# Computing Non-Repetitive Sequences Using the Lovász Local Lemma 

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## Introduction

## Background

- The Lovász local lemma (LLL) is an existence theorem with many uses within the probabilistic method (Erdős and Lovász, 1975).
- There is a probabilistic algorithm for finding witnesses to the LLL (Moser and Tardos, 2010).
- This algorithm can be simulated to compute infinite witnesses (Rumyantsev and Shen, 2014).
- This effective version has been applied in complexity theory (Csima, Dzhafarov, Hirschfeldt, Jockusch, Solomon, and Westrick, 2019; Liu, Monin, and Patey, 2018).
- The LLL can by extended in myriad ways. We effectivise a version of the LLL inspired by the Lefthanded LLL (Pegden, 2011).


## Non-Repetitive Sequences

## Classical Existence of Non-Repetitive Sequences

The following theorem of classical combinatorics, says "there exists of a sequence such that repetitions of long blocks are far apart."

## Theorem (Beck, 1981)

For each $\varepsilon>0$ there is an $N_{\varepsilon}$ and an infinite $\{0,1\}$-valued sequence such that any two identical blocks $[k, k+n)$ and $\left[\ell, \ell+n\right.$ ) of length $n>N_{\varepsilon}$ have distance $\ell-k$ greater than $(2-\varepsilon)^{n}$.

## Example

In the string

$$
a_{0} a_{1} a_{2} \ldots a_{11}=011010010001
$$

the only pair of identical blocks of size 4 are $[3,7)$ and $[6,10)$.

Question: Can we compute such a sequence?

## Classical Existence of Non-Repetitive Sequences

## Theorem (Beck, 1981)

For each $\varepsilon>0$ there is an $N_{\varepsilon}$ and an infinite $\{0,1\}$-valued sequence such that any two identical intervals of length $n>N_{\varepsilon}$ have distance greater than $(2-\varepsilon)^{n}$.

- Question: How to compute such a sequence?
- Existence is given by the infinite LLL.
- Natural choice: use effective version of the LLL given by Rumyantsev and Shen (2014).


## Effective Local Lemma Setup

- Let $\mathcal{X}=\left\{x_{1}, x_{2}, \ldots\right\}$ be a set of computable random variables with finite ranges and uniformly computable probability distributions.
- Let $\mathcal{A}=\left\{A_{1}, A_{2}, \ldots\right\}$ be a set of events such that
- Each $A \in \mathcal{A}$ is determined by a finite set of variables $\operatorname{vbl}(A) \subset \mathcal{X}$.
- The code numbers for $\operatorname{vbl}\left(A_{i}\right)$ are uniformly computable with respect to $i$.
- For each $A \in \mathcal{A}$, the set of neighbors $\Gamma(A)=\{B \in \mathcal{A}: \operatorname{vbl}(A) \cap \operatorname{vbl}(B) \neq \emptyset\}$ is finite.
- For each $x_{i},\left\{A_{j}: x_{i} \in \operatorname{vbl}\left(A_{j}\right)\right\}$ is finite and has code number uniformly computable with respect to $i$.


## Computable Local Lemma

Recall that $\Gamma(A)=\{B \in \mathcal{A}: \operatorname{vbl}(A) \cap \operatorname{vbl}(B) \neq \emptyset\}$
Theorem (Rumyantsev and Shen, 2014)
Suppose there exists a rational constant $\alpha \in(0,1)$ and a computable real-valued function $z: \mathcal{A} \rightarrow(0,1)$ such that, for each $A \in \mathcal{A}$,

$$
\operatorname{Pr}(A) \leq \alpha z(A) \prod_{B \in \Gamma(A)}(1-z(B)) .
$$

Then there exists a computable assignment of the variables in $\mathcal{X}$ that makes all events $A \in \mathcal{A}$ false.

## Setup for Building a Non-Repetitive Sequence

## Theorem (Beck, 1981)

For each $\varepsilon>0$ there is an $N_{\varepsilon}$ and an infinite $\{0,1\}$-valued sequence such that any two identical intervals of length $n>N_{\varepsilon}$ have distance greater than $(2-\varepsilon)^{n}$.

- Let $x_{i}$ be the value the $i^{\prime}$ th bit in the sequence.
- Let $A_{k, \ell, n}$ be the event that blocks $[k, k+n)$ and $[\ell, \ell+n)$ are identical (assume $k<\ell$ ).
- Let $\mathcal{A}=\left\{A_{k, \ell, n}: \ell-k<(2-\varepsilon)^{n}\right\}$.
- $\operatorname{vbl}\left(A_{k, \ell, n}\right)=[k, k+n) \cup[\ell, \ell+n) . \operatorname{Pr}\left(A_{k, \ell, n}\right)=2^{-n}$.
- $\Gamma\left(A_{k_{0}, \ell_{0}, n_{0}}\right)=\left\{A_{k, \ell, n} \in \mathcal{A}: \operatorname{vbl}\left(A_{k, l, n}\right) \cap \operatorname{vbl}\left(A_{k_{0}, \ell_{0}, n_{0}}\right) \neq \emptyset\right\}$


## Unsatisfied Conditions

We run into a the following issues with this setup.

- Each $x_{i}$ appears in $\operatorname{vbl}\left(A_{k, \ell, n}\right)$ for infinitely many $A_{k, \ell, n} \in \mathcal{A}$.
- Each $A_{k_{0}, \ell_{0}, n_{0}} \in \mathcal{A}$ has infinitely many neighbors $A_{k, \ell, n}$.

There are two sources:
(1) Fix $k, \ell$. Increase $n$.

- Can be fixed by modifying $\mathcal{A}$ to be $\left\{A_{k, \ell, n}: n\right.$ is least such that $\left.\ell-k<(2-\varepsilon)^{n}\right\}$
(2) Fix $k$. Increase $\ell$ and $n$.

The latter is not as readily fixed. To resolve them, we modify the Moser-Tardos algorithm.

## The Resample Algorithm

## Moser-Tardos Algorithm

The Moser-Tardos algorithm, also known as the resample algorithm, looks for a valuation of the variables in $\mathcal{X}=\left\{x_{1}, x_{2}, \ldots\right\}$ that makes each event in $\mathcal{A}=\left\{A_{1}, A_{2}, \ldots\right\}$ false.

## Algorithm

Start with a random sample of the variables in $\mathcal{X}$ and proceed in stages.

- At each stage, resample each $x \in \operatorname{vbl}(A)$ for some true event $A$.
- If all $A \in \mathcal{A}$ are false at any stage, then the algorithm stops doing anything.
- Prioritize events $A_{i}$ with lower indices.


## Example Stage

## Example

Suppose $\left\{x_{0}, \ldots, x_{11}\right\}$ are independent fair coin flips and that $A_{k, \ell, n} \in \mathcal{A}$ for $(k, \ell, n)=(3,6,4)$ and $(k, \ell, n)=(0,5,4)$. If the current valuation is

$$
x_{0}, x_{1}, \ldots, x_{11}=011010010001,
$$

then $A_{3,6,4}$ is true. So, the resample algorithm takes new random samples for each $x_{i} \in \operatorname{vbl}\left(A_{3,6,4}\right)=[3,7) \cup[6,10)=[3,10)$. Suppose the resulting valuation is

$$
x_{0}, x_{1}, \ldots, x_{11}=011010110101 .
$$

This valuation makes $A_{0,5,4}$ true.
Resampling $A_{3,6,4}$ caused $A_{0,5,4}$ to go from false (good) to true (bad).

## Ingredients of Computable LLL

## Theorem (Constructive Lovász Local Lemma (Moser and Tardos, 2010))

Suppose the set of events $\mathcal{A}$ depending on variables $\mathcal{X}$ satisfy the conditions of the local lemma. Let $\tau_{n}$ be the first stage of the resample algorithm at which each of $A_{1}, A_{2}, \ldots, A_{n}$ is false. Then, $\mathbb{E}\left(\tau_{n}\right)<\infty$ for each $n$.

## Lemma (Rumyantsev and Shen, 2014)

Suppose the set of events $\mathcal{A}$ depending on variables $\{X\}$ satisfy the setup and conditions for the computable local lemma. Then, there is a computable function $f: \mathbb{N}^{2} \rightarrow \mathbb{N}$ such that

$$
\operatorname{Pr}\left(x_{i} \text { is resampled after stage } s\right) \leq f(i, s)
$$

and $\lim _{s \rightarrow \infty} f(i, s)=0$ for every $i$.

## A Computable Witness

## Lemma (Rumyantsev and Shen, 2014)

Suppose the hypotheses and conclusions of the previous lemma hold.
Then,
(1) With probability 1, the resample algorithm converges to a witness to the infinite LLL on $\mathcal{A}$ and $\mathcal{X}$.
(2) At least one of these witnesses is computable.

To compute initial segment $x_{1}, \ldots x_{n}$ of a witness:

- Simulate the resample algorithm for every possible resampling.
- Do this for enough steps $s$ to approximate the probability distribution on the final values of $x_{1}, \ldots, x_{n}$.
- Pick a valuation of $x_{1}, \ldots, x_{n}$ that has approximate probability greater than $\sum_{i=1}^{n} f(i, s)$.


## Analysis of the Moser-Tardos Algorithm

To understand why the resample algorithm converges, we track the causality behind each resampling.

## Tracking "Blame" in the Resample Algorithm

Suppose $\Gamma\left(A_{i}\right)=\left\{A_{i-1}, A_{i}, A_{i+1}\right\}$ and the first few events resampled by the Moser-Tardos algorithm begin with

$$
A_{1}, A_{2}, A_{5}, A_{4}, A_{3}, A_{6}, A_{5}, A_{5}, A_{4}, A_{6} .
$$

Reversing this initial segment yields

$$
A_{6}, A_{4}, A_{5}, A_{5}, A_{6}, A_{3}, A_{4}, A_{5}, A_{2}, A_{1} .
$$

Then, we can track "blame" for $A_{6}$ being true at the last step via the string
$A_{6}, A_{4}, A_{5}, A_{5}, A_{6}, A_{x}, A_{4}, A_{5}, A_{k}, A_{k}=A_{6}, A_{5}, A_{5}, A_{6}, A_{4}, A_{5}$.
We track "blame" for $A_{5}$ being true at the third step with the following sequence of length 1

$$
A_{5}, A_{x}, A_{x}=A_{5} .
$$

## Key Features of Moser-Tardos Algorithm

Let $A_{i_{1}}, A_{i_{2}}, A_{i_{3}}, \ldots$ be the log of events resampled by the a run of the Moser-Tardos algorithm.

- For each $A_{i_{j}}$ in the log, we can track the "blame" of why $A_{i_{j}}$ is true via a Moser-Tree
- The probability of large Moser trees appearing approaches zero
- The $\alpha$ in the statement of the computable local lemma gives us a computable bound on the rate.
- If $x_{i} \in \operatorname{vbl}(A)$ for finitely many $A$ then
$\operatorname{Pr}\left(x_{i}\right.$ is resampled after stage $\left.s\right)$ therefore also approaches zero in a uniformly computable way.

What if $x_{i} \in \operatorname{vbl}(A)$ for infinitely many $A$ ?

## What if $x_{i} \in \operatorname{vbl}(A)$ for Infinitely Many A?

If $x_{i} \in \operatorname{vbl}(A)$ for infinitely many $A$, then

- Large Moser trees are still just as rare,
- but small Moser trees can still cause $x_{i}$ to be resampled at late stages $s$.


## Example

Suppose $\mathcal{A}$ and $\mathcal{X}$ satisfy all conditions of the computable local lemma except that

$$
\left[x_{1} \in \operatorname{vbl}\left(A_{j}\right)\right] \Leftrightarrow\left[j=2^{n}\right] .
$$

If the singleton Moser tree

$$
A_{2 k}
$$

can occur for any $k$, then $x_{1}$ can be resampled arbitrarily late despite not being part of a long Moser tree.

## Modifying the Resample Algorithm

## Restricting the Resample Set

As in the previous slide, suppose that

$$
\left[x_{1} \in \operatorname{vbl}\left(A_{j}\right)\right] \Leftrightarrow\left[j=2^{n}\right]
$$

but also that

$$
\operatorname{Pr}\left(A_{j} \mid x_{1}=0\right)=\operatorname{Pr}\left(A_{j} \mid x_{0}=1\right)=\operatorname{Pr}\left(A_{j}\right)
$$

## Idea

We should able to get all of the previous results for a modified resample algorithm in which we only resample $\operatorname{vbl}\left(A_{2^{n}}\right) \backslash\left\{x_{1}\right\}$ when $A_{2^{n}}$ is the least true event in $\mathcal{A}$.

## Restricting the Resample Set

## Idea

In general, for each $A \in \mathcal{A}$, specify a subset $\operatorname{rsp}(A) \subset \operatorname{vbl}(A)$ of variables to resample. Then, we should redefine the neighborhood relation $\Gamma$ by

$$
\Gamma(A)=\{B \in \mathcal{A}: \operatorname{rsp}(A) \cap \operatorname{rsp}(B) \neq \emptyset\}
$$

for our analysis of this modified resample algorithm.

- Let $\operatorname{stc}(A)=\operatorname{vbl}(A) \backslash \operatorname{rsp}(A)$.
- We also require that $\max (\operatorname{stc}(A))<\min (\operatorname{rsp}(A))$.


## Priority of Events for Resampling

- Not resampling the full set $\operatorname{vbl}(A)$ of variables that determine $A$ obfuscates the "blame tracking" feature of the Moser-sequences
- To recover this, we resample events in a specific order


## Idea

Fix a linear order $\prec$ on $\mathcal{A}$ such that

$$
[\max (\operatorname{rsp}(A))<\max (\operatorname{rsp}(B))] \Rightarrow[A \prec B] .
$$

Where max is calculated based on the indices of the variables in $\mathcal{X}$.

The modified resample algorithm chooses the $\prec$-least true event to resample at each stage.

## Constraints on rsp(A)

- Let $A \ll B$ if and only if $A \prec B$ and $A \notin \Gamma(B)$ (i.e. $\operatorname{rsp}(A)$ and $\operatorname{rsp}(B)$ are disjoint).
- To ensure that $\ll$ is transitive, we also impose that $\operatorname{rsp}(A)$ be an interval $[\min (\operatorname{rsp}(A)), \max (\operatorname{rsp}(A))]$.
- We also require analogous conditions on $\operatorname{rsp}(A)$ as we did for $\operatorname{vbl}(A)$ in the effectivisation of the original LLL.


## Computable "Lefthanded" LLL

Under all of the conditions previously described,

## Theorem (M.)

Suppose there is $P^{*}(\mathcal{A}) \rightarrow[0,1]$ such that for each $A \in \mathcal{A}$ and each valuation $\mu$ of the variables in $\operatorname{stc}(A)$,

$$
P^{*}(A) \geq \operatorname{Pr}(A \mid x=\mu(x) \text { for all } x \in \operatorname{stc}(A))
$$

Furthermore, suppose there is computable $z: \mathcal{A} \rightarrow(0,1)$ and $\alpha \in(0,1)$ such that, for each $A \in \mathcal{A}$,

$$
P^{*}(A) \leq \alpha z(A) \prod_{B \in \Gamma(A)}(1-z(B)) .
$$

Then, there is a computable valuation of $\mathcal{X}$ under which each $A \in \mathcal{A}$ is false.

## Computable Non-Repetitive Sequences

Thus, we can compute a sequence whose long identical intervals are far apart:

## Corollary

For each $\varepsilon>0$ there is an $N_{\varepsilon}$ and a computable $\{0,1\}$-valued sequence such that any two identical intervals of length $n>N_{\varepsilon}$ have distance greater than $(2-\varepsilon)^{n}$.

We can also compute a sequence whose adjacent intervals are very different.

## Corollary

For each $\varepsilon>0$, there is an $N_{\varepsilon}$ and a computable $\{0,1\}$-valued sequence such that any two adjacent intervals of length $n>N_{\varepsilon}$ have share at most $\left(\frac{1}{2}-\varepsilon\right) n$ many entries.

Game versions of these applications appear in (Pegden, 2011).

## Outlook

## Further Questions

- Can we compute winning strategies to the games studied by Pegden (2011)?
- Almost: we can compute a winning sequence of moves from an opposing strategy
- Can we make conditions for the computable version of the "Lefthanded" LLL closer to the conditions of the original?
- Is the computable "Lefthanded" LLL useful in complexity theory?

Thank you!

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