**Problem 1:** Suppose P is a probability measure defined on  $S = \{1, 2, 3, 4, 5\}$  with mass function

$$\begin{array}{c|cccc} s & p(s) \\ \hline 1 & 0.1 \\ 2 & 0.3 \\ 3 & 0.2 \\ 4 & 0.3 \\ 5 & 0.1 \\ \end{array}$$

Compute the information content I(s) of each sample point and the entropy H(P).

## We compute

$$\begin{array}{c|cccc} s & p(s) & I(s) \\ \hline 1 & 0.1 & 3.322 \\ 2 & 0.3 & 1.737 \\ 3 & 0.2 & 2.322 \\ 4 & 0.3 & 1.737 \\ 5 & 0.1 & 3.322 \\ \end{array}$$

where the surprisals are rounded to three places after the decimal point. The entropy is  $H(P) \approx 2.171$ .

**Problem 2:** Let  $X \sim \mathcal{B}er(\theta)$  for  $\theta \in [0,1]$ . Compute a formula for H(X) in terms of  $\theta$ .

We compute:

$$H(X) = -\sum_{x=0}^{1} p(x) \log_2(p(x)) = -(1-\theta) \log_2(1-\theta) - \theta \log_2(\theta).$$

**Problem 3:** Suppose P and Q are probability measures defined on  $S = \{1, 2, 3, 4, 5\}$  with mass functions

Compute the cross entropies  $H_P(Q)$  and  $H_Q(P)$ .

$$H_P(Q) = -\sum_{s=1}^5 p(s) \log_2(q(s)) \approx 3.258, \quad H_Q(P) = -\sum_{s=1}^5 q(s) \log_2(p(s)) \approx 2.337.$$

**Problem 4:** Suppose P and Q are probability measures defined on  $S = \{1, 2, 3, 4, 5\}$  with mass functions

$$\begin{array}{c|cccc} s & p(s) & q(s) \\ \hline 1 & 0.1 & 0.05 \\ 2 & 0.3 & 0.15 \\ 3 & 0.2 & 0.7 \\ 4 & 0.3 & 0.03 \\ 5 & 0.1 & 0.07 \\ \hline \end{array}$$

Compute  $D(P \parallel Q)$  and  $D(Q \parallel P)$ .

$$D(P \parallel Q) = \sum_{s=1}^{5} p(s) \log_2 \left( \frac{p(s)}{q(s)} \right) \approx 1.087 \quad \text{and} \quad D(Q \parallel P) = \sum_{s=1}^{5} q(s) \log_2 \left( \frac{q(s)}{p(s)} \right) \approx 0.929.$$

**Problem 5:** While it is **not** true that  $D(P \parallel Q) = D(Q \parallel P)$  for **all** P and Q, there are examples of special distributions P and Q for which equality does hold. Find examples.

For  $\theta \in [0,1]$ , define  $P_{\theta}$  and  $Q_{\theta}$  on the sample space  $S = \{0,1\}$  by

$$\begin{array}{c|cccc}
s & p_{\theta}(s) & q_{\theta}(s) \\
\hline
0 & \theta & 1 - \theta \\
1 & 1 - \theta & \theta
\end{array}$$

One then easily proves  $D(P_{\theta} \parallel Q_{\theta}) = D(Q_{\theta} \parallel P_{\theta})$ .

**Problem 6:** For each  $\phi \in [0,1]$ , consider the communication channel with transition matrix

$$\mathbf{K} = \begin{bmatrix} 1 - \phi & \phi \\ \phi & 1 - \phi \end{bmatrix}$$

These are called binary symmetric channels. Suppose that  $X \sim \mathcal{B}er(\alpha)$  for some  $\alpha \in [0, 1]$ , with the range of X enumerated as  $x_0 = 0$  and  $x_1 = 1$ . Show that X and the communication channel determine a random variable Y with range  $y_0 = 0$  and  $y_1 = 1$ . Determine its distribution.

Since the range of Y is  $\{0,1\}$ , it must be Bernoulli, with  $Y \sim \mathcal{B}er(\beta)$  for some  $\beta \in [0,1]$ . We need to determine the parameter  $\beta$ . But notice that the probability vectors encoding the mass functions of X and Y have the form

$$\pi(X)^{\intercal} = \begin{bmatrix} 1 - \alpha & \alpha \end{bmatrix}$$
 and  $\pi(Y)^{\intercal} = \begin{bmatrix} 1 - \beta & \beta \end{bmatrix}$ .

So, if we conceptualize the entries in the transition matrix as the conditional probabilities, then by the Law of Total Probability we must have

$$\boldsymbol{\pi}(Y)^{\mathsf{T}} = \boldsymbol{\pi}(X)^{\mathsf{T}} \mathbf{K} = \begin{bmatrix} 1 - \alpha & \alpha \end{bmatrix} \begin{bmatrix} 1 - \phi & \phi \\ \phi & 1 - \phi \end{bmatrix} = \begin{bmatrix} (1 - \alpha)(1 - \phi) + \alpha\phi & (1 - \alpha)\phi + \alpha(1 - \phi) \end{bmatrix}.$$

Thus,  $\beta = \phi + \alpha - 2\phi\alpha$ .

**Problem 7:** Suppose X and Y are Bernoulli random variables with joint mass function given by

$$\begin{array}{c|cccc} p(x,y) & y = 0 & y = 1 \\ \hline x = 0 & 0.3 & 0.1 \\ x = 1 & 0.36 & 0.24 \end{array}$$

(a) Compute the transition matrix of the communication channel induced from X and Y.

This just means that we need to compute the conditional mass function p(y|x). We begin by computing the marginal mass p(x):

$$\begin{array}{c|c}
x & p(x) \\
\hline
0 & 0.4 \\
1 & 0.6
\end{array}$$

Then the transition matrix is given by

$$\mathbf{K} = \left\{ \begin{array}{c|cc} p(y|x) & y = 0 & y = 1 \\ \hline x = 0 & 0.75 & 0.25 \\ x = 1 & 0.6 & 0.4 \end{array} \right\}.$$

(b) Compute the mutual information I(X,Y).

$$I(X,Y) \approx 0.018$$