

Math 3235

Probability

Theory

09/15/22

Expectation of a r.v.

$X$

$p_X(x)$

p. m. f.

$$E(X) = \sum_x x p_X(x)$$

Function of a r.v.

$X$  is a r.v.

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

Then

$Y = f(X)$  is a r.v. itself.

Since  $X$  is discrete, we need

no conditions on  $f$ .

$$P_Y(y) = \sum_{x \mid f(x)=y} P_X(x) = \sum_{x \in f^{-1}(y)} P_X(x)$$

$$x \mid f(x)=y \iff x \in f^{-1}(y)$$

Th: If  $Y = f(X)$  where  $X$  is a discrete r.v. Then

$$\mathbb{E}(f(X)) = \sum_x f(x) p_X(x)$$

Proof

$$\begin{aligned} \mathbb{E}(Y) &= \sum_y y p_Y(y) = \\ &= \sum_y y \sum_{x \in f^{-1}(y)} p_X(x) = \\ &= \sum_y \sum_{x \in f^{-1}(y)} f(x) p_X(x) = \\ &= \sum_x f(x) p_X(x). \quad \text{Q.E.D.} \end{aligned}$$

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Poisson r.v.  $X$  par.  $\lambda$

$$Y = e^X$$

$$\mathbb{E}(Y) = \sum_x e^x \frac{\lambda^x}{x!} e^{-\lambda} =$$

$$= \sum_x \frac{(e\lambda)^x}{x!} e^{-\lambda} = e^{\lambda} (e - 1)$$

Th:  $X$  is a r.v.

$$\text{a) if } \mathbb{P}(X \geq 0) = 1 \implies \mathbb{P}(X=0) = 1 \\ \mathbb{E}(X) = 0$$

$$\text{b) } \mathbb{E}(aX + b) = a\mathbb{E}(X) + b$$

Proof

$$\text{b) } \mathbb{E}(aX + b) = \sum_x (ax + b) p(x) = \\ = a \sum_x x p(x) + b \sum_x p(x) = \\ = a\mathbb{E}(X) + b$$

$$\text{a) if } \mathbb{P}(X \geq 0) = 1 \implies$$

$$x \in \text{Im}(X) \implies x \geq 0$$

$$\mathbb{E}(X) = 0 \implies x \mathbb{P}(X=x) = 0 \\ \forall x > 0$$

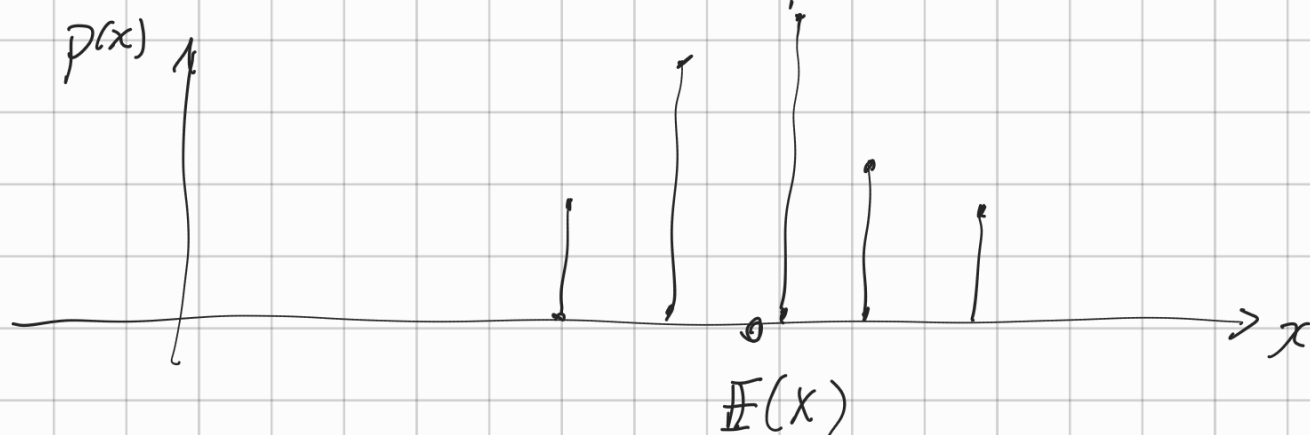
$$\mathbb{P}(X=0) = 1$$

# Variance

$$V(X) = \text{var}(X) =$$

$$\mathbb{E}((X - \mathbb{E}(X))^2)$$

Expected value: measure of position.



Measure of Variability.

How distant from the Exp. val.

is a "normal" value of the r.v.,

in average?

$$\mathbb{E}(X - \mathbb{E}(X)) =$$

$$\mathbb{E}(X) - \mathbb{E}(X) = 0$$

Not very useful !!

$$\mathbb{E} \left( (X - \mathbb{E}(X))^2 \right) = \text{var}(X)$$

$$\sigma_X = \sqrt{\text{var}(X)}$$

standard deviation.

$$\text{var}(X) \geq 0$$

$$\text{var}(X) = 0 \Rightarrow \mathbb{P}(X - \mathbb{E}(X) = 0) = 1$$

$X = \mathbb{E}(X)$  with prob. 1.

This means that  $X$  takes a unique value with prob. 1.

How To compute variances?

$$\text{var}(X) = \sum_x (x - \mathbb{E}(X))^2 p(x) =$$

$$\sum_x (x^2 - 2\mu x + \mu^2) p(x) =$$

where we set  $\mathbb{E}(X) = \mu$ .

$$\sum_x x^2 p(x) - 2\mu \sum_x x p(x) + \mu^2 \sum_x p(x) =$$

$$\mathbb{E}(X^2) - 2\mu \mathbb{E}(X) + \mu^2$$

$$\text{var}(X) = \mathbb{E}(X^2) - \mathbb{E}(X)^2$$

since  $\text{var}(X) \geq 0$

$$\mathbb{E}(X^2) \geq \mathbb{E}(X)^2$$

$$\text{var}(aX + b) = ?$$

$$\begin{aligned}\text{var}(aX) &= \mathbb{E}(a^2 X^2) - \mathbb{E}(aX)^2 \\ &= a^2 \text{var}(X)\end{aligned}$$

$$\text{var}(X + b) = \mathbb{E}\left(\left(X + b - \mathbb{E}(X + b)\right)^2\right)$$

$$\text{but } \mathbb{E}(X + b) = \mathbb{E}(X) + b$$

$$\text{var}(X + b) = \text{var}(X)$$

$$\text{var}(aX + b) = a^2 \text{var}(X)$$

$$Y = aX + b$$

$$\sigma_Y = |a| \sigma_X$$

Example

$$\sigma_{-X} = \sigma_X$$

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Examples.

Bernoulli  $p$

$$\mathbb{E}(X) = p$$

$$\text{var}(X) = p - p^2 = pq$$

$\text{var}(X)$  is minimal for  $p = 0$  or  $p = 1$

$\text{var}(X)$  is maximal for  $p = 0.5$

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Binomial r.v. with par  $N, p$

$$\mathbb{E}(X^2) = \mathbb{E}(X(X-1)) + \mathbb{E}(X)$$

$$\begin{aligned} \text{var}(X) &= \mathbb{E}(X(X-1)) + \mathbb{E}(X) - \mathbb{E}(X)^2 \\ &= \mathbb{E}(X(X-1)) - \mathbb{E}(X)(\mathbb{E}(X)-1) \end{aligned}$$

$$\mathbb{E}(X(X-1)) = \sum_x x(x-1) \binom{N}{x} p^x q^{N-x}$$

$$\begin{aligned}
 x(x-1) \binom{N}{x} &= \frac{N!}{x!(N-x)!} x(x-1) = \\
 &= \frac{N!}{(x-2)!(N-x)!} = \\
 &= N(N-1) \frac{(N-2)!}{(x-2)!(N-x)} = \\
 &= N(N-1) \binom{N-2}{x-2}
 \end{aligned}$$

Thus

$$\sum_{x=0}^N x(x-1) \binom{N}{x} p^x q^{N-x} =$$

$$N(N-1) \sum_{x=2}^N \binom{N-2}{x-2} p^x q^{N-x} =$$

$$\Rightarrow N(N-1) p^2 \sum_{z=0}^{N-2} \binom{N-2}{z} p^z q^{N-2-z} = N(N-1) p^2$$

$$\begin{aligned}
 \text{var}(X) &= N(N-1) p^2 \sim N p (N p - 1) = \\
 &= N p (1 - p)
 \end{aligned}$$



$X$  is Poisson par  $\lambda$

$$\begin{aligned} E(X(X-1)) &= \sum_{x \geq 0} x(x-1) \frac{\lambda^x}{x!} e^{-\lambda} = \\ &= \lambda^2 \sum_{x \geq 2} \frac{\lambda^{x-2}}{(x-2)!} e^{-\lambda} = \\ &= \lambda^2 \end{aligned}$$

$$\begin{aligned} \text{var}(X) &= E(X(X-1)) + E(X) - E(X)^2 = \\ &= \lambda \end{aligned}$$

Other way

$$\begin{array}{ccc} N p (1-p) & \text{with} & p = \frac{\lambda}{N} \\ \downarrow & & \\ \lambda \left(1 - \frac{\lambda}{N}\right) & \xrightarrow{N \rightarrow \infty} & \lambda \end{array}$$

$X$  geometric  $p$

$$E(X) = \frac{1}{p}$$

$$\begin{aligned} E(X(X-1)) &= pq \sum_{x \geq 0} x(x-1) q^{x-2} = \\ &= pq \frac{d^2}{dq^2} \left( \sum_{x \geq 2} q^x \right) = \\ &= \frac{2q}{(1-q)^3} = \frac{2q}{p^3} \end{aligned}$$

$$\text{var}(X) = q p^{-2}$$