

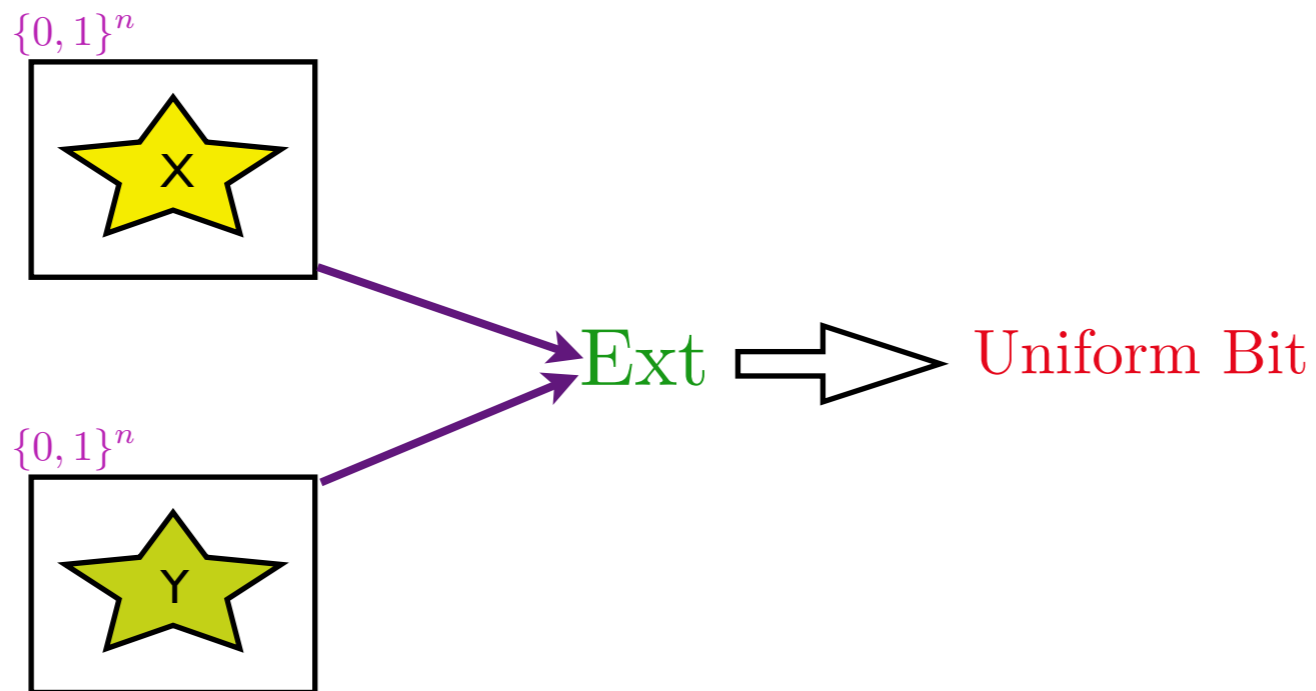
Part 3: Explicit 2-Source Extractors

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STOC 2018: Extractor Workshop

2-Source Extractor

X, Y are independent (n, k) -sources



More formally,

$$\left| \Pr[\text{Ext}(X, Y) = 1] - \frac{1}{2} \right| \leq \epsilon$$

Ramsey Graphs

K-Ramsey graph: No independent set or clique of size K .

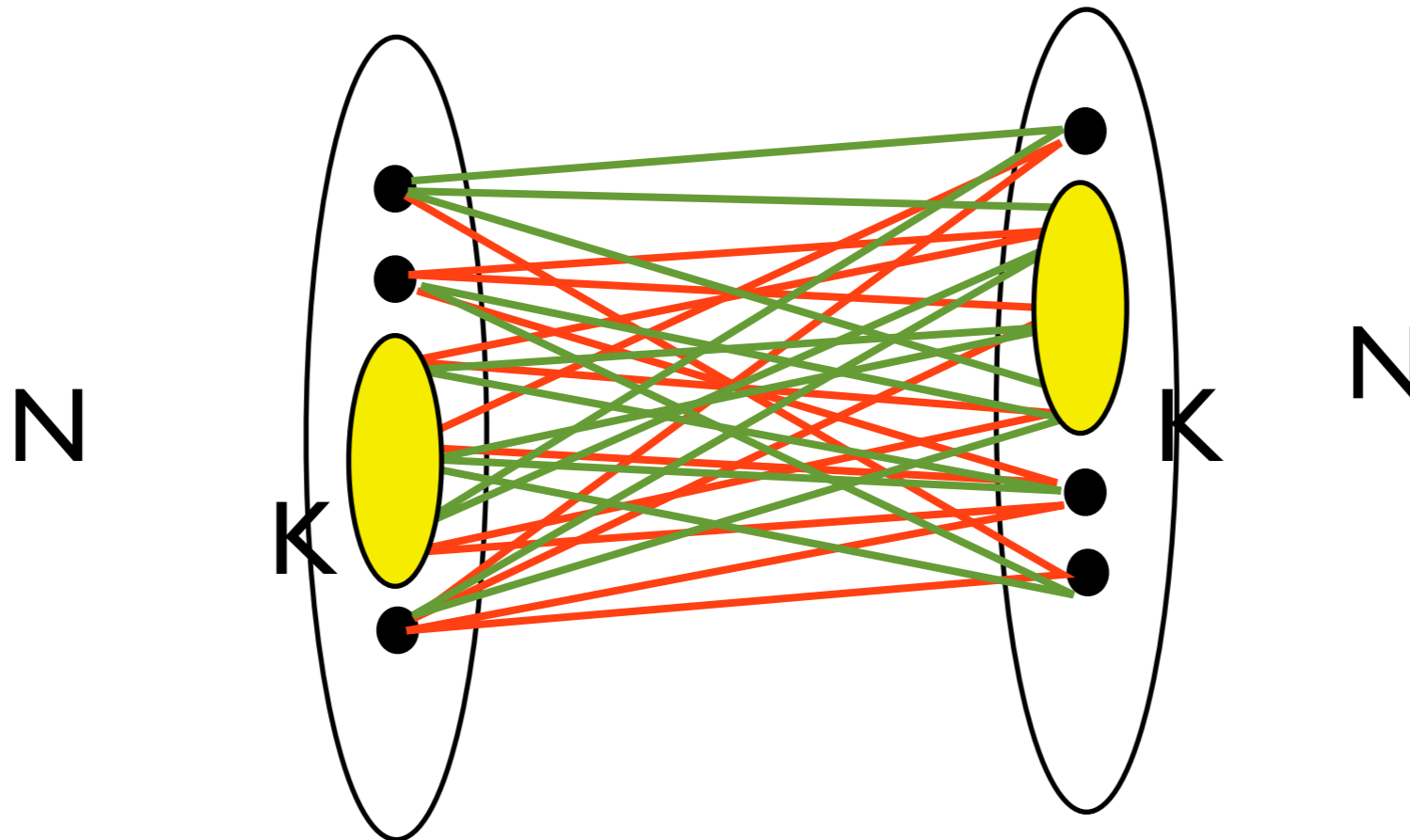
Bipartite K-Ramsey graph: Bipartite graph with no complete or empty $K \times K$ sub-graph.

Erdos (1947): Existence of K-Ramsey graphs on N vertices for $K > (2+o(1)) \log N$.

Explicit Constructions?

Ramsey Graphs

$$N=2^n, \quad K=2^k$$



Explicit 2-Source Extractors

Reference	k_1
Chor-Goldreich 88	$>0.5n$
Bourgain 05	$\geq 0.499n$
Recent advances	$\log n \ o(\log \log n)$

Explicit Ramsey Graphs ($N=2^n$, $K=2^k$)

Reference	K	Bipartite
Erdős 47 (existential)	$\geq 2 \log N$	Yes
Hadamard Matrix	\sqrt{N}	Yes
Frankl-Wilson81	$2^{\sqrt{\log N}}$	No
Best 2-Source Extractors	$(\log N)^{o(\log \log \log N)}$	Yes

Tools and Techniques

Resilient Function

$$f: \{0, 1\}^n \rightarrow \{0, 1\}$$

(q, ϵ) -resilient function: For any subset of q co-ordinates, probability f is **fixed** on **uniform sampling** of the remaining co-ordinates is $\geq 1 - \epsilon$.

Example: MAJORITY is $(n^{0.49}, \epsilon)$ -resilient.

PARITY is **NOT** (q, ϵ) -resilient for any $q > 0, \epsilon < 1$.

Resilient Functions

- **[C-Zuckerman]** Explicit $(n^{0.99}, \epsilon)$ -resilient functions.
- **[Meka]** Explicit $(n/\log^2 n, \epsilon)$ -resilient functions.

Strong Seeded Extractors

$$\text{Ext}: \{0, 1\}^n \times \{0, 1\}^d \rightarrow \{0, 1\}^m$$

X : (n, k) -source

Y : U_d

$$\text{Ext}(X, Y), Y \approx_{\varepsilon} U_m, Y$$

Explicit Construction: $d = O(\log(n/\varepsilon))$, $m = \Omega(k)$.

t-Non-Malleable Extractors

$\text{nmExt}:\{0,1\}^n \times \{0,1\}^d \rightarrow \{0,1\}$,
non-malleable

$\exists S \subseteq \{0,1\}^d, \quad |S| > (1-\epsilon)2^d$,
for any distinct seeds s_1, \dots, s_t in S

$\text{nmExt}(X, s_1), \dots, \text{nmExt}(X, s_t) \approx_{\epsilon} U_t$

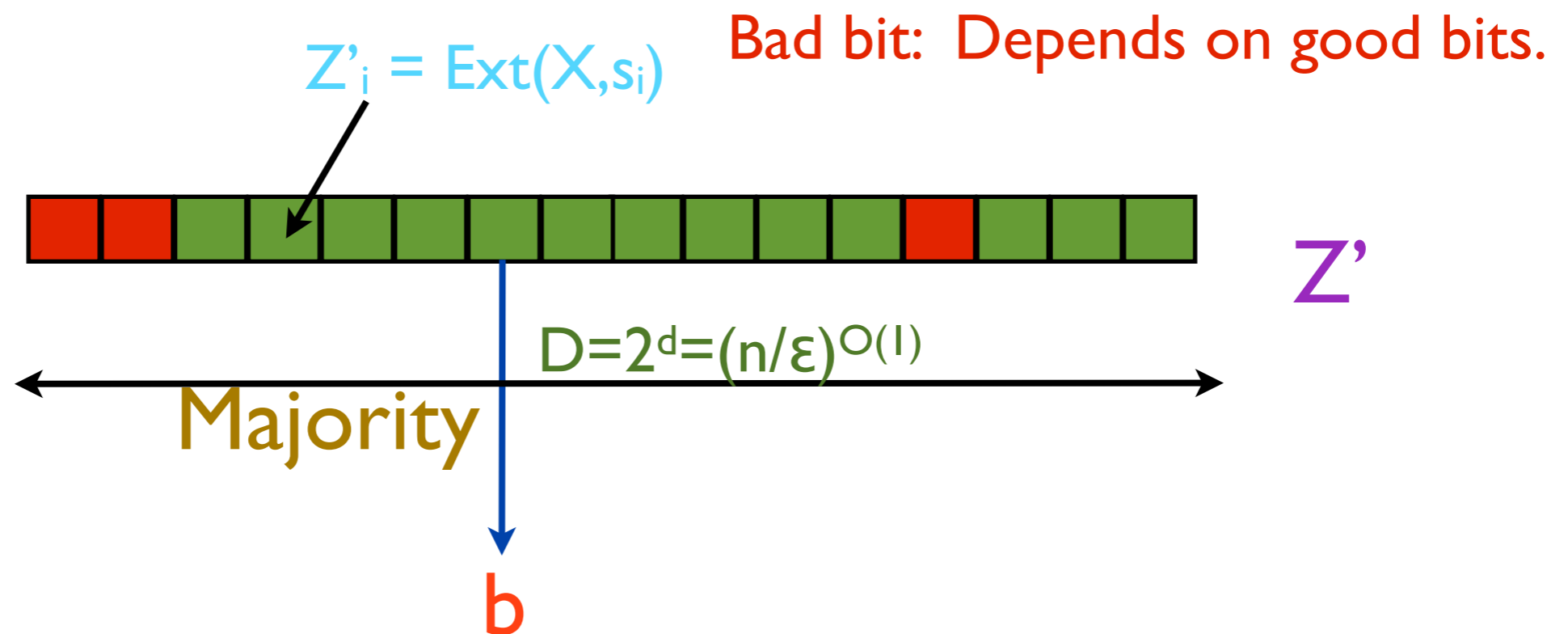
Explicit: $k = t \log(n/\epsilon) \log \log(n/\epsilon)$,
 $d = O(t^2 \log(n/\epsilon) (\log \log(n/\epsilon)))$

A Preliminary Attempt

$\text{Ext}: \{0, 1\}^n \times \{0, 1\}^d \rightarrow \{0, 1\}$: Strong-seeded extractor
Min-entropy k , error ϵ .

X : (n, k) -source

$(1 - \epsilon)$ fraction of the bits in Z' are uniform



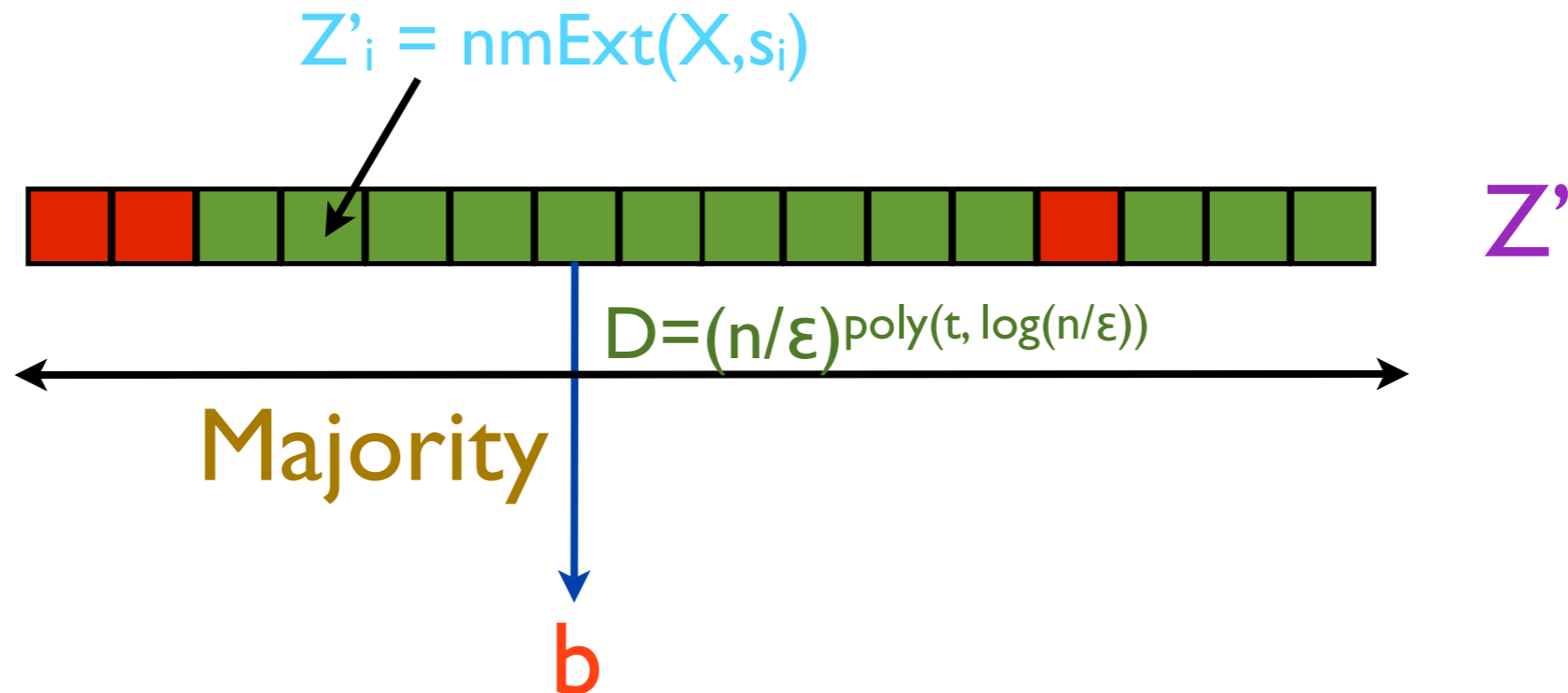
Does not work: The uniform bits are arbitrarily correlated

A Second Attempt

Idea: Make the uniform bits almost t -wise independent

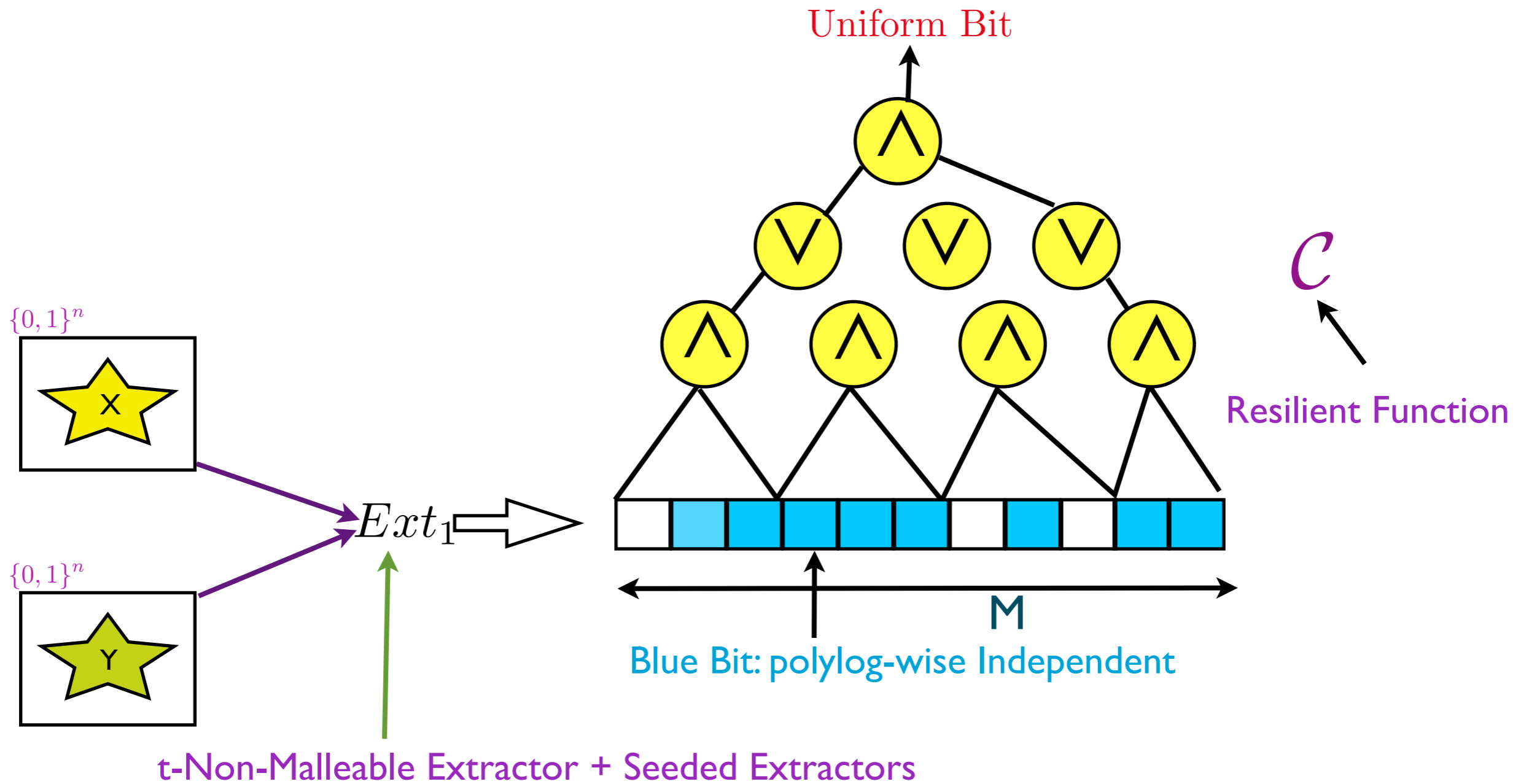
Use a t -non-malleable extractor

X : (n,k) -source



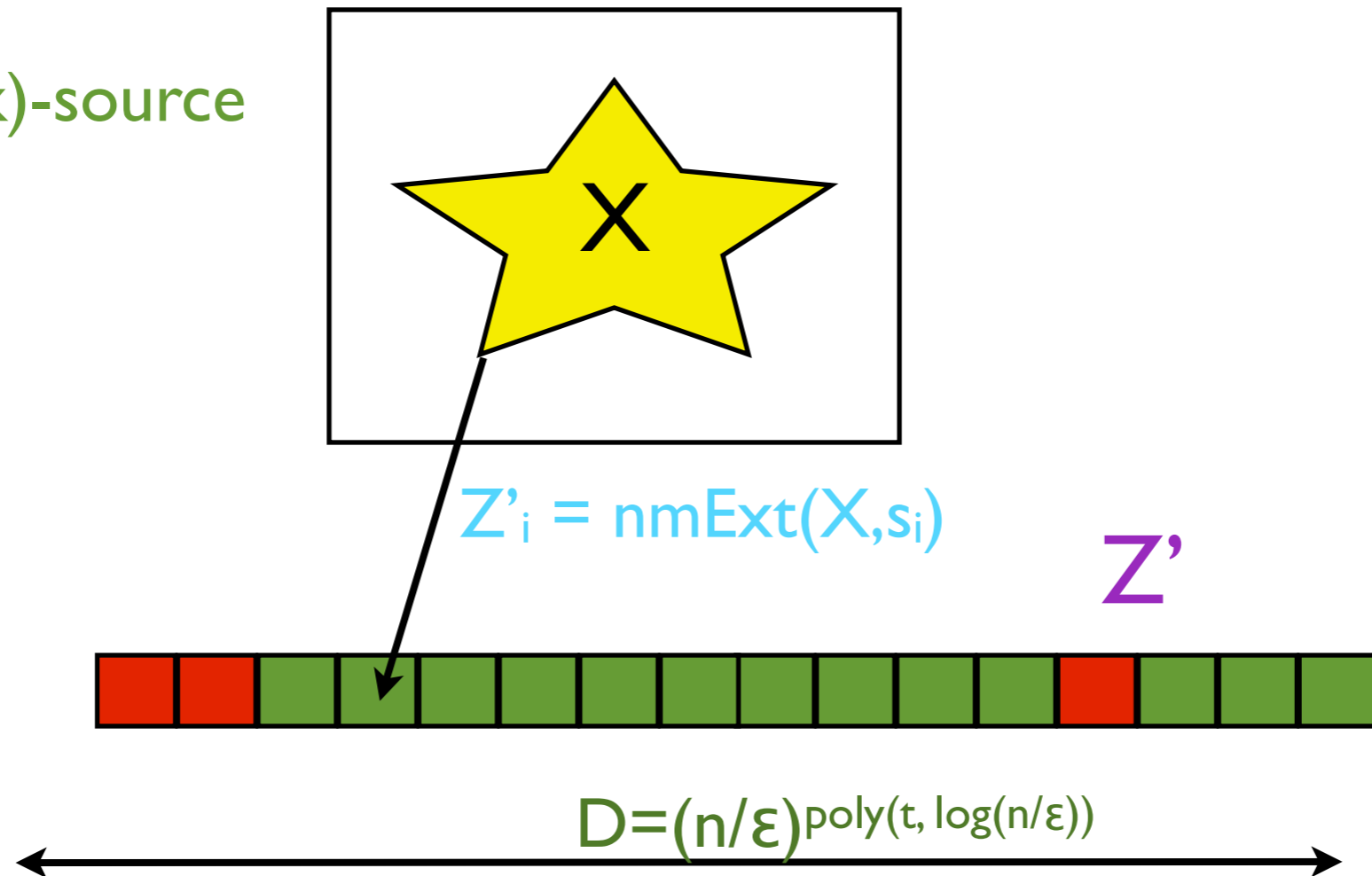
Does not work: $> D^{0.5}$ bad bits.
(not surprisingly! since we have 1 source)

A (very) High Level Idea of our Construction



Executing Step I

X: (n,k)-source

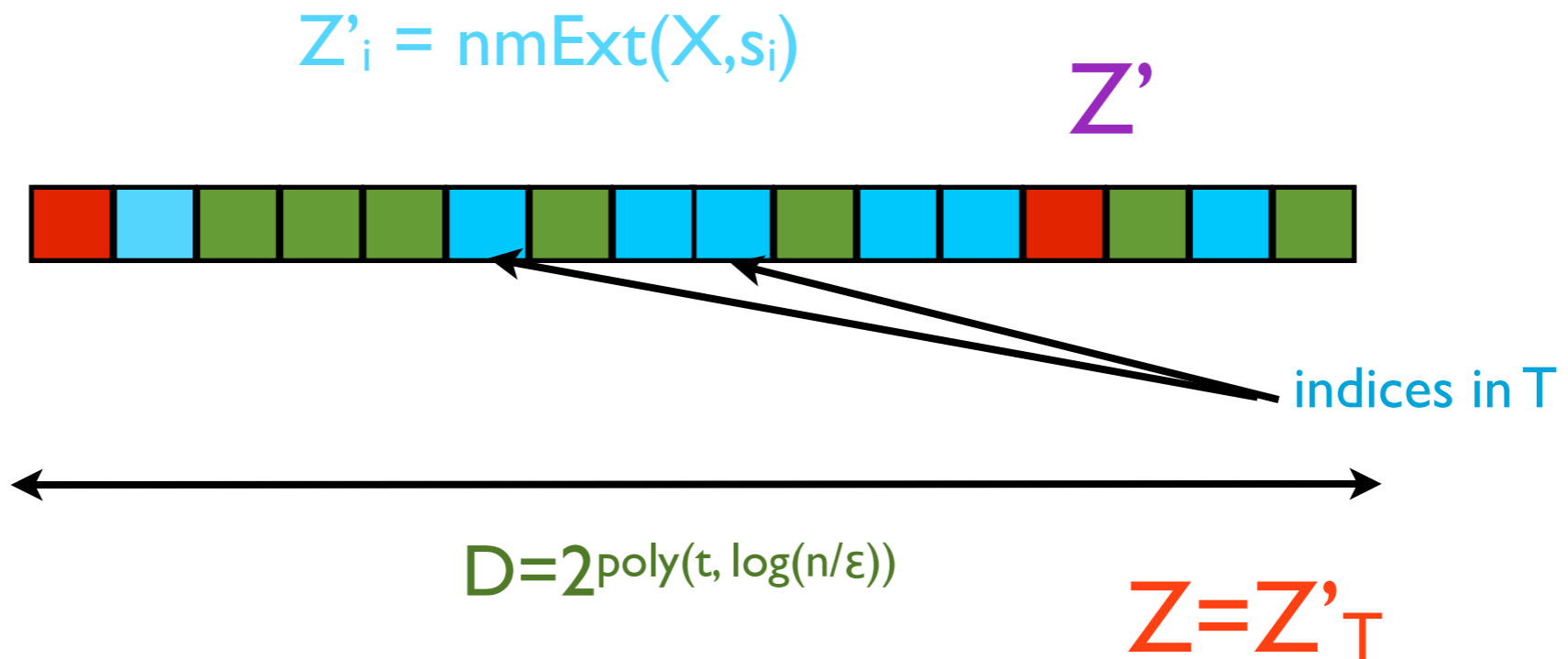


The good bits of Z are almost t -wise independent

of bad indices $\leq \epsilon D$

Idea: Sample a pseudorandom subset T of $[D]$ using Y .

Executing Step I



Pseudorandom Subset:

$$T = \{\text{Ext}(Y, r_1), \dots, \text{Ext}(Y, r_M)\}, M = 2^{O(\log(n/\epsilon'))}$$

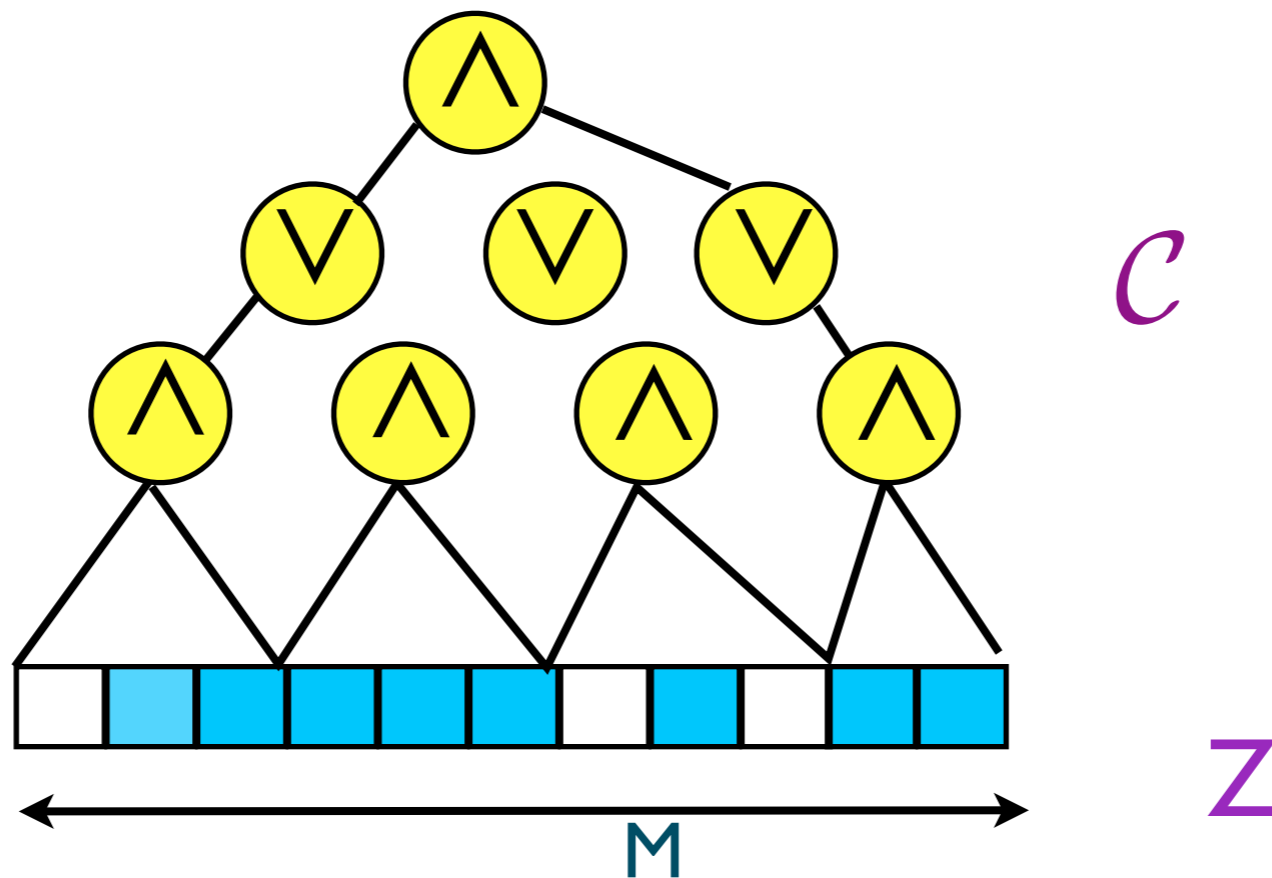
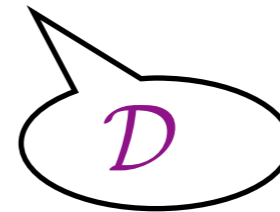
$$\text{No. of bad indices in } T: (\epsilon + \epsilon')M < M^{1-\delta}$$

Executing Step 2

No. of bad indices $< M^{1-\delta}$

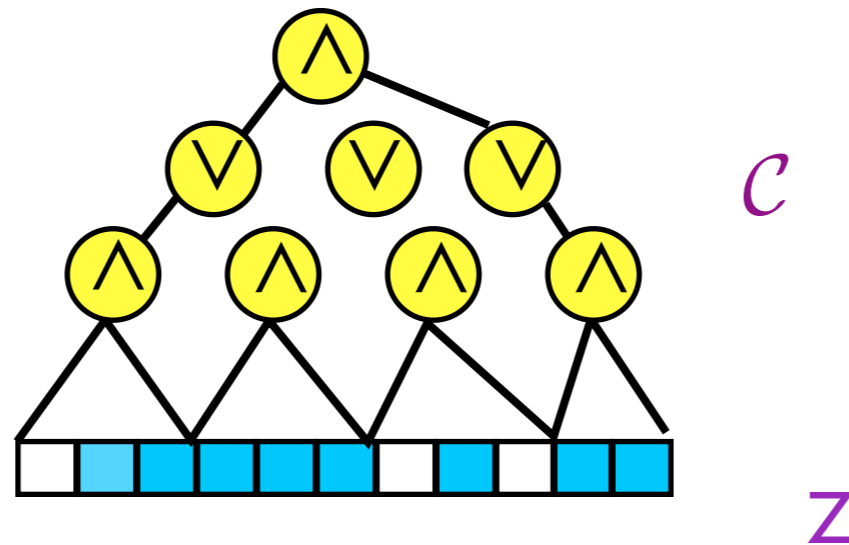
The good bits of Z are ($t = \text{polylog}(M)$)-wise independent.

$Pr_{x \sim \mathcal{D}}[\mathcal{C}(x) \text{ is not fixed}] = ?$



Executing Step 2

Step 2a: Good bits can be assumed to uniform, independent



$$\mathcal{C}'(x) := (\mathcal{C}(x \circ \vec{0}) \neq \mathcal{C}(x \circ \vec{1}))$$

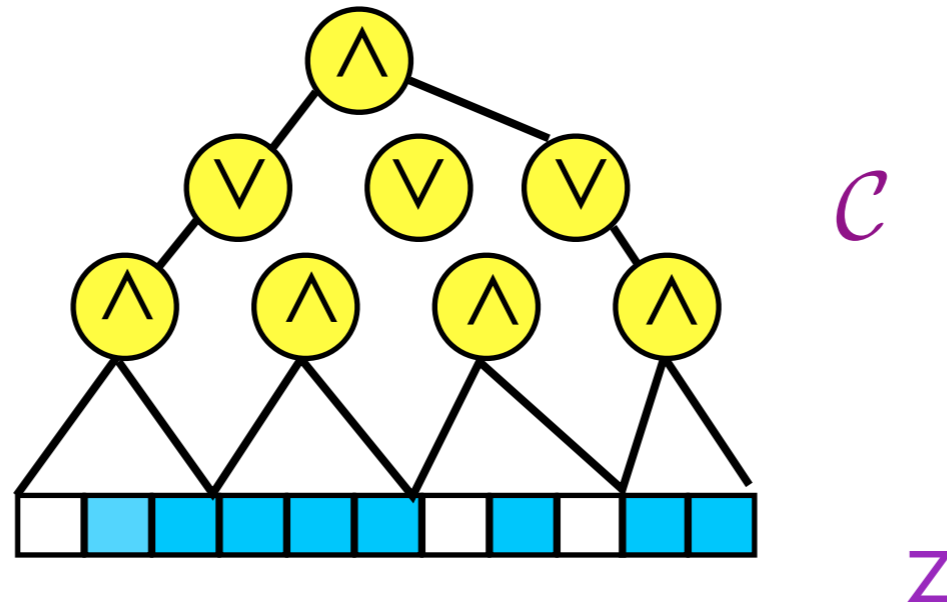
If \mathcal{C} is monotone,

$$\Pr_{x \sim \mathcal{D}}[\mathcal{C}(x) \text{ is not fixed}] = \Pr_{x \sim \mathcal{D}}[\mathcal{C}'(x) = 1]$$

(Braverman 07) If \mathcal{C} is in AC^0 ,

$$|\Pr_{x \sim \mathcal{D}}[\mathcal{C}'(x) = 1] - \Pr_{x \sim U}[\mathcal{C}'(x) = 1]| \leq \epsilon$$

Executing Step 2



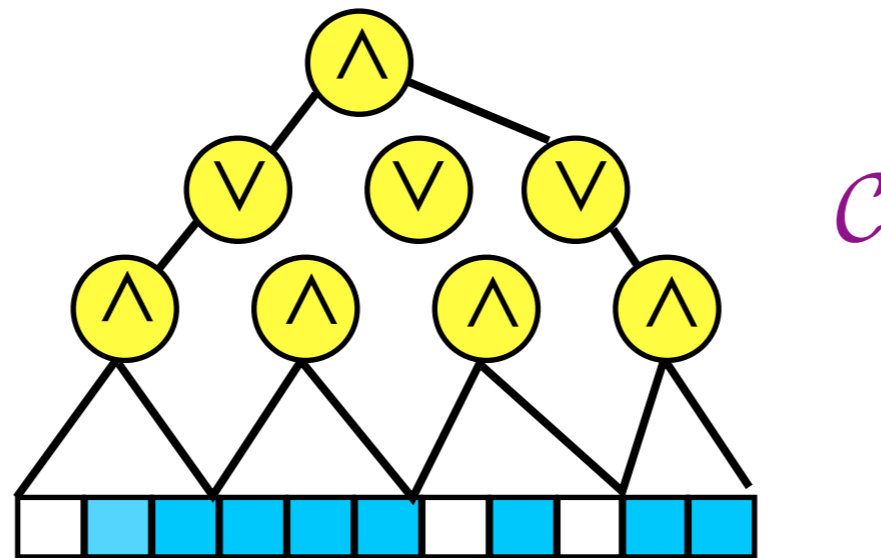
$$\mathcal{C}'(x) := (\mathcal{C}(x \circ \vec{0}) \neq \mathcal{C}(x \circ \vec{1}))$$

Thus, if \mathcal{C} is in AC^0 and monotone,

$$|\Pr_{x \sim \mathcal{D}}[\mathcal{C}(x) \text{ is not fixed}] - \Pr_{x \sim U}[\mathcal{C}(x) \text{ is not fixed}]| \leq \epsilon$$

Thus, we can assume good bits are **independent, uniform**.

Executing Step 2



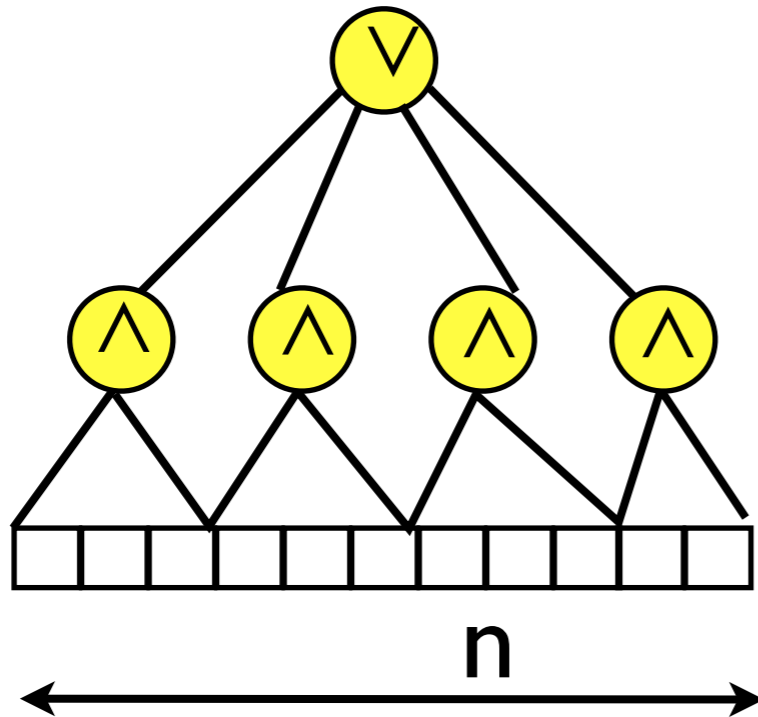
Remaining Task:

Explicit construction of a monotone C in AC^0 on n bits s.t:

- (1) C is $(n^{1-\delta}, \epsilon)$ -resilient and
- (2) almost unbiased.

Ajtai-Linial Function

Tribes function:



AL Function: $T_1 \wedge T_2 \dots \wedge T_n$

T_i : randomly-negated Tribes on randomly chosen partitions.

Resilient to coalitions of size $O(n/\log^2 n)$.

Problems: (1) Probabilistic: Naive derandomization takes time $n^{O(n^2)}$.

(2) Not Monotone.

Derandomizing Ajtai-Linial

Key Ingredient: An explicit construction of a collection of partitions of $[n]$ s.t:


$$n^{1-\delta}$$

(1) Any small subset of $[n]$ has small intersection with most partitions.



Used to bound influence

(2) The partitions are pairwise pseudorandom:
- the intersection of any two blocks is bounded.



Used to bound bias

Derandomized via extractors!

Open Questions

- **Negligible error:** All constructions known achieve error $1/n^{\Omega(1)}$; not enough for cryptographic applications.
- **Extractors for other sources:** Low degree varieties, Circuit sources, ...
 - Applications to circuit lower bounds
- **Optimal Ramsey graphs**
- **Other applications of Non-malleable extractors**
- **Simpler constructions**

Thank You!