

## Review.

Let  $X$  be a discrete r.v.

The prob. mass function of  $X$ :

$$p(a) = P\{X=a\}, \quad a \in \mathbb{R}.$$

$$\bullet \sum_{x: p(x) > 0} p(x) = 1.$$

Expected Value.

$$E[X] = \sum_{x: p(x) > 0} x p(x).$$

Variance.

$$V(X) := E[(X-\mu)^2], \quad \text{where } \mu = E[X].$$

It describe how  $X$  is spread out from its mean.

Standard deviation:  $\sqrt{V(X)}$ .

$$\bullet \text{ Prop. } V(X) = E[X^2] - \mu^2, \quad \text{where } \mu = E[X].$$

$$\begin{aligned} \text{pf. } V(X) &= E[(X-\mu)^2] \\ &= \sum_{x: p(x) > 0} (x-\mu)^2 \cdot p(x) \end{aligned}$$

$$\begin{aligned}
&= \sum_{x: p(x) > 0} (x^2 - 2x\mu + \mu^2) p(x) \\
&= \sum_{x: p(x) > 0} x^2 p(x) - 2\mu \sum_{x: p(x) > 0} x p(x) \\
&\quad + \mu^2 \sum_{x: p(x) > 0} p(x) \\
&= E[X^2] - 2\mu^2 + \mu^2 \\
&= E[X^2] - \mu^2. \quad \square
\end{aligned}$$

#### § 4.6. Bernoulli r.v. and Binomial r.v.

##### (i) Bernoulli r.v.

Consider a random experiment, whose outcome can be classified as either a success, or a failure.

Define  $X = \begin{cases} 1 & \text{if the outcome is a success,} \\ 0 & \text{if the outcome is a failure.} \end{cases}$

Let  $p = P\{X=1\}$ , then  $P\{X=0\} = 1-p$ .

It has a prob. mass function:  $\begin{cases} p(1) = p, \\ p(0) = 1-p. \end{cases}$

We call  $X$  a Bernoulli r.v.

- $E[X] = p$

$$E[X^2] = p$$

$$V(X) = E[X^2] - E[X]^2 = p - p^2$$

(2) Binomial r.v.

Consider  $n$  independent trials, each of them results in either a success with prob  $p$ , or a failure with prob.  $(1-p)$ .

Let  $X =$  the number of the successes that appear in the  $n$ -trials.

We call  $X$  a Binomial r.v. with parameters  $(n, p)$ .

- Example:  $n=2$ .

possible outcomes  $\{(S, S), (S, F), (F, S), (F, F)\}$

$$P\{(S, S)\} = P(E_1 E_2) = P(E_1) P(E_2) = p \cdot p$$

where  $E_1$  is the event that the outcome of the first trial is  $S$

and  $E_2$  is the event that the outcome of the second trial is  $S$ .

Similarly  $P\{(S, F)\} = P\{(F, S)\} = p(1-p)$ ,  $P\{(F, F)\} = (1-p)^2$ .

Hence,

$$P\{X=1\} = P\{(S,F), (F,S)\} = 2 \cdot p(1-p).$$

- Prob. mass function for a general Binomial r.v. with parameters  $(n, p)$   
For  $i=0, 1, \dots, n$ , we have

$$P\{X=i\} = \binom{n}{i} \cdot p^i (1-p)^{n-i}$$

Reason: The prob. of a special sequence of outcomes containing  $i$  successes and  $(n-i)$  failures, is equal to  $p^i (1-p)^{n-i}$

But there are in total  $\binom{n}{i}$  such sequences, so

$$P\{X=i\} = \binom{n}{i} p^i (1-p)^{n-i}$$

$$\text{(Recall } \binom{n}{i} = \frac{n!}{i!(n-i)!} = \frac{n(n-1)\dots(n-i+1)}{i(i-1)\dots 1}$$

and

$$(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^i y^{n-i} \quad \text{(Binomial formula)}$$

Prop. Let  $X$  be a Binomial r.v. with parameters  $(n, p)$ .

Let  $k \geq 1$  be an integer. Then

$$E[X^k] = np \cdot E[(Y+1)^{k-1}]$$

where  $Y$  is a Binomial r.v. with parameters  $(n-1, p)$ .

Pf. By def,

$$E[X^k] = \sum_{i=0}^n i^k \cdot \binom{n}{i} p^i (1-p)^{n-i}$$

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

$$\text{(Using } i \binom{n}{i} = n \binom{n-1}{i-1} \text{)}$$

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

$$= np \sum_{i=1}^n i^{k-1} \binom{n-1}{i-1} p^{i-1} (1-p)^{n-i}$$

Letting  $j = i-1$

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

$$= np \cdot E[(Y+1)^{k-1}]$$

□

$$\text{Cor. } E[X] = np \cdot E[(Y+1)^0] = np.$$

$$E[X^2] = np \cdot E[(Y+1)]$$

$$= np (E[Y] + 1)$$

$$= np ((n-1)p + 1)$$

$$\text{Var}(X) = E[X^2] - E[X]^2$$

$$= np((n-1)p + 1) - (np)^2$$

$$= n(p - p^2).$$