

An inner bound technique on the normalized set of entropy vectors

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Entropy vectors

- Let $\mathbf{X} = (X_1, \dots, X_N)$ be N discrete random variables with finite support.
- Let $h(\mathbf{X}_A)$ be the entropy of the subset of rvs $\mathbf{X}_A = (X_i, i \in A)$ for some non-empty subset $A \subseteq \{1, \dots, N\} \equiv [N]$.
- Let $\mathbf{h} = (h(\mathbf{X}_A), A \subseteq [N])$ be the vector of entropies of each non-empty subset $A \subseteq [N]$. Note \mathbf{h} has $2^N - 1$ entries.
 - Example: for $N = 3$, $\mathbf{h} = (h_1, h_2, h_3, h_{12}, h_{13}, h_{23}, h_{123})$.
- A vector $\mathbf{h} \in \mathbb{R}^{2^N - 1}$ is called entropic if its elements are the entropies for some joint distribution \mathbf{p} on the N rvs \mathbf{X} .
- The entropy vector region (EVR) Γ_N^* is the set of all entropic vectors.
- Normalize by the number of bits for the support m : $\bar{\mathbf{h}} = \mathbf{h}/\log_2 m$, and define Ω_N^* as the set of normalized entropy vectors (Hassibi and Shadback 2007).

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Two representations of joint distributions on binary rvs

Lemma 1. The joint distribution \mathbf{p} on N binary rvs

$$\mathbf{p} = (p_{[N]}(\mathbf{x}) = \mathbb{P}(\mathbf{X} = \mathbf{x}), \mathbf{x} \in \{0, 1\}^N),$$

and the collection of probabilities of each subset of rvs each taking the value zero

$$\mathbf{q} = (p_A(\mathbf{0}) = \mathbb{P}(\mathbf{X}_A = \mathbf{0}), A \subseteq [N])$$

are equivalent in that there is a bijection between them.

Specifying the zero probabilities for all strict subsets leaves one degree of freedom, the joint probability of zero.

Example: $N = 2$ rvs. There exists an invertible matrix such that $\mathbf{q} = \mathbf{M}\mathbf{p}$:

$$\begin{bmatrix} p_{(1)}(0) \\ p_{(2)}(0) \\ p_{(1,2)}(00) \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} p_{(1,2)}(00) \\ p_{(1,2)}(01) \\ p_{(1,2)}(10) \\ p_{(1,2)}(11) \end{bmatrix}.$$

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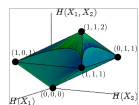
A finite terminating algorithm to determine membership in the set of binary entropy vectors

- The algorithm considers a sequence of subsets of increasing cardinality and finds the marginal distributions on these subsets consistent with the given entropy vector candidate.
- The set of all possible binary distributions on a given set $B \subseteq A$ of size $|B| = k - 1$ consistent with the given entropies is stored in the set \mathcal{Q}_B .
- Find candidate joint distributions \mathbf{p}_A on \mathbf{X}_A consistent with their marginals \mathbf{p}_B for each $B \subseteq A$.
- For each candidate joint disbn \mathbf{p}_A set the remaining free variable using the specified entropy h_A (either 0, 1, 2 values).
- Return all joint disbns \mathbf{p} on \mathbf{X} consistent with the given entropy vector \mathbf{h} (if any).

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An easy proof of Yeung's result that $\mathcal{P}_2 = \Omega_2^*$ (and similarly $\mathcal{P}_3 = \Omega_3^*$)

- Let Φ_N be the collection of all entropy vectors for N binary rvs.
- The Shannon-type inequalities Γ_N plus cardinality constraints ($\mathcal{B}_N = \{\mathbf{h} : h_i \leq 1, i = 1, \dots, N\}$), denoted $\mathcal{P}_N = \Gamma_N \cap \mathcal{B}_N$, form a polytope in entropy space that outer bounds $\text{conv}(\Phi_N)$.
- For $N = 2$, \mathcal{P}_2 is generated by the vertices below, each of which is in Φ_2 (using the algo.), as can be seen with the following constructions:
 - (0, 0, 0): $(X_1, X_2) = (0, 0)$ with probability one.
 - (0, 1, 1): X_1 deterministic, X_2 uniform.
 - (1, 0, 1): X_1 uniform, X_2 deterministic.
 - (1, 1, 1): $X_1 = X_2$ with probability one, X_1 uniform.
 - (1, 1, 2): X_1, X_2 independent and uniform.



Therefore $\mathcal{P}_2 = \text{conv}(\Phi_2)$.

- Further, since $\text{conv}(\Phi_N) \subseteq \Omega_N^* \subseteq \mathcal{P}_N$, we have $\mathcal{P}_2 = \Omega_2^*$.
- Similar result and proof for $N = 3$ (both results originally due to Yeung).

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But, not all normalized entropy vectors are binary: $\text{conv}(\Phi_N) \neq \Omega_N^*$ for $N \geq 4$

- Enumerating the vertices of \mathcal{P}_4 gives a vertex at

$$\frac{h_1}{1} \quad \frac{h_2}{1/2} \quad \frac{h_3}{1/2} \quad \frac{h_4}{1/2} \quad \frac{h_{12}}{1} \quad \frac{h_{13}}{1} \quad \frac{h_{14}}{1} \quad \frac{h_{23}}{1} \quad \frac{h_{24}}{1} \quad \frac{h_{34}}{1} \quad \frac{h_{123}}{1} \quad \frac{h_{124}}{1} \quad \frac{h_{134}}{1} \quad \frac{h_{234}}{1} \quad \frac{h_{1234}}{1}$$

- Evaluation of this point \mathbf{h} using the membership algorithm asserts this point is not in Φ_4 .
- But, this point is in Ω_4^* using the following construction: fix Z_1, Z_2 to be independent uniform bits, then take

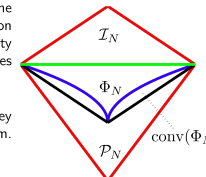
$$X_1 = (Z_1, Z_2), \quad X_2 = (Z_1 \oplus Z_2, 0), \quad X_3 = (Z_1, 0), \quad X_4 = (Z_2, 0).$$

- Can argue this point is an extreme point of Ω_4^* , and from here we can argue the result.

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An algorithm to generate an inner bound for a given outer bound

- Enumerate the extreme points of the polytope \mathcal{P}_N , by using a double description algorithm to convert the linear inequality representation into the generating vertices representation.
- For each of these vertices, determine if they lie in Φ_N , using the membership algorithm. Keep only those vertices lying in Φ_N .
- Take the convex hull of these vertices to get the polytope \mathcal{I}_N . \mathcal{I}_N can be expressed in normal linear inequality form by using the double description method again.



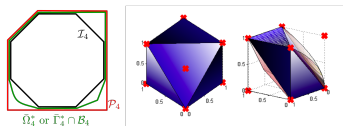
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Every Shannon type inequality is tight for $N = 4$

Running the above algorithm using the \mathcal{P}_4 outer bound gives an inner bound with the following property.

Theorem 1. For $N = 4$ every Shannon type inequality is tight on a full $(2^N - 2 = 14)$ dimensional exposed face of $\text{conv}(\Phi_4)$ and hence Ω_4^* . Furthermore, for every such full dimensional exposed face \mathcal{F}_I of $\text{conv}(\Phi_4)$ that is contained in an exposed face \mathcal{F}_P of \mathcal{P}_4 , the containment is strict ($\mathcal{F}_I \subsetneq \mathcal{F}_P$).

Thus all Shannon inequalities are necessary for $N \geq 4$, but each such Shannon inequality can be improved!



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Key properties of the inner bound algorithm

Theorem 2. Given any polytopic outer bound \mathcal{O}_N to Ω_N^* , the inner bound algorithm provides after a finite number of computations a polytopic inner bound $\mathcal{I}_N(\mathcal{O}_N)$ to $\text{conv}(\Phi_N)$ and hence Ω_N^* and Γ_N^* . Every exposed face of \mathcal{O}_N which is also an exposed face of $\text{conv}(\Phi_N)$ will also be an exposed face of $\mathcal{I}_N(\mathcal{O}_N)$. Such an exposed face will also necessarily be an exposed face of Ω_N^* .

The inner bound algorithm applies to any outer bound, i.e., not just \mathcal{P}_N . In particular, augmenting the Shannon outer bound with the recently discovered non-Shannon-type inequalities yields a better outer bound.

The quality of the inner bound improves with the quality of the outer bound. That is, given a sequence of increasingly tight outer bounds, our algorithm generates a corresponding sequence of increasingly tight inner bounds, in the sense that any face of the outer bound that is tight on an exposed face of Φ_N will generate a face of the inner bound that is also tight.

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Extensions and future work

- We further applied this general technique to the Shannon outer bound augmented with the non-Shannon type inequality due to Zhang and Yeung (1998), and established that none of the non-Shannon exposed faces are tight on a full face of $\text{conv}(\Phi_4)$.
- We have also extended the inner bound algorithm to the set of entropy vectors under one or more distribution constraints, meaning we are restricted to a subset of the full entropy vector space.
- An important open question is whether or not $\text{conv}(\Phi_N)$ is a polytope. If it is, then it admits a finite characterization in both a listing of its generating vertices and the inequalities characterizing its exposed faces, and it is of interest to obtain these listings. This is the subject of our current work.

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