

1)

Definition of Expectation

$$E[X] = \sum_{i=1}^N x_i p_i \text{ where } p_i \text{ is } P(x_i)$$

Assuming equal probability for every  $x_i$ 

$$E[X] = (1/N) \sum_{i=1}^N x_i \\ = \mu$$

Definition of Variance

$$\begin{aligned} \text{Var}(X) &= E[(X - \mu)^2] \\ &= E[(X - E[X])^2] \text{ using the above} \\ &= E[X^2 - 2XE[X] + E^2[X]] \\ &= E[X^2] - 2E[X]E[X] + E^2[X] \\ &\quad \uparrow \text{By the commutative property} \\ &= E[X^2] - 2E^2[X] + E^2[X] \end{aligned}$$

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



We know that

$$E_{x|y}(x) = \int_x f(x) p(x|y) dx$$

Wrapping the  $E_y$  term around this expression

$$E_y(E_{x|y}(x)) = \int_y \int_x f(x) p(x|y) dx p(y) dy$$

Since  $p(y)$  is constant with respect to  $x$  we can move it inside the integral with respect to  $x$ .

$$\begin{aligned} E_y(E_{x|y}(x)) &= \int_y \int_x f(x) p(x|y) p(y) dx dy \\ &= \int_x \int_y f(x) p(x|y) p(y) dy dx \end{aligned}$$

We now see that that the integration over  $y$  marginalizes the conditional probabilities.

$$E_y(E_{x|y}(x)) = \int_x f(x) p(x) dx$$

Therefore, by definition

$$E_y(E_{x|y}(x)) = E(x)$$



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For a pdf to be proper, its integral over its defined range must be equal to one.

$$\int p(x) dx = 1$$

a) Lets see if this is true:

$$\begin{aligned} p(x) &= 3x^2|_0^1 \\ \int_0^1 p(x) dx &= \int_0^1 3x^2 dx \\ &= [x^3]_0^1 \end{aligned}$$

$$\boxed{\int p(x) dx = 1} \quad \text{so this is a proper pdf.}$$

if  $Y = X^2$  then  $p(y) = p^2(x)$

The difference is where the bounds on  $y$  will be

$$\begin{aligned} \int_a^b 9x^4 dx &= 1 \\ \left[ \frac{9}{5} x^5 \right]_a^b &= 1 \end{aligned}$$

$$\boxed{b^5 - a^5 = 5/9, \quad 0 \leq a < b}$$

Anywhere this holds true but for a range from  $0 \rightarrow$  we let  $a = 0$ .

$$b^5 = \frac{5}{9}, \quad b = \sqrt[5]{5/9}$$

$$\text{So: } \boxed{P(y) = P(x^2) = 9x^4, \quad [0, \sqrt[5]{5/9}]}$$

b)