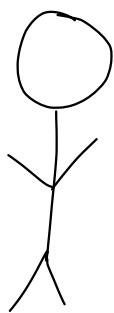


Private Information Retrieval

- Motivations
- 2-Server PIR
- Single Server PIR
- Modern Results

Motivation



Query: Peanut Allergies
Response: Peanut Webpage

"Search"

Database of Medical Conditions
"Advertisement Platform"

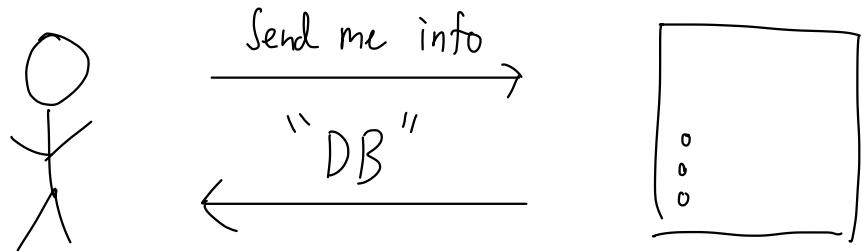
- Database server learns your query and can sell this information to Ad networks or health providers.
- Could lead to more targeted ads, higher premiums, targeted pricing

No privacy!!!

- If only we could query databases without the server learning any information about our query!

We can with Private Information Retrieval!

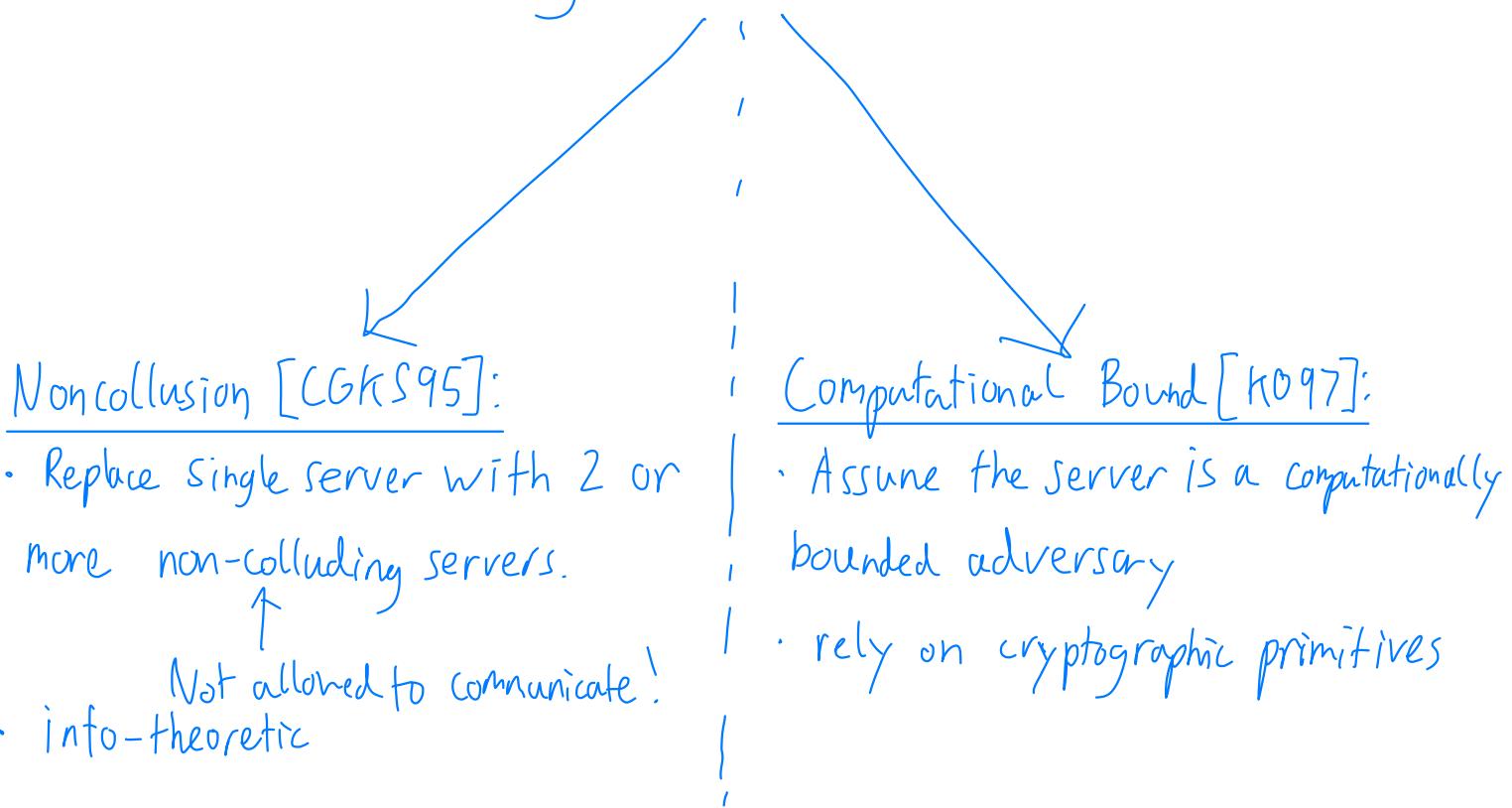
First (trivial) Construction



- The server could just send the whole database!
- Issue: Communication is linear!
- Question: Can we obtain sublinear communication without the server learning our query?

Not Quite! — if $\text{comm} < |\text{DB}|$, then we lose info about some entry j ; an unbounded server could distinguish which entry!

Getting around the issue!

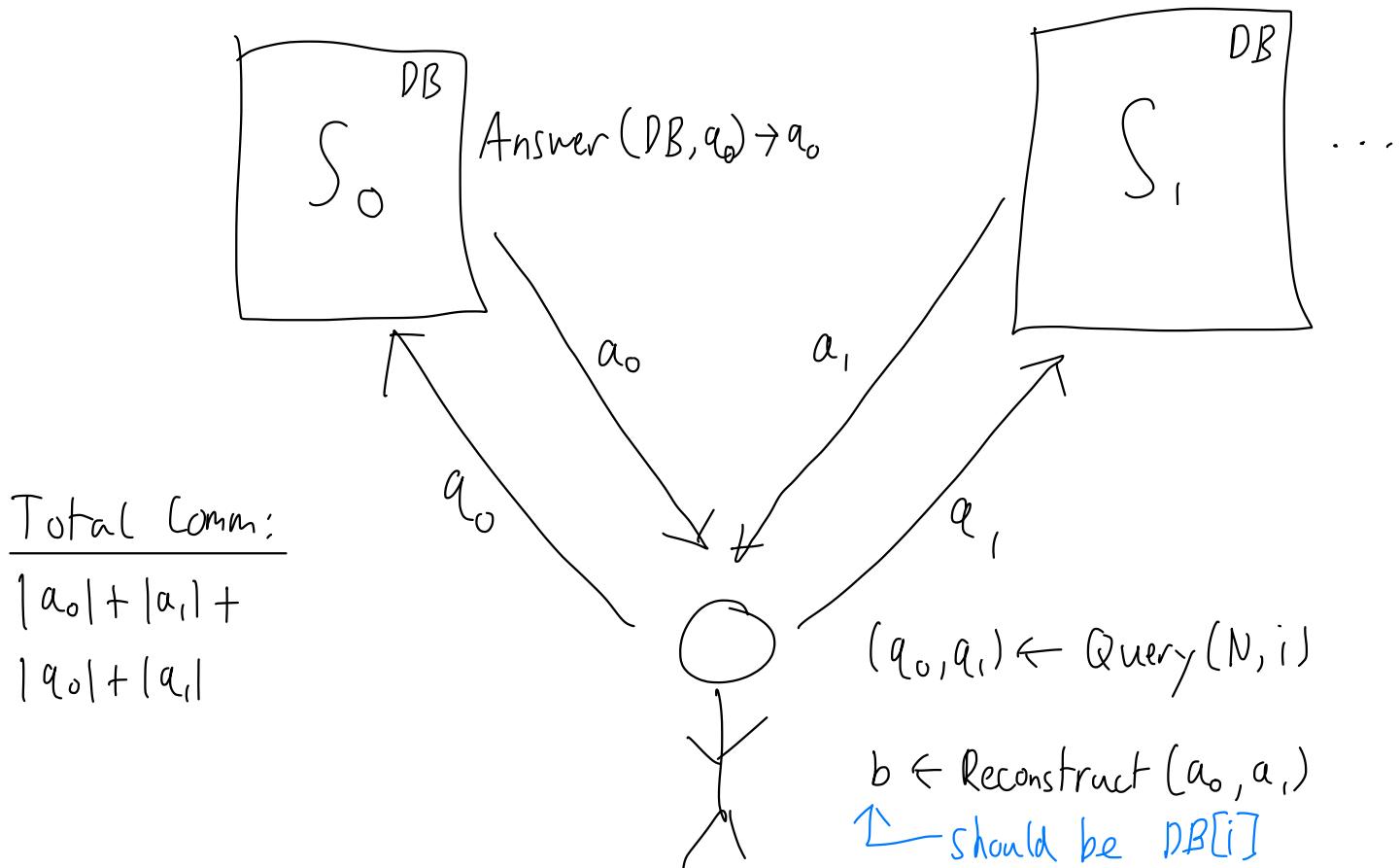


TWO Server PIR

- We'll refer to the servers as S_0, S_1 , and treat the database $DB \in \{0,1\}^N$ (i.e. Entry $j \in [N] = DB[j] \in \{0,1\}$)

A two server PIR scheme is a tuple of eff algs:

- Query($N, i \rightarrow (q_0, q_1)$: $N = |DB|$ and i is the query index.
 q_0, q_1 are query strings sent to S_0, S_1 , respectively.
- Answer($DB, q \rightarrow a$: $DB \in \{0,1\}^N$, q is the query string from the client and a is the server response.
- Reconstruct($a_0, a_1 \rightarrow b$: a_0, a_1 are responses from S_0, S_1 , and b is the reconstructed entry.



Correctness:

$\forall N \in \mathbb{N}, i \in [N], DB \in \{0,1\}^N,$

$$\Pr \left[b = DB[i] \quad \begin{array}{l} (q_0, q_1) \in \text{Query}(N, i) \\ a_0 \leftarrow \text{Answer}(DB, q_0) \\ a_1 \leftarrow \text{Answer}(DB, q_1) \\ b \leftarrow \text{Reconstruct}(a_0, a_1) \end{array} \right] = 1$$

Security:

\forall polybounded $N \in \mathbb{N}, DB \in \{0,1\}^N$, pairs $(i, j) \in [N]^2$, $b \in \{0,1\}$,
the following distributions are indistinguishable,

$$\left\{ \text{Query}(DB, i)[b] \right\} \approx \left\{ \text{Query}(DB, j)[b] \right\}$$

↑
could be perfect or computational

Note: b selects only one of the server queries.

Warmup: A trivial construction!

ith basis vector
↓

$\text{Query}(N, i)$: Sample $q_0 \in \{0,1\}^N$ and $q_1 \leftarrow q_0 \oplus e_i$. Output (q_0, q_1)

$\text{Answer}(DB, q)$: Output $a \leftarrow \langle DB, q \rangle \leftarrow$ inner product

$\text{Reconstruct}(a_0, a_1)$: Output $b \leftarrow a_0 \oplus a_1$.

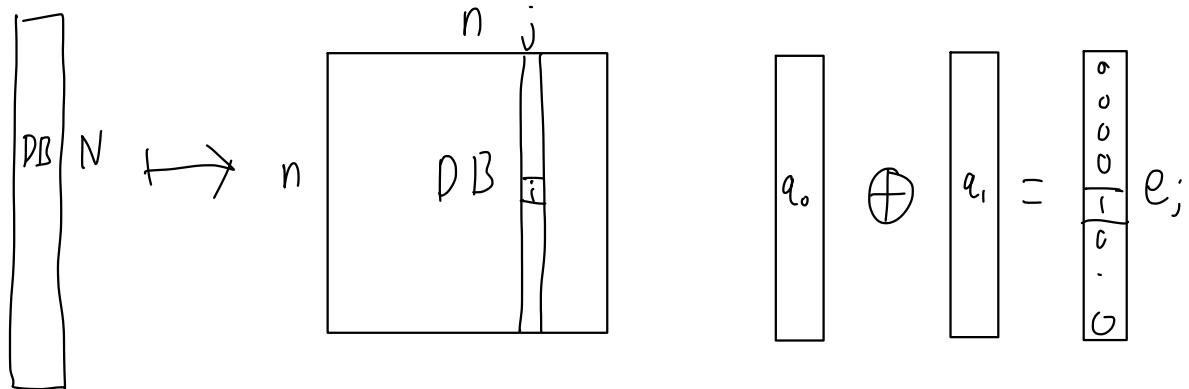
$$\begin{aligned}
 \text{Correctness: } a_0 \oplus a_i &= \langle DB, a_0 \rangle \oplus \langle DB, a_i \rangle \\
 &= \langle DB, a_0 \oplus a_i \rangle \\
 &= \langle DB, e_i \rangle = DB[i]
 \end{aligned}$$

Security: Separately, a_0 and a_i are uniformly random N bit vectors. Thus, the distributions will be identical!

Issue: Communication is still linear! (swapped direction)
 (upload vs download)
 with just sending DB

CGKS Construction

- Assume $N = n^2$ is a perfect square for simplicity.



Query($N, (i, j)$): Sample $a_0 \in \{0, 1\}^n$ and $a_i \leftarrow a_0 \oplus e_j$. Output (a_0, a_i)

Answer(DB, q): Compute vector $a \leftarrow DB \cdot q$. Output a .

Reconstruct(a_0, a_i): Output $b \leftarrow (a_0 \oplus a_i)[i]$

Matrix $n \times n$
 vector $n \times 1$ product

$$\begin{aligned}
 \text{Correctness: } a_0 \oplus a_1 &= DB \cdot q_0 \oplus DB \cdot q_1 \\
 &= DB \cdot (q_0 \oplus q_1) \\
 &= DB \cdot e_j = DB_j \text{ "jth col of } DB
 \end{aligned}$$

$$DB_j[i] = DB[i, j].$$

Security: Similarly, both q_0 and q_1 are uniformly random vectors.

Communication: $O(\sqrt{N})$ bits! (upload = download)

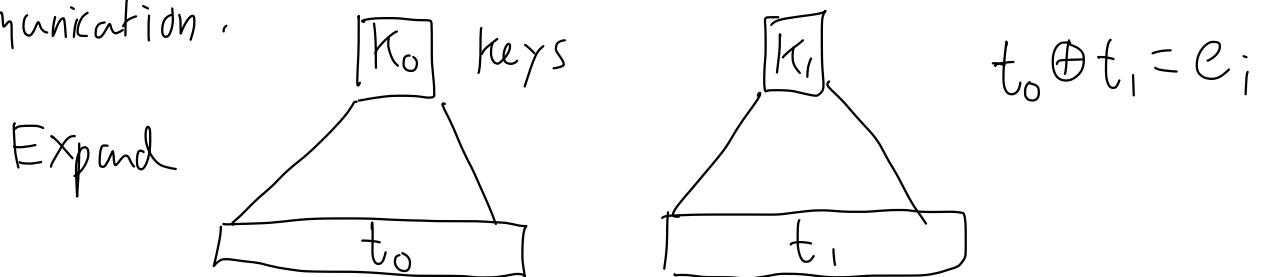
Can we do better in the two server setting?

- When considering unbounded servers (info-theoretic), [DG15] obtain $\tilde{O}(N^{\sqrt{\log \log N / \log N}}) = \tilde{O}(N^{o(1)})$ comm from locally-decodeable codes.



[BGIL15]

- In computationally bounded setting, Distributed Point functions (constructed from PRGs) gives us a 2-server PIR with $O_\lambda(\log N)$ communication.



Computational Single Server PIR

- [K097] leverages linearly-homomorphic encryption (El-gamal, etc)

$\forall k \in K, \forall m_0, m_1 \in M,$

$$\text{Enc}(k, m_0) + \text{Enc}(k, m_1) = \text{Enc}(k, m_0 + m_1)$$

Query($N, (i, j)$): $k \leftarrow K$. Output query

$$q \leftarrow (\text{Enc}(k, e_j[1]), \text{Enc}(k, e_j[2]), \dots, \text{Enc}(k, e_j[n]))$$

vector of ciphertexts

Answer(DB, q): Output $a \leftarrow DB \cdot q$.

$$\begin{array}{c}
 DB, \quad DB_1, \quad DB_2, \dots \\
 \boxed{DB} \quad \left[\begin{matrix} c_1 \\ \vdots \\ c_n \end{matrix} \right] = \left[\begin{matrix} 1 & & & \\ 0 & 1 & & \\ \vdots & \vdots & \ddots & \\ 0 & 0 & \dots & 1 \end{matrix} \right] c_1 + \left[\begin{matrix} 0 & & & \\ 1 & 0 & & \\ \vdots & \vdots & \ddots & \\ 0 & 0 & \dots & 0 \end{matrix} \right] c_2 + \dots \\
 \xrightarrow{\text{Linear Homomorphism}} \left[\begin{matrix} c_1 + c_2 + \dots \\ c_1 + \dots \\ 0 + c_2 + \dots \\ \vdots \end{matrix} \right] \leq n \text{ additions}
 \end{array}$$

$\text{Reconstruct}(k, a)$: Decrypt ciphertexts,

$$u \leftarrow (\text{Dec}(k, a_1), \text{Dec}(k, a_2), \dots, \text{Dec}(k, a_n))$$

Output $u[i]$.

Correctness: Follows from linear homomorphism of Enc ,

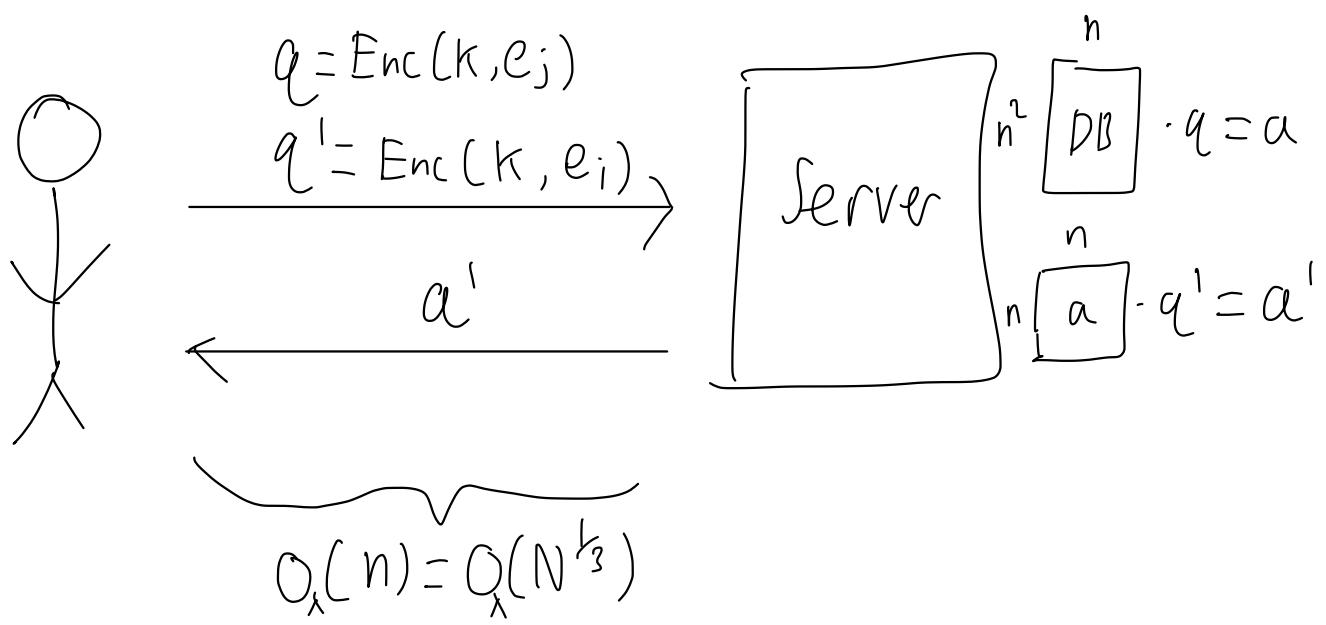
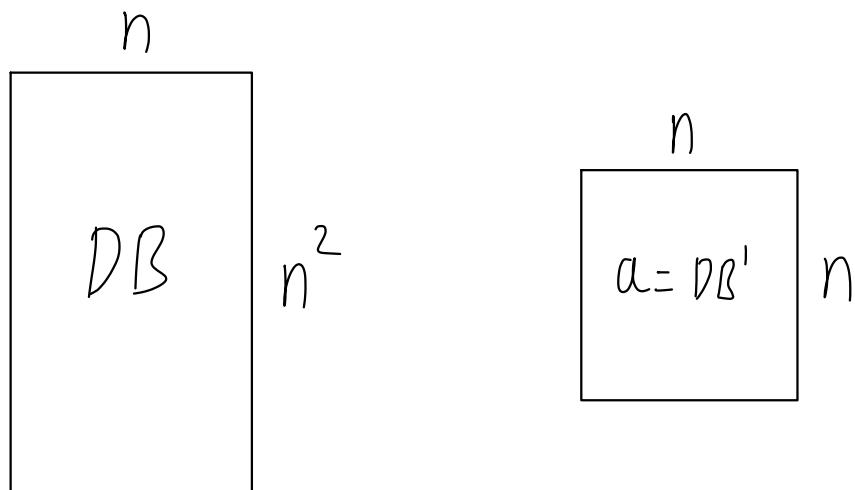
$$a = DB \cdot q = \begin{bmatrix} \sum_t DB_{1,t} \cdot q_t \\ \vdots \\ \vdots \\ \sum_t DB_{j,t} \cdot q_t \end{bmatrix} = \begin{bmatrix} \text{Enc}(k, DB_{1,j}) \\ \text{Enc}(k, DB_{2,j}) \\ \vdots \\ \vdots \\ \text{Enc}(k, DB_{n,j}) \end{bmatrix}$$

Security: Follows from semantic security of Enc , notably that the Enc of an e_j should be computationally indistinguishable from $e_{j' \neq j}$.

Comm: | Ciphertext | = $O(\lambda)$ so overall comm is $O(\lambda n) = O(\lambda \sqrt{N})$

Can we do better?

Idea: Recursion! Notice the client only needs 1 ciphertext, so why not treat the server response as another database DB' . $N = n^3$



Can we recurse further?

Issue: $|Entry| = 1 \text{ bit}$



$|Ciphertext| = \lambda \text{ bit}$

So, everytime we recurse we blow up communication by a factor of λ . $|a| = \lambda^2 n^2$, $|a'| = \lambda^2 n$

Solution: Leverage a Lin-hom scheme which

- good rate: $\frac{|m|}{|c|}$ is closer to 1
- large message spaces

[Damgård-Jurik]

Best Comp Single Server: $\text{Poly} \log(n)$ comm

[CMS99, Lip05]

from QR, DHT, LWE
assumptions