

# Discrete Mathematics and Its Applications 2 (CS147)

*Lecture 12: Conditional expectation, coupon collector's problem*

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Information!

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A dice is repeatedly thrown until it lands on a 6. Let  $T$  be the number of rolls it takes for a dice to roll a 6, and let  $A$  be the event that all dice rolls in a sequence are even. What is  $\mathbb{E}(T|A)$ ?

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### Solution

We know the information the sequence are *even*, i.e., 2, 4, 6.

$\mathbb{E}(T|A) \Leftrightarrow$  find the expected number of throws until the result is 6.

$\Rightarrow p = 1/3$

$\Rightarrow \mathbb{E}(T|A) = 1/p = 3$ .



## Conditioning on a random variable

- $\mathbb{E}(X|A)$  is a value ( $A$  is an event)
- $\mathbb{E}(X|Y)$ : the expected value of  $X$  conditioned on  $Y$  is itself a **random variable**
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### Informal understanding

- ▶  $E(X)$ : average - the best estimate of a  $X$  given no information about it
- ▶  $\mathbb{E}(X|Y)$ : we have already known the information from  $Y$ , how to give a good estimation for  $X$ ? a function of  $Y$  that best approximates  $X$

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### Property

- ▶  $\mathbb{E}(X|X) = X$
- ▶  $\mathbb{E}(X|Y) = \mathbb{E}(X)$  if  $X, Y$  are independent.
- ▶  $\mathbb{E}(aX + bY|Z) = a\mathbb{E}(X|Z) + b\mathbb{E}(Y|Z)$  for two constants  $a, b$ .

# Law of total expectation

## Theorem

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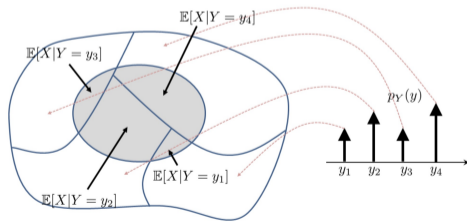
$$\mathbb{E}[\mathbb{E}[X|Y]] = \mathbb{E}[X]$$

## Proof.

$$\begin{aligned}\mathbb{E}[\mathbb{E}[X|Y]] &= \sum_y \mathbb{E}[X|Y = y] \Pr(Y = y) \\ &= \sum_y \left( \sum_x x \Pr(X = x|Y = y) \right) \Pr(Y = y) \\ &= \sum_y \sum_x x \Pr(X = x, Y = y) \\ &= \sum_x x \sum_y \Pr(X = x, Y = y) = \sum_x x \Pr(X = x).\end{aligned}$$

□

total expectation theorem



## Examples (I)

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Let  $X$  and  $Y$  be the values of independent six-sided dies. What is  $\mathbb{E}(X|X + Y)$ ?

Intuition: we know the information of  $X + Y$  and want to estimate  $X$ .

- ▶  $\mathbb{E}(X + Y|X + Y) = X + Y$
- ▶  $\mathbb{E}(X + Y|X + Y) = \mathbb{E}(X|X + Y) + \mathbb{E}(Y|X + Y) = 2\mathbb{E}(X|X + Y)$  by symmetry

Then we have  $\mathbb{E}(X|X + Y) = (X + Y)/2$ . □

## Examples (II)

### Example

We roll two standard 6-sided dice, let  $X_1$  and  $X_2$  be the numbers we obtain and  $X = X_1 + X_2$ . Compute  $\mathbb{E}[X_1|X = 8]$ .

### Solution

$$\mathbb{E}[X_1|X = 8] = \sum_{i=1}^6 i\Pr(X_1 = i|X = 8) = \sum_{i=2}^6 i\Pr(X_1 = i|X = 8),$$

where  $X_1 \neq 1$  for the condition  $X = X_1 + X_2 = 8$ . Otherwise  $X_2 = 7$ , which is unrealistic.



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where  $X_1 \neq 1$  for the condition  $X = X_1 + X_2 = 8$ . Otherwise  $X_2 = 7$ , which is unrealistic. Then, the event  $\{X_1 = i|X = 8\}$  for any  $i \in \{2, 3, 4, 5, 6\}$  is an equal-probability event, so

$$\mathbb{E}[X_1|X = 8] = \sum_{i=1}^6 i\Pr(X_1 = i|X = 8) = \sum_{i=2}^6 i\Pr(X_1 = i|X = 8) = \frac{1}{5} (2 + 3 + 4 + 5 + 6) = 4.$$

# Coupon collector's problem

## Problem

*We repeatedly sample from a set of  $N$  distinct objects until at least one copy of each distinct object is obtained. Denote  $T$  as the number of draws until the every  $\{1, 2, \dots, N\}$  is seen, what is  $\mathbb{E}(T)$ ?*

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Recall Geometric distribution:

- ▶ Fails in the first  $n - 1$  times

$$\Pr(X = n) = (1 - p)^{n-1} p$$

- ▶ Success at the  $n$ -th time

## Illustration

- ▶ Sample a new object: ✓
- ▶ Sample a repeated object: ✗

⇒ success to sample a new object before previous (repeated) objects

⇒ the first occurrence of success for a new object

example with 5 coupons: [3] [1] [3] [5] [5] [3] [1] [1] [3] [2] [3] [2] [2] [4]

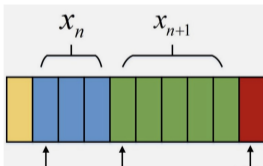
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## Solution

- ▶  $\Pr(\text{find the first unique coupon}) = \frac{N}{N} = 1$
- ▶  $\Pr(\text{find the second unique coupon}) = \frac{N-1}{N}$
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## Solution

Let  $x_n$  as how many times we need to collect the  $n$ -th **unique** coupon after collecting  $(n-1)$ -th **unique** coupons.

$$\Rightarrow x_n \sim \text{Geo}(p_n).$$

$$\Rightarrow p_n = \frac{N-n+1}{N}.$$

$$\begin{aligned}\mathbb{E}(T) &= \mathbb{E}\left(\sum_{n=1}^N x_n\right) = \sum_{n=1}^N \mathbb{E}(x_n) = \frac{N}{N} + \frac{N}{N-1} + \dots + \frac{N}{1} \\ &= N\left(1 + \frac{1}{2} + \dots + \frac{1}{N}\right) = NH_N \approx N(\log N + \gamma).\end{aligned}$$

## Next lecture...

Tail probability: We wish to create an upper bound  $R$  such that  $T$  exceed  $R$  with a low probability

$$\Pr(T \geq R) \leq \textit{small} .$$