# **Alchemy:** Query Optimization for Private Data Federations

### <u>Setting</u>

- $\rightarrow$  Multiple private data owners
- $\rightarrow$  Query the union of their records
- $\rightarrow$  Query results may be precise or differentially private

#### Problem

- $\rightarrow$  Execution behavior leaks statistics
- $\rightarrow$  Secure oblivious query execution is slow (SMCQL)
- $\rightarrow$  Traditional query optimization utilizes private information
- $\rightarrow$  Non-Oblivious privacy guarantees are hard

#### <u>Goals</u>

- $\rightarrow$  Fast privacy-preserving query execution
- → Leverage differentially private-statistics gathered with MPC to offer semi-oblivious, private query processing
- → Reason symbolically about constraints from the schema to constrain intermediate results sizes
  > End to and privacy guarantees
- $\rightarrow$  End-to-end privacy guarantees

#### Solution

- → Combine relational constraints with DP statistics
- $\rightarrow$  Provide upper bounds for output cardinalties and runtimes
- $\rightarrow$  Private query processing specific optimization rules

### **Challenges**

- → Histogram selection and parameterization
- $\rightarrow$  End to End Privacy Guarantees in both results and stats
- → Maximize result accuracy and query processing efficiency

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# → Traditional data federations require trust between data owners or a trusted third party → In our setting data owners do not trust each other

# Secure Multiparty Computation

→ Group of N data owners want to securely compute function f on the union of their data
 → Share only *encrypted data* for computation
 → Can compute arbitrary functions

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