| EE376A/STATS376A Information Theory |
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| Lecture 5: Variable Length Lossless Compression 5-01/23/2018 |
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In this lecture we investigate strategies for variable length lossless compression. Several variable length encoding schemes are shown and their merits are discussed. Then a procedure for producing prefix codes for a dyadic source is presented. Lastly, Shannon Codes are presented as a generalization of the dyadic encoding procedure.

## 1 Review From Previous Class

$$
\begin{equation*}
U_{1}, U_{2}, \ldots U_{n} \sim \operatorname{iid} U \in \mathcal{U} \tag{1}
\end{equation*}
$$



Figure 1: Asymptotic Equipartition Property

$$
\begin{align*}
P\left(U^{n} \in A_{\epsilon}^{(n)}\right) & \approx 1  \tag{2}\\
\forall u^{n} \in A_{\epsilon}^{(n)}: p\left(u^{n}\right) & \approx 2^{-n H(U)}  \tag{3}\\
\left|A_{\epsilon}^{(n)}\right| & \approx 2^{n H(U)} \tag{4}
\end{align*}
$$

This tells us that we need $H(U)$ bits per source symbol for a near-lossless fixed length scheme. If fewer bits are used (say $\alpha<H(U)$ bits), then the size of the set encoded by those $\alpha$ bits is given by

$$
\left|B_{n}\right|=2^{n \alpha}
$$

and we showed last time that the probability of a source sequence being within the set $B_{n}$ goes to 0 and hence such a scheme makes an error with probability close to 1 .

## 2 Variable Length Lossless Compression Examples

### 2.1 Example 1

Given an alphabet with four letters in it

$$
\mathcal{U}=\{a, b, c, d\}
$$

The probability of each letter and a coding scheme is given in the table below:

| $u$ | $p(u)$ | codeword $c(u)$ | length $l(u)$ |
| :---: | :---: | :---: | :---: |
| a | $1 / 2=2^{-1}$ | 0 | 1 |
| b | $1 / 4=2^{-2}$ | 10 | 2 |
| c | $1 / 8=2^{-3}$ | 110 | 3 |
| d | $1 / 8=2^{-3}$ | 111 | 3 |

The code is $\{c(u)\}_{u \in \mathcal{U}}$. Note that any code can be represented by the nodes in a binary tree. For example, a binary tree representation of this code is shown in figure 2 .


Figure 2: Binary tree representation
Note: For this code, $l(u)=\log \frac{1}{p(u)}$, therefore

$$
\begin{equation*}
\bar{l}=\mathbb{E}[l(U)]=H(U) \tag{5}
\end{equation*}
$$

Thus, this is a lossless scheme that achieves the ideal bits per source symbol, the entropy $H(U)$ (we'll show in the next lecture that the entropy is fundamental limit of compression for variable length lossless schemes as well). Additionally, it is a prefix code which makes the encoding and decoding process very simple. Note that a code is a prefix code if and only if all the codewords are leaves in the binary tree representation of the code.

### 2.2 Example 2

Consider a new scheme for the same source (binary tree representation shown in figure 3).

| $U$ | codeword $c(U)$ | length $l(u)$ |
| :---: | :---: | :---: |
| a | 0 | 1 |
| b | 1 | 1 |
| c | 01 | 2 |
| d | 11 | 3 |



Figure 3: The binary tree representation of the code

Note: The length of the code is $\bar{l}<H(U)$, i.e. better than the entropy. But, consider the encoding of three source symbols $a b d$ and $c b b$ :

$$
\begin{align*}
& a b d \xrightarrow{\text { encoded to }} 0111  \tag{6}\\
& c b b \xrightarrow{\text { encoded to }} 0111 \tag{7}
\end{align*}
$$

Thus, different source sequences can have the same encoding, and this code cannot be decoded uniquely.
Definition 1. A code is uniquely decodable (UD) if every sequence of source symbols is mapped to a distinct binary representation.

Definition 2. A prefix code is a code where no codeword is the prefix of any other.
Note: A prefix code is uniquely decodable and can be decoded efficiently and on the fly (i.e. without needing entire binary sequence before decoding starts)

### 2.3 Exercise

Consider the code for the same source

| $U$ | codeword $c(U)$ | length $l(u)$ |
| :---: | :---: | :---: |
| a | 10 | 2 |
| b | 00 | 2 |
| c | 11 | 2 |
| d | 110 | 3 |

Show that, though this code is not a prefix code, it is actually uniquely decodable. For the purpose of this exercise, you don't need to know the source distribution.

## Proof

Part of HW 2.

## 3 Prefix code for dyadic distributions

Definition 3. A source is dyadic if

$$
\begin{equation*}
p(u)=2^{-n_{u}} \tag{8}
\end{equation*}
$$

where $n_{u}$ is an integer $\forall u \in \mathcal{U}$.
Suppose we find a code such that

$$
\begin{equation*}
l(u)=n_{u}=\log \frac{1}{p(u)} \tag{9}
\end{equation*}
$$

then

$$
\begin{equation*}
\bar{l}=\mathbb{E}[l(U)]=H(U) \tag{10}
\end{equation*}
$$

Before proceeding to a prefix code for dyadic sources, we prove a technical lemma which says that the number of symbols with the lowest probability is even for a dyadic source. In particular, a dyadic source cannot have a single symbol with the lowest probability.

Lemma 4. Assume $\mathcal{U}$ is dyadic with $|\mathcal{U}| \geq 2$, and let $n_{\max }=\max _{u \in \mathcal{U}} n_{u}$. Then the number of symbols $u \in \mathcal{U}$ with $n_{u}=n_{\text {max }}$ is even.

Proof:

$$
\begin{align*}
1 & =\sum_{u} p(u)  \tag{11}\\
& =\sum_{u} 2^{-n_{u}}  \tag{12}\\
& =\sum_{n=1}^{n_{\max }}\binom{\# \text { of symbols } u}{\text { with } n_{u}=n} 2^{-n} \tag{13}
\end{align*}
$$

Multiplying both sides by $2^{n_{\text {max }}}$,

$$
\begin{align*}
\underbrace{2^{n_{\max }}}_{\text {even }} & =\sum_{n=1}^{n_{\max }}\binom{\# \text { of symbols } u}{\text { with } n_{u}=n} \cdot 2^{n_{\max }-n}  \tag{15}\\
& =\underbrace{\sum_{n=1}^{n_{\max }-1}\binom{\# \text { of symbols } u}{\text { with } n_{u}=n} \cdot \underbrace{n_{\max }-n}_{\text {even }}}_{\text {even }}+\binom{\# \text { of symbols } u}{\text { with } n_{u}=n_{\max }} \cdot 1 \tag{16}
\end{align*}
$$

Therefore, the number of letters $u$ with $n_{u}=n_{\max }$ must be even.

### 3.1 Procedure for constructing prefix codes

Consider the following procedure:

- Choose 2 symbols with $n_{u}=n_{\text {max }}$ and merge them into a single symbol.
- We now have a new symbol with twice the probability.
- The new source is also dyadic.
- So repeat the first step until we are left with just one symbol.

Note: This procedure induces a binary tree and the codewords are the leaves of the constructed binary tree. As noted earlier, a code with all codewords on the leaf nodes is a prefix code.

### 3.2 Example of procedure



Figure 4: Procedure for generating a prefix code
Note: If $p(u)=2^{-n_{u}}$, then the node $u$ will participate in $n_{u}$ merges (since each merge doubles the probability). Thus, the distance of $u$ from the root is $n_{u}$, which is also equal to $l(u)$.
Conclusion: Our procedure yields a prefix code with

$$
\begin{equation*}
l(u)=n_{u}=\log \frac{1}{p(u)} \tag{17}
\end{equation*}
$$

and, in particular, $\bar{l}=H(U)$.

## 4 Shannon Codes

For a general source let

$$
\begin{equation*}
n_{u}^{*}=\left\lceil\log \frac{1}{p(u)}\right\rceil \quad \forall u \in \mathcal{U} \tag{18}
\end{equation*}
$$

## Note:

$$
\begin{align*}
\sum_{u \in \mathcal{U}} 2^{-n_{u}^{*}} & =\sum_{u \in \mathcal{U}} 2^{-\left\lceil\log \frac{1}{p(u)}\right\rceil}  \tag{19}\\
& \leq \sum_{u \in \mathcal{U}} 2^{-\log \frac{1}{p(u)}}  \tag{20}\\
& =\sum_{u \in \mathcal{U}} p(u)=1 \tag{21}
\end{align*}
$$

Consider a new source $p^{*}(u)=2^{-n_{u}^{*}}$. While $p^{*}(u)$ does not sum to 1 over $\mathcal{U}$, we can add new symbols to extend the source to $\mathcal{U}^{*} \supseteq \mathcal{U}$ such that $p^{*}(u)$ is dyadic over $\mathcal{U}^{*}$ and $\sum_{u \in \mathcal{U}^{*}} p^{*}(u)=1$.
Using the technique in the previous section, we can now construct a prefix code for the source $p^{*}$ such that

$$
\begin{equation*}
l(u)=\log \frac{1}{p^{*}(u)}=n_{u}^{*}=\left\lceil\log \frac{1}{p(u)}\right\rceil \quad \forall u \in \mathcal{U} \tag{22}
\end{equation*}
$$

Note that the code is defined over $\mathcal{U}^{*}$ but we only care about the symbols in $\mathcal{U}$.
Note: This code is known as the Shannon Code.

The expected length for this code is

$$
\begin{align*}
\bar{l} & =\sum_{u} p(u) l(u)  \tag{23}\\
& =\sum_{u} p(u)\left\lceil\log \frac{1}{p(u)}\right\rceil  \tag{24}\\
& \leq \sum_{u} p(u)\left(\log \frac{1}{p(u)}+1\right)  \tag{25}\\
& =H(U)+1 \tag{26}
\end{align*}
$$

So we are within 1 bit per source symbol of the entropy. To get even closer to the entropy, we can work with blocks of source symbols.

For the memoryless source $U_{1}, U_{2}, \ldots$ iid $\sim \mathcal{U}$, we can construct a Shannon code for $U^{N}=\left(U_{1}, U_{1}, \ldots, U_{N}\right)$. Then

$$
\begin{equation*}
\frac{1}{N} \mathbb{E}\left[l\left(U^{N}\right)\right] \leq \frac{1}{N}\left(H\left(U^{N}\right)+1\right)=H(U)+\frac{1}{N} \tag{27}
\end{equation*}
$$

Thus as $N$ becomes large, the bits used per source symbol tends toward the entropy.

