

# Applying MPC at Scale

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# Papers

- ❑ ANONIZE: A Large-Scale Anonymous Survey System
- ❑ Blind Seer: A Scalable Private DBMS

# ANONIZE GOAL: Anonymity vs Authenticity

- Scenario: Medical Survey
- Authenticity: Only Legitimate users can give the survey
- Anonymity: No link between user and survey data for honest feedback

# Issues

- Scenario: College Course Review
  - Side channel attack can be used to get the order in which students completed the survey.
  - Third parties may not be secure. Eg. Cornell incident.
  - Cryptographic voting technique – current techniques:
    - Users check-out a single use “token” from server – not related to user.
    - User completes the survey using the token.
    - TIME attack – clear link.

# Literature: PRF (Wiki)

- In cryptography, a pseudorandom function family, abbr. PRF, is a collection of efficiently-computable functions which emulate a random oracle in the following way: no efficient algorithm can distinguish (with significant advantage) between a function chosen randomly from the PRF family and a random oracle (a function whose outputs are fixed completely at random).
- The guarantee of a PRF is that all its outputs appear random, regardless of how the corresponding inputs were chosen, as long as the function was drawn at random from the PRF family

# Literature: NIZK (Wiki)

- Non-interactive zero-knowledge (NIZK) proof systems are fundamental cryptographic primitives used in many constructions of various cryptographic protocols.

# Definition of Ad-hoc survey system

- An *ad-hoc survey scheme*  $\Gamma$  is an 8-tuple of PPT algorithms and interactive PPTs ( $\text{GenRA}, \text{GenSA}, \text{Reg}^{\text{RA}}, \text{Reg}^U, \text{GenSurvey}, \text{authorized}, \text{Submit}, \text{Check}$ ) where
- $\text{GenRA}(1^n)$  outputs a key-pair  $\text{pk}_{\text{RA}}, \text{sk}_{\text{RA}}$ .
- $\text{GenSA}(1^n)$  outputs a key-pair  $\text{pk}_{\text{SA}}, \text{sk}_{\text{SA}}$ .
- $\text{Reg}^{\text{RA}}(\text{sk}_{\text{RA}}, 1^n, \text{pk}_{\text{RA}}, id_i)$  is an interactive PPT that outputs either success or fail.
- $\text{Reg}^U(1^n, \text{pk}_{\text{RA}}, id)$  is an interactive PPT that outputs a bitstring  $\text{cred}_{id}$  or fail.
- $\text{GenSurvey}(1^n, \text{sid}, L, \text{sk}_{\text{SA}})$  outputs a bitstring  $\text{pk}_{\text{sid}}$ . Here  $\text{sid}$  is a unique arbitrary identifier and  $L$  is a description of the set of users eligible to participate in the survey.
- $\text{authorized}(\text{pk}_{\text{SA}}, \text{sid}, \text{pk}_{\text{sid}}, id)$  outputs either YES or NO.
- $\text{Submit}(1^n, \text{sid}, \text{pk}_{\text{sid}}, m, \text{cred}_{id})$  outputs  $\text{Sub} = (\text{tok}, m, \text{tokauth})$ .
- $\text{Check}(\text{pk}_{\text{RA}}, \text{pk}_{\text{SA}}, \text{sid}, \text{pk}_{\text{sid}}, \text{Sub})$  outputs either accept or fail.

# Summary

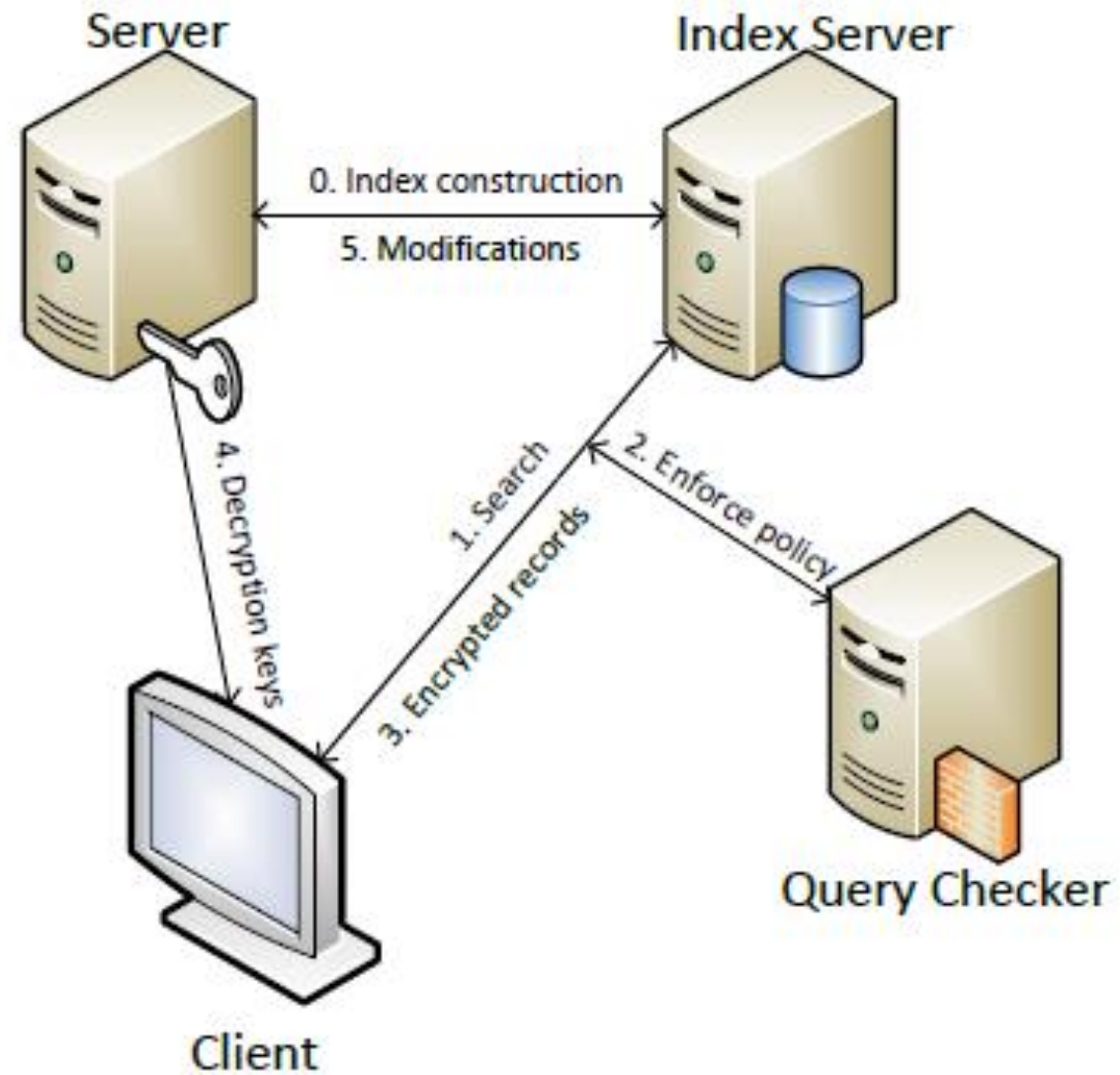
- Securely implements a million-person anonymous surveys using a single modern workstation.



# Blind Seer(BLoom filter INdex SEarch of Encrypted Results) Goal

- Almost all earlier applications considered limited queries.
- Provided weak guarantee of privacy.
- We have other solutions for privacy like FHE, ORAM and MPC which are way too expensive to be applied to practical uses (polynomial time).
- Therefore we try to solve : **Query privacy in secure DBMS** keeping scalability, functionality (efficient sublinear search for arbitrary Boolean queries) and usability in mind.

# Model



# Approach

- In large problem, we find small privacy critical parts & solve them securely.
- Large problem (encrypted search on large DB) is done by traversing an encrypted search tree.
- Input is secure; intermediate and output are in plain text revealing some information (to make search sublinear).
- Search tree traversal decision is made by Secure Function Evaluation (SFE) using Yao's GC protocol.
- Bloom filters are used to store keywords in each tree node. It gives constant time access and invariant traversal patterns.

## Literature:

**Notations.** Let  $[n] = \{1, \dots, n\}$ . For  $\ell$ -bit strings  $a$  and  $b$ , let  $a \vee b$  (resp.,  $a \wedge b$  and  $a \oplus b$ ) denote the bitwise-OR (resp. bitwise-AND and bitwise-XOR) of  $a$  and  $b$ . Let  $S = (i_1, i_2, \dots, i_\eta)$  be a sequence of integers. We define a projection of  $a \in \{0, 1\}^\ell$  on  $S$  as  $a \downarrow_S = a_{i_1} a_{i_2} \dots a_{i_\eta}$ ; for example, with  $S = (2, 4)$ , we have  $0101 \downarrow_S = 11$ . We also define a filtering of  $a = a_1 a_2 \dots a_\ell$  by  $S$  as  $a \uparrow_S = b_1 b_2 \dots b_\ell$  where  $b_j = a_j$  if  $j \in S$ , or  $b_j = 0$  otherwise; for example, with  $S = (2, 4)$ , we have  $1110 \uparrow_S = 0100$ .

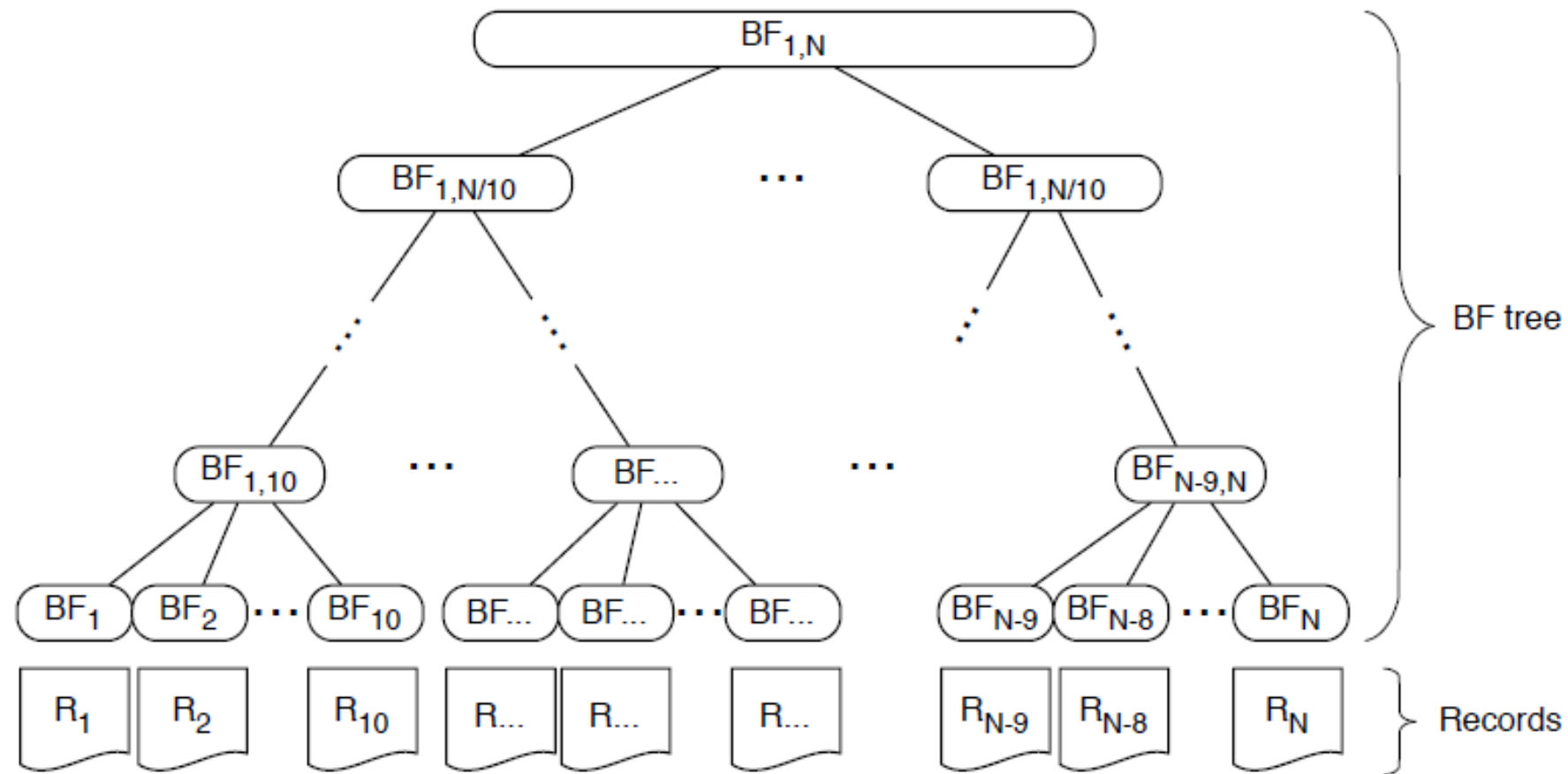
## Literature:

**Bloom filter (BF).** A Bloom filter is a well-known data structure that facilitates efficient search. The filter  $B$  is a string initialized with  $0^\ell$  and associated with a set of  $\eta$  different hash functions  $\mathcal{H} = \{h_i : \{0, 1\}^* \rightarrow [\ell]\}_{i=1}^\eta$ . For a keyword  $\alpha \in \{0, 1\}^*$ , let  $\mathcal{H}(\alpha)$  the sequence of the hash results of  $\alpha$ , i.e.,

$$\mathcal{H}(\alpha) = (h_1(\alpha), h_2(\alpha), \dots, h_\eta(\alpha)).$$

To add a keyword  $\alpha$  to the filter, the hash result  $\mathcal{H}(\alpha)$  is added to it, that is,  $B := B \vee (1^\ell \sharp \mathcal{H}(\alpha))$ .

# Bloom Filter Search tree



# Bloom Filter Search tree

- Key data structure enabling sublinear search is a BF search tree for the database records
- There is only one global search tree for the entire database
- In the search tree, each leaf is associated with each database record, and each node  $v$  is associated with a Bloom filter  $B_v$
- Filter  $B_v$  contains all the keywords from the (leaf) records that the node  $v$  have (as itself or as its descendants)
- Example, if a node contains a record that has Jeff in the fname field, a keyword = 'fname:Jeff' is inserted to  $B_v$

# Search using BF search tree

Search( $\alpha \wedge \beta, v$ ):

If the BF  $B_v$  contains  $\alpha$  and  $\beta$ , then

If  $v$  is a leaf, then output  $\{v\}$ .

Otherwise, output  $\bigcup_{c: \text{children of } v} \text{Search}(\alpha \wedge \beta, c)$ .

Otherwise, output  $\emptyset$ .

- $\alpha$  and  $\beta$  are the keywords to search
- Example: Search( $\alpha, \beta$ , root) will give all entries with the 2 keywords



# Preprocessing – Encrypting DB

- Server  $S$  shuffles and encrypts its records
- $S$  creates the BF search tree using the new records.
- Each node gets a Bloom filter  $B_v$  which is encrypted using PRF  $F$  with key  $k$ .
- $S$  sends the encrypted DB and search tree to  $IS$ .
- Client gets the PRF key  $k$  and the hash functions for the Bloom filter generation.

# Traversing Search tree privately

- Search procedure starts with transforming the query into Boolean circuit
- Client and IS compute the circuit via secure computation.
- If the circuit outputs true, it will again evaluate on that node recursively till it reaches a leaf node or else it terminates at that node.
- If the client gets the leaf node successfully, IS will send it  $(\psi(i), r_i, \tilde{R}_i)$ .
- Client sends  $\psi(i)$  to S and gets  $s'_{\psi(i)} = s_i r_i$ .
- C decrypts  $\tilde{R}_i$  using  $s_i$  and obtains the output record.

# Summary

- Instead of complete privacy we go for a weaker privacy as a trade off for greater efficiency.
- Performance of just 1.2x to 3x slowdown was achieved.
- The performance is due to the search tree traversal access patterns being revealed to the parties.

# References

- Blind Seer: A Scalable Private DBMS, SP14
- ANONIZE: A Large-Scale Anonymous Survey System, SP14

Thank You!