

# Efficient Searchable Encryption Through Compression



Ioannis Demertzis  
University of Maryland  
yannis@umd.edu

Rajdeep Talapatra  
University of Maryland  
rajdeep@umd.edu

Charalampos Papamanthou  
University of Maryland  
cpap@umd.edu



## Motivation

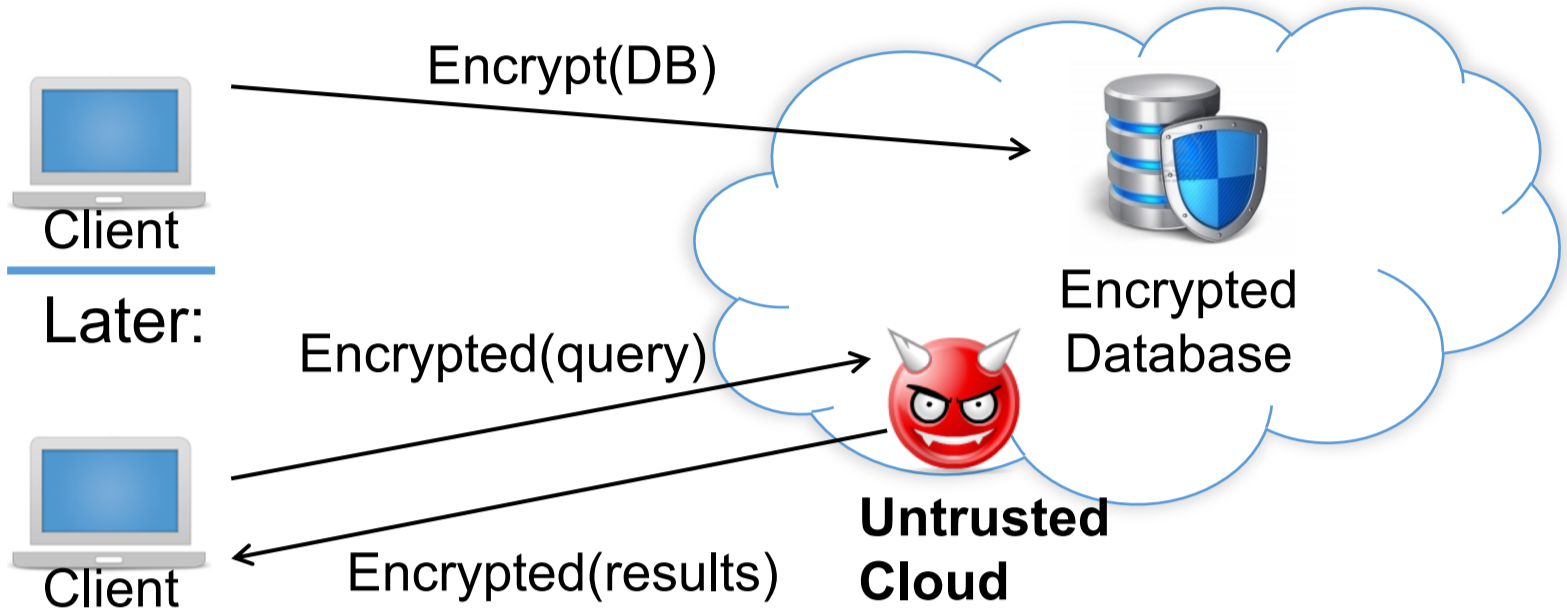
### Cloud Computing:

- Offers: **near infinite scalability**, **easy & ubiquitous** access ...
- Serious **security** and **privacy concerns** when outsourcing and querying on private company or personal data

### General Data Protection Regulation (GDPR) Era:

- GDPR is more flexible when data are encrypted
- Standard Encryption **limits** usability, efficiency, functionality

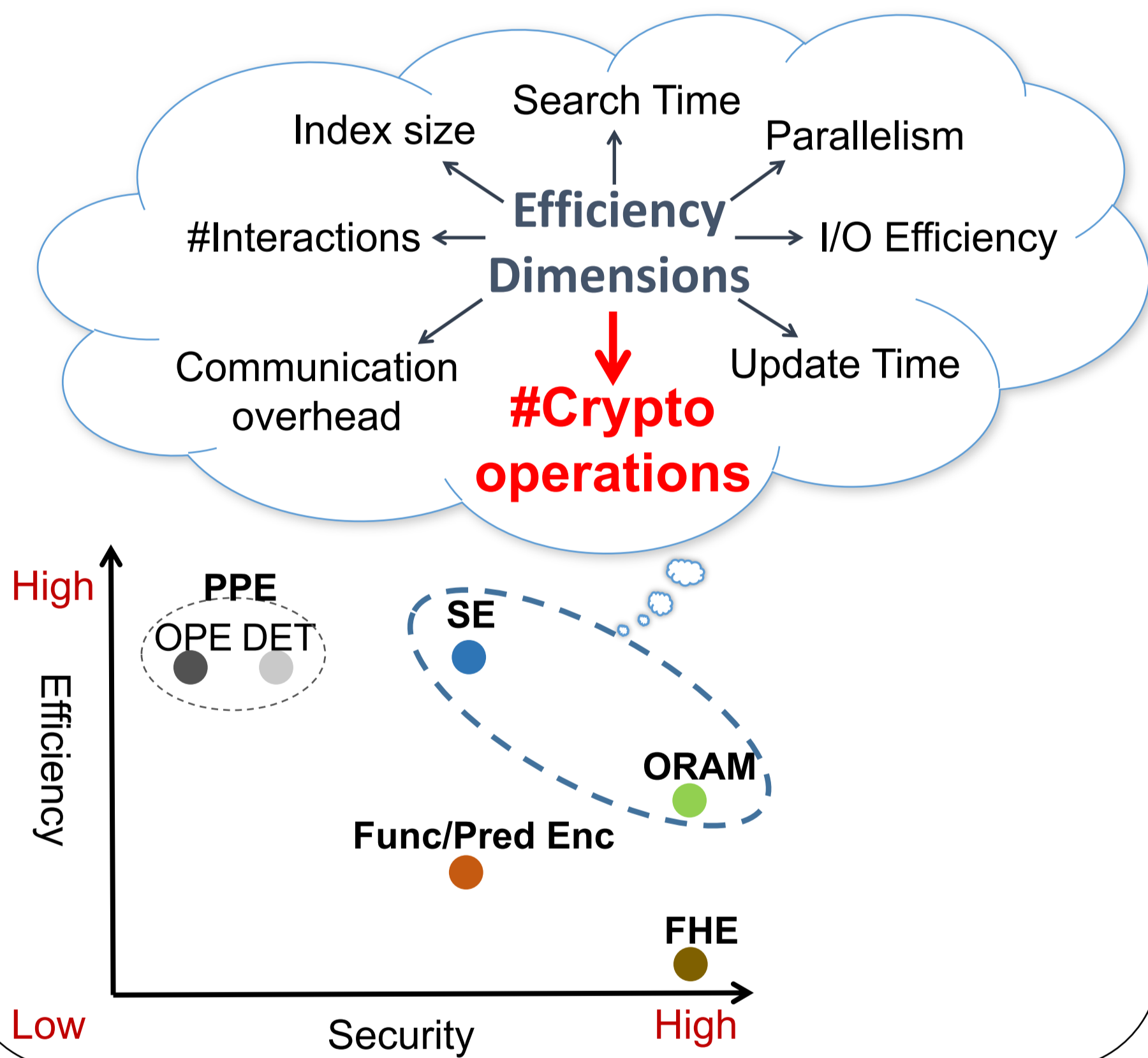
### Solution: Searchable Encryption (SE)/Oblivious SE (OSE)



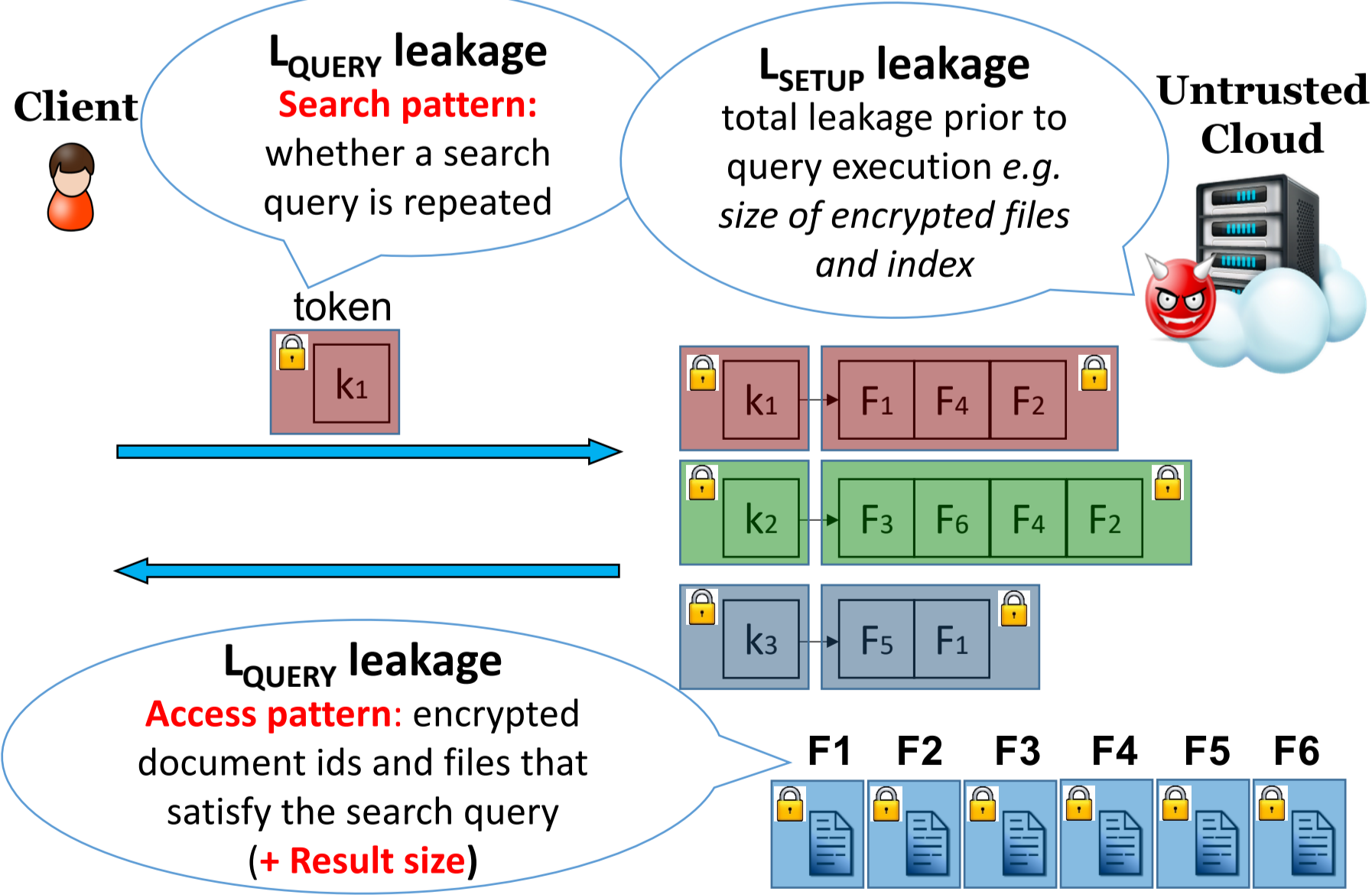
### Our contribution:

- New efficiency dimension for SE: **#cryptographic operations** to retrieve query results
- Use any set of lossless **compression** algorithms and any **SE/OSE** schemes as black-boxes to improve their search efficiency without affecting their security
- Keyword search time savings: up to **188x** (Enron dataset)
- Database search: up to **62x** (location description attribute), **170-203x** (binary attributes).
- Our OSE approach reduces the index search time to access **1M** tuples from **21 hours** to **20 minutes**

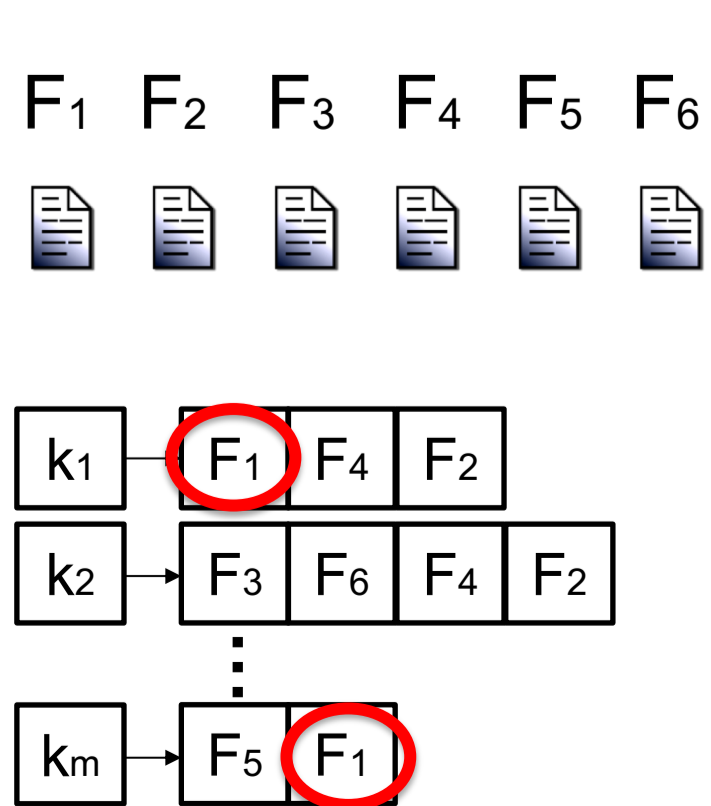
## Related Work



## SE/OSE



### Keyword Search



### Database Search

T <sub>1</sub>	John	Smith	CMU	27	\$3,000
T <sub>2</sub>	Alice	Lu	UCLA	28	\$4,000
...	...	...	...	...	...
T <sub>N</sub>	Bruce	William	UMD	30	\$2,000

27	T <sub>1</sub>	T <sub>4</sub>	T <sub>20</sub>	
28	T <sub>2</sub>	T <sub>6</sub>	T <sub>12</sub>	T <sub>50</sub>
...	...	...	...	
30	T <sub>13</sub>	T <sub>N</sub>		

SE → L<sub>QUERY</sub>: **Access + Search**    OSE → L<sub>QUERY</sub>: **Result size + Search**  
 OSE → L<sub>QUERY</sub>: **Result size**

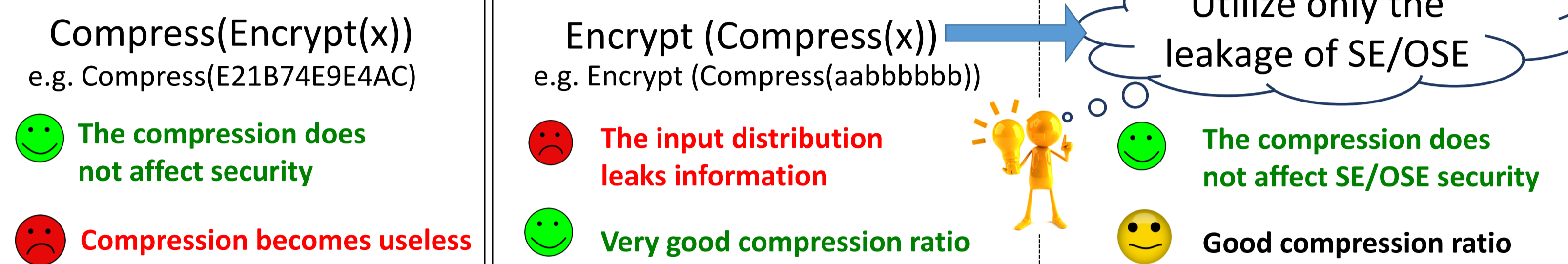
## Our Approach

General Idea of microSE/microOSE

microSE/microOSE with leakage  $\Lambda'$ , where  $\Lambda' \leq \Lambda$



## Combining Encryption with Compression



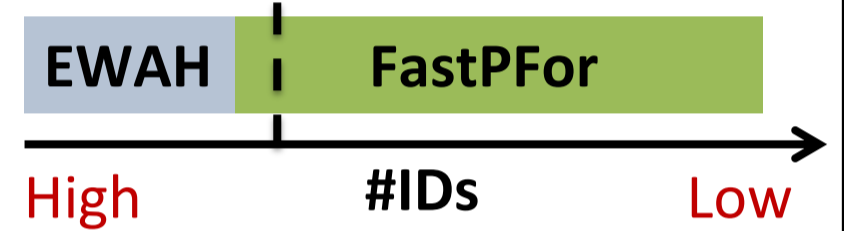
## Construction's Highlights

Prior SE schemes: each identifier is a random string of  $k$  bits → **Compression of a sequence of random strings is useless**

**Solution:** Assign an id chosen uniformly at random from the range  $[0, N-1]$  ( $N = \#tuples$ ) → **Achieves good compression** + **The compression does not affect SE's security**

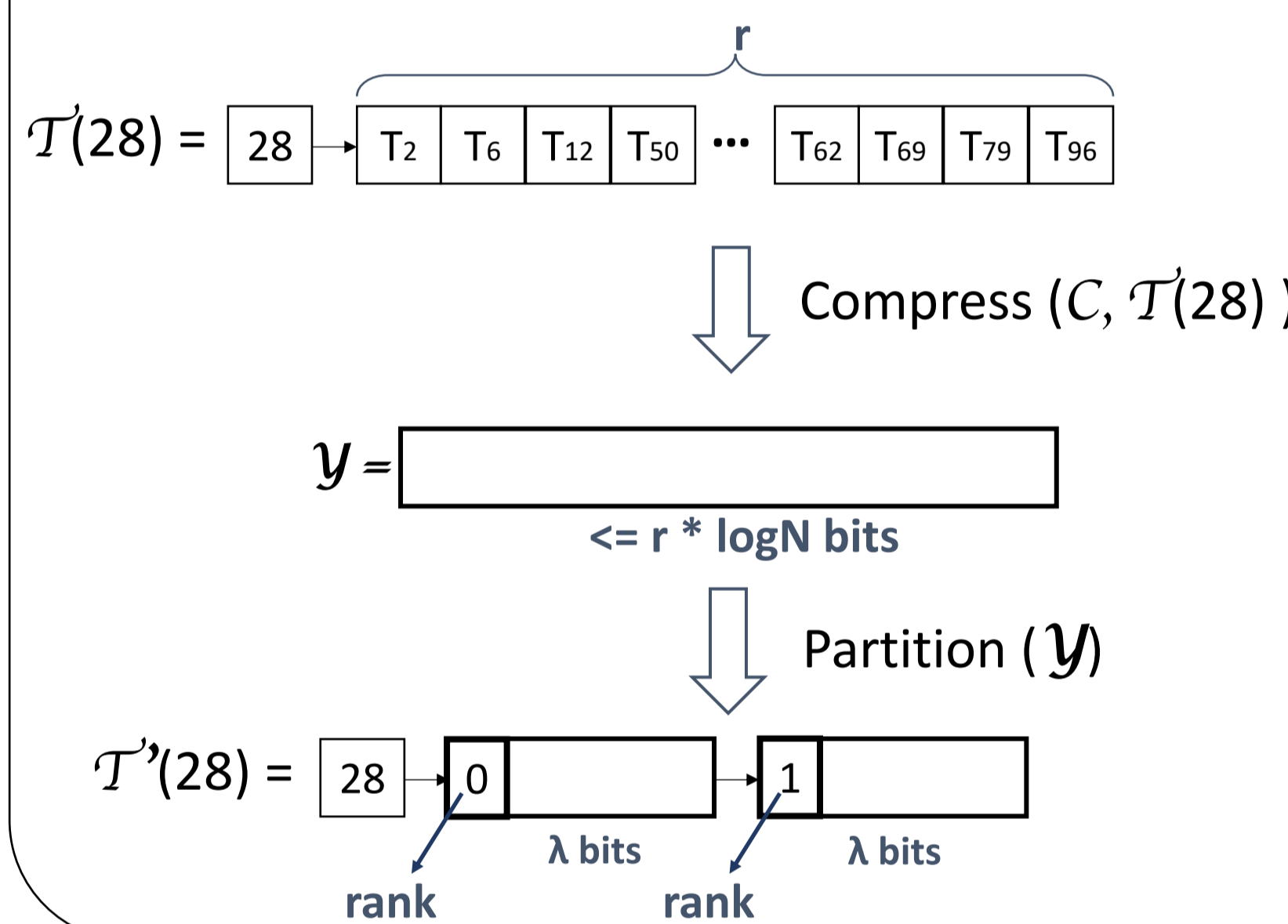
**Problem:** Compress the sequence of identifiers that are uniformly distributed in the range  $[0, N-1]$

- Chosen compression algorithms based on [WLP+17]:
  - EWAH
  - FastPFor



[WLP+17]: "Jianguo Wang, Chunbin Lin, Yannis Papakonstantinou, and Steven Swanson. An Experimental Study of Bitmap Compression vs. Inverted List Compression. SIGMOD 2017."

### microSE



### microSE/microOSE

**microSE:** Assume that  $|T'(28)| = |T'(32)|$  then  $|T'(28)| \geq |T'(30)|$

By leaking the **search pattern** the adversary knows that  $T'(28)$  and  $T'(30)$  are different queries

**microOSE:**  $|T'(28)|$  has to be equal to  $|T'(30)|$  since the **search pattern** is **not** leaked !!!

- Solution 1:** Compress\*(C, T'(28))
- Solution 2:** Store the overflowed lists in a local stash on the client side

## Experiments

### Setup

- Our SE Scheme (**microSE**) + used PiBas [CJS+14] as black-box SE → **microSE(PiBas)**
- Our OSE Scheme (**microOSE+Solution2\***) + used PathORAM [SDS+12] as black-box → **microOSE(PathORAM)**
- Java implementation using JavaX.crypto and Bouncy Castle
- 64bit machine with Intel Xeon E5-2676v3 with 64GB RAM

\*Local stash for solution 2 was always smaller than the stash of PathORAM

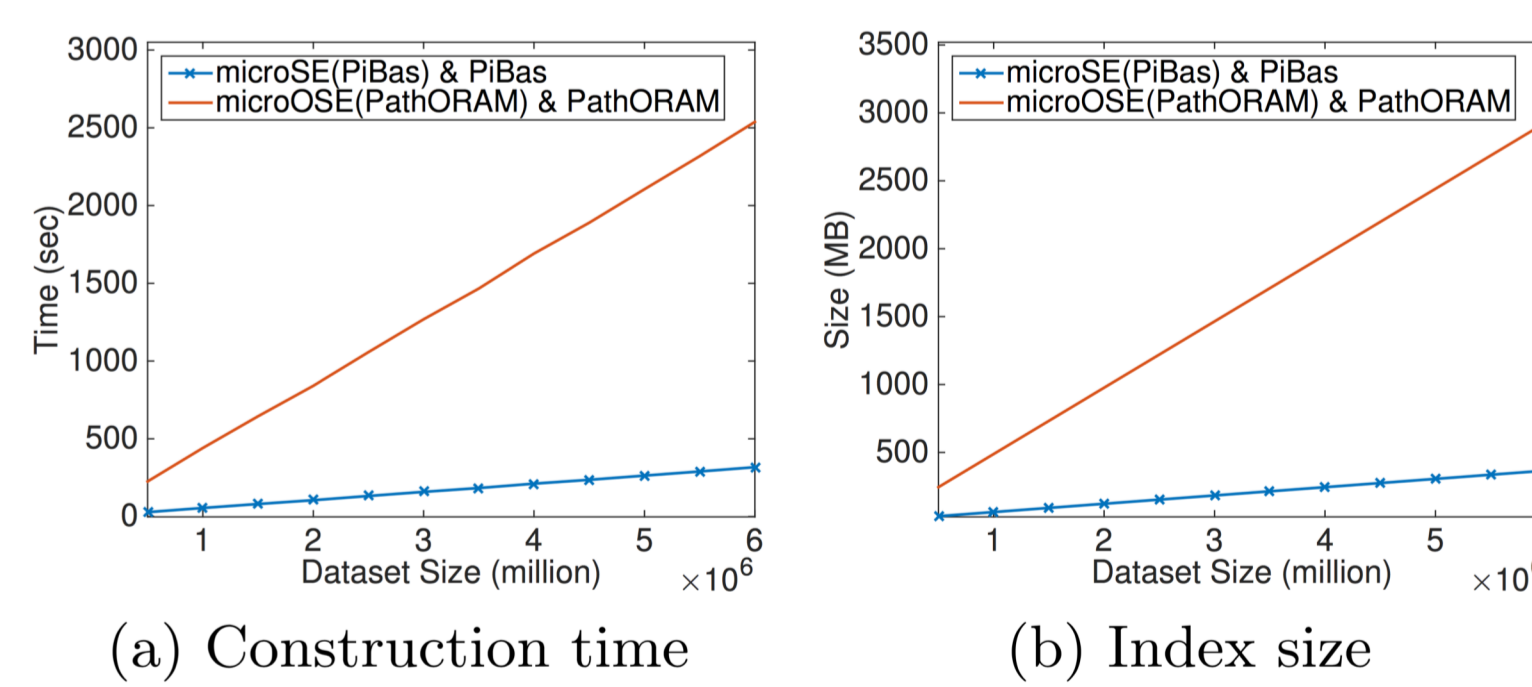
### Database Search

- Chicago crime incidents dataset with 6,123,276 records
  - 22 attributes with different distributions: ID, Case Number, Date, Block, ICR, Primary Type, Description, Location Description, Arrest, Domestic, Beat, District, Ward, Community Area, FBI Code, X Coordinate, Y Coordinate, Year, Updated On, Latitude, Longitude, Location

### Keyword Search

- Enron email dataset with 30,109 emails of 150 employees of the Enron corporation sent between 2000-2002

### Setup Costs

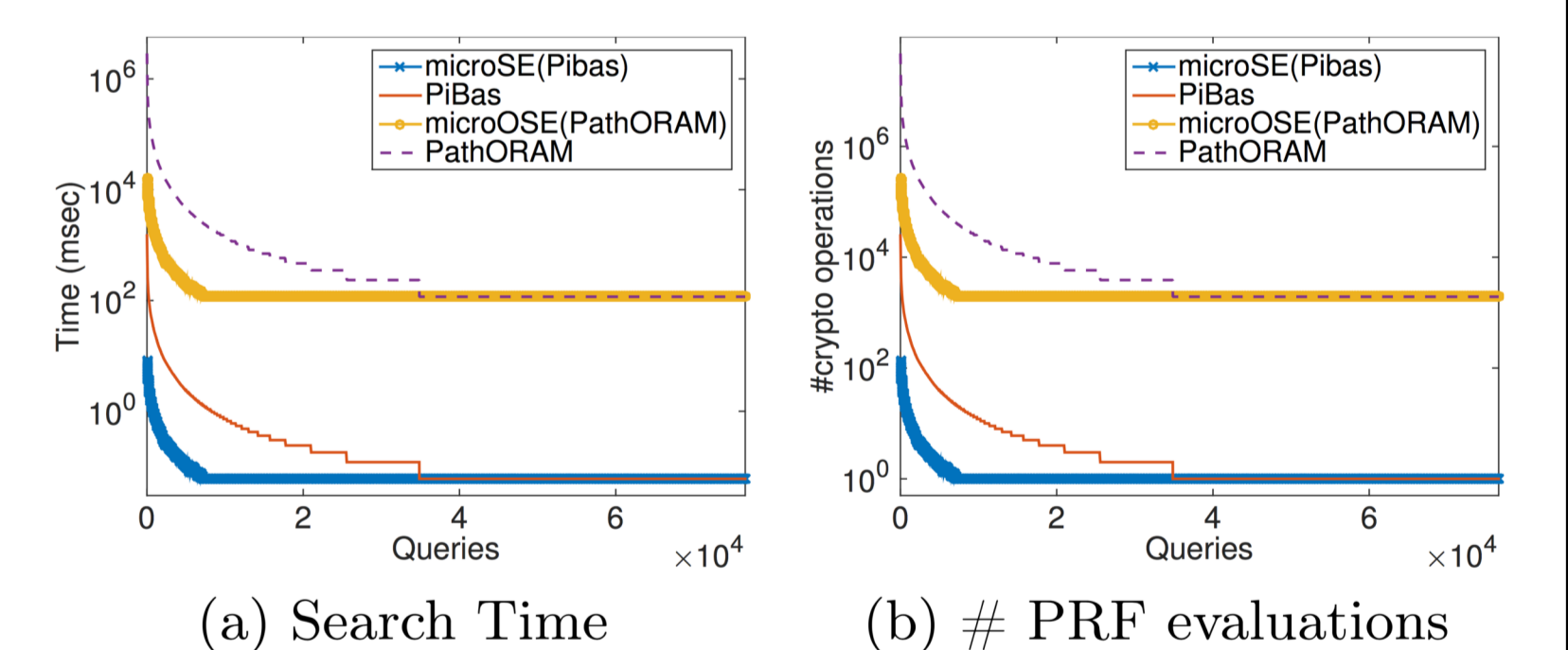


(a) Construction time

(b) Index size

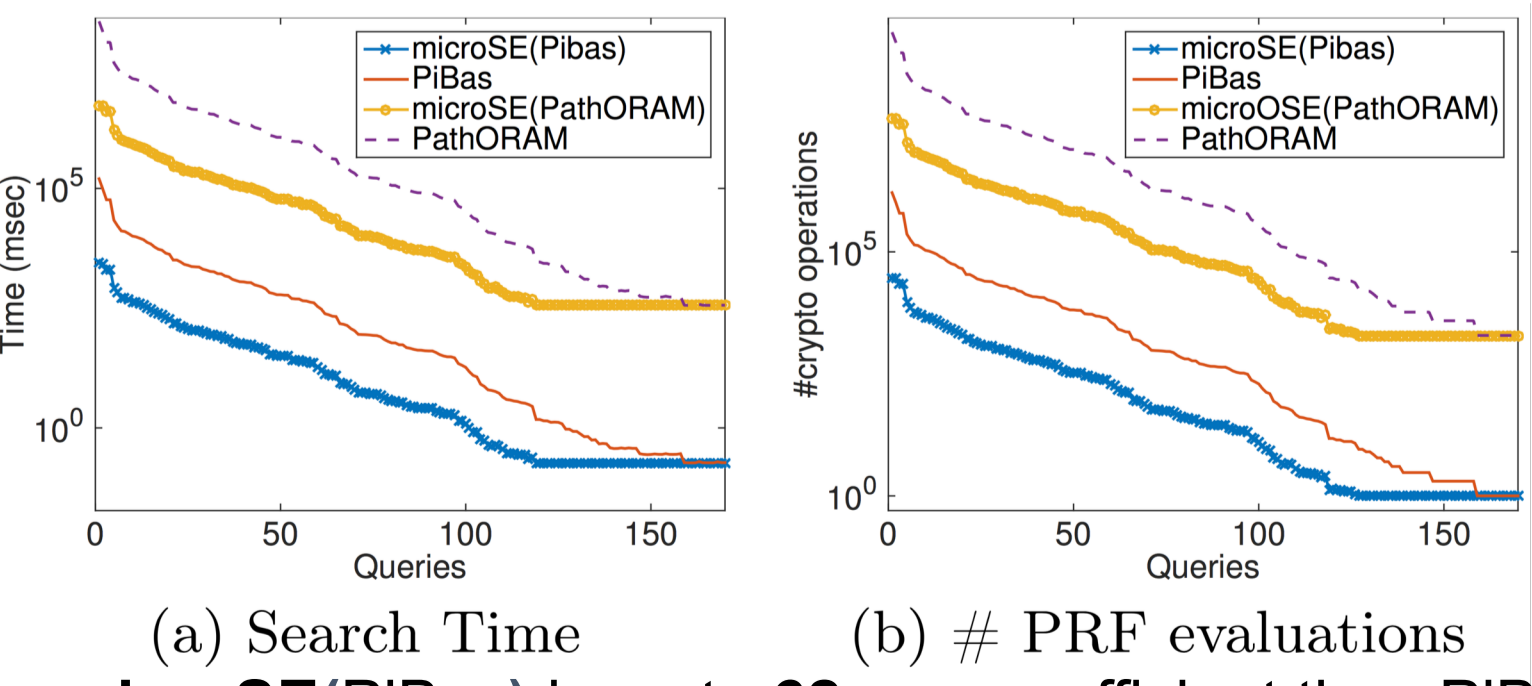
Crime Dataset

### Search Time – Keyword Search



• **microSE(PiBas) & microOSE(PathORAM)** are up to **188x** more efficient than PiBas and PathORAM

### Search Time – Database Search

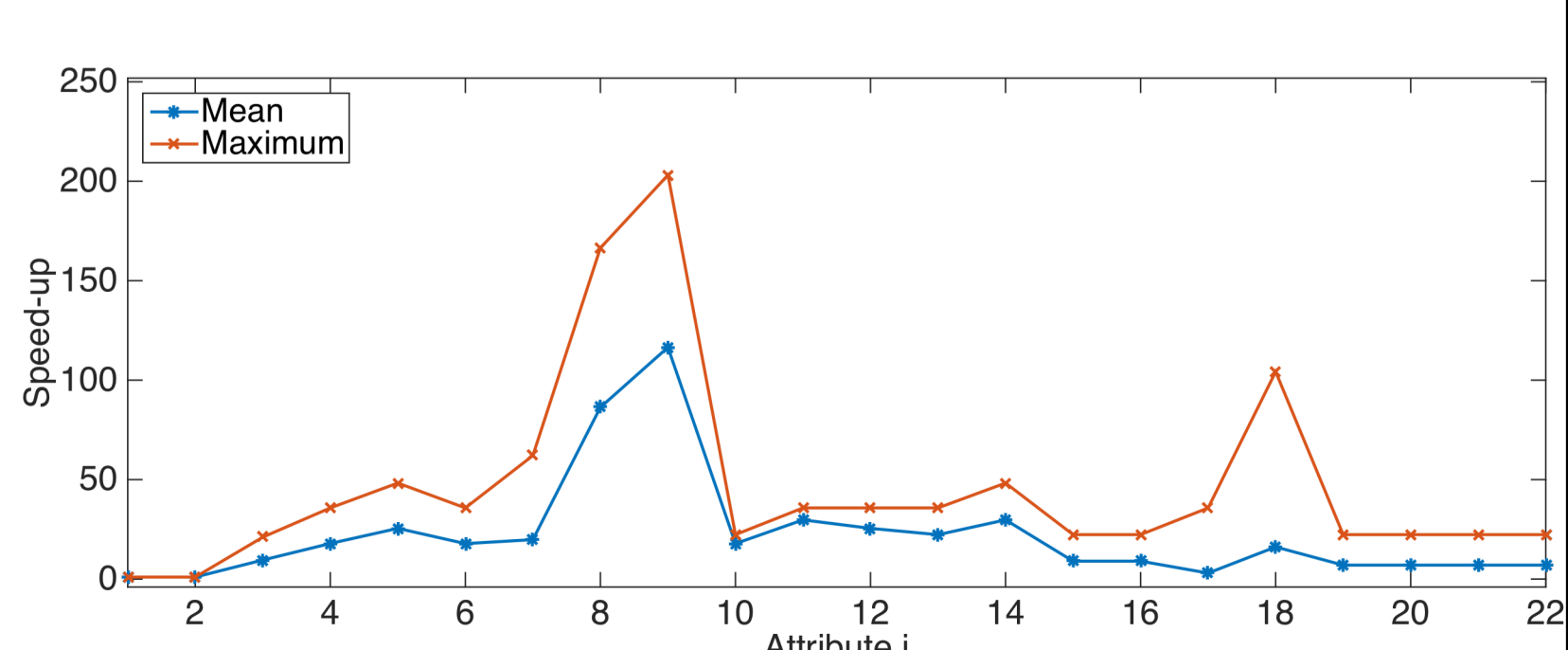


(a) Search Time

(b) # PRF evaluations

• **microSE(PiBas)** is up to **62x** more efficient than PiBas  
 • **microOSE(PathORAM)** accesses 1M tuples in **20 minutes** vs **21 hours** of PathORAM

### Search Time – Database Search



Crime Dataset – All 22 attributes