



# SEARCH PROBLEMS

[CONTINUED ...]

# ASSIGNMENT/QUIZ 1

- Later this week, you will have assignment 1.
- Release Date: Friday, August 9.
- Due: Monday, August 12.
- Covers Weeks 1 to 3.
- Multiple-choice quiz over Wattle.

***Keep an eye on Wattle announcements for further instructions!***

# EGG DROPPING PROBLEM

**Input:** There is a building with  $n$  floors.

**Output:** You want to find the **highest** floor that if you drop an egg from it doesn't break.

**Assumptions:**

- If the egg breaks when dropped from floor  $k$ , then it would also have broken from any floor above that.
- If the egg survives a fall, then it will survive any fall below that.

**Objective:** *Minimize the number of required drop tests (trials).*

## SCENARIO 3

What if you are given 2 identical eggs?

### Binary Search Approach

- Drop the egg from floor  $\lceil \frac{n}{2} \rceil$ .
- If it breaks, then we have to apply the strategy with 1 egg for floors  $1, \dots, \lceil \frac{n}{2} \rceil - 1$ .
- This takes  $\Omega(n)$  in the worst case.

**Can we do better?**

## SCENARIO 3: A MORE OPTIMAL SOLUTION

- Roughly speaking, we want to minimize our maximum regret.
- Suppose we drop our egg from floor  $d$ .
- If it breaks, we step through  $(d - 1)$  floors one by one.
- If not, instead of jumping another  $d$  floors, we should step up just  $d - 1$  floors because we have one less drop available.
- Thus, the next floor should be  $d + (d - 1)$ .
- If it breaks, we do the floor-by-floor search, otherwise we jump to  $d + (d - 1) + (d - 2)$ , and so on.

## SCENARIO 3: A MORE OPTIMAL SOLUTION

- The number of drops in this algorithm is at most  $d$  by design.
- However, to ensure that we will cover all floors, we need
$$d + (d - 1) + (d - 2) + \cdots + 1 \geq n$$
- Thus, we are interested in the minimum  $d$  which satisfies the above equation.
- The left-hand side is  $\frac{d(d+1)}{2}$ . It is easy to see that there is a  $d$  in  $O(\sqrt{n})$ , which satisfies this equation.
- Thus, we need  $O(\sqrt{n})$  trials.



# PROBABILISTIC ANALYSIS

[CLRS SEC. 5.2]

# PROBABILISTIC ANALYSIS

Two randomized setups:

- The algorithm is deterministic, but the input is from a random distribution.
- Randomized algorithm: the algorithm itself makes random choices.

Our objective is to analyze the **expected running time** for such algorithms.



# TOPICS

- Probability Refresher
- Probabilistic Analysis of Deterministic Algorithms
- Probabilistic Analysis of Randomized Algorithms

# PROBABILISTIC MODELS

Probabilistic Modelling Assumptions:

- Experiments with random outcome
- Which outcome is likely to happen is quantifiable

Probabilistic modelling consists of three components:

- Sample Space ( $S$ ): Set of all possible outcomes.
- Events ( $E$ ): Subsets of sample space.
- Probability ( $P$ ): “Quantifies” the likelihood of an event.

# PROBABILISTIC MODELS

Probability: A function that maps events to real numbers satisfying the following 3 axioms:

- Non-negativity:  $P(A) \geq 0$  for any event  $A \in \mathbf{E}$
- $P(\bigcup_{i=1}^n A_i) = \sum_{i=1}^n P(A_i)$  where  $A_i \in \mathbf{E}$  are disjoint (mutually exclusive) events.
- Normalization:  $P(S) = 1$ .

# EXAMPLE



Flipping a fair coin, head (H) or tail (T).

$$S = \{H, T\}$$

$E = \{\emptyset, \{H\}, \{T\}, \{H, T\}\}$  and the probabilities are:

$$P(\emptyset) = 0, P(\{H\}) = P(\{T\}) = 0.5, P(\{H, T\}) = 1$$

For compactness of writing, we usually do not list zero and one probabilities. Thus, we write the above as

$$S = \{H, T\} ; P(\{H\}) = P(\{T\}) = 0.5$$

# EXAMPLE

Biased dice

- Each even face is 2 times as likely as each odd face.
- All even faces are equally likely.
- All odd faces are equally likely.

What is the probability that the outcome is less than 4 ?

$$P(\{1\}) = P(\{3\}) = P(\{5\}) = x$$

$$P(\{2\}) = P(\{4\}) = P(\{6\}) = 2x$$

$$3x + 3 * 2x = 1 \rightarrow x = 1/9.$$

$$P(\{1\} \cup \{2\} \cup \{3\}) = P(\{1\}) + P(\{2\}) + P(\{3\}) = x + 2x + x = 4x = \frac{4}{9}.$$



# RANDOM VARIABLES

Probability assigns likelihood to subsets of the sample space (events).

In many cases, it's easier if the sample space has an ordering structure.

**Random variable**  $X$  is a function that maps sample space  $S$  to a number.

# RANDOM VARIABLES: DISCRETE

S: Students of COMP3600/6466

P: We choose a student uniformly at random

X: The program a student is enrolled in

Suppose the programs are indexed, e.g., BAC: 0, BIT: 1, Eng: 2, PhB: 3, MCompSci: 4

$X(\text{student 1}) = 3$  ;  $X(\text{student 2}) = 0$  ;  $X(\text{student 3}) = 0$  ; ....

$P(X = 0)$  means the probability that the randomly chosen student is enrolled in BAC.

$$P(X = 0) = \frac{\# \text{ BAC students}}{\# \text{ students}}$$

# EXPECTED VALUE

Expected value is the weighted average of possible values of  $X$ , where the weight is the probability:

$$E[X] = \sum_{\forall x} x P(X = x)$$
$$E[g(X)] = \sum_{\forall x} g(x) P(X = x)$$

Linearity of expectation for random variables  $X$  and  $Y$  and constants  $a$  and  $b$ :

$$E[aX + b] = aE[X] + b$$
$$E[X + Y] = E[X] + E[Y]$$



# INDICATOR RANDOM VARIABLES

Suppose we have sample space  $S$  and an event  $A$ , then:

- The indicator random variable associated with event  $A$  is

$$I\{A\} = \begin{cases} 1 & A \text{ occurs} \\ 0 & A \text{ does not occur} \end{cases}$$

- Let  $X_A = I\{A\}$ , then  $E[X_A] = P(\{A\})$

Proof:

$$E[X_A] = E[I\{A\}] = 1 * P(\{A\}) + 0 * P(S \setminus \{A\}) = P(\{A\})$$

# INDICATOR RANDOM VARIABLES: EXAMPLES

$$S = \{H, T\}; P(\{H\}) = P(\{T\}) = 0.5$$

$$X_H = I\{\{H\}\} = \begin{cases} 1 & \{H\} \text{ occurs} \\ 0 & \{T\} \text{ occurs} \end{cases}$$

$$E[X_H] = E[I\{H\}] = 0.5$$

# INDICATOR RANDOM VARIABLES: EXAMPLES

Expected number of heads obtained when flipping a fair coin  $n$  times?

Let  $x_i$  be the indicator random variable that head occurs in the  $i^{\text{th}}$  flip, and random variable  $X$  be the number of heads in  $n$  flips.

$$\text{➤ } x_i = I\{\{H_i\}\} = \begin{cases} 1 & \text{head occurs in the } i^{\text{th}} \text{ flip} \\ 0 & \text{tail occurs in the } i^{\text{th}} \text{ flip} \end{cases}$$

$$\begin{aligned} \text{➤ } E[X] &= E\left[\sum_{i=1}^n x_i\right] \\ &= \sum_{i=1}^n E[x_i] \quad (\text{based on linearity of expectation}) \\ &= \sum_{i=1}^n 0.5 \quad (\text{based on results in previous slide}) \\ &= 0.5n. \end{aligned}$$

# NUMBER OF MATCHES

From a randomly shuffled deck of size  $n$ , cards are laid out on a table, face up from left to right, and then another shuffled deck is laid out so that each of its cards is beneath a card of the first deck.

What is the expected number of matches of the card above and the card below?

# NUMBER OF MATCHES

Each card in the first deck has 1 chance in  $n$  of matching its paired card.

Label the cards from 1 to  $n$ . Then, define the indicator random variable  $x_i$  to be 1 if and only if card number  $i$  is matched. Then

$$E[x_i] = P(x_i = 1) = \frac{1}{n}$$

The total number of matches is  $X = \sum_{i=1}^n x_i$ . Thus, the expected number of matches is

$$E[\sum_{i=1}^n x_i] = \sum_{i=1}^n E[x_i] = \sum_{i=1}^n \frac{1}{n} = 1.$$

# COUPON COLLECTOR PROBLEM

There are  $n$  types of coupons and we want to collect at least one from each type. In each trial, we are given a coupon uniformly at random among all types of coupons.

**Question:** How many coupons do we need to buy in expectation to have all  $n$  types?

# COUPON COLLECTOR PROBLEM

Let random variable  $X$  be the number of trials we need. We are interested in  $E[X]$ .

Let  $x_i$  be the number of coupons we need to buy to get a new coupon while having  $i - 1$  types of coupons. Then, using linearity of expectation

$$E[X] = E\left[\sum_{i=1}^n x_i\right] = \sum_{i=1}^n E[x_i].$$

# COUPON COLLECTOR PROBLEM

If we own  $i - 1$  coupons, the probability of picking a new coupon is exactly

$$p_i = 1 - \frac{i - 1}{n} = \frac{n - i + 1}{n}$$

Random variable  $x_i$  is **geometrically distributed** with parameter  $p_i$ . Thus, we have

$$E[x_i] = \frac{1}{p_i} = \frac{n}{n - i + 1}$$



# COUPON COLLECTOR PROBLEM

$$E[X] = \sum_{i=1}^n E[x_i] = \sum_{i=1}^n \frac{n}{n-i+1} = n \sum_{j=1}^n \frac{1}{j}$$

$\sum_{j=1}^n \frac{1}{j}$  is the  $n$ -th **Harmonic Number** and equal to  $\log n + c$  for some constant  $c$ .

Hence,  $E[X] = n \log n + cn$ .