Information Theory over Multisets

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Outline

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Communication model [1]

Information Source

Encoder Signal received Dencoder Moise Source

Noise Source

Short review of Shannon information theory

- Established the fundamental natural limits on communication
- **▶** Source entropy [1]

$$H(X) = \sum_{i} P_i H_i = -\sum_{i,j} P_i p_i(j) \log p_i(j)$$
 (1)

► Channel capacity [1] The capacity C of a discrete channel is given by

$$C = \lim_{T \to \infty} \frac{\log N(T)}{T}$$

where N(T) is the number of allowed signals of duration T.

Multiset source entropy

Consider a discrete information source which produces multiset messages:

- A message is a multiset of symbols.
- ▶ A multiset is a string equivalence class.
- ▶ The entropy rate of such a source is proved to be zero in [2]:

$$H(X_{multiset}) = \lim_{n \to \infty} \frac{1}{n} H(\{X_i\}_{i=1}^n) = 0$$

Information content of a multiset

The information content of an outcome (multiset) x is

$$h(x) = \log \frac{1}{P(x)} = \log \frac{\prod_{i=1}^{n} m_i!}{(\sum_{i=1}^{n} m_i)! \prod_{i=1}^{n} p_i^{m_i}}$$

Definition according to [3].

Proof.

$$h(x = x_1^{m_1} x_2^{m_2} \dots x_n^{m_n}) = \log \frac{1}{P[x]} = \log \frac{1}{m_1, m_2, \dots, m_n} \prod_{i=1}^n p_i^{m_i} = \log \left(\frac{1}{\sum_{i=1}^n m_i !} \prod_{i=1}^n p_i^{m_i} \right) = \log \frac{\prod_{i=1}^n m_i !}{(\sum_{i=1}^n m_i)! \prod_{i=1}^n p_i^{m_i}} = \log \frac{\prod_{i=1}^n m_i !}{(\sum_{i=1}^n m_i)! \prod_{i=1}^n p_i^{m_i}}$$

Encoding length

We consider a set X of N symbols, an alphabet A, and the length of encoding I, therefore:

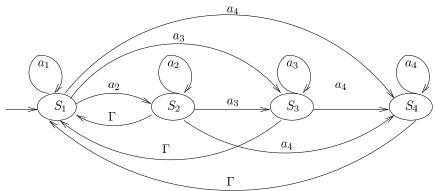
$$X = \{x_i = a_1^{n_1} a_2^{n_2} \dots a_b^{n_b} \mid \sum_{j=1}^b n_j = I, a_j \in A\}$$

Theorem

Non-uniform encodings over multisets are shorter than uniform encodings over multisets.

Channel Capacity in base 4

We consider that a sequence of multisets is transmitted along the channel. The capacity of such a channel is computed for base 4, then some properties of it for any base are presented.



According to Shannons' Capacity Theorem we get $b_{ij}^{(a_k)} = t_k$ because we consider that the duration to produce a_k is the same for each $(i,j) \in E$. The determinant equation is

Channel Capacity

Theorem

The multiset channel capacity is zero, C = 0.

$$\begin{vmatrix} W^{-t_1} - 1 & W^{-t_2} & W^{-t_3} & \cdots & W^{-t_b} \\ 0 & W^{-t_2} - 1 & W^{-t_3} & \cdots & W^{-t_b} \\ 0 & 0 & W^{-t_3} - 1 & \cdots & W^{-t_b} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \cdots & 0 & W^{-t_{b-1}} - 1 & W^{-t_b} \\ 0 & 0 & 0 & \cdots & W^{-t_b} - 1 \end{vmatrix} = 0$$

$$W = \frac{1}{\sqrt[t]{x}} \quad \Rightarrow \quad C = -\frac{1}{t} \log_b x. \tag{2}$$

$$(1-x)^b=0 \Rightarrow W=1 \Rightarrow C=0.$$

Conclusion

- we derive a formula for the information content of a multiset
- as future work:
 - further explore the properties of multiset-based communication systems
 - compare these to similar results for string-based communication systems

References

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