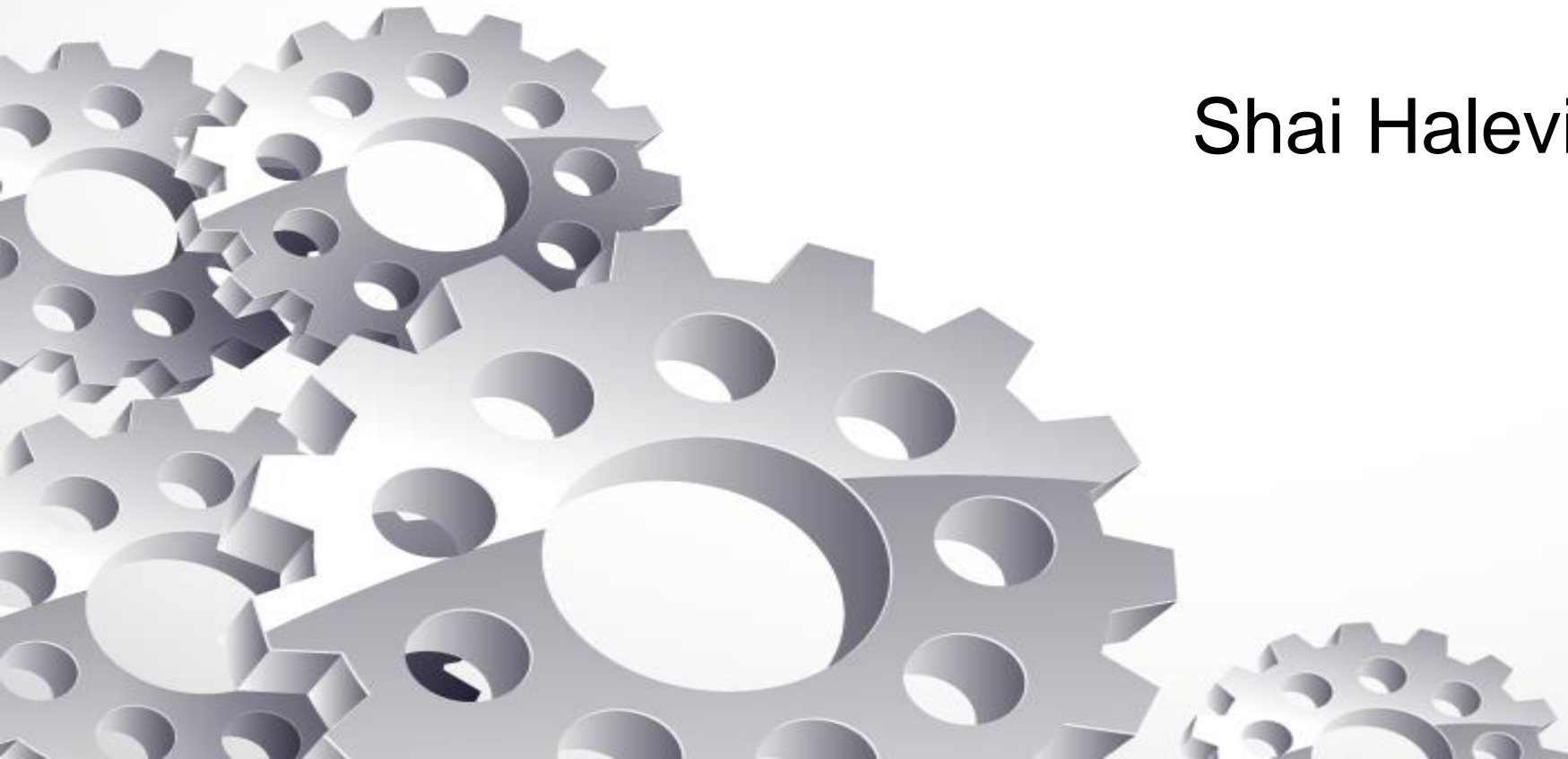


Advanced Cryptography: Promise and Challenges

Shai Halevi, IBM Research

ACM-CCS, October 2018



What's "Advanced Cryptography"?

- Cryptography beyond encryption, signatures
 - **Protecting computation, not just data**



What's "Advanced Cryptography"?



- Cryptography beyond encryption, signatures
 - **Protecting computation, not just data**

I'll mention three technologies:

– Zero-Knowledge Proofs (ZKP) 

– Secure Multi-Party Computation (MPC) 

– Homomorphic Encryption (HE) 

What's "Advanced Cryptography"?



- Cryptography beyond encryption, signatures
 - **Protecting computation, not just data**

I'll mention three technologies:

- Zero-Knowledge Proofs (ZKP)
- Secure Multi-Party Computation (MPC)
- Homomorphic Encryption (HE)

Not in this talk:

- Searchable Encryption
- Oblivious RAM (ORAM)
- Attribute-Based Encryption (ABE)
- ...

Advanced Cryptography is Needed



Advanced Cryptography is



Needed



Fast enough
to be useful



Advanced Cryptography is



Needed



Fast enough
to be useful



Not "generally
usable" yet



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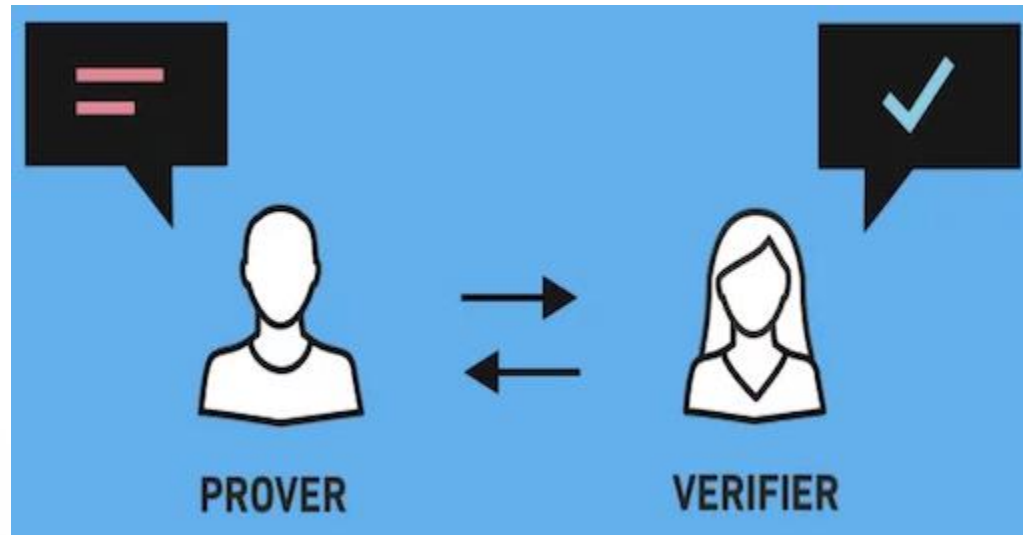
Advanced Crypto Tools

- Zero-Knowledge (ZK)
- Secure Multi-Party Computation (MPC)
- Homomorphic Encryption (HE)

Zero Knowledge Proofs



- I have a secret
 - I can convince you of some properties of my secret
 - Without revealing it



- Available (in principle) since the 80's [GMR'85]

Zero Knowledge Proofs

- I have a secret
 - I can convince you of some properties of my secret
 - Without revealing it
- Example: my secret is my purchase history



Zero Knowledge Proofs

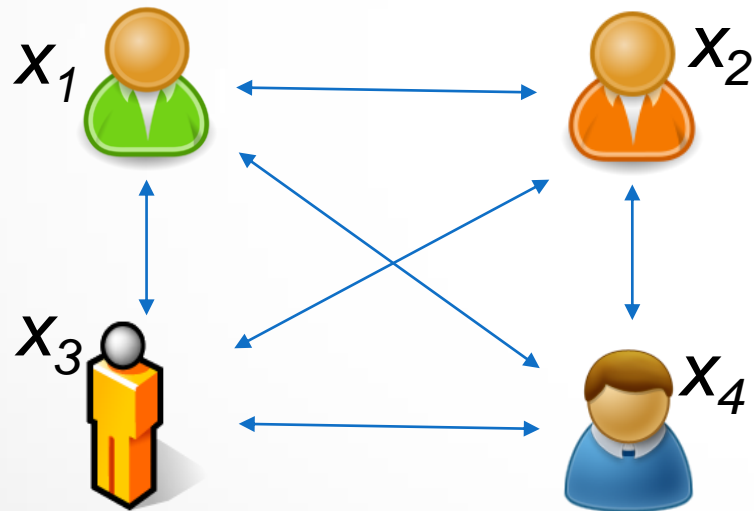
- I have a secret
 - I can convince you of some properties of my secret
 - Without revealing it
- Example: my secret is my purchase history
 - I can prove to Hood that I bought 10 gallons of milk this month
 - so I can get a coupon
 - Without revealing anything else



Secure Multi-Party Computation



- We all have our individual secrets
 - We can compute a function of these secrets
 - Without revealing them to each other (or anyone else)



Goal:

Correctness: Everyone computes $y = f(x_1, \dots, x_n)$

Privacy: Nothing but the output is revealed

- Available (in principle) since the 80's [Yao'86, GMW'86]

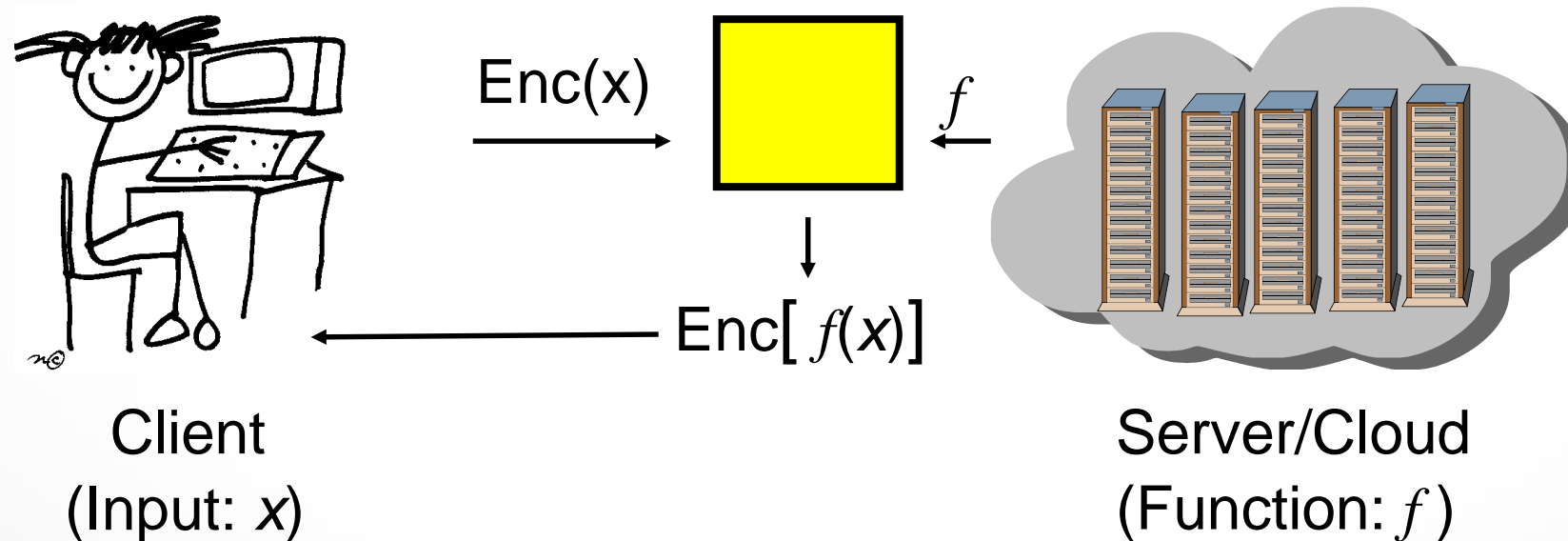
Secure Multi-Party Computation



- We all have our individual secrets
 - We can compute a function of these secrets
 - Without revealing them to each other (or anyone else)
- Example: medical data
 - Evaluating the effectiveness of a treatment
 - $f(\text{patient1Data}, \text{patient2Data}, \dots) = \text{effective/not-effective}$**
 - Data for different patients held by different clinics
 - Can compute this without revealing any private data

Homomorphic Encryption

- Data can be processed in encrypted form
 - Result is also encrypted



- Available (in principle) for <10 years [Gen'09]

Homomorphic Encryption

- Data can be processed in encrypted form
 - Result is also encrypted
- Example: location services
 - I encrypt my location, send to Yelp
 - Yelp compute an encrypted table lookup
 - $T[\text{cityBlock\#}] = \text{reviews for nearby coffee shops}$
 - I get back encrypted recommendation for coffee shops within two blocks



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The Promise of Advanced Cryptography

Blindfold Computation



- The ability to process data without ever seeing it



The Need for Advanced Cryptography

Your Privacy for Sale



- We give up information in return for services
 - E.g., location for directions, restaurant recommendation, health data for "personalized medicine", financials for tax and investment services, purchase history for better ads and coupons, ...

Your Privacy for Sale



- We give up information in return for services
 - E.g., location for directions, restaurant recommendation, health data for "personalized medicine", financials for tax and investment services, purchase history for better ads and coupons, ...
- Personalized services **require** personal information
 - or so we are told

Your Privacy for Sale



- We give up information in return for services
 - E.g., location for directions, restaurant recommendation, health data for "personalized medicine", financials for tax and investment services, purchase history for better ads and coupons, ...
- Personalized services **require** personal information
 - or so we are told
- What happens once we give away this information?



Data Abuse is the New Normal



- The entire IT industry is busy making it easier
 - Larger collections, better ways to link, process them



- Data abuse, not “data breach”
 - Overwhelming motivation to use whatever data can be found
 - If the data is available, it will be (ab)used

Data Abuse is the New Normal



- The entire IT industry is busy making it easier
 - Larger collections, better ways to link, process them



- It will only get worse
 - We cannot provide opportunity for easy abuse, seriously expect it not to happen

Data Abuse is the New Normal

- The entire IT industry is busy making it easier
 - Larger collections, better ways to link, process them



IT, security professionals

- We need all the tools we can get to push back
 - “Advanced Crypto” is an under-utilized tool in our box

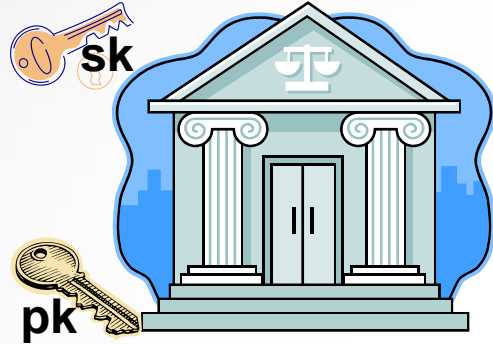
The Promise of Advanced Cryptography

Blindfold Computation



- The ability to process data without ever seeing it
 - Personalized services without access to private information
 - You cannot abuse data that's not there

Example: Anonymous Credentials using ZK



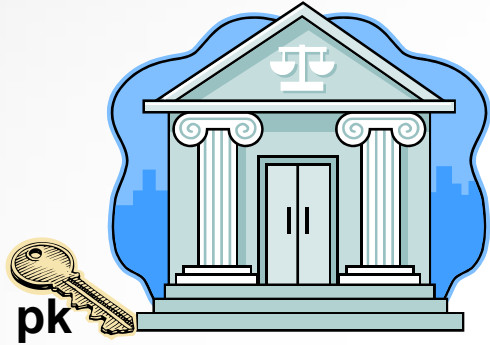
Issuing a
certificate




Name: Stick Person
DoB: August 1, 1988
Eye color: Black

Digital Signature: D2A6B1..8F

Example: Anonymous Credentials using ZK



“**D2A6B1..8F** is a valid signature wrt **pk** on a statement that includes a birthdate before 2000 and the picture ”



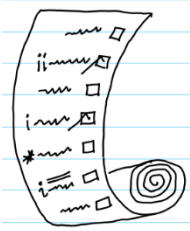
Prove in zero-knowledge



Example: No-Fly-List Using 2PC



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Police has
a list of
suspects



Airline has
a list of
passengers

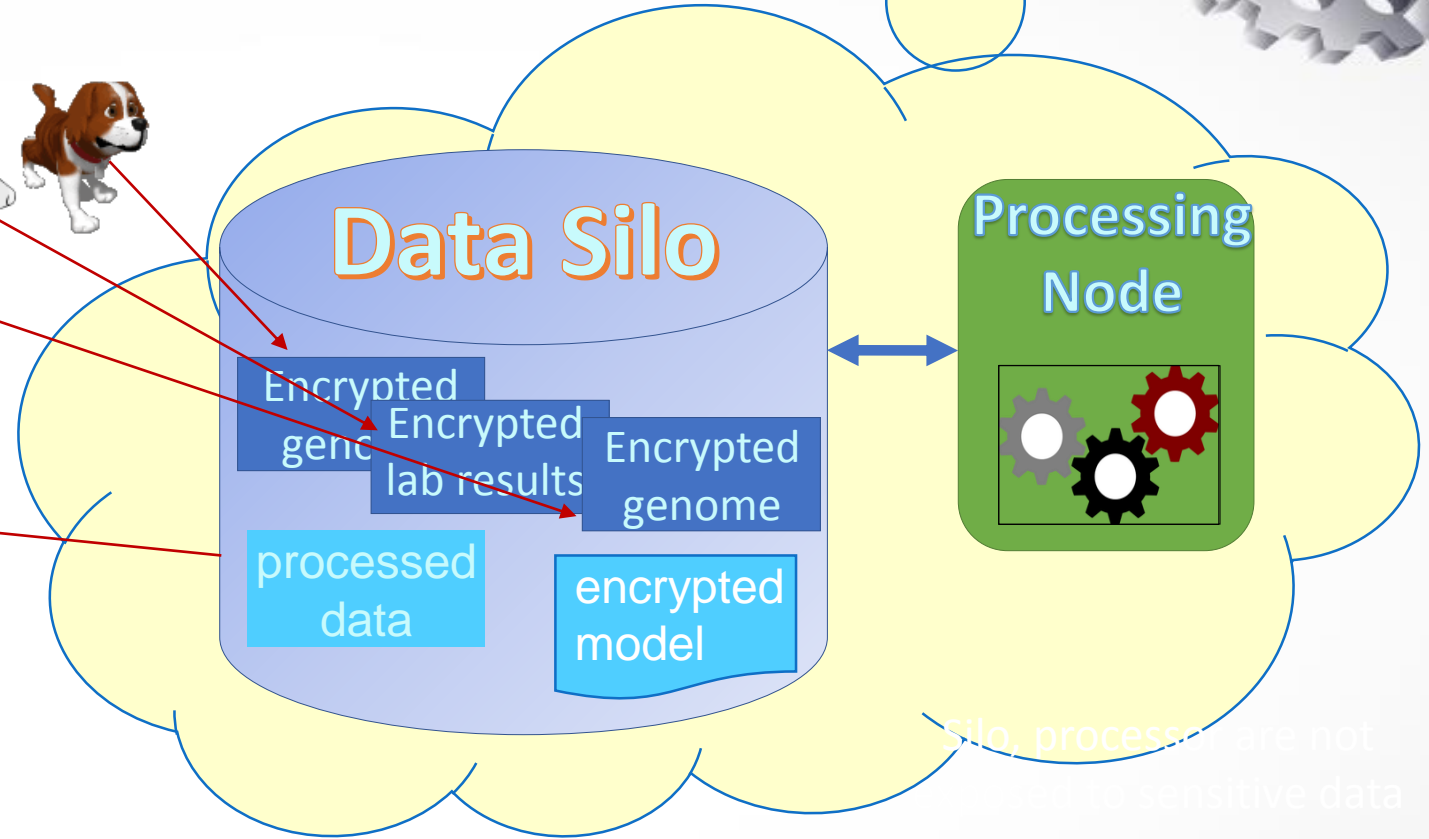
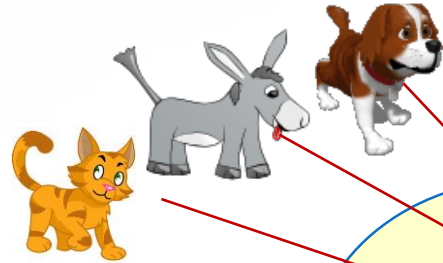


Output is the
intersection of
the two lists

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HE for Medical Data in the Cloud

Anyone can
Recovering results in the clear requires secret key, only processed results should be decrypted



Silo, processing are not
sensitive data

- “Silos” of encrypted data, each controlled by a key
 - Lots of stored data, small parts of it are in process at any time

The Promise of Advanced Cryptography

Blindfold Computation



- Also useful for “more traditional” security issues
 - E.g., key and credential management, protecting commercial secrets, collaboration on sensitive data, ...



Fast Enough to be Useful

Performance of Advanced Cryptography

- Improving performance has been a major research topic over the last 30 years
 - Tremendous progress, many orders of magnitude
- For most tasks, there is a cryptographic solution with adequate performance
 - Although designing it may take a team of experts



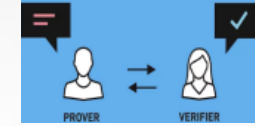
Some Speed Examples



- Lots of examples, meant to demonstrate feasibility of doing “many things” with reasonable performance
 - It’s okay to feel a little dizzy after example #17,352...
- The point is not to compare them
 - They operate in very different settings: “general-purpose” vs. specific functions, different security guarantees, different performance profiles, etc.

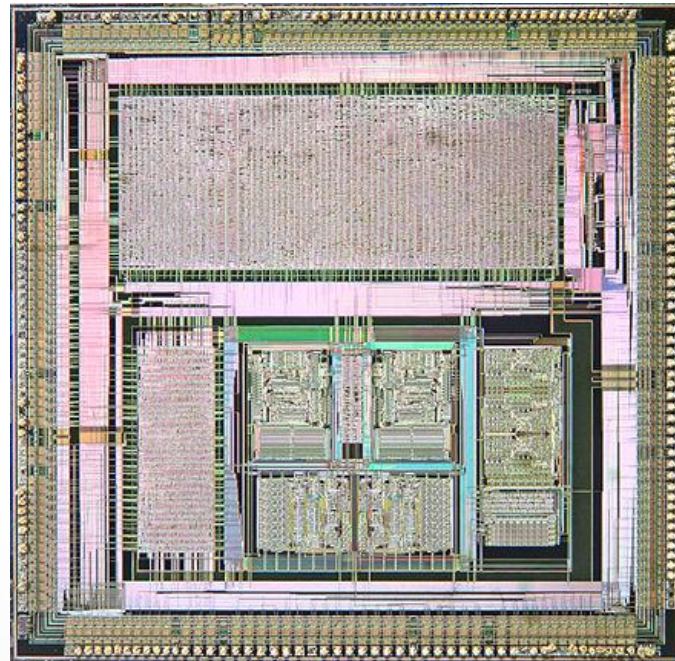
My apologies if I didn’t include your awesome work in this list

Some ZK Speed Examples



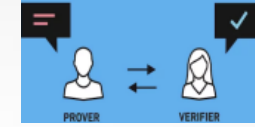
- Proving a 100,000-gate predicate in 1.8sec

Improved Non-Interactive Zero Knowledge with Applications [...]
(KKW, CCS 2018)



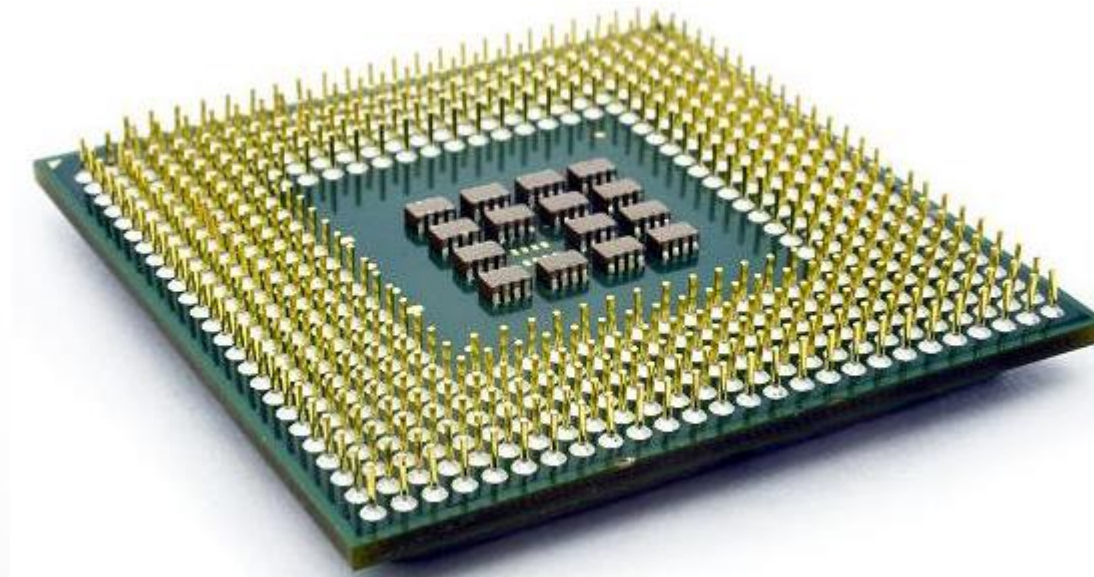
From this conference

Some ZK Speed Examples



- Proving a 2^{27} -gate predicate on a 64-cluster in ~1.5 hours

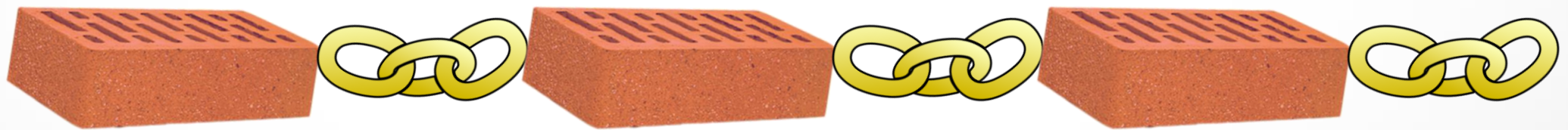
DIZK: A Distributed Zero Knowledge Proof System
(WZCPS, USENIX Security 2018)



Some ZK Speed Examples

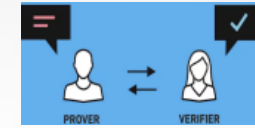


- “I know a pre-image of $0xA4E...1$ under SHA”
 - Proving at 100 pre-images/sec, verifying at 5000/sec
- Useful, e.g., for blockchains
 - Ligerio: Lightweight Sublinear Arguments Without a Trusted Setup (AHIV, CCS 2017)
- Useful, e.g., for blockchains



- Can prove things about the hash values in the blocks

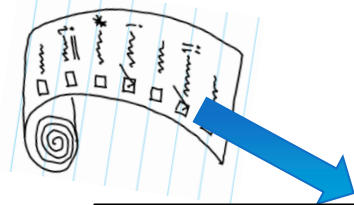
Some ZK Speed Examples



- DNA match against a database (zk-STARK, [BBHR, 2018])



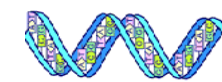
Police has
a forensic
DNA database



Public
commitment
0x3b2a108a

“the sample whose
hash is 0xe677d398
does not match
anything in the
database whose hash
is 0x3b2a108a”

Presidential
candidate has
a DNA sample

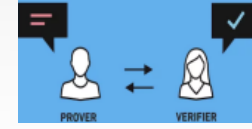


Public
commitment
0xe677d398



- Size-100,000 DB, proving in ~1 hour, verifying in milliseconds

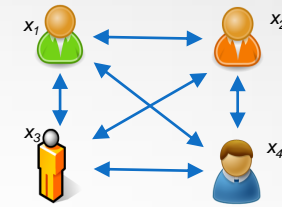
ZK Proofs in the Wild



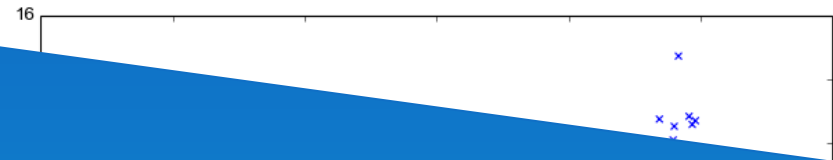
- Digital currencies (zCoin, Zcash, ...)
 - Proving that I have sufficiently many unspent coins on the ledger
 - Constructing proof in ~1min*, verification in a few msec
- Anonymous credentials (e.g., idemix)
 - Proving that I possess a credential, takes 1-30 seconds
- Private payments in the Brave browser (using Anonize)
- Tax bracket proofs (Deloitte/QEDit)
 - Commitments to my financial data posted to ledger
 - Then I can prove that I belong to a certain tax bracket
- ...

* Soon to be much faster

Some MPC Speed Examples

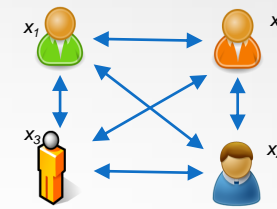


- Linear regression with



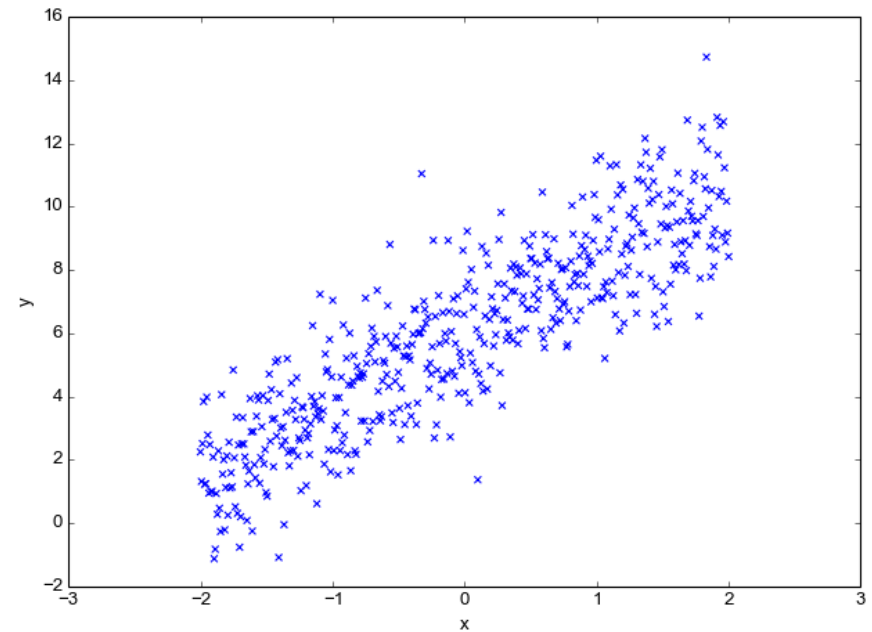
- For most protocols, the bottleneck is communication rather than computation
 - So performance is measured for LAN vs WAN

Some MPC Speed Examples



- 10-party linear regression with 4M inputs in 5sec over LAN

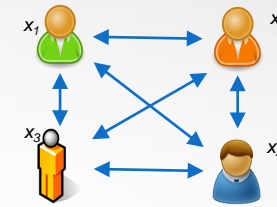
An End-to-End System for Large Scale P2P MPC-as-a-Service [...]
(BHKL, CCS 2018)



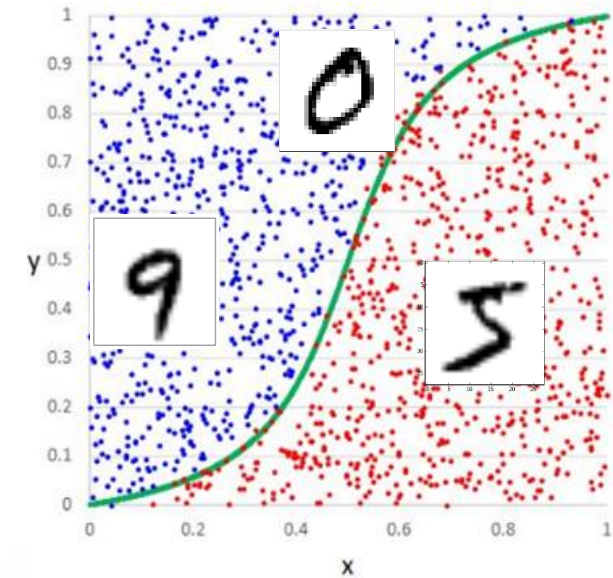
Data is shared among the parties, each holding 400,000 points

Cherry picked from this conference

Some MPC Speed Examples



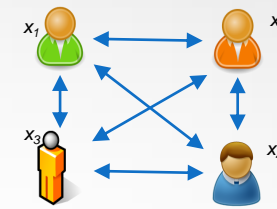
- 10-party regression with 4M inputs in 5sec over LAN
- 4-party logistic regression training in ~5 days over WAN
 - NANOPI: Extreme-Scale Actively-Secure Multi-Party Computation (ZCSH, CCS 2018)



Benchmarked on MNIST data:
1K rows x 784 columns

Cherry picked from this conference

Some MPC Speed Examples



- 10-party regression with 4M inputs in 5sec over LAN
- 4-party logistic regression training in ~5 days over WAN
- 2-party 16x16 Gaussian elimination in 16sec over WAN

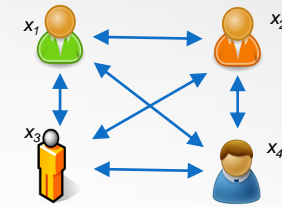
HyCC: Compilation of Hybrid Protocols
for Practical Secure Computation
(BDK, CCS 2018)

$$\begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} & b_1 \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} & b_m \end{bmatrix}$$

The matrix is shared between the two parties

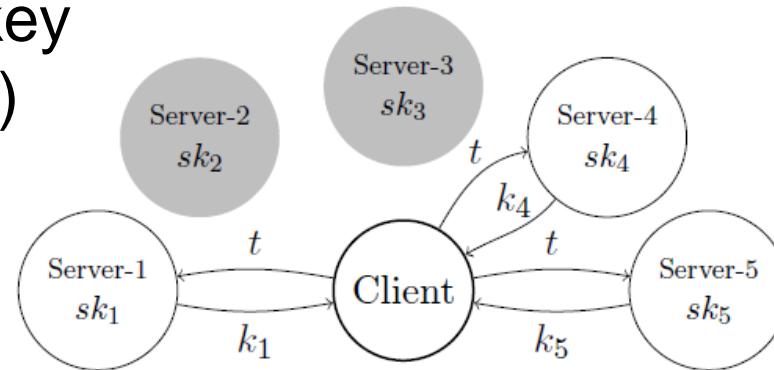
Cherry picked from this conference

Some MPC Speed Examples



- 10-party regression with 4M inputs in 5sec over LAN
- 4-party logistic regression training in ~5 days over WAN
- 2-party 16x16 Gaussian elimination in 16sec over WAN
- **12-party distributed AES >50,000 enc/sec on WAN**

DiSE: Distributed Symmetric-key
Encryption (AMMP, CCS 2018)



Encryption key is
secret-shared
among the servers

Cherry picked from this conference

More MPC Systems, Use-Cases



- Tax Fraud Detection System (Sharemind)
 - Analyzing one month of the Estonian economy in ten days
 - “How the Estonian Tax and Customs Board Evaluated a Tax Fraud Detection System Based on Secure Multi-party Computation” (BJSV, FC 2015)
- Virtual HSMs (Unbound), MPC replacing hardware
 - RSA, ECDSA, AES,..., comparable speed to hardware HSM

More MPC Systems, Use-Cases



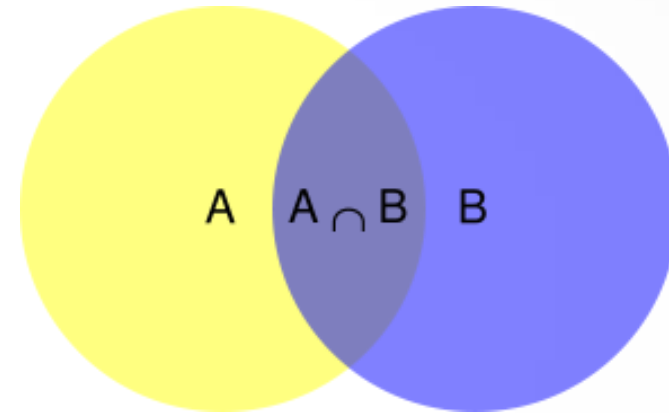
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- Virtual HSMs (Unbound), MPC replacing hardware
 - RSA, ECDSA, AES,..., comparable speed to hardware HSM
- Similar patients in a genomic database (iDASH 2016)
 - Best 5 matches against 4000 patients, 1000 markers, in ~30sec
“Privacy-Preserving Search of Similar Patients in Genomic Data”
(AHLR, PoPETS 2018)
- Clearing-price auction on Hyperledger Fabric, 10-20sec
 - “Initial Public Offering (IPO) on Permissioned Blockchain using Secure Multiparty Computation” (BDHHJMZ 2018)

HE Speed Examples



- Set intersection, size- 2^{20} by size-512 sets in 1 sec

Labeled PSI from Fully Homomorphic Encryption with Malicious Security (CHLR, CCS 2018)



From this conference

HE Speed Examples



- Set intersection, size- 2^{20} by size-512 sets in 1 sec
- Multiplying two 64x64 “real matrices” in ~9 seconds

Secure Outsourced Matrix Computation and Application to Neural Networks (JKLS, CCS 2018)

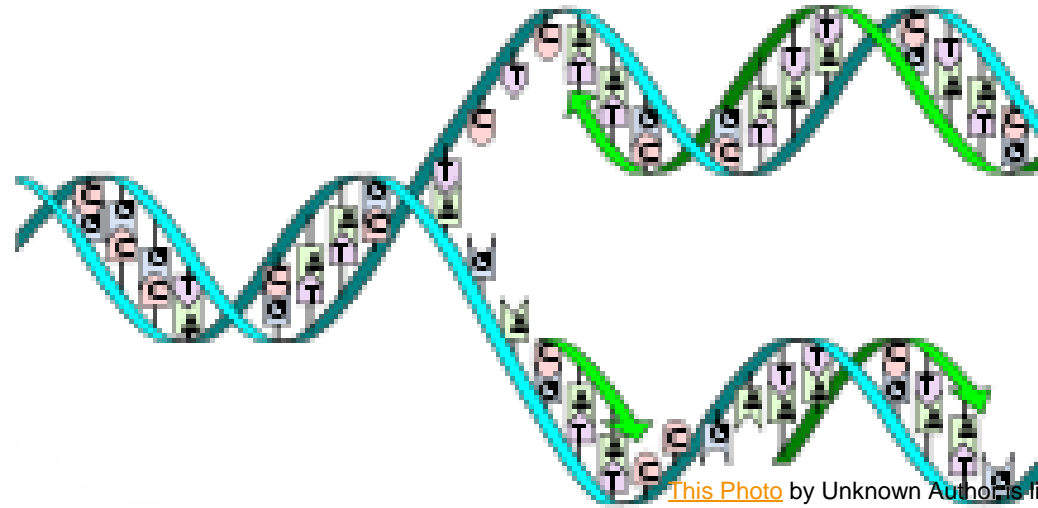
$$\begin{array}{ccc} \left[\begin{array}{ccc} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{array} \right] & \left[\begin{array}{ccc} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{array} \right] & = & \left[\begin{array}{ccc} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{array} \right] \\ \mathbf{A} & \mathbf{B} & & \mathbf{C} \end{array}$$

From this conference

More HE Speed Examples



- Computing similarity of two 1M-vectors in minutes
 - Similarity of encrypted genome sequences (iDASH 2015)

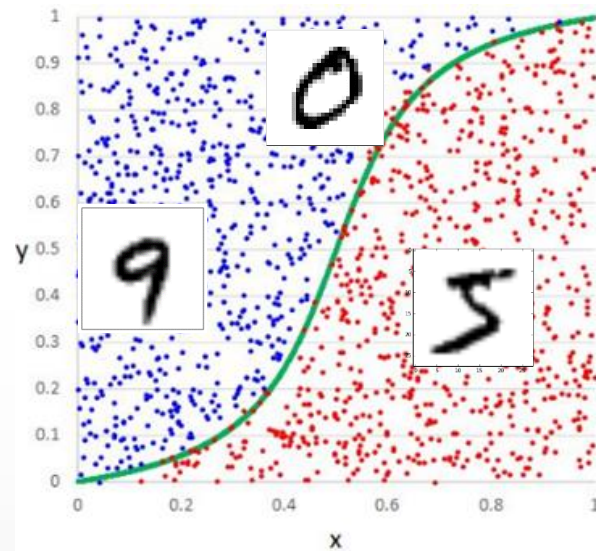


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More HE Speed Examples



- Computing similarity of two 1M-vectors in minutes
- Inference of simple models on encrypted data
 - 1000 predictions/minute, CNN on MNIST optical characters
“Crypto-Nets: Neural Networks over Encrypted Data” (DGLLNW, ICML 2016)
 - 8000 predictions/second on 100-feature LR model



More HE Speed Examples



- Computing similarity of two 1M-vectors in minutes
- Inference of simple models on encrypted data
 - 1000 predictions/minute, CNN on MNIST optical characters
 - 8000 predictions/second on 100-feature LR model
- Training a logistic-regression model on genome data
 - Under 10 minutes with 10-15 features, ~1000 rows (iDASH 2017)
 - “Logistic Regression Model Training based on the Approximate Homomorphic Encryption” (KSKLC, BMC Medical Genomics 2018)
 - 15-30 minutes to train 30,000 models w/ 5 features (iDASH 2018)

Such awesome performance, how come we're not seeing these tools everywhere?



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Not “Generally Usable” Yet

Complexity of Advanced Cryptography



- Distributed computing is already complex enough, “advanced crypto” adds secrecy considerations
- Good performance requires extreme optimizations
 - Straightforward implementation will be exceedingly slow
 - Small application-level changes can make a big difference in how to best optimize for it
- Tension between simplicity/usability and performance

Implementations

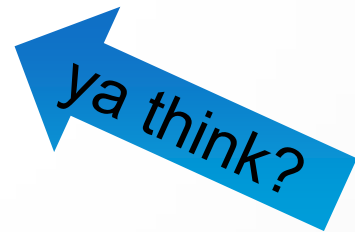


- Many software libraries for ZKP / MPC / HE
 - Most of them open-source
- Very hard to compare them, decide which technology/implementation to use for what purpose
 - Different tools, data models, computation models, performance profiles, security guarantees, ...
 - Hardly any accepted benchmarks
- Many of the libraries are written for speed, not usability

Code Quality

- Most code written in C/C++
 - By researchers with limited C/C++ experience

```
parts.push_back(CtxtPart(*ptr,handle));  
if (negative) parts.back().Negate(); // not thread-safe??
```



Example: Secure-MPC Communication

- Communication between parties is a bottleneck in many protocols for secure multi-party computation
 - To optimize, many MPC libraries work with sockets
 - The library expects to be “in charge” of IP-address:port

```
int main(int argc, char** argv)
{
    const char* addr = "127.0.0.1";
    int port = 7766;

    if (m_nPID == SERVER_ID) { //Play as OT sender
        InitSender(addr, port, glock);
        OTextSnd* sender = InitOTextSnd(prot, m_nBaseOTs, m_nChecks, usemeocr, ftype, crypt);
        [...]
    }
    else { //Play as OT receiver
        InitReceiver(addr, port, glock);
        OTextRec* receiver = InitOTextRec(prot, m_nBaseOTs, m_nChecks, usemeocr, ftype, crypt);
        [...]
    }
}
```

Example: Secure-MPC Communication

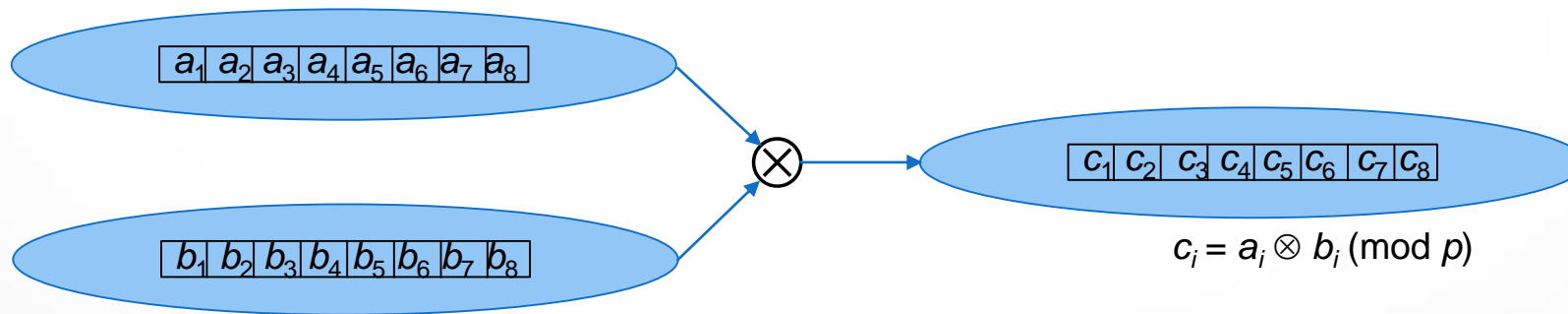


- Communication between parties is a bottleneck in many protocols for secure multi-party computation
 - To optimize, many MPC libraries work with sockets
- What if my system has its own communication layer?
 - E.g. working over https, gRPC, ...
- Retrofitting existing libraries to use “abstract channels” is a lot of work, may degrade performance
 - Your best option is to look for another library

Example: Data Encoding for HE



- Ciphertext operations in contemporary HE are slow
- “Ciphertext packing” to gain in performance
 - Each ciphertext encrypts a vector of plaintext element
 - Ciphertext operations effect element-wise operations



- Vector-size is a parameter, depends on the algebra

Example: Data Encoding for HE



- Lots of flexibility in setting the parameters
 - Determine plaintext modulus, vector-size, more
 - Choosing the right parameters is an art form
- Even with parameters set, where to put each piece of data requires a careful design
 - Could get orders-of-magnitude performance difference between different packing schemes
- Almost no tool support for making these choices

Taming the Complexity



- How to make advanced cryptography usable to non-expert programmers?
- Usable “toolbox libraries” for common tasks
 - Low level: arithmetic, sorting, linear algebra, ...
 - Mid level: graphs algorithms, set intersection, ML tools, ...
 - Domain specific tasks (medical, financial, ...)
- Design libraries as “middleware”
 - One component in larger systems
 - Don’t assume that the library “owns” the relevant resources

Taming the Complexity



- How to make advanced cryptography usable to non-expert programmers?
- Frameworks, compiler support
 - Some work over last 10+years
 - e.g., Fairplay, Sharemind, Obliv-C, ...
 - Considerable work reported in this conference
 - *An End-to-End System for Large Scale P2P MPC-as-a-Service[...]* (BHKL)
 - *HyCC: Compilation of hybrid protocols for Practical Secure[...]* (BDK)
 - *Generalizing the SPDZ Compiler For Other Protocols* (ABFKLOT)
 - *ALCHEMY: A Language and Compiler for HE [...]* (CPS)

Time to Put These Tools to Use

- The need is acute
- Push back against IT systems that put us in a fishbowl
- **Personalized services are possible without access to personal information**
 - Don't believe people telling you they're too slow



Time to Put These Tools to Use



- Cryptographers must put emphasis on usability and “mundane” software engineering aspects
 - Although improving performance is still important
- System builders should try to use what tools exist
 - Complain bitterly to your fellow cryptographers if their tools are too hard to use
- For now, keep designing one-off systems
 - Hopefully, some generalizations will emerge
 - These technologies are best suited for that type of applications

Time to Put These Tools to Use



- Some starting points to access these technologies:
 - Zero-Knowledge: <https://zkp.science/>
 - Secure-MPC: <https://github.com/rdragos/awesome-mpc> and <http://www.multipartycomputation.com>
 - HE: <http://homomorphicencryption.org/>
- We really need HOWTO documents
 - With application focus
 - Any volunteers to write them?

Incentives for Blindfold Computation?



- Customer demand?
 - Seems unlikely
- Government regulation?
 - Maybe, in some cases
- Developers wanting to do the right thing?
 - That's us, we have some choice in the systems that we build
 - Don't build systems that require users to hand over their data
 - It will be abused

Summary: Advanced Cryptography is



Needed

Fast enough
to be useful

Not "generally
usable" yet



- Can help prevent data abuse
- An under-utilized tool



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We are making some progress

Summary: Advanced Cryptography is



Needed

Fast enough
to be useful

Not "generally
usable" yet



- Can help prevent data abuse
- Still an under-utilized tool



Questions?



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We are making some progress

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