

Foundations of Differentially Oblivious Algorithms

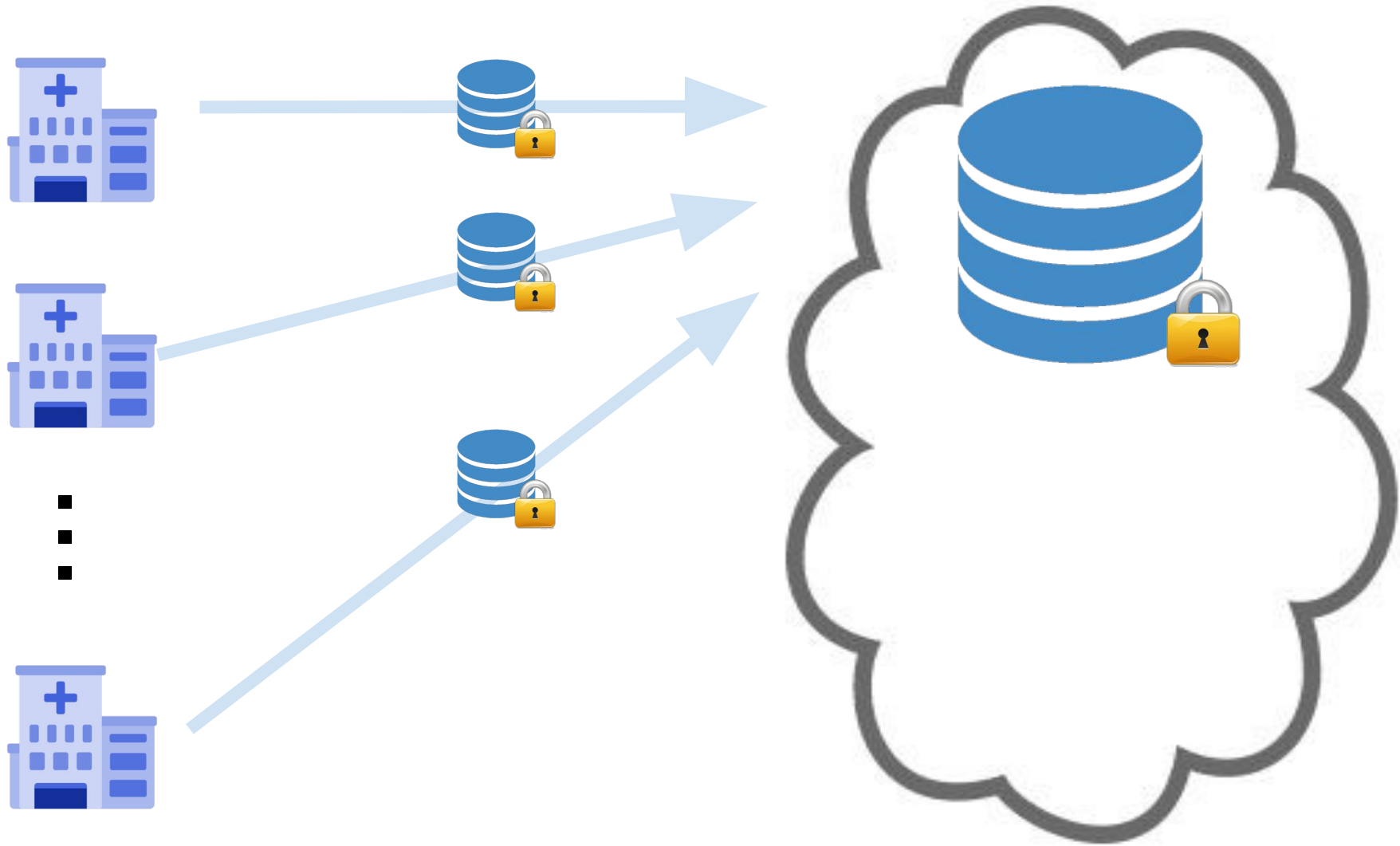
Elaine Shi

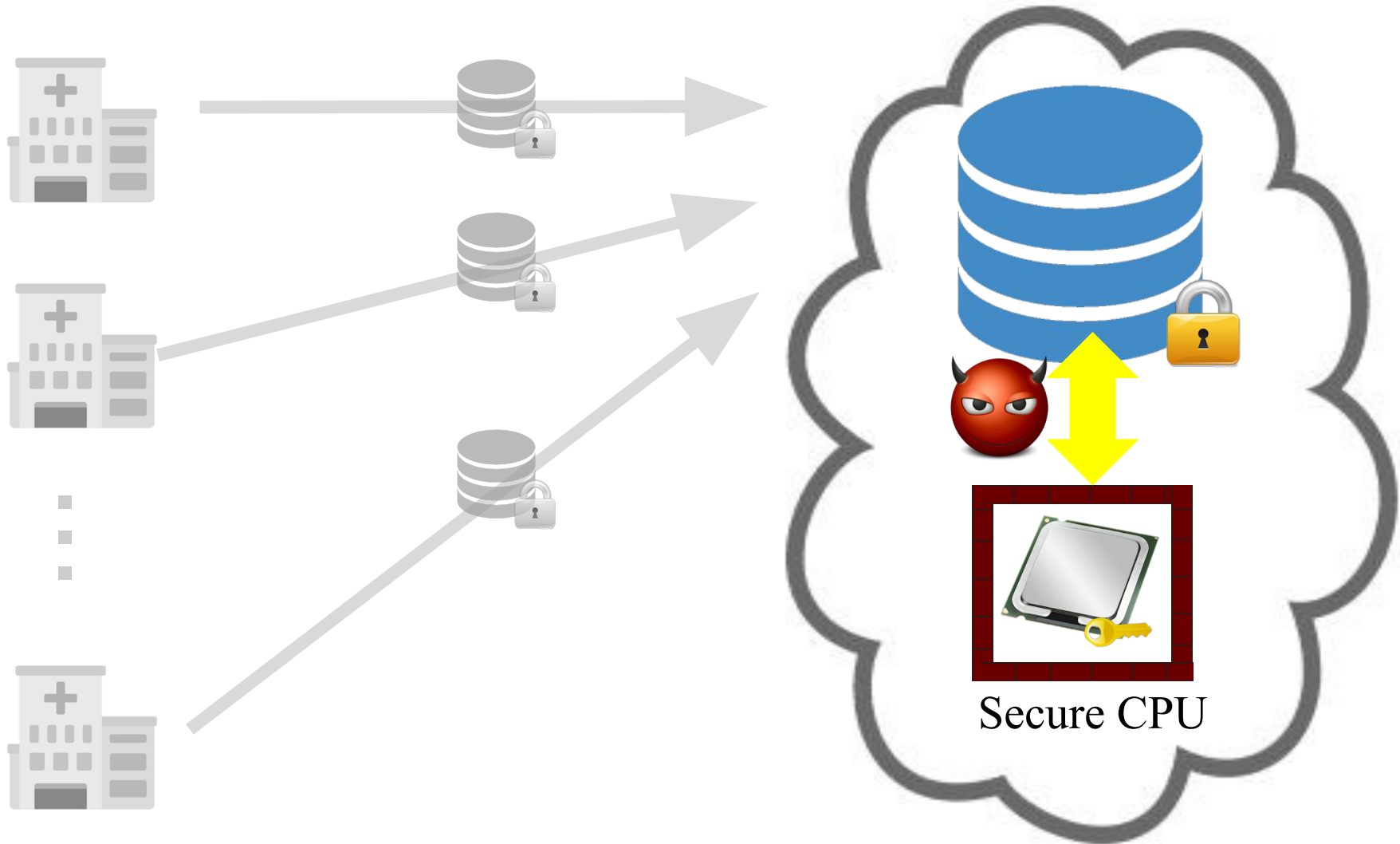
Based on [CCMS'18] and [LSX'18]



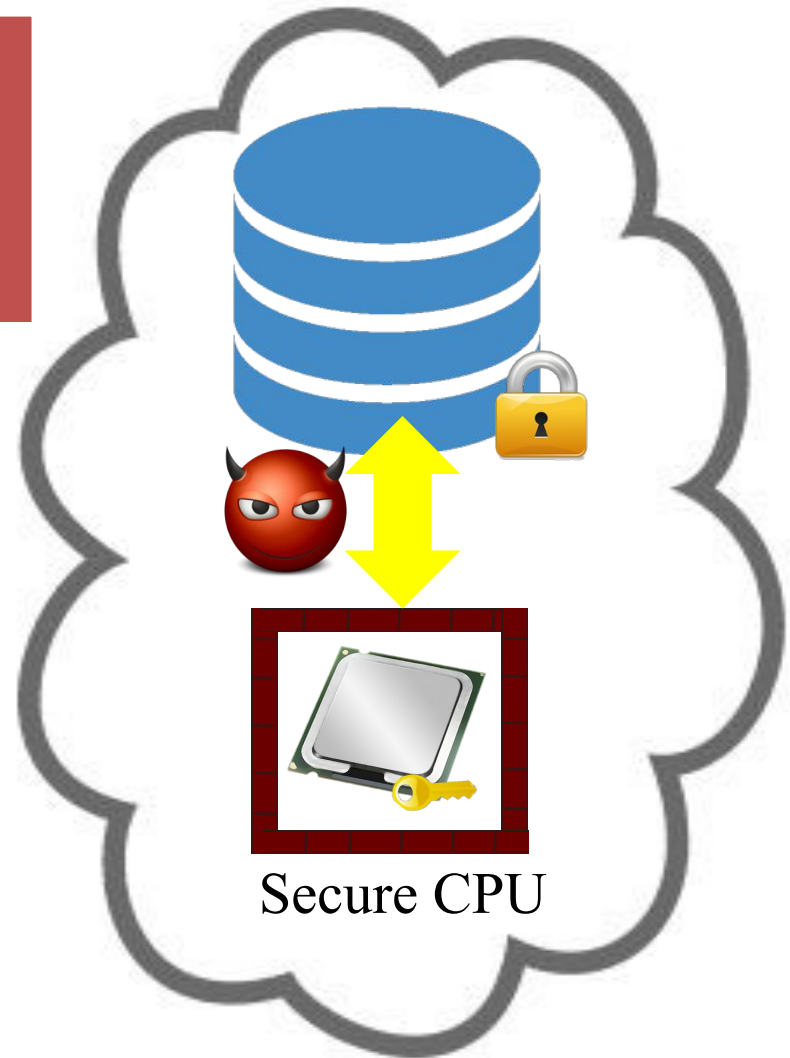
⋮







Access patterns to even encrypted data leak sensitive information.

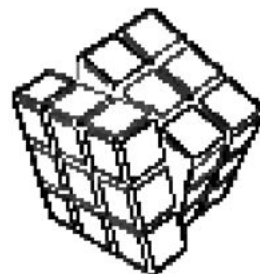


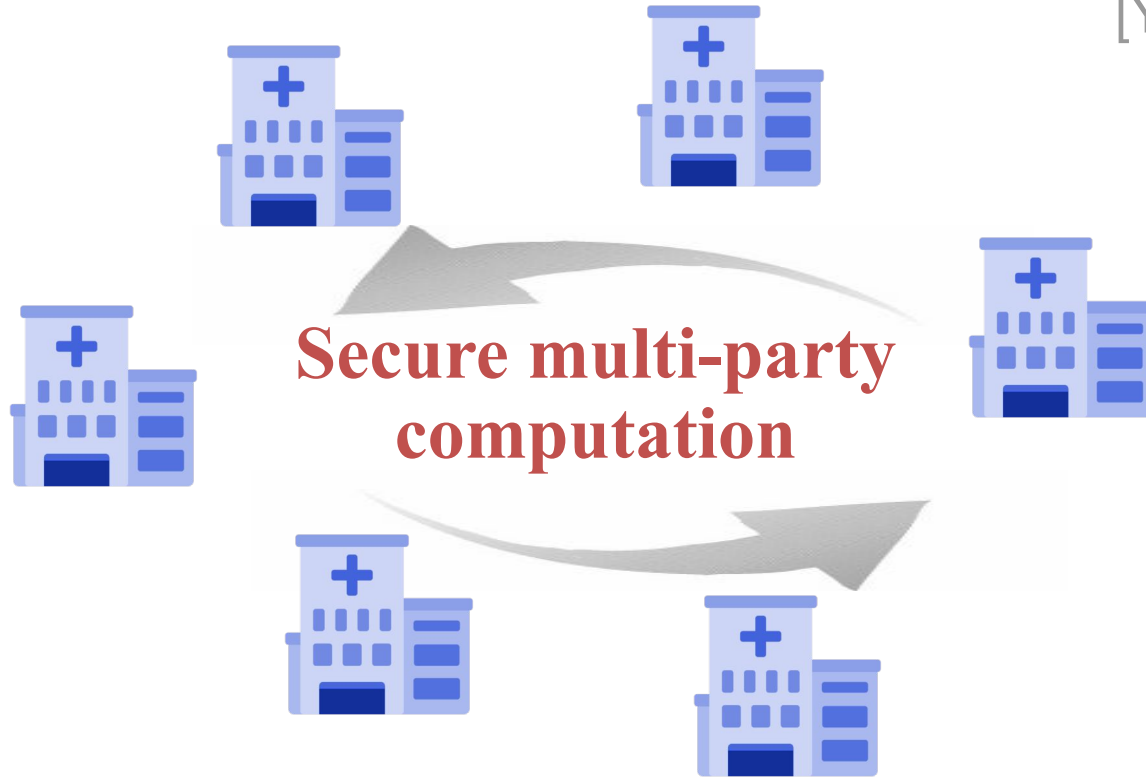
Access Pattern Attack: Computing on JPEG Image

Original



Recovered





Access Pattern Leakage in MPC

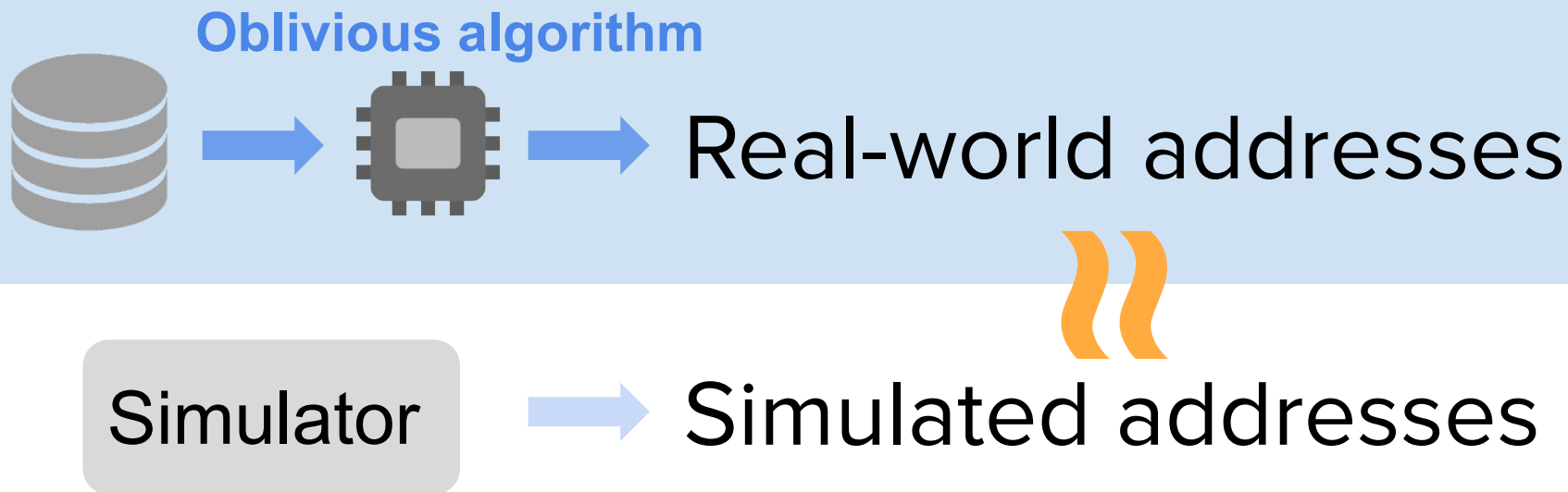


Oblivious RAM

An algorithmic approach that
provably obfuscates access patterns



Oblivious RAM





Oblivious RAM

“Encrypting the access patterns”



Oblivious RAM

“Encrypting the access patterns”

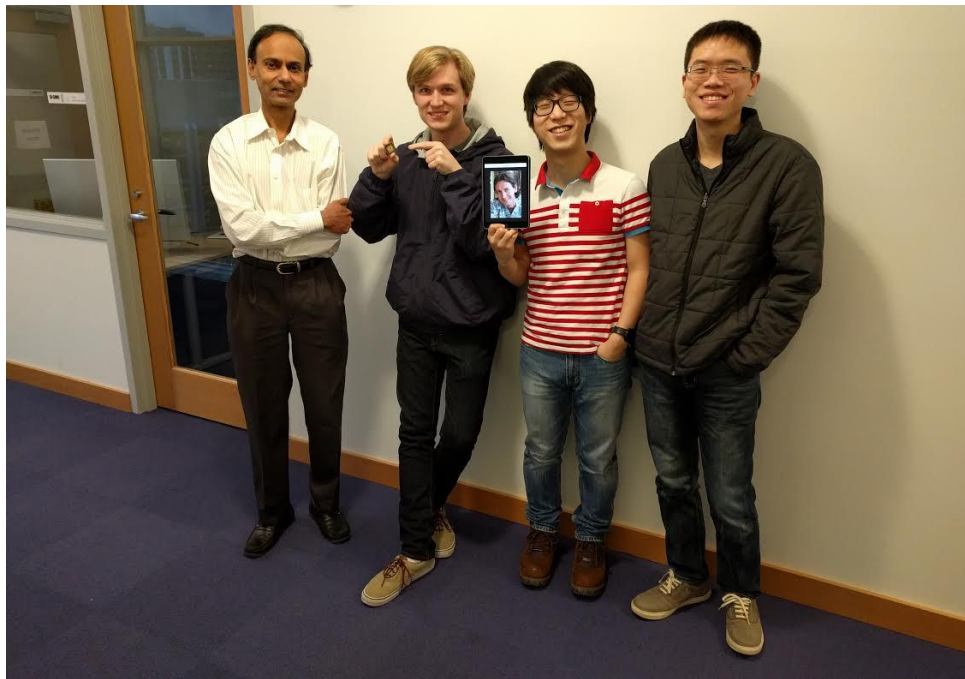
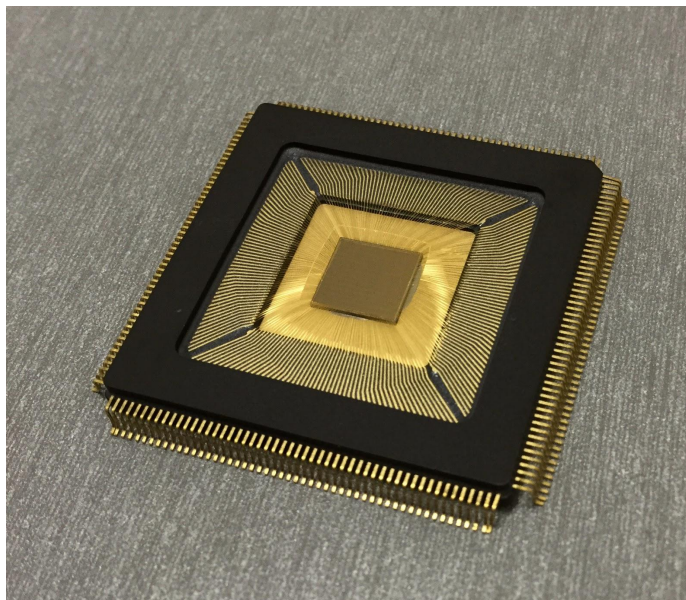
- Permute data in memory
- Shuffle data upon accesses

ORAM State of the Art



Any program can be made oblivious with
 $O(\log N)$ to $O(\log^2 N)$ overhead

[Otoroma, Circuit ORAM, ...]



ORAM State of the Art



Any program can be made oblivious with $O(\log N)$ to $O(\log^2 N)$ overhead



$\Omega(\log N)$ is necessary

[GO'96, LN'18]

ORAM State of the Art



Any program can be made oblivious with $O(\log N)$ to $O(\log^2 N)$ overhead

Implicit assumption:

Runtime is fixed and known

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- Must pad to worst-case runtime
- Can incur even **linear** overhead

Implicit assumption:

Runtime is fixed and known

Relax the obliviousness notion?

- Still provide meaningful privacy
- Significantly improve efficiency

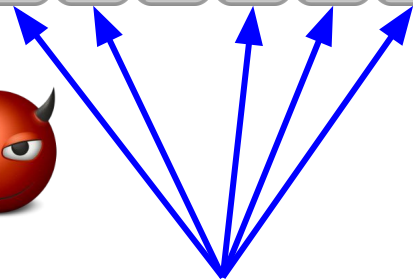




Differential Obliviousness

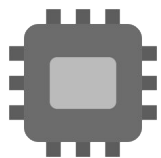
Inspired by differential privacy [Dwork et al. 05]

Memory



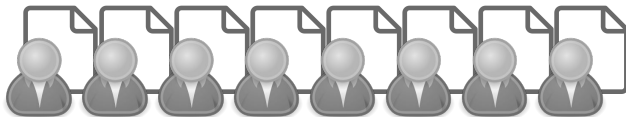
Algorithm

(e.g., compaction,
sorting)

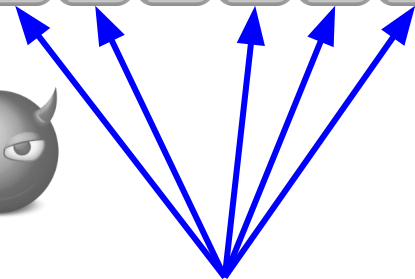


randomized

Database

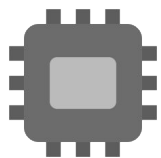


Memory



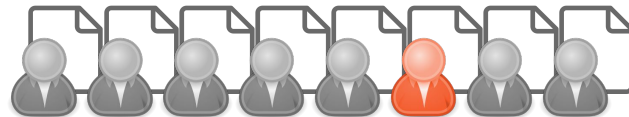
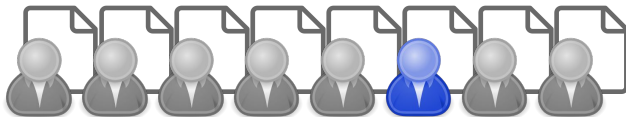
Algorithm

(e.g., data analytics)

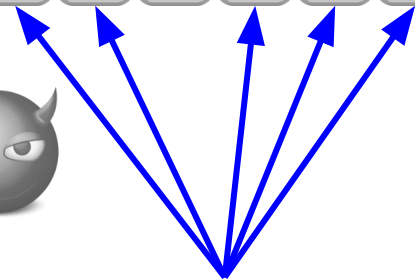


Neighboring input DBs

Database

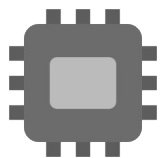


Memory

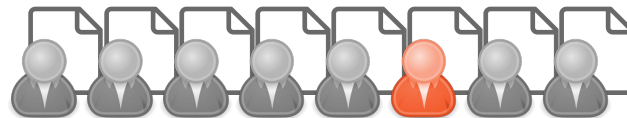
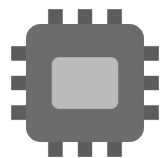
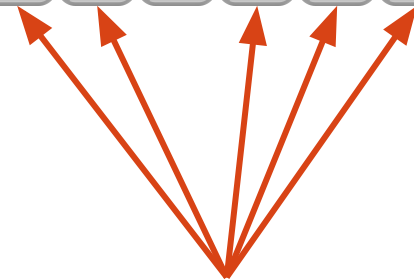
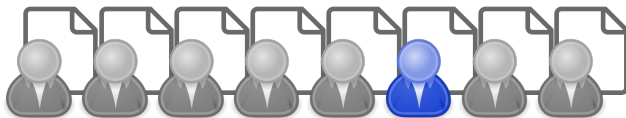


Algorithm

(e.g., data analytics)



Database



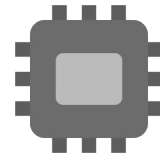
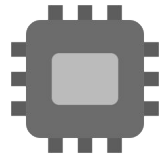
Access patterns on **neighboring** DBs must be **close**

Memory

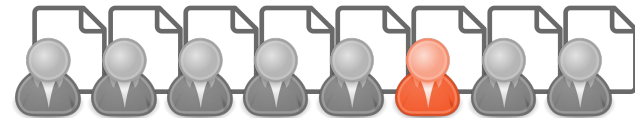
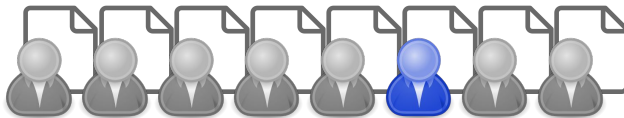


Algorithm

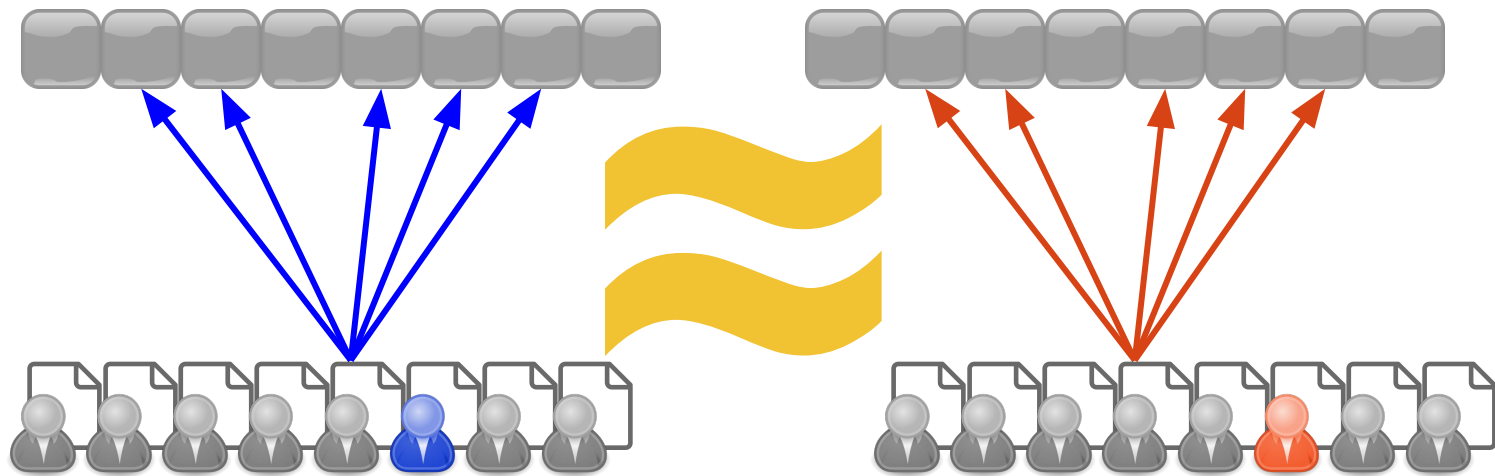
(e.g., data analytics)



Database



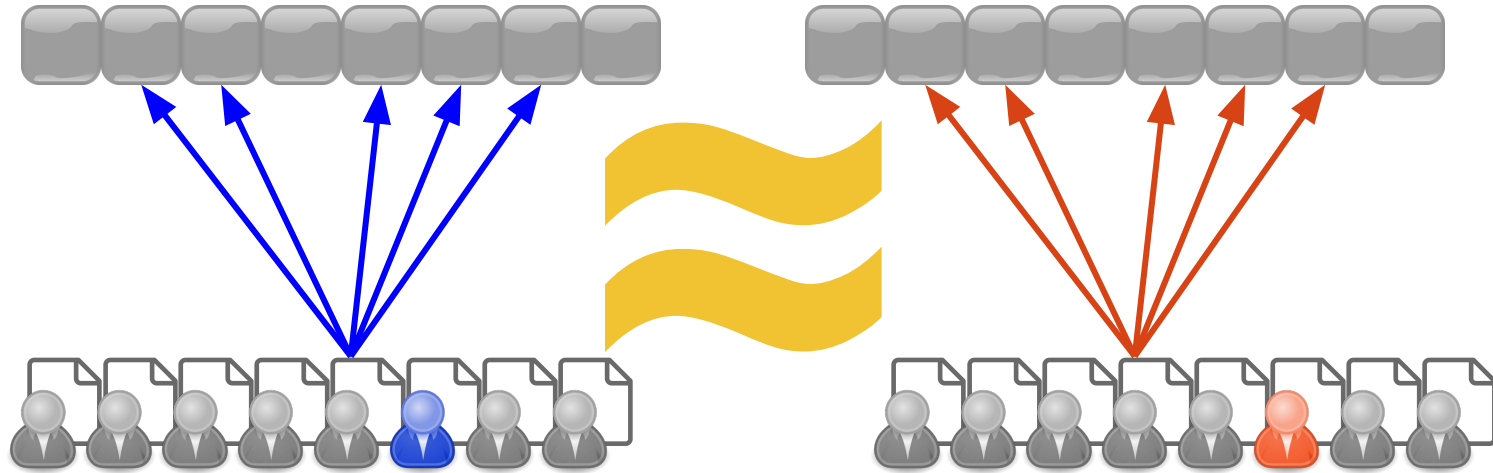
Access patterns on **neighboring** DBs must be **close**



$$\Pr[\text{blue arrows} \in S] \leq e^{\epsilon} \Pr[\text{red arrows} \in S] + \delta$$

This must hold for any S

(ϵ, δ) -Differential Obliviousness



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(ϵ, δ) -Differential Obliviousness

- 1 What is being relaxed?
- 2 Still provide meaningful privacy?
- 3 Overcome obliviousness barriers?

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What is being relaxed?

Closeness needs to hold only for neighboring DBs

$$\Pr[\text{blue arrows} \in S] \leq e^{\epsilon} \Pr[\text{red arrows} \in S] + \delta$$

This must hold for any S

What is being relaxed?

Closeness needs to hold only for neighboring DBs

Allow multiplicative, non-negl. loss

$$\Pr[\text{blue arrows} \in S] \leq e^{\epsilon} \Pr[\text{red arrows} \in S] + \delta$$

This must hold for any S

Does not require padding
to worst-case runtime

$$\Pr[\text{blue arrows} \in S] \leq e^{\epsilon} \Pr[\text{red arrows} \in S] + \delta$$

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(ϵ, δ) -Differential Obliviousness

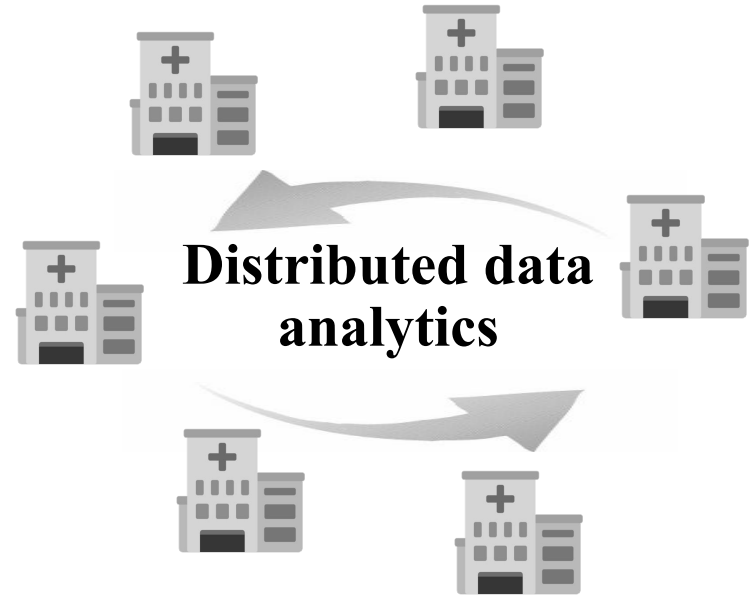
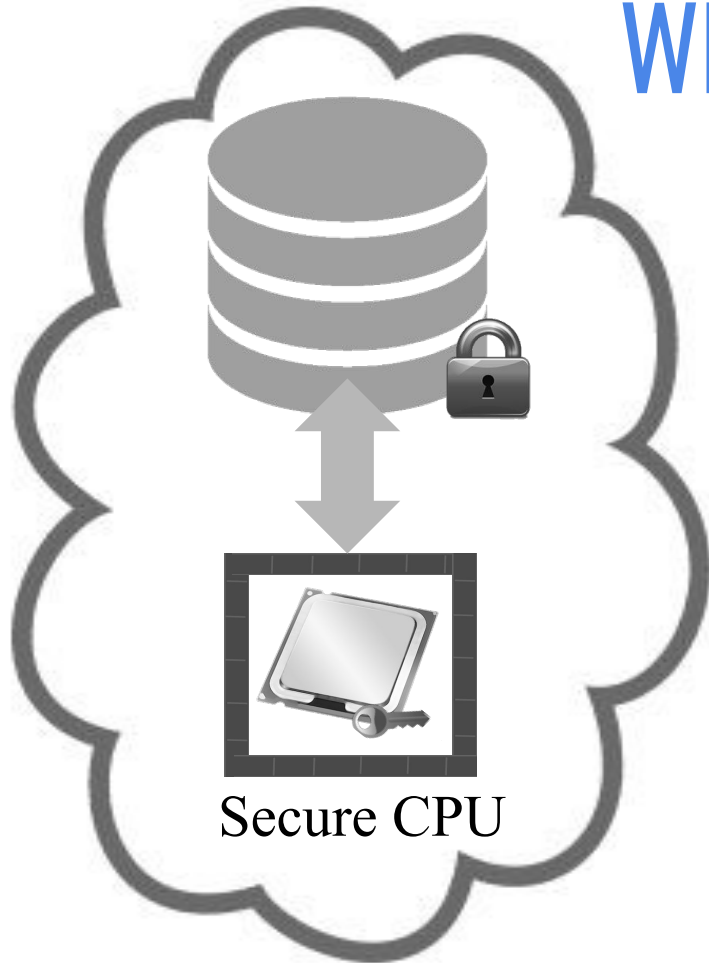
- 1 What is being relaxed?
- 2 Still provide meaningful privacy?**
- 3 Overcome obliviousness barriers?



Bad idea if you are protecting your
Bitcoin signing key!

2 Still provide meaningful privacy?

When does DO make sense?



Typical parameters

Constant

Negl. in **N**

$$\Pr[\text{blue arrows} \in S] \leq e^{\epsilon} \Pr[\text{red arrows} \in S] + \delta$$

This must hold for any S

(ϵ, δ) -Differential Obliviousness

- 1 What is being relaxed?
- 2 Still provide meaningful privacy?
- 3 **Overcome obliviousness barriers?**

Stable Compaction

Obliviousness

$\Omega(N \log N)$

necessary

Differential

Obliviousness

$O(N \log \log N)$

Stable Compaction



Stable Compaction: Why do we care?

- Simple yet non-trivial
- Frequent algorithmic building block
- Warmup scheme in paper

Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm



Stable Compaction: **insecure** algorithm

Completes in $O(N)$ time

Leaks exact progress

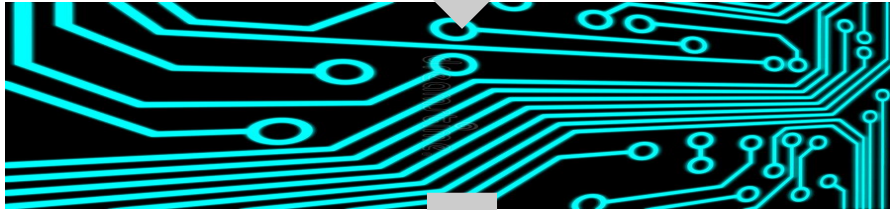
Stable Compaction: **oblivious** algorithm



Stable Compaction: **oblivious** algorithm



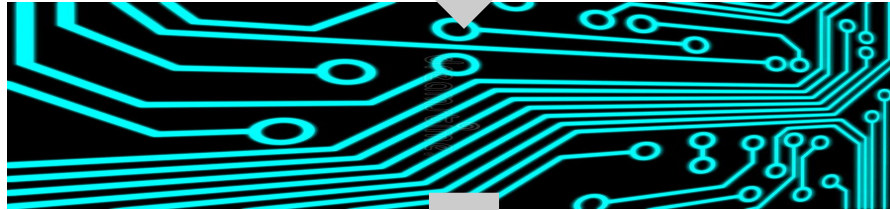
Sorting
network



Takes $N \log N$ time



Sorting
network

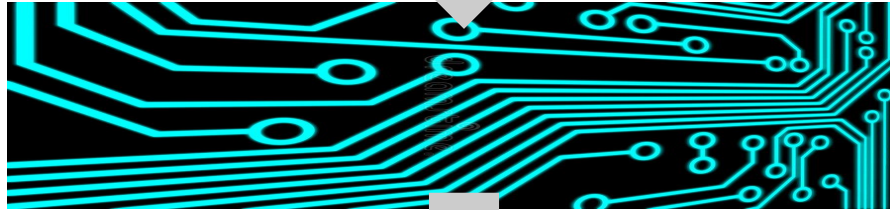


$N \log N$ time is necessary for obliviousness

Assumption: algorithm does not perform encoding on the kitties



Sorting
network



Stable Compaction

Obliviousness

$\Omega(N \log N)$

necessary

**Differential
Obliviousness**

$O(N \log \log N)$



Obliviousness

Cannot leak progress

Differential Obliviousness

Leak rough notion of
progress



polylog(N) batch

DP oracle

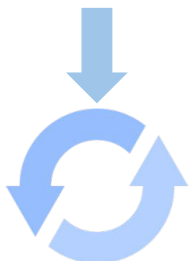
2~5 kitties so far





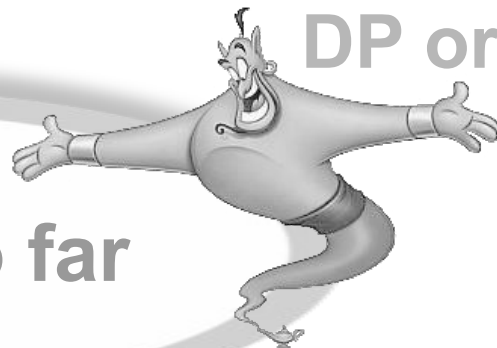
polylog(N) batch

O-sort



2~5 kitties so far

DP oracle



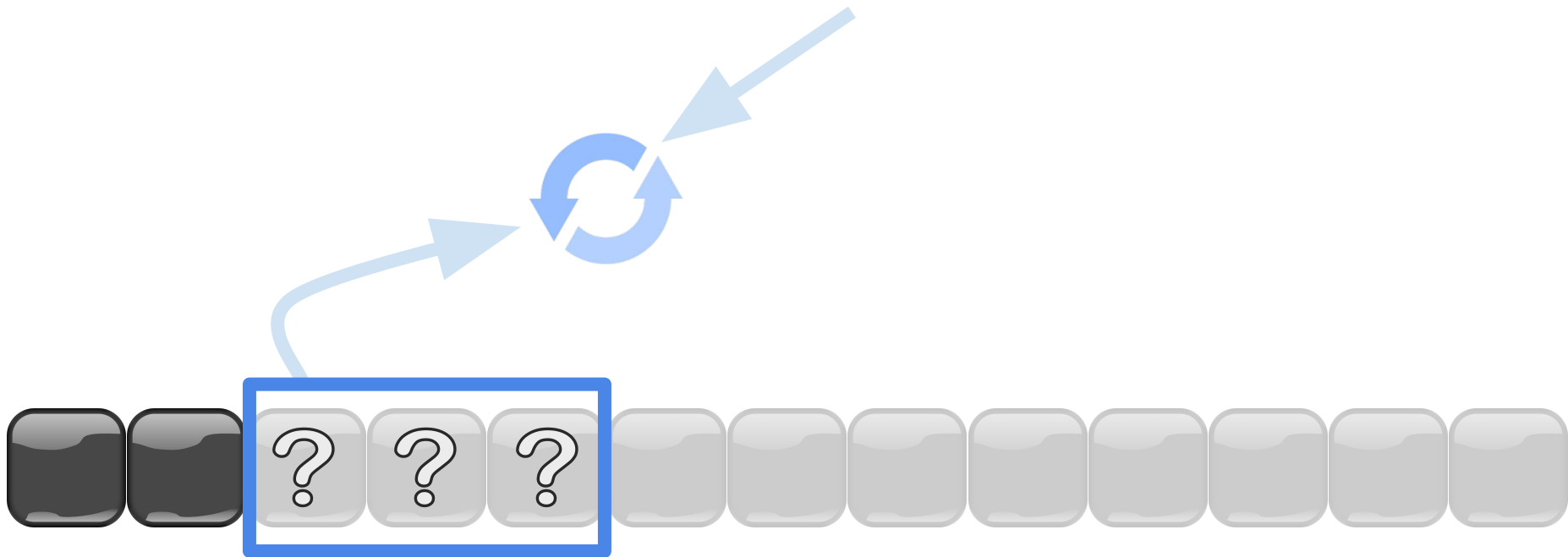


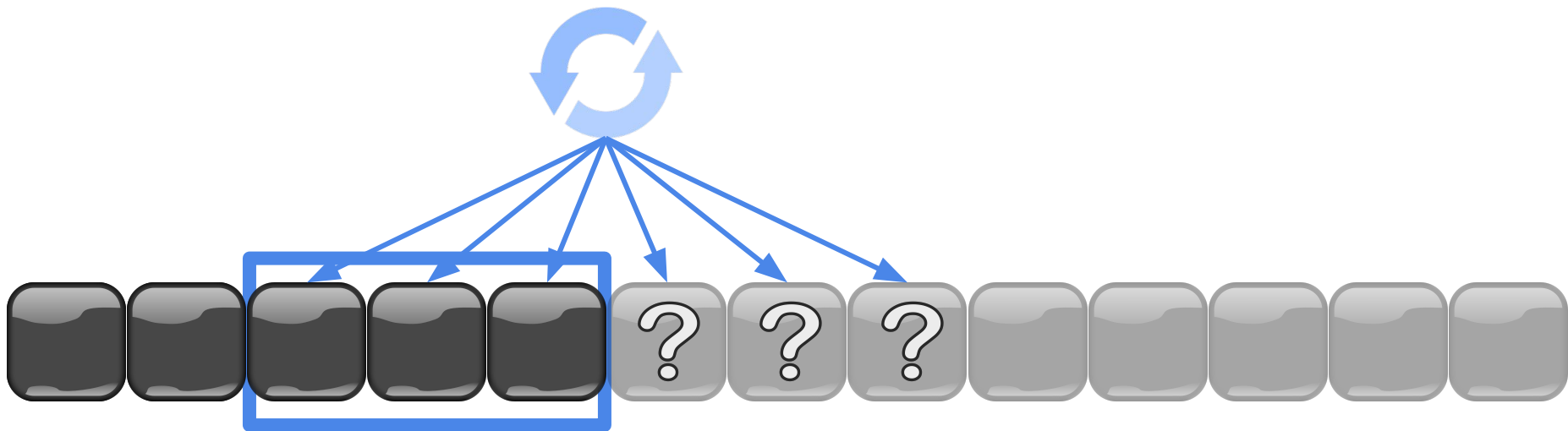
$\text{polylog}(N)$ batch

O -sort



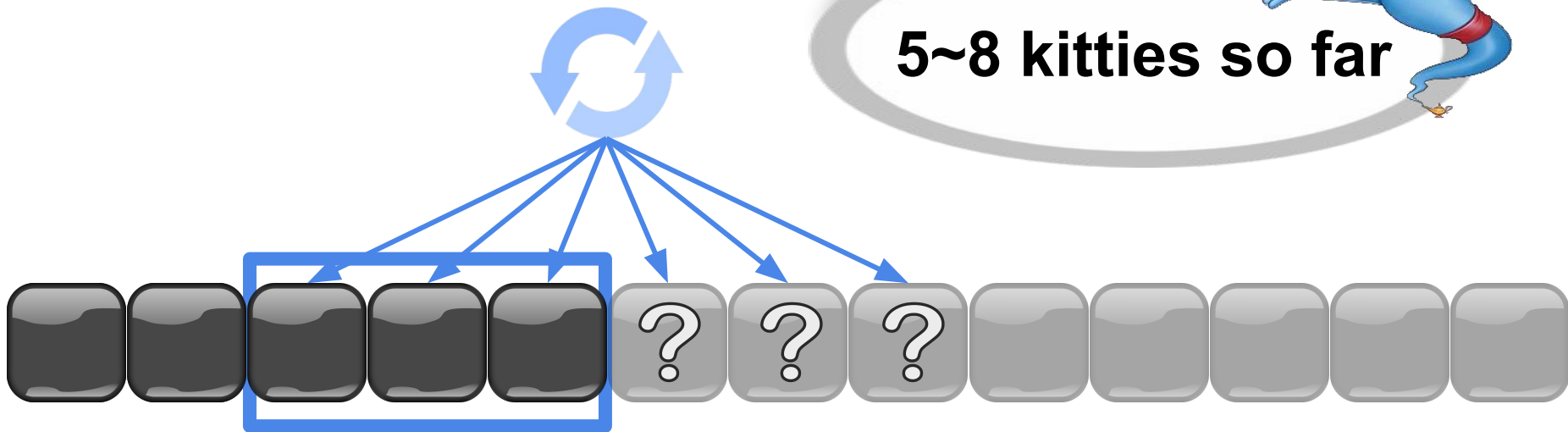
$\text{polylog}(N)$ error,
DP estimate

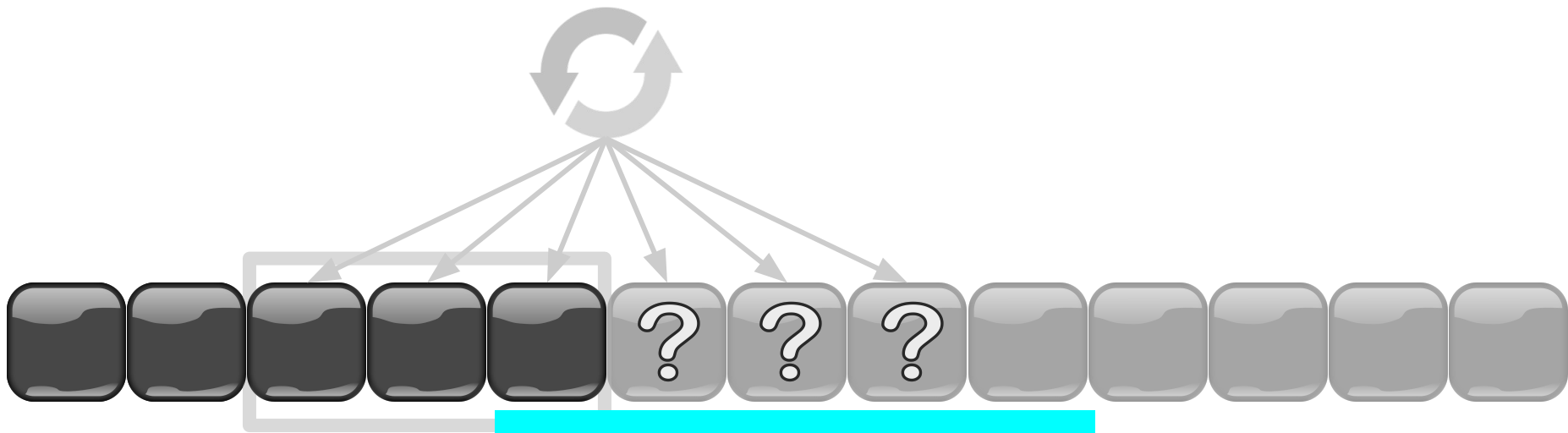






5~8 kitties so far

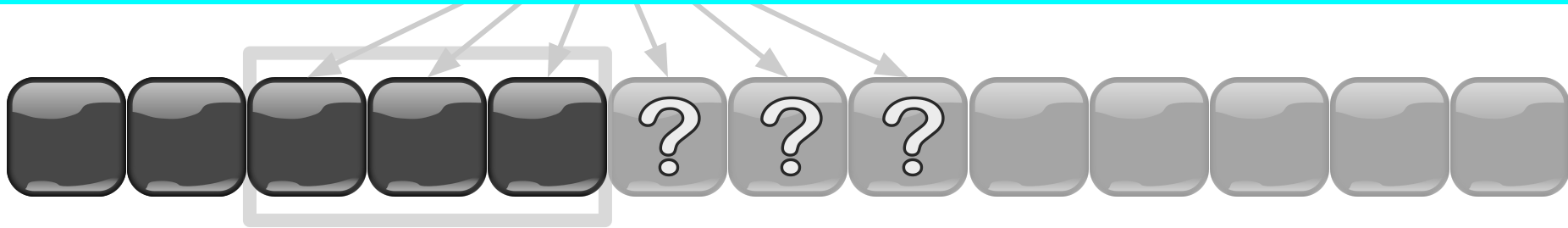




**polylog(N) error,
DP estimate**



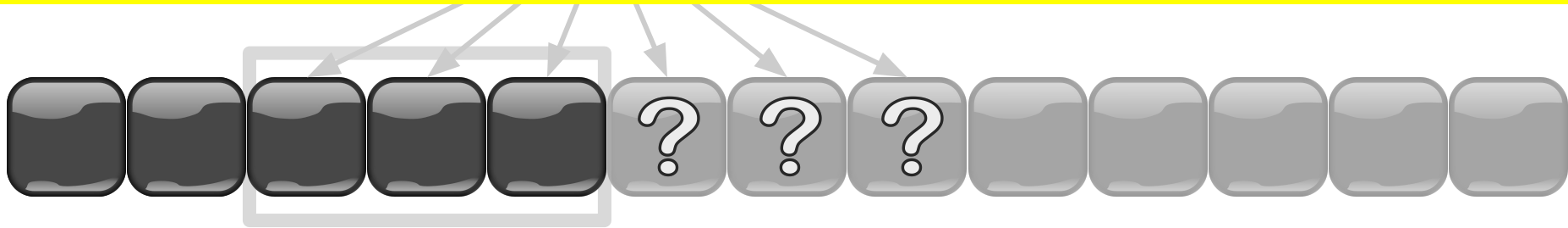
Completes in
 $O(N \log \log N)$
time





Need:

Oblivious and **DP** alg. that estimates
all prefix sums, with **polylog** error





Naive algorithm:

- Compute all **N** prefix sums
- Add **independent** noise to each

All prefix sums -- DP and Oblivious

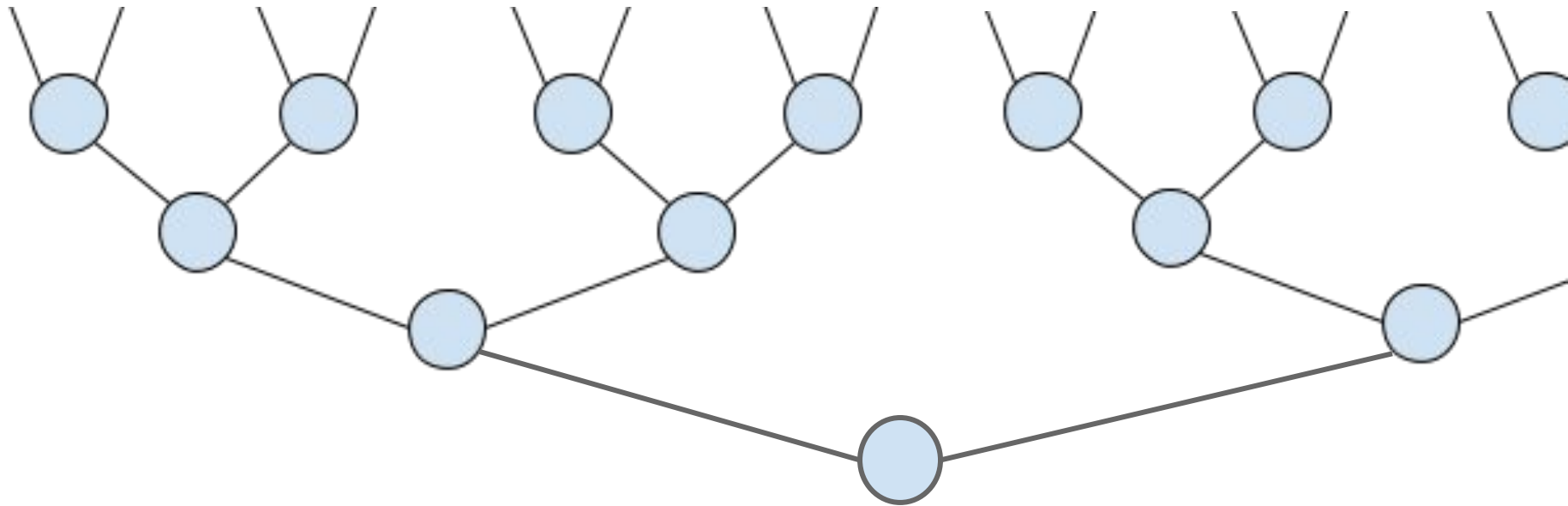


Naive algorithm:

- Compute all N prefix sums
- Add *independent* noise to each

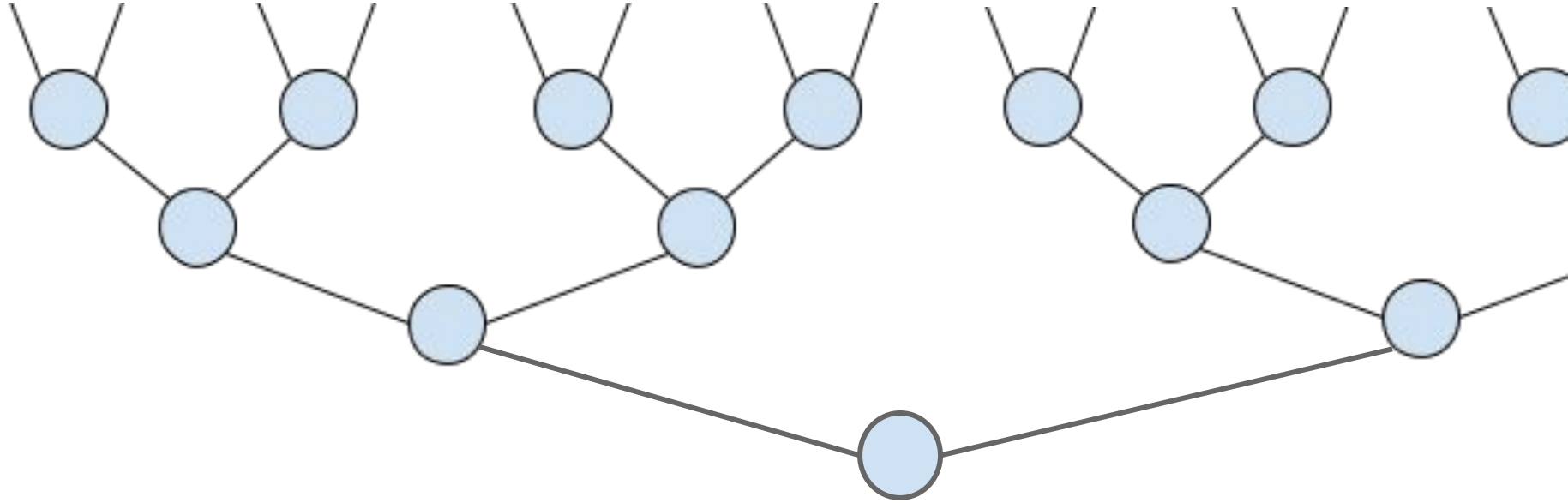


Incurs $\Theta(N)$ error

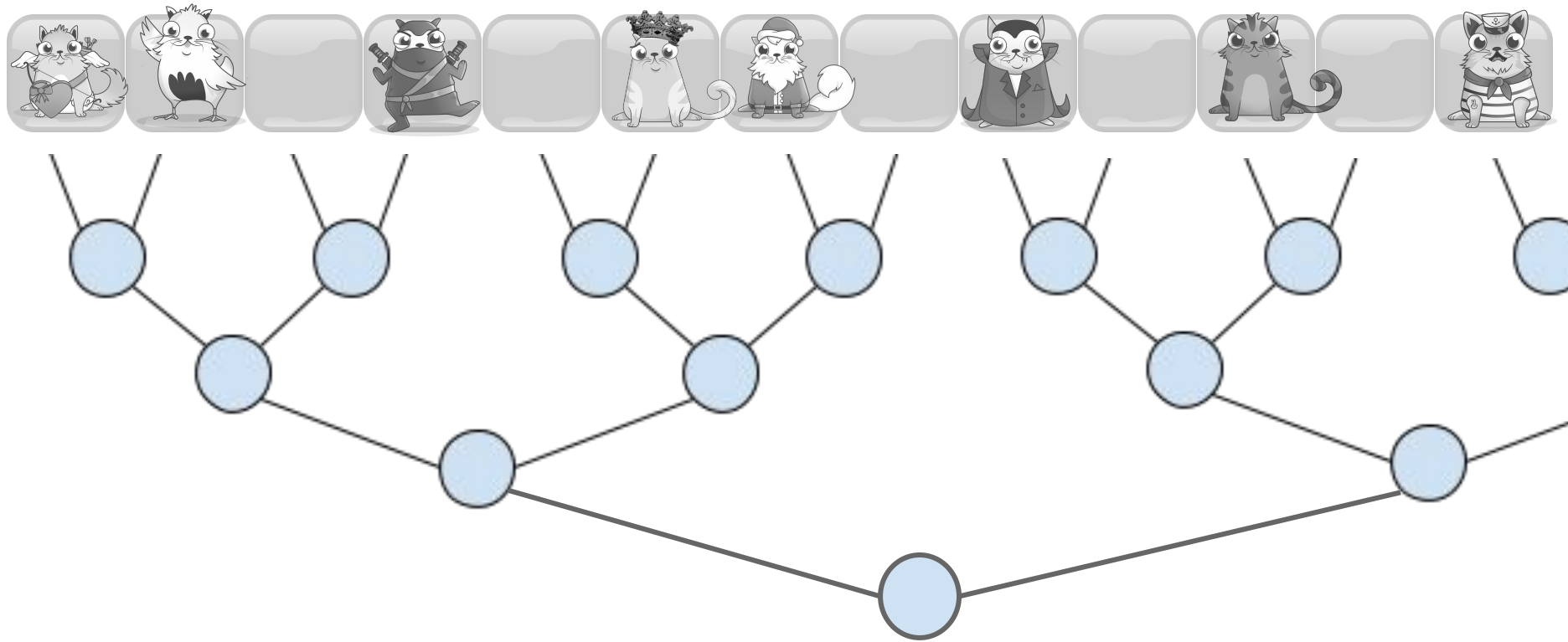


[Dwork et al. 10, CS'10]

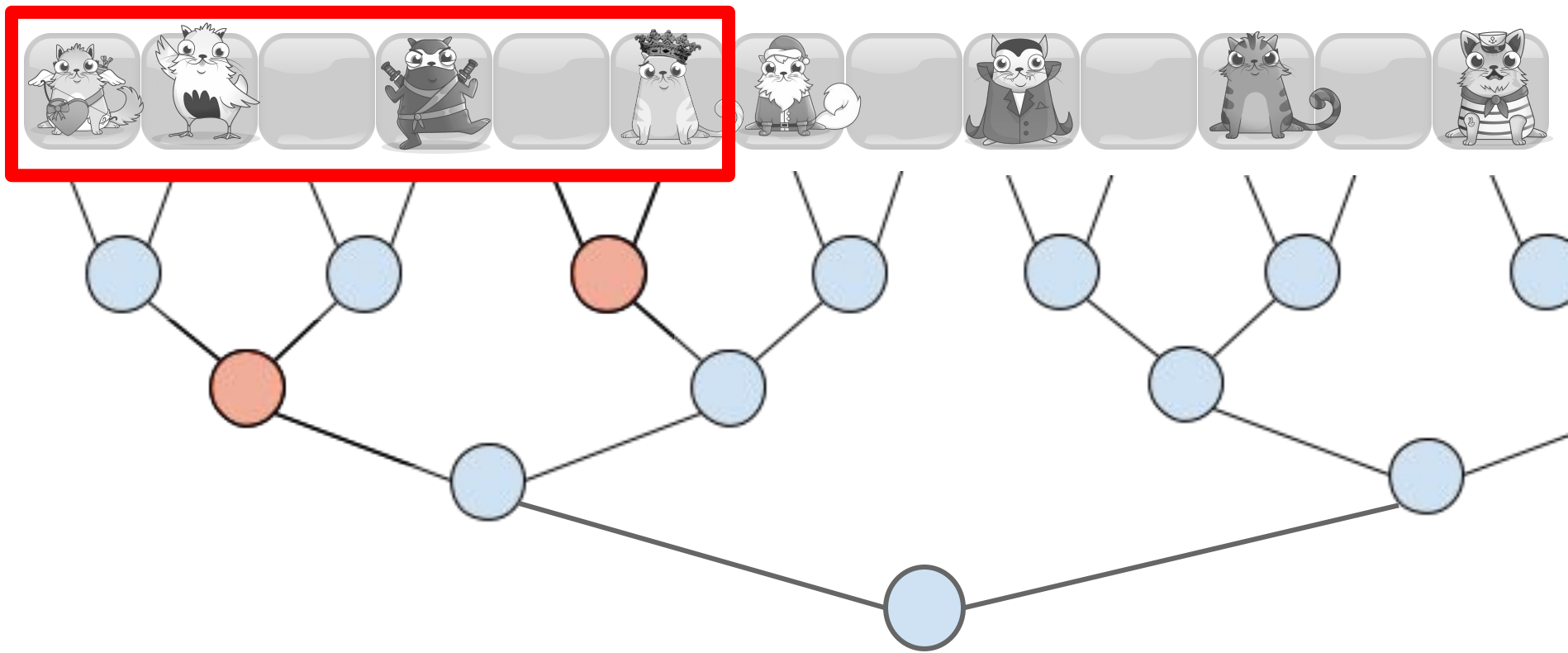
All prefix sums -- DP and Oblivious



- Every node in the tree represents a range
- Compute DP estimate for every node in the tree



- Every input appears in only $\log N$ nodes!
- Achieve only $\Theta(\log N)$ error per node!



- Every prefix sum is the sum of $\log N$ nodes
- Achieve $\text{poly } \log N$ error for each prefix sum

Summary: Leak rough notion of progress



Non-trivial combination of DP and oblivious algorithms



Apply oblivious alg to
small bins



Make DP mechanisms
oblivious

Putting it altogether

There exists an $O(N \log \log N)$
time, $(\Theta(1), \text{negl}(N))$ -DO algorithm
that realizes stable compaction

Is this necessary?

There exists an $O(N \log \log N)$
time, $(\Theta(1), \text{negl}(N))$ -DO algorithm
that realizes stable compaction

$(\epsilon, 0)$ -Differentially Oblivious Stable
Compaction:

$N \log N$ is necessary

even when ϵ is arbitrarily large!



Other Results in Our Paper

An iceberg floating in the ocean, with a small portion visible above the water and a much larger, more complex shape submerged below. The background is a deep blue sea under a cloudy sky.

- **lower bounds for obliviousness**
- **merging, range query DB**
 - Differentially oblivious algorithms with $O(\log \log N)$ blowup.
 - $\Omega(\log N)$ blowup necessary for full obliviousness.

Closely Related Works

[Wagh et al.] DP-ORAM, achieve $O(1)$ gain

[Kellaris et al.] DP for length, otherwise fully oblivious

[Mazloom et al.] DP access patterns for MPC



This is just a beginning.

Differential obliviousness for generic programs?

Composition?

Alternative notions?

Practical performance?



This is just a beginning.

Differential obliviousness for generic programs?

Composition?

Alternative notions?

Practical performance?

Thank you!

elaine@cs.cornell.edu