#### Jukka Suomela

Aalto University · Helsinki · Finland jukkasuomela.fi

# Algorithms that design algorithms?

## Computer science: *what can be automated?*

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Today: can we automate the study of distributed computing?

• **Question:** is there an efficient distributed algorithm for solving task *X* in model *M*?

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## Toy example: Locally checkable problems in cycles



#### • Computer network: cycle of n computers

- globally consistent orientation
- each node has one "successor" and one "predecessor"



## Setting

- Computer network: cycle of n computers
- Model of computing: LOCAL model
  - synchronous communication rounds
  - time = number of rounds until all nodes stop
  - unbounded message size
  - unlimited local computation
  - unique identifiers



## Setting

- Computer network: cycle of n computers
- Model of computing: LOCAL model
- Problem: any discrete problem you can define with local constraints
  - finite number of output labels
  - relation that tells which label sequences are valid

- independent set = no two neighbors selected
- maximal = cannot greedily add more



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$$\rightarrow 1 \rightarrow 0 \rightarrow 1 \rightarrow 0 \rightarrow 1 \rightarrow 0 \rightarrow 1 \rightarrow 0 \rightarrow 1 \rightarrow 0$$

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#### • Example: maximal independent set

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good

• Valid if you only see these: ???, ???, ????

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• Example: maximal independent set • independent set = no two neighbors selected not greedily add more Problem specification good

• Valid if you only see these: 001, 010, 100, 101

## Valid label sequences

- 2-coloring: 12, 21
- 3-coloring: 12, 21, 13, 31, 23, 32
- Independent set: 01, 10, 00

All possible output labelings in a window of size k

- *Maximal independent set:* **001, 010, 100, 101**
- Distance-2 coloring with 3 colors: 123, 132, 213, 231, 312, 321

## Fully automatic



• Write down the specification of any locally checkable problem X

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- Then you can *find efficiently* 
  - distributed round complexity of X
  - asymptotically optimal distributed algorithm for X



This algorithm solves X in time O(log\* n)

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Polynomial time (in the size of problem description)











Example: X = maximal independent set problem

1 0 0

0 0 1



1 0 0

0 0 1

1 0 1

010???

Compatible neighborhoods for adjacent nodes

1 0 0

0 0 1

1 0 1

0 1 0 1 0

Compatible neighborhoods for adjacent nodes

1 0 0

0 0 1

0

1



Compatible neighborhoods for adjacent nodes

1 0 0

0 0 1





1 0 0





1 0 0

1 0 1



0 1 0 - 1 0 0 1 0 1  $1 \quad 0 \quad 0 \quad \rightarrow \quad 0 \quad 0 \quad 1$ 

1 0 0

1 0 1



1 0 0

1 0 1


1 0 0

0 0 1

1 0 1



1 0 0

0 0 1

1 0 1



1 0 0

1 0 1

0 0 1





1 0 0

1 0 1

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## Let's draw this graph







































distance-2 coloring





















Algorithm: ?



## self-loop J solvable in O(1) rounds

## **Algorithm:** Constant output (e.g. here all-0)





Proof: ?





**Proof:** No self-loop  $\rightarrow$  any solution breaks symmetry everywhere





**Proof:** No self-loop  $\rightarrow$  any solution breaks symmetry everywhere  $\rightarrow$  can be used to find 3-coloring





**Proof:** No self-loop  $\rightarrow$  any solution breaks symmetry everywhere  $\rightarrow$  can be used to find 3-coloring  $\rightarrow$  not possible in  $o(\log^* n)$  rounds

























distance-2 coloring



Let's study walks that start and end here




























































































87

Can you find a self-returning walk of length 8?



97

Can you find a self-returning walk of length 9?



#### Self-returning walk of length k

## *k* = 5, 6, 7, 8, 9, ...





Decidable in polynomial time (how?)



solvable in
O(log\* n) rounds



Algorithm: ???

solvable in
O(log\* n) rounds



Find markers separated by ≥ k<sub>0</sub> hops Use flexible configuration around markers Follow a selfreturning walk to fill in between markers 0



solvable in
O(log\* n) rounds

#### **Algorithm:**

- split in blocks of length  $\geq k_0$
- use the flexible configuration at each block boundary
- fill in between boundaries by following a self-returning walk

"Flexible": for all  $k \ge k_0$ there is a selfreturning walk of length k solvable in



solvable in O(log\* n) rounds **Proof:** Not flexible  $\rightarrow$  must use the same non-flexible configuration at least twice far from each other; not compatible for all distances  $\rightarrow$  global coordination needed  $\rightarrow$  not possible in o(n) rounds













distance-2 coloring











distance-2 coloring



independent set





*O*(log\* *n*)





distance-2 coloring



independent set





*O*(log\* *n*)





O(n)

## Fully automatic

- Write down the specification of any locally checkable problem X
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This algorithm solves X in time O(log\* n)

# Can we generalize beyond directed cycles?



Grids





#### solution ≈ execution history of a **finite automaton**







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solution ≈ execution history of a **Turing machine** 



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#### Grids

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#### Grids

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Many questions undecidable

## Undecidable # hopeless

## Normal forms

Any algorithm **A** that solves a locally checkable problem X fast can be written as  $\mathbf{A} = \mathbf{B} \circ \mathbf{C}_{\mathbf{k}}$ 

- $C_k$  = distance-k coloring
- **B** = finite function that maps colored neighborhoods to local outputs

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**Proof idea:** Coloring  $\approx$  locally unique identifiers. If *A* fails with such fake identifiers, it also fails in some small graph with some real identifiers.
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### For each *k* = 1, 2, 3, ...:

- check all possible candidate functions **B**
- if any of them is good  $\rightarrow$  fast algorithm found!

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#### **Undecidability:** *don't know when to stop if fast algorithms don't exist*

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C<sub>k</sub> = distance-k
 B = finite funct
 Computational complexity:
 neighborhoods
 typically doubly-exponential in k

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- We can make it more feasible in practice:
  - more "compact" normal forms, e.g. distance-k coloring  $\rightarrow$  ruling set
  - represent "candidate B is good for this value of k" as a Boolean formula and use modern SAT solvers to find such a B

- Example: *4-coloring in grids*
- Computers were much faster than human beings in figuring out that this is solvable in  $O(\log^* n)$  rounds

#### Cycles, paths

#### solution ≈ execution history of a **finite automaton**



#### Grids

#### solution ≈ execution history of a **Turing machine**

Many questions undecidable (but there is hope!)

#### Cycles, paths

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#### **Grids + beyond**

solution ≈ execution history of a **Turing machine** 

Bad news apply to any graph family that contains large grids

#### Cycles, paths

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#### **Grids + beyond**

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What is here between paths and grids?

# Big picture: meta-computational questions and algorithms synthesis

# Meta questions

. . .

- Designing algorithms that design algorithms?
- Studying the computational complexity of studying computational complexity?
- Using computation (in practice) to understand computation (in theory)?

# **Verification & synthesis**

### Algorithm verification:

- given problem *P* and algorithm *A*
- does A solve P?

### Algorithm synthesis:

- given problem P
- find an algorithm A that solves P?

# **Verification & synthesis**

- Algorithm verification often hard
  - recall: halting problem
- Algorithm synthesis can be easier!
  - verification must handle arbitrary algorithms
    synthesis can produce "nice" algorithms

# Conclusions

### **Take-home messages**

- Algorithm design can be made systematic and mechanical, even computers can do it!
  - we need the right *representations* for computational problems & algorithms
  - this is **not machine learning** but is this Al?
- Key concepts:
  - meta-computational problems
  - algorithm *verification* vs. algorithm *synthesis*