Understanding Computation with Computation

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Joint work with...

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Algorithm synthesis

- Computer science: what can be automated?
- Can we *automate our own work*?
- Can we outsource algorithm design to computers?
 - input: problem specification
 - **output:** asymptotically optimal algorithm

Verification and synthesis

- Verification:
 - given problem P and algorithm A
 - does A solve P ?
- Synthesis:
 - given problem P
 - find an algorithm A that solves P?

Verification and synthesis

- Algorithm verification often difficult
 - easy to run into e.g. halting problem
- Algorithm synthesis is entirely hopeless?
- Not necessarily!
 - verifying *arbitrary* algorithms in model *M*
 - synthesising only "*nice*" algorithms in model *M*

Setting

- Our focus: distributed algorithms
 - multiple nodes working in parallel
 - complicated interactions between nodes
 - possibly also faulty nodes, adversarial behaviour
- Computational techniques in algorithm design can outperform human beings

Setting

- We do theory, not practice
- Desired outputs:
 - algorithm design & analysis
 - Iower-bound proofs
- We want provably correct algorithms, not something that "seems to work"

Four success stories...

Success stories (1/4)

- Fault-tolerant digital clock synchronisation
 - nodes have to count clock pulses modulo c in agreement: all nodes say "this is pulse k"
 - self-stabilising algorithms: reaches correct behaviour even if the starting state is arbitrary
 - Byzantine fault tolerance: some nodes may be adversarial

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4 nodes

- 1 faulty node
- 3 states per node

always stabilises in at most 7 steps



Efficient computerdesigned solution for the **base case**

human-designed recursive step

efficient solution for the **general case**



Success stories (2/4)

- **Theorem:** any triangle-free *d*-regular graph has a cut of size $\left(\frac{1}{2} + \frac{0.281}{\sqrt{d}}\right)m$
 - prior bound: $\left(\frac{1}{2} + \frac{0.177}{\sqrt{d}}\right)m$ (Shearer 1992)
- Proof: we design a simple randomised distributed algorithm that finds such cuts (in expectation)

Pick a random cut, change sides if at least $\left[\frac{d+\sqrt{d}}{2}\right]$ neighbours on the same side



Success stories (3/4)

- Classical symmetry-breaking primitive:
 - input: directed path coloured with n colours
 - output: directed path coloured with 3 colours
- Prior work: $\frac{1}{2} \log^{*}(n) \pm O(1)$ rounds
- New result: exactly ¹/₂ log^{*}(n) rounds for infinitely many n

Success stories (4/4)

- Any locally checkable labelling problem
 - maximal independent set, colouring ...
- Setting: cycles, 2-dimensional grids, ...
- Complexity is O(1), $\Theta(\log^* n)$, or $\Theta(n)$
- Synthesis possible for class Θ(log* n)

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Key challenges

Key challenges

- A combinatorial search problem
 - find an object A that satisfies these constraints...
- How to make the problem finite?
 - so that the problem is *solvable at least in principle*
- How to solve it in practice?
 - how to avoid combinatorial explosion

Key challenges

- Much easier to make the problem finite if we fix some parameters:
 - algorithm for *n* = 10 nodes?
 - algorithm for any *n*, but maximum degree $\Delta = 10$?
- How to generalise?

How to generalise

1. Computer-inspired algorithms

• computer solves *small cases*, generalise the idea

2. Generalise by induction

• computer solves the *base case*, prove inductive step

3. Direct synthesis for the general case

• sit down and relax

How to generalise

- 1. Computer-inspired algorithms
 - example: large cuts
- 2. Generalise by induction
 - example: *clock synchronisation*
- 3. Direct synthesis for the general case
 - example: O(log* n)-time algorithms

A simple example

- Computer network = directed *n*-cycle
 - nodes labelled with O(log n)-bit identifiers
 - each round: each node exchanges (arbitrarily large) messages with its neighbours and updates its state
 - each node has to output its own part of the solution
 - time = number of rounds until all nodes stop
 - equivalently: *time = distance* (how far to look)

- LCL problems:
 - solution is globally good if it looks good in all local neighbourhoods
 - examples: vertex colouring, edge colouring, maximal independent set, maximal matching...
 - cf. class NP: solution easy to verify, not necessarily easy to find



- 2-colouring: inherently global
 - **O**(*n*) rounds
- 3-colouring: local
 - **O(log*** *n*) rounds



- Given an algorithm, it may be very difficult to verify
 - easy to encode e.g. halting problem
 - running time can be any function of *n*
- However, given an LCL problem, it is very easy to synthesise optimal algorithms!



- LCL problem ≈ set of feasible local neighbourhoods in the solution
- Can be encoded as a graph:
 - node = neighbourhood
 - edge = "compatible" neighbourhoods
 - walk ≈ sliding window





Neighbourhood v is "flexible" if for all sufficiently large k there is a walk $v \rightarrow v$ of length k ₃

- equivalent: there are walks of coprime lengths
- "12" is flexible here, $k \ge 2$





- Verification hard but synthesis easy:
 - construct graph, analyse its structure
- "Compactification":
 - any LCL problem can be represented concisely as a graph
 - seemingly open-ended problem of finding an efficient algorithm is reduced to a simple graph problem

Beyond cycles

Beyond cycles

- Classification undecidable on 2D grids
 - "is this problem solvable in O(log* n)"
- But 1 bit of advice is enough!
 - just tell me whether it is solvable in time O(log* n)
 - then I can find an optimal algorithm at least in principle, but often also in practice
 - key insight: "normal form" for any such algorithm

(92) (49)(26)33 (57 74 0 1 0 0 0 0 (62)(55)MIS 79 8 (48)(24) 0 0 0 0 1 0 f (60)(67) (15) (30)0 31 21 3 0 0 0 1 U (23)(47)(98)5 (95)0 17 0 0 0 1 0 0 (88)(87 (80)(25) (38)(20)(64) · 1 0 0 0 0 0 ' **1** ' (99)(91)(51)(69) 45 (61)0 0 0 0 0 1 (39)2 (58) (53)(63)(40)(16)0 1 0 0 0 0 1 **O**(log* *n*) **O(1)**

Key tools

- Domain-specific part:
 - constructing the concise representation
 - algorithms for *enumerating* all possible "neighbourhoods", "configurations", etc.
- Generic part:
 - efficient SAT solvers (and other solvers)
 - e.g. lingeling, picosat, akmaxsat

High-throughput algorithmics

- We can use computers to mass-produce data on computational complexity:
 - here are 2¹⁶ computational problems...
 - try to synthesise fast algorithms for all of them!
 - see where computers fail
 - find a *concise representation* of unsolvable cases
 - excellent starting point for human beings

Future

- How far can we push these techniques?
 - immediate next steps: distributed algorithms in much more general graph families
- More focus on *meta-algorithmics*?
 - how to design algorithms for designing algorithms
- Algorithms for *lower bounds*?

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