Bonus Lecture #5: SNARKs Under the Hood

COMS 4995-001: The Science of Blockchains

URL: https://timroughgarden.org/s25/

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Goals for Bonus Lecture #5

1. Review of NP and SNARKs.

- SNARK = succinct, noninteractive argument of knowledge
- short (<< witness length) & easy-to-verify proofs of an NP statement

2. General probabilistic verification and the PCP Theorem.

every NP problem can be probabilistically verified

3. PCP Theorem → SNARKs.

can derive SNARKS from one of the deepest results in theory CS

4. Bird's-eye view of modern SNARK constrictions.

front ends, back ends, polynomial commitments, polynomial IOPs

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- -x is a "yes" instance if there exists a witness w with C(x,w)=1
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Idea of a SNARK: proof that x a "yes" instance, with proof length << witness length and verification time << time to compute C(x,w).

Definition: a SNARK for an NP problem (defined by C) is a way to generate short and easy-to-verify proofs π of existence of a witness.

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 - "prover time," ideally O(RT of C) with non-astronomical hidden constant
- if x a "no" instance, computationally infeasible to find π s.t. $V(x, \pi)$ ="yes"
 - i.e., practically impossible to convince verifier of a false statement

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Key ingredients: (cf., matrix multiplication)

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Question: how to probabilistically verify arbitrary computations?

Amazing fact: every NP problem can be probabilistically verified.

Problem: 3-SAT

Input: A list of Boolean decision variables x_1, x_2, \ldots, x_n ; and a list of constraints, each a disjunction of at most three literals.

Output: A truth assignment to x_1, x_2, \ldots, x_n that satisfies every constraint, or a correct declaration that no such truth assignment exists.

For example, there's no way to satisfy all eight of the constraints

$$x_1 \lor x_2 \lor x_3 \quad x_1 \lor \neg x_2 \lor x_3 \quad \neg x_1 \lor \neg x_2 \lor x_3 \quad x_1 \lor \neg x_2 \lor \neg x_3$$
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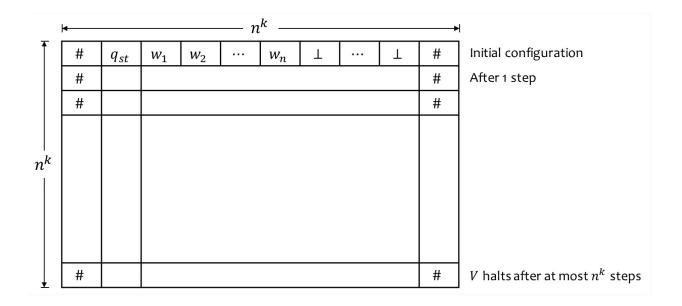
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Proof of the Cook-Levin Theorem

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- proof idea: decision variables = state of memory at each time step of computation; constraints = computation proceeds according to C



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Upshot: to probabilistically verify every NP problem, enough to probabilistically verify 3-SAT.

- verify an arbitrary NP problem by first converting it to 3-SAT

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Because SAT is NP-complete: every NP problem L can be likewise probabilistically verified. [Convert L to 3-SAT, use PCP theorem.]

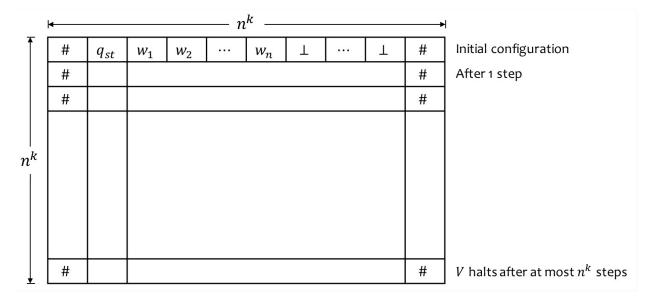
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Arbitrary NP problem (first attempt):

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Good news: (i) same correctness guarantees as matrix multiplication (ii) fast verification [O(t) evaluations of h, random accesses to y]

Bad news: proof y is not succinct (at least as large as witness length).

- y is essentially a redundantly encoded satisfying truth assignment

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Revised attempt: instead of posting y:

- prover forms a Merkle tree T with leaves = bits of y, posts root r of T
- also posts Merkle proofs revealing the answer to each query in each S_i , where S_i derived from $h(x \parallel y \parallel i)$ [or even $h(x \parallel y \parallel i)$ Merkle pfs for S_1, \dots, S_{i-1})]
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Result: correctness same as before (assuming no hash fn collisions), proof size now O(log RT(C)). [assuming t = O(1)]

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- not immediately practical (too much work to generate PCP proof y)
- but the conceptual basis for modern SNARK constructions

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Practical SNARKs: require careful joint optimization of the front end, the polynomial commitment scheme, and the polynomial IOP.