

# Verifiable Computation with Massively Parallel Interactive Proofs

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# Outsourcing

- Many applications require outsourcing computation to untrusted service providers.
  - Main motivation: Commercial cloud computing services.
  - Also, weak peripheral devices; fast but faulty co-processors.
  - Volunteer Computing (SETI@home, World Community Grid, etc.)
- User requires a guarantee that the cloud performed the computation correctly.
- One solution: require cloud to *prove* correctness of answer.

# Goals of Verifiable Computation

- Provide user with a correctness guarantee, without requiring her to perform the requested computations herself.
  - Ideally user will not even maintain a local copy of the data.
  - User may have resorted to the cloud in the first place because she has more data than she can store.
- Minimize the amount of extra bookkeeping the cloud has to do to prove the integrity of the computation.
- Ideally our protocols will be secure against arbitrarily malicious clouds, but sufficiently lightweight for use in more benign settings.

# Interactive Proofs

- Two Parties: Prover  $P$  and Verifier  $V$ .
- Think of  $P$  as powerful,  $V$  as weak.  $P$  solves a problem, tells  $V$  the answer.
  - Then  $P$  and  $V$  have a conversation.
  - $P$ 's goal: convince  $V$  the answer is correct.
- Requirements:
  - 1. Completeness: An honest  $P$  can convince  $V$  she's telling the truth.
  - 2. Soundness:  $V$  will catch a lying  $P$  with high probability no matter what  $P$  says to try to convince  $V$  (Secure even if  $P$  is computationally unbounded).



# Interactive Proofs

- IPs have revolutionized Complexity Theory in the last 25 years.
  - $\text{IP}=\text{PSPACE}$  [Shamir 90].
  - PCP Theorem e.g. [AS 98]. Hardness of approximation.
  - Zero Knowledge Proofs.
- But IPs have had very little impact in real delegation scenarios.
  - Why?
  - Not due to lack of applications!

# Interactive Proofs

- Old Answer: Most results on IPs dealt with hard problems, needed **P** to be too powerful.
  - But recent constructions focus on “easy” problems (e.g. “Interactive Proofs for Muggles” [GKR 08]).
  - Allows **V** to run **very** quickly, so outsourcing is useful even though problems are “easy”.
  - **P** does not need “much” more time to prove correctness than she does to solve the problem in the first place!



# Interactive Proofs

- Why does GKR not yield a practical protocol out of the box?
  - $\text{P}$  has to do a lot of extra bookkeeping (**cubic** blowup in runtime).
  - Naively,  $\text{V}$  has to retain the full input.
  - Substantial overhead due to finite field arithmetic and other technical issues.



# Engineering Practical IPs

## [CMT12, TRMP12]

# A Two-Pronged Approach

- The present paper is part of a recent line of work aiming to develop practical IPs [CCMT12, CMT10, CTY12, CMT12]
- Ideal: General purpose implementation allowing to verify arbitrary computation.
  - Based on general-purpose “Interactive Proofs for Muggles” construction [GKR 08].
- Also develop highly optimized protocols for specific important problems.
  - Reporting queries (what value is stored in memory location  $x$  of my database?)
  - Matrix multiplication.
  - Graph problems like perfect matching.
  - Certain kinds of linear programs.
  - Etc.

# Main Results: Part 1

- Can save  $V$  substantial amounts of space essentially for free.
  - Reason: GKR protocol (and several others) only requires  $V$  to store a fingerprint of the data.
  - This fingerprint can be computed in a single, light-weight pass over the input.
  - Fingerprint serves as a sort of "secret" that  $V$  can use to catch the cloud in a lie.
- Fits cloud computing well: pass by  $V$  can occur while uploading data to cloud.
- $V$  never needs to store entirety of data!
- The fingerprint is a few KBs in size, even if the input contains terabytes of data.

# Main Results: Part 2

- Can save  $V$  substantial amounts of time.
- E.g. when multiplying two  $512 \times 512$  matrices,  $V$  requires .12s to process the input, while naive matrix multiplication takes about .70 seconds.
- Savings for  $V$  will be much larger on at larger input sizes, when applying our implementation to more time-intensive computations than matrix multiplication (because  $V$ 's runtime grows quasi-linearly with input size; she just needs to compute a fingerprint of the input).

# Main Results: Part 3

- We've come a long way in making  $\mathbf{P}$  more efficient.
- In [CMT12], we brought the runtime of  $\mathbf{P}$  down from **cubic** in the size of a circuit computing the function of interest, to quasilinear in the size of the circuit.
- Lots of additional engineering in the implementation (helps make  $\mathbf{V}$  fast too).
  - Choosing the “right” finite field to work over.
  - Using the “right” circuits.
  - Etc.
- Practically speaking, this is still not good enough on its own.
  - $256 \times 256$  matrix multiplication takes  $\mathbf{P}$  about 27 minutes for our previous single-threaded implementation.

# Main Results: Part 4 (Focus of [TRMP12])

- Our implementation is extremely amenable to parallelization.
- Holds for both **P** and **V** (although **V** runs quickly even without parallelization, see Insight 2).

Problem	<b>P</b> time (single-threaded)	<b>P</b> time (GPU)	<b>V</b> time	Rounds	Communication
$F_2$ (n=2^20)	29.8 s	0.36 s	.19 s	118	2.5 KB
MatMult (256 x 256)	27.6 Minutes	39.6.s	.04 s	3910	91.6 KB

If **V** also has a GPU, we get close to 100-fold speedups for **V** relative to single-threaded implementation.

# Main Results: Part 4 (Focus of [TRMP12])

- Main challenge to parallelizing and scaling to large inputs was the memory-intensive nature of  $\textcolor{red}{P}$ 's computation in the GKR protocol.
  - Naïve  $n \times n$  matrix multiplication only requires  $O(n^2)$  space.
  - $\textcolor{red}{P}$  has to store a circuit of size  $O(n^3)$  (we use 40 bytes per gate).
  - Even  $256 \times 256$  matrix multiplication over 1.5 GBs of space.
  - Took steps to mitigate this issue despite limited device memory.

# Related Work

- Setty, McPherson, Blumberg, and Walfish [NDSS 12] implement an *argument* system original due to Ishai, Kushilevitz, and Ostrovsky [CCC 07].
  - Bring the runtime of the cloud down by a factor of  $10^{20}$  relative to a naive implementation.
  - Advantages of our implementation: save  $\mathbf{V}$  time even when outsourcing a single computation, secure against computationally unbounded clouds.
- Canetti, Riva, and Rothblum [CCS 12] give highly practical protocols which are secure when there are *two* clouds, at least one of whom is honest.
- Ben-Sasson, Chiesa, Genkin, and Tromer working toward practical PCPs.

# Conclusions

- Interactive Proofs and other protocols for verifiable computation represent some of the most celebrated results in complexity theory.
- They have the potential to mitigate trust issues in cloud computing, but were wildly impractical until recently.
- We can already save the user a lot of time and space.
- The main remaining bottleneck is the extra bookkeeping the cloud must do to provide integrity guarantees.
- Parallelization helps mitigate this issue, but there is still much work to be done.

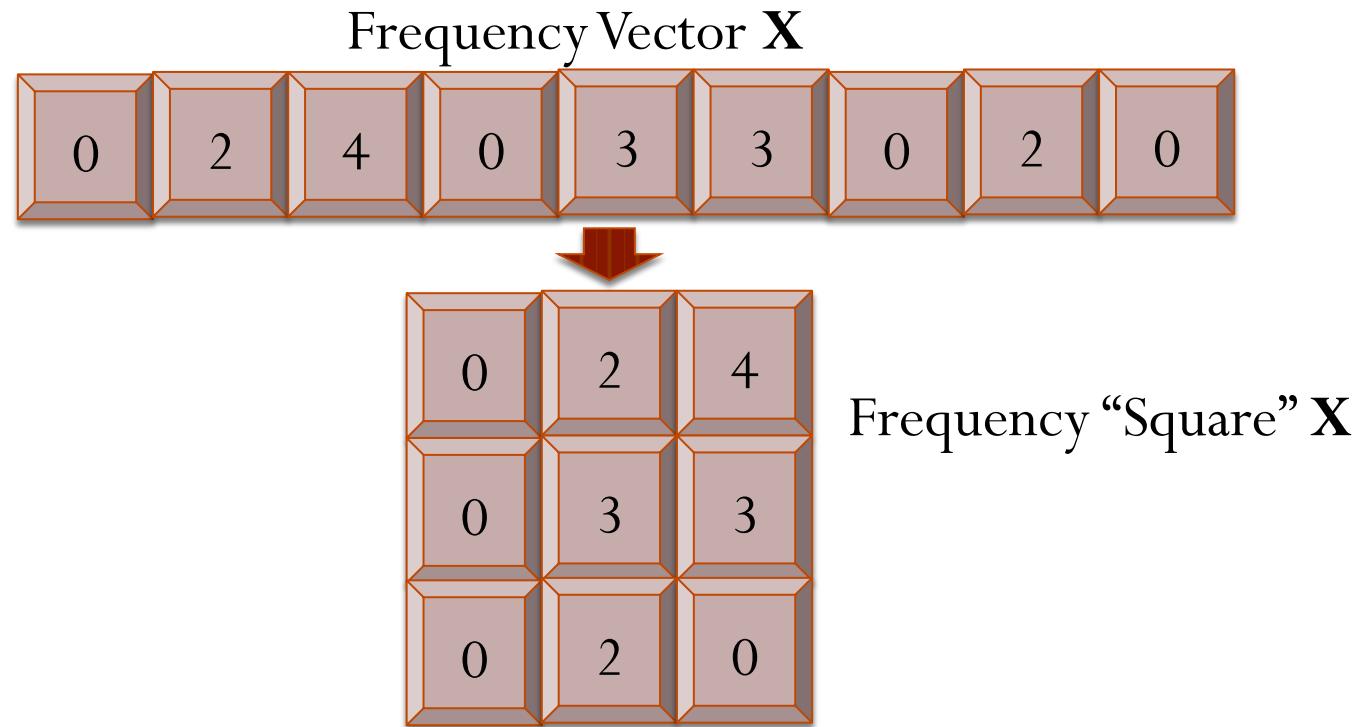
Thank you!

# Sample Variance of Data Stream

- The (scaled) sample variance of a data stream is defined as follows:
  - Let  $\mathbf{X}$  be the frequency vector of the stream  
( $X_i$  is number of occurrences of  $i$  in the stream)
  - $F_2(\mathbf{X}) = \sum_i X_i^2$
- [CCM 09/CCMT 12] give a one-message protocol for  $F_2$ , requiring  $O(\sqrt{n})$  communication and  $O(\sqrt{n})$  space for  $V$ .
- This is optimal.

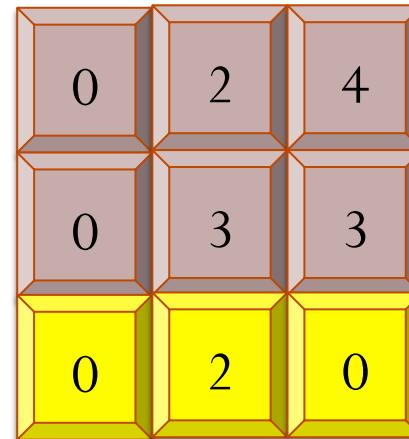
# Sample-Variance Protocol

- Recall:  $F_2(\mathbf{X}) = \sum_i X_i^2$
- View universe  $[n]$  as  $[\sqrt{n}] \times [\sqrt{n}]$ .



- First idea: Have  $P$  send the answer “in pieces”:
  - $F_2(\text{row 1}).F_2(\text{row 2})$ . And so on. Requires  $\sqrt{n}$  communication.
- $V$  exactly tracks a row at random (denoted in yellow) so if  $P$  lies about any piece,  $V$  has a chance of catching her. Requires space  $\sqrt{n}$ .

Frequency Square X



$P$  sends

$$20 = 2^2 + 4^2$$

$$18 = 3^2 + 3^2$$

$$4 = 2^2$$

- Problem: If  $P$  lies in only one place,  $V$  has small chance of catching her.
- We would like the following to hold: if  $P$  lies about even one piece, she will have to lie about many.
- Solution: Have  $P$  commit (succinctly) to second frequency moment of rows of an **error-corrected encoding** of the input.
- Need  $V$  to evaluate any row of the encoding in a streaming fashion. Can do this for “low-degree extension” code. Note: this code is *systematic*, meaning the first  $n$  symbols are just the input itself.

Input is  
embedded in  
encoding  
(low-degree  
extension)

### Error-corrected Encoding of Frequency Square X

0	2	4
0	3	3
0	2	0
0	-1	-5
0	-6	-12
0	-13	-21

These values  
will all lie on  
low-degree  
polynomial s(X)



H sends

$$20 = 2^2 + 4^2$$

$$18 = 3^2 + 3^2$$

$$4 = 2^2$$

$$26 = (-1)^2 + (-5)^2$$

$$180 = (-6)^2 + (-12)^2$$

$$610 = (-13)^2 + (-21)^2$$