## Program Representations (1)

#### Bar-Ilan Winter School on Verifiable Computation Class 7 January 5, 2016

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Let's clarify a few things from class 5:

• What are the constraints C' versus the constraints C?

• How does the assignment *z* (satisfying or not) affect V's checks?

• How and why do QAPs dramatically improve the picture?

## Attempt 3: Use long PCPs interactively (summary) [IKO07, SMBW12]



Achieves simplicity, with good constants ...

```
... but pre-processing is required (because |q_i| = |v|)
```

... and prover's work is quadratic; address that shortly

# Attempt 4: Use long PCPs non-interactivelypreprocessing[BCIOP13]SNARG



Query process now happens "in the exponent"

... pre-processing still required (again because  $|q_i| = |v|$ )

... prover's work still quadratic; addressing that soon

#### Recap

	efficient (short) PCPs	arguments, CS proofs	arguments w/ preprocessing	SNARGs w/ preprocessing
who	ALMSS92, AS92, BGSHV, Dinur,	Kilian92, Micali94	iko07, smbw12, svpbbw12	Groth10, GGPR12, BCIOP13,
what	classical PCP	commit to PCP by hashing	commit to long PCP using linearity	encrypt queries to a long PCP
security	unconditional	CRHFs	linearly HE	knowledge-of- exponent
why/why not	not efficient for V	constants are unfavorable	simple	simple, non- interactive
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(Thanks to Rafael Pass.)

Final attempt: apply linear query structure to GGPR's QAPs [Groth10, Lipmaa12, GGPR12]



Addresses the issue of quadratic costs.

PCP structure implicit in GGPR. Made explicit in [BCIOP13, SBVBBW13].

Summary of published argument implementations



- standard assumptions
- amortize over batch
- interactive

- non-falsifiable assumptions
- amortize indefinitely
- non-interactive, ZK, ...

QAPs play the same role (but much, much better!) as "Q(z) plus the  $[z, z \otimes z]$  encoding" (which is from [ALMSS92]; see [SMBW12, Apdx A] for a self-contained listing). This works because QAPs have a linear query structure, meaning that the query is a vector and the response is the dot product with a fixed vector).

Onto the front-end.....



This session: front-end techniques

• Key ideas: arithmetization, the convenience of non-determinism, data-dependent control flow, the price of generality, amortization

Recall the technical role of the front-end: given computation **f**, produce constraints **C**, where **C** is degree-2 constraints over  $\mathbb{F}$  and variables (X, Y, Z) s.t.  $\forall x,y: \exists w \text{ s.t. } y=\mathbf{f}(x,w) \Leftrightarrow \mathbf{C}(X=x,Y=y)$  is satisfiable









This session: front-end techniques

- Key ideas: arithmetization, the convenience of non-determinism, data-dependent control flow, the price of generality, amortization
- Focus on "non-deterministic ASICs"; provides intuition for the rest

#### Rest of this session

(1) Arithmetization: from programs to constraints

(2) Enhancing expressiveness: data-dependent control flow

(3) Costs and comparisons

We will walk through the process of transforming a program into equivalent constraints (arithmetization):

- How program structures translate.
- How the translation is automated by a C compiler.
- How the translation targets the format required by the back-end.

A lot of this is folklore (not many references, but see Braun's thesis [Braun12] and the appendices of Ginger [SVPBBW12]).

We will work over the field  $\mathbb{F}_p$  (the integers mod a prime, p). Let's begin with a warmup ...

Assignment allocates a fresh constraint variable (circuit wire):

Boolean functions turn into arithmetic:

// assume x1 and x2 are 0-1 valued y = x1 AND x2;  $\implies$ y = x1 OR x2;  $\implies$ 

EXERCISE: Fill in the equivalent constraints for the functions below:

y =	= NO]	[ x1;	$\implies$
y =	= x1	NAND $x2;$	$\implies$
y =	= x1	NOR x2;	$\implies$
y =	= x1	XOR x2;	$\implies$

Equality checks are efficient:

// x1 and x2 need not be Boolean  $z3 = (z1 != z2) ? 1 : 0; \Longrightarrow$ 

Observe: the constraints exploit "non-determinism" . . . even though the computation is deterministic.

EXERCISE: Fill in the equivalent constraints for the function below:

 $y = (x1 == x2) ? 1 : 0; \implies$ 

Conditionals require constraints (or gates) for each branch:

if (x1)  

$$y = x2;$$
  
else  
 $y = x3;$ 

EXERCISE: Fill in the equivalent constraints for the excerpt below:

if 
$$(z1 == 3)$$
  
 $z2 = 10;$   
else if  $(z1 == 5)$   
 $z2 = 20;$   
else  
 $z2 = 30;$ 

EXERCISE: Fill in the equivalent constraints for the excerpt below:

// assume z1, z2 are already defined  
if (z3 == 9)  
 z1 = z1 + 6; 
$$\implies$$
  
else  
 z2 = z2 + 10;

Loops are unrolled:

$$\left\{\begin{array}{l} Z = 0, \\ Z_0 = Z + 1, \\ Z_1 = Z_0 + 1, \\ \vdots \\ Z_9 = Z_8 + 1 \end{array}\right\}$$

Loop bounds must be static (for now).

EXERCISE (primitive load): (a) Write a program in pseudocode that takes two inputs: an array of some fixed size (which you can represent as a vector of variables) and an index in the array. Return the value at the specified index in the array. (b) Translate your program into constraints. (c) What's the most efficient set (smallest number) of constraints that you can produce for this program?

EXERCISE (Challenge!): Your solution to the previous exercise probably had O(m) constraints, where *m* is the size of the input array. Can you lower the number of constraints to  $O(\log m)$ ? (This will also require changing the input specification.)

Negative numbers require care. ( $\mathbb{F}_p$  has no notion of "less than zero".)

What about order comparisons (such as x1 < x2)?

if 
$$(x1 < x2)$$
  
y = 3;  
else  
y = 4;  

$$= \begin{cases} M\{C_{<}\}, \\ M(Y-3) = 0, \\ (1-M)\{C_{>=}\}, \\ (1-M)(Y-4) = 0 \end{cases}$$

$$C_{<} = \begin{cases} B_{0}(1 - B_{0}) &= 0, \\ B_{1}(2 - B_{1}) &= 0, \\ \cdots & \\ \end{array}$$

Cost: O(w), where w is bit width of variables.

#### EXERCISE: Write down constraints for z3 = z2 | z1, where | is bitwise or.

EXERCISE (Challenge!): So far, we have presumed that the original computation was working over the integers; we then mapped integer operations into  $\mathbb{F}_p$ , and from there to constraints. Extend this model to rational numbers: let the program work (in principle) over  $\mathbb{Q}$ , identify a suitable finite field for the constraints, and describe how to translate operations to constraints.

Hint: Show that there is a choice of p for which a computation over  $\mathbb{Q}/p$  (the quotient field of  $\mathbb{F}_p$ ) is isomorphic to a computation over  $\mathbb{Q}$ . How will you handle the order comparisons (<, etc.)?

The foregoing process is automated. A compiler for (a subset of) C:

- Transforms the input program to *single assignment*
- Uses "pseudoconstraints" for some of the assignments
- Outputs constraints and annotations (hints for the prover)

By tracking the sizes of intermediate values, the compiler:

- Infers lower bound on prime *p*.
  - Example: for matrix multiplication, compiler is told that inputs are signed N bits. Compiler can infer that p must be at least  $m \cdot 2^{2N}$ .
- Produces only necessary bitwise constraints.

For more about the mechanics of compilation, see Braun's thesis [Braun12]; a summary is in Pantry [BFRSBW13;  $\S2$ ,  $\S7$ ]. See also Ginger [SVPBBW12] and Pinocchio [PGHR13].

The compiler must obey the constraint format required by the back-end:

- Degree-2, and possibly also:
- Quadratic form, meaning  $p_A \cdot p_B = p_C$ , where each *p* is a degree-1 polynomial. This is needed for QAP-based back-ends [GGPR13].

EXERCISE: Assuming C consists of degree-2 constraints, describe a (straightforward) reduction from  $M\{C\}$  to a set of degree-2 constraints. What is the cost of the reduction, in terms of extra variables and constraints introduced?

EXERCISE: Consider the constraint  $\{3 \cdot Z_1Z_2 + 2 \cdot Z_3Z_4 + Z_5 - Z_6 = 0\}$ . Replace this with three constraints in quadratic form.

EXERCISE: What is the cost, in terms of the number of extra variables and constraints, of transforming a set of degree-2 constraints C to a set C' in quadratic form? What is the worst case? Do "usual" computations experience the worst case?

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```
("Quadratic Form" = "R1CS")
```

Question: what are the R1CS constraints for matrix multiplication?

Digression: What is Freivalds algorithm for matrix multiplication?

(1) Arithmetization: from programs to constraints

(2) Enhancing expressiveness: data-dependent control flow

(3) Costs and comparisons

What happens when loops are nested?

$$i=0; \\for (j=0; j<10; j++) {
i++; \\for (k=0; k<2; k++) {
i=i*2; } \\} \\Z=Z_{1} + 1, // j == 0 \\Z_{1} = Z_{0} \cdot 2, // k == 0 \\Z_{2} = Z_{1} \cdot 2, // k == 1 \\Z_{3} = Z_{2} + 1, // j == 1 \\Z_{4} = Z_{3} \cdot 2, // k == 0 \\Z_{5} = Z_{4} \cdot 2, // k == 1 \\...$$

Inner loop unrolls into every iteration of outer loop.

What happens when loops are nested?

$$i=0; \\for (j=0; j<10; j++) {
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i=i*2; } \\} \\Z_0 = Z + 1, // j == 0 \\Z_1 = Z_0 \cdot 2, // k == 0 \\Z_2 = Z_1 \cdot 2, // k == 1 \\Z_3 = Z_2 + 1, // j == 1 \\Z_4 = Z_3 \cdot 2, // k == 0 \\Z_5 = Z_4 \cdot 2, // k == 1 \\...$$

Inner loop unrolls into every iteration of outer loop.

What if the loop bounds were data-dependent?

```
"a5b2" \Rightarrow "aaaaabb"
```



- 1. Read (inchar, length) pair.
- 2. Emit inchar, length times.

```
"a5b2" \Rightarrow "aaaaabb"
```

At one extreme, a single character's run length could be OUTLENGTH.

```
"a5b2" \Rightarrow "aaaaabb"
```

At the other extreme, every character's run length could be 1, and the outer loop would iterate OUTLENGTH times.

```
"a5b2" \Rightarrow "aaaaabb"
```

Thus, the compiler must unroll the inner loop to OUTLENGTH<sup>2</sup> iterations, even though the computation is linear in OUTLENGTH.

#### Observations:

- 1. Loop nests are equivalent to finite state machines (FSMs) ...
- 2. ... but FSMs are more efficiently represented in constraints

#### Idea: transform loop nests into FSMs.



How can a compiler perform such a transformation systematically?

```
i = j = 0;
while (j < OUTLENGTH) {
    inchar = input[i++];
    length = input[i++];
    do {
        output[j++] = inchar;
        length--;
    } while (length > 0);
}
```



• Identify vertices: straight line code segments.



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  - 1 transitions to 2 unconditionally.
  - 2 self-transitions when length > 0.
  - 2 transitions to 1 when length  $\leq 0$ .

Step 2: from the control flow graph

$$1$$
  
length <= 0  
$$2$$
  
length > 0

Step 2: from the control flow graph, output the finite state machine.

i = j = 0;state = 1;length <= 0</pre> while (j < OUTLENGTH) {</pre> if (state == 1) { inchar = input[i++]; length = input[i++]; length > 0state = 2;} if (state == 2) { output[j++] = inchar; length--; if (length <= 0) { state = 1;} } }

Step 2: from the control flow graph, output the finite state machine.

The technique generalizes to break, continue, arbitrary nesting, sequential loops, etc.

The whole thing works by source-to-source translation: from a program with tested loops to one in FSM form, and from there into constraints.

The technique is detailed in Buffet [WSRBW15]; it is inspired by, and extends, loop flattening from the parallel compilers literature [GF95, KNP05, YCFVEEGH08, Knijnenburg98, Polychron87].

#### Caveats:

- Programmer must tell compiler # of steps to unroll the FSM.
- No "program memory"  $\Rightarrow$  no function pointers.

EXERCISE: Transform the code below to a FSM. Assume that a bound is known on the total number of iterations that your FSM will take.

```
// assume k is initialized earlier
// assume x is user-supplied input
while (j < MAX1) {
  k = k + 1;
   for (i = 0; i < x; i++) {
     if (i + j == k) {
       break;
     }
    j = j + 1;
   }
  j = j + 2;
}
```

A more general solution to data-dependent loop bounds preview [BCTV14b, BCGTV13]

The state variable in the FSM is like a coarse program counter ...

... what if the constraints modeled a program counter, registers, etc.?



Great programmability: handles all of C (but still requires bounded execution, because programmer selects # of CPU steps.)

An important question, when considering expressiveness, is how one represents RAM computations inside the circuit or constraint formalism. There are multiple approaches to this problem; time permitting, we may cover this topic.

For now, note that [BCTV14b] has an innovative solution, based on permutation networks, and assuming the "CPU approach". Buffet [WSRBW15] borrows this solution and adapts it to the "ASIC approach".

A self-contained, short description of [BCTV14b]'s solution is in section 2.3 of [WSRBW15].

(1) Arithmetization: from programs to constraints

(2) Enhancing expressiveness: data-dependent control flow

(3) Costs and comparisons

Costs arise from the front-end, the back-end, and their interaction

Goals:

- Understand concrete costs
- Understand the different amortization regimes
- Understand current trade-offs

Plan:

- Compare front-ends, by holding back-end constant
- Compare back-ends on two different circuits
- Examine various metrics (mostly running times)
- Examine the amortization regimes

## Front-end comparison

Back-end: libsnark, which is BCTV's [BCTV14b] optimized implementation of Pinocchio/GGPR [PGHR13, GGPR13].

Front-ends: implementations or re-implementations of

- Zaatar (ASIC) [SBVBPW13]
- BCTV (CPU) [BCTV14b]
- Buffet (ASIC) [WSRHBW15]



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Front-ends: implementations or re-implementations of

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- Buffet (ASIC) [WSRHBW15]

Evaluation platform: cluster at Texas Advanced Computing Center (TACC) Each machine runs Linux on an Intel Xeon 2.7 GHz with 32GB of RAM.

- (1) What are the verifier's costs?
- (2) What are the prover's costs?

Proof length	288 bytes	
V per-instance	$6 \text{ ms} + ( \mathbf{x}  +  \mathbf{y} ) \cdot 3 \mu \text{s}$	
V pre-processing	C  • 180 μs	
P per-instance	C ·60 μs + C log  C ·0.9μs	
P's memory requirements	$O( C \log C )$	
( C : circuit size)		

(3) How do the front-ends compare to each other?

(4) Are the constants good or bad?

How does the prover's cost vary with the choice of front-end?

Extrapolated prover execution time, normalized to Buffet



All of the front-ends have terrible concrete performance

Extrapolated prover execution time, normalized to native execution



The maximum input size is far too small to be called practical

	Zaatar	BCTV	Buffet
approach	ASIC	CPU	ASIC
m × m mat. mult	215	7	215
merge sort m elements	256	32	512
KMP str len: m substr len: k	m=320, k=32	m=160, k=16	m=2900, k=256

The data reflect a "gate budget" of  $\approx 10^7$  gates.

Pre-processing costs 10-30 minutes; proving costs 8-13 minutes

## Back-end comparison

- Data are from our re-implementations and match or exceed published results.
- All experiments are run on the same machines (2.7Ghz, 32GB RAM). Average 3 runs (experimental variation is minor).
  - For a few systems, we extrapolate from detailed microbenchmarks
- Benchmarks: 128×128 matrix multiplication and clustering algorithm

1. What is the per-instance verification cost?

2. What are the cross-over points?



3. What is the server's overhead versus native execution?

Verification cost sometimes beats (unoptimized) native execution.





number of instances

#### The prover's costs are rather high.



### Amortization comparison (of built systems)

Systems [CMT12, VSBW13, Thaler13] derived from [GKR08] require little or no amortization (but have some expressivity limitations)

Of the schemes that handle arbitrary circuits (that is, those based on arguments), preprocessing costs amortize differently. Ordered best to worst:

- 1. Bootstrapped GGPR-based SNARKs [BCTV14a, CTV15]
  - Constant preprocessing; amortize over all computations (but concrete costs to prover are extremely high).
- 2. BCTV [BCTV14b]: "CPU" front-end + non-interactive GGPR back-end
  - Amortize over all future computations of the same length
- 3. Pinocchio [PGHR13]: "ASIC" front-end + non-interactive GGPR back-end
  - Amortize over all future uses of a given computation
- 4. Zaatar [SBVBPW13]: "ASIC" front-end + interactive GGPR/IKO back-end
  - Amortize over a batch of instances of a given computation

## Summary of concrete performance

- Front-end: generality brings a concrete price (but better in theory)
- Verifier's "variable costs": genuinely inexpensive
- Verifier's "pre-processing": depends on setting
- Prover's computational costs: mostly disastrous
- Memory: creates scaling limit for verifier and prover

Performance is plausibly acceptable in certain settings ...

- It must be "regular" (to avoid setup costs), or there must be many identical instances (to amortize setup costs)
- The given computation needs to be small

... But none of the systems is at true practicality

#### Summary of front-ends



circuit is unrolled CPU execution [BCGTV13, BCTV14a, BCTV14b, CTV15]



each line translates to gates/constraints [SVPBBW12, SBVBPW13, VSBW13, PGHR13, BFRSBW13, BCGGMTV14, BBFR14, FL14,

KPPSST14, WSRBW15, CFHKKNPZ15]

"CPU"

- Verbose (costly)
- Good amortization
- Great programmability

"ASIC"

- Concise
- Amortization worse
- Programmability not bad

#### Summary of key concepts and points

- 1. Arithmetization: how to translate programs to equations
- Non-deterministic circuit/constraint models make this easier
- The process can be automated
- 2. Data-dependent control flow can be provided naturally in either a "CPU" front-end or an "ASIC" front-end
- Likewise for RAM operations
- 3. There are trade-offs among expressiveness, amortization behavior, and performance
- None of the implementations have achieved genuine practicality

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