

# Expectations of Random Variables

$$\underline{X}: S \rightarrow \mathbb{R}$$

Discrete:  $S = \text{finite}$

$$f_{\underline{X}}(r) = \text{Prob}(\underline{X} = r)$$

~~||||~~  $E(\underline{X}) = \text{expectation of } X$

$$= \sum_{\substack{\text{all values} \\ r_i \text{ of } \underline{X}}} r_i f_{\underline{X}}(r_i)$$

Continuous Case:

$$f_{\underline{X}}: \mathbb{R} \rightarrow \mathbb{R} \quad \text{continuous}$$
$$\text{Prob}(a \leq \underline{X} \leq b) = \int_{t=a}^{t=b} f(t) dt$$

$$E(\underline{X}) = \int_{t=-\infty}^{t=\infty} t f(t) dt$$



Example: Fix  $n \geq 1$

$$S = \{ A = (A_1, \dots, A_n); A_i = 0 \text{ or } 1 \}$$

For each  $i \in \{1, \dots, n\}$  have

$$\underline{X}_i: S \rightarrow \{0, 1\} \subseteq \mathbb{R}$$

$$\underline{X}_i((A_1, \dots, A_n)) = A_i$$

$$f_{\underline{X}_i}: \mathbb{R} \rightarrow [0, 1]$$

$$f_{\underline{X}_i}(r) = \begin{cases} 0 & \text{if } r \notin \{0, 1\} \\ p_i & \text{if } r = 1 \\ 1 - p_i & \text{if } r = 0 \end{cases}$$

$$E(\underline{X}_i) = \sum_{r=0,1} r \cdot f_{\underline{X}_i}(r)$$

$$= 0 \cdot (1 - p_i) + 1 \cdot p_i = p_i$$



Have  $\underline{X}_i : S \rightarrow \{0, 1\}$   
"  $\{ \mathbf{a} = (a_1, \dots, a_n) ; a_i \in \{0, 1\} \}$

$\underline{X} = \underline{X}_1 + \dots + \underline{X}_n : S \rightarrow \{0, \dots, n\} \subseteq \mathbb{R}$

$\underline{X}((a_1, \dots, a_n)) = \# \{ i ; a_i = 1 \}$

$f_{\underline{X}}(k) = \text{Pr}(\underline{X} = k)$   
= Prob of getting exactly  
 $k$  "heads" = "ones".

Theorem: Given Random Variable

$\underline{X}_1, \dots, \underline{X}_n : S \rightarrow \mathbb{R}$

one has

$$E(\underline{X}_1 + \dots + \underline{X}_n) = E(\underline{X}_1) + \dots + E(\underline{X}_n)$$



Example:  $E(X_i) = p_i$

$$E(X_1 + \dots + X_n) = p_1 + \dots + p_n$$

coin flip example.

$$S = \{(a_1, \dots, a_n) ; a_i = 0 \text{ or } 1\}$$

Sketch of Proof (Discrete case)

Joint density function  $f$

$$X_1, \dots, X_n : S \rightarrow \mathbb{R}.$$

Discrete case  $f : \mathbb{R}^n \rightarrow [0, 1]$

so for all  $r_1, \dots, r_n \in \mathbb{R}$ ,

$$\text{Prob}(X_1 = r_1, X_2 = r_2, \dots, X_n = r_n)$$

$$= f(r_1, \dots, r_n)$$



Continuous case:

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

$$\text{Prob}(a_1 \leq \underline{X}_1 \leq b_1, \dots, a_n \leq \underline{X}_n \leq b_n)$$

$$= \int_{a_1}^{b_1} \dots \int_{a_n}^{b_n} f(t_1, \dots, t_n) dt_1 \dots dt_n$$

Important Fact:

Discrete case

$$f_{\underline{X}_1}(r_1) = \sum_{r_2, \dots, r_n \in \mathbb{R}} f(r_1, r_2, \dots, r_n)$$

$$= \sum_{r_2, \dots, r_n \in \mathbb{R}} \text{Prob}(\underline{X}_1 = r_1 \text{ and } \underline{X}_2 = r_2, \dots, \underline{X}_n = r_n)$$

$$= \text{Prob}(\underline{X}_1 = r_1)$$



Proof that if  $\underline{X} = \underline{X}_1 + \dots + \underline{X}_n$  then

$$E(\underline{X}) = E(\underline{X}_1) + \dots + E(\underline{X}_n)$$

$f = f_{\underline{X}}$  joint density for  $\underline{X}$ .

$$E(\underline{X}) = \sum_{r \in \mathbb{R}} \text{Prob}(\underline{X} = r) \cdot r$$

$$= \sum_{r_1, \dots, r_n \in \mathbb{R}} \text{Prob}(\underline{X}_1 = r_1, \dots, \underline{X}_n = r_n) \cdot (r_1 + \dots + r_n)$$

$$= \sum_{i=1}^n \sum_{r_i \in \mathbb{R}} r_i \cdot \sum_{\text{all choices of } r_j \text{ for } j \neq i} \text{Prob}(\underline{X}_1 = r_1, \dots, \underline{X}_n = r_n)$$

$$= \sum_{i=1}^n \left( \sum_{r_i \in \mathbb{R}} r_i f_{\underline{X}_i}(r_i) \right)$$

$$= \sum_{i=1}^n E(\underline{X}_i)$$



Ex:  $n$  Independent flips of a coin  
which has prob.  $p$  of  
coming up heads

$$S = \{(t_1, \dots, t_n); t_i = 0 \text{ or } 1\}.$$

$X_i$ :  $\downarrow$   $E(X_i) = p$  all  $i$   
 $\{0, 1\}$

$$E(\underbrace{X_1 + \dots + X_n}_X) = \sum_{i=1}^n E(X_i) = np$$

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

$$E(X) = \sum_{k=0}^n f_X(k) \cdot k = np$$

$$\sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k} \cdot k$$



## Connection To Epidemics

$S$  = all people who now  
would test positive for  
COVID-19.

Historical

$$\underline{X}: S \rightarrow \mathbb{Z}_{\geq 0} \subseteq \mathbb{R}$$

$\underline{X}(A)$  = how many other  
people are infected  
by person  $A$ .

Projected

$$\underline{X}: S \rightarrow \mathbb{R}_{\geq 0}$$

$\underline{X}(A)$  = estimated for the expected  
number of new infections  
resulting from  $A$ .



Important:

$$\overline{E(X)} =$$

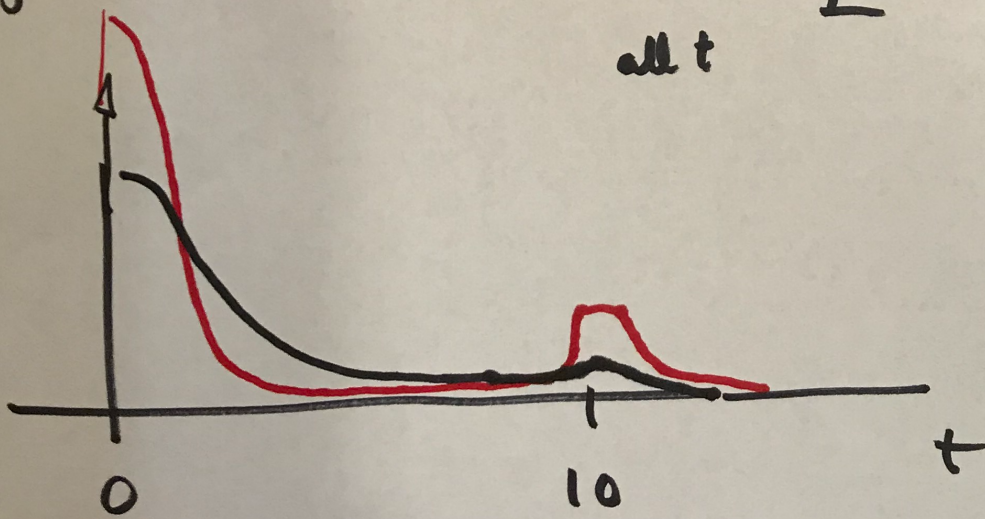
average (expected) number of new infections caused by each person who never has covid-19.

Projected case:  $\underline{X}: S \rightarrow \mathbb{R}_{\geq 0}$

$f_{\underline{X}}$  = density

$$\int_{\text{all } t} f_{\underline{X}}(t) dt = \underline{1}$$

$$y = f_{\underline{X}}(t)$$



$$E(\underline{X}) = \int_{t=0}^{\infty} t f_{\underline{X}}(t) dt$$

↓  
 $E(\underline{X})$

$t=0$