信息论作业1

史泽宇

2020年2月26日

题目 1

- 1. 两枚骰子总点数之和为 7 的情况有 $\{(1,6),(2,5),(3,4),(4,3),(5,2),(6,1)\}$ 6 种情况,总共可能产生 $6\times 6=36$ 种情况。所以投掷总点数之和为 7 的概率为 $\frac{18}{18}$,其自信息为 $\log_2\frac{18}{3}\approx 2.58bit$ 。
- 2. 两枚骰子总点数之和为 12 的情况有 $\{(1,1)\}$ 1 种情况。所以投掷总点数之和为 12 的概率为 $\frac{1}{36}$,其自信息为 $\log_2 36 \approx 5.17 bit$ 。

题目 2 本题目如果直接计算多个离散概率的互信息,脑补了一下计算复杂度,可能会比较高,查阅了部分资料后 [1],得到如下结论(论文上说 Elements of Information Theory 上面有,但是我没找到):

Corollary 1 Let $Y = X_1 + X_2$, where X_1 and X_2 are independent, then $I(X_1; Y) = H(Y) - H(X_2)$

$$I(X_1; X_1 + X_2) = H(X_1 + X_2) - H(X_1 + X_2 | X_1)$$
(1)

$$= H(X_1 + X_2) - H(X_2|X_1) \tag{2}$$

$$= H(X_1 + X_2) - H(X_2) \tag{3}$$

$$=H(Y)-H(X_2) \tag{4}$$

其中 (1) 到 (2) 的步骤不是很明白,想请老师答疑解惑一下。本题目的解答中很大程度上依靠了 (1) 到 (2) 化简方法。这里令 X_1 表示第一颗骰子的结果, X_2 表示第二颗骰子的结果, X_3 表示第三颗骰子的结果,所以 $X=X_1,Y=X_1+X_2,Z=X_1+X_2+X_3$ 。

$$H(X) = H(X_1) = H(X_2) = H(X_3)$$
(5)

$$=\sum_{1}^{6} \frac{1}{6} \log_2 6 \tag{6}$$

$$= \log_2 6bit \tag{7}$$

$$=2.5850bit \tag{8}$$

$$H(Y) = H(X_1 + H_2) = H(X_2 + X_3) = H(X_1 + X_3)$$
(9)

$$=\sum_{p}^{\Omega} p \log_2 \frac{1}{p} \tag{10}$$

$$= 3.2744bit \tag{11}$$

$$\Omega = \{ \frac{1}{36}, \frac{2}{36}, \frac{3}{36}, \frac{4}{36}, \frac{5}{36}, \frac{6}{36}, \frac{5}{36}, \frac{4}{36}, \frac{3}{36}, \frac{2}{36}, \frac{1}{36} \}$$
 (12)

1.

$$H(Z|Y) = H(X_1 + X_2 + X_3|X_1 + X_2)$$
(13)

$$= H(X_3|X_1 + X_2) \tag{14}$$

$$=H(X_3) \tag{15}$$

$$=2.5850bit (16)$$

2.

$$H(X|Y) = H(X_1|X_1 + X_2) (17)$$

$$= H(X_1) - I(X_1; X_1 + X_2)$$
(18)

$$= H(X_1) - H(X_1 + X_2) + H(X_2)$$
(19)

$$=1.8955bit \tag{20}$$

3.

$$H(Z|X,Y) = H(X_1 + X_2 + X_3|X_1, X_1 + X_2)$$
(21)

$$=H(X_3|X_1,X_1+X_2) (22)$$

$$=H(X_3) \tag{23}$$

$$=2.5850bit (24)$$

4.

$$H(X, Z|Y) = H(X|Y) + H(Z|X, Y)$$
 (25)

$$=4.4805bit$$
 (26)

5.

$$H(Z|X) = H(X_1 + X_2 + X_3|X_1)$$
(27)

$$= H(X_2 + X_3 | X_1) (28)$$

$$= H(X_2 + X_3) (29)$$

$$= 3.2744bit \tag{30}$$

题目 3

1.

$$P(0) = \sum_{i=1}^{8} P(u_i)P(0|u_i)$$
(31)

$$=\frac{4(1-p)}{8} + \frac{4p}{8} \tag{32}$$

$$=\frac{1}{2}\tag{33}$$

$$= \frac{1}{2}$$

$$P(u_1|0) = \frac{P(u_1)P(0|u_1)}{P(0)}$$
(33)

$$=\frac{\frac{1}{8}(1-p)}{\frac{1}{2}}\tag{35}$$

$$=\frac{1-p}{4}\tag{36}$$

$$I(u_1;0) = \log_2 \frac{P(u_1|0)}{P(u_1)} \tag{37}$$

$$= \log_2 \frac{\frac{1-p}{4}}{\frac{1}{8}} \tag{38}$$

$$= 1 + \log_2(1 - p)bit (39)$$

2.

$$P(00) = \sum_{i=1}^{8} P(u_i)P(00|u_i)$$
(40)

$$=\frac{1}{4}\tag{41}$$

$$= \frac{1}{4}$$

$$P(u_1|00) = \frac{P(u_1)P(00|u_1)}{P(00)}$$
(41)

$$=\frac{1-p^2}{2}\tag{43}$$

$$I(u_1; 00) = \log_2 \frac{P(u_1|00)}{P(u_1)} \tag{44}$$

$$= 2 + 2\log_2(1 - p)bit \tag{45}$$

参考文献 4

3.

$$P(000) = \sum_{i=1}^{8} P(u_i)P(000|u_i)$$
(46)

$$=\frac{1}{8}\tag{47}$$

$$= \frac{1}{8}$$

$$P(u_1|000) = \frac{P(u_1)P(000|u_1)}{P(000)}$$
(48)

$$= (1-p)^3 \tag{49}$$

$$I(u_1; 000) = \log_2 \frac{P(u_1|000)}{P(u_1)} \tag{50}$$

$$= 3 + 3\log_2(1 - p)bit \tag{51}$$

4.

$$P(0000) = \sum_{i=1}^{8} P(u_i)P(0000|u_i)$$
(52)

$$=\frac{(1-p)^4+p^4+6(1-p)^2p^2}{8}$$
 (53)

$$P(u_1|0000) = \frac{P(u_1)P(0000|u_1)}{P(0000)}$$
(54)

$$= \frac{(1-p)^4}{(1-p)^4 + p^4 + 6(1-p)^2 p^2}$$
 (55)

$$I(u_1; 0000) = \log_2 \frac{P(u_1|0000)}{P(u_1)}$$
(56)

$$= \log_2 \frac{8(1-p)^4}{(1-p)^4 + p^4 + 6(1-p)^2 p^2} bit$$
 (57)

参考文献

[1] Mokshay Madiman. On the entropy of sums. In 2008 IEEE Information Theory Workshop, pages 303-307. IEEE, 2008.