Exercise 12.1

Let $|\psi\rangle = a|0\rangle + b|1\rangle$. Then, $|\varphi\rangle = b|0\rangle - a|1\rangle$.

We require a gate which transforms between the basis of $|\psi\rangle$ and $|\varphi\rangle$, and the computational basis. This can be performed using the unitary, $U = U^d agger = \begin{bmatrix} a & b \\ b & -a \end{bmatrix}$. Using U we construct the following circuit which performs the cloning.

Exercise 12.2

$$\sum_{y} \sqrt{E_{y}} \otimes U_{y}(\sigma \otimes |0\rangle \langle 0|) \sqrt{E_{y}} \otimes U_{y} = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y\rangle \langle 0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{E_{y}} \otimes |0 + y| = \sum_{y} \sqrt{E_{y}} \sigma \sqrt{$$

Exercise 12.3

From the Holevo bound $H(X:Y) \leq S(\rho)$, however $S(\rho)$ is bounded by $\log 2^n = n$, hence no more than n bits can be transferred.

Exercise 12.4

Let $\rho_i = |X_i\rangle \langle X_i|$, and $\rho = \frac{1}{4}\sum_i \rho_i = \frac{I}{2}$, which is the maximally mixed state, hence $S(\rho) = 1$ and the $S(\rho_i) = 0$ as the ρ_i are pure states. Then,

$$\chi = S(\rho) - \sum_{i} p_i S(\rho_i) = 1$$

We have tetrahedral symmetry for the states $|X_i\rangle$, hence we consider POVMs with the same symmetry. Consider $E_i = \frac{1}{2} |X_i\rangle \langle X_i|$, and $F_i = \frac{1}{2} |\psi_i\rangle \langle \psi_i|$, where

$$|\psi_1\rangle = |1\rangle$$

$$|\psi_2\rangle = \sqrt{\frac{1}{3}}(\sqrt{2}e^{\pi i/3}|0\rangle + |1\rangle)$$

$$|\psi_3\rangle = \sqrt{\frac{1}{3}}(\sqrt{2}e^{\pi i}|0\rangle + |1\rangle)$$

$$|\psi_4\rangle = \sqrt{\frac{1}{3}}(\sqrt{2}e^{5\pi i/3}|0\rangle + |1\rangle)$$

Using H(X:Y) = H(Y) - H(Y|X), $P(X=x) = \frac{1}{4}$, $p(Y=y|x=x) = tr(\rho_x E_y)$ and Bayes theorem we get,

$$H(Y) = -\sum_{y=0}^{\infty} p(Y = y) \log p(Y = y) =$$

$$-\sum_{y} \left(\sum_{x} p(X=x) p(Y=y|X=x) \right) \log \left(\sum_{x} p(X=x) p(Y=y|X=x) \right)$$

$$H(Y|X) = -\sum_{x} p(X=x) \sum_{y} p(Y=y|X=x) \log p(Y=y|X=x)$$

The calculation can be found in 12.3.py and gives $H(X:Y)_E \approx 0.208$ and $H(X:Y)_F \approx 0.415$. Therefore, F_i is the desired POVM.

As shown in DOI: 10.1109/TIT.1978.1055941 the maximum information is $\log \frac{4}{3} \approx 0.415$.

Exercise 12.5

We know that $p(x \in T(n, \epsilon)) \ge 1 - \delta$, hence $1 - p(x \in T(n, \epsilon)) \le \delta$. The expected number of bits required for a single symbol when the xs are ϵ -typical will be,

$$E(n,\epsilon) = H(X)p(x \in T(n,\epsilon)) + \frac{\log d^n}{n}p(x \notin T(n,\epsilon)) \le Rp(x \in T(n,\epsilon)) + \log d(1 - p(x \in T(n,\epsilon))) \le R + \delta \log \delta$$

$$Rp(x \in T(n, \epsilon)) + \log d(1 - p(x \in T(n, \epsilon))) \le R + \delta \log \delta$$

Hence,

$$E(n, \epsilon) - R \le \delta \log d$$

Exercise 12.6

$$C_X = p^{n-wt(X)}(1-p)^{wt(X)}$$

Where, wt(X) is the Hamming weight.

Exercise 12.7

This is the circuit outlined in Box 12.4 but with the basis change gate V set to I.

Exercise 12.8

Taking $p_J = p_{j_1} p_{j_2} \dots p_{j_n}$ and using Exercise 9.20 for the concavity of fidelity, from Equation 9.141 we have,

$$\bar{F} \geq F\left(\sum_{J} p_{J} \rho_{J}, D^{n} \circ C^{n}\right) = F\left(\sum_{J} p_{j_{1}} |\psi_{j_{1}}\rangle \langle \psi_{j_{1}}| \otimes p_{j_{2}} |\psi_{j_{2}}\rangle \langle \psi_{j_{2}}| \dots p_{j_{n}} |\psi_{j_{n}}\rangle \langle \psi_{j_{n}}|, D^{n} \circ C^{n}\right) = F(\rho^{\otimes n}, D^{n} \circ C^{n}) = F(\rho, D^{n} \circ C^{n})^{n} \geq (1 - 2\delta)^{n}$$

Where the last inequality is only true for $R > S(\rho)$.

Exercise 12.9

- Exercise 12.10
- Exercise 12.11
- Exercise 12.12
- Exercise 12.13
- Exercise 12.14
- Exercise 12.15
- Exercise 12.16
- Exercise 12.17
- Exercise 12.18
- Exercise 12.19
- Exercise 12.20