

The Log-Rank Conjecture: New Equivalent Formulations

Lianna Hambardzumyan

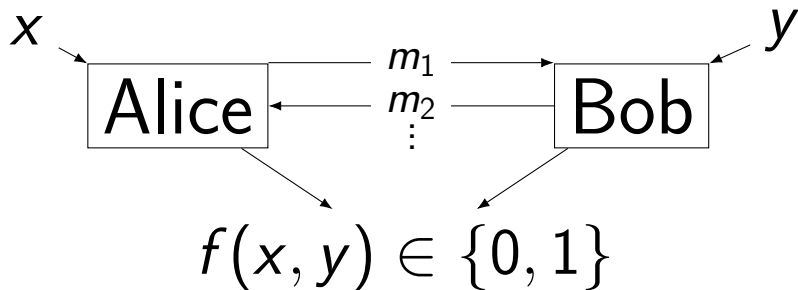
Shachar Lovett

Morgan Shirley

Outline

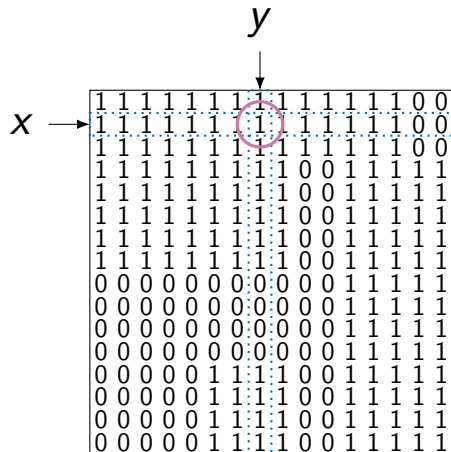
- ▶ What is the Log-Rank Conjecture?
- ▶ Equiv. formulation 1: signed rectangle rank
- ▶ Equiv. formulation 2: cross-intersecting sets

Communication Complexity

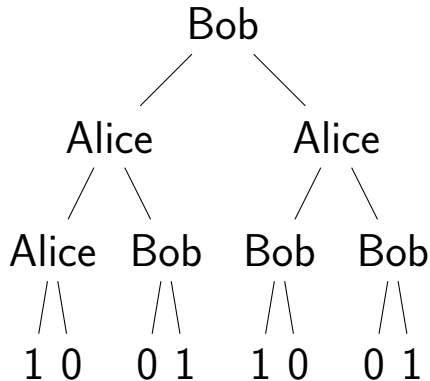


How much information must be exchanged to compute f ?

Communication as matrices



Communication as matrices



1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1	1	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	1	1	1	1	0	0	1	1	1
0	0	0	0	0	0	1	1	1	1	0	0	1	1	1
0	0	0	0	0	0	1	1	1	1	0	0	1	1	1
0	0	0	0	0	0	1	1	1	1	0	0	1	1	1

$$\# \text{ all-ones rectangles} \leq \# \text{ of protocol leaves} \leq 2^{D^{\text{cc}}}$$

Connection between communication and rank

[illegible]

$$D^{\text{cc}}(M) \geq \log \text{rank}(M)$$

$$D^{\text{cc}}(M) \geq \log \text{binrk}(M) \geq \log \text{rank}(M)$$

(Binary rank is also called *partition number*)

$$D^{\text{cc}}(M) \leq \text{poly log binrk}(M)$$

The Log-Rank Conjecture

Conjecture [LS88]: for all Boolean matrices,
 $D^{\text{cc}}(M) \leq \text{poly log rank}(M)$

Because D^{cc} and $\log \text{binrk}$ are polynomially-related, this is equivalent to

Conjecture: for all Boolean matrices,
 $\log \text{binrk}(M) \leq \text{poly log rank}(M)$

Why do we care?

- ▶ It would answer questions about the structure of low-rank matrices
- ▶ Hardness of approximation for D^{cc} :
 - ▶ D^{cc} NP-hard to compute [HIL25]
 - ▶ Binary rank NP-hard to compute [JR93]
 - ▶ Rank poly-time to compute (Gaussian elimination)
- ▶ Graph theory implications: chromatic number vs. adjacency matrix rank

How is rank different from binary rank?

- ▶ Rank is **algebraic**, binary rank is **combinatorial**
- ▶ Rank allows **cancellations**, binary rank is **monotone**

What happens when we relax only one of these?

Monotone but algebraic

Non-negative rank: How many *non-negative* rank-one matrices are needed to sum to M ?

$$\log \text{binrk}(M) \leq \text{poly log nnr}(M)$$

Combinatorial but with cancellations

Definition: The signed rectangle rank of M , denoted $\text{srr}(M)$, is the minimum s such that we can write

$$M = \sum_{i=1}^s \epsilon_i R_i$$

where R_i are binary rank-one matrices and $\epsilon_i \in \{-1, 1\}$.

Main result

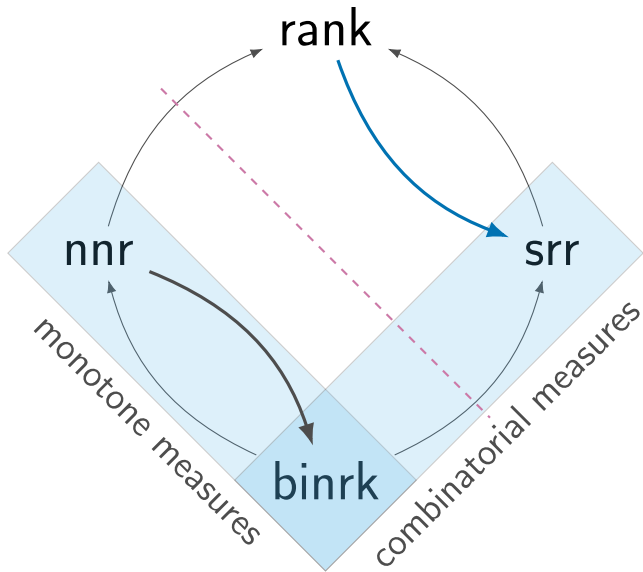
Theorem: For every Boolean matrix M of rank r ,

$$\text{srr}(M) \leq O(r \log r).$$

So the Log-Rank Conjecture is *equivalent* to:

Conjecture: for all Boolean matrices,

$$\log \text{binrk}(M) \leq \text{poly log srr}(M)$$



Cancellations vs. monotonicity
is the core of the Log-Rank Conjecture

Secondary result

We also prove that the Log-Rank Conjecture is equivalent to a conjecture in combinatorics.

Main result: proof idea

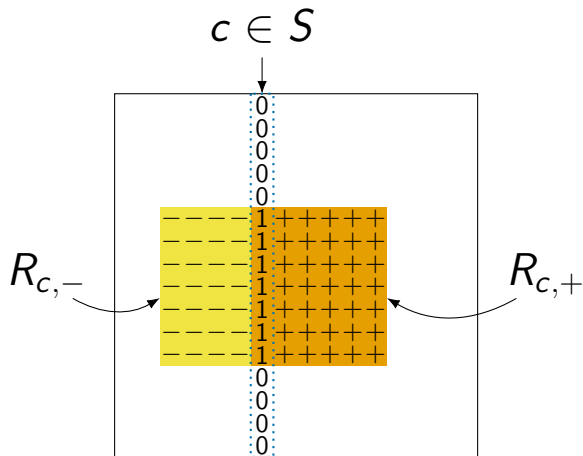
Theorem: For every Boolean matrix M of rank r ,

$$\text{srr}(M) \leq O(r \log r).$$

Idea: Find a set of columns S such that:

- ▶ $|S| \leq O(r \log r)$
- ▶ Every column in M is a ± 1 linear combination of columns from S

Main result: proof idea



$$\text{srr}(M) \leq 2|S| \leq O(r \log r)$$

Main result: proof idea

Idea: Find an independent set of columns S such that:

- ▶ $|S| \leq O(r \log r)$

- ▶ **Claim 1**

- ▶ Every column in M is a ± 1 linear combination of columns from S

- ▶ **Claim 2**

Main result: independent sets of columns

Definition: A set of columns S is *independent* if all subsets of S have different sums.

Weaker than linear independence.

$$\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{bmatrix} 0 & 1 & 0 & 1 & 1 & 1 & 2 & 2 & 1 & 1 & 2 & 2 & 2 & 3 & 2 & 3 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 2 & 1 & 1 & 2 & 1 & 1 & 2 & 2 \\ 0 & 1 & 1 & 0 & 0 & 2 & 1 & 1 & 1 & 1 & 0 & 2 & 2 & 1 & 1 & 2 \end{bmatrix}$$

Main result: outline

Claim 1: If S is an independent set of M , then $|S| \leq O(r \log r)$.

Claim 2: If S is a maximal independent set of M , then every column in M is a ± 1 linear combination of columns from S .

Main result: Claim 1

Claim 1: If S is an independent set of M , then $|S| \leq O(r \log r)$.

$$\begin{bmatrix} 0 & 1 & 0 & 1 & 1 & 1 & 2 & 2 & 1 & 1 & 2 & 2 & 2 & 3 & 2 & 3 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 2 & 1 & 1 & 2 & 1 & 1 & 2 & 2 \\ 0 & 1 & 1 & 0 & 0 & 2 & 1 & 1 & 1 & 1 & 0 & 2 & 2 & 1 & 1 & 2 \end{bmatrix}$$

- ▶ $\# \text{ columns} = \# \text{ unique columns} = 2^{|S|}$
- ▶ $\text{rank} \leq r$
- ▶ $\text{entries} \in \{0, \dots, |S|\}$

Main result: Claim 1

Claim 1: If S is an independent set of M , then $|S| \leq O(r \log r)$.

$$\text{rank} \leq r \quad \text{entries} \in \{0, \dots, |S|\}$$

- ▶ r rows span the matrix
- ▶ Columns uniquely identified by r entries
- ▶ At most $(|S| + 1)^r$ unique columns

Main result: Claim 1

Claim 1: If S is an independent set of M , then $|S| \leq O(r \log r)$.

$$(|S| + 1)^r \leq \# \text{ unique columns} = 2^{|S|}$$

Solving for $|S|$ gives $|S| \leq O(r \log r)$.

Main result: Claim 2

Claim 2: If S is a maximal independent set of M , then every column in M is a ± 1 linear combination of columns from S .

For any column c of M :

- ▶ If $c \in S$, we're done :)
- ▶ Else, $S \cup \{c\}$ not independent
 - ▶ $\sum A = \sum B + c$
 - ▶ $c = \sum A - \sum B$
 - ▶ done here too :)

Main result

Theorem: For every Boolean matrix M of rank r ,

$$\text{srr}(M) \leq O(r \log r).$$

Claim 1: If S is an independent set of M , then $|S| \leq O(r \log r)$.

Claim 2: If S is a maximal independent set of M , then every column in M is a ± 1 linear combination of columns from S .

Cross-intersecting set families

A pair of set families $\mathcal{S}, \mathcal{T} \subseteq 2^{[d]}$ are $\{a, b\}$ -*cross-intersecting* if

$$\forall (S, T) \in \mathcal{S} \times \mathcal{T}, |S \cap T| \in \{a, b\}.$$

Cross-intersecting set families conjecture

Conjecture [L21,SS22]: For any a, b and any $\{a, b\}$ -cross-intersecting families $\mathcal{S}, \mathcal{T} \subseteq 2^{[d]}$, there exist subfamilies $\mathcal{A} \subseteq \mathcal{S}, \mathcal{B} \subseteq \mathcal{T}$ such that

- ▶ Every pair $(A, B) \in \mathcal{A} \times \mathcal{B}$ has the same intersection size,
- ▶ $|\mathcal{A}| \cdot |\mathcal{B}| \geq 2^{\text{poly log } d} \cdot |\mathcal{S}| \cdot |\mathcal{T}|$.

A similar conjecture

Conjecture [NW95]: Any Boolean matrix M has an all-zeroes or all-ones rectangle of density $2^{\text{poly log rank}(M)}$.

Theorem [NW95]: The above conjecture is equivalent to the log-rank conjecture.

Our result on the CIS conjecture

Using the Main Theorem, we prove:

Theorem: The cross-intersecting set conjecture implies the Log-Rank Conjecture.

(Therefore, the conjectures are equivalent!)

Concluding thoughts

- ▶ Cancellations vs. monotonicity: “right way” to think about the LRC?
- ▶ Pessimist’s view: CIS conjecture seems hard
- ▶ Optimist’s view: new reason to care about CIS conjecture
- ▶ Left out of talk: methods naturally extend to tensors and tensor rank, related to NIH LRC
- ▶ Open problem: improve $O(r \log r)$ to $O(r)$? Our method doesn’t work for this [Lin65]