

Research Faculty Summit 2018

Systems | Fueling future disruptions



Oasis: Privacy-Preserving Smart Contracts at Scale

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The Value of Data Analytics and Machine Learning

Data analysis and machine learning has many applications, huge potential impact



"Data is the New Oil"







Dia Sent Mail Spam (372) Trash





What are biggest problems affecting data today?

Data breaches are becoming more common



Sport

Weather

Shop Earth

Travel

More

Most Data Is Siloed



Users Are Losing Control of Their Data





♠ > Technology Intelligence

Millions of private Gmail messages read by third parties

Blockchain: a Transformative Technology







Openness & transparency

No reliance on a central party

Automatic enforcement of agreements

The future of blockchain Credit scoring ٠ Decentralized exchange ٠ Decentralized hedge fund ٠ Medical diagnostics ٠ Personalized medicine ٠ **Private auctions** ٠ Cryptokitties Payments Tokens Internet of Things applications ٠ $\square \bigtriangledown$ ethereum

Fraud detection

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Motivating example: Fraud detection



Banks would all benefit by combining data to train better model Can't do this today because:

- Privacy concerns
- Regulatory risk
- Misaligned incentives

Motivating example: Fraud detection



Oasis: Privacy-preserving Smart Contracts at Scale Our Solution Properties of Our Solution

Privacy-preserving Smart contract



Oasis Blockchain

- Automatic enforcement of codified privacy requirements
- Without relying on any central party
- Scale to real-world applications including machine learning
- Easy to use for developers without privacy expertise

Privacy-Preserving Smart Contracts At Scale



Outline

1. Confidentiality-Preserving smart contract execution

2. Privacy-preserving analytics & machine learning

3. Scalable smart contract execution

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Confidentiality-preserving Smart Contract Execution



Secure computation techniques



Secure Hardware



Ekiden: Confidentiality-preserving Smart Contracts

Ekiden: A Platform for Confidentiality-Preserving, Trustworthy, and Performant Smart Contract Execution

Fan Zhang

University of California, Berkeley

Raymond Cheng University of California, Berkeley

Warren He

University of California, Berkeley

Ari Juels

Cornell Tech

National University of Singapore **Cornell University** Nicholas Hynes

Andrew Miller University of Illinois,

Dawn Song University of California, Berkeley

Jernej Kos

Noah Johnson

University of California, Berkeley

https://arxiv.org/abs/1804.05141

Urbana-Champaign

- Smart contract execution using secure computation:
 - Secure Enclave (e.g. Intel SGX)
 - Cryptographic protocols: secure MPC or Zero-knowledge Proofs

• Security proof: Universal Composability

Ekiden: Sample Applications

Application	Secret Input/Output	Secret State
Machine Learning	Training data, predictions	Model
Thermal Modeling	Sensor data, temperature	Building model
Token (Rust)	Transfer(from, to, amount)	Account balances
Poker	Players' cards	Shuffled deck
Cryptokitties	Random mutations	Breeding algorithm
Ethereum VM	Input and output	Contract state

Secure Enclave as a Cornerstone Security Primitive

- Strong security capabilities
 - Authenticate itself (device)
 - Authenticate software
 - Guarantee the integrity and privacy of execution
- Platform for building new security applications
 - Couldn't be built otherwise for the same practical performance
 - Many examples
 - Haven [OSDI'14], VC3 [S&P'15], M2R[USENIX Security'15], Ryoan [OSDI'16], Opaque [NSDI'17]

Trusted hardware timeline

Closed source



- Can be publicly analyzed and verified
- Can be manufactured by any manufacturer
- First release: Fall 2018

Challenges in Secure Hardware

- How secure can it be? Under what threat models?
- What would you entrust with secure hardware?
 - Your bitcoin keys
 - Financial data
 - Health data
- Can we create trustworthy secure enclave as a cornerstone security primitive?
 - Widely deployed, enable secure systems on top
 - A new secure computation era

Path to Trustworthy Secure Enclave

- Open source design
 - Provides transparency & enables high assurance
 - Builds a community
- Formal verification
- Secure supply-chain management

Keystone Enclave

- What is the Keystone Enclave?
 - Open-source Trusted Execution Environment (TEE) based on RISC-V
- Strong Memory Isolation
 - ISA-enforced memory access management
 - Separate virtual memory management without relying on the OS
- Simple and Portable
 - Exploits standard RISC-V ISA primitives: PMP, TVM
- Remote Attestation
 - Extends MIT Sanctum's remote attestation
- Open Source
 - Full software/hardware stack will be released
 - Run on many platforms: QEMU, Amazon AWS FPGA (FireSim), HiFive Unleashed, ...



Keystone Goals and Roadmap

Website: https://keystone-enclave.org



Open-source Hardware Enclave

Overview

Keystone is an open-source project for building <u>trusted execution environments</u> (TEE) with secure hardware enclaves, based on the <u>RISC-V</u> architecture. Our goal is to build a secure and trustworthy open-source secure hardware enclave, accessible to everyone in industry and academia.

Why do we need secure hardware enclaves?

Secure computation is a powerful abstraction, protecting the integrity and confidentiality of computations over secret data. While there are already many applications for secure computing, it will continue to grow in importance. First, the shift towards cloud computing has driven high demand for security in the cloud, because

- 1. Chain of Trust
 - Secure boot
 - Remote attestation
 - Secure key provisioning (PUF)
- 2. Memory Isolation
 - Physical memory protection
 - Page table isolation
- 3. Defense against Physical Attack
 - Memory encryption
 - Memory address bus encryption
- 4. Defense against Side-channel Attack
 - Isolated architecture
- 5. Formal Verification
- 6. Deployment
 - RISC-V QEMU
 - Amazon AWS FPGAs (FireSim)
 - HiFive Unleashed
- 7. Tape Out to Chip
- 8. Secure supply-chain management

Timeline



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Privacy Risks in Analytics

How many trips were taken in New York last year? How many trips did Joe take last week?

Reflects a trend

Reflects an individual

Access control policies cannot enable the use of data while protecting the privacy of individuals

Data Anonymization



Data Anonymization



Reidentification attacks

Netflix prize (Narayanan et al.)

NYC taxi data (Anthony Tockar)





Do Neural Networks Remember Training Data?

Can Attackers Extract Secrets (in Training Data) from (Querying) Learned Models?

N Carlini, C Liu, J Kos, Ú Erlingsson, and D Song. "The Secret Sharer: Measuring Unintended Neural Network Memorization & Extracting Secrets". 2018.

Extracting Social Security Number from Language Model

- Learning task: train a language model User son Enron Email dataset
 - Containing actual people's credit card and social security numbers
- New attacks: can extract 3 of the 10 secrets completely by querying trained models
- New measure "Exposure" for memorization
 - Used in Google Smart Compose

User	Secret Type	Exposure	Extracted?
A	CCN	52	\checkmark
В	SSN	13	
	SSN	16	
С	SSN	10	
	SSN	22	
D	SSN	32	\checkmark
F	SSN	13	
3	CCN	36	
G	CCN	29	
	CCN	48	\checkmark

Preventing Memorization

				Testing	Estimated
Differential Privacy: a formal notio	n	Optimizer	ε	Loss	Exposure
of privacy to protect sensitive input	ts				
		RMSProp	0.65	1.69	1.1
		RMSProp	1.21	1.59	2.3
Solution: train a differentially-	DP	RMSProp	5.26	1.41	1.8
private neural network	th]	RMSProp	89	1.34	2.1
 Exposure is lower empirically 	Wi	RMSProp	$2 imes 10^8$	1.32	3.2
 Attack upable to extract cocrete 		RMSProp	1×10^9	1.26	2.8
• Allack unable to extract secrets)	SGD	∞	2.11	3.6
	DP				
	0 L	SGD	N/A	1.86	9.5
	Ž	RMSProp	N/A	1.17	31.0

Differential Privacy: a Formal Privacy Definition



- Outcome is the same with or without Joe's data
 - Holds for *every* user and *every* database
- Immune to re-identification attacks
- Parameterized by ε (the *privacy budget*)

Real-world Use of Differential Privacy

- Previous work on differential privacy is either:
 - Theoretical
 - Targeted for specialized applications
 - Google: top websites visited
 - Apple: top emojis used
- No previous real-world deployments of differential privacy for general-purpose analytics

Challenges for Practical General-purpose Differential Privacy for SQL Queries

- Usability for non-experts
- Broad support for analytics queries
- Easy integration with existing data environments

No existing system addresses these issues

Collaboration with Uber: address practical deployment challenges

Chorus: a Framework for Privacy-preserving Analytics

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• Usable by non-experts

- Analyst does not need to understand differential privacy
- Chorus automatically enforces differential privacy for SQL queries



Broad support for analytics queries

- Modular design to support wide variety of mechanisms
- Implemented mechanisms support 93% of queries in our workload



• Easy integration with existing data environments

- Chorus works with standard SQL databases
- Designed for real-world use
 - Deployment underway at Uber

Optio: Privacy-Preserving Machine Learning

- Optio provides automatic differential privacy guarantees
 - Rewriting and verifying analytics and ML pipelines
 - Type system to enforce privacy policies



Real-world Deployment at Uber

- Ongoing deployment for analytics
 - Differential privacy
 - GDPR
- Plans for public-facing systems
- Open-source release: https://github.com/uber/sql-differential-privacy



Kara

A Privacy-Preserving Tokenized Data Market for Medical Data



Medical data is locked in "Data Silos". Goal: Incentivize doctors and patients to share data and improve medical research!



Kara

Meet Kara!

- Kara is a privacy-preserving tokenized data market
- Easy, fast and secure way for doctors and patients to earn tokens by sharing data
- Data is stored securely and privately in Oasis Blockchain Platform
- Researchers, doctors, industry can look for certain diseases / categories and pay to train their models with privacy-preserving machine learning



How it works

Doctors / Patients

Kara

Researchers



Nick Hynes, Raymond Cheng, Noah Johnson, David Dao, Dawn Song. "A Demonstration of Sterling: A Privacy-Preserving Data Marketplace" in VLDB'18 (Demo Track)

David Dao, Dan Alistarh, Claudiu Musat, Ce Zhang. "DataBright: Towards a Global Exchange for Decentralized Data Ownership and Trusted Computation"

Oasis: Example use cases



Oasis Labs Just Launched!

MIT Technology Review

aics+ The Download Magazine Events More+

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\equiv Forbes

Meet Oasis Labs, the blockchain startup Silicon Valley is buzzing about

Big Hitter Crypto Funds Pile Into Privacy-Enhanced Smart Contract Startup Oasis Labs



Oasis Labs Building Cloud Computing on Blockchain With \$45 Million

Backers include a16zcrypto, Accel Partners, Binance, Polychain, Metastable

# WIRED	
BUSINESS	CULTURE
HOW A START BLOCKCHAIN	UP IS USING THE TO PROTECT





Crypto and venture's biggest names are backing a new distributed ledger project called Oasis Labs

DEAN TAKAHASHI @DEANTAK JULY 9, 2018 3:00 AM

Oasis Testnet

Interested in building an application on Oasis?

Join our private testnet! https://www.oasislabs.com/developers

Oasis Labs

Building a privacy-first, high performance cloud computing platform on blockchain.

We're hiring! www.oasislabs.com Thank you!

