing row  $k$  with probability  $p_k$
- A mixed strategy for Column is a probability distribution  $Q = (q_1, \dots, q_n)$  on their set of pure strategies.

# Mixed Strategies

## Example

- Column plays Rock Paper Scissors with  $Q = (1/4, 1/2, 1/4)$ 
  - Rock 1/4 the time, Paper 1/2 the time, Scissors 1/4 the time
- Now Row can view each strategy as a lottery

|          | Rock | Paper | Scissors |
|----------|------|-------|----------|
| Rock     | 0    | -1    | 1        |
| Paper    | 1    | 0     | -1       |
| Scissors | -1   | 1     | 0        |

$1/4 \quad 1/2 \quad 1/4$

- If Row plays Rock:  $E = \frac{1}{4} \cdot (0) + \frac{1}{2} \cdot (-1) + \frac{1}{4} \cdot (1) = -\frac{1}{4}$ 
  - On average, Row will lose one point per four games.

# Mixed Strategies

|          | Rock | Paper | Scissors |
|----------|------|-------|----------|
| Rock     | 0    | -1    | 1        |
| Paper    | 1    | 0     | -1       |
| Scissors | -1   | 1     | 0        |

$\frac{1}{4}$        $\frac{1}{2}$        $\frac{1}{4}$

- If Row plays Rock:  $E = \frac{1}{4} \cdot (0) + \frac{1}{2} \cdot (-1) + \frac{1}{4} \cdot (1) = -\frac{1}{4}$ 
  - On average, Row will lose one point per four games.
- If Row plays Paper:  $E = \frac{1}{4} \cdot (1) + \frac{1}{2} \cdot (0) + \frac{1}{4} \cdot (-1) = 0$
- If Row plays Scissors:  $E = \frac{1}{4} \cdot (-1) + \frac{1}{2} \cdot (1) + \frac{1}{4} \cdot (0) = \frac{1}{4}$
- What should Row do? Play Scissors.

# Mixed Strategies

## Discussion Question

Is “Play Rock every time” a mixed strategy?

## Definition

- A **basic mixed strategy** is a probability distribution with every probability except one equal to 0.
  - Write  $P_i$  for Row's basic mixed strategy that sets  $p_i = 1$  and always plays row  $i$
  - Write  $Q_j$  for Column's basic mixed strategy that sets  $q_j = 1$  and always plays column  $j$ .
- 
- We'll start by studying what happens when one player uses a pure strategy against the other player's mixed strategy
  - Later we'll generalize.

# Mixed Strategies and Expected Value

## Lemma

Consider a  $m \times n$  matrix game.

- If Row plays  $P_i$  against Column's mixed  $Q = (q_1, \dots, q_n)$ , then the expected value of the payoff is

$$E(P_i, Q) = q_1 u_{i,1} + q_2 u_{i,2} + \dots + q_n u_{i,n}.$$

- Similarly, if Column plays  $Q_j$  against Row's mixed  $P = (p_1, \dots, p_m)$ , then the expected value of the payoff is

$$E(P, Q_j) = p_1 u_{1,j} + p_2 u_{2,j} + \dots + p_m u_{m,j}.$$

## Corollary

If  $P_i$  and  $Q_j$  are basic mixed strategies, then  $E(P_i, Q_j) = u_{i,j}$ .