

Information Theory with Applications, Math6397

Lecture Notes from October 2, 2014

taken by Nathaniel Hammen

Last Time:

- Huffman's Code, also for k-ary trees
- Discrete Memoryless channels
- Channel coding with fixed length transition codes
- Jointly typical sequences

3 Coding for discrete channels (continued)

3.1 The discrete memoryless channel (continued)

3.1.9 Theorem. *Given a discrete memoryless source $\{X_j\}_{j=1}^{\infty}$ and a discrete memoryless channel $\gamma : \mathbb{A} \times \Omega \rightarrow \mathbb{B}$, denoting $Y_j = \gamma(X_j)$ for all $j \in \mathbb{N}$, then if $\ln(\mathbb{P}_{X_1}(X_1))$, $\ln(\mathbb{P}_{Y_1}(Y_1))$ and $\ln(\mathbb{P}_{X_1, Y_1}(X_1, Y_1))$ are integrable, we have*

$$-\frac{1}{n} \ln(\mathbb{P}_{X_1, \dots, X_n}(X_1, \dots, X_n)) \rightarrow H(X_1)$$

$$-\frac{1}{n} \ln(\mathbb{P}_{Y_1, \dots, Y_n}(Y_1, \dots, Y_n)) \rightarrow H(Y_1)$$

and

$$-\frac{1}{n} \ln(\mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(X_1, \dots, X_n, Y_1, \dots, Y_n)) \rightarrow H(X_1, Y_1)$$

Proof. By independence,

$$\ln(\mathbb{P}_{X_1, \dots, X_n}(X_1, \dots, X_n)) = \ln \left(\prod_{j=1}^n \mathbb{P}_{X_j}(X_j) \right) = \sum_{j=1}^n \ln(\mathbb{P}_{X_j}(X_j))$$

$$\ln(\mathbb{P}_{Y_1, \dots, Y_n}(Y_1, \dots, Y_n)) = \ln \left(\prod_{j=1}^n \mathbb{P}_{Y_j}(Y_j) \right) = \sum_{j=1}^n \ln(\mathbb{P}_{Y_j}(Y_j))$$

$$\begin{aligned}\ln(\mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(X_1, \dots, X_n, Y_1, \dots, Y_n)) &= \ln\left(\prod_{j=1}^n \mathbb{P}_{X_j, Y_j}(X_j, Y_j)\right) \\ &= \sum_{j=1}^n \ln(\mathbb{P}_{X_j, Y_j}(X_j, Y_j))\end{aligned}$$

We then get convergence by the strong law of large numbers:

$$\begin{aligned}-\frac{1}{n} \sum_{j=1}^n \ln(\mathbb{P}_{X_j}(X_j)) &\rightarrow -\mathbb{E}[\ln(\mathbb{P}_{X_1}(X_1))] = H(X_1) \\ -\frac{1}{n} \sum_{j=1}^n \ln(\mathbb{P}_{Y_j}(Y_j)) &\rightarrow -\mathbb{E}[\ln(\mathbb{P}_{Y_1}(Y_1))] = H(Y_1) \\ -\frac{1}{n} \sum_{j=1}^n \ln(\mathbb{P}_{X_j, Y_j}(X_j, Y_j)) &\rightarrow -\mathbb{E}[\ln(\mathbb{P}_{X_1, Y_1}(X_1, Y_1))] = H(X_1, Y_1)\end{aligned}$$

□

Since the strong implies the weak law of large numbers, we have an immediate consequence for the probability of F_δ^n .

3.1.10 Corollary. *Let $\epsilon > 0$. By choosing n sufficiently large, we can achieve $\mathbb{P}(F_\delta^n) > 1 - \epsilon$*

3.1.11 Theorem (Shannon-McMillan for channels). *Given a discrete memoryless source $\{X_j\}_{j=1}^\infty$, and a discrete memoryless channel $\gamma : \mathbb{A} \times \Omega \rightarrow \mathbb{B}$, $\ln(\mathbb{P}_{X_1}(X_1))$, $\ln(\mathbb{P}_{Y_1}(Y_1))$ and $\ln(\mathbb{P}_{X_1, Y_1}(X_1, Y_1))$ integrable, and $\delta > 0$, then for all sufficiently large n , F_δ^n satisfies*

1. $\mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}((F_\delta^n)^c) < \delta$
2. $(1 - \delta) \exp(n(H(X, Y) - \delta)) < |F_\delta^n|$
3. $|F_\delta^n| < \exp(n(H(X, Y) + \delta))$
4. *If $(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n$, then*

$$\exp(-n(H(X, Y) + \delta)) < \mathbb{P}(x_1, \dots, x_n, y_1, \dots, y_n) < \exp(-n(H(X, Y) - \delta))$$

Proof. 1. By the preceding corollary.

2. To prove the second assertion, use the above to get

$$1 - \delta < \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(F_\delta^n) = \sum_{(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n} \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(x_1, \dots, x_n, y_1, \dots, y_n)$$

By the definition of a jointly typical sequence, if $(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n$, then

$$-\frac{1}{n} \ln \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(x_1, \dots, x_n, y_1, \dots, y_n) > H(X_1, Y_1) - \delta$$

so, by exponentiating,

$$\begin{aligned}
1 - \delta &< \sum_{(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n} \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(x_1, \dots, x_n, y_1, \dots, y_n) \\
&< \sum_{(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n} e^{-n(H(X_1, Y_1) - \delta)} \\
&= |F_\delta^n| e^{-n(H(X_1, Y_1) - \delta)}
\end{aligned}$$

and hence $(1 - \delta) \exp(n(H(X, Y) - \delta)) < |F_\delta^n|$.

3. By the definition of a jointly typical sequence, if $(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n$, then

$$-\frac{1}{n} \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(x_1, \dots, x_n, y_1, \dots, y_n) < H(X_1, Y_1) + \delta$$

so, by exponentiating,

$$\begin{aligned}
1 &\geq \sum_{(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n} \mathbb{P}_{X_1, \dots, X_n, Y_1, \dots, Y_n}(x_1, \dots, x_n, y_1, \dots, y_n) \\
&> \sum_{(x_1, \dots, x_n, y_1, \dots, y_n) \in F_\delta^n} e^{-n(H(X_1, Y_1) + \delta)} \\
&= |F_\delta^n| e^{-n(H(X_1, Y_1) + \delta)}
\end{aligned}$$

and thus $|F_\delta^n| < \exp(n(H(X, Y) + \delta))$.

4. The last assertion follows directly from the definition of a jointly typical sequence, and was used in proving both parts of the second assertion. □

3.1.12 Question. What is the transmission rate of a channel?

3.1.13 Definition (capacity). Given a discrete memoryless channel γ with transition probabilities $\{\mathbb{W}(\bullet|a)\}_{a \in \mathbb{A}}$, then we define the capacity of γ to be $C = \max_{\mathbb{P}_X} I(X; Y)$ where $Y = \gamma(X)$.

3.1.14 Theorem (Channel coding). Consider a discrete memoryless channel $\gamma : \mathbb{A} \times \Omega \rightarrow \mathbb{B}$, and $C = \max_{\mathbb{P}_X} I(X; Y)$ with $Y = \gamma(X)$. Let $\epsilon > 0$, $4\epsilon > \tau > 0$. Then there is a sequence of (n, m_n) fixed-length transmission codes $(\mathcal{C}_n, \phi_n, \psi_n)$ such that for sufficiently large n , $\frac{1}{n} \ln m_n > C - \tau$ and the averaged error probability $P_e(\psi_n \circ \gamma \circ \phi_n \neq id) < \epsilon$.

Proof. For this proof, we will abbreviate (X_1, \dots, X_n) as $X^{\otimes n}$, (Y_1, \dots, Y_n) as $Y^{\otimes n}$, and $(X_1, \dots, X_n, Y_1, \dots, Y_n)$ as $(X, Y)^{\otimes n}$. By choosing N_0 appropriately, we have $\{m_n\}$ integers with $C - \tau < \frac{1}{n} \ln m_n \leq C - \frac{\tau}{2}$. Let $\delta = \frac{\tau}{8}$. Choose \mathbb{P}_X which achieves the capacity, let \mathbb{Q}_Y be the corresponding output distribution, i.e. $\mathbb{Q}_Y(b) = \sum_{a \in \mathbb{A}} \mathbb{P}_X(a) \mathbb{W}(b|a)$. We demonstrate the existence of a code sequence in 3 steps.

Step 1: Select m_n input sequences $\{c_j\}_{j=1}^{m_n}$ of length n from \mathbb{A} randomly, according to $\mathbb{P}_{X^{\otimes n}}$. Note that duplicates are possible. Identify $\mathcal{C}_n = \{1, 2, \dots, m_n\}$ with codes sequences $\phi_n(j) = c_j$ and choose

$$\psi_n(y) = \begin{cases} j & \text{if } \phi(j) = c_j, (c_j, y) \in F_\delta^n, \text{ and } \nexists(c', y) \in F_\delta^n \text{ with } c' \neq c_j \\ \text{arbitrary} & \text{otherwise} \end{cases}$$

Step 2: We estimate the averaged error probability for the randomly generated code. Let $\Lambda(c_m)$ be the probability of having a decoding error given fixed input sequence c_m . The error could come from $(c_m, y) \notin F_\delta^n$ (not being typical) or from multiple source symbols mapped to y within F_δ^n . We have

$$\Lambda(c_m) \leq \sum_{\substack{y \in \mathbb{B}^n \\ (c_m, y) \notin F_\delta^n}} \mathbb{W}(y|c_m) + \sum_{m' \neq m} \sum_{\substack{y \in \mathbb{B}^n \\ (c_{m'}, y) \in F_\delta^n}} \mathbb{W}(y|c_m)$$

with the notation $y = (y_1, \dots, y_n)$ with $\mathbb{W}(y|c_m) = \prod_{k=1}^n \mathbb{W}(y_k|(c_m)_k)$.

If we average with respect to the choice of codewords, governed by $\mathbb{P}_{X^{\otimes n}}$, then we obtain the expected error probability.

$$\begin{aligned} \mathbb{E}_{X^{\otimes n}}[\Lambda(c_m)] &= \sum_{a \in \mathbb{A}^n} \mathbb{P}_{X^{\otimes n}}(a) \Lambda(a) \\ &\leq \sum_{a \in \mathbb{A}^n} \left(\sum_{\substack{y \in \mathbb{B}^n \\ (a, y) \notin F_\delta^n}} \mathbb{P}_{X^{\otimes n}}(a) \mathbb{W}(y|a) + \sum_{m' \neq m} \sum_{\substack{y \in \mathbb{B}^n \\ (c_{m'}, y) \in F_\delta^n}} \mathbb{P}_{X^{\otimes n}}(a) \mathbb{W}(y|a) \right) \\ &= \mathbb{P}_{(X, Y)^{\otimes n}}((F_\delta^n)^c) + \sum_{a \in \mathbb{A}^n} \sum_{m' \neq m} \sum_{\substack{y \in \mathbb{B}^n \\ (c_{m'}, y) \in F_\delta^n}} \mathbb{P}_{X^{\otimes n}, Y^{\otimes n}}(a, y) \\ &= \mathbb{P}_{(X, Y)^{\otimes n}}((F_\delta^n)^c) + \sum_{m' \neq m} \sum_{\substack{y \in \mathbb{B}^n \\ (c_{m'}, y) \in F_\delta^n}} \mathbb{P}_{Y^{\otimes n}}(y) \end{aligned}$$

To be continued...