

Multiobjective optimization and the entropy region

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Outline

- 1 Polymatroids and regions
- 2 Entropy inequalities
- 3 The structure of H_4 and the natural coordinates
- 4 Multiobjective optimization
- 5 Solving the optimization problem

Polymatroids

- The **ground set** N is any finite set, $N = \{1, 2, \dots, N\}$.
- The **rank function** f assigns non-negative values to the subsets $I \subseteq N$, that is, $f : 2^N \rightarrow \mathbb{R}^{\geq 0}$.
- $\langle f, N \rangle$ is a **polymatroid** if it satisfies the Shannon inequalities:

$$f(\emptyset) = 0,$$

$$f(A) \geq f(B) \text{ if } A \supseteq B,$$

$$f(A) + f(B) \geq f(A \cup B) + f(A \cap B).$$

- $\langle f, N \rangle$ is a **matroid** if $f(A)$ is integer, and $f(A) \leq |A|$.
- $\langle f, N \rangle$ is **entropic** if $f(A) = \mathbf{H}(\xi_A)$, where $(\xi_i : i \in N)$ are discrete random variables with some joint distribution.
- Pointwise limit of entropic polymatroids are **almost entropic**.

Regions

The rank function f is a vector indexed by non-empty subsets of N .

- $\mathbf{H}_N \subseteq \mathbb{R}^{2^N-1}$ is the region of **polymatroids**.
– a full-dimensional closed convex pointed cone.
- $\mathbf{H}_N^{\text{ent}} \subseteq \mathbf{H}_N$ is the **entropy region**.
- $\text{cl}(\mathbf{H}_N^{\text{ent}})$ is the **pointwise closure** of $\mathbf{H}_N^{\text{ent}}$.

Theorem (Zhang–Yeung 1998, Matúš 2007)

- $\text{cl}(\mathbf{H}_N^{\text{ent}})$ is a convex full-dimensional cone in \mathbb{R}^{2^N-1} .
- The interior of $\text{cl}(\mathbf{H}_N^{\text{ent}})$ is entropic.
- $\mathbf{H}_2^{\text{ent}} = \text{cl}(\mathbf{H}_2^{\text{ent}}) = \mathbf{H}_2$.
- $\mathbf{H}_3^{\text{ent}} \neq \text{cl}(\mathbf{H}_3^{\text{ent}}) = \mathbf{H}_3$.
- $\mathbf{H}_N^{\text{ent}} \neq \text{cl}(\mathbf{H}_N^{\text{ent}}) \neq \mathbf{H}_N$ for $N \geq 4$.
- $\text{cl}(\mathbf{H}_N^{\text{ent}})$ is **not** polyhedral for $N \geq 4$. □

The boundary of the entropy region

Definition

$\mathbf{H}_N^k \subseteq \mathbf{H}_N^{\text{ent}}$ is the subregion where the distribution $(\xi_i : i \in N)$ has alphabet size k .

Facts

\mathbf{H}_N^k is closed; $\mathbf{H}_N^k \subseteq \mathbf{H}_N^{k+1}$; and $\mathbf{H}_N^{\text{ent}} = \bigcup_k \mathbf{H}_N^k$. □

Research Problems

- ❶ For fixed N , is the convergence $\mathbf{H}_N^k \rightarrow \mathbf{H}_N^{\text{ent}}$ uniform?
- ❷ Give an estimate for the thickness of $\mathbf{H}_N^{\text{ent}} - \mathbf{H}_N^k$ (in different metrics) as a function of k .
- ❸ Give a description of $\text{cl}(\mathbf{H}_N^{\text{ent}}) - \mathbf{H}_N^{\text{ent}}$ in the case $N = 3$. Where is it fractal-like?

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Searching for new entropy inequalities

Known methods to get new entropy inequalities are:

- 1 Zhang–Yeung method (1998)
- 2 Makarychev *et al.* technique (2002)
- 3 Matúš' polymatroid convolution (2007)
- 4 Maximum entropy extension (2014)

Equivalence of #1 and #2 for **balanced** inequalities was shown by Tarik Kaced (2013).

Research problem

Show that methods #3 and #4 are actually **stronger** than the other two.

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Show that methods #3 and #4 are actually **stronger** than the other two.

We focus on method #1, others raise similar issues.

Zhang–Yeung method

In nutshell

- 1 Start with a pool of some (at least four) random variables;
- 2 split the random variables into two sets: \vec{x}_1 and \vec{y} ,
- 3 make an independent copy \vec{x}_2 of \vec{x}_1 over \vec{y} to get the new pool of random variables $\langle \vec{x}_1, \vec{x}_2, \vec{y} \rangle$;
- 4 iterate steps 2 and 3 several times;
- 5 collect the constraints:
 - Shannon inequalities for the final variable set;
 - equalities among entropy values expressing:
 - all conditional independence; identical distribution of (\vec{x}_1, \vec{y}) and (\vec{x}_2, \vec{y}) ; symmetry of \vec{x}_1 and \vec{x}_2 over \vec{y} ;
- 6 extract all consequences for the original variables.

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- 6 extract all consequences for the original variables.

Numerically intractable even for three full iterations.

Zhang–Yeung method

Remedy (Dougherty *et al*)

- discard some of the copied variables in \vec{x}_2 ; and/or
- glue together some variables in \vec{x}_2 .

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Example **copy string** with three iterations, **initial** random variables **abcd** and **auxiliary** variables **rstuv**:

```
rs=cd:ab; tu=cr:ab; v=(cr):abtu
```

The set of **constraints** is composed of

- all Shannon inequalities,
- all conditional independence, and
- equality arising from identical distributions and symmetry,

written for entropies of the subsets of the initial and auxiliary variables (**abcd+rstuv**).

Zhang–Yeung method – geometrical view

Given a copy string for initial variables **abcd**, we use the notation

- $\mathbf{x} \in \mathbb{R}^p$ for the entropies of subsets of **abcd** ($p = 15$);
- $\mathbf{y} \in \mathbb{R}^m$ for the vector of all other entropies;
- $\mathcal{M}(\mathbf{x}, \mathbf{y})$ for the collection of constraints.

$\mathcal{M}(\mathbf{x}, \mathbf{y})$ is **linear** and **homogeneous**, thus can be written as

$$P\mathbf{x} + M\mathbf{y} \geq 0$$

for some $p \times n$ and $m \times n$ matrices P and M determined by the copy string.

- $\mathcal{P} = \{ (\mathbf{x}, \mathbf{y}) \in \mathbb{R}^{p+m} : \mathbf{x} \geq 0, \mathbf{y} \geq 0, P\mathbf{x} + M\mathbf{y} \geq 0 \}$
is the **feasible region**, a convex pointed polyhedral cone;
- $\mathcal{Q} = \{ \mathbf{x} \in \mathbb{R}^p : \text{for some } \mathbf{y} \in \mathbb{R}^m, (\mathbf{x}, \mathbf{y}) \in \mathcal{P} \}$
is the **projection of \mathcal{P}** , a convex, pointed polyhedral cone.

⇒

Geometrical view

- ⇒
- $\mathcal{P} = \{(\mathbf{x}, \mathbf{y}) \in \mathbb{R}^{p+m} : \mathbf{x} \geq 0, \mathbf{y} \geq 0, P\mathbf{x} + M\mathbf{y} \geq 0\}$,
 - $\mathcal{Q} = \{\mathbf{x} \in \mathbb{R}^p : \text{for some } \mathbf{y} \in \mathbb{R}^m, (\mathbf{x}, \mathbf{y}) \in \mathcal{P}\}$.

Linear consequences of $P\mathbf{x} + M\mathbf{y} \geq 0$ are the non-negative linear combinations of the rows of (P, M) . Such an inequality bounds \mathcal{Q} iff in it all \mathbf{y} coordinates are zero. Thus the collection of linear inequalities bounding \mathcal{Q} – the **dual cone of \mathcal{Q}** – is

- $\mathcal{Q}^\circ = \{P^T\mathbf{v} \in \mathbb{R}^p : \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \geq 0, M^T\mathbf{v} = 0\}$.

Observations

- If $\mathbf{x} \in \mathbb{R}^p$ is entropic, then for some $\mathbf{y} \in \mathbb{R}^m$, $(\mathbf{x}, \mathbf{y}) \in \mathcal{P}$. Therefore the entropy region is contained in the projection \mathcal{Q} .*
- The “strongest” entropy inequalities which can be extracted from a copy string are the extremal rays of \mathcal{Q}° .*



Creating new information inequalities

In theory it is as easy as ...

- 1 Choose your favorite **copy string**.
- 2 **Generate** the matrices (P, M) describing the linear homogeneous constraints arising from your copy string.
- 3 **Compute** the extremal rays of Q° using your favorite computer algebra package.

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In practice there are annoying nuisances ...

- 1 When things get interesting, M becomes really large (over 28000 Shannon inequalities just for 4+7 variables).
- 2 Even if the size is not a problem, M is highly degenerate (hated by all packages).
- 3 The computational problem is numerically unstable (and integer arithmetic takes ages).

Improving the performance

Use what is known about H_4

Where to look:

- [1] Frantisek Matúš and Milan Studený,
Conditional independencies among four
random variables I,
*Combinatorics, Probability and
Computing*, no 4, (1995) pp. 269-278.



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Entropy expressions

Fix four random variables as a, b, c, d .

For any subset J of $abcd$, J also denotes its entropy, $H(J)$.

Definition

For any permutation of the variables a, b, c, d we define

- $(a, b) \stackrel{\text{def}}{=} a + b - ab; \quad \Leftarrow \text{mutual info}$
- $(a, b | c) \stackrel{\text{def}}{=} ac + bc - abc - c; \quad \Leftarrow \text{cond. mutual info}$
- $(a, b | cd) \stackrel{\text{def}}{=} acd + bcd - abcd - cd;$
- $(a | bcd) \stackrel{\text{def}}{=} abcd - bcd; \quad \Leftarrow \text{cond. entropy}$
- $[abcd] \stackrel{\text{def}}{=} -(a, b) + (a, b | c) + (a, b | d) + (c, d). \quad \Leftarrow \text{Ingleton}$

The **Ingleton** expression is symmetric in ab and cd :

$$[abcd] = [\overset{\curvearrowright}{bacd}] = [ab\overset{\curvearrowright}{dc}] = [\overset{\curvearrowright}{ba}\overset{\curvearrowright}{dc}].$$

Why Ingleton is so important

Definition

$\square \subset cl(\mathbf{H}_4^{\text{ent}})$ where **all six** Ingleton expressions are ≥ 0 ;

$\square_{ab} \subset cl(\mathbf{H}_4^{\text{ent}})$ where $[abcd] \leq 0$, i.e., this Ingleton is violated;

$\square_{ac} \subset cl(\mathbf{H}_4^{\text{ent}})$ where $[acbd] \leq 0$; etc.

Theorem (Matus – Studeny, 1995)

- $cl(\mathbf{H}_4^{\text{ent}}) = \square \cup \square_{ab} \cup \dots \cup \square_{cd}$.
- Any two of $\square, \square_{ab}, \dots, \square_{cd}$ have disjoint interior; common points are on the boundary of \square .
- \square is a full dimensional closed polyhedral cone, bounded by the six Ingleton, and certain other Shannon facets.
- Internal points and vertices of \square are linearly representable.
- $\square_{ab}, \dots, \square_{cd}$ are isomorphic; isomorphisms are provided by permutations of $abcd$.



If ...

If we know \square_{ab} , then
we know everything.*

*at least about $cl(\mathbf{H}_4^{\text{ent}})$.



The case of five variables

Research problem

Give a similar decomposition of the 31-dimensional cone H_5 .

- H_5 has a 120-fold symmetry;
- it has 117978 vertices ^[2];
- the vertices fall into 1319 equivalence classes^[2] (into 15 equivalence classes in case of four variables);
- the linearly representable core of H_5 is known precisely^[3].

[2] M. Studený, R. R. Bouckaert, T. Kočka:
Extreme supermodular set functions over five variables

[3] R. Dougherty, C. Freiling, K. Zeger:
Linear rank inequalities on five or more variables

Natural coordinates

$\square_{ab} \subset \mathbf{H}_4$ is contained in the simplicial cone determined by these facets (proved in [1]):

1. $[abcd]$,
- 2., 3. $(a, b | c), (a, b | d)$,
- 4–7. $(a, c | b), (b, c | a), (a, d | b), (b, d | a)$,
- 8., 9. $(c, d | a), (c, d | b)$,
10. (c, d) ,
11. $(a, b | cd)$,
- 12–15. $(a | bcd), (b | acd), (c | adb), (d | abc)$,

Natural coordinates

Use the facet equations as the coordinates for the entropies.

Entropy inequalities in natural coordinates

There are **six** natural coordinate systems corresponding to the six non-equivalent Ingleton expressions. Each entropy inequality can be written using any of the natural coordinates.

General form of a linear inequality

$$\lambda_1[abcd] + \lambda_2(a, b|c) + \lambda_3(a, b|d) + \dots + \lambda_{15}(d|abc) \geq 0. \quad (1)$$

Claim

- ① $\lambda_2 \geq 0, \lambda_3 \geq 0, \dots, \lambda_{15} \geq 0$.
- ② *The Ingleton coeff is > 0 in some natural coordinate system.*
- ③ *Can be strengthened by setting $\lambda_{12}, \lambda_{13}, \lambda_{14}, \lambda_{15}$ to zero.*

Proof.

- ① (1) must hold for the entropic vector $(0, \dots, 0, 1, 0, \dots)$.
- ② If not, then all points satisfying (1) are in \square .
- ③ Equivalent to balancing (1). □

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What we've had

⇒

- $\mathbf{x} \in \mathbb{R}^p$ are the entropies of **abcd**,

- $\mathbf{y} \in \mathbb{R}^m$ are all other entropies,

- the constraints are given by the matrices (P, M) ,

- $\mathcal{P} = \{(\mathbf{x}, \mathbf{y}) \in \mathbb{R}^{p+m} : \mathbf{x} \geq 0, \mathbf{y} \geq 0, P\mathbf{x} + M\mathbf{y} \geq 0\}$,

- $\mathcal{Q} = \{\mathbf{x} \in \mathbb{R}^p : \text{for some } \mathbf{y} \in \mathbb{R}^m, (\mathbf{x}, \mathbf{y}) \in \mathcal{P}\}$.

- $\mathcal{Q}^\circ = \{P^T \mathbf{v} \in \mathbb{R}^p : \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \geq 0, M^T \mathbf{v} = 0\}$,

What we've had, and what we've got

- ⇒
- $\mathbf{x} \in \mathbb{R}^p$ are the entropies of **abcd** in **natural coordinates**,
 - $\mathbf{y} \in \mathbb{R}^m$ are all other entropies,
 - the constraints are given by the matrices (P, M) ,
 - $\mathcal{P} = \{(\mathbf{x}, \mathbf{y}) \in \mathbb{R}^{p+m} : \mathbf{x} \geq 0, \mathbf{y} \geq 0, P\mathbf{x} + M\mathbf{y} \geq 0\}$,
 - $\mathcal{Q} = \{\mathbf{x} \in \mathbb{R}^p : \text{for some } \mathbf{y} \in \mathbb{R}^m, (\mathbf{x}, \mathbf{y}) \in \mathcal{P}\}$.
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The gains are

- 1 the first (Ingleton) coordinate in \mathcal{Q}° can be fixed to be 1;
- 2 the last four coordinates in \mathcal{Q}° can be requested to be zero.

These conditions can be moved from P to M to get (P_*, M_*) .

The relevant part of \mathcal{Q}° with coordinates $\lambda_2, \dots, \lambda_{11}$ is

- $\mathcal{Q}^* = \{P_*^T\mathbf{v} \in \mathbb{R}^{10} : \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \geq 0, M_*^T\mathbf{v} = \mathbf{e}_{\text{Ingl}}\}$.



The optimization problem

$$\Rightarrow \bullet Q^* = \{ P_*^T \mathbf{v} \in \mathbb{R}^{10} : \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \geq 0, M_*^T \mathbf{v} = \mathbf{e}_{\text{Ing}} \},$$

where \mathbf{e}_{Ing} is the Ingleton unit vector.

Observations

a) If $\lambda \in Q^*$, then $\lambda \geq 0$.

b) Q^* is upward closed: if $\lambda \in Q^*$, and $\lambda \leq \lambda'$, then $\lambda' \in Q^*$.

The vertices of Q^* are the coefficients of the “strongest” entropy inequalities which can be extracted from the copy string,

The optimization problem

$$\Rightarrow \bullet Q^* = \{ P_*^T \mathbf{v} \in \mathbb{R}^{10} : \mathbf{v} \in \mathbb{R}^n, \mathbf{v} \geq 0, M_*^T \mathbf{v} = \mathbf{e}_{\text{Ing}} \},$$

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Observations

- a) If $\lambda \in Q^*$, then $\lambda \geq 0$.
- b) Q^* is upward closed: if $\lambda \in Q^*$, and $\lambda \leq \lambda'$, then $\lambda' \in Q^*$.

The vertices of Q^* are the coefficients of the “strongest” entropy inequalities which can be extracted from the copy string, and the vertices of Q^* are the solutions of

Multiobjective optimization problem

Find the minimum of: $P_*^T \mathbf{v} \in \mathbb{R}^{10}$ ⇐ 10 objectives

subject to: $\mathbf{v} \geq 0$, and $M_*^T \mathbf{v} = \mathbf{e}_{\text{Ing}}$ ⇐ constraints

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Benson's outer approximation algorithm

The problem is to find the vertices of the polytope

$$Q^* = \{ P_*^T \mathbf{v} : \mathbf{v} \geq 0, M_*^T \mathbf{v} = \mathbf{e} \}.$$

Benson's idea: Given the internal point $\mathbf{x}_i \in Q^*$, and the external point $\mathbf{x}_o \notin Q^*$, find

$$\max_{\mu} \{ 0 \leq \mu \leq 1 : \mu \mathbf{x}_o + (1 - \mu) \mathbf{x}_i \in Q^* \}.$$

- a) *This is an $n + 11$ -dimensional LP problem.*
- b) *The (dual of the) solution gives a proof for maximality, which is a facet of Q^* separating \mathbf{x}_i and \mathbf{x}_o .*

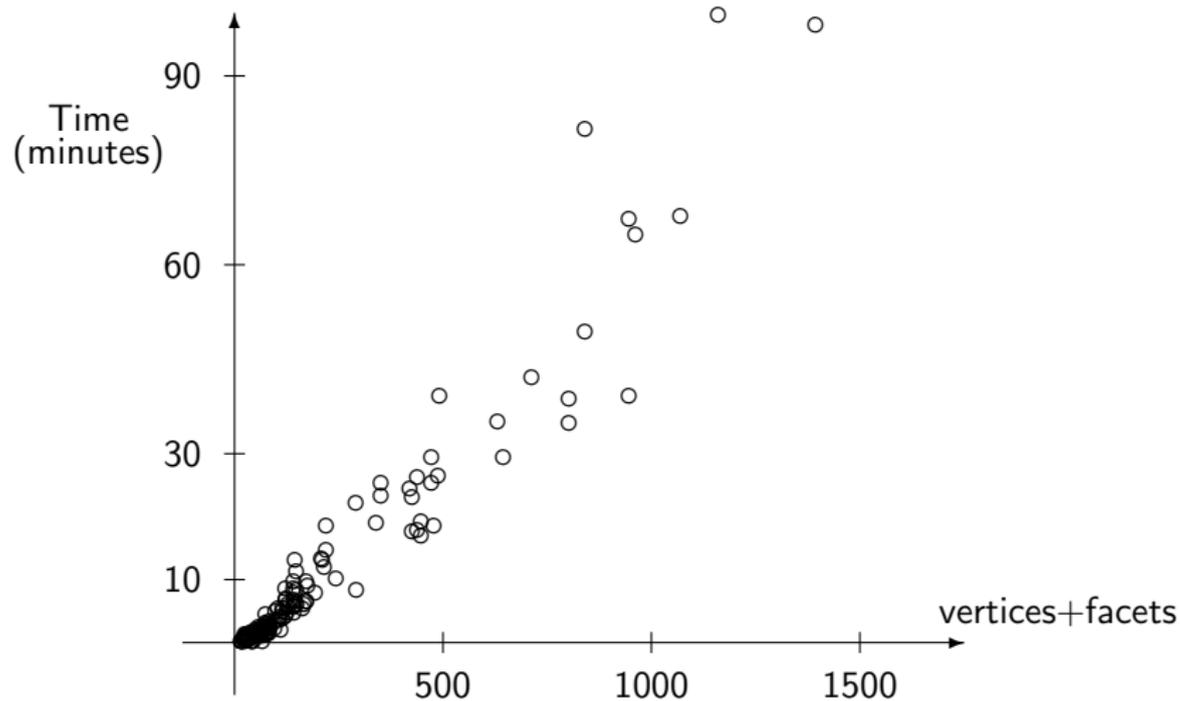
The algorithm

Use this idea to get all facets of Q^* , maintaining the vertices of the approximating polytope bounded by the facets obtained so far.

Some results for Dougherty *et al* out of 133

Copy string	Size of M_*	Vertices	Facets	Time
r=c:ab;s=r:ac;t=r:ad	561×80	5	20	0:01
rs=cd:ab;t=r:ad;u=s:adt	1509×172	40	132	6:19
rs=cd:ab;t=a:bc;u=(cs):abrt	1569×178	47	76	6:51
rs=cd:ab;t=a:bc;u=b:adst	1512×178	177	261	17:40
rs=cd:ab;t=a:bc;u=t:acr	1532×178	85	134	18:27
rs=cd:ab;t=(cr):ab;u=t:acs	1522×172	181	245	22:58
r=c:ab;st=cd:abr;u=a:bcrt	1346×161	209	436	29:18
rs=cd:ab;t=a:bc;u=c:abrst	1369×166	355	591	38:59
rs=cd:ab;t=a:bc;u=c:abrt	1511×178	363	599	1:04:32
rs=cd:ab;t=a:bc;u=s:abcdt	1369×166	355	591	1:07:01
rs=cd:ab;t=a:bc;u=(at):bc	1555×177	484	676	1:39:30
rs=cd:ab;t=a:bc;u=a:bcst	1509×177	880	1238	4:30:26
rs=cd:ab;t=a:bc;u=a:bdrt	1513×177	2506	2708	5:11:25

Running time vs. vertices + facets



Some results with five auxiliary variables

Copy string	Size of M_*	Vertices	Facets	Time
<code>rs=cd:ab;tu=cr:ab;v=(cs):abtu</code>	4055×370	19	58	1:10:10
<code>rs=ad:bc;tu=ar:bc;v=r:abst</code>	4009×370	40	103	3:24:37
<code>rs=cd:ab;t=(cr):ab;uv=cs:abt</code>	3891×358	30	102	3:34:31
<code>rs=cd:ab;tu=cr:ab;v=t:adr</code>	3963×362	167	235	9:20:19
<code>rs=cd:ab;tu=dr:ab;v=b:adsu</code>	4007×370	318	356	13:20:08
<code>rs=cd:ab;tv=dr:ab;u=a:bcrt</code>	4007×370	318	356	14:34:42
<code>rs=cd:ab;tu=cs:ab;v=a:bcrt</code>	4007×370	297	648	22:02:39
<code>rs=cd:ab;t=a:bc;uv=bt:acr</code>	3913×362	779	1269	37:15:33
<code>rs=cd:ab;tu=cr:ab;v=a:bcstu</code>	3987×362	4510	7966	427:43:30
<code>rs=cd:ab;tu=cs:ab;v=a:bcrtu</code>	3893×362	10387	13397	716:36:32

Using five auxiliary variables, more than 260 new entropy inequalities were generated. One of them is

$$2[abcd] + (a, b | c) + 3(a, c | b) + (b, c | a) + 3(c, d | a) \geq 0.$$



Thank you for your attention