

Teoria da Informação - Lista 01B

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Review questions

1. (a) O valor esperado é a soma produto da probabilidade e o valor da variável aleatória.

$$E(X) = \sum_{s \in S} p(s)X(s)$$

É uma representação análoga ao centro de massa.

- (b)
- $\{(1, 1)\}$
 $X = 1 \Rightarrow p(X = 1) = \frac{1}{36}$
 - $\{(2, 1), (1, 2), (2, 2)\}$
 $X = 2 \Rightarrow p(X = 2) = \frac{3}{36}$
 - $\{(3, 1), (3, 2), (1, 3), (2, 3), (3, 3)\}$
 $X = 3 \Rightarrow p(X = 3) = \frac{5}{36}$
 - $\{(k, 1), (k, 2), \dots, (k, k)\}$
 $X = k \Rightarrow p(X = k) = \frac{2k-1}{36}$

Portanto o valor esperado é:

$$E(X) = \sum_{k=1} \frac{2k-1}{36} \cdot k = 1 \cdot \frac{1}{36} + \dots + 6 \cdot \frac{11}{36} = \frac{161}{36} \approx 4.47$$

2. (a) O significado do ensaio de Bernoulli é a probabilidade ter k sucessos em n tentativas.
- (b)

$$p(X = k) = C(n, k)p^k q^{n-k}$$

(c) Mostrando uma equivalência necessária para o desenvolvimento.

$$\begin{aligned} C(n, k) &= \frac{n!}{(n-k)!k!} = \frac{n}{k} \cdot \frac{(n-1)!}{(n-k+1-1)(k-1)!} = \\ &= \frac{n}{k} \cdot \frac{(n-1)!}{((n-1)-(k-1))(k-1)!} = \frac{n}{k} \cdot C(n-1, k-1) \Rightarrow \\ C(n, k) &= \frac{n}{k} \cdot C(n-1, k-1) \Rightarrow k \cdot C(n, k) = n \cdot C(n-1, k-1) \end{aligned}$$

Lembrando que $q = 1-p$. Seja n o número de ensaios e k o número de sucessos. Lembrando que $p(X = k) = C(n, k) = p^k q^{n-k}$. Então:

$$\begin{aligned} E(X) &= \sum_{k=1}^n X(k)p(X = k) = \\ &= \sum_{k=1}^n k \cdot C(n, k)p^k q^{n-k} = \sum_{k=1}^n n \cdot C(n-1, k-1)p^k q^{n-k} = \\ &= n \sum_{k=1}^n C(n-1, k-1)p^k q^{n-k} \underbrace{=}_{j=k-1} n \sum_{j=0}^{n-1} C(n-1, j)p^{j+1} q^{n-j+1} = \\ &= np \underbrace{\sum_{j=0}^{n-1} C(n-1, j)p^j q^{n-j+1}}_{(p+q)^{n-1}} = np(p+q)^{n-1} = np(p+(1-p))^{n-1} = \\ &= np(1)^{n-1} = np \end{aligned}$$

3. (a) Linearidade da esperança é, dada duas variáveis aleatórias X e Y :

$$E(X + Y) = E(X) + E(Y)$$

4.

$$\begin{aligned} p(F|E) &= \frac{p(E|F)p(F)}{p(E|F)p(F)+p(E|\bar{F})p(\bar{F})} = \\ &= \frac{\frac{1}{3} \cdot \frac{2}{3}}{\frac{1}{3} \cdot \frac{2}{3} + \frac{1}{4} \cdot (1 - \frac{2}{3})} = \frac{\frac{1}{3} \cdot \frac{2}{3}}{\frac{1}{3} \cdot \frac{2}{3} + \frac{1}{4} \cdot \frac{1}{3}} = \frac{8}{11} \end{aligned}$$

5. Seja X uma variável aleatória no espaço S . Variância de X é:

$$V(X) = \sum_{s \in S} (X(s) - E(X))^2 p(s)$$

$$\begin{aligned} V(X) &= \sum_{s \in S} (X(s) - E(X))^2 p(s) = \\ &= \sum_{s \in S} (X(s)^2 - 2X(s)E(X) + E(X)^2) p(s) \end{aligned}$$

Utilizando a linearidade.

$$\begin{aligned}
& \sum_{s \in S} X(s)^2 p(s) - 2 \sum_{s \in S} X(s) E(X) p(s) + \sum_{s \in S} E(X)^2 p(s) = \\
& \underbrace{\sum_{s \in S} (X(s))^2 p(s)}_{E(X^2)} - 2 \underbrace{E(X) \sum_{s \in S} X(s) p(s)}_{E(X)} + \underbrace{E(X)^2 \sum_{s \in S} p(s)}_1 = \\
& E(X^2) - 2(E(X))^2 + (E(X))^2 = E(X^2) - E(X)^2 \Rightarrow \\
& V(X) = E(X^2) - E(X)^2
\end{aligned}$$

6. (a)

$$\begin{aligned}
& V\left(\sum_{i=1}^n X_i\right) \\
&= E\left(\left(\sum_{i=1}^n X_i\right)^2\right) - E\left(\sum_{i=1}^n X_i\right)^2 \\
&= V\left(X_1 + \sum_{i=2}^n X_i\right) \\
&= E\left(\left(X_1 + \sum_{i=2}^n X_i\right)^2\right) - E\left(X_1 + \sum_{i=2}^n X_i\right)^2 \\
&= E\left(X_1^2 + 2X_1 \sum_{i=2}^n X_i + \left(\sum_{i=2}^n X_i\right)^2\right) \\
&\quad - \left(E(X_1) + E\left(\sum_{i=2}^n X_i\right)\right)^2 \\
&= E(X_1^2) + \underbrace{E\left(2X_1 \sum_{i=2}^n X_i\right)}_{2E(X_1)E\left(\sum_{i=2}^n X_i\right)} + E\left(\left(\sum_{i=2}^n X_i\right)^2\right) \\
&\quad - \left(E(X_1)^2 + 2E(X_1)E\left(\sum_{i=2}^n X_i\right) + \left(E\left(\sum_{i=2}^n X_i\right)\right)^2\right) \\
&= E(X_1^2) + 2E(X_1)E\left(\sum_{i=2}^n X_i\right) + E\left(\left(\sum_{i=2}^n X_i\right)^2\right) \\
&\quad - E(X_1)^2 - 2E(X_1)E\left(\sum_{i=2}^n X_i\right) - \left(E\left(\sum_{i=2}^n X_i\right)\right)^2 \\
&= E(X_1^2) + E\left(\left(\sum_{i=2}^n X_i\right)^2\right) \\
&\quad - E(X_1)^2 - \left(E\left(\sum_{i=2}^n X_i\right)\right)^2 \\
&= \underbrace{E(X_1^2) - E(X_1)^2}_{V(X_1)} \\
&\quad + \underbrace{E\left(\left(\sum_{i=2}^n X_i\right)^2\right) - \left(E\left(\sum_{i=2}^n X_i\right)\right)^2}_{V\left(\sum_{i=2}^n X_i\right)}
\end{aligned}$$

Se refizermos a operação para as somas restantes $\sum_{i=2}^n X_i$, teremos:

$$V\left(\sum_{i=1}^n X_i\right) = V(X_1) + V(X_2) + \dots + V(X_n) = \sum_{i=1}^n V(X_i)$$

Exercises

7. Caixa 1: 2 brancas e 3 azuis.

Caixa 2: 4 brancas e 1 azuis.

B : pegar bola branca

A : pegar bola azul

X_1 : pegar na caixa 1

X_2 : pegar na caixa 2

$$p(X_1|A) = \frac{p(A|X_1) \cdot p(X_1)}{p(A|X_1)p(X_1) + p(A|X_2)p(X_2)} = \frac{\frac{3}{5} \cdot \frac{1}{2}}{\frac{3}{5} \cdot \frac{1}{2} + \frac{1}{5} \cdot \frac{1}{2}} = \frac{3}{3+1} = \frac{3}{4}$$

8. De princípio, definimos:

- H : HIV positivo
- \bar{H} : HIV negativo
- T : teste indica HIV positivo
- \bar{T} : teste indica HIV negativo

Pelo enunciado, temos:

$$\begin{aligned}p(H) &= 0.08 \\p(\bar{H}) &= 1 - p(H) = 1 - 0.08 = 0.92 \\p(T|H) &= 0.98 \\p(\bar{T}|H) &= 1 - p(T|H) = 0.02 \\p(T|\bar{H}) &= 0.03 \\p(\bar{T}|\bar{H}) &= 1 - p(T|\bar{H}) = 0.97\end{aligned}$$

(a) Utilizando o Teorema de Bayes.

$$\begin{aligned}p(H|T) &= \frac{p(T|H)p(H)}{p(T|H)p(H) + p(T|\bar{H})p(\bar{H})} \\&= \frac{0.98 \cdot 0.08}{0.98 \cdot 0.08 + 0.03 \cdot 0.92} \approx 0.739\end{aligned}$$

(b)

$$p(\overline{H}|T) = 1 - p(H|T) \approx 1 - 0.739 = 0.261$$

(c) Utilizando o Teorema de Bayes.

$$\begin{aligned} p(H|\overline{T}) &= \frac{p(\overline{T}|H)p(H)}{p(\overline{T}|H)p(H) + p(\overline{T}|\overline{H})p(\overline{H})} \\ &= \frac{0.02 \cdot 0.08}{0.02 \cdot 0.08 + 0.97 \cdot 0.92} \approx 0.002 \end{aligned}$$

(d)

$$p(\overline{H}|\overline{T}) = 1 - p(H|\overline{T}) \approx 1 - 0.002 = 0.998$$