

Sec 3.3

Math 320

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Exercise. 3.14

(i)

$$f_X(\alpha) = \begin{cases} \frac{1}{2} & \text{if } \alpha = a, b, c \\ \frac{1}{6} & \text{if } \alpha = d \\ \frac{1}{3} & \text{if } \alpha = e, f \\ 0 & \text{elsewhere} \end{cases}$$

(ii)

$$\frac{1}{2} + \frac{1}{6} \cdot 2 + \frac{1}{3} \cdot 3.5 = 2$$

(iii)

$$\text{var}(x) = (1-2)^2 \cdot \frac{1}{2} + (0)^2 \cdot \frac{1}{6} + (1.5)^2 \cdot \frac{1}{3} = \frac{5}{4}$$

Exercise. 3.15

Let X be a discrete set, and have probability distribution $p(x = i) = g_X$. Suppose h is a continuous function s.t. $h : \mathbb{R} \rightarrow \mathbb{R}$, and $\exists \epsilon > 0$ between each point in X where there are countable number of points. Thus, because h is continuous, $\exists \epsilon' > 0$ between each point. Let $A = \{\epsilon'\}_{i \in I}$. Choosing the $\inf(A)$ will yield an ϵ' , such that for every point p , $\exists E(p, \frac{\epsilon'}{2})$, such that the neighborhood E contains no other points in X . Thus, $h(X)$ is a random variable, and by LOTUS,

$$E[h(X)] = \sum_{i \in I} h(i)p(x = i) = \sum_{i \in I} h(i)g_X(i)$$

Exercise. 3.16

$$\begin{aligned} \text{var}[X] &= E((X - \mu)^2) = E(X^2 - 2X\mu + \mu^2) \\ &= E(X^2) - 2\mu E(X) + E(\mu^2) \\ &= E(X^2) - \mu^2 \end{aligned}$$

Exercise. 3.17

$$\begin{aligned}
\text{var}(\alpha x + \beta y) &= E(\alpha x + \beta y)^2 - (E(\alpha x + \beta y))^2 \\
&= E(\alpha^2 x^2 + \alpha x \beta y + \alpha \beta x y + \beta^2 y^2) - (\alpha E(x) + \beta E(y))^2 \\
&= \alpha^2 E(x^2) + 2\alpha\beta E(xy) + \beta^2 E(y^2) - \alpha E(x)^2 - 2\alpha\beta E(x)E(y) - \beta^2 E(y)^2 \\
&= \alpha^2 (E(x^2) - E(x)^2) + 2\alpha\beta (E(xy) - E(x)E(y)) + \beta^2 (E(y^2) - E(y)^2) \\
&= \alpha^2 \text{var}[x] + 2\alpha\beta (E(xy) - E(x)E(y)) + \beta^2 \text{var}[y]
\end{aligned}$$

As desired, and if x, y are independent, we know that $E(xy) - E(x)E(y) = 0$, and thus

$$\text{var}(\alpha x + \beta y) = \alpha^2 \text{var}[x] + \beta^2 \text{var}[y]$$

Exercise. 3.18

Note, by LOTUS $E\left(\frac{1}{X+1}\right)$ is a random variable.

$$\begin{aligned}
E\left(\frac{1}{X+1}\right) &= \sum_{i=0}^n \frac{1}{i+1} \binom{n}{i} p^i (1-p)^{n-i} \\
&= \sum_{i=0}^n \frac{n!}{(i+1)!(n-i)!} p^i (1-p)^{n-i} \\
&= \sum_{i=1}^{n+1} \frac{n!}{i!(n-i+1)!} p^{i-1} (1-p)^{n-i+1} \\
&= \frac{1}{p(n+1)} \sum_{i=1}^{n+1} \frac{(n+1)!}{i!(n+1-i)!} p^i (1-p)^{n+1-i} \\
&= \frac{1}{p(n+1)} \sum_{i=1}^{n+1} \left(\binom{n}{i} + 1, i \right) p^i (1-p)^{n+1-i} \\
&= \frac{1}{p(n+1)} \left(\sum_{i=0}^{n+1} \binom{n}{i} + 1, i \right) p^i (1-p)^{n+1-i} - (1-p)^{n+1} \\
&= \frac{1}{p(n+1)} \cdot (1 - (1-p)^{n+1}) \\
&= \frac{1 - (1-p)^{n+1}}{p(n+1)}
\end{aligned}$$

Exercise. 3.19

$$\begin{aligned}
E(X) &= \sum_{x \in X} x \cdot P(X = x) \\
&= \sum_{i \in I} x \sum_{x \in X} \frac{x P(X = x) \cap B_i P(B_i)}{P(B_i)} \\
&= \sum_i \sum_x x P(X = x | B_i) P(B_i) \\
&= \sum_i E(X | B_i) P(B_i)
\end{aligned}$$