Proving the correct execution of concurrent services in zero-knowledge

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Proving correct executions

Verifies that code obeys a desired specification (first three talks) A cryptographic proof that desired code was correctly executed (this talk)

Neither subsumes the other

Consider a cloud-hosted wallet service (e.g., Square, WeChat Pay)



Verify trace by replay

API

- Issue (...)
- Transfer(...)
- Withdraw(...)

Issues with verifiability via record-and-replay

- 1. Sacrifices privacy: exposes requests and the internal state to a verifier
 - For example: account balances in the wallet app

Verifiable state machines address both problems

- 2. Verification via replay is expensive
 - Verifiers must reexecute all requests
 - Recorded trace can be large \rightarrow network costs are high

A verifiable state machine:



- Proofs are zero-knowledge: they do not reveal requests, responses, or the state
- Proofs are succinct: each proof is small and verification is inexpensive
- If the service errs, verifiers output reject (except for a small probability, <1/2¹²⁸)



Prior work suffers from two major issues

- 1. Producing proofs about storage operations is computationally expensive Several seconds to minutes of CPU-time/operation
- 2. They can only produce proofs about sequential executions \rightarrow each request must be processed before the next

For the wallet service app (on a single CPU core): Pantry [SOSP13] achieves < 0.15 requests/second Geppetto [S&P15] achieves < 0.002 requests/second

Our system: Spice

- Features a new storage primitive: 29—2000x more efficient
- Supports concurrent request processing, with transactional semantics
- Includes a toolchain:



- We built three apps: a wallet service, payment network, and a dark pool
- Throughput: 488—1048 reqs/sec (512 CPU-cores)
 - This is 18,000—685,000 higher throughput than prior work

Rest of this talk

- Background
- Overview of Spice
- Experimental results

Background: Pantry[SOSP13]

Under Pantry, a service is expressed using:





- Bitwise operations
- Conditional control flow
- Volatile memory (with pointers)
- Loops (with bounded iterations)

storage primitives

• Key-value ops: get, put, etc.

Mechanics of Pantry [SOSP13] to produce proofs



Background: Pantry[SOSP13]

Under Pantry, a service is expressed using:



- Arithmetic operations
- Bitwise operations
- Conditional control flow
- Volatile memory (with pointers)
- Loops (with bounded iterations)

An attempt:

Value get(Key k) { Value v = service_get(k); return v;

key-value store maintained by service

clientId	balance
1	200
2	500
3	100

service supplies state

service could supply incorrect values

Merkle trees provide the necessary building block



Issues with using Merkle trees for key-value stores

1. Cost of a get/put is logarithmic in the size of the state

- 2. The root of the tree serves as a contention point
 - \rightarrow supports only sequential executions

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Spice in a nutshell:

subset of C

...



- Arithmetic and bitwise ops
- Conditionals, loops, memory

Compile and apply argument protocol

Succinct zero-knowledge proof



[Blum et al. FOCS91, Clarke et al. ASIACRYPT03, Arasu et al. SIGMOD17]



read-set is a not subset of write-set

read-set is a subset of write-set

service's state



An equivalent of Merkle root struct set-root { set-hash rs; // set-hash of read-set set-hash ws;

$$Set-Hash\left(\begin{array}{c} A, \\ B \end{array} \right) = Set-Hash\left(\begin{array}{c} A \end{array} \right) * Set-Hash\left(\begin{array}{c} B \end{array} \right)$$
$$= Set-Hash\left(\begin{array}{c} B \end{array} \right) * Set-Hash\left(\begin{array}{c} A \end{array} \right)$$

Takeaways on set-based storage:

get, put add an element to read-set and write-set



cost of a get, put is a constant

service periodically proves
read-set \subset write-set



cost is linear in state size, but amortized over a batch

non-conflicting set operations commute



multiple writers and concurrent request processing

We built transactions and apps atop set-based storage



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Evaluation questions

- 1. How does Spice compare with the prior state-of-the-art?
- 2. What is the end-to-end performance of apps built with Spice?

Evaluation testbed:

Azure D64s_v3 instances: 32 CPUs, 2.4 Ghz Intel Xeon, 256 GB RAM, running Ubuntu 17.04

(1) How does Spice compare to prior work?

A million key-value pairs

Transactions with a single operation, keys chosen with a uniform distribution Metric: number of ops/second (i.e., proofs/sec)

	get	put
Pantry [SOSP13]	0.078	0.039
Pantry++	0.15	0.076
Geppetto [S&P15]	0.002	0.002
Spice (1-thread)	3.6	3.6
Spice (512-threads)	1,366	1,370

(2) End-to-end performance with varying #CPUs



- TPS is 18,000—685,000x better than prior state-of-the-art
- Verification throughput: >1,000 proof verifications/sec (4 CPU-cores)

Limitations of Spice

 CPU-cost to produce proofs remains large (compared to executions without proofs): >1000x

- 2. Spice amortizes the cost of producing a proof (that read-set \subset write-set) over a batch of requests
 - Introduces latency for producing proofs (7.5 minutes in our experiments)
 - Tunable, but lower latency increases CPU-costs

Summary

- Verifiable state machines add verifiability to services—without compromising their privacy
- Spice is a substantial progress toward building verifiable state machines
 - 18,000—685,000x better performance (over prior state-of-the-art)
 - Spice supports realistic apps with thousands of transactions/sec
- We predict: Spice or a variant will be a key tool in building secure systems