

Low-Error Two-Source extractors from efficient non-malleable extractors

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Joint work with
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Today's talk

- * Two-source extractors.
- * Non-malleable extractors.
- * Current constructions of two-source extractors via non-malleable extractors and where they fail in achieving small error.
- * Constructing low-error two-source extractors given “good” non-malleable extractors.

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Two-source extractors

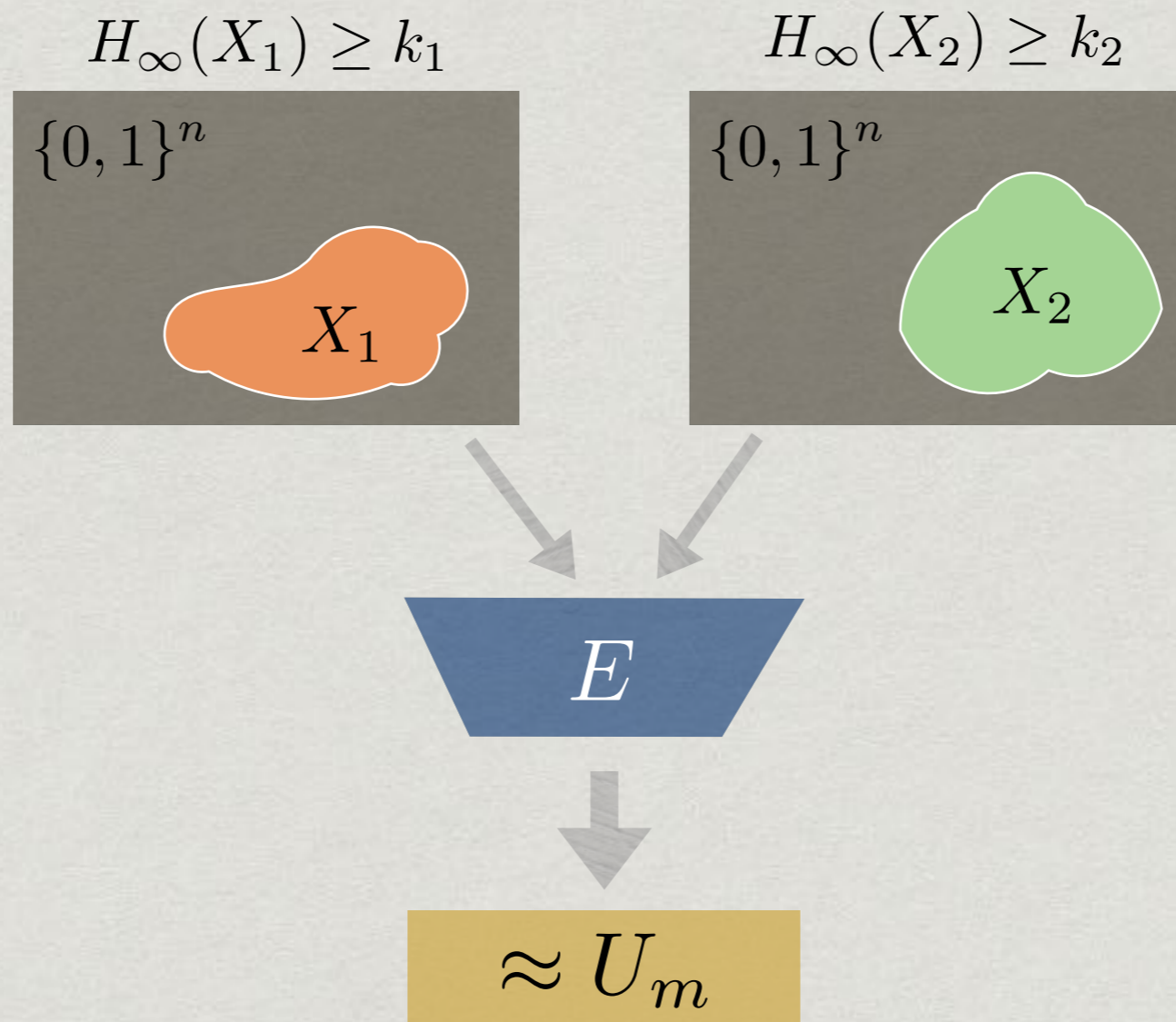
Two-source extractors

- * We say that a source X over $\{0,1\}^n$ has k min-entropy if for every x , $\Pr[X=x] \leq 2^{-k}$. This is how we model *weak* sources.
- * Alternatively, we can think of a weak source X as uniformly distributed over a subset of size 2^k .

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- * Alternatively, we can think of a weak source X as uniformly distributed over a subset of size 2^k .
- * Given two **independent** weak source X_1 and X_2 , we want to extract almost-uniform bits (potentially, almost all the entropy).

Two-source extractors



$$E(X_1, X_2) \approx_\varepsilon U_m$$

Two-source extractors

- * Known results for **constant** error.
- * Omitted here: many constructions of multi-source extractors.

	min-entropy
Non-explicit	$\log n + O(1)$
[CG88]	$(\frac{1}{2} + \delta)n$
[Raz05]	$(\frac{1}{2} + \delta)n, O(\log n)$
[Bourgain05]	$0.499n$
[CZ15]	$\text{polylog}(n)$
[BDT16]	$\log^{1+o(1)} n$
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A closer look at the error

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- * Only the constructions of Chor-Goldreich, Raz and Bourgain achieve this.
- * We will soon see where recent constructions fall short.
- * Viewing it differently: We want the construction to run in time $\text{polylog}(1/\varepsilon)$ instead of $\text{poly}(1/\varepsilon)$.

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Our goal: Low-error two-source extractors, even for δn min-entropy.

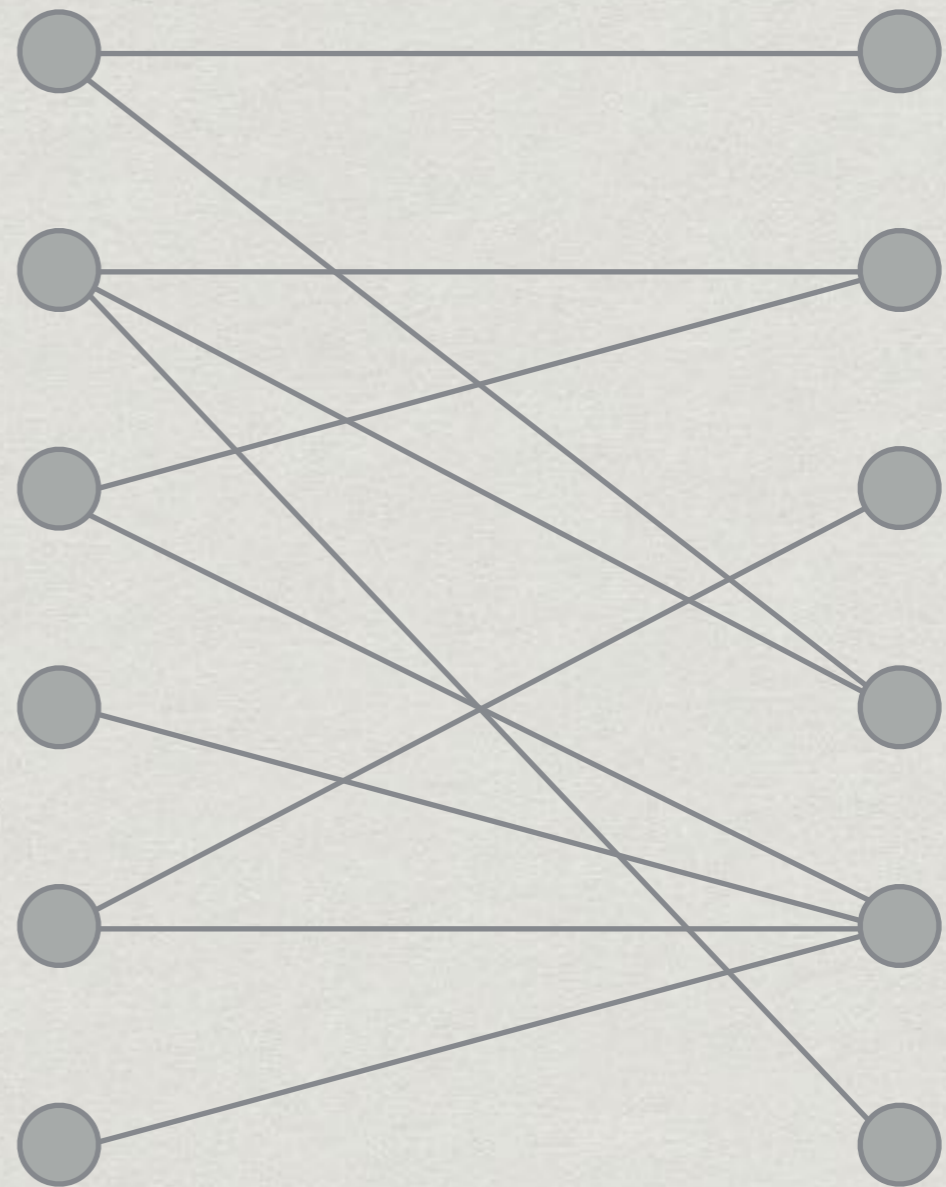
(Preferably outputting many bits as well, but it often goes together...)

Bipartite Ramsey graphs

- * The very-high error case is also interesting...
- * In every $N \times N$ bipartite graph there is a $\frac{1}{2} \log N \times \frac{1}{2} \log N$ monochromatic subgraph (a bipartite clique or an independent set).

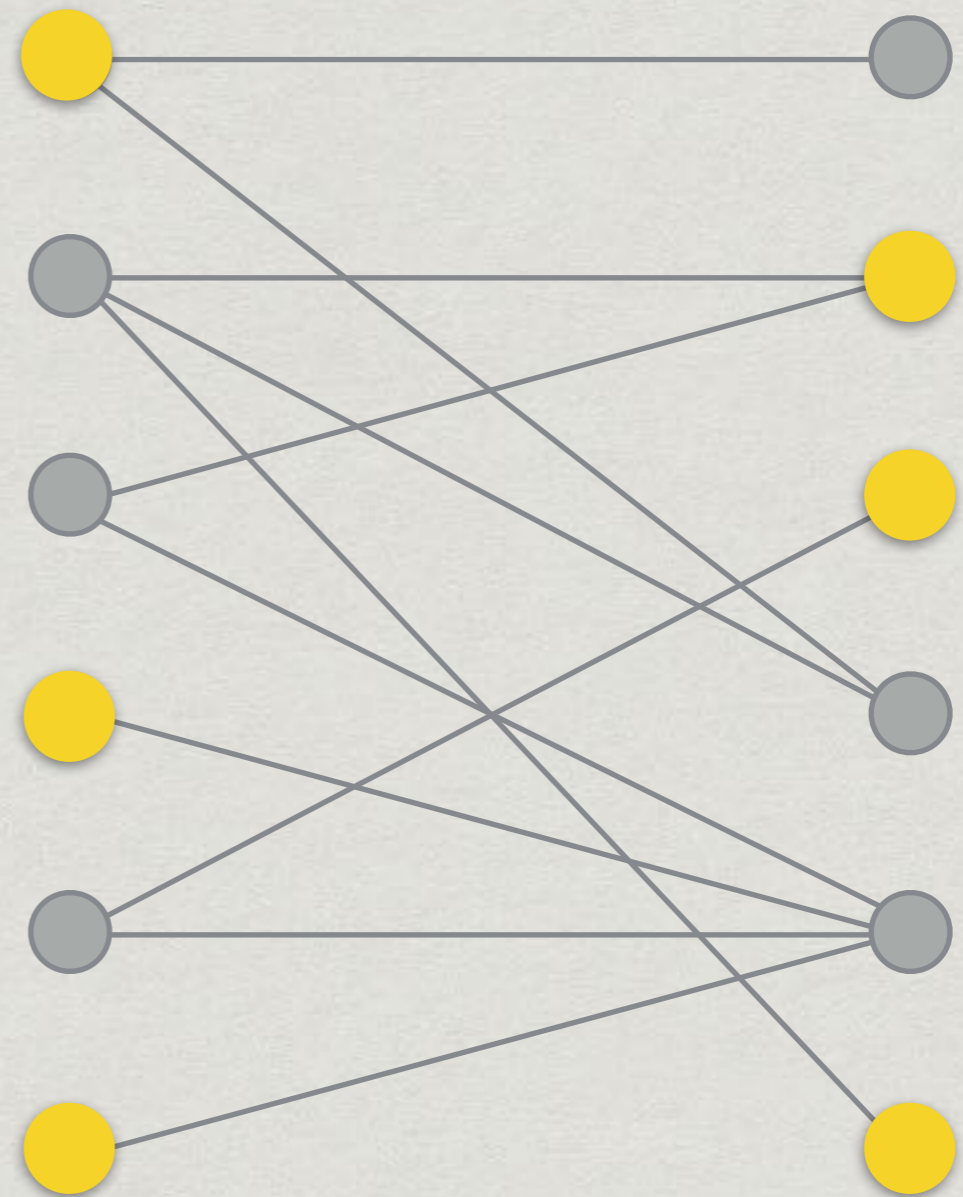
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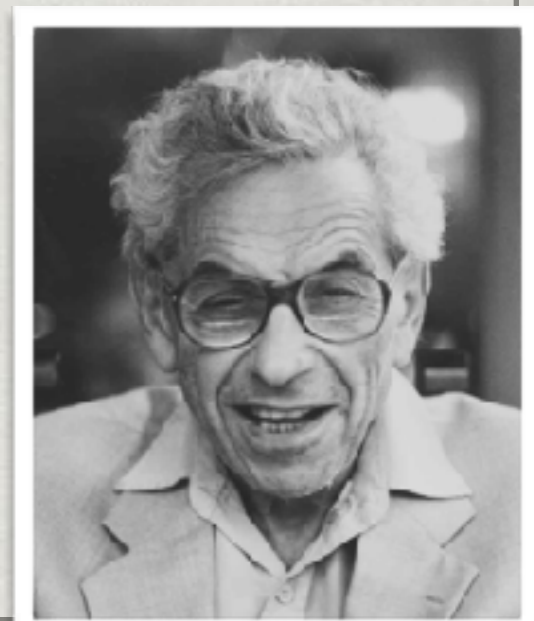
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Bipartite Ramsey graphs

- * Erdős (1947) — there exists an $N \times N$ bipartite graph with **no** $K \times K$ monochromatic subgraphs, for $K=2\log N$.
- * A random graph has this property.
- * The Erdős \$100 challenge — find such an explicit graph, even for $K=O(\log N)$.
- * Still open...



Bipartite Ramsey graph

- * We can view every bipartite graph naturally as a function $E:[N]\times[N]\rightarrow\{0,1\}$.
- * The bipartite Ramsey problem: construct explicit matrices with no $K\times K$ **constant** sub-matrices.

N

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

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Bipartite Ramsey graph

- * We can view every bipartite graph naturally as a function $E:[N] \times [N] \rightarrow \{0,1\}$.
- * The bipartite Ramsey problem: construct explicit matrices with no $K \times K$ **constant** sub-matrices.
- * The low-error two-source extractors problem: Insist on **unbiased** sub-matrices, with a **very small bias**.

N

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0	0	0	1	1	0	0	1	0
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0	1	0	0	0	0	1	0	0
1	1	0	1	0	1	0	1	0
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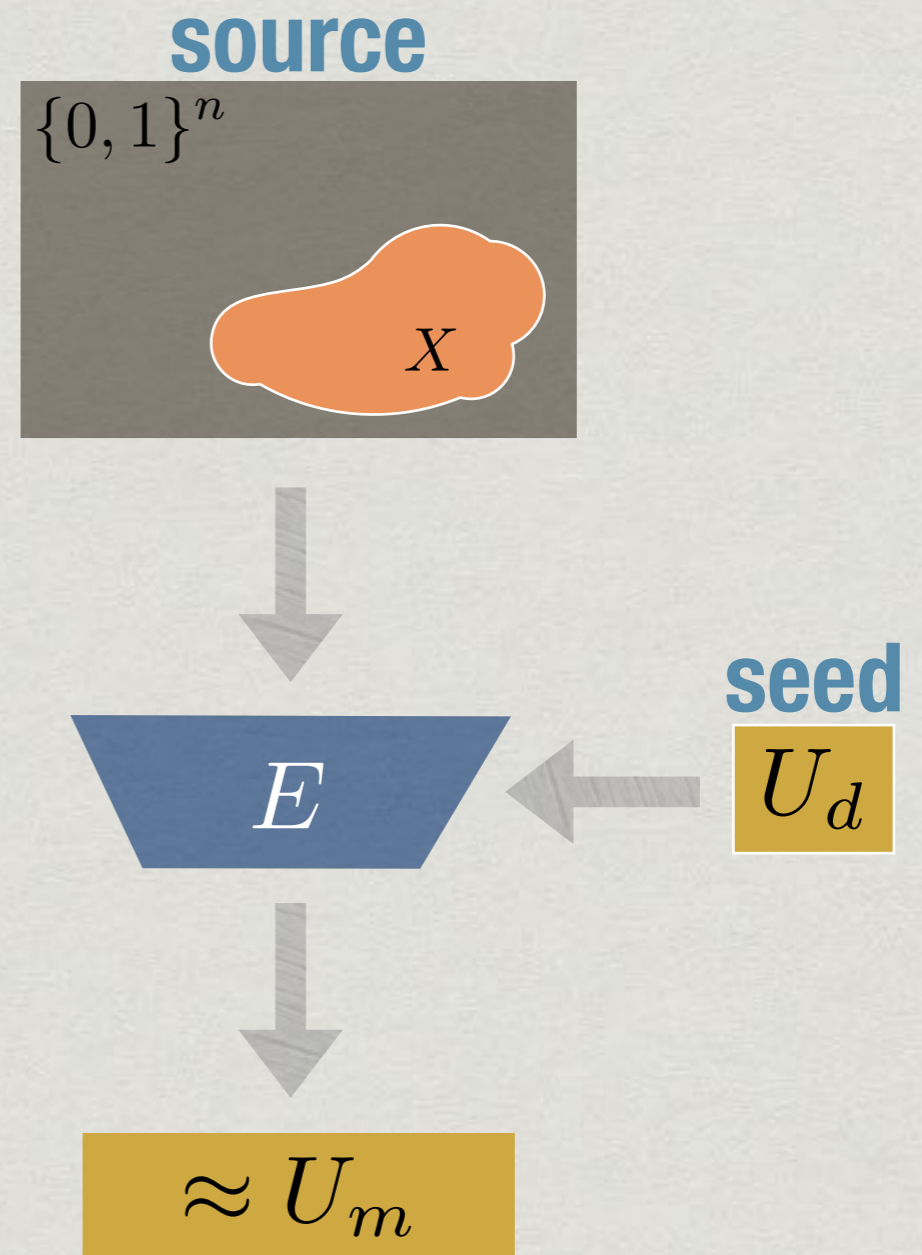
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Seeded extractors

- * A special case of two-source extractors is when one source is completely uniform, the *seed*.
- * The seed length can be as small as $2\log(n/\epsilon)$.



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- * We say a seeded extractor is **strong** if the output is uniform even given the seed: $(E(X, Y), Y) \approx_\varepsilon (U, Y)$.

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- * We say a seeded extractor is **strong** if the output is uniform even given the seed: $(E(X, Y), Y) \approx_\varepsilon (U, Y)$.
- * Equivalently, for every source X with entropy at least k there exists a set of **good** seeds of density at least $1-\varepsilon$ such that for every good seed $y \in \{0, 1\}^d$, $E(X, y) \approx_\varepsilon U$.
- * We have good strong seeded extractors [LRVW03, GUV07, ...].

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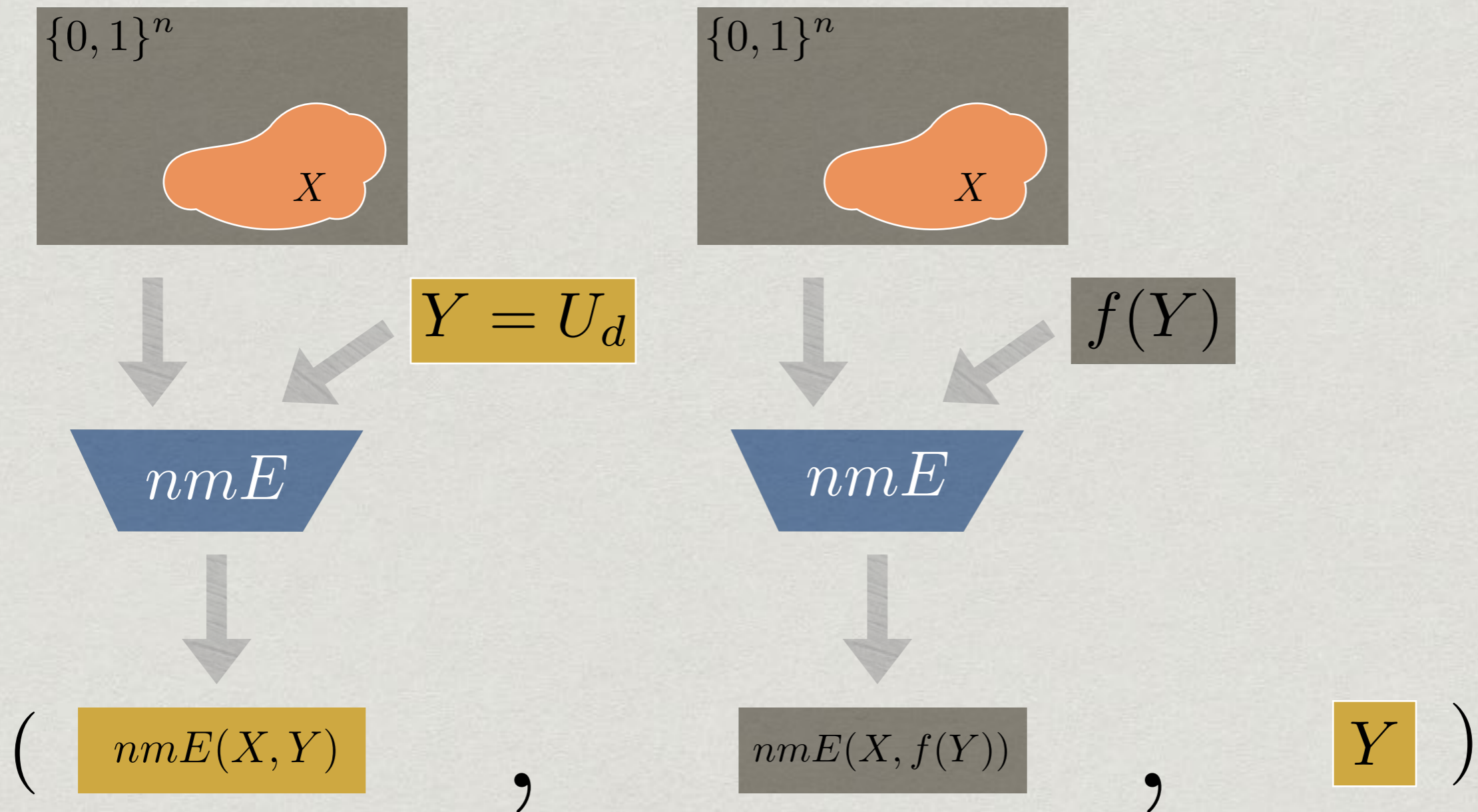
- * A generalization of strong seeded-extractors.
- * An adversary cannot distinguish between the output $\text{nmE}(X, Y)$ and a uniform string, even given the seed Y and the output of nmE on t **correlated** seeds.

Non-malleable extractors [DW09]

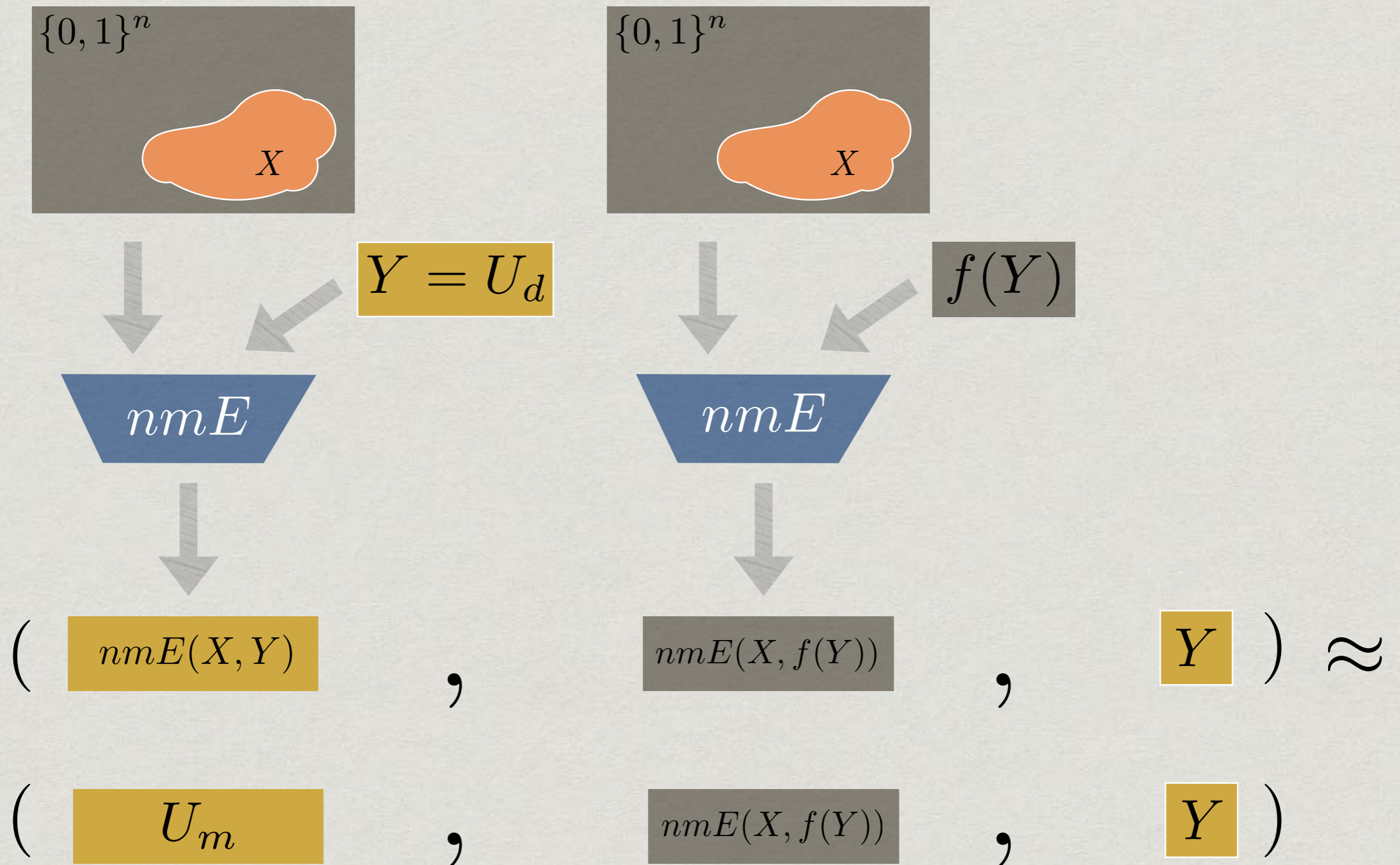
- * A generalization of strong seeded-extractors.
- * An adversary cannot distinguish between the output $\text{nmE}(X, Y)$ and a uniform string, even given the seed Y and the output of nmE on t **correlated** seeds.
- * $(\text{nmE}(X, Y), \text{nmE}(X, f_1(Y)), \dots, \text{nmE}(X, f_t(Y)), Y)$ is ε -close to $(U, \text{nmE}(X, f_1(Y)), \dots, \text{nmE}(X, f_t(Y)), Y)$.

Non-malleable extractors

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Non-malleable extractors

- * Known explicit constructions for $t=1$ (a partial list). A reduction by [Cohen16] allows us to go to an arbitrary t by roughly paying a factor of t in the entropy and t^2 in the seed-length.

	seed length	min-entropy
[CRS12,DLWZ11]	$\log(n/\varepsilon)$	$(1/2+\delta)n$
[Li12]	$\log(n/\varepsilon)$	$0.499n$
[CGL15]	$\log^2(n/\varepsilon)$	$\Omega(d)$
[Cohen16]	$\log(n/\varepsilon)\log(\log(n)/\varepsilon)$	$\Omega(d)$
[CL16]	$\log^{1+o(1)}(n/\varepsilon)$	$\Omega(d)$
[Cohen16]	$\log(n)+\log(1/\varepsilon)\text{poly}(\log\log(1/\varepsilon))$	$\Omega(d)$
[Li16]	$\log(n)+\log(1/\varepsilon)\log\log(1/\varepsilon)$	$\Omega(d)$

Non-malleable extractors

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- * We will use an equivalent definition (up to some loss in the error) [CZ15,Cohen16].
- * nmE is a n.m. extractor if every source induces a set of **good** seeds of high density such that the output of the extractor on a good seed is close to uniform even conditioned on its output on t other distinct seeds.

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- * nmE is a n.m. extractor if every source induces a set of **good** seeds of high density such that the output of the extractor on a good seed is close to uniform even conditioned on its output on t other distinct seeds.
- * For every X there exists a set of G of density at least $1-\varepsilon$ such that for every $y \in G$ and any $y_1, \dots, y_t \in \{0,1\}^d \setminus \{y\}$ it holds that $(\text{nmE}(X, \mathbf{y}), \text{nmE}(X, \mathbf{y}_1), \dots, \text{nmE}(X, \mathbf{y}_t))$ is ε -close to $(U, \text{nmE}(X, \mathbf{y}_1), \dots, \text{nmE}(X, \mathbf{y}_t))$.

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Current constructions of two-source extractors

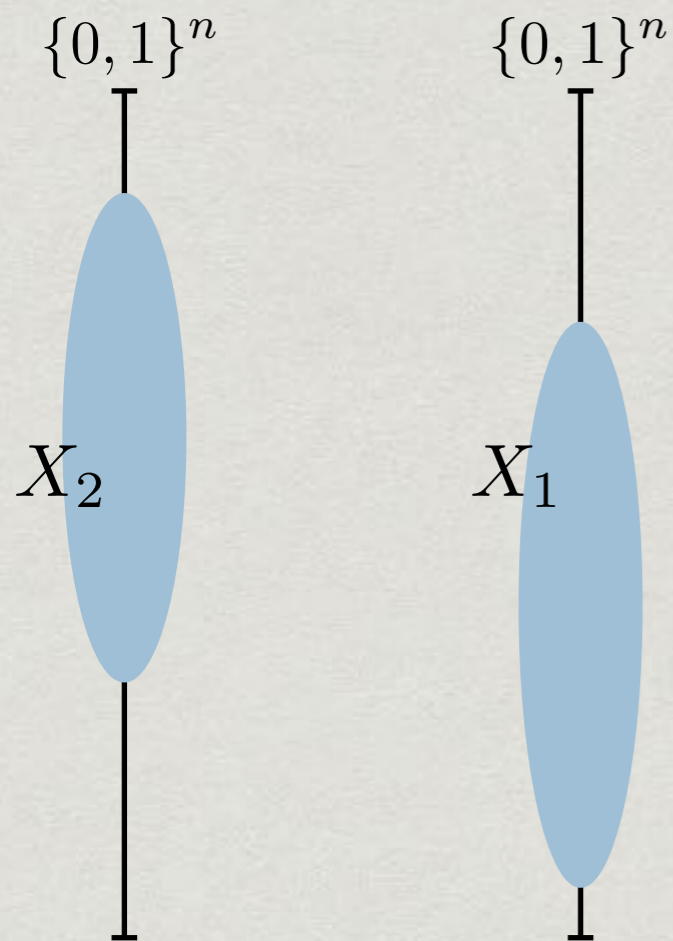
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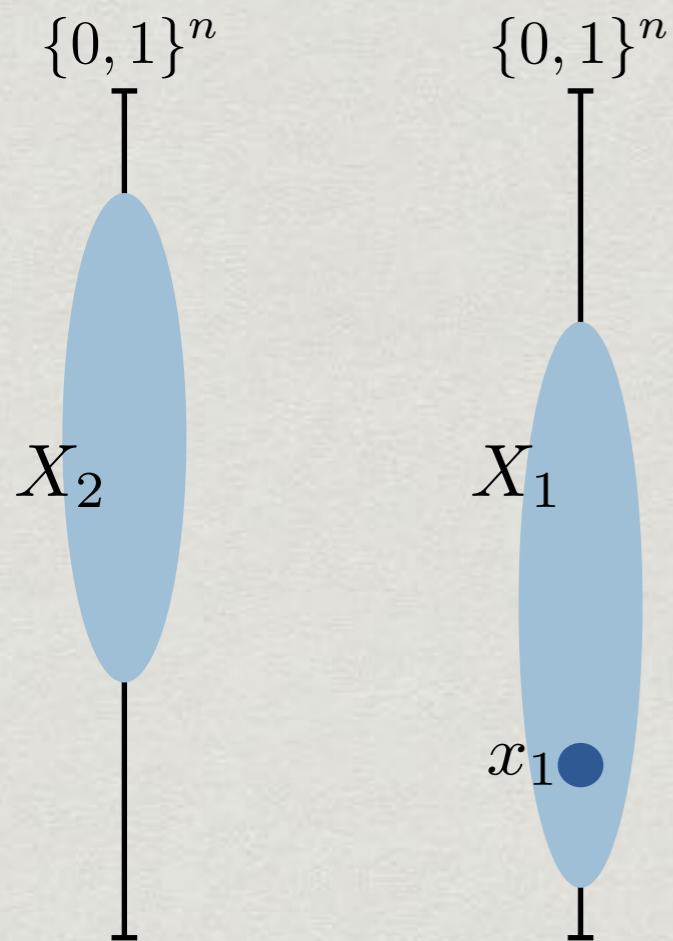
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- * All recent constructions of two-source extractors use non-malleable extractors as a central ingredient.
- * A bird's-eye view of these constructions: Given two inputs x_1 and x_2 ,
 - * Generate a table of $\text{nmE}(x_1, i)$ for all seeds $i \in \{0, 1\}^d$.
 - * Using x_2 , sample a subset of the rows.
 - * Apply a *resilient* function on the reduced table.

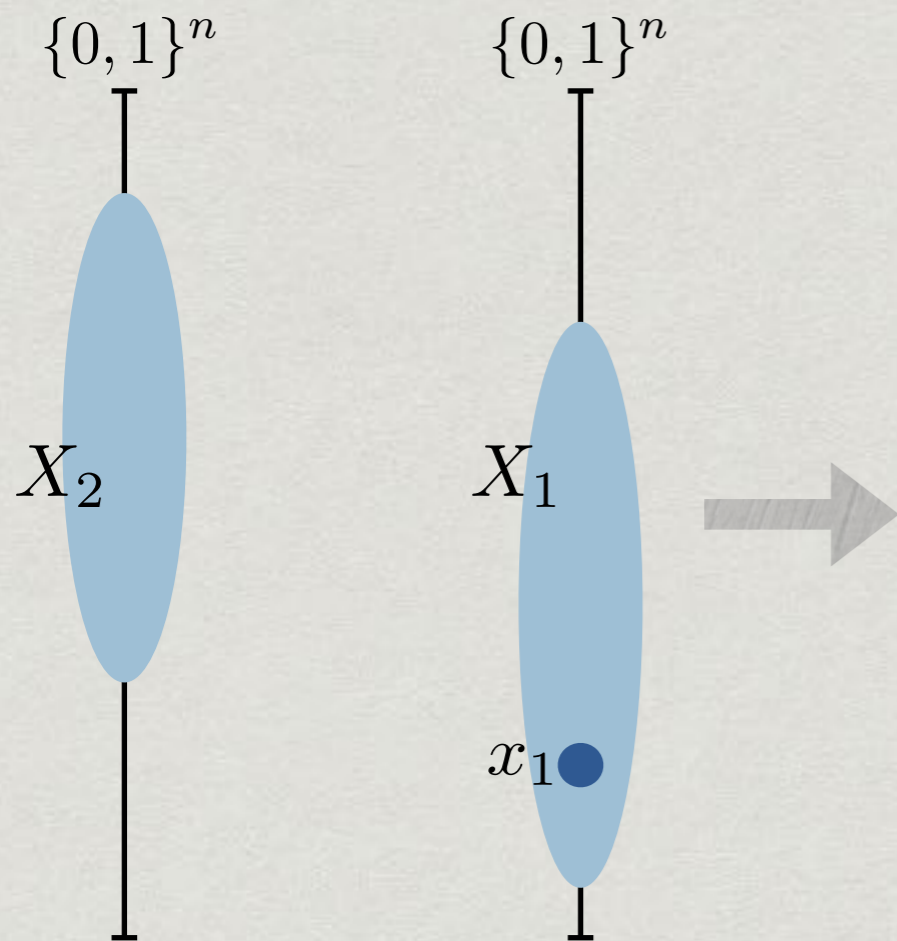
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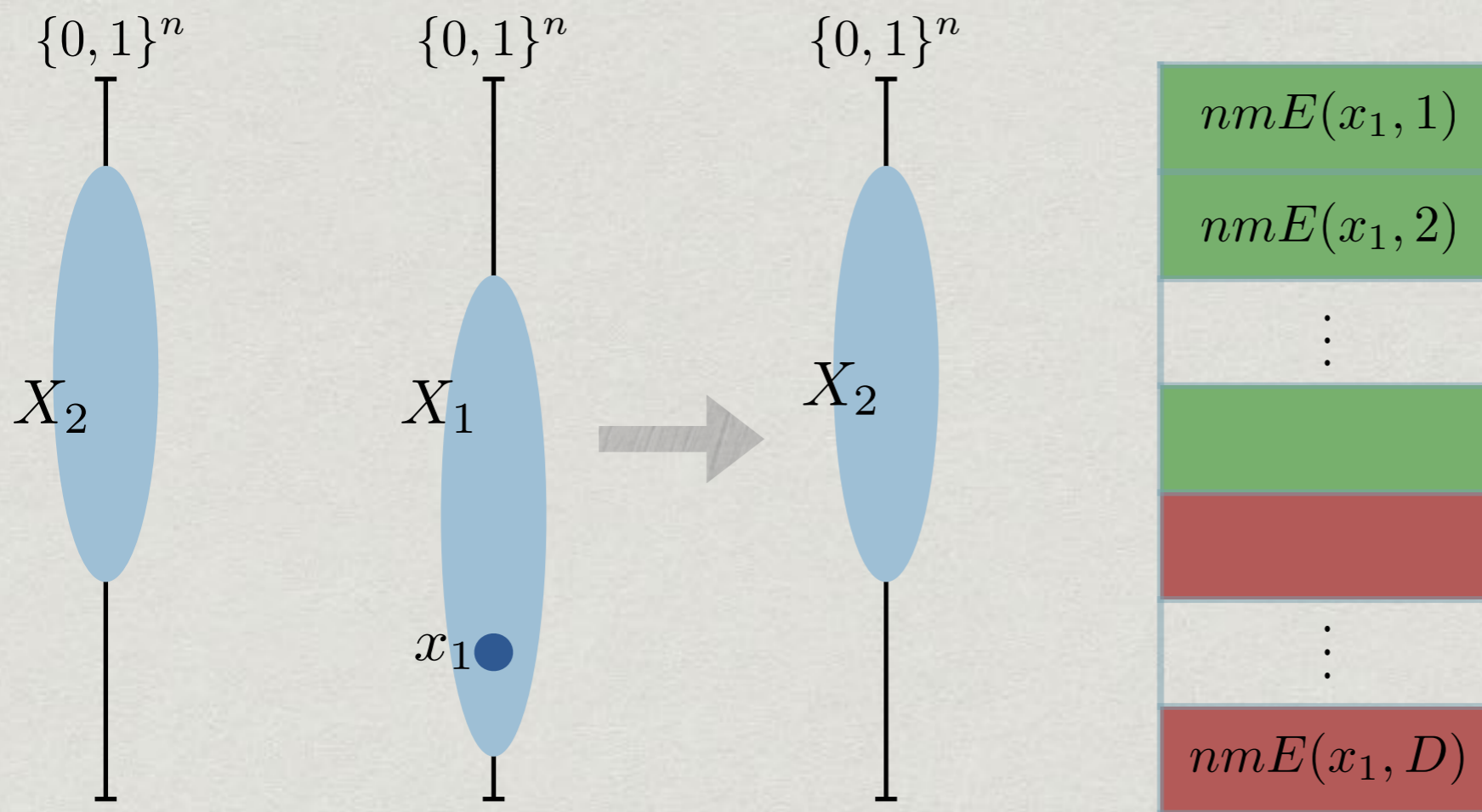
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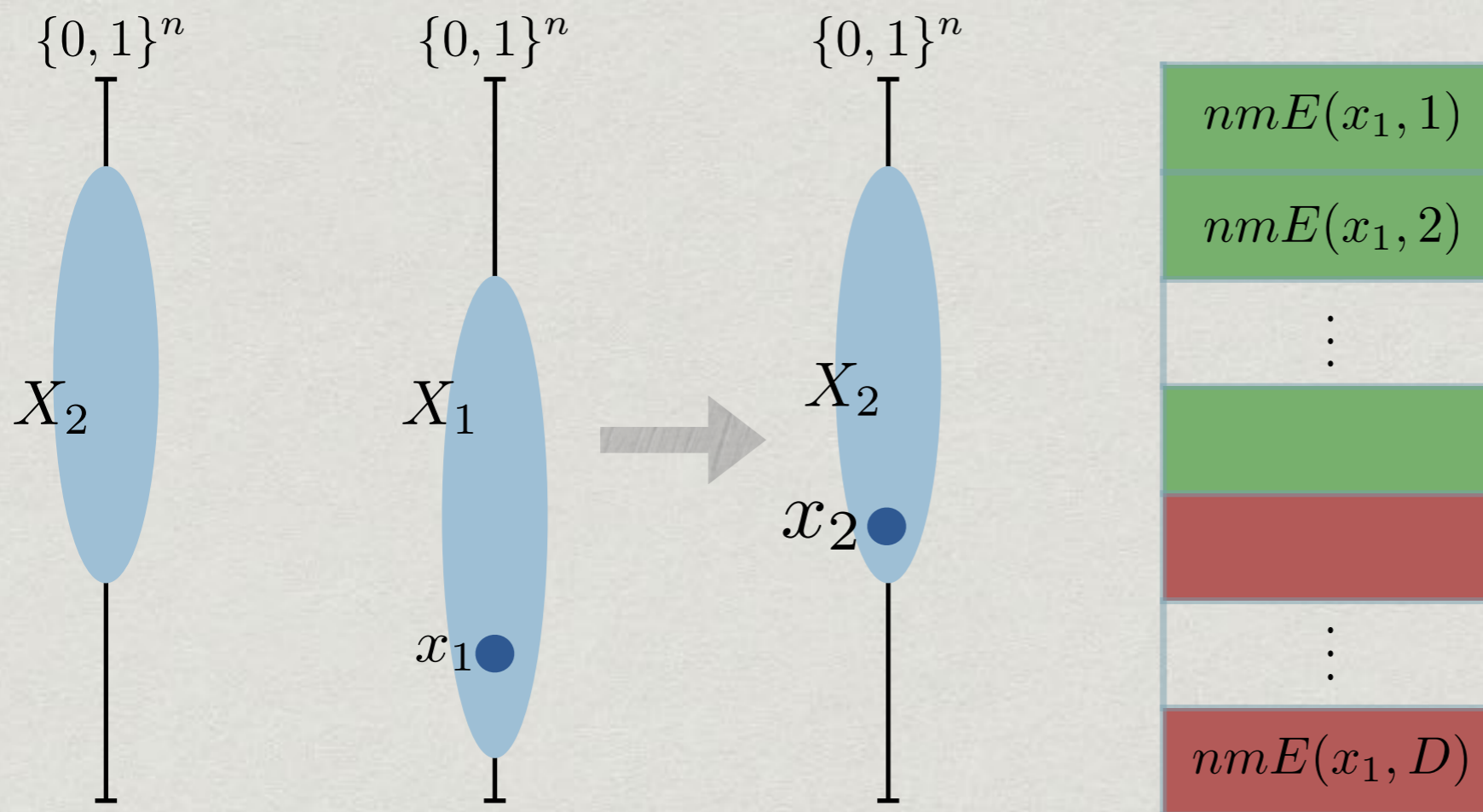
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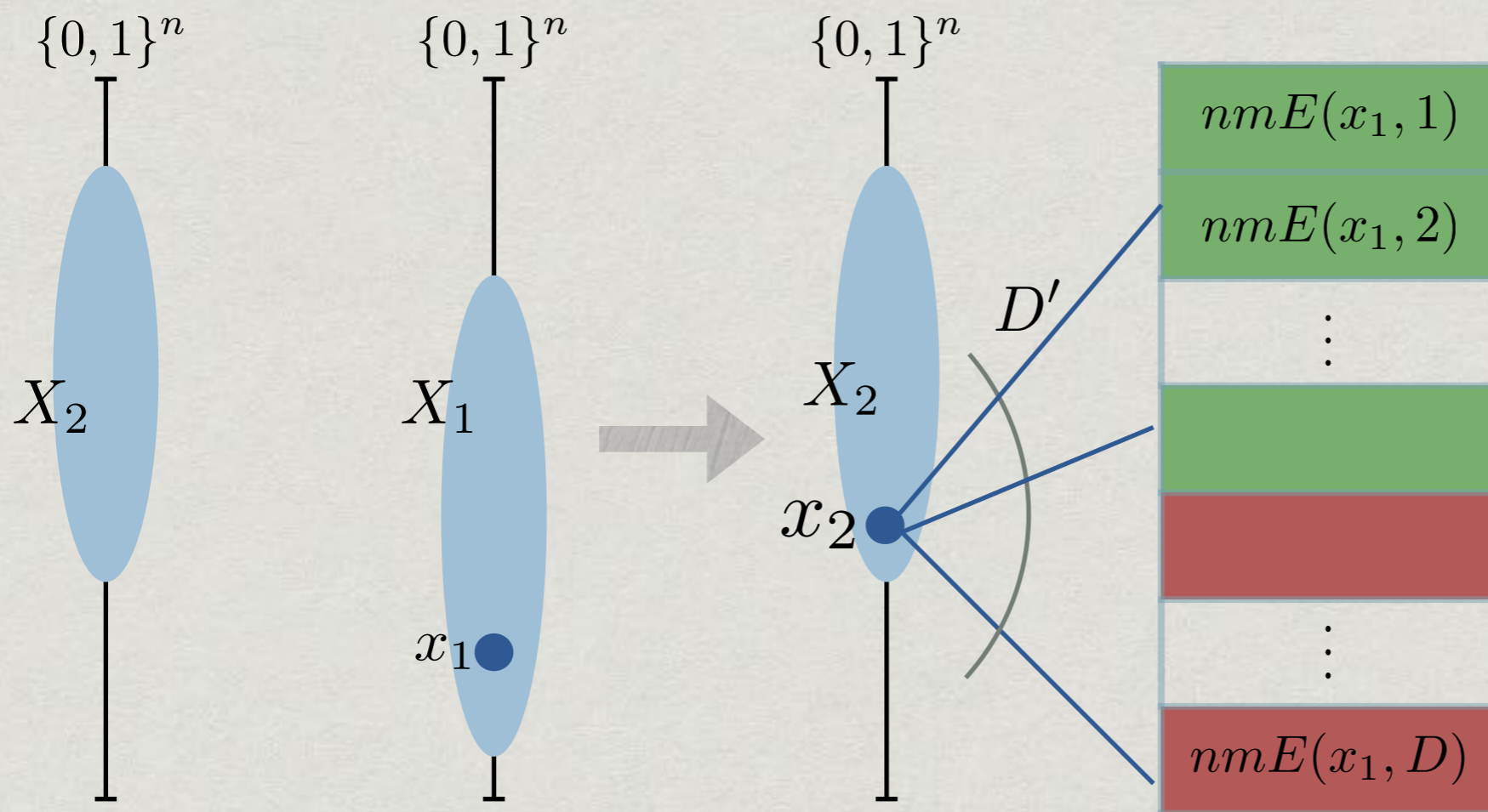
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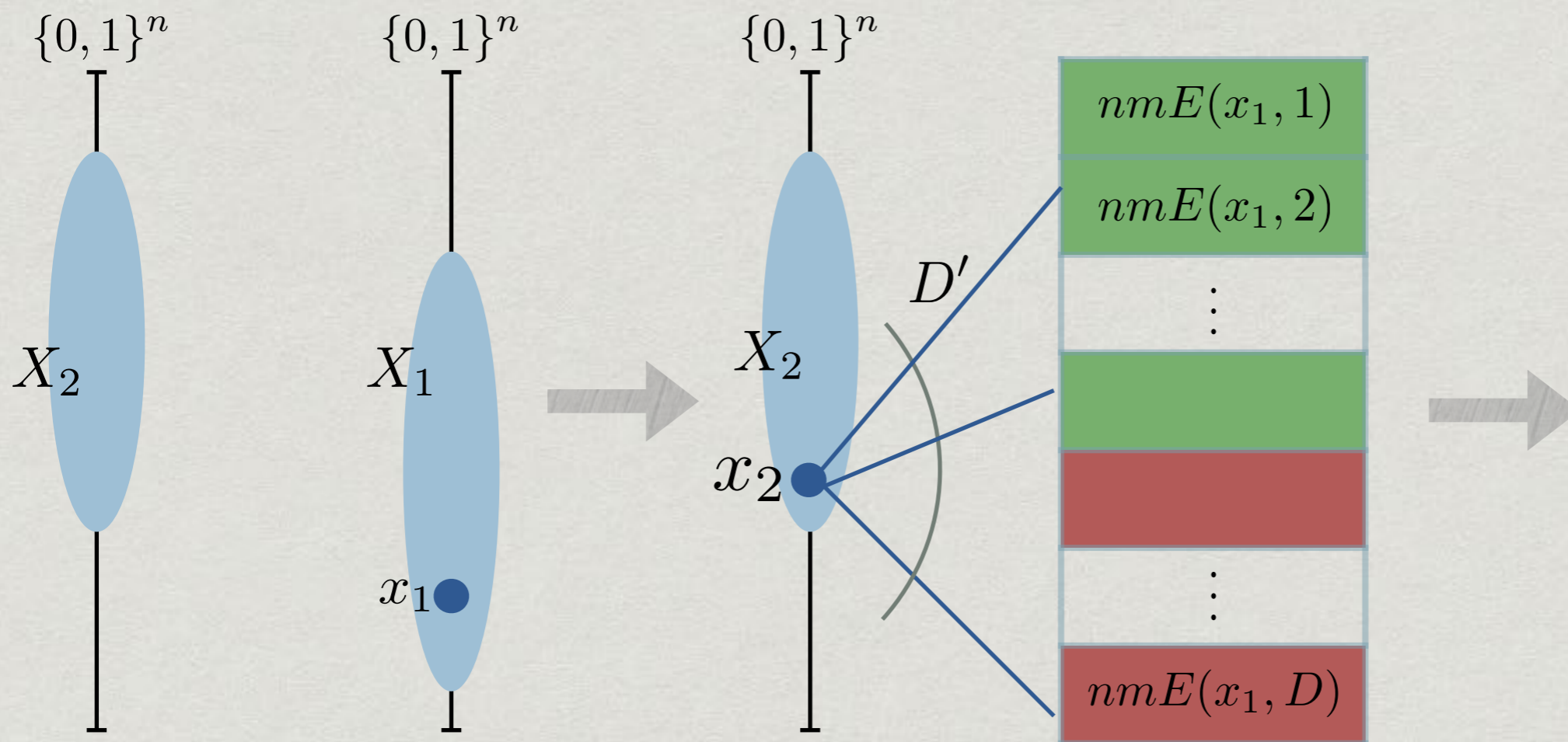
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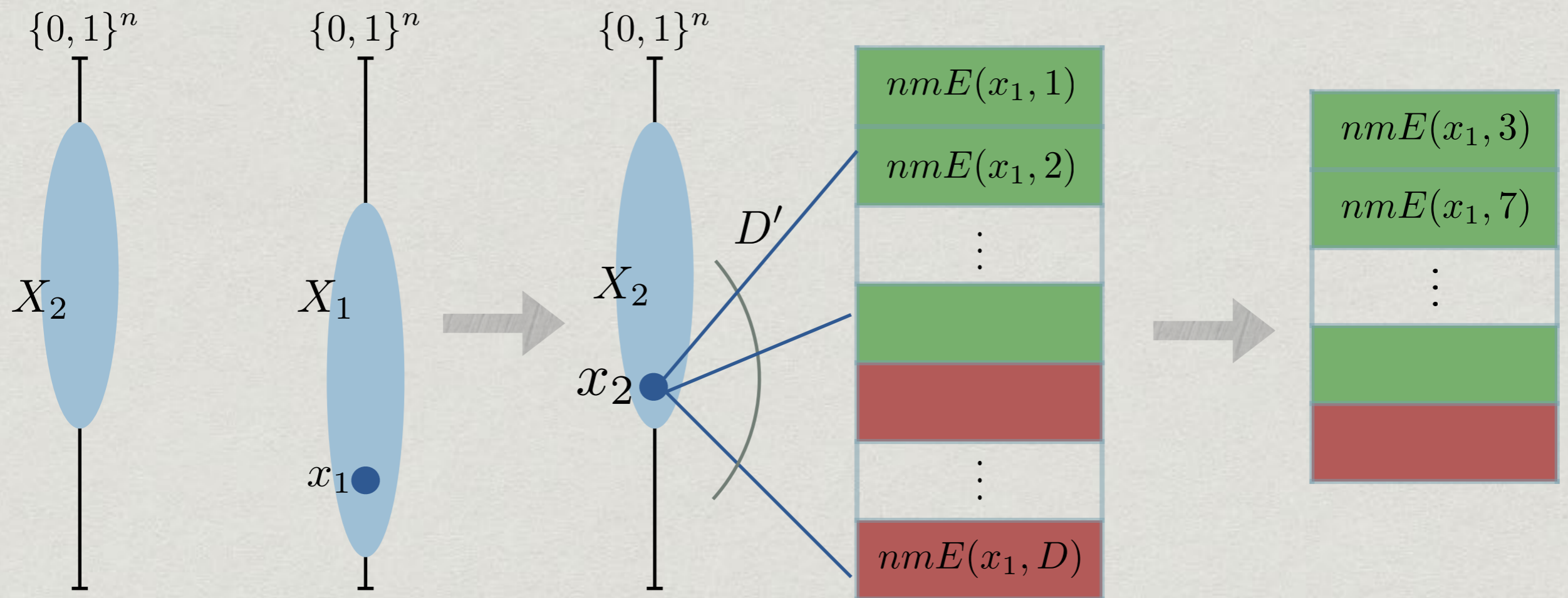
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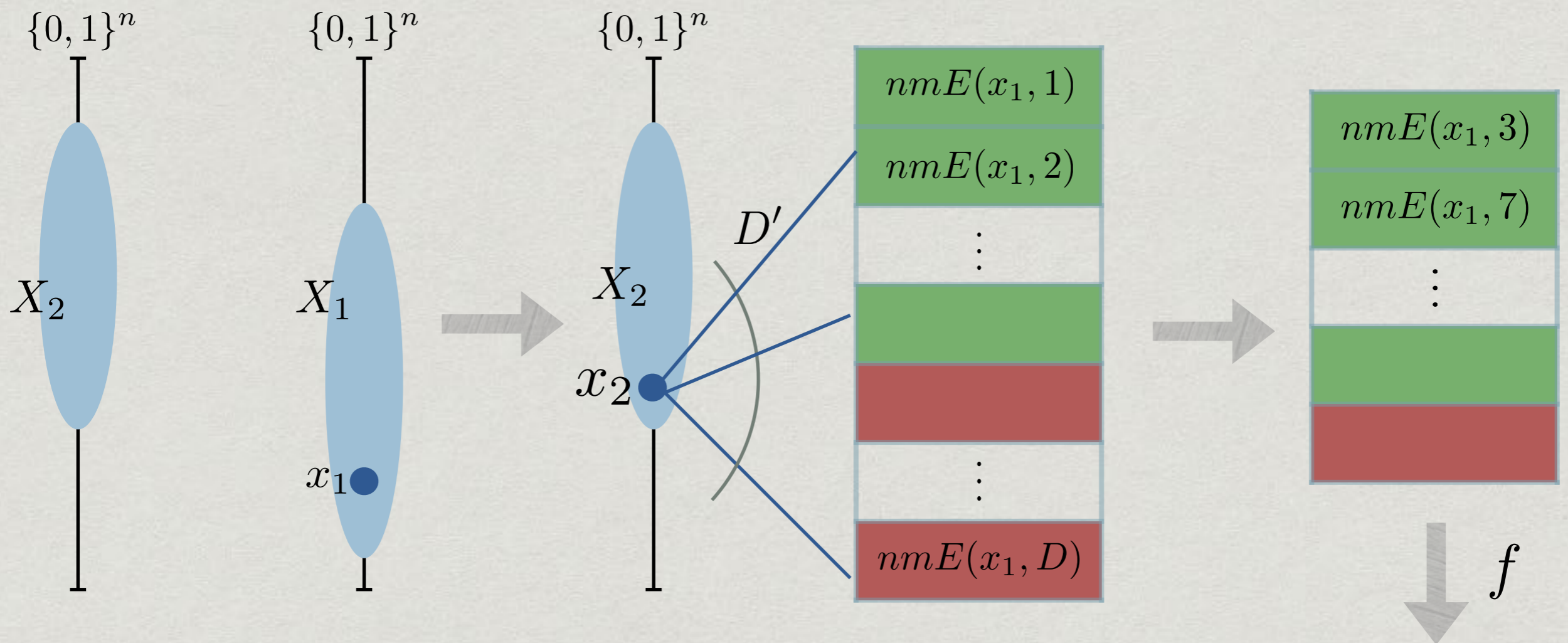
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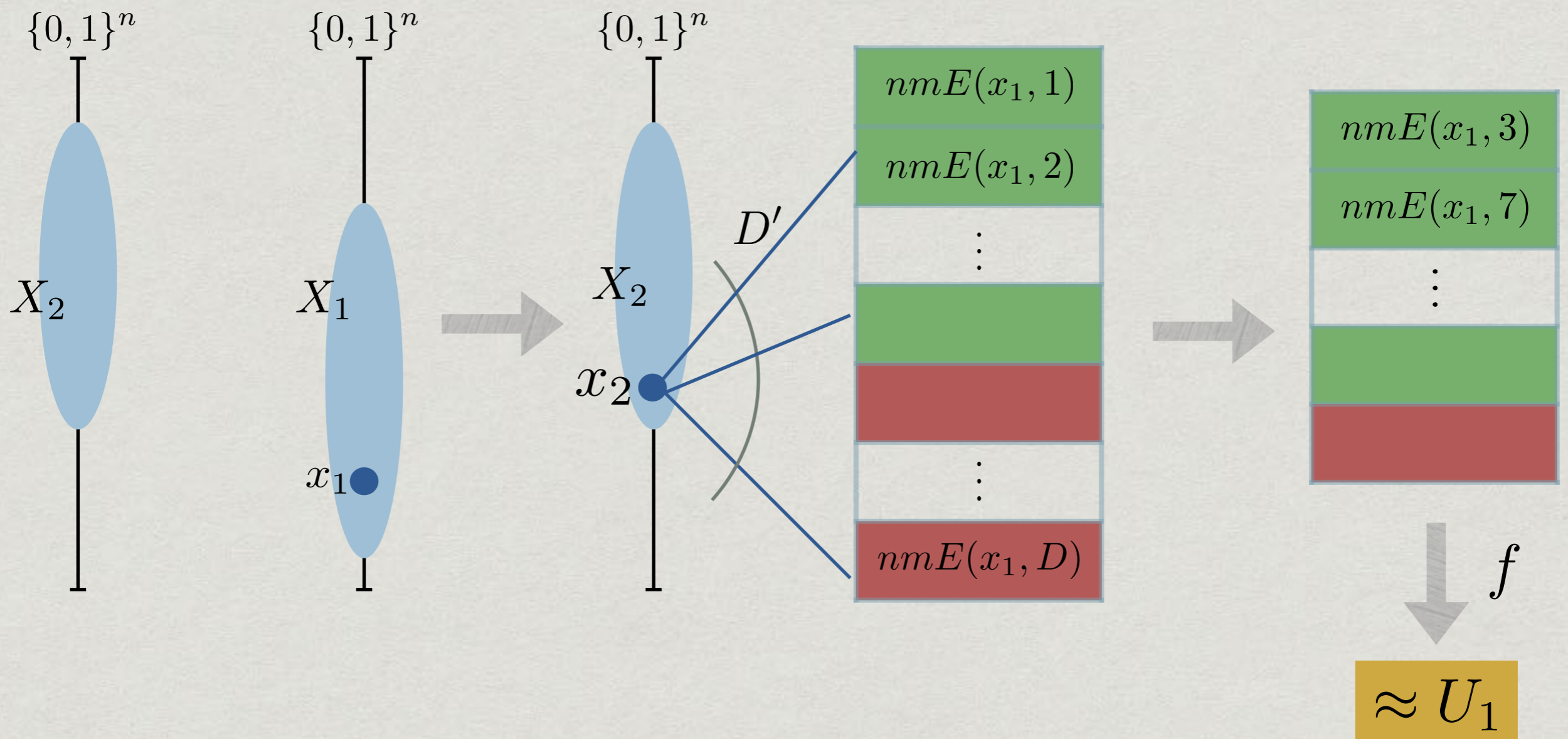
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- * We need f to be **resilient**:
 - * Say we have D' players. ε -fraction of them are malicious, and the rest are t -wise independent.
 - * The honest players draw their random bit and later the malicious players draw as they wish.
 - * With high probability, the outcome is not biased — the malicious players cannot substantially bias the outcome.

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- * A corollary of [KKL88] — even one malicious player can bias the output with probability at least $\log D'/D'$.
- * We cannot hope for an error smaller than $1/D'$, and D' is the size of our table.
- * Thus, the running time is at least $1/\epsilon$.

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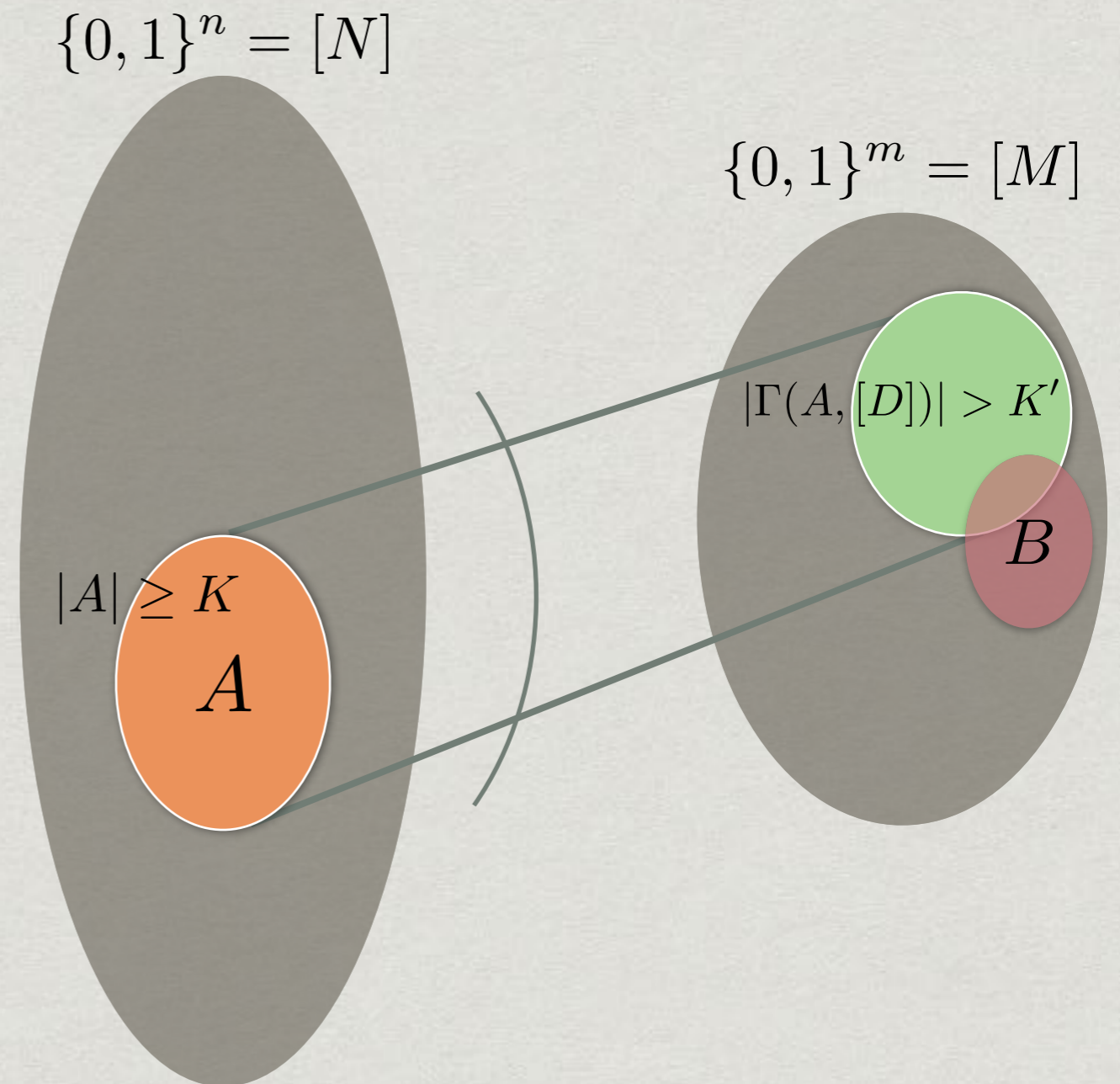
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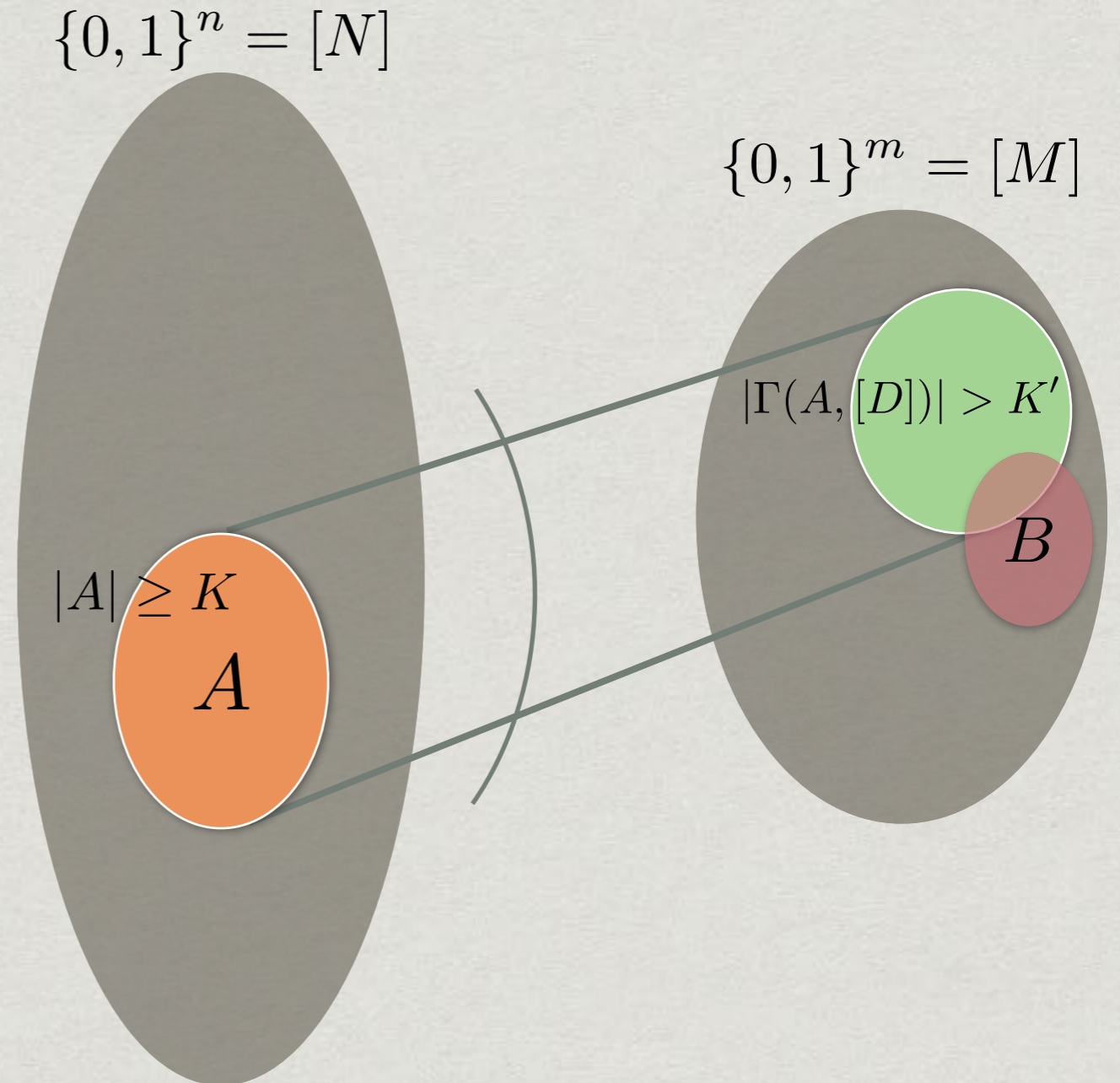
- * We should abandon resilient functions if we want to get a small error.
- * Instead of trying to sample and then employ t -wise in the good rows, let's just try and **hit** a good row.
- * As usual, we hit with a disperser...

Dispersers



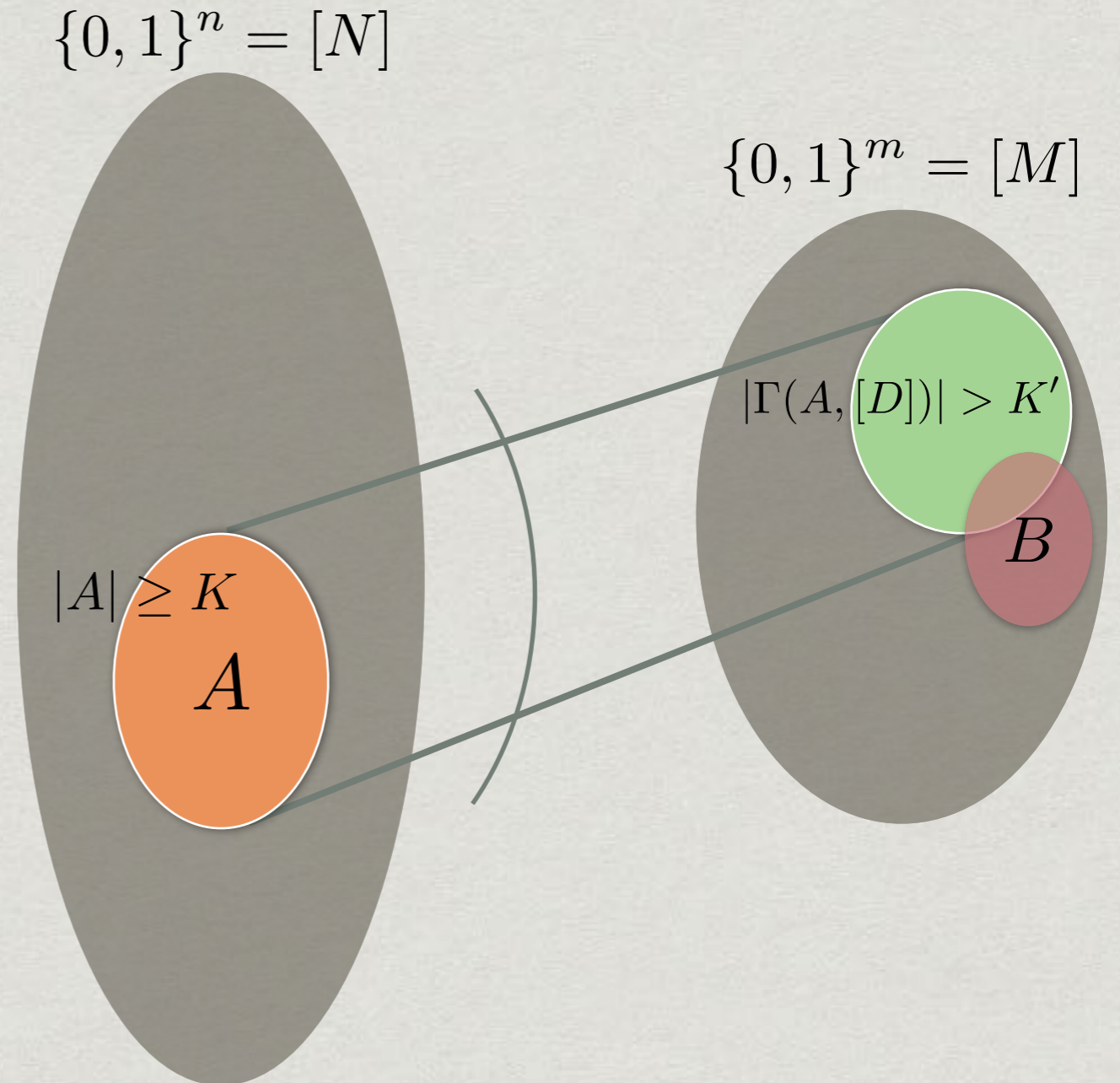
Dispersers

- * $\Gamma: \{0,1\}^n \times [D] \rightarrow \{0,1\}^m$ is a (K, K') -disperser if for every set A of cardinality at least K , Γ maps A to a set of cardinality greater than K' .



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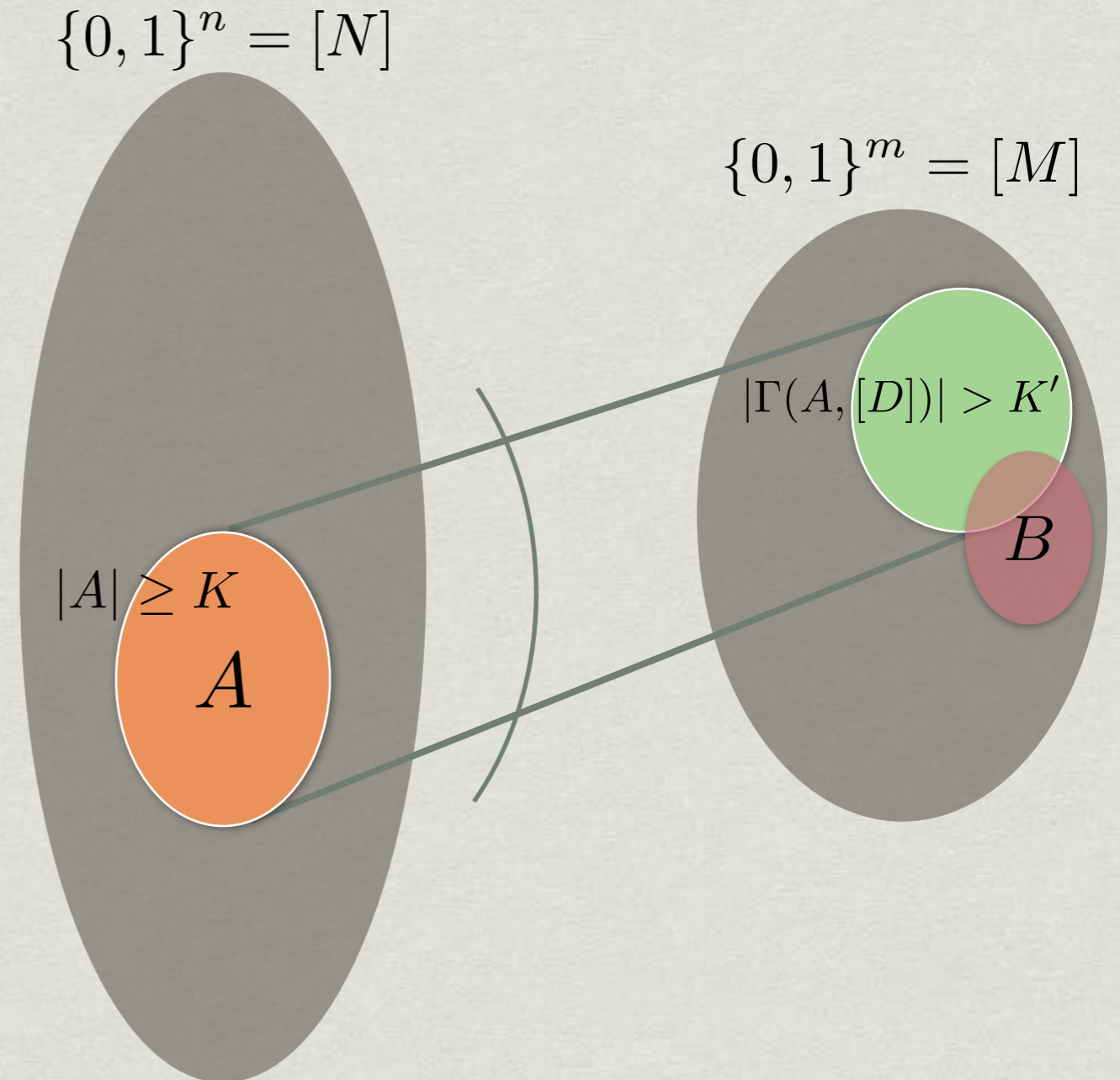
- * $\Gamma: \{0,1\}^n \times [D] \rightarrow \{0,1\}^m$ is a (K, K') -disperser if for every set A of cardinality at least K , Γ maps A to a set of cardinality greater than K' .
- * We are interested in the case where K' is small compared to 2^m . That is, we want to avoid **small** bad sets.



Dispersers

- * Used to reduce error in one-sided probabilistic algorithms.
- * [RT]: When K' is not too large, say $K' = \epsilon M$, the lower bound on the degree is

$$D = \Omega \left(\frac{\log \frac{N}{K}}{\log \frac{1}{\epsilon}} \right)$$



Zuckerman's disperser

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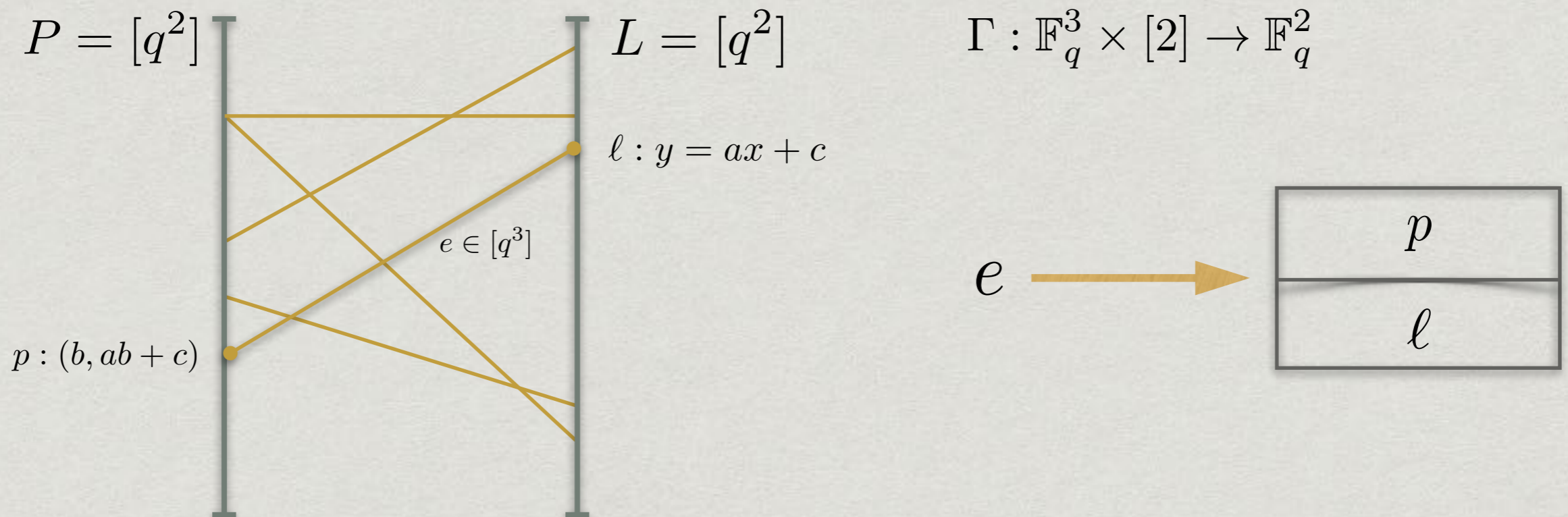
- * Quite amazingly, when $K=N^\delta$ for a constant δ , there exist explicit constructions that achieve this bound [BKSSW05,Raz05,Zuck06].

Zuckerman's disperser

- * Quite amazingly, when $K=N^\delta$ for a constant δ , there exist explicit constructions that achieve this bound [BKSSW05,Raz05,Zuck06].
- * The key ingredient in Zuckerman's construction: A points-lines incidence graph.

Zuckerman's disperser

The input source is distributed, over $[q]^3$, among the edges of the graph.



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Zuckerman's disperser

- * This gives a degree-2 disperser, and we can recurse.
- * For $K=N^\delta$, where δ is arbitrary, the dependence is

$$D = (1/\delta)^{O(1)} \frac{n}{\log \frac{1}{\epsilon}}$$

- * Also, the output length is determined by the number of recursion steps, and we have $m=\delta^{O(1)}n$.

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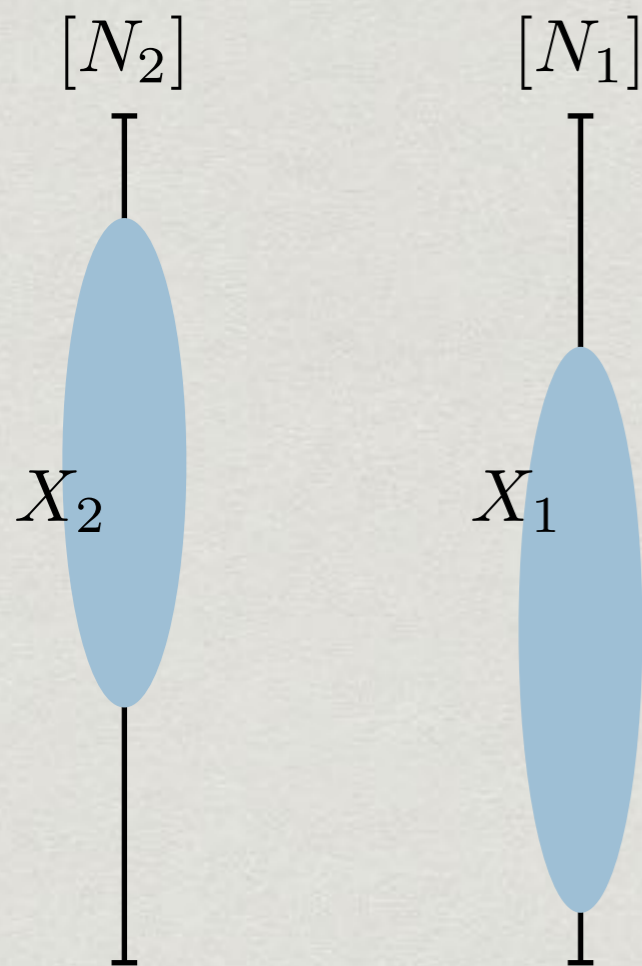
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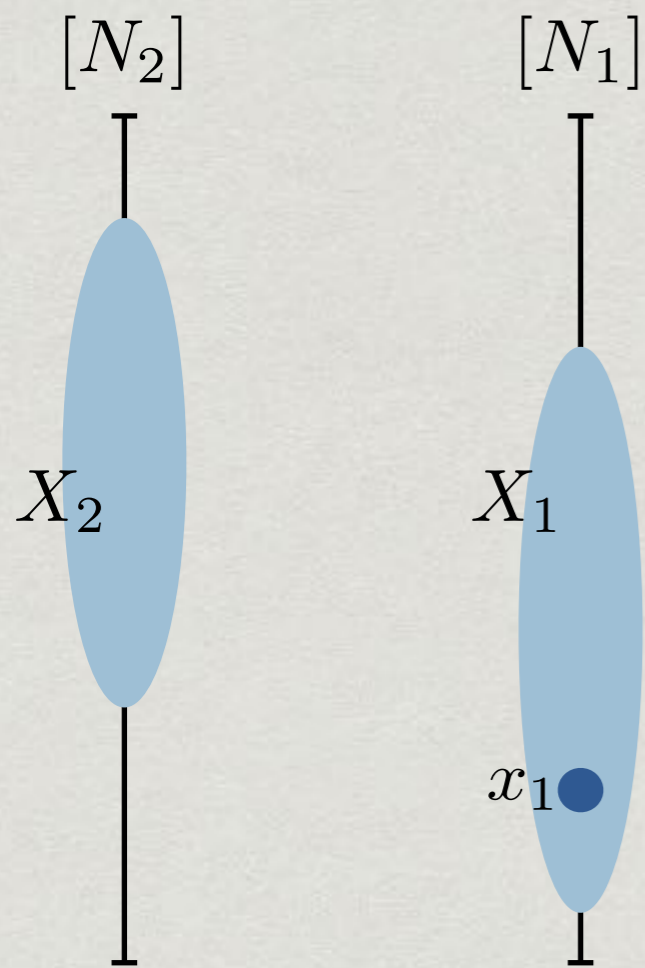
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 - * $\Gamma: [N_2] \times [t+1] \rightarrow [D]$, a $(\varepsilon k_2, \varepsilon D)$ -disperser.
- * On input x_1, x_2 , output $\bigoplus_{i \in [t+1]} \text{nmE}(x_1, \Gamma(x_2, i))$.

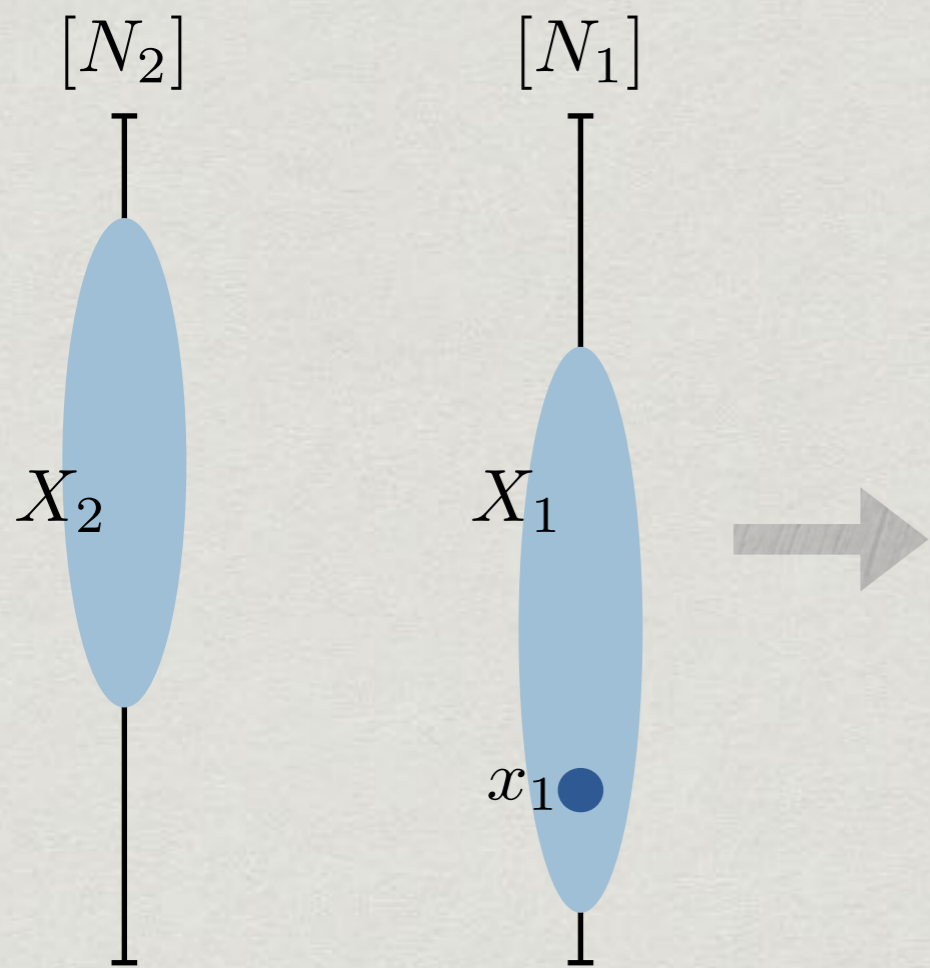
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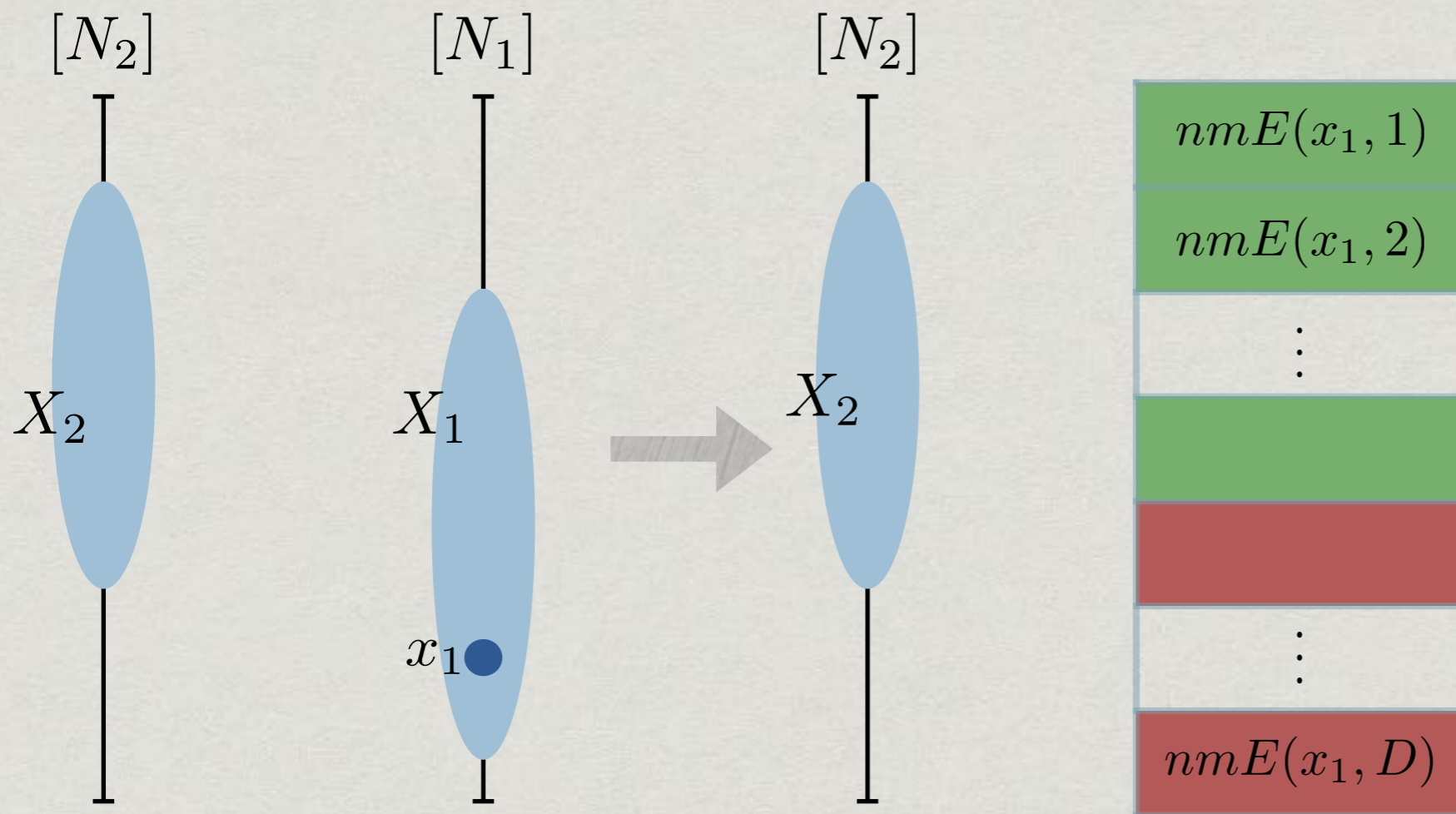
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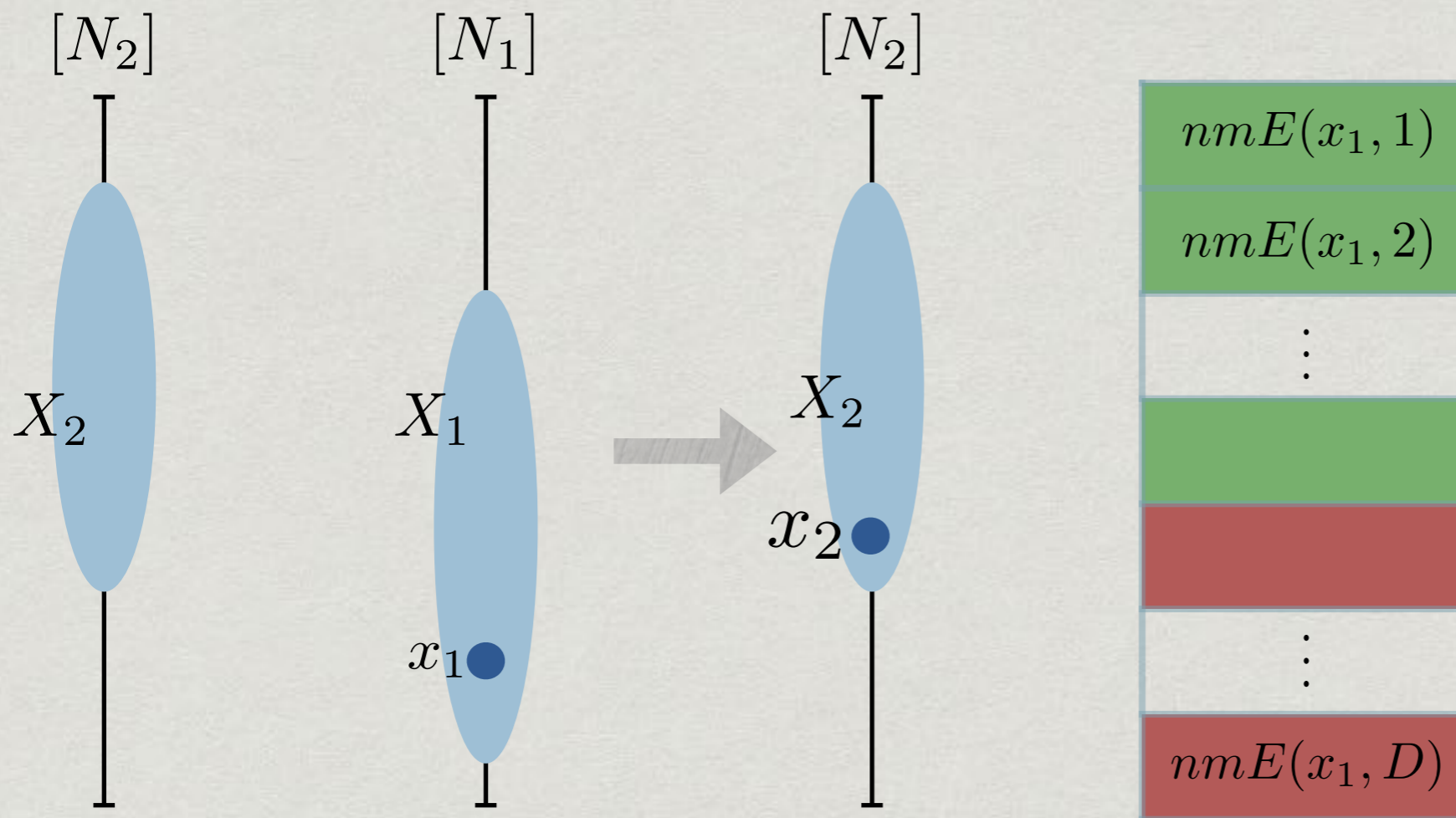
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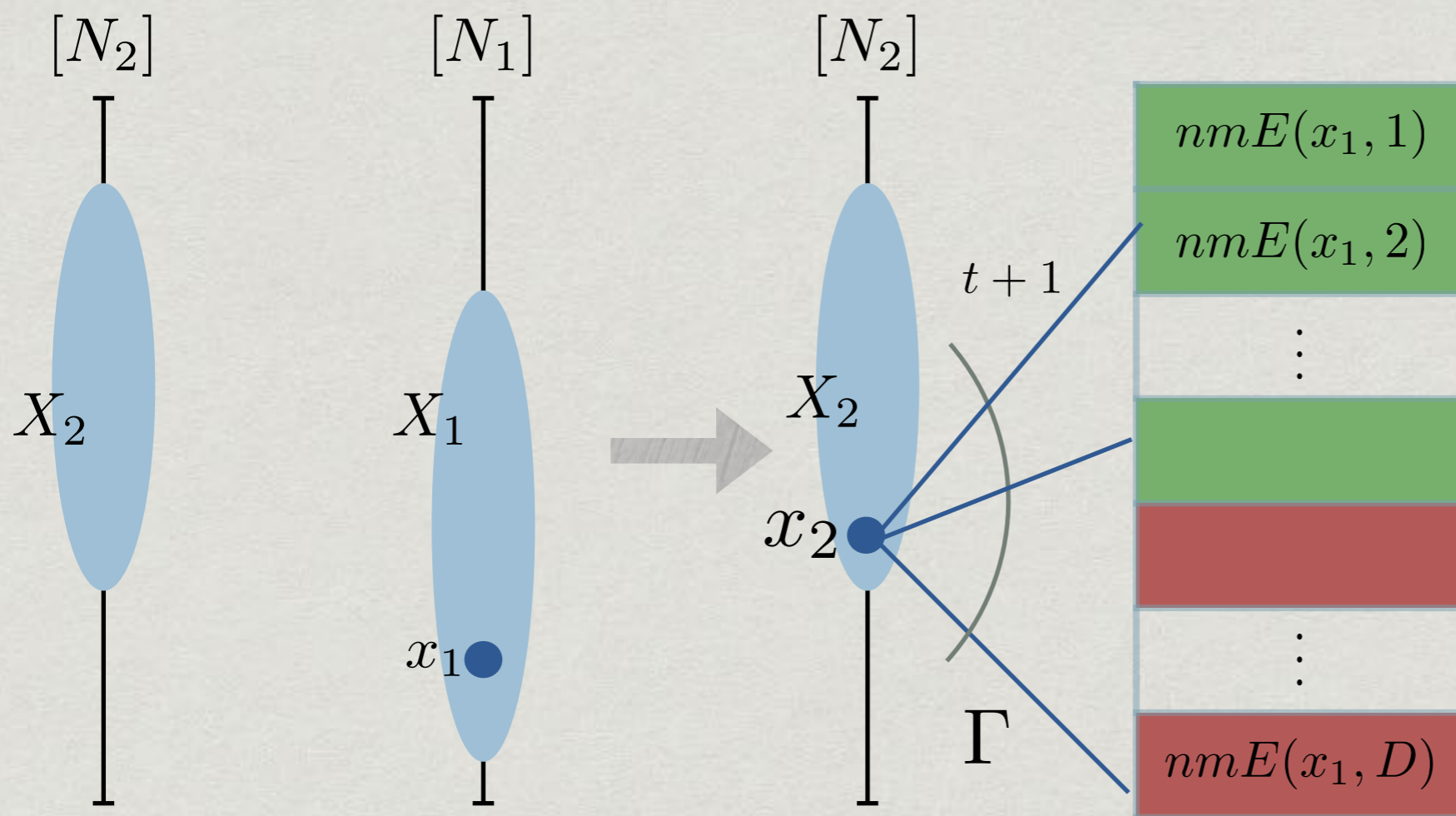
Our reduction



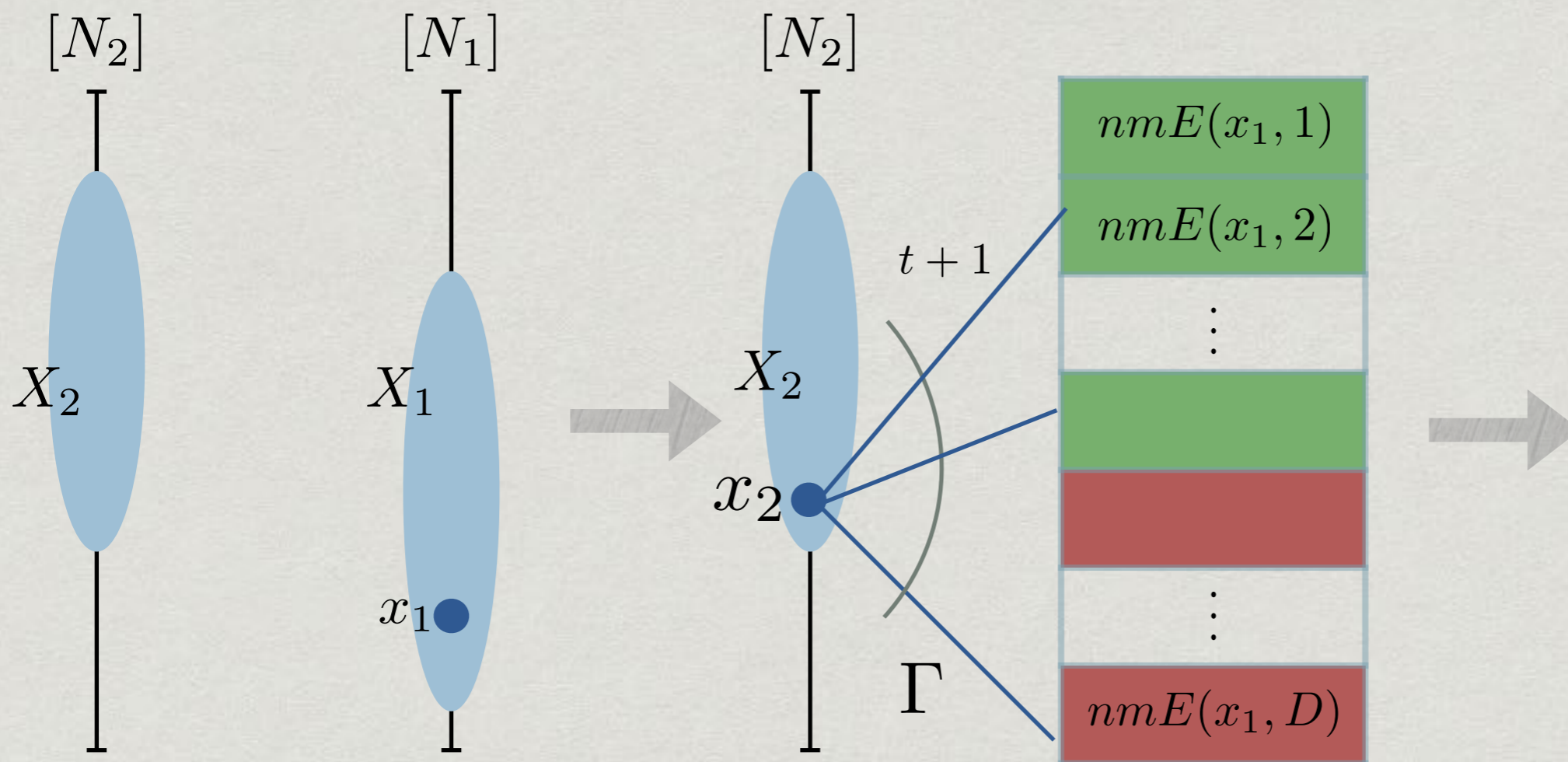
Our reduction



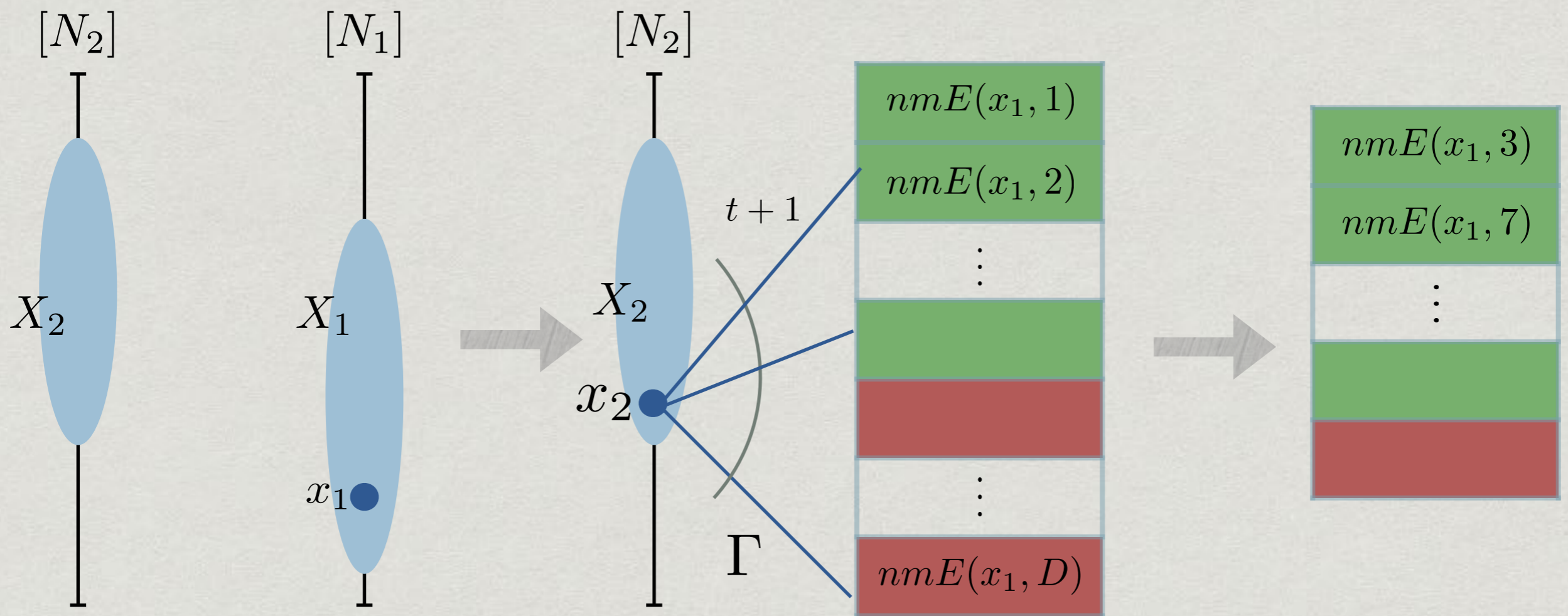
Our reduction



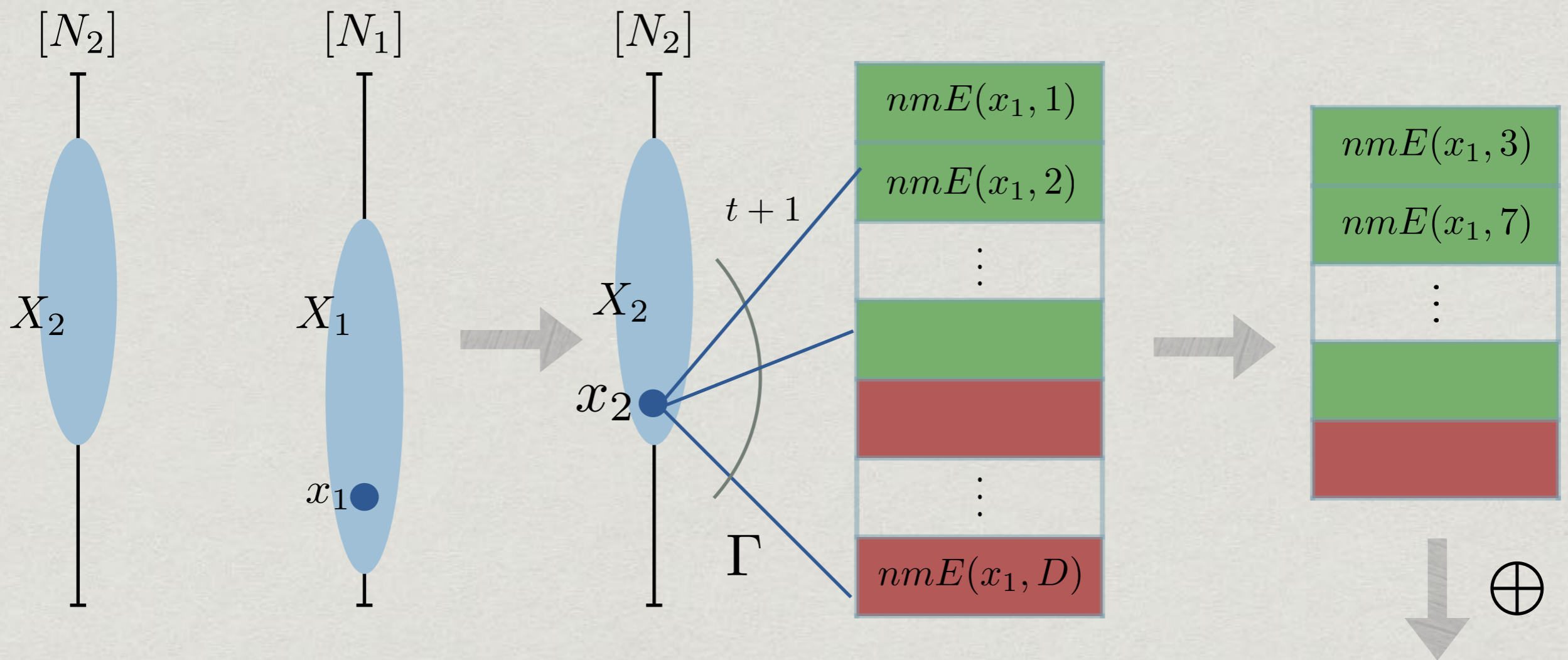
Our reduction



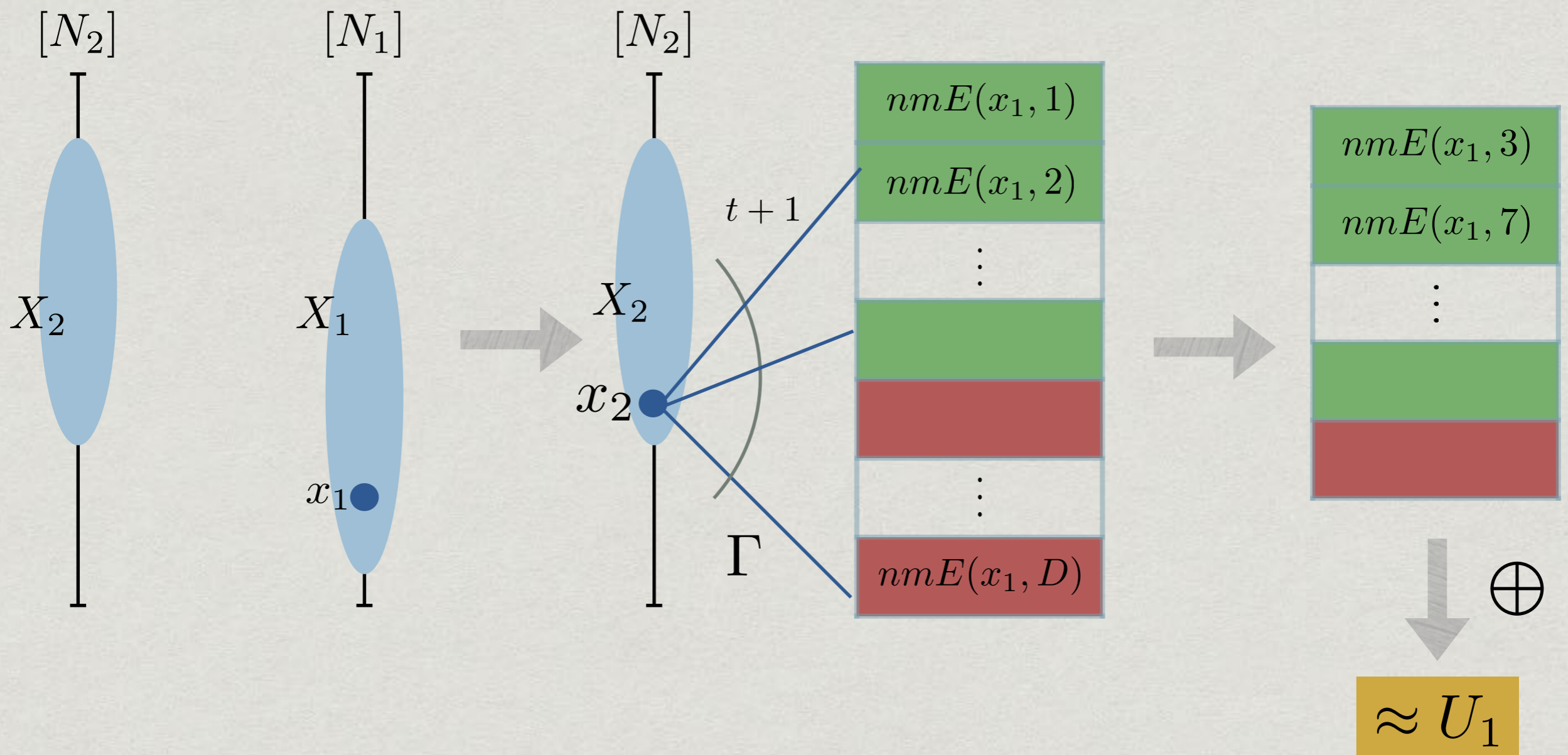
Our reduction



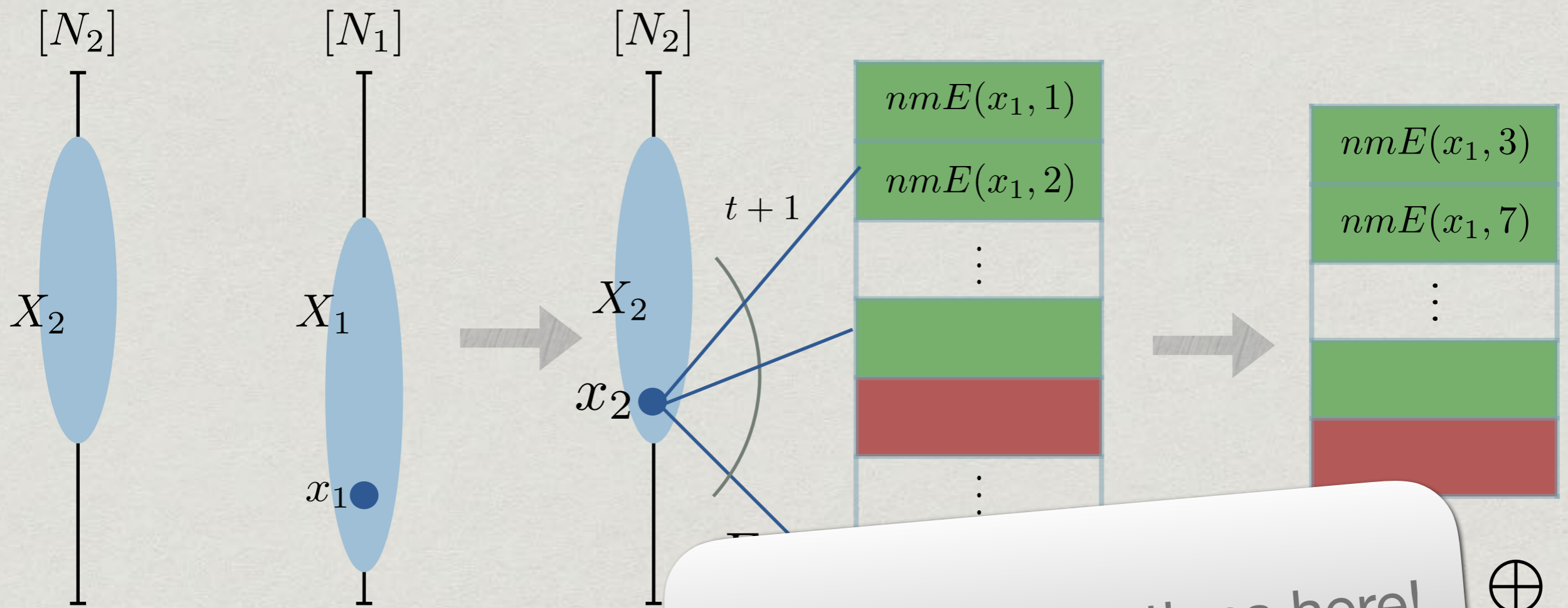
Our reduction



Our reduction



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No resilient functions here!

\oplus

J_1

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- * By the properties of Γ , the number of elements x_2 for which $\Gamma(x_2, [t+1])$ contains only **bad** seeds is at most εK_2 .
- * Thus, with probability at least $1-\varepsilon K_2/K_2=1-\varepsilon$, the input x_2 samples $t+1$ seeds of nmE , one of which, **y**, is **good**.

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- * Hence, the parity of these random variables is also close to uniform, and the overall error is 2ε .

Our reduction

- * So, if the n.m. extractor can support small error (and existing constructions can), we get a construction with a small error.

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- * The parity is not resilient... What happened here?
 - * Instead of sampling (with a good sampler) D' rows from the table and applying a resilient function, we pick a drastically smaller sample set — of size $t+1$.
 - * Instead of requiring that the number of malicious players is small, we have the weaker requirement that not *all* of the players in our sample set are malicious.

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But does it work?

- * Or, *when* does it work? We have no option but to look closer into the parameters.
- * First, note that the disperser dictates n_2 , the length of the second source, and typically it is smaller than n_1 .
- * A potential circular hazard — the degree of Γ should be at least $t+1$, but the degree of Γ also depends on the seed length of the n.m. extractor, which in turn depends on t ...

But does it work?

- * Let's check this circularity on the board...

Our result

- * We see that the seed length of the n.m. extractor plays a crucial role. Say there exists an explicit n.m. extractor with seed length d and supports entropy k_1 . Our results:
 - * If $d = ct \log(n_1/\epsilon)$ for a small enough constant c , there exists an explicit two-source extractor with small error for entropies k_1 and $k_2 = an_2$ (for every constant a).

Our result

- * We see that the seed length of the n.m. extractor plays a crucial role. Say there exists an explicit n.m. extractor with seed length d and supports entropy k_1 . Our results:
 - * If $d = t^\gamma \log(n_1/\epsilon)$ for a small enough constant γ , there exists an explicit two-source extractor with small error for entropies k_1 and $k_2 = n_2^\beta$ for some constant β .

Our result

- * We see that the seed length of the n.m. extractor plays a crucial role. Say there exists an explicit n.m. extractor with seed length d and supports entropy k_1 . Our results:
 - * If $d = \log(n_1/\varepsilon) + O(\mathbf{t})$, there exists an explicit two-source extractor with small error for entropies k_1 and $k_2 = n_2^\beta$ for every constant β .

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- * Non-explicitly, our constraints on d are easily satisfied. The seed length of a probabilistic construction is **$d=2\log(n/\varepsilon)+O(\log t)$** .

Good n.m. extractors

- * Non-explicitly, our constraints on d are easily satisfied. The seed length of a probabilistic construction is $d=2\log(n/\varepsilon)+O(\log t)$.
- * Taking a closer look on recent constructions of non-malleable extractors, we see that $d=\Omega(k)$ and $k=\tilde{O}(t^2\log(n/\varepsilon))$.
- * Very roughly, this coupling between d and k is inherent when you do alternating extraction.

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Good n.m. extractors

- * To summarize...
- * Due to [CZ15,BDT16] we know that n.m. extractors with short seed length supporting small entropies give rise to good two-source extractors with constant error.
- * This work: N.m. extractors also give rise to two-source extractors with small error, as long as the seed-length's dependency on t is good.

Good n.m. extractors

- * The moral: Keep constructing non-malleable extractors, with techniques that go beyond alternating extraction.

Thanks for listening.