

# Global Optima from Local Algorithms

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# The LLL Setting

- Probability space + Set of  $m$  “bad” events  $B = \{\mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_m\}$ .
- If  $\{\mathcal{E}_i\}$  are independent,  $\Pr[\text{Nothing bad happens}] = \prod_{i=1}^m (1 - p_i)$ .
- But what if avoiding some bad events **boosts** some other bad events ?

**Example:**  $\Omega = \{0, 1\}^3$  with uniform measure,  $F = (x_1 \vee x_2) \wedge (\overline{x_2} \vee x_3)$ .

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## General LLL (Erdős, Lovász '75)

If each  $\mathcal{E}_i$  is **mutually independent of all** events in  $B \setminus (\Gamma(i) \cup \mathcal{E}_i)$  and there exist  $\{x_i\} \in [0, 1)$  such that

$$x_i \geq \frac{\Pr[\mathcal{E}_i]}{\prod_{j \in \Gamma(i)} (1 - x_j)}$$

then  $\Pr[\text{Nothing bad happens}] \geq \prod_{i=1}^m (1 - x_i)$ .

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## Asymmetric LLL (Very Handy Corollary)

If each  $\mathcal{E}_i$  is mutually independent of all events in  $B \setminus (\Gamma(i) \cup \mathcal{E}_i)$   
and

$$\sum_{j \in \Gamma(i)} \Pr[\mathcal{E}_j] \leq \frac{1}{4}$$

then  $\Pr[\text{Nothing bad happens}] > 0$ .

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## Lopsided LLL (Erdős, Spencer '91)

If each  $\mathcal{E}_i$  is **never boosted** by any conditioning on events in  $B \setminus (\Gamma(i) \cup \mathcal{E}_i)$  and there exist  $\{x_i\} \in [0, 1)$  such that

$$x_i \geq \frac{\Pr[\mathcal{E}_i]}{\prod_{j \in \Gamma(i)} (1 - x_j)}$$

then  $\Pr[\text{Nothing bad happens}] \geq \prod_{i=1}^m (1 - x_i)$ .

# A Tight Example and a Breakthrough

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Every  $k$ -CNF formula where each clause shares variables with at most  $\Delta \leq 2^k/e$  other clauses is satisfiable.

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There exist **unsatisfiable** formulas with  $\Delta = (1 + \delta_k)2^k/e$ , where  $\delta_k \rightarrow 0$ .

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Algorithmic LLL: a s.t.a can be found efficiently if  $\Delta \leq 2^{k/4}$

[Beck 91], [Alon 91], [Molloy, Reed 98], [Czumaj, Scheideler 00], [Srinivasan 08]



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## Theorem (Moser '09)

If  $\Delta(F) < 2^{k-5}$  a sat assignment can be found in  $O(|V| + |C| \log |C|)$ .

Moser's ideas, with more care, yield  $2^k/e$ .

[Messner, Thierauf 11]

# The Moser-Tardos Variable Setting

Assume that  $\mu$  is a **product** probability space over  $n$  variables, i.e.,  

$$\mu(\omega) = \mu_1(\omega_1) \cdots \mu_n(\omega_n) \text{ for all } \omega \in \Omega.$$

## Theorem (Moser Tardos 09)

If  $\mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_m$  satisfy the conditions of the **original** LLL, an element  $\omega \in \Omega$  such that none of the  $\{\mathcal{E}_i\}$  holds can be found in time  $O(m)$ .

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## Resample

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- 1: Sample  $\omega \in \Omega$  according to  $\mu$
- 2: **while** some bad event occurs **do**
- 3:     Select **any** occurring bad event  $\mathcal{E}_i$
- 4:     Resample the variables of  $\mathcal{E}_i$  **according to**  $\mu$
- 5: **return**  $\omega$

Easy since  $\mu$  is product

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- Covers majority of **existing** LLL applications
- It can be parallelized [Moser, Tardos 09]
- It can be derandomized [Chandrasekaran, Goyal, Haeupler 09]
- May work even if  $m$  is exponential [Haeupler, Saha, Srinivasan 10]
- Shearer's condition suffices (weaker than LLL) [Kolipaka, Szegedy 11]

# Limitations of the Moser Tardos Proof

The proof depends **heavily** on  $\mu$  being a product measure

- Does not extend to Lopsided LLL
- Hard to extend even to simple non-product spaces  
e.g., uniform measure on permutations [Harris, Srinivasan 13]

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- Product measures require a **variable decomposition**.
- Product measures are **algorithmically** limiting:

Resamplings are state-independent

## Example

LLL requires  $e\Delta$  colors on graphs with max degree  $\Delta$  instead of just  $\Delta + 1$  colors

What to do?

# Get Rid of the Measure

# No Measure

Let  $\Omega$  be an **arbitrary** finite set, such as:

- $\{0, 1\}^n$
- $P_n$
- $\text{Ham}(G) :=$  **Hamiltonian** cycles of graph  $G$ .
- $\text{Col}_q(G) :=$  **Valid** edge- $q$ -colorings of graph  $G$ .

Permutations of  $[n]$

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Let  $F = \{f_1, f_2, \dots, f_m\}$  be **arbitrary** subsets of  $\Omega$  called **flaws**.

**Examples** (Flaw  $f_i$  is the subset of....)

- $\{0, 1\}^n$  that violates clause  $c_i$ .
- $P_n$  in which  $\pi(i) = i$ .
- $\text{Col}_q(G)$  in which cycle  $C_i$  is bichromatic.

SAT

Derangements

Acyclic Edge Coloring



# The LLL as a Random Walk

## How to Find Flawless Objects

- Specify a directed graph  $D$  on  $\Omega$  such that:
  - Every flawed object has outdegree at least 1.
  - Every flawless object has outdegree 0.
- Start at an arbitrary  $\sigma_1 \in \Omega$
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## Definition (writing $\sigma \in \Omega$ instead of $\omega \in \Omega$ to emphasize “state” view)

- For  $\sigma \in \Omega$ , let  $U(\sigma) = \{f_i : \sigma \in f_i\}$ .  $\sigma$  is flawless iff  $U(\sigma) = \emptyset$ .
- $\forall \sigma \in \Omega, \forall f \ni \sigma$  define a non-empty set  $A(f, \sigma) \subseteq \Omega$  of **actions**.
- Each  $\tau \in A(f, \sigma)$  becomes an arc  $\sigma \xrightarrow{f} \tau$  in  $D$ .  $D$  is a multidigraph

# The Classics

$k$ -**CNF** formula  $F = c_1 \wedge \cdots \wedge c_m$  with  $n$  variables.

- $\Omega = \{0, 1\}^n$ .
- $f_i = \{\sigma \in \Omega : \sigma \text{ violates clause } c_i\}$ .
- $A(f_i, \sigma) = \{\text{The } 2^k \text{ mutations of } \sigma \text{ through } \text{var}(c_i)\}$ .

$$\Delta(F) < 2^k/e$$

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**$q$ -coloring** graph  $G(V, E)$  with  $n$  vertices.

- $\Omega = [q]^n$ .
- $f_{u,v} = \{\sigma \in \Omega : \text{col}(u) = \text{col}(v)\}$
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- $A(f_{u,v}, \sigma) = \{\text{Only conflict-free mutations}\}$

$$q > e\Delta$$

$$q \geq \Delta + 1$$

# Measuring Flaws

## Persistence

Let  $D_i$  be the subgraph of arcs labeled  $f_i$ . The **persistence** of  $f_i$  is

$$A_i = \max_{\sigma \in f_i} \frac{\text{In}(\sigma)}{\text{Out}(\sigma)}$$

## Potential Causality

Write  $i \rightarrow j$  if there is **any**  $\sigma \xrightarrow{f_i} \tau$  such that  $f_j \in U(\tau) \setminus (U(\sigma) \setminus f_i)$

$i \rightarrow i$  if  $f_i$  is still present in  $\tau$  and  $i \rightarrow j$  if  $f_j$  is a "new" flaw in  $\tau$

## Transience

The digraph  $D$  is **transient** if for every  $i \in [m]$ ,

$$\sum_{j \leftarrow i} A_j < \frac{1}{e} .$$

# Main Result

- Let  $\sigma_1 \in \Omega$  be arbitrary.
- For  $t = 1, 2, \dots$ 
  - Let  $f_i$  be a random flaw present in  $\sigma_t$ .
  - Address  $f_i$  by taking a uniformly random action in  $A(f_i, \sigma)$ .

## Theorem

Let  $T_0 = \log |\Omega| + |U(\sigma_1)|$ . If  $D$  is transient the probability that the walk does not reach a sink within  $t = O(T_0 + s)$  steps is less than  $2^{-s}$ .

## Features

- No need for a uniform sample to start off.
- Running time depends on  $|U(\sigma_1)|$ , not  $|F|$ .  $\Omega$ -sampling  $\implies T_0 = O(|F|)$ .

Flaw-choice: we can do better than random

Left-handed version

## General LLL

... If there exist positive real numbers  $\{x_i\}$  such that for all  $i \in [m]$ ,

$$\Pr(A_i) \prod_{j \in \{i\} \cup \Gamma(i)} (1 + x_j) \leq x_i \dots$$

## Theorem (Main result)

.... If there exist positive real numbers  $\{x_f\}$  such that for every  $f \in F$ ,

$$A_f \prod_{g \in \Gamma(f)} (1 + x_g) < x_f \dots$$

In fact, it suffices to have

$$A_f \sum_{S \in \text{Ind}(\Gamma(f))} \prod_{g \in S} x_g < x_f \dots$$



# Rainbow Perfect Matchings in a Complete Graph

Input: an edge-colored  $K_{2n}$ .      Output: a rainbow perfect matching

- Let  $\Omega$  be the set of all perfect matchings of  $K_{2n}$ .
- For each pair of vertex-disjoint edges  $\{e_i, e_j\}$  with the same color let

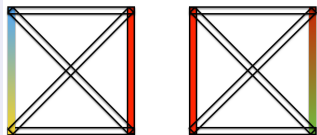
$$f_{i,j} = \{M \in \Omega : \{e_i, e_j\} \subset M\} .$$

**Result:** If no color is used more than  $n/(2e)$  times...  $O(n^2 \log n)$  time.

## Algorithm Design via Atomicity

**Algorithm:** swap **both** same-color edges out.

**Atomicity:** Consider  $M \xrightarrow{f_{i,j}} M'$ . Add to  $M'$  edges  $\{e_i, e_j\}$  and then close the two cycles.

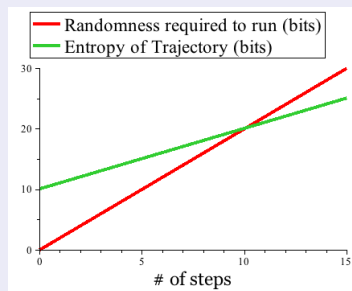


**Remark:** Swapping out only one of the two edges does not work.

## Proof

## Entropic Method

- Atomicity  $\Rightarrow$  if each state transition emits the flaw addressed, the trajectory up to moment  $t$  can be reconstructed from:  $I(t) = \langle f^1, f^2, \dots, f^t, \sigma_t \rangle$
- Amenability  $\Rightarrow$  Every  $t$ -step trajectory must consume at least  $L(t)$  bits
- Transcience  $\Rightarrow H(I_t) < L(t)$  for  $t > T_0$
- $\Rightarrow \exists$  likely trajectories of length  $< t$



# Summary + Future Work

- No probability function
- No variables
- State-dependent actions

Just  $\Omega$  and subsets

Arbitrary  $\Omega$  and subsets

Non-trivial Algorithms

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## Future Work

- Quantum version
- Backtracking algorithms
- Sampling

# Thanks!