

LECTURE 5

TODAY : - MARKOVIAN SOURCES

- LZ Compression Theorem

Source: My notes from Spring 2006 → Gallager's Notes 1994

RECALL MARKOVIAN SOURCES (= Hidden Markov Model):

if $X_1 \dots X_n \dots$ is a hidden Markov Model
 if $Z \in \mathcal{M}$ (rxn state space)

- Hidden Markov Model given by matrix $M \in \mathbb{R}^{k \times k}$
 & $P_x^{(1)} \dots P_x^{(k)}$ dist on \mathcal{X}

st.- Output of HMM = $X_1 \dots X_n \dots$ ~~the~~ generated as follows

where ~~X_i~~ ~~is~~ $Z_1 \sim \pi(m)$
 \uparrow
 stationary

$Z_1, Z_2 \dots Z_n \dots$ generated according to \uparrow
 $Z_{i-1} \xrightarrow{M} Z_i$

$X_i \sim P_x^{(z_i)}$

[all mc's in lecture "irreducible" + "aperiodic"]

Entropy of Chain:

$$\lim_{n \rightarrow \infty} H(X_n | X_1 \dots X_{n-1}) \triangleq H(M)$$

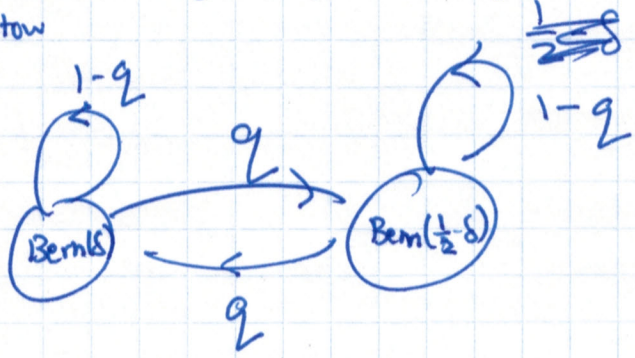
⑧ limit exists since

$$\begin{aligned}
 H(X_n | X_1 \dots X_{n-1}) &\leq H(X_n | X_2 \dots X_{n-1}) \\
 &= H(X_{n-1} | X_1 \dots X_{n-2})
 \end{aligned}$$

} time invariant

Note limit exists but not known to be computable.

Eg. $H(M) = ?$ (as function of q, δ).
below



Theorem: \forall Markov chain M , $\forall \epsilon$, $\exists n_0 \forall n \geq n_0$

$$Pr_{X_1 \dots X_n} \left[|LZ(X_1 \dots X_n) - (H(M) + \epsilon) \cdot n| \leq \epsilon \right]$$

[Very "qualitative". Finite length analysis = "great project"]

Key Ingredients in Proof:

- ① Universal-Huffman-Compressor achieves "optimal compression"
- ② \Rightarrow Finite State Compressor Compresses.
- ③ LZ no worse than finite state compressor.



"Universal-Huffman-Compressor" $C_{Huff, \ell} (X_1 \dots X_n); X_i \in \Omega$

① ~~Comp~~ for ~~every~~
divide $X = B_1 \dots B_{n/\ell}$ $B_i \in \Omega^\ell$

② ~~Can~~ Compute frequencies ~~for~~ $\{f_w\}_{w \in \Omega^\ell}$
among $B_1 \dots B_{n/\ell}$

③ ~~Huffman-code~~ $(B_1 \dots B_{n/\ell};$
 $Z_i = \text{Huffman}(B_i; \{f_w\})$

④ Output $\{f_w\}$
 $Z_1 \dots Z_{n/\ell}$

Theorem': \forall m.c. M , $\forall \epsilon$, $\exists n_0 \forall n \geq n_0$

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$$\Pr_{x_1 \dots x_n} \left[\left| \text{Univ. Huff}(x_1 \dots x_n) \right| > (H(M) + \epsilon) \cdot n \right] < \epsilon$$

key insight to Proof of Theorem': AEP holds for Markov Sources

AEP Theorem: \forall m.c. M , $\forall \epsilon \exists l$ & set

$S \subseteq \Sigma^l$ s.t. the following hold

① $|S| \leq \frac{(H(M) + \epsilon) \cdot n}{2}$

~~② $\frac{1}{2^{(H(M) + \epsilon) \cdot n}} \leq \Pr_{(x_1 \dots x_n) \in S} [(x_1 \dots x_n) \in S]$~~

② $\forall (x_1 \dots x_n) \in S$

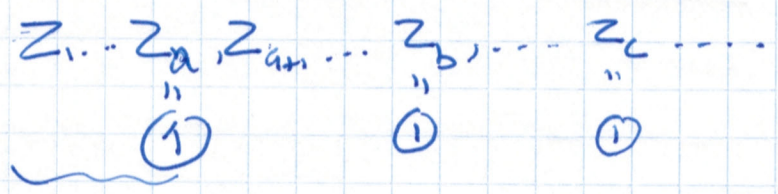
$$\frac{1}{2^{(H(M) + \epsilon) \cdot n}} \leq \Pr [(x_1 \dots x_n) = (x_1 \dots x_n)] \leq \frac{1}{2^{(H(M) - \epsilon) \cdot n}}$$

③ $\Pr_{(x_1 \dots x_n)} [(x_1 \dots x_n) \notin S] \leq \epsilon$

Proof of AEP Theorem: Omitted

Ideas: ~~Consider states~~ $Z_e, Z_{2e}, Z_{3e}, \dots$

① Prove for Markov Chains by decoupling walk into sequences that end at state ① & do not contain ① in between. Sequences now i.i.d.



↑ lengths i.i.d.
 Expected entropy iid → both converge } → ratio converges

② To get to Hannon's, insert $Z_e, Z_{2e} \dots$ at periodic intervals; results in slightly higher entropy but not much higher; now we have Markov chain!

AEP Theorem \Rightarrow Theorem'

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Essentially implies w.p. $1-\epsilon$ we get strings ^{B:} of frequency

$\approx 2^{-H(m) \cdot l} \Rightarrow$ Shannon coding will associate strings
of length $\leq (H(m) + \epsilon) \cdot l$

w.p. ϵ use string of length $\approx l$

\Rightarrow Expected compression length

$$= \frac{n}{l} (\epsilon + H(m) + \epsilon) l$$

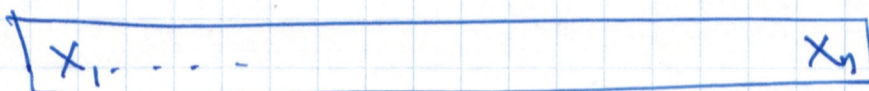
$$= (H(m) + 2\epsilon) n$$

□

[Full proof will involve union bound over $S \dots$]

Part II: Finite State Compressors:

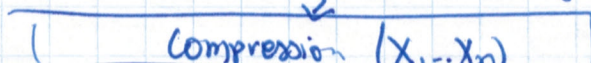
$C =$ finite state compressor if



↑ ← left to right read access



↓ ← left to right write access.



can be converted to

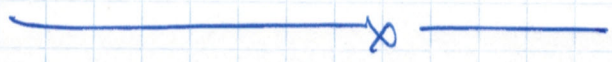
Proposition: Univ.-Huffman is an Ω^L -state Finite State Compressor. [with preprocessing of M].

Proof: ~~Figure out~~ ~~before~~ expected (proof obvious)

Proof:



Theorem' \Rightarrow M can be compressed by Ω^L -state compressor.



Theorem'': if M compressed by S-state compressor then M compressed by Lempel-Ziv.



Key Ingredient in this proof

$$C(X_1 \dots X_n) \cong \max \left\{ \frac{n}{|T|} \mid \exists Y_1 \dots Y_{|T|} \text{ distinct in } \Sigma^* \text{ s.t. } X_1 \dots X_n = Y_1 \circ Y_2 \circ \dots \circ Y_{|T|} \right\}$$

Intuition:

~~Rough~~ ~~idea~~ (1) $C(X_1 \dots X_n) \approx \frac{n}{\log n}$ \Rightarrow $X_1 \dots X_n$ not compressible by any thing

(2) $C(X_1 \dots X_n) = \text{small} \Rightarrow$ LZ compresses well.

Formal Claims

Let $t = C(x_1 \dots x_n)$ then

$$(1) \quad \cancel{t} n \geq t \log \frac{t}{4} \quad \text{(intermediate)} \\ \text{[} t \text{ is growing]}$$

(2) $\forall s$ state compressors C^s

$$|C^s(x_1 \dots x_n)| \geq t \cdot \log \left(\frac{t}{s^2} \right)$$

$$= t \log t - t \log s^2$$

$$\geq (1 - o(1)) t \log t$$

$$\left[\begin{array}{l} \text{if } s = O(1) \\ \& t = \omega(1) \end{array} \right]$$

$$(3) \quad |C^{LZ}(x_1 \dots x_n)| \leq (1 + o(1)) t \log t$$

$$\Rightarrow |C^{LZ}(x_1 \dots x_n)| \leq (1 + o(1)) \cdot |C^s(x_1 \dots x_n)|$$

$$\leq (1 + o(1)) \cdot H(m) \cdot n \quad \left[\begin{array}{l} \text{using} \\ \text{Theorem'} \end{array} \right]$$

Proofs of Claims

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① Follow from # strings of length $l \leq \Omega^l$;

$Y_1 \dots Y_t$ distinct

$$\Rightarrow \sum |Y_i| = \sum_k k \cdot \#\{i \text{ s.t. } |Y_i| = k\}$$
$$\geq \sum_{k=0}^{\lfloor \log_{\Omega} l \rfloor} k \cdot \Omega^k + (l - \Omega^{\lfloor \log_{\Omega} l \rfloor}) (\log_{\Omega} l + 1)$$

$$\geq l \cdot \log_{\Omega} l \cdot (1 - o(1))$$

③ $|C^{LZ}(x_1 \dots x_n)| \leq t(\lceil \log t \rceil + 1)$

④ $\leq t \cdot (1 + o(1)) \cdot \log t$

② let $x_1 \dots x_n = Y_1^0 Y_2^0 \dots Y_t^0$, Y_i distinct

- let $Y_1' \dots Y_t'$ be outputs of FSM while traversing $Y_1 \dots Y_t$.

- Note $Y_1' \dots Y_t'$ need not be distinct;

However (key) if $Y_i^a = Y_j^b$ both start in state a & end in b , then $Y_i' \neq Y_j'$

- $T_{ab} \triangleq \{ j \mid \text{FSM starts at } a \text{ \& ends at } b \text{ on } Y_j \}$

$$- L_{ab} \triangleq \sum_{j \in T_{ab}} |Y_j|$$

- $|L_{ab}| \geq |T_{ab}| \log \frac{|T_{ab}|}{4}$ (log (1) applied to binary strings).

$$- |C^S(X_1 \dots X_n)| = \sum_{j=1}^n |Y_j|$$

$$= \sum_{a,b} L_{ab}$$

$$\geq \sum_{a,b} |T_{ab}| \log \frac{|T_{ab}|}{4}$$

$$\geq \frac{t}{|S|^2} \cdot \log \frac{t}{|S|^2 \cdot 4} \cdot |S|^2$$

$$= t \cdot \log \frac{t}{|S|^2 \cdot 4} \quad \square$$