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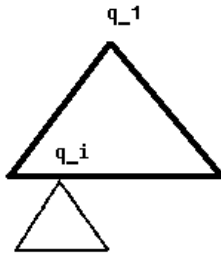
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Document URL: http://www.stanford.edu/~tanqf/lowerbound_1.pdf (old URL)

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Motivation: Since Gallager upperbounded the Redundancy using P_{\max} , probability of most likely source letter, we consider a dual problem of lowerbounding the Redundancy using P_{\min} , probability of the least likely source letter.

Diagram:



Denote $tree^{q_i}$ as the tree formed by collapsing everything below the node with probability q_i to just a point. Similarly denote $tree_{q_i}$ as the subtree under the node with probability q_i .

By the law of total probability we have $L = L(tree^{q_i}) + q_i L(tree_{q_i})$

For the entropy, we can write it as

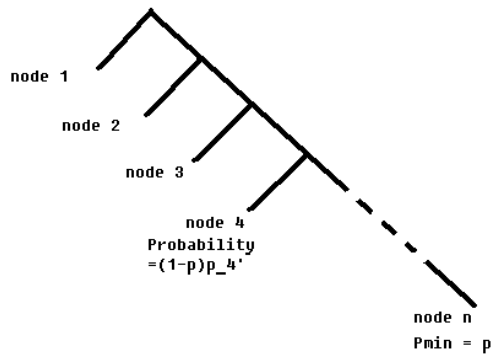
$$\begin{aligned} H &= -\sum_k q_k \log q_k \\ &= -\sum_{k \leq i} q_k \log q_k - \sum_{k > i} q_k \log q_k \\ &= -\sum_{k \leq i} q_k \log q_k - \sum_{k > i} \frac{q_k}{q_i} \log \frac{q_k}{q_i} + q_i \log q_i \end{aligned}$$

Thus we see that

$$\begin{aligned}
 R(\text{tree}) &= L - H \\
 &= L(\text{tree}^{q_i}) + q_i L(\text{tree}_{q_i}) - \left[-\sum_{k \leq i} q_k \log q_k - \sum_{k > i} \frac{q_k}{q_i} \log \frac{q_k}{q_i} + q_k \log q_i \right] \\
 &\geq R(\text{tree}^{q_i})
 \end{aligned}$$

Thus to find a lower bound, we can collapse the tree to the following form:

Diagram:



Where $P_{\min} = p$. Suppose each leaf occurs with probability $p_i (1-p)$, where

$$\sum_{i=1}^{n-1} p_i = 1.$$

We can write the redundancy as

$$\begin{aligned}
 R &= L - H \\
 &= \left[pn + \sum_{i=1}^{n-1} p_i (1-p)i \right] - \left[-p \log_2 p - \sum_{i=1}^{n-1} p_i (1-p) \log_2 p_i (1-p) \right] \\
 &= \left[pn + \sum_{i=1}^{n-1} p_i (1-p)i \right] - \left[-p \log_2 p - \sum_{i=1}^{n-1} p_i (1-p) \log_2 p_i - \sum_{i=1}^{n-1} p_i (1-p) \log_2 (1-p) \right] \\
 &= \left[pn + \sum_{i=1}^{n-1} p_i (1-p)i \right] - \left[-p \log_2 p - (1-p) \sum_{i=1}^{n-1} p_i \log_2 p_i - (1-p) \log_2 (1-p) \sum_{i=1}^{n-1} p_i \right] \\
 &= \left[pn + \sum_{i=1}^{n-1} p_i (1-p)i \right] - \left[(1-p)H(p_1, \dots, p_{n-1}) + H_{\text{binary}}(p) \right] \\
 &= pn - H_{\text{binary}}(p) + (1-p) \left[\sum_{i=1}^{n-1} p_i i + H(p_1, \dots, p_{n-1}) \right]
 \end{aligned}$$

Thus we have the following problem:

$$\min_{p_1, \dots, p_{n-1}} R \Rightarrow \min_{p_1, \dots, p_{n-1}} \sum_{i=1}^{n-1} p_i i + H(p_1, \dots, p_{n-1})$$

To incorporate the condition that $\sum_{i=1}^{n-1} p_i = 1$ we use Lagrange multipliers:

$$\text{Write } f = \sum_{i=1}^{n-1} p_i i + H(p_1, \dots, p_{n-1}) + \lambda(1 - \sum_{i=1}^{n-1} p_i).$$

Note that if we do $\frac{\partial f}{\partial \lambda} = 0 \Rightarrow \sum_{i=1}^{n-1} p_i = 1$ we get back our original constraint.

By the symmetry of the problem, we need only to take the partial derivative with respect to lets say p_m for some $0 \leq m \leq n-1$.

$$\begin{aligned} \frac{\partial f}{\partial p_m} &= 0 \\ \Rightarrow m - \log_2 p_m - \log_2 e - \lambda &= 0 \\ \Rightarrow \log_2 p_m &= m - \log_2 e - \lambda \\ \Rightarrow p_m &= 2^{m - \log_2 e - \lambda} \end{aligned}$$

Plugging this back into our constraint $\sum_{i=1}^{n-1} p_i = 1$ we get

$$\begin{aligned} \sum_{i=1}^{n-1} 2^{i - \log_2 e - \lambda} &= 1 \\ \Rightarrow (2^{-\log_2 e - \lambda}) \sum_{i=1}^{n-1} 2^i &= 1 \\ \Rightarrow (2^{-\log_2 e - \lambda})(2^n - 1) &= 1 \\ \Rightarrow \lambda &= \log_2 \left[(2^n - 1) \left(\frac{1}{e} \right) \right] \end{aligned}$$

Putting this into $p_m = 2^{m - \log_2 e - \lambda}$ we get

$$p_m = 2^{m - \log_2 e - \log_2 \left[(2^n - 1) \left(\frac{1}{e} \right) \right]} = \frac{2^m}{2^n - 1}.$$

Thus we can see that the distribution that minimizes the redundancy is given by

$$\frac{2}{2^n - 1}(1-p), \frac{2^2}{2^n - 1}(1-p), \dots, \frac{2^{n-1}}{2^n - 1}(1-p), p$$

Putting this back into $R = pn - H_{\text{binary}}(p) + (1-p) \left[\sum_{i=1}^{n-1} p_i^i + H(p_1, \dots, p_{n-1}) \right]$ we get

R

$$\begin{aligned}
 &= pn - H_{\text{binary}}(p) + (1-p) \left[\sum_{i=1}^{n-1} \frac{2^i}{2^n - 1} i - \sum_{i=1}^{n-1} \frac{2^i}{2^n - 1} \log_2 \frac{2^i}{2^n - 1} \right] \\
 &= pn - H_{\text{binary}}(p) + (1-p) \left[\sum_{i=1}^{n-1} \frac{2^i}{2^n - 1} i - \sum_{i=1}^{n-1} \frac{2^i}{2^n - 1} i + \log_2(2^n - 1) \right] \\
 &= pn - H_{\text{binary}}(p) + (1-p) \log_2(2^n - 1)
 \end{aligned}$$

which is the lower bound we desire.