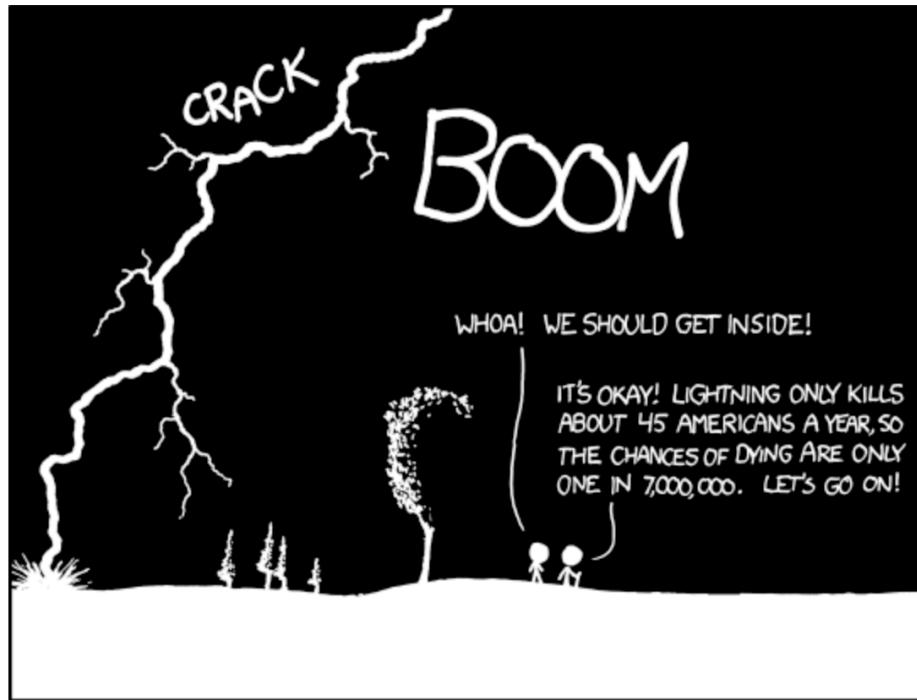


# Stat 88: Probability & Math. Statistics in Data Science



<https://xkcd.com/795/>

Lecture 17: 3/1/2021  
Expectation by conditioning

## Lec 17 : First recall from last time

$S \setminus X$	1	2	3	$P_S(s)$
1	$\frac{1}{4} \cdot \frac{1}{4} = \frac{1}{16}$	0	0	$Y_{16}$
2	$\frac{1}{4} \cdot \frac{1}{2} = \frac{1}{8}$	$\frac{1}{8}$	0	$Y_8 = \frac{1}{8}/\frac{1}{16}$
3	$\frac{1}{4} \cdot \frac{1}{2} = \frac{1}{8}$	$\frac{1}{8} (\frac{1}{2} \cdot \frac{1}{2})$	$\frac{1}{16} P(X=3, Y=1)$	$Y_4 = \frac{6}{16}$
4	$\frac{1}{4} \cdot \frac{1}{4} = \frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{16} P(X=3, Y=2)$	$Y_2 = \frac{4}{16}$
5	0	$\frac{1}{2} \cdot \frac{1}{4} = \frac{1}{8}$	$\frac{1}{8}$	$Y_1 = \frac{1}{16}$
6	0	0	$\frac{1}{16}$	$Y_0 = \frac{1}{16}$
	$\frac{4}{16}$	$\frac{8}{16}$	$\frac{4}{16}$	

Joint dsn of  $X$  and  $S$

$$E(X) = 1 \cdot \frac{4}{16} + 2 \cdot \frac{8}{16} + 3 \cdot \frac{4}{16} = 2$$

Writing down the dsn of  $g(S)$

$s$	1	2	3	4	5	6
$P(S=s)$	$\frac{1}{16}$	$\frac{4}{16}$	$\frac{6}{16}$	$\frac{4}{16}$	$\frac{1}{16}$	
$E(X S=s)$	1	1.5	2	2.5	3	

$$g(s) = E(X|s)$$

r.v.; what varies  $\rightarrow s$ .

$$E(g(s)) = 1 \cdot \frac{1}{16} + (1.5) \left( \frac{4}{16} \right) + (2) \left( \frac{6}{16} \right) + (2.5) \left( \frac{4}{16} \right) + 3 \left( \frac{1}{16} \right)$$

$$= \frac{1+6+12+10+3}{16} = \frac{32}{16} = 2.$$

$$E(g(s)) = E(E(X|s)) = E(X)$$

Given	$P(X=1)$	$P(X=2)$	$P(X=3)$	$E(X S=s)$
$s=2$	1	0	0	1
$s=3$	$\frac{1}{2}$	$\frac{1}{2}$	0	$\frac{3}{2} = 1.5$
$s=4$	$\frac{1}{6}$	$\frac{4}{6}$	$\frac{1}{6}$	$\frac{1 \cdot \frac{1}{6} + 2 \cdot \frac{4}{6} + 3 \cdot \frac{1}{6}}{\frac{1}{6} + \frac{4}{6} + \frac{1}{6}} = \frac{12}{6} = 2$
$s=5$	0	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{5}{2} = 2.5$
$s=6$	0	0	1	$3 - 1 = 3$

$g(s) = E(X|S=s)$  is a function of  $s$ , not  $x$   
 $E(X|S)$  is a r.v.

$$X = \begin{cases} 1 & \text{w/pr. } \frac{1}{4} \\ 2 & \text{w/pr. } \frac{1}{2} \\ 3 & \text{w/pr. } \frac{1}{4} \end{cases}$$

$$F_x = F_Y, X, Y \text{ idl.}$$

$$S = X+Y$$

## Expectation by Conditioning

cond'l = conditional

- In the example we just worked out, once we fix a value  $s$  for  $S$ , then we have a distribution for  $X$ , and can compute its expectation using that distribution that depends on  $s$ :  $E(X | S = s) = \sum_x x \underbrace{P(X = x | S = s)}_{\text{cond'l dsn of } X} \Big| S=s$ , with the sum over all values of  $X$ .
- Note that  $E(X | S = s)$  is a **function of  $s$** . We can think of  $E(X | S)$  as a rv.  
 $E(X | S) \leftarrow \text{r.v.}$
- This means that if we want to compute  $E(X)$ , we can just take a weighted average of these conditional expectations  $E(X | S = s)$ :

$$E(X) = \sum_s E(X | S = s) P(S = s)$$

- This is the **law of iterated expectation**.

$$E(E(X | S)) = E(X)$$

averages out the  $X$   
averages out the  $S$

## Law of iterated expectation

- Note that  $E(X | S = s)$  is a function of  $s$ . That is, if we change the value of  $s$  we get a different value. (It is not a function of  $x$ , though.)
- Therefore we can define the function  $g(s) = E(X | S = s)$ , and the random variable  $g(S) = E(X | S)$ .
- In general, recall that  $E(g(S)) = \sum_s g(s)f(s) = \sum_s g(s)P(S = s)$ .
- How can we use this to find the expected value of the rv  $g(S)$ ?

$$E(g(S)) = \underbrace{\sum_s E(X | S=s) P(S=s)}_{\text{ }} = E(X)$$

To prove this explicitly:

$$\text{write out } E(X | S=s) = \sum_x x \cdot P(X=x | S=s)$$

now you have a double sum,  
write out  $P(X=x | S=s) = \frac{P(X=x, S=s)}{P(S=s)}$

$$g(S) = E(X|S), \quad E(g(S)) = E(X)$$

Examples from the text: Time to reach campus

- 2 routes to campus, student prefers route A (expected time = 15 minutes) and uses it 90% of the time. 10% of the time, forced to take route B which has an expected time of 20 minutes. What is the expected duration of her trip on a randomly selected day?

$X$  = duration of the trip  $\begin{cases} 15 \text{ m} \\ 20 \text{ mins} \end{cases}$

$$E(X) = 15(0.9) + 20(0.1) = 15.5 \text{ minutes.}$$

Conditioning  $X$  on

$S$  = Route taken

$$E(X) = E(E(X|S=s)P(S=s))$$

$$= (15 \cdot P(S=A) + 20 P(S=B))$$

$$= 15(0.9) + 20(0.1)$$

## Catching misprints

$E(X | S=s)$  is NOT a r.v.  
a fixed #  
 $E(X | S)$  is a r.v.

- The number of misprints is a rv  $N \sim \text{Pois}(5)$  dsn. Each misprint is caught before printing with chance 0.95 independently of all other misprints. What is the expected number of misprints that are caught before printing?

$$P(\text{catching a misprint before printing}) = 0.95$$

let  $M = \# \text{ of misprints caught before printing}$

Sp. we know document has 8 misprints.

$$M \sim \text{Bin}(8, 0.95), E(M | N=8) = 8 \cdot 0.95$$

What about if  $N = n$ ?  $M \sim \text{Bin}(n, 0.95), E(M | N=n)$

$$E(M) = E(E(M | N))$$

(unchain of  $E$ )  
dependent on values of  $N$

$$= \sum_{n=0}^{\infty} E(M | N=n) \cdot P(N=n) = \sum_{n=0}^{\infty} 0.95^n \cdot P(N=n)$$

$$= 0.95 \sum_{n=0}^{\infty} n \cdot P(N=n) = (0.95)^5 \cdot 5$$

$$\mathbb{E}(\mathbb{E}(M|N=n)) = \sum_n \left( \sum_m m \cdot P(M=m|N=n) \right) P(N=n)$$

$$g(N) = \mathbb{E}(M|N)$$

$$\mathbb{E}(g(N)) = \sum_n g(n) P(N=n)$$

### Exercise 5.7.13

- A survey organization in a town is studying families with children. Among the families that have children, the distribution of the number of children is shown below. Suppose each child has chance 0.51 of being male, independently of all other children. What is the expected number of male children in a family picked at random from those that have children?

$$\mathbb{E}(N) = (1)(0.2) + 2(0.4) + 3(0.2) + 4(0.15) + 5(0.05)$$

$n$	1	2	3	4	5
prop. with $n$ children	0.2	0.4	0.2	0.15	0.05

$M = \# \text{ of male children in a randomly picked family}$

$N = \# \text{ of children}$

What is the dsn of  $M|N=n$ ,  $M \sim \text{Bin}(n, 0.51)$

$$\mathbb{E}(M|N=n) = 0.51n$$

$$\begin{aligned} \mathbb{E}(M) &= \mathbb{E}(\mathbb{E}(M|N)) = \sum_{n=1}^5 \mathbb{E}(M|N=n) P(N=n) \\ &= \sum_{n=1}^5 0.51n \cdot P(N=n) = (0.51)(1 \cdot 0.2) + 2(0.4) + 3(0.2) + \dots \\ &\approx 1.25 \end{aligned}$$

$$= 0.51 \sum_{n=1}^{\infty} n P(N=n) = 0.51 \cdot E(N)$$

## Expectation of a Geometric waiting time

- $X \sim \text{Geom}(p)$  :  $X$  is the number of trials until the first success
- $P(X = k) = (1 - p)^{k-1} p, k = 1, 2, 3, \dots$
- Let  $x = E(X)$  going to solve for  $x$
- Recall that  $P(X > 1) = P(\text{first trial is } F) = 1 - p$
- We can split the possible situations into when the first trial is a success and the first trial is a failure, and condition on this and compute the conditional expectation:

$X = T_1$  :

$Y = \text{result of first trial.}$   
either first trial  
S or F

$$E(X) = E(X | X = 1)P(X = 1) + E(X | X > 1)P(X > 1)$$

$$= 1 \cdot p + (x+1)(1-p)$$

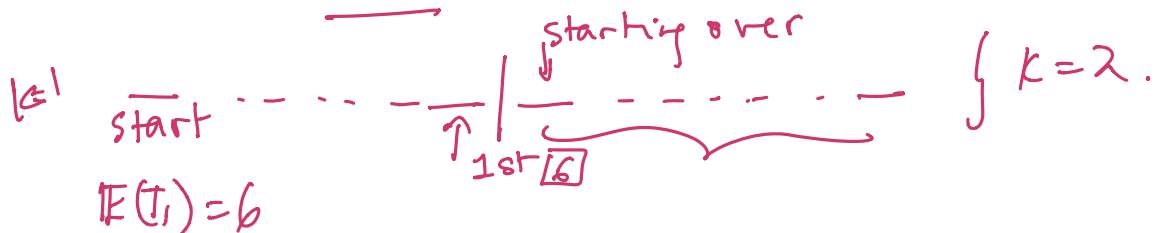
just like starting over  
but with 1 step more  
Exp # of trials =  $x + 1$

$$x = p + (x+1)(1-p)$$

$$x = p + x + 1 - px - p \Rightarrow x = 1/p = E(X)$$

## Expected waiting time until $k$ sixes have been rolled

- Let  $T_k$  be the waiting time until the  $k$ th six is rolled. (fani die  $P(6) = \frac{1}{6}$ )
- For example, consider  $k = 1$  (geometric), expected waiting time = 6 rolls =  $\frac{1}{\frac{1}{6}}$
- What about if  $k = 2?$   $k = 3?$   $k = 10?$



$$T_2 = T_1 + T_1 \rightarrow E(T_2) = E(T_1) \cdot 2 \\ = 12$$

$$E(T_{10}) = 60 = 10 \cdot E(T_1)$$

$\underbrace{\text{wait for } 1^{\text{st}} S}_{T_1} \bigg| \underbrace{\text{wait for } 2^{\text{nd}} S}_{T_1} \bigg| \underbrace{\text{wait for } 3^{\text{rd}} S}_{T_1} \dots$

- General formula for expected waiting time until  $k$ th success where  $P(S) = p$

$$= \underline{k \cdot \frac{1}{p}}$$

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