

Lecture 16:

Expectation and Variance - Part III

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Properties of variance

► $\text{Var}(X) \geq 0$. $\text{Var}(X) = E[(X - \mu_X)^2]$

$$= \sum_x (x - \mu_X)^2 p_X(x)$$
$$\geq 0 \quad \left(\because \underbrace{(x - \mu_X)^2}_{\text{square of a real no.}} \geq 0, \underbrace{p_X(x)}_{\text{by axioms of probability}} \geq 0 \text{ for all } x \right)$$

► $\text{Var}(c) = 0$ for any constant c .

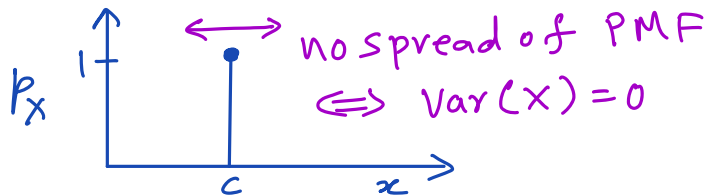
Let X be an r.v. s.t. $\underbrace{P(X=c)=1}_{\downarrow} \Rightarrow E(X) = \sum_x x P(X=x)$

$$= c \cdot P(X=c)$$
$$= c.$$

Then, for r.v. $(X-c)^2$, $\underbrace{P((X-c)^2=0)=1}_{\downarrow}$

$$\Rightarrow E[(X-c)^2] = 0 \cdot 1 = 0$$

Note: $\text{Var}(c)=0$ means that when the r.v. X is a constant c (i.e., $P(X=c)=1$) then the spread of PMF is zero:



Properties of variance

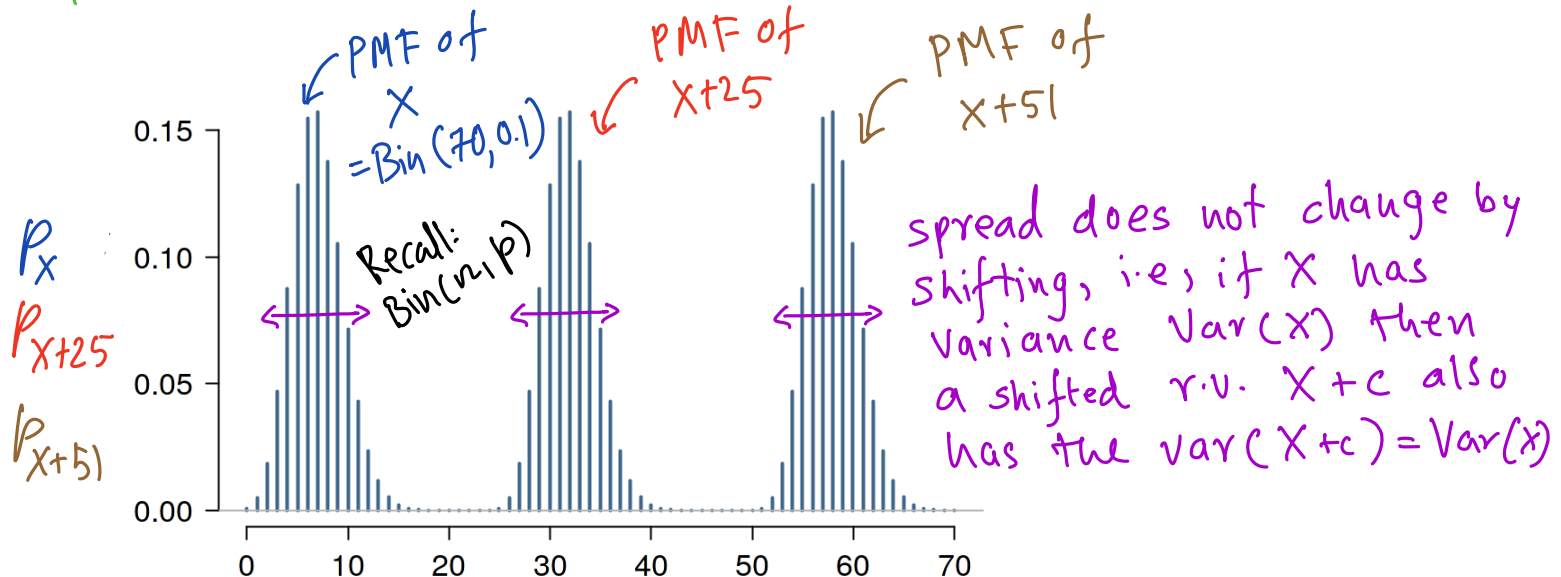
shifting X by c

- $\text{Var}(X + c) = \text{Var}(X)$ for any constant c .

$$E(X + c) = \mu_X + c.$$

$$\begin{aligned}\text{Var}(X + c) &= E[(X + c - E(X + c))^2] \\ &= E[(X + c - \mu_X - c)^2] = E[(X - \mu_X)^2] \\ &= \text{Var}(X).\end{aligned}$$

- Note: spread does not change by "shifting".



Properties of variance

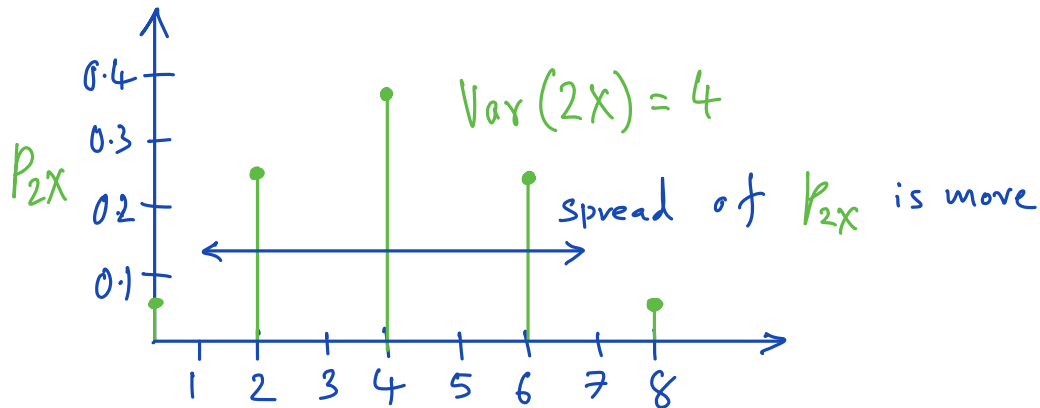
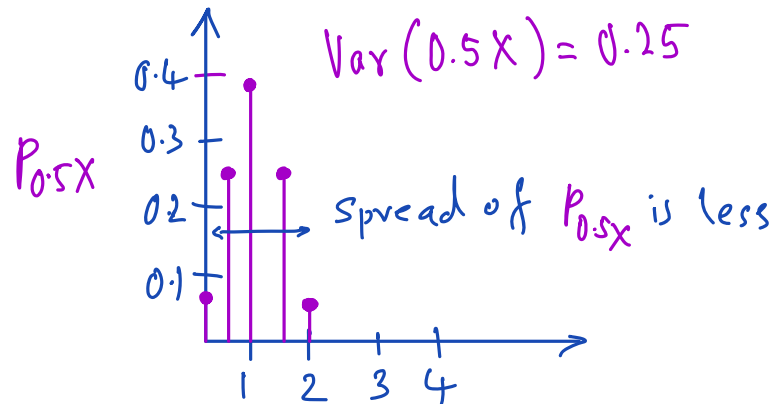
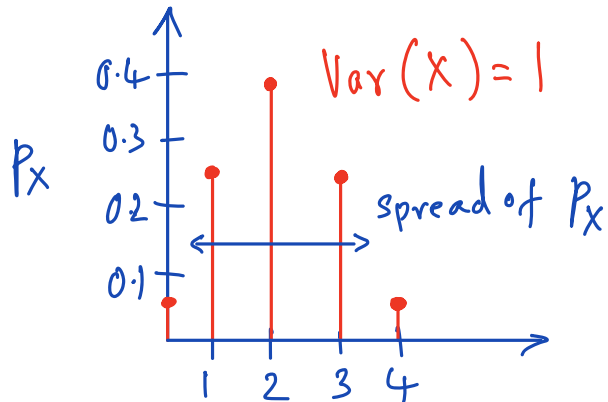
- $\text{Var}(cX) = c^2 \text{Var}(X)$ for any constant c .

$$\begin{aligned}\text{Var}(cX) &= E[(cX - E(cX))^2] \\ &= E[c^2(X - \mu_X)^2] \\ &= c^2 E[(X - \mu_X)^2] \quad \because \text{linearity of } E(\cdot). \\ &= c^2 \text{Var}(X).\end{aligned}$$

- That is, if we scale X by c then the variance $\text{Var}(cX)$ of the resulting r.v. cX is $\text{Var}(X)$ scaled by c^2 .
- So, depending on the value of c , the spread for cX either reduces or increases (if $c=1$ then the spread remains the same.)
- The next example demonstrates this.

Properties of variance

- **Example:** Let $X \sim \text{Bin}(4, .5)$. Plot PMFs of X , $2X$, $.5X$ and compare their variance.



Functions of independent random variables

- **Theorem:** Suppose that X and Y are independent r.v.s. Then $W = g(X)$ and $Z = h(Y)$ are also independent.

Proof:

$$P_{W,Z}(\omega, z) = P(W = \omega, Z = z) = P\left(\left[\bigcup_{x: g(x) = \omega} \{X = x\}\right] \cap \left[\bigcup_{y: h(y) = z} \{Y = y\}\right]\right)$$
$$= P\left(\bigcup_{\substack{x: g(x) = \omega \\ y: h(y) = z}} \{X = x, Y = y\}\right)$$

Note: You may skip the proof (will not be asked in exam). But remember the theorem statement.

The proof is just to satisfy your curiosity.

$$= \sum_{x: g(x) = \omega} \sum_{y: h(y) = z} P_{X,Y}(x, y)$$

\Downarrow since $P_{X,Y}(x, y) = P_X(x) \cdot P_Y(y)$

$$= \sum_{x: g(x) = \omega} P_X(x) \sum_{y: h(y) = z} P_Y(y) = P_W(\omega) \cdot P_Z(z).$$

Properties of variance

- If X and Y are independent, then

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y).$$

$$\begin{aligned}\text{Var}(X+Y) &= E[(X+Y - E(X+Y))^2] \\&= E[(X+Y - \mu_X - \mu_Y)^2] \quad \because \text{linearity of } E(\cdot) \\&= E[(X - \mu_X)^2 + (Y - \mu_Y)^2 + 2(X - \mu_X)(Y - \mu_Y)] \\&= \text{Var}(X) + \text{Var}(Y) + 2E[(X - \mu_X)(Y - \mu_Y)] \\&\quad \Downarrow \text{functions of independent r.v.s are independent} \\&= \text{Var}(X) + \text{Var}(Y) + 2E(X - \mu_X) \cdot E(Y - \mu_Y) \\&= \text{Var}(X) + \text{Var}(Y) + 2\underbrace{(\mu_X - \mu_X)}_0 \cdot \underbrace{(\mu_Y - \mu_Y)}_0 \\&= \text{Var}(X) + \text{Var}(Y)\end{aligned}$$

Example

► Let X be a random variable with the PMF

$$p_X(-3) = 1/6, p_X(6) = 1/2, p_X(9) = 1/3.$$

(a) Find $\mu_{g(X)} = E(g(X))$, where $g(X) = (2X + 1)^2$.

(b) Find $SD(g(X))$.

$$\begin{aligned} (a) \quad \mu_{g(X)} &= E[(2X+1)^2] \\ &= \frac{25}{6} + \frac{169}{2} + \frac{361}{3} = 209. \end{aligned}$$

$$\begin{aligned} (b) \quad \sigma_{g(X)}^2 &= E[(2X+1)^2 - 209]^2 \\ &= \sum_x [(2x+1)^2 - 209] \cdot p_X(x) \\ &= (25 - 209)^2 \cdot \frac{1}{6} + (169 - 209)^2 \cdot \frac{1}{2} + (361 - 209)^2 \cdot \frac{1}{3} \\ &= 14144 \Rightarrow SD(g(X)) = \sqrt{\sigma_{g(X)}^2} = 118.9. \end{aligned}$$