

Secure Data Types: A Simple Abstraction for Confidentiality-Preserving Data Analytics

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Introduction

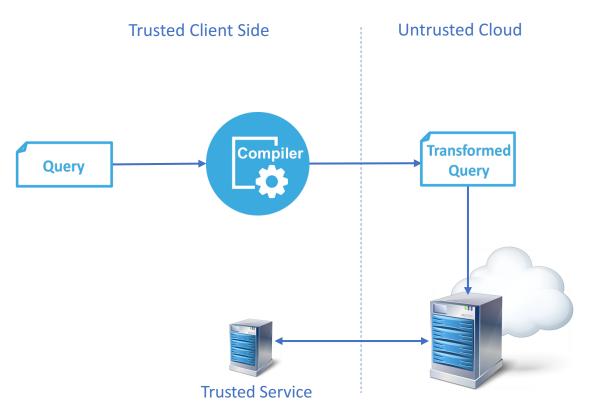


Requirement: Confidentiality–preserving query execution

Preserving Confidentiality

- Fully homomorphic encryption (FHE)
 - Can express arbitrary computations
 - High overhead for complex queries
- Partially homomorphic encryption (PHE)
 - Allows specific operations over encrypted data
 - E.g., addition, multiplication, comparisons, pattern match
 - Mutually incompatible (limited expressiveness)

Current PHE-based Solutions



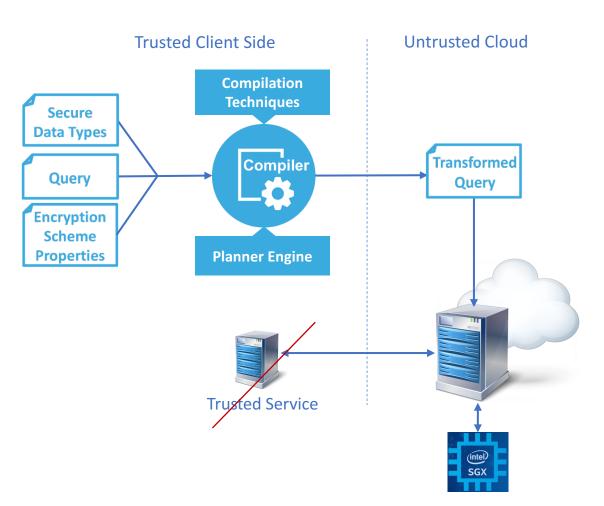
Drawbacks

- Compilation transparent to data constraints
- 2. Compilation largely ignores encryption scheme properties

	ASHE [OSDI'16]	Paillier [EUROCRT'99]
E(x) + E(y)	1	/
E(x) + y		/
E(x) × y		/
Performance	symmetric	asymmetric
Security	high	high

- 3. No/Limited use of trusted service
 - a) Give up (CryptDB [SOSP'11])
 - b) Split execution (Monomi [VLDB'13])
 - c) Re-encryption (Crypsis [ASE'14])

Cuttlefish



Secure data types (SDTs)

 Capture constraints and structure of data

Encryption scheme properties

 Capture supported operations, performance and security guarantees of encryption schemes

→ Compilation techniques

More optimized queries

→ Planner engine

- More efficient deployment
- Can utilize trusted hardware

Secure Data Types

- Sensitivity levels
 - high, low, public
 - Accounts for different security guarantees offered by cryptosystems
- Data range
 - +/- numbers
 - Fixed ranges, e.g., 100-200
- Composite types
 - Values containing multiple parts, e.g., dates, addresses, phones
 - E.g., composite [(4:int[+])-(2:int[range(1-12)])-(2:int[range(1-31)])]
- Also: decimal accuracy, uniqueness, tokenization, enumerated types, etc.

Compilation Techniques

- Expression rewriting
 - Simplify expressions involving composite types

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• E.g., d \ge 2010-01-01 AND d < 2011-01-01

\rightarrow y \ge 2010 OR (y == 2010 \text{ AND m} \ge 01) ...

\rightarrow y == 2010
```

- Condition expansion
 - Expand conditions to aggressively filter rows, based on range information

• E.g.,
$$x + y > c$$

 $\rightarrow y > (c - max(x))$ AND $x + y > c$ Short-circuit

• Similarly for $[+, -, \times, /]$ and $[==, >, \geq, <, \leq]$

Compilation Techniques (cont.)

- Selective encryption
 - Choose encryption scheme that does not require use of trusted service

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• E.g., (x + y) × z where z is public

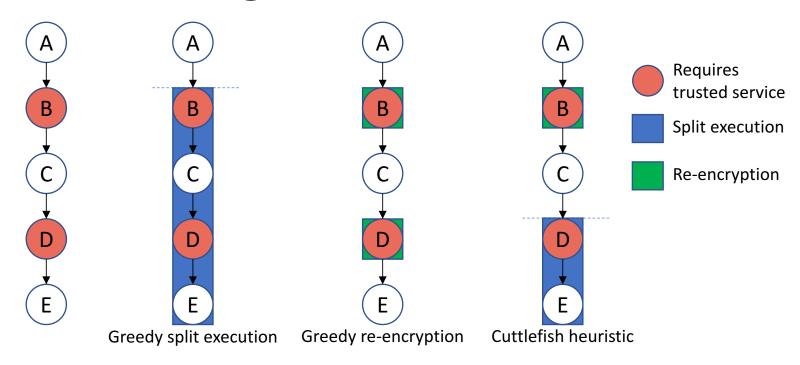
→ (ashe(x) + ashe(y)) × z

→ (paillier(x) + paillier(y)) × z
```

	ASHE [OSDI'16]	Paillier [EUROCRT'99]
E(x) + E(y)	/	1
E(x) + y		1
E(x) × y		1
Performance	symmetric	asymmetric

See paper for more compilation techniques

Planner Engine



- Cuttlefish Heuristic
 - Use a cost model to choose between re-encryption and split execution at each step
- Utilize trusted hardware, if available, to deploy an in-cloud reencryption service

Evaluation

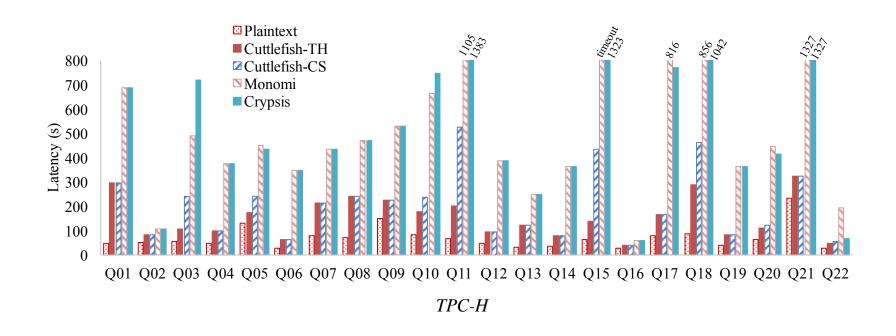
Cuttlefish

- Apache Spark 2.1
- Cuttlefish-TH: trusted service deployed using trusted hardware (Intel SGX)
- Cuttlefish-CS: trusted service deployed using remote client side

Setup

- TPC-H and TPC-DS (subset) at scale 100
- Cloud: 20 AWS m4.xlarge instances (4 CPUs and 16GB memory)
- Client: 1 AWS c4.2xlarge instance (8 CPUs and 15GB memory)

System Performance



Average overhead compared to plaintext

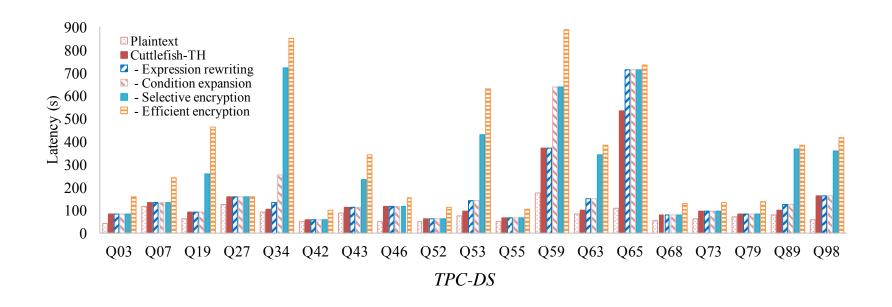
Cuttlefish-TH: 2.34 ×

Cuttlefish-CS: 3.05 ×

Average performance gains

- 3.35× faster than Monomi
- 3.71× faster than Crypsis

Compilation Techniques Performance



Average overhead compared to plaintext

- With Compilation techniques: 1.69 ×
- Without Compilation techniques: 4.23 ×

Conclusion

Cuttlefish enables efficient data analytics in public clouds

- Secure data types
 - Capture constraints and structure of data
- Compilation techniques
 - Enable more efficient queries
- Planner engine
 - Optimized use of trusted service

Thank you!