

ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

Handout 28

Information Theory and Coding

Solutions to Homework 11

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PROBLEM 1.

- (a) Suppose \mathbf{x} and \mathbf{x}' are two codewords in \mathcal{C} . Then for $\forall i = 0, 1, \dots, m-1$,

$$\begin{aligned} x_0 + x_1\alpha_i + \dots + x_{n-1}\alpha_i^{n-1} &= 0 \\ x'_0 + x'_1\alpha_i + \dots + x'_{n-1}\alpha_i^{n-1} &= 0 \end{aligned}$$

Therefore,

$$(x_0 + x'_0) + (x_1 + x'_1)\alpha_i + \dots + (x_{n-1} + x'_{n-1})\alpha_i^{n-1} = 0 \quad \text{for } \forall i = 0, 1, \dots, m-1.$$

which shows $\mathbf{x} + \mathbf{x}'$ is also a codeword.

- (b) $x(D) = x_0 + x_1D + \dots + x_{n-1}D^{n-1}$ is a polynomial of degree (at most) $n-1$ and (x_0, \dots, x_{n-1}) is a codeword if $\alpha_0, \alpha_1, \dots, \alpha_{m-1}$ are m of its roots. This means

$$x(D) = (D - \alpha_0)(D - \alpha_1) \dots (D - \alpha_{m-1})h(D) = g(D)h(D)$$

for some $h(D)$. Note that $h(D)$ can have degree (at most) $n-m-1$. On the other side, there is a one-to-one correspondence between the codewords of \mathcal{C} and degree $n-1$ polynomials. Since $g(D)$ is fixed for all codewords, a polynomial $x(D)$ corresponding to a codeword \mathbf{x} is determined by choosing the coefficients of $h(D) = h_0 + h_1D + \dots + h_{n-m-1}D^{n-m-1}$. Since $h_j \in \mathcal{X}$ for $j = 0, 1, \dots, n-m-1$ we have q^{n-m} different $h(D)$ s and, thus, q^{n-m} codewords.

- (c) For every column vector $\mathbf{u} = [u_0, u_1, \dots, u_{m-1}]^T$, $\mathbf{A}\mathbf{u} = [u(1), u(\beta), \dots, u(\beta^{n-1})]^T$. Consequently, $\mathbf{A}\mathbf{u} = \mathbf{0}$ means $u(D)$ has n roots which is impossible (since it is a polynomial of degree $m-1 < n$).

- (d) Using the same reasoning as in (c) one can verify that $\mathbf{x} = (x_1, \dots, x_n)$ is a codeword iff $\mathbf{x}\mathbf{A} = \mathbf{0}$. This means \mathbf{A} is the parity-check matrix of the code \mathcal{C} . Since the code is linear, using Problem 4 of Homework 11 we know that has minimum distance d iff every $d-1$ rows of \mathbf{H} are linearly independent and some d rows are linearly dependent. That \mathbf{A} has rank m implies there are no m linearly dependent rows thus $d \geq m+1$. On the other side, we know from the Singleton bound that a code with q^{n-m} codewords and block-length n has minimum distance $d \leq m+1$. Thus we conclude that $d = m+1$.

PROBLEM 2.

- (a) For every $0 \leq p \leq 1$, define $\bar{p} := 1 - p$. We have:

$$h_2(\bar{p}) = -\bar{p} \log \bar{p} - p \log p = -p \log p - \bar{p} \log \bar{p} = h_2(p). \quad (1)$$

On the other hand, it is easy to check that for every $0 \leq p', p'' \leq 1$, we have:

$$\bar{p}' * p'' = p' * \bar{p}'' = \overline{p' * p''} \quad \text{and} \quad \bar{p}' * \bar{p}'' = p' * p''.$$

Now (1) implies that

$$h_2(\bar{p}' * p'') = h_2(p' * \bar{p}'') = h_2(\overline{p' * p''}) = h_2(p' * p''). \quad (2)$$

Let $p' = \mathbb{P}[X_1 = 1]$ and $p'' = \mathbb{P}[X_2 = 1]$. We have the following:

- $\mathbb{P}[X_1 \oplus X_2 = 1] = \mathbb{P}[X_1 = 1]\mathbb{P}[X_2 = 0] + \mathbb{P}[X_1 = 0]\mathbb{P}[X_2 = 1] = p'\overline{p''} + \overline{p'}p'' = p' * p''$. Therefore, $H(X_1 \oplus X_2) = h_2(p' * p'')$.
- Since $H(X_1) = h_2(p_1)$, then we have either $p' = p_1$ or $p' = 1 - p_1$. I.e., we have $p_1 = p'$ or $p_1 = 1 - p' = \overline{p'}$.
- Since $H(X_2) = h_2(p_2)$, then we have either $p'' = p_2$ or $p'' = 1 - p_2$. I.e., we have $p_2 = p''$ or $p_2 = 1 - p'' = \overline{p''}$.

Now (2) implies that $H(X_1 \oplus X_2) = h_2(p' * p'') = h_2(p_1 * p_2)$.

(b) We have $H(X_1|Y) = \sum_{y \in \mathcal{Y}} H(X_1|Y=y)\mathbb{P}_Y(y) = \sum_{y \in \mathcal{Y}} h_2(p_1(y))q(y)$.

Now for every $y \in \mathcal{Y}$, X_1 and X_2 are independent conditioned on $Y = y$. Moreover, $H(X_1|Y=y) = h_2(p_1(y))$ and $H(X_2|Y=y) = H(X_2) = h_2(p_2)$ since X_2 and Y are independent. Therefore, part (a) implies that $H(X_1 \oplus X_2|Y=y) = h_2(p_1(y) * p_2)$.

We conclude that

$$\begin{aligned} H(X_1 \oplus X_2|Y) &= \sum_{y \in \mathcal{Y}} H(X_1 \oplus X_2|Y=y)\mathbb{P}_Y(y) \\ &= \sum_{y \in \mathcal{Y}} h_2(p_1(y) * p_2)q(y) = \sum_{y \in \mathcal{Y}} h_2(p_2 * p_1(y))q(y). \end{aligned}$$

(c) Note that $p_2 * p = p(1 - p_2) + p_2(1 - p) = \beta p + p_2$, where $\beta = 1 - 2p_2 \geq 0$. Let $g(p) = \frac{\frac{\partial}{\partial p} h_2(p_2 * p)}{\frac{\partial}{\partial p} h_2(p)} = \frac{\frac{\partial}{\partial p} h_2(\beta p + p_2)}{\frac{\partial}{\partial p} h_2(p)} = \frac{\beta h_2'(\beta p + p_2)}{h_2'(p)}$. We have

$$\begin{aligned} g'(p) &= \frac{\beta^2 h_2''(\beta p + p_2) h_2'(p) - \beta h_2''(p) h_2'(\beta p + p_2)}{h_2'(p)^2} \\ &= \frac{\beta h_2''(\beta p + p_2) h_2''(p)}{h_2'(p)^2} \left[\beta \frac{h_2'(p)}{h_2''(p)} - \frac{h_2'(\beta p + p_2)}{h_2''(\beta p + p_2)} \right]. \end{aligned}$$

Note that $h_2'(p) = \log \frac{1-p}{p}$ and $h_2''(p) = \frac{-1}{p(1-p)\ln 2}$, which implies that $h_2''(\beta p + p_2) \leq 0$ and $h_2''(p) \leq 0$. Therefore, $\frac{\beta h_2''(\beta p + p_2) h_2''(p)}{h_2'(p)^2} \geq 0$ and so it is sufficient to show that we have $\beta \frac{h_2'(p)}{h_2''(p)} - \frac{h_2'(\beta p + p_2)}{h_2''(\beta p + p_2)} \geq 0$. Now define $\alpha = 1 - 2p$. It is easy to check the following:

- $p = \frac{1}{2}(1 - \alpha)$.
- $1 - p = \frac{1}{2}(1 + \alpha)$.
- $\beta p + p_2 = \frac{1}{2}(1 - \alpha\beta)$.
- $1 - (\beta p + p_2) = \frac{1}{2}(1 + \alpha\beta)$.

Therefore, we have

$$\beta \frac{h_2'(p)}{h_2''(p)} = -\beta(\ln 2)p(1-p) \log \frac{1-p}{p} = -\frac{\beta \ln 2}{4}(1 - \alpha^2) \log \frac{1 + \alpha}{1 - \alpha},$$

and

$$\frac{h_2'(\beta p + p_2)}{h_2''(\beta p + p_2)} = -(\ln 2)(\beta p + p_2)(1 - \beta p - p_2) \log \frac{1 - \beta p - p_2}{\beta p + p_2} = -\frac{\ln 2}{4}(1 - (\alpha\beta)^2) \log \frac{1 + \alpha\beta}{1 - \alpha\beta}.$$

Using the formula $\log(1+x) = \sum_{k \geq 1} (-1)^{k-1} \frac{x^k}{k}$, we get

$$\begin{aligned} \log \frac{1+x}{1-x} &= \log(1+x) - \log(1-x) = \left(\sum_{k \geq 1} (-1)^{k-1} \frac{x^k}{k} \right) - \left(\sum_{k \geq 1} (-1)^{k-1} \frac{(-x)^k}{k} \right) \\ &= \sum_{k \geq 1} ((-1)^{k-1} + 1) \frac{x^k}{k} = 2 \sum_{\substack{k \geq 1 \\ k \text{ is odd}}} \frac{x^k}{k}. \end{aligned}$$

Therefore,

$$\begin{aligned} -(1-x^2) \log \frac{1+x}{1-x} &= -2 \sum_{\substack{k \geq 1 \\ k \text{ is odd}}} \frac{x^k}{k} + 2 \sum_{\substack{k \geq 1 \\ k \text{ is odd}}} \frac{x^{k+2}}{k} = -2x - 2 \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \frac{x^k}{k} + 2 \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \frac{x^k}{k-2} \\ &= -2x + 2 \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) x^k. \end{aligned}$$

Hence,

$$\begin{aligned} \beta \frac{h'_2(p)}{h''_2(p)} &= -\frac{\beta \ln 2}{4} (1-\alpha^2) \log \frac{1+\alpha}{1-\alpha} = \frac{\beta \ln 2}{4} \left[-2\alpha + 2 \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) \alpha^k \right] \\ &= -\frac{\alpha \beta \ln 2}{2} + \frac{\ln 2}{2} \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) \beta \alpha^k, \end{aligned}$$

and

$$\begin{aligned} \frac{h'_2(\beta p + p_2)}{h''_2(\beta p + p_2)} &= -\frac{\ln 2}{4} (1 - (\alpha\beta)^2) \log \frac{1+\alpha\beta}{1-\alpha\beta} = \frac{\ln 2}{4} \left[-2\alpha\beta + 2 \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) (\alpha\beta)^k \right] \\ &= -\frac{\alpha\beta \ln 2}{2} + \frac{\ln 2}{2} \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) \beta^k \alpha^k. \end{aligned}$$

We conclude that

$$\beta \frac{h'_2(p)}{h''_2(p)} - \frac{h'_2(\beta p + p_2)}{h''_2(\beta p + p_2)} = \frac{\ln 2}{2} \sum_{\substack{k \geq 3 \\ k \text{ is odd}}} \left(\frac{1}{k-2} - \frac{1}{k} \right) (\beta - \beta^k) \alpha^k \stackrel{(*)}{\geq} 0,$$

where (*) follows from the fact that $\beta = 1 - 2p_2 \leq 1$ which implies that $\beta^k \leq \beta$. Therefore, $g'(p) \geq 0$ and so $g(p)$ is increasing. We conclude that the function f is convex.

(d) We have

$$\begin{aligned} H(X_1 \oplus X_2 | Y) &= \sum_{y \in \mathcal{Y}} h_2(p_2 * p_1(y)) q(y) = \sum_{y \in \mathcal{Y}} h_2(p_2 * h_2^{-1}(H(X_1 | Y = y))) q(y) \\ &= \sum_{y \in \mathcal{Y}} f(H(X_1 | Y = y)) q(y) \stackrel{(*)}{\geq} f\left(\sum_{y \in \mathcal{Y}} H(X_1 | Y = y) q(y) \right) \\ &= f(H(X_1 | Y)) = h_2(p_2 * h_2^{-1}(H(X_1 | Y))) = h_2(p_2 * p_1) = h_2(p_1 * p_2), \end{aligned}$$

where (*) follows from the convexity of the function f .

(e) For every $y_1 \in \mathcal{Y}_1$, let $0 \leq p_1(y_1) \leq \frac{1}{2}$ be such that $H(X_1|Y_1 = y_1) = h_2(p_1(y_1))$ and let $q_1(y_1) = \mathbb{P}_{Y_1}(y_1)$. Similarly, for every $y_2 \in \mathcal{Y}_2$, let $0 \leq p_2(y_2) \leq \frac{1}{2}$ be such that $H(X_2|Y_2 = y_2) = h_2(p_2(y_2))$ and let $q_2(y_2) = \mathbb{P}_{Y_2}(y_2)$. For every $y_1 \in \mathcal{Y}_1$, define the mapping $f_{y_1} : [0, 1] \rightarrow \mathbb{R}$ as $f_{y_1}(h) = h_2(p_1(y_1) * h_2^{-1}(h))$. Part (c) implies that f_{y_1} is convex for every $y_1 \in \mathcal{Y}_1$. We have

$$\begin{aligned}
H(X_1 \oplus X_2|Y_1, Y_2) &= \sum_{y_1 \in \mathcal{Y}_1} \sum_{y_2 \in \mathcal{Y}_2} h_2(p_1(y_1) * p_2(y_2)) \mathbb{P}_{Y_1, Y_2}(y_1, y_2) \\
&= \sum_{y_1 \in \mathcal{Y}_1} \sum_{y_2 \in \mathcal{Y}_2} h_2(p_1(y_1) * p_2(y_2)) q_1(y_1) q_2(y_2) \\
&= \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) \sum_{y_2 \in \mathcal{Y}_2} h_2(p_1(y_1) * h_2^{-1}(H(X_2|Y_2 = y_2))) q_2(y_2) \\
&= \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) \sum_{y_2 \in \mathcal{Y}_2} f_{y_1}(H(X_2|Y_2 = y_2)) q_2(y_2) \\
&\stackrel{(*)}{\geq} \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) f_{y_1} \left(\sum_{y_2 \in \mathcal{Y}_2} H(X_2|Y_2 = y_2) q_2(y_2) \right) \\
&= \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) f_{y_1}(H(X_2|Y_2)) = \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) h_2(p_1(y_1) * h_2^{-1}(H(X_2|Y_2))) \\
&= \sum_{y_1 \in \mathcal{Y}_1} q_1(y_1) h_2(p_1(y_1) * p_2) = \sum_{y_1 \in \mathcal{Y}_1} h_2(p_2 * h_2^{-1}(H(X_1|Y_1 = y_1))) q_1(y_1) \\
&= \sum_{y_1 \in \mathcal{Y}_1} f(H(X_1|Y_1 = y_1)) q_1(y_1) \stackrel{(**)}{\geq} f \left(\sum_{y_1 \in \mathcal{Y}_1} H(X_1|Y_1 = y_1) q_1(y_1) \right) \\
&= f(H(X_1|Y_1)) = h_2(p_2 * h_2^{-1}(H(X_1|Y_1))) = h_2(p_2 * p_1) = h_2(p_1 * p_2),
\end{aligned}$$

where (*) follows from the convexity of the functions $\{f_{y_1} : y_1 \in \mathcal{Y}_1\}$ and (**) follows from the convexity of f .

PROBLEM 3.

- (a) Any codeword of \mathcal{C} is of the form $\langle \mathbf{a}, \mathbf{a} \oplus \mathbf{b} \rangle$ with $\mathbf{a} \in \mathcal{C}_1$ and $\mathbf{b} \in \mathcal{C}_2$. Given two codewords $\langle \mathbf{u}', \mathbf{u}' \oplus \mathbf{v}' \rangle$ and $\langle \mathbf{u}'', \mathbf{u}'' \oplus \mathbf{v}'' \rangle$ of \mathcal{C} , their sum is $\langle \mathbf{u}, \mathbf{u} \oplus \mathbf{v} \rangle$ with $\mathbf{u} = \mathbf{u}' \oplus \mathbf{u}''$ and $\mathbf{v} = \mathbf{v}' \oplus \mathbf{v}''$. Since \mathcal{C}_1 and \mathcal{C}_2 are linear codes $\mathbf{u} \in \mathcal{C}_1$ and $\mathbf{v} \in \mathcal{C}_2$. Thus the sum of any two codewords of \mathcal{C} is a codeword of \mathcal{C} and we conclude that \mathcal{C} is linear.
- (b) If $(\mathbf{u}, \mathbf{v}) \neq (\mathbf{u}', \mathbf{v}')$, then either $\mathbf{u} \neq \mathbf{u}'$, or, $\mathbf{u} = \mathbf{u}'$ and $\mathbf{v} \neq \mathbf{v}'$. In either case $\langle \mathbf{u} | \mathbf{u} \oplus \mathbf{v} \rangle \neq \langle \mathbf{u}' | \mathbf{u}' \oplus \mathbf{v}' \rangle$: in the first case the first halves differ, in the second case the second halves differ. Thus no two of the (\mathbf{u}, \mathbf{v}) pairs are mapped to the same element of \mathcal{C} , and the code has exactly $M_1 M_2$ elements. Its rate is $\frac{1}{2n} \log(M_1 M_2) = \frac{1}{2} R_1 + \frac{1}{2} R_2$.
- (c) As $\mathbf{v} = \mathbf{u} \oplus \mathbf{u} \oplus \mathbf{v}$,

$$w_H(\mathbf{v}) = w_H(\mathbf{u} \oplus \mathbf{u} \oplus \mathbf{v}) \leq w_H(\mathbf{u}) + w_H(\mathbf{u} \oplus \mathbf{v})$$

by the triangle inequality. Noting that the right hand side is $w_H(\langle \mathbf{u} | \mathbf{u} \oplus \mathbf{v} \rangle)$ completes the proof.

- (d) If $\mathbf{v} = \mathbf{0}$ we have $\langle \mathbf{u} | \mathbf{u} \oplus \mathbf{v} \rangle = \langle \mathbf{u} | \mathbf{u} \rangle$ which has twice the Hamming weight of \mathbf{u} . Otherwise (c) gives $w_H(\langle \mathbf{u} | \mathbf{u} \oplus \mathbf{v} \rangle) \geq w_H(\mathbf{v})$.

- (e) Since \mathcal{C} is linear its minimum distance equals the minimum weight of its non-zero codewords. If $\langle \mathbf{u}|\mathbf{u} \oplus \mathbf{v} \rangle$ is non-zero either $\mathbf{v} \neq \mathbf{0}$, or, $\mathbf{v} = \mathbf{0}$ and $\mathbf{u} \neq \mathbf{0}$. By (d), in the first case $w_H(\langle \mathbf{u}|\mathbf{u} \oplus \mathbf{v} \rangle) \geq w_H(\mathbf{v}) \geq d_1$, in the second case $w_H(\langle \mathbf{u}|\mathbf{u} \oplus \mathbf{v} \rangle) \geq 2w_H(\mathbf{u}) \geq 2d_2$. Thus $d \geq \min\{2d_1, d_2\}$.
- (f) Let \mathbf{u}_0 be the minimum weight non-zero codeword of \mathcal{C}_1 and let \mathbf{v}_0 be the minimum weight non-zero codeword of \mathcal{C}_2 . Note that $\langle \mathbf{u}_0|\mathbf{u}_0 \rangle$ is a non-zero codeword of \mathcal{C} (corresponding to the choice $\mathbf{u} = \mathbf{u}_0, \mathbf{v} = \mathbf{0}$). It has weight $2d_1$. Similarly, $\langle \mathbf{0}|\mathbf{v}_0 \rangle$ is also a non-zero codeword of \mathcal{C} (corresponding to the choice $\mathbf{u} = \mathbf{0}, \mathbf{v} = \mathbf{v}_0$). It has weight d_2 . Consequently $d \leq \min\{2d_1, d_2\}$. In light of (e) we find $d = \min\{2d_1, d_2\}$.

This method of constructing a longer code from two shorter ones is known under several names: ‘Plotkin construction’, ‘bar product’, ‘ $(u|u+v)$ construction’ appear regularly in the literature. Compare this method to the ‘obvious’ method of letting the codewords to be $\langle \mathbf{u}|\mathbf{v} \rangle$. The simple method has the same block-length and rate as we have here, but its minimum distance is only $\min\{d_1, d_2\}$. The factor two gained in d_1 by the bar product is significant, and many practical code families can be built from very simple base codes by a recursive application of the bar product. Notable among them are the family of Reed–Muller codes.

PROBLEM 4.

- (a) We have

$$\begin{aligned}
Q_{i+1} &= \sqrt{Z_{i+1}(1 - Z_{i+1})} = \begin{cases} \sqrt{Z_i^2(1 - Z_i^2)} & \text{w.p. } 1/2 \\ \sqrt{(2Z_i - Z_i^2)(1 - 2Z_i + Z_i^2)} & \text{w.p. } 1/2 \end{cases} \\
&= \begin{cases} \sqrt{Z_i^2(1 - Z_i)(1 + Z_i)} & \text{w.p. } 1/2 \\ \sqrt{(2 - Z_i)Z_i(1 - Z_i)^2} & \text{w.p. } 1/2 \end{cases} \\
&= \begin{cases} \sqrt{Z_i(1 - Z_i)}\sqrt{Z_i(1 + Z_i)} & \text{w.p. } 1/2 \\ \sqrt{Z_i(1 - Z_i)}\sqrt{(2 - Z_i)(1 - Z_i)} & \text{w.p. } 1/2 \end{cases} \\
&= \sqrt{Z_i(1 - Z_i)} \begin{cases} \sqrt{Z_i(1 + Z_i)} & \text{w.p. } 1/2 \\ \sqrt{(2 - Z_i)(1 - Z_i)} & \text{w.p. } 1/2 \end{cases} \\
&= Q_i \begin{cases} f_1(Z_i) & \text{w.p. } 1/2 \\ f_2(Z_i) & \text{w.p. } 1/2 \end{cases},
\end{aligned}$$

where $f_1(z) = \sqrt{z(z+1)}$ and $f_2(z) = \sqrt{(2-z)(1-z)}$.

- (b) We have

$$f_1'(z) = \frac{2z+1}{2\sqrt{z(z+1)}}$$

so

$$\begin{aligned}
f_1''(z) &= \frac{4\sqrt{z(z+1)} - (2z+1)\frac{2(2z+1)}{2\sqrt{z(z+1)}}}{\left(2\sqrt{z(z+1)}\right)^2} \\
&= \frac{4z(z+1) - (2z+1)^2}{4(z(z+1))^{\frac{3}{2}}} = \frac{-1}{4(z(z+1))^{\frac{3}{2}}} \leq 0.
\end{aligned}$$

Therefore, f_1 is concave. By noticing that $f_2(z) = f_1(1 - z)$, we obtain:

$$\begin{aligned} f_1(z) + f_2(z) &= f_1(z) + f_1(1 - z) = 2 \left(\frac{1}{2} f_1(z) + \frac{1}{2} f_1(1 - z) \right) \\ &\stackrel{(*)}{\leq} 2 f_1 \left(\frac{1}{2} z + \frac{1}{2} (1 - z) \right) = 2 f_1 \left(\frac{1}{2} \right) = 2 \sqrt{\frac{1}{2} \left(\frac{1}{2} + 1 \right)} \\ &= 2 \sqrt{\frac{1}{2} \cdot \frac{3}{2}} = 2 \frac{\sqrt{3}}{2} = \sqrt{3}, \end{aligned}$$

where $(*)$ follows from the concavity of f_1 . We have

$$\mathbb{E}[Q_{i+1} \mid Z_0, \dots, Z_i] = \frac{1}{2} f_1(Z_i) Q_i + \frac{1}{2} f_2(Z_i) Q_i = \frac{1}{2} (f_1(Z_i) + f_2(Z_i)) Q_i \leq \rho Q_i,$$

where $\rho = \frac{\sqrt{3}}{2} < 1$.

- (c) We will show the claim by induction on $i \geq 0$. For $i = 0$, we have $Z_0 = z_0$ with probability 1. Therefore, $\mathbb{E}Q_0 = \sqrt{z_0(1 - z_0)}$.

It is easy to see that the function $[0, 1] \rightarrow \mathbb{R}$ defined by $z \rightarrow \sqrt{z(1 - z)}$ achieves its maximum at $z = \frac{1}{2}$, and so $\mathbb{E}Q_0 = \sqrt{z_0(1 - z_0)} \leq \sqrt{\frac{1}{2} \left(1 - \frac{1}{2} \right)} = \frac{1}{2}$. Therefore, the claim is true for $i = 0$.

Now suppose that the claim is true for $i \geq 0$, i.e., $\mathbb{E}Q_i \leq \frac{1}{2} \rho^i$. We have

$$\mathbb{E}Q_{i+1} = \mathbb{E}[\mathbb{E}[Q_{i+1} \mid Z_0, \dots, Z_i]] \stackrel{(*)}{\leq} \mathbb{E}[\rho Q_i] = \rho \mathbb{E}[Q_i] \stackrel{(**)}{\leq} \rho \cdot \frac{1}{2} \rho^i = \frac{1}{2} \rho^{i+1},$$

where $(*)$ follows from Part (b) and $(**)$ follows from the induction hypothesis. We conclude that $\mathbb{E}Q_i \leq \frac{1}{2} \rho^i$ for every $i \geq 0$.

- (d) By noticing that $\delta < z < 1 - \delta$ if and only if $z(1 - z) > \delta(1 - \delta)$, we get:

$$\begin{aligned} \mathbb{P}[Z_i \in (\delta, 1 - \delta)] &= \mathbb{P}[Z_i(1 - Z_i) > \delta(1 - \delta)] = \mathbb{P}[\sqrt{Z_i(1 - Z_i)} > \sqrt{\delta(1 - \delta)}] \\ &= \mathbb{P}[Q_i > \sqrt{\delta(1 - \delta)}] \stackrel{(*)}{\leq} \frac{\mathbb{E}Q_i}{\sqrt{\delta(1 - \delta)}} \stackrel{(**)}{\leq} \frac{\rho^i}{2\sqrt{\delta(1 - \delta)}}, \end{aligned}$$

where $(*)$ follows from the Markov inequality and $(**)$ follows from Part (c). Now since $\rho < 1$, we have $\frac{\rho^i}{2\sqrt{\delta(1 - \delta)}} \rightarrow 0$ as $i \rightarrow \infty$. We conclude that

$$\mathbb{P}[Z_i \in (\delta, 1 - \delta)] \rightarrow 0 \text{ as } i \text{ gets large.}$$

PROBLEM 5. As we should never represent a 0 with a 1, we are restricted to conditional distributions with $p_{V|U}(1|0) = 0$. Consequently, the possible $p_{V|U}$ are of the type

$$p_{V|U}(0|0) = 1 \quad p_{V|U}(1|0) = 0, \quad p_{V|U}(0|1) = \alpha \quad p_{V|U}(1|1) = 1 - \alpha,$$

and parametrized by $\alpha \in [0, 1]$. For $p_{V|U}$ as above, we have $\Pr(V = 1) = \frac{1}{2}(1 - \alpha)$, and

$$\mathbb{E}[d(U, V)] = \sum_{u,v} p_U(u) p_{V|U}(v|u) d(u, v) = \alpha/2,$$

$$I(U; V) = H(V) - H(V|U) = h_2\left(\frac{1}{2}(1 - \alpha)\right) - \frac{1}{2} h_2(\alpha) =: f(\alpha).$$

Thus $R(D) = \min\{f(\alpha) : 0 \leq \alpha \leq \min\{1, 2D\}\}$, with $f(\alpha) = h_2(\frac{1}{2}(1 - \alpha)) - \frac{1}{2}h_2(\alpha)$. It is not difficult to check that f is a decreasing function on the interval $[0, 1]$, and thus consequently

$$R(D) = \begin{cases} h_2(\frac{1}{2} - D) - \frac{1}{2}h_2(2D), & 0 \leq D < \frac{1}{2} \\ 0, & D \geq \frac{1}{2}. \end{cases}$$

Note that for $D \geq \frac{1}{2}$ we can represent any u with a constant, namely $v = 0$, with average distortion $1/2$.