

### 3.3 Expected values

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Consider a university with 15'000 students. Let  $X$  be the number of courses for which a random student is registered. <sup>(Assume)</sup> The pf of  $X$  is:

$x$	0	1	2	3	4	5	6	7
$p(x_i)$	0.01	0.03	0.13	0.25	0.33	0.17	0.02	
# of students	$0.01 \times 15'000$ $= 150$	450	1950	3750	5850	2550	300	

The average number of courses per student is

$$\frac{1 \cdot 150 + 2 \cdot 450 + \dots + 7 \cdot 300}{15'000}$$

$$= 1 \cdot 0.01 + 2 \cdot 0.03 + \dots + 7 \cdot 0.02 = 4.57.$$

Rem: To compute the average of  $X$  over the population, we did not have to use the population size. Knowledge of the pf was enough.

Def. Let  $X$  be a discrete rv with pf  $p$ . The expected value (or mean value) of  $X$  is

$$EX := \sum_{x \in R} x \cdot p(x).$$

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Ex: For the above example,

$$\begin{aligned} EX &= \sum x p(x) = 1 \cdot p(1) + 2 \cdot p(2) + \dots + 7 \cdot p(7) \\ &= 1 \cdot 0.01 + 2 \cdot 0.03 + \dots + 7 \cdot 0.02 \\ &= 4.57. \end{aligned}$$

Let  $X$  be a Bernoulli rv with pf  $p(x) = p$  and  $p(0) = 1 - p$ , where  $p \in (0, 1)$ . Then,

$$EX = \sum x p(x) = 0 \cdot p(0) + 1 \cdot p(1) = p.$$

Let  $X$  be a rv with pf

$$p(x) := \begin{cases} \frac{k}{x^2}, & x = 1, 2, \dots \\ 0, & \text{else.} \end{cases}$$

Then,  $k = \frac{\pi^2}{6}$  ensures that  $\sum_{x \in \mathbb{N}} p(x) = 1$ . (You don't need to understand why. It is not easy.) The expected value of  $X$  is

$$\begin{aligned} EX &= \sum_x x p(x) = k \sum_{x \in \mathbb{N}} \frac{1}{x}. \quad \longrightarrow \text{"harmonic series"} \\ &= \infty. \end{aligned}$$

So, when  $p(x)$  does not decrease fast enough as  $x$  increases, then  $X$  does not have a <sup>finite</sup> expectation. We say  $X$  is heavy-tailed.

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## The expected value of a function

Suppose the cost of a diagnostic vehicle test depends on the number of cylinders  $X$  in the vehicle via

$$h(X) = 20 + 3X + 0.5X^2. \quad \text{If } X \text{ is a rv, then so}$$

is  $Y = h(X)$ . Suppose the pf of  $X$  is:

$x$	4	6	8
$p(x)$	0.5	0.3	0.2

 $\Rightarrow$ 

$y$	40	56	76
$p(y)$	0.5	0.3	0.2

$$\begin{aligned}\Rightarrow EY &= \sum_y y p(y) = 40 \cdot 0.5 + 56 \cdot 0.3 + 76 \cdot 0.2 \\ &= h(4) \cdot 0.5 + h(6) \cdot 0.3 + h(8) \cdot 0.2 \\ &= \sum_x h(x) p(x).\end{aligned}$$

Prop: let  $X$  be a rv with pf  $p$ . Then,

$$E h(X) = \sum_{x \in \mathbb{R}} h(x) p(x). \quad (!?)$$

Example: A store purchased three computers at 500\$ a piece. It will try to sell them at 1000\$ each. If it cannot it will return them to the manufacturer who will refund 200\$ to the store. What is the expected profit of the store if the pf of the number of computers sold  $X$  is

$x$	0	1	2	3
$p(x)$	0.1	0.2	0.3	0.4

The profit is  $h(x) = 1000x + 200(3-x) - 1500$  So,

$$\begin{aligned} E h(x) &= \sum h(x) p(x) = h(0)p(0) + h(1)p(1) + h(2)p(2) + h(3)p(3) \\ &= -900 \cdot 0.1 + (-100) \cdot 0.2 + 700 \cdot 0.3 + 1500 \cdot 0.4 \\ &= 700. \end{aligned}$$

The expected value of a linear function

Prop:  $E(ax+b) = a EX + b$ . "Linearity of expectation" (!)

Proof:  $E(ax+b) = \sum (ax+b) p(x) = a \sum x p(x) + b \sum p(x)$   
 $= a EX + b$

□

Cor: i  $E(ax) = a EX$  ii  $E(x+b) = EX + b$ .

Example: In the previous example  $h(x) = 800x - 900$

and  $EX = 0.2 + 0.6 + 1.2 = 2.0$

$$\Rightarrow E h(x) = 800 EX - 900 = 700.$$

The variance of  $X$

The expected value of  $X$  describes where the distr. of  $X$  is centered. The variance will describe its spread or variability:

Def: let  $X$  be a discrete rv. with pt  $p$  and mean value  $\mu$ .

The variance of  $X$  is

$$\sigma_x^2 = \sigma^2 = V(X) = \sum_{x \in \mathbb{R}} (x - \mu)^2 p(x) = E((X - \mu)^2).$$

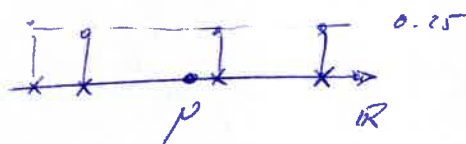
The standard deviation<sup>(sd)</sup> of  $X$  is

$$\sigma_x = \sqrt{\sigma_x^2}.$$

Rem:  $(X - \mu)^2$  is the squared deviation of  $X$  from its mean st  $\sigma_x^2$  is the expected squared deviation.



small variance / sd



large variance / sd.

Ex Consider a discrete rv  $X$  with pt

$x$	1	2	3	4	5	6
$p(x)$	0.3	0.25	0.15	0.05	0.1	0.15

The expected value of  $X$  is

$$EX = \sum x p(x) = 0.3 + 0.5 + \dots + 0.9 = 2.85$$

The variance of  $X$  is

$$V(x) = \sum (x - 2.85)^2 p(x) = (1 - 2.85)^2 \cdot 0.3 + (2 - 2.85)^2 \cdot 0.25 + \dots + (6 - 2.85)^2 \cdot 0.15 = \dots = 3.2275$$

Prop:  $V(X) = \sum_{x \in \Omega} x^2 p(x) - \mu^2 = E(X^2) - E(X)^2$

Example: In the example above:

$$\rightarrow \sigma^2 = E(x^2) - \mu^2 = 11.35 - (2.85)^2 = 3.2275.$$

We have that

If  $h(x) = ax + b$ , then

Prop:  $V(aX + b) = a^2 V(X)$  ,  $\sigma_{aX+b} = |a| \sigma_X$  [Pv]

Cor: ~~iii~~ i.  $\sigma_{aX} = |a| \sigma_X$  , ii.  $\sigma_{X+b} = \sigma_X$  .

Rem: Please, don't forget the absolute value!

### 3.4 The binomial probability distr.

Many experiments conform exactly or approximately to the following requirements:

- i. The experiment consists of a sequence of  $n$  other experiments, called trials.
- ii. Each trial has two possible outcomes. Let's say  $S$  and  $F$ .  $\rightarrow$  labels are arbitrary
- iii. The trials are independent: the outcome of any trial does not influence the outcome of any other trial.
- iv.  $P(S)$  is constant from trial to trial.

We call such an example a binomial experiment.

Ex. Tossing a coin  $n \in \mathbb{N}$  times is a binomial experiment.

ii. Drawing  $n$  balls from an urn containing  $k$  blue ones and  $l$  red ones with replacement, is a binomial experiment.

→ Replace the ball after every draw.

$$P(\text{a blue ball is drawn}) = \frac{k}{k+l}$$

$$P(\text{a red ball is drawn}) = \frac{l}{k+l}$$

iii. Think of drawing 6 balls without replacement from an urn containing 35 blue ones and 15 red ones. Then,

i. 
$$P(B \text{ on second draw} \mid B \text{ on first draw})$$
$$= \frac{34}{49} \approx 0.6939$$

ii. 
$$P(B \text{ on second draw})$$
$$= P(B \text{ on second} \mid B \text{ on first}) P(B \text{ on first})$$
$$+ P(B \text{ on second} \mid R \text{ on first}) P(R \text{ on first})$$
$$= \frac{34}{49} \cdot \frac{35}{50} + \frac{35}{49} \cdot \frac{15}{50} = \frac{35}{50} \left( \frac{34}{49} + \frac{15}{49} \right)$$
$$= \frac{35}{50} \approx 0.7$$

So, the trials are not independent. This can also be seen from



$$P(B \text{ on sixth} | BBBB) = \frac{30}{45} = \frac{2}{3} \approx 0.6$$

$$P(B \text{ on sixth} | RRRR) = \frac{35}{45} = \frac{7}{9} \approx 0.7$$

iv. Now draw 6 balls without replacement from an urn containing  $3.5 \cdot 10^6$  ~~red~~ <sup>blue</sup> balls and  $1.5 \cdot 10^6$  red balls. Then,

$$P(B \text{ on sixth} | BBBB) = \frac{3'499'985}{4'999'995} \approx 0.7000,$$

$$P(B \text{ on sixth} | FFFFF) = \frac{3'500'000}{4'999'995} \approx 0.7000.$$

So, here trials are effectively independent.

Upshot Sampling / Drawing with replacement always yields independent trials. Sampling / Drawing without replacement does so approximately if the population is very large. !

Def. i The binomial rv  $X$  associated with a binomial experiment is

$X$  is # of S among  $n$  trials

[1.9]

ii. The pf of a binomial rv depends on  $n$  and  $p$

the probability  $p$  of  $S$ : we denote it by  $b(x; n, p)$ .

Ex Consider 20 male and 10 female Monarch butterflies. Pick four butterflies with replacement. ~~What is the probability that~~ Let  $X$  be the number of male butterflies picked. What is the  $pt$  of  $X$ ?

$X$  can take the values 0, 1, 2, 3, 4:

$$P(X=0) = P(FFFF) = P(F)^4 = \left(\frac{10}{10+20}\right)^4 = \frac{1}{5^4}$$

$$\begin{aligned} P(X=1) &= P(MFFF \text{ or } FMFF \text{ or } FFMF \text{ or } FFFM) \\ &= 4 P(M) P(F)^3 = 4 \frac{2}{5} \frac{1}{5^3} \end{aligned}$$

$$P(X=2) = \binom{4}{2} P(M)^2 P(F)^2 = \binom{4}{2} \frac{2^2}{5^2} \frac{1}{5^2}$$

$$P(X=3) = \binom{4}{3} \frac{2^3}{5^3} \frac{1}{5}$$

$$P(X=4) = \binom{4}{4} \frac{2^4}{5^4}$$

The above is an experiment with replacement; i.e., a binomial experiment. So,

$P(X=n) = b(n; 4, \frac{2}{3}) = \begin{cases} \binom{4}{n} \left(\frac{2}{3}\right)^n \left(\frac{1}{3}\right)^{4-n} & n=0,1,2,3,4 \\ 0 & \text{else} \end{cases}$

Theorem:

$$b(x; n, p) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & \text{if } x = 0, 1, \dots, n \\ 0 & \text{else} \end{cases}$$

Proof:

$$b(x; n, p) = P(X = x) = \binom{\text{number of sequences of length } n \text{ consisting of } x \text{ S's}}{\text{probability of any such sequence}}$$

$$= \binom{n}{x} p^x (1-p)^{n-x}$$

probability of  $x$  S's

probability of  $n-x$   $F_s$

Def: If  $X$  has pdf  $b(x; n, p)$ , we write  $X \sim \text{Binom}(n, p)$ .

The cdf of  $X$  is

$$B(x; n, p) := P(X \leq x) = \sum_{y=0}^x b(y; n, p).$$

Example: Suppose 20% of all copies of a particular textbook fail a certain binding strength test. Let  $X$  be the number of copies that fail the test among 15 randomly selected copies.

- a. What is the prob. that at most 8 fail the test?
- b. \_\_\_\_\_ exactly \_\_\_\_\_
- c. \_\_\_\_\_ at least \_\_\_\_\_
- d. What is the probability that between 4 and 7 (inclusive) fail the test?

a. ~~P(X)~~  $X \sim \text{Binom}(15, 0.2)$

$$P(X \leq 8) = B(8; 15, 0.2) = \sum_{x=0}^8 b(x; 15, 0.2)$$

$\approx 0.999$  → calculator

b.  $P(X = 8) = b(8; 15, 0.2) = \binom{15}{8} 0.2^8 0.8^7 \approx 0.0035$

c.  $P(X > 8) = 1 - P(X \leq 8) = 1 - P(X \leq 7)$   
 $= 1 - B(7; 15, 0.2) \approx 0.004.$

d.  $P(X \in \{4, 5, 6, 7\}) = P(X \leq 7) - P(X \leq 3)$   
 $= B(7; 15, 0.2) - B(3; 15, 0.2)$   
 $\approx 0.348.$

Rem: Learn how to do this with your calculator or with Table A.1 in Devore's book.