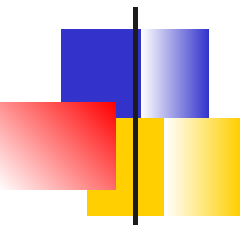




UCMERCED

# Bi-Level Online Aggregation On Raw Data



**Yu Cheng<sup>+</sup>, Weijie Zhao<sup>\*</sup>, Florin Rusu<sup>\*</sup>**

<sup>+</sup>: Amobee. Inc.

<sup>\*</sup>: University of California, Merced



# Outline

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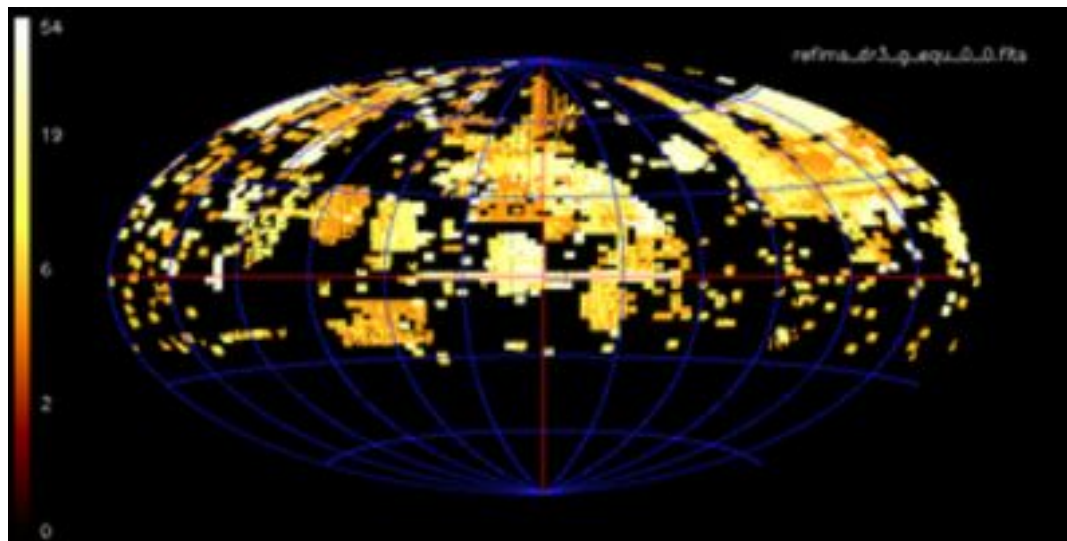
- Background
- Problem
- OLA-RAW
- Evaluation



# Palomar Transient Factory (PTF)

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The Palomar Transient Factory (PTF) project aims to identify and automatically classify transient astrophysical objects such as variable stars and supernovae in realtime





# Illustrative Example

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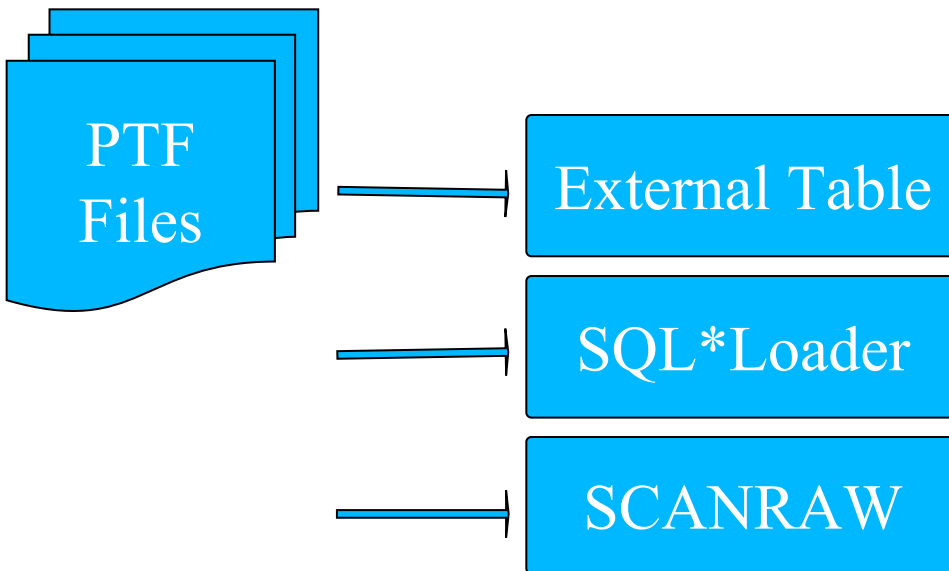
- Supernova identification



PTF  
Files

```
SELECT AGGREGATE(expression) AS agg  
FROM candidate  
WHERE predicate  
HAVING agg < threshold
```

# Existing Solutions



Time to query	Execution	Storage
instant	slow	zero
loading	fast	full replication
instant	fast	adaptive



# Illustrative Example

---

- Supernova identification



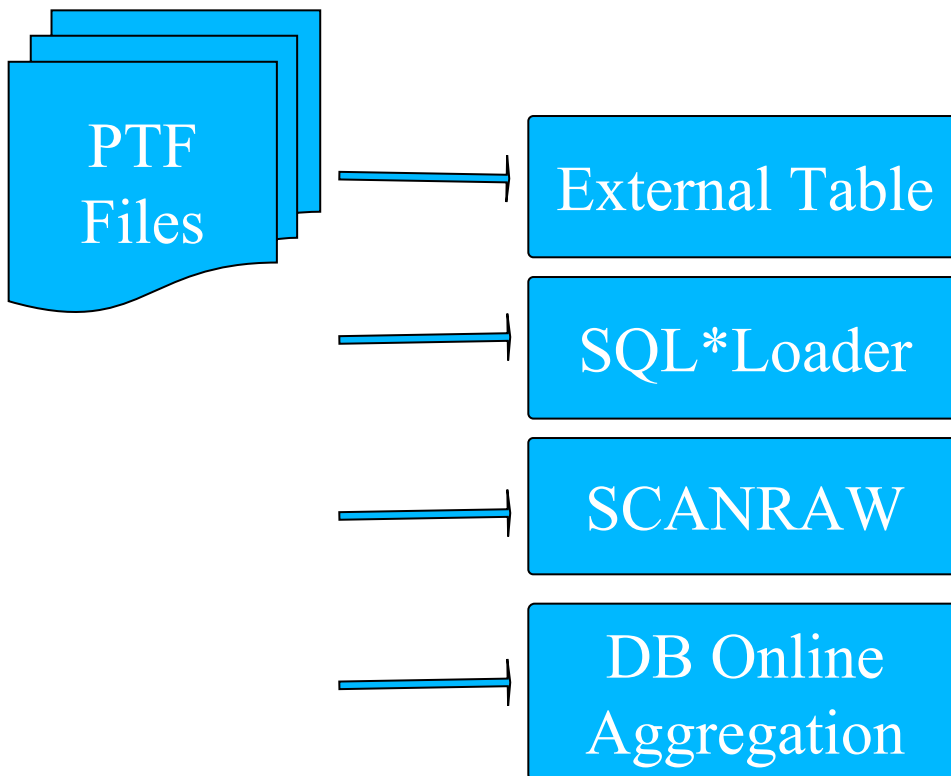
PTF  
Files

```
SELECT AGGREGATE(expression) AS agg
FROM candidate
WHERE predicate
HAVING agg > threshold
WITH ACCURACY  $\alpha$ 
```



# Existing Solutions

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Time to query	Execution	Storage
instant	slow	zero
loading	fast	full replication
instant	fast	adaptive
loading + shuffling	faster	full replication

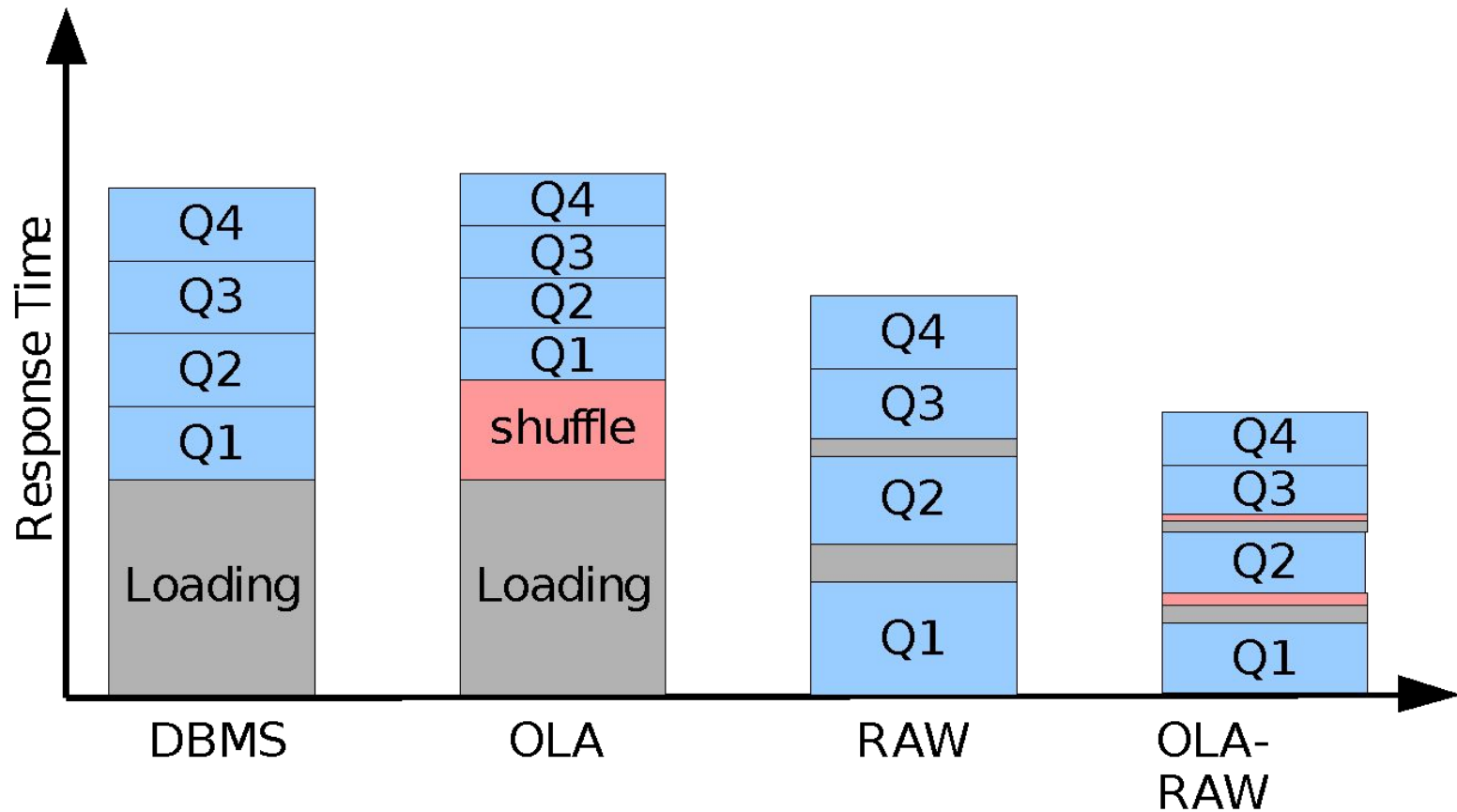


# Research Problem

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- Can we find a better solution to execute approximate queries in-situ over raw files?
  - Instant access to data  
**In-situ data processing**
  - Generate results faster  
**Online aggregation (OLA)**
  - Minimize used storage  
**In-memory synopsis**

# High Level Approach





# Related Work

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- Adaptive partial loading [Idreos et al., CIDR 2011]  
Only load necessary attributes before query starts
- NoDB [Alagianis et al., SIGMOD 2012]  
Instead of loading, build index and cache necessary attributes in memory
- Invisible loading [Abouzied et al., EDBT/ICDT 2013]  
Portion of necessary data is loaded into database for every query
- Data vaults [Ivanova et al., SSDBM 2012]  
Memory cache for complex data in scientific repositories
- SCANRAW [Cheng and Rusu, SIGMOD 2014]  
Load data using spare system resources without affecting query processing



# OLA-RAW

---

- ❖ OnLine Aggregation for RAW data processing
  - How to generate random samples from raw files?
  - Design a feasible architecture to combine online aggregation with in-situ data processing
  - Find an efficient method to maintain extracted samples



# OLA-RAW

---

## ❖ OnLine Aggregation for RAW data processing

- How to generate random samples from raw files?  
**Bi-Level Sampling**
- Design a feasible architecture to combine online aggregation with in-situ data processing
- Find an efficient method to maintain extracted samples

# Sampling and Estimator

Raw Data

1	1	1	1	2	28	6	9	9	12	8	25	3	3	3	3
---	---	---	---	---	----	---	---	---	----	---	----	---	---	---	---

Cluster Sampling

9	12	8	25
---	----	---	----

1	1	1	1
---	---	---	---

Bi-Level Sampling

2	28	6	9
---	----	---	---

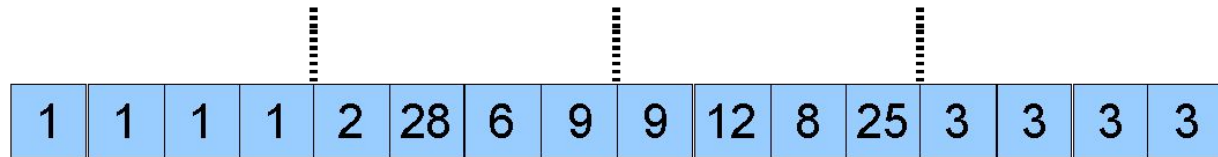
9	12	8	25
---	----	---	----

1	1	1	1
---	---	---	---

$$\hat{\tau} = \frac{N}{n} \sum_{i=1}^n \hat{y}_i \quad \hat{y}_i = \frac{M_i}{m_i} \sum_{j=1}^{m_i} y_{ij}$$

# Sampling and Estimator

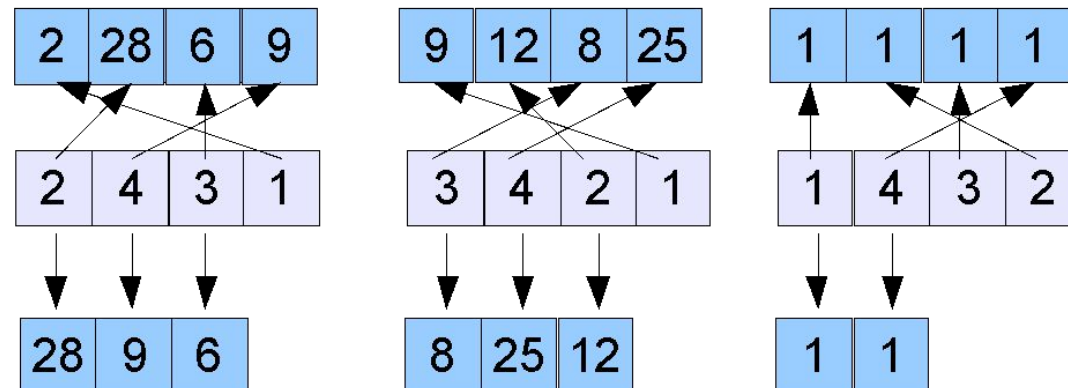
Raw Data



Cluster Sampling



Bi-Level Sampling





# Sampling and Estimator

---

$$Var(\hat{\tau}) = N(N - n) \frac{s_u^2}{n} + \frac{N}{n} \sum_{i=1}^n M_i(M_i - m_i) \frac{s_i^2}{m_i}$$

$$s_u^2 = \frac{1}{n-1} \sum_{i=1}^n (\hat{y}_i - \hat{\mu}_1)^2, \quad s_i^2 = \left( \frac{1}{m_i-1} \right) \sum_{j=1}^{m_i} (y_{ij} - \bar{y}_i)^2$$

where  $i = 1, \dots, n$ , and  $\hat{\mu}_1 = (1/n) \sum_{i=1}^n \hat{y}_i$ .

- $n$  : number of chunks
- $m$  : number of processed tuples



# OLA-RAW

---

## ❖ OnLine Aggregation for RAW data processing

- How to generate random samples from raw files?

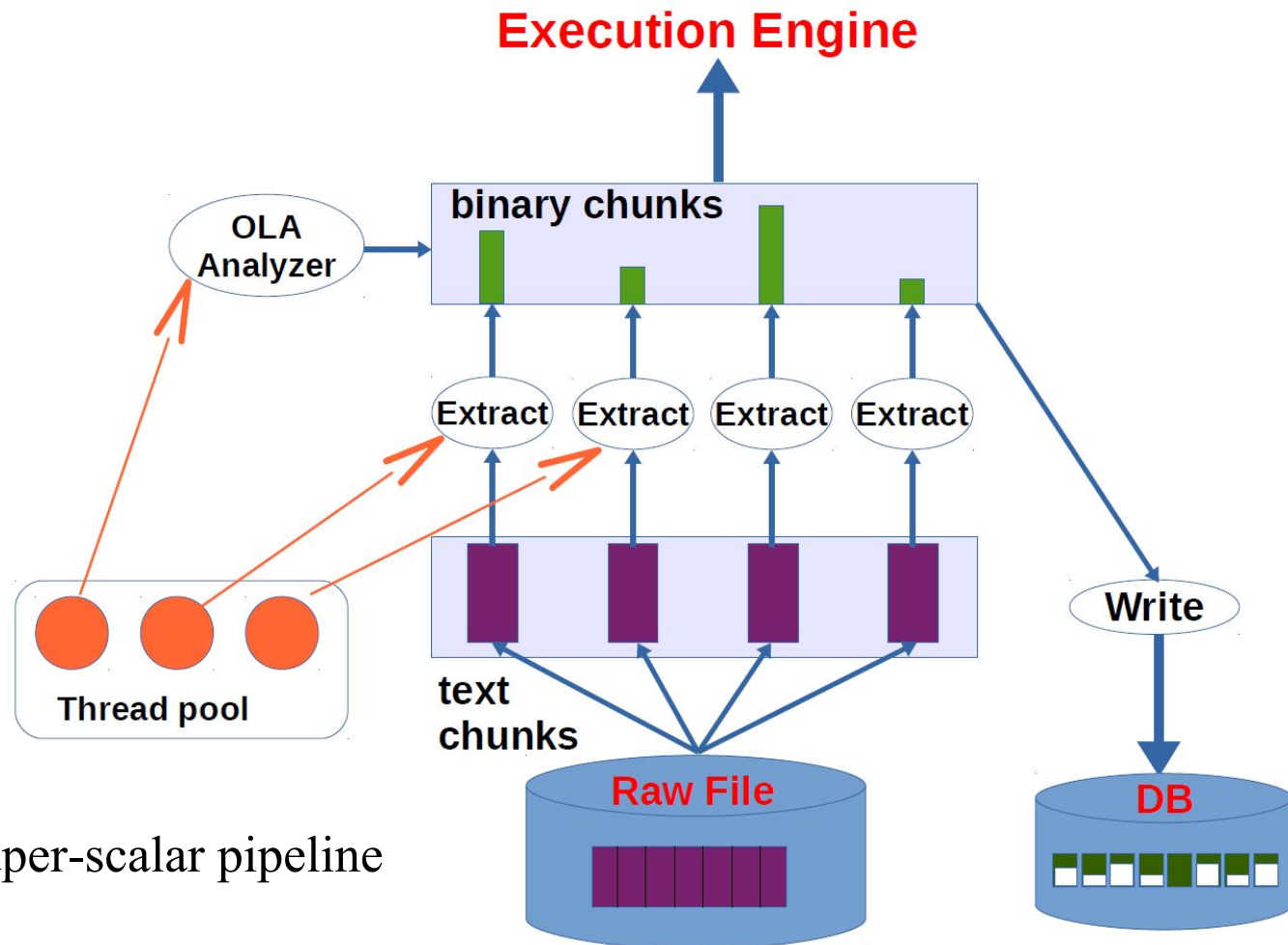
### Bi-Level Sampling

- Design a feasible architecture to combine online aggregation with in-situ data processing

### OLA-RAW

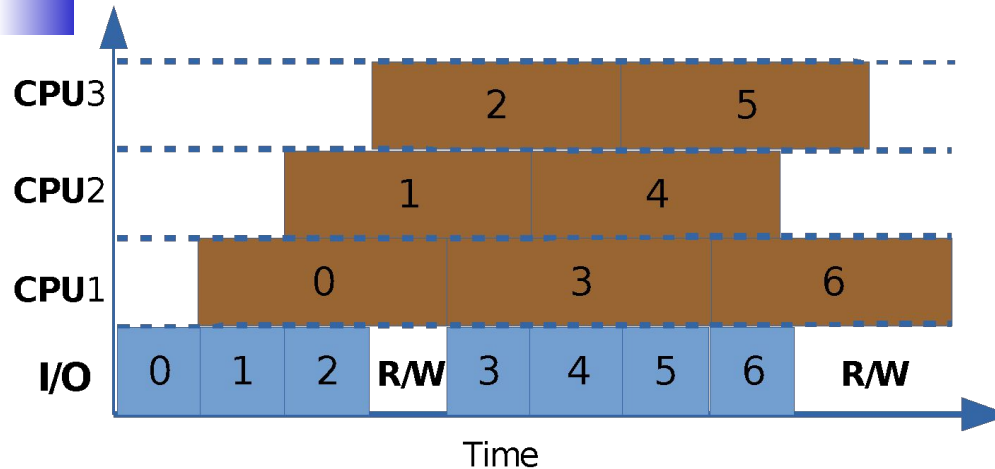
- Find an efficient method to maintain processed samples

# Architecture

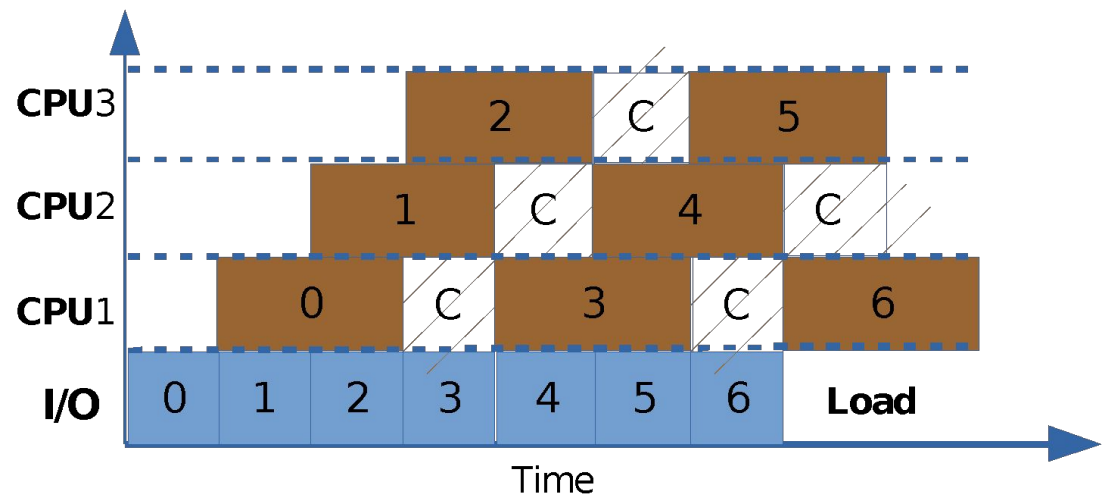


- Parallel super-scalar pipeline

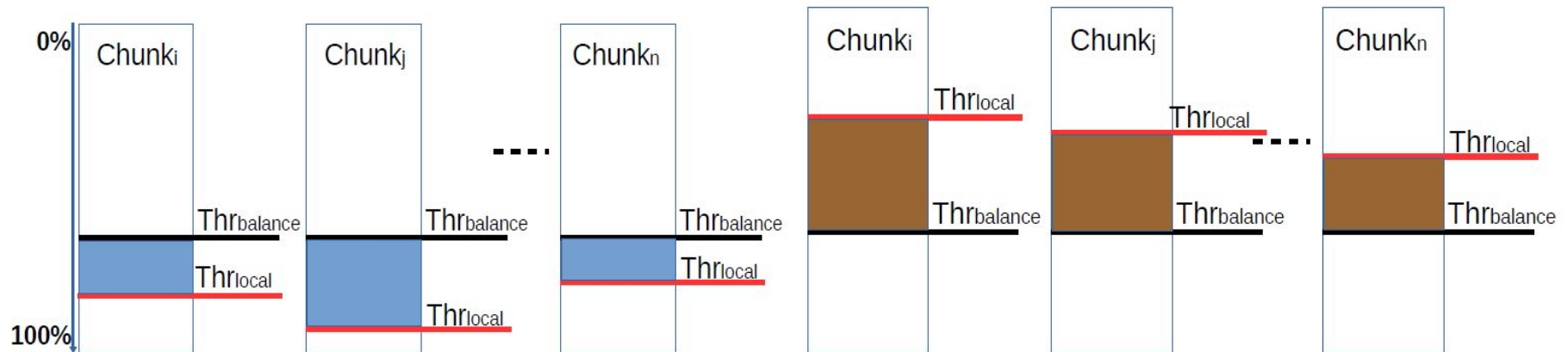
# Where Does the Time Go?



- I/O-bound
- permutation generation
- process more tuples



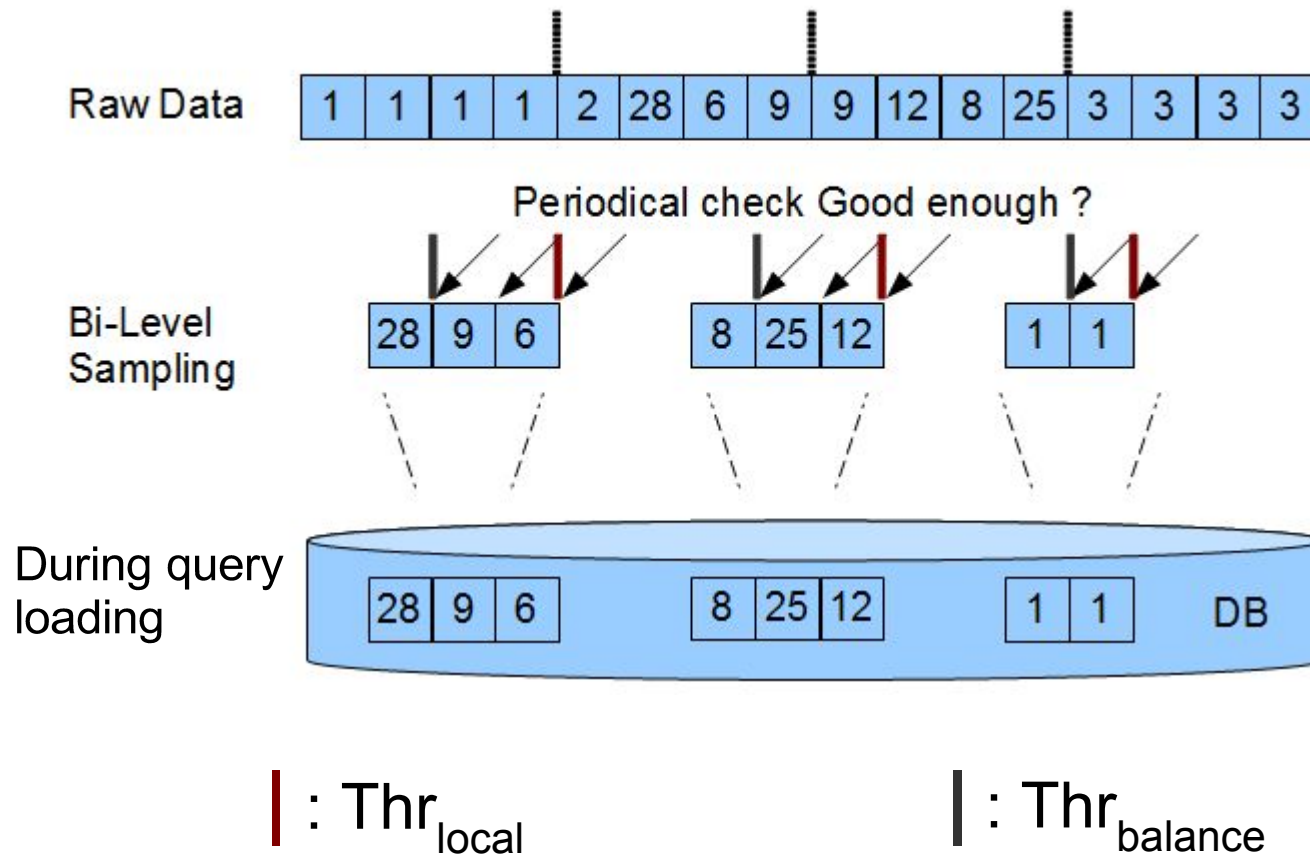
# How many samples are enough?



- Make sure to generate good enough estimation by accessing raw data only once
- Generate accurate estimate for each chunk

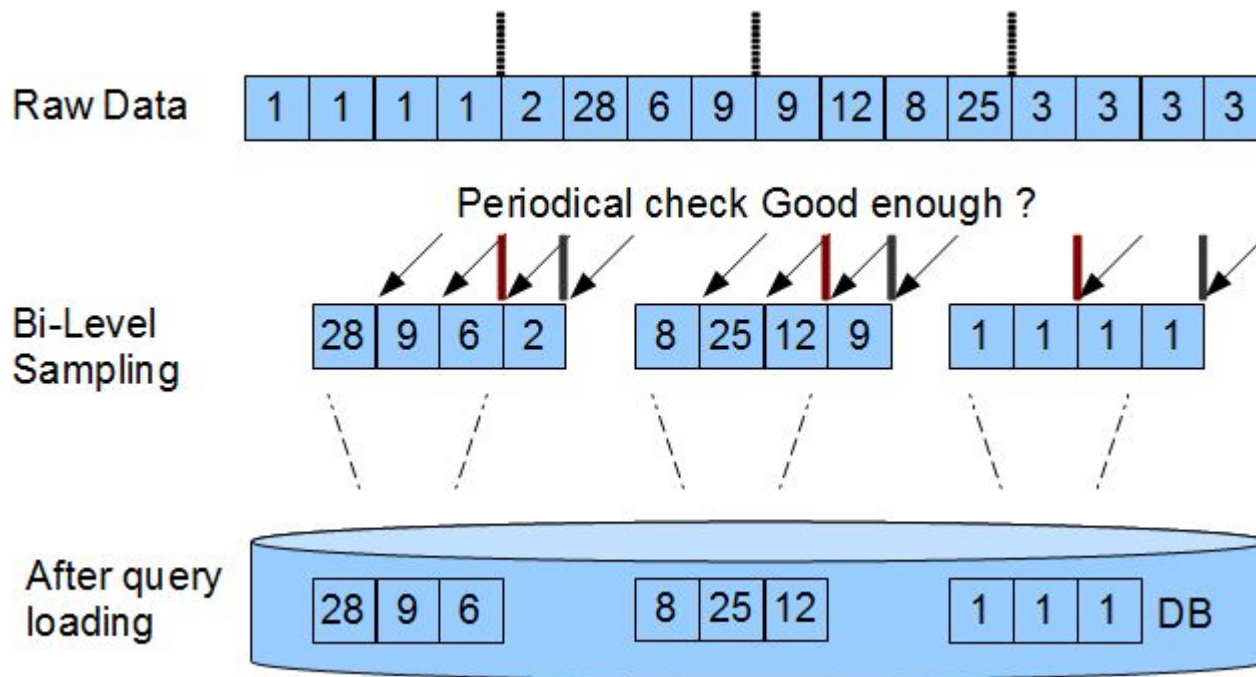
# Query Processing

CPU-bound process



# Query Processing

## IO-bound process



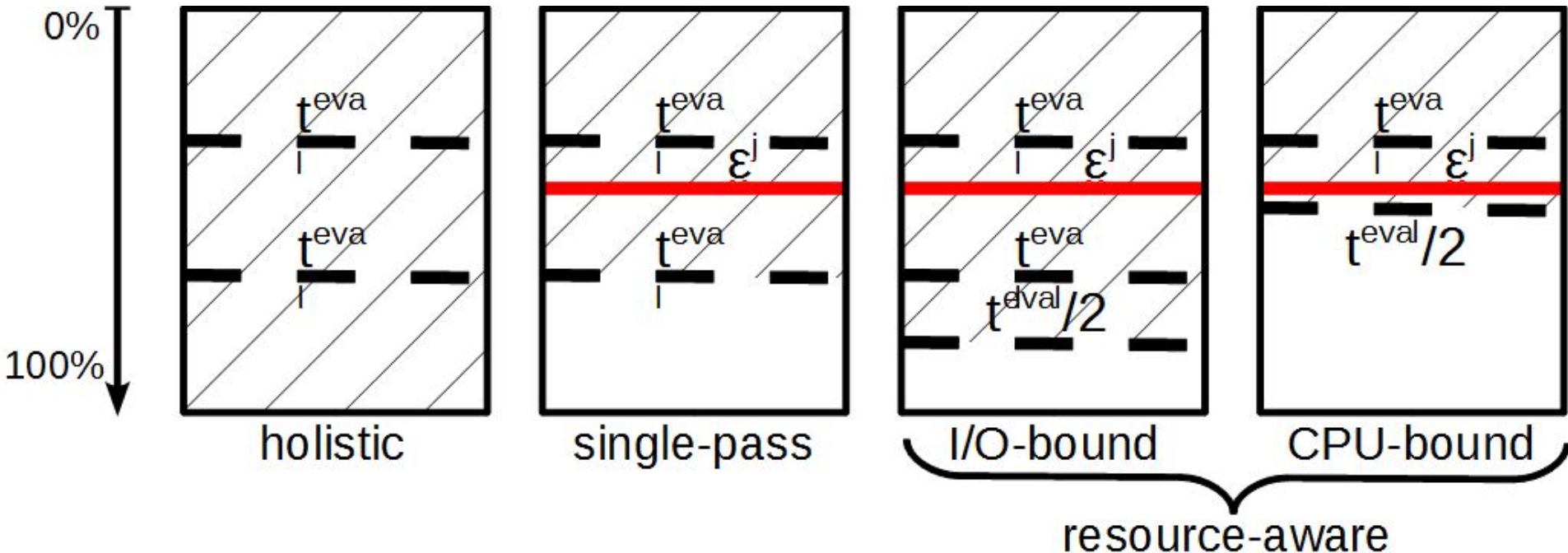
| :  $\text{Thr}_{\text{local}}$

| :  $\text{Thr}_{\text{balance}}$

# Sampling Strategy

## ❖ Parallel sampling procedure

Result order  $\neq$  Random chunk order  $\rightarrow$  Inspection paradox





# OLA-RAW

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## ❖ OnLine Aggregation for RAW data processing

- How to generate random samples from raw files?

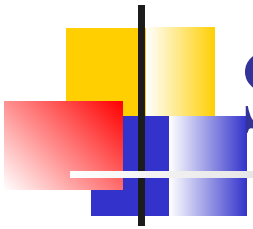
Bi-Level Sampling

- Design a feasible architecture to combine online aggregation with in-situ data processing

OLA-RAW

- Find an efficient method to maintain processed samples

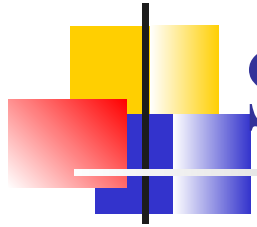
In-memory sample synopsis



# Sample Maintenance

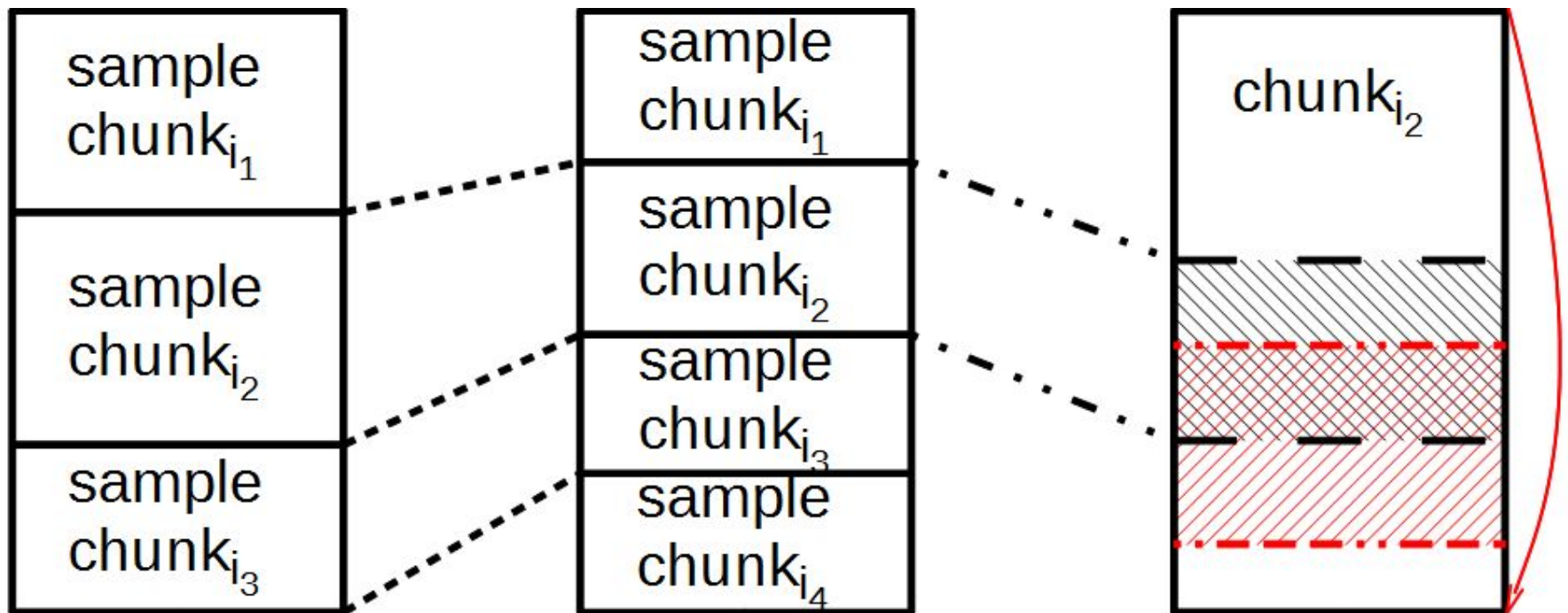
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- What kind of samples should be preserved?  
**Variance-driven**
- When to load the samples?  
**During query or loading after query processing**
- How to make sure the additional samples have not been selected before?  
**Permutation seeds + offset**



# Sample Maintenance

## ❖ Variance-driven sample swap policy





# Evaluation

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Data : The PTF dataset with 1 billion transient detection tuples. Each tuple has 8 attributes, 6 of which are real numbers with 10 decimal digits

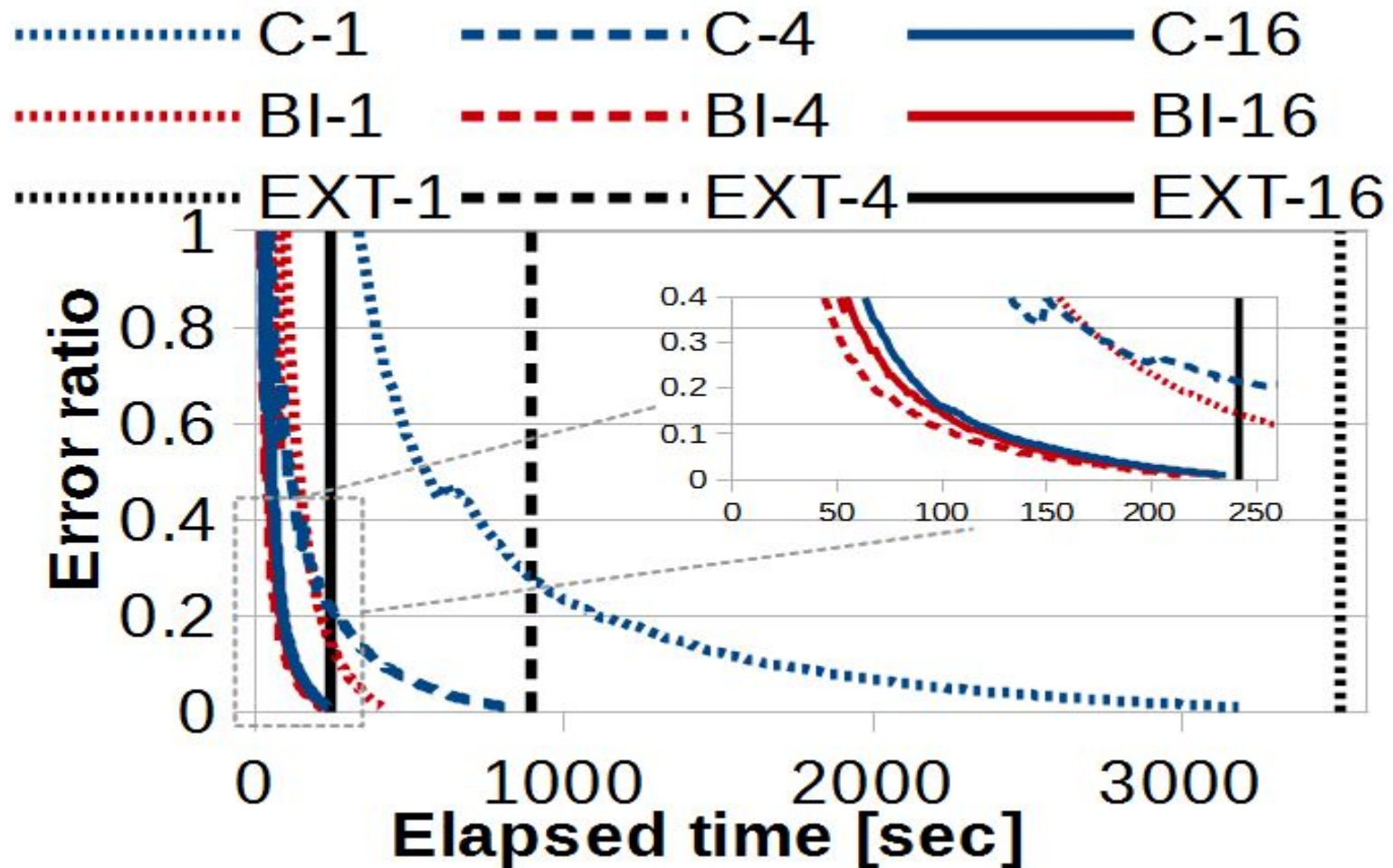
Query :

```
SELECT SUM (  $\sum_{j=1}^K C_{ij}$  ) FROM FILE
```

System : 2 AMD 8-core processors, 40 GB of memory, 4 disks in RAID-0 with I/O throughput 450 MB/s

Illustration: 16 attributes,  $2^{26}$  lines, 20GB

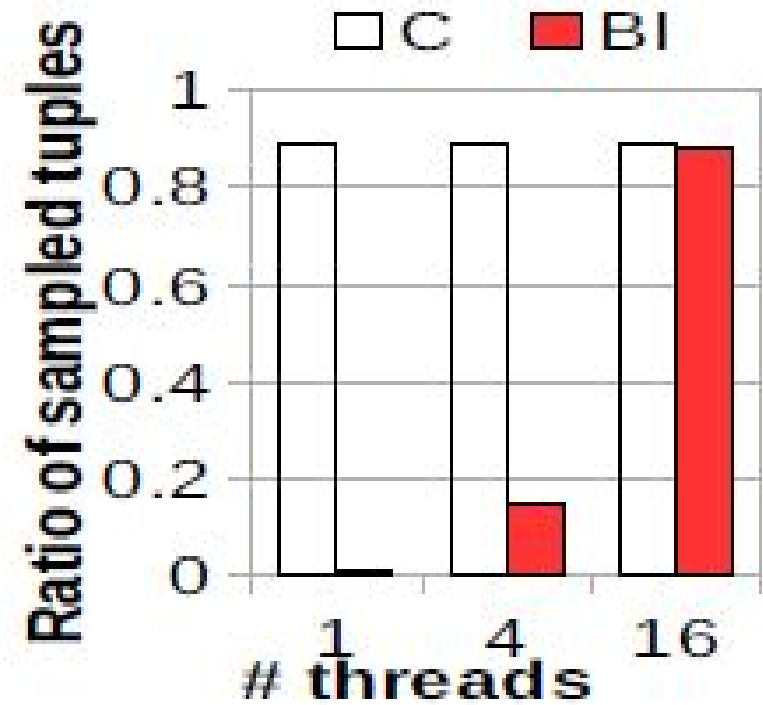
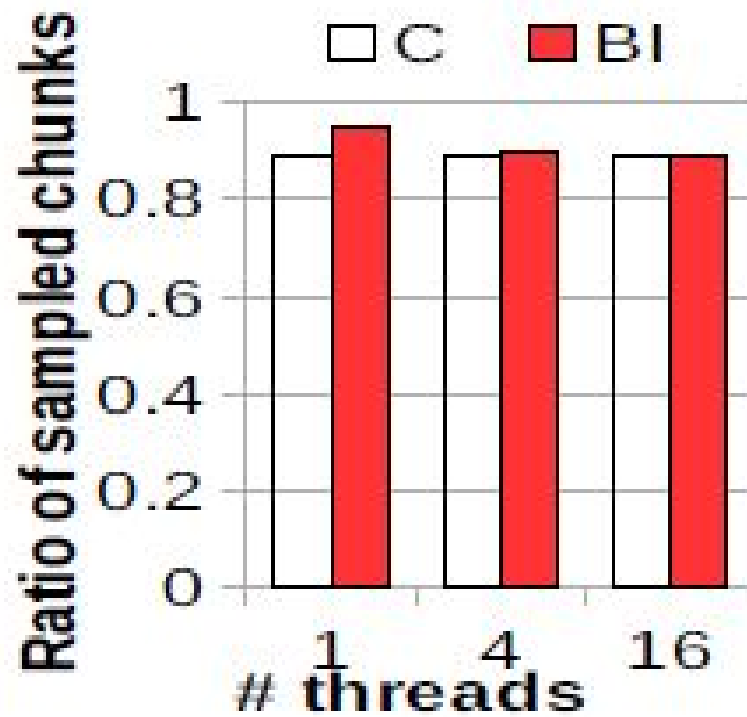
# Query Execution Time



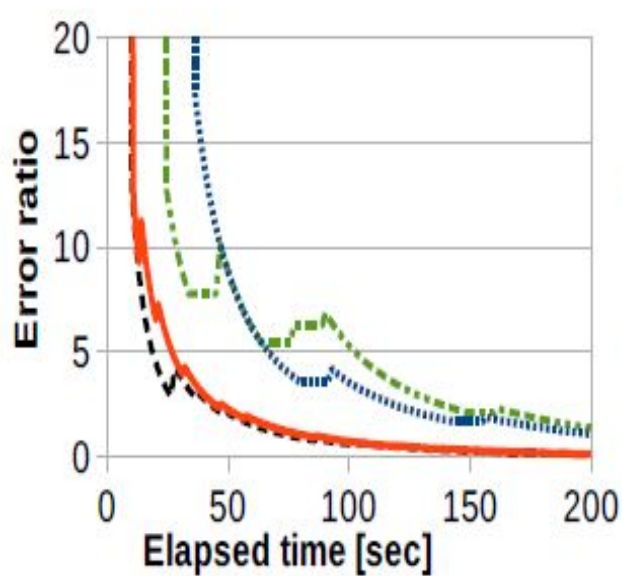


# Sample Size

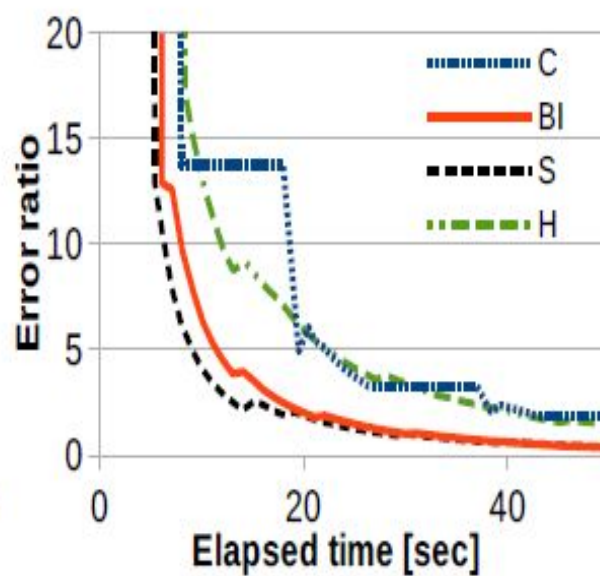
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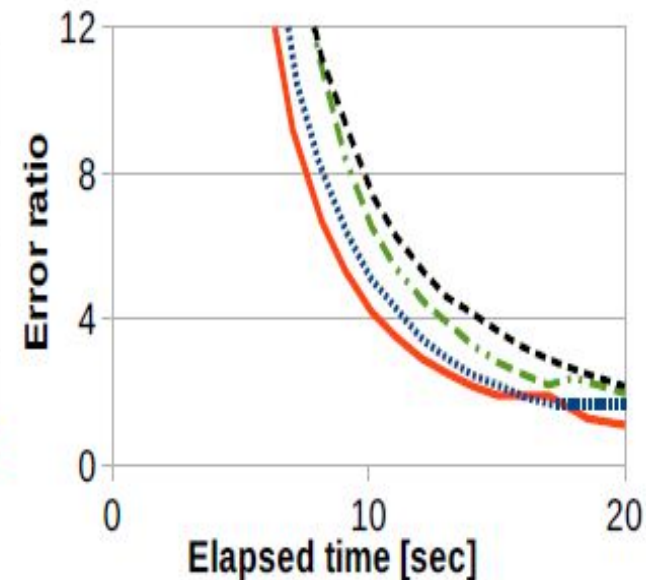
# Parallel Sampling Comparison



1 thread

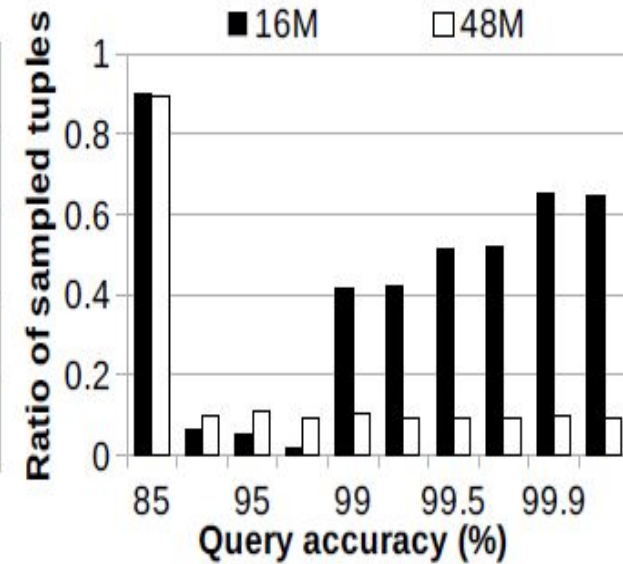
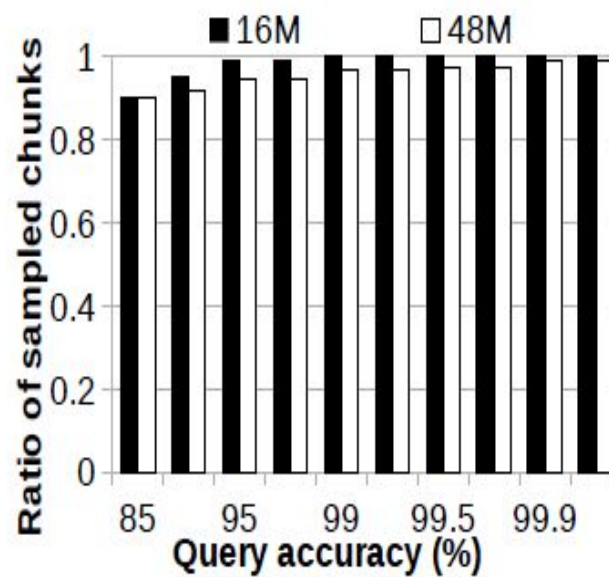
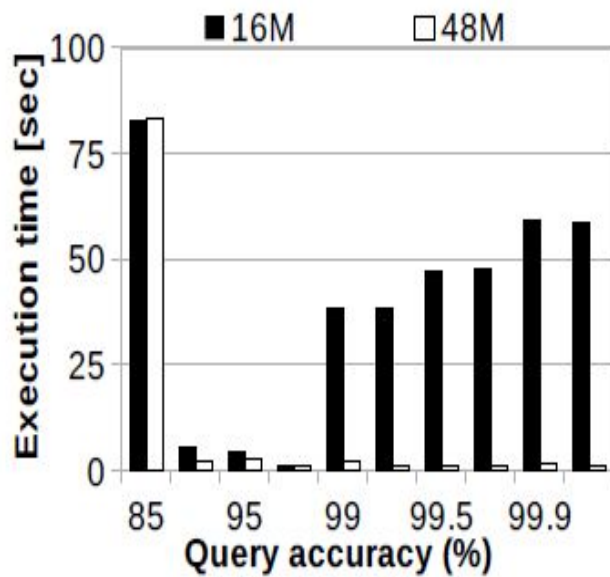


4 threads

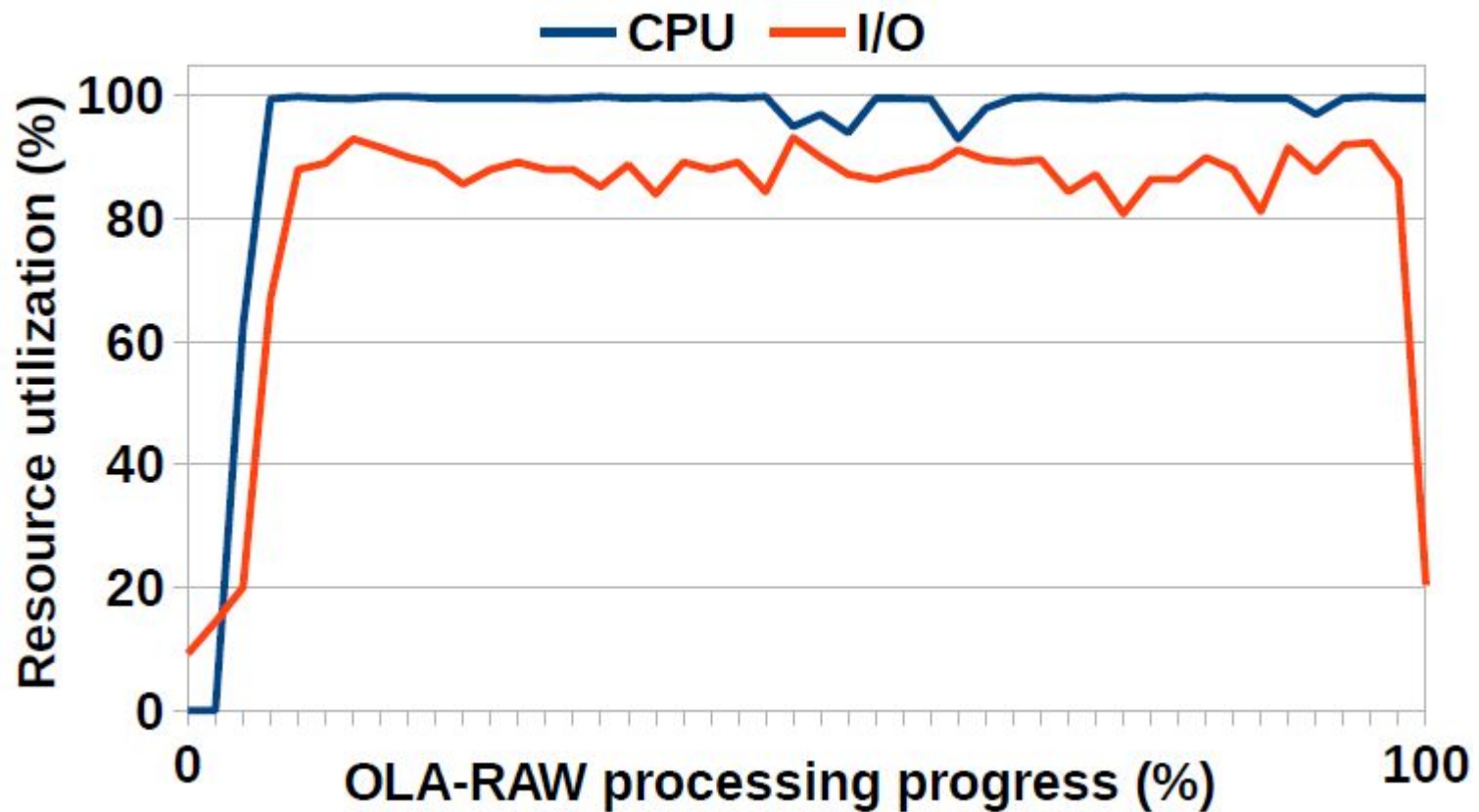


16 threads

# Sample Synopsis



# Resource Utilization

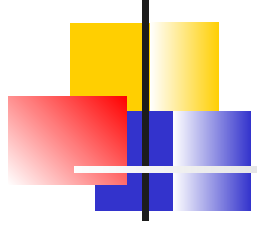




## Conclusions

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- OLA-RAW is a novel resource-aware bi-level sampling method for parallel on-line aggregation over raw data
- OLA-RAW is an efficient scheme for data exploration that avoids unnecessary work



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**Thank you!**

**Questions?**