Software Architecture III
Leveraging Nature to Build Better Systems

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Outline

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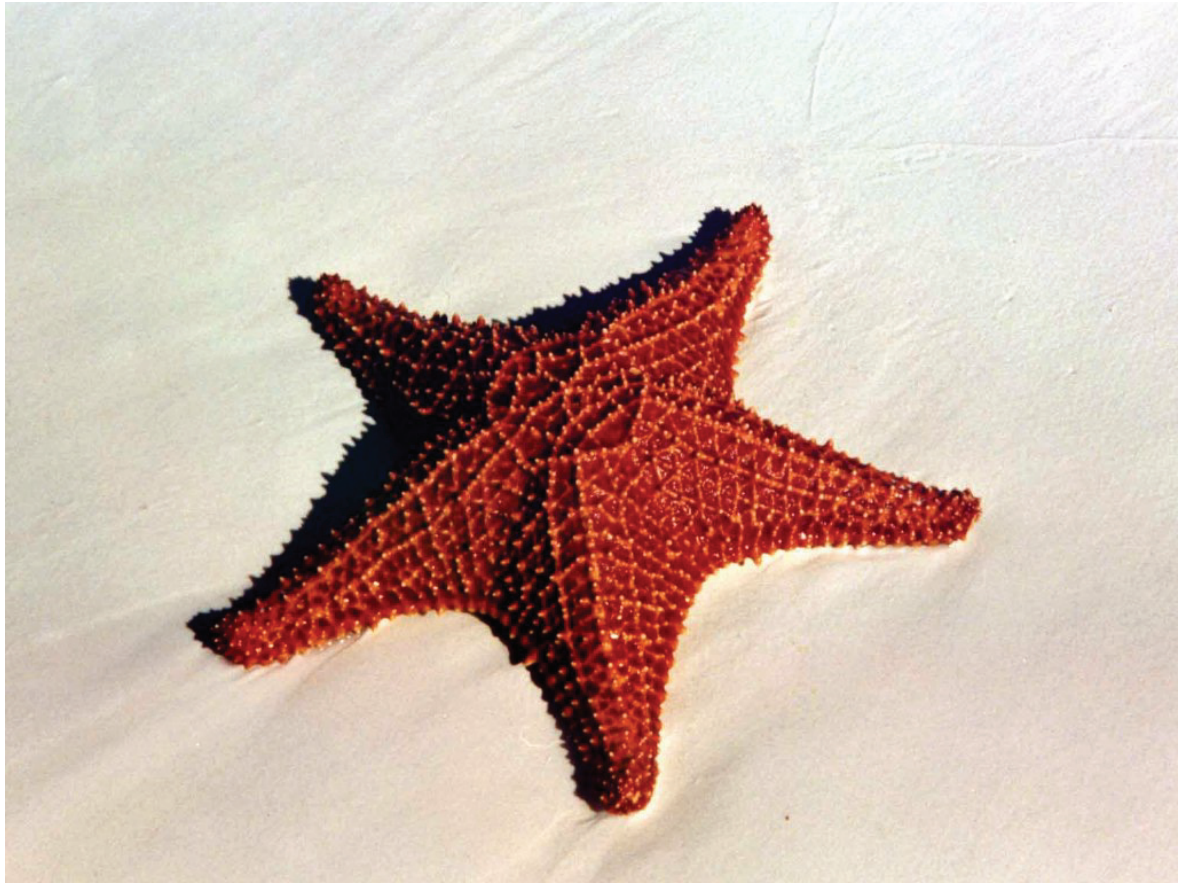
Systems in Nature

- Resilient to
  - death
  - malfunction
  - malicious agents
- Self-healing
- Fault-tolerant



In Contrast: Software

- Less-complex systems
- Fault-tolerance is "intelligently designed"
- Not expected to recover from catastrophes

Genetic Algorithms

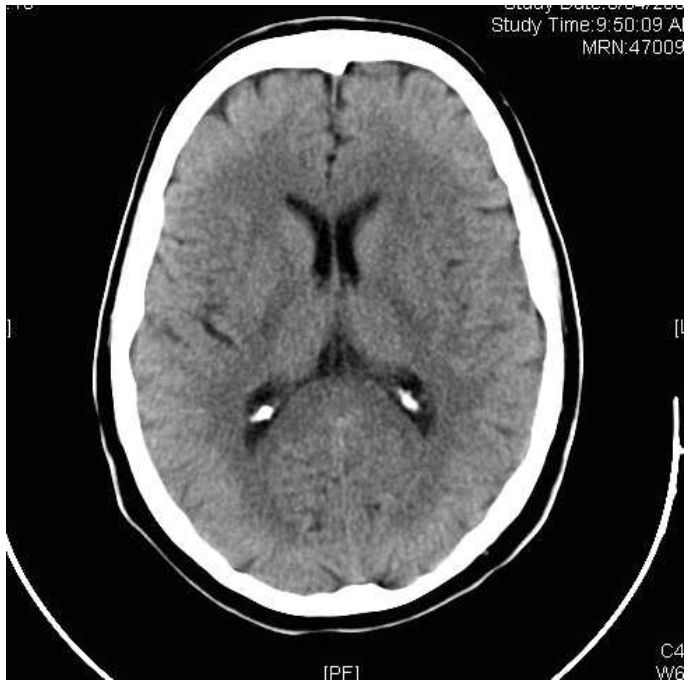
Have been used to:

- Design of fighter-planes airfoils [HO03]
- Train scheduling
- Automatic software bug patching [WNGF09]
- Data mirroring [RKCM09]

Neural Networks

Have been used to:

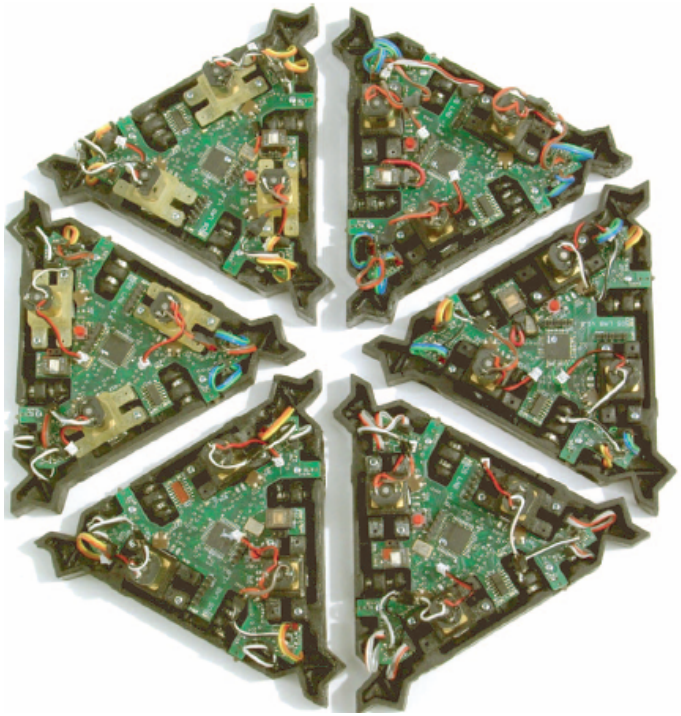
- Classification
- Sales forecasting / marketing
- Medical diagnoses [SKR01]
- Credit evaluation [Wes00]



Distributed Robotics

Have been used to:

- Search and rescue scenarios [MEB+10]
- Vacuum design
- Sensor networks [AAC+00]
- Education



Robofish

Outline

Leonard M. Adleman

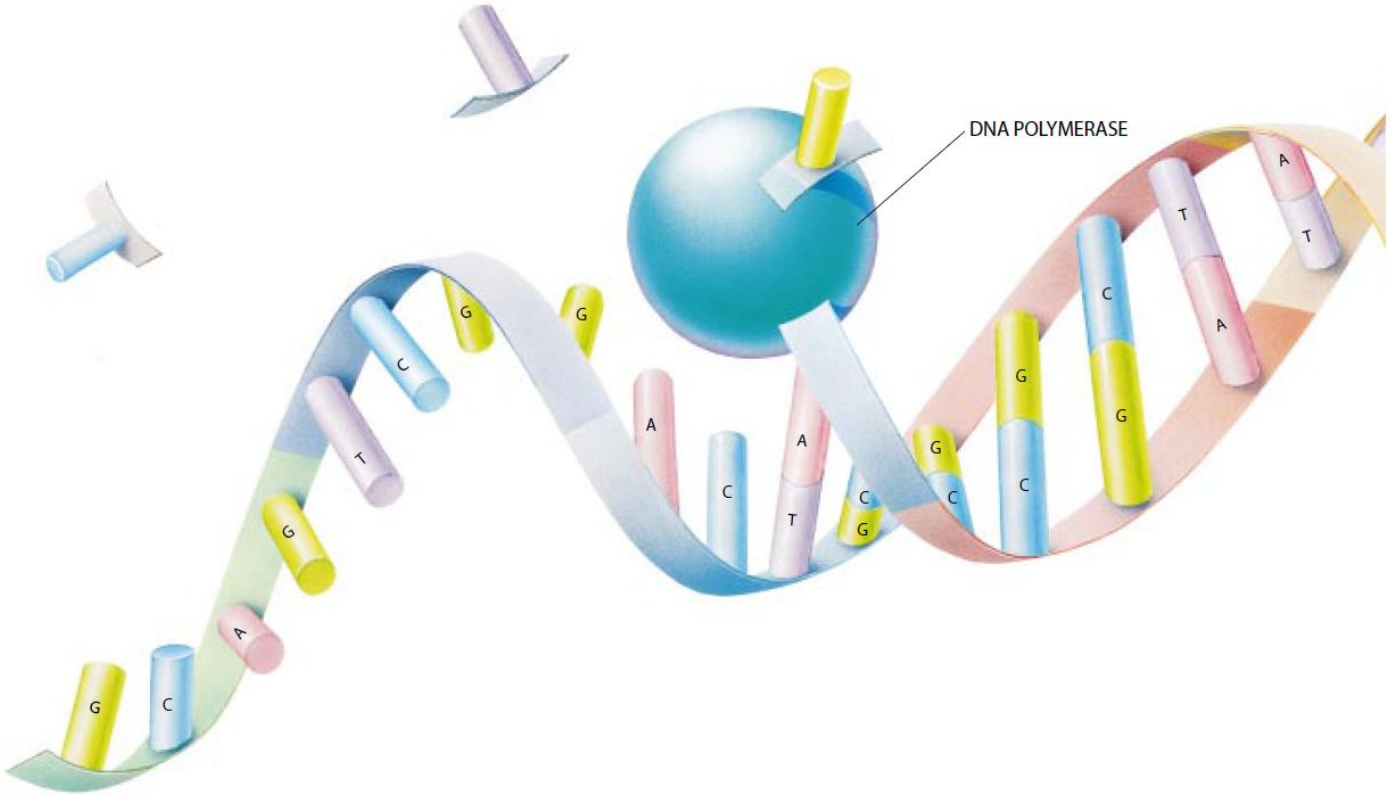
"The manipulation of DNA to solve mathematical problems is redefining what is meant by 'computation'."

A Bit of History

Adleman's research

- RSA public key cryptosystem [RSA78]
- Computer viruses [Adl90]
- HIV modeling [AW93]
- DNA computation [Adl94]

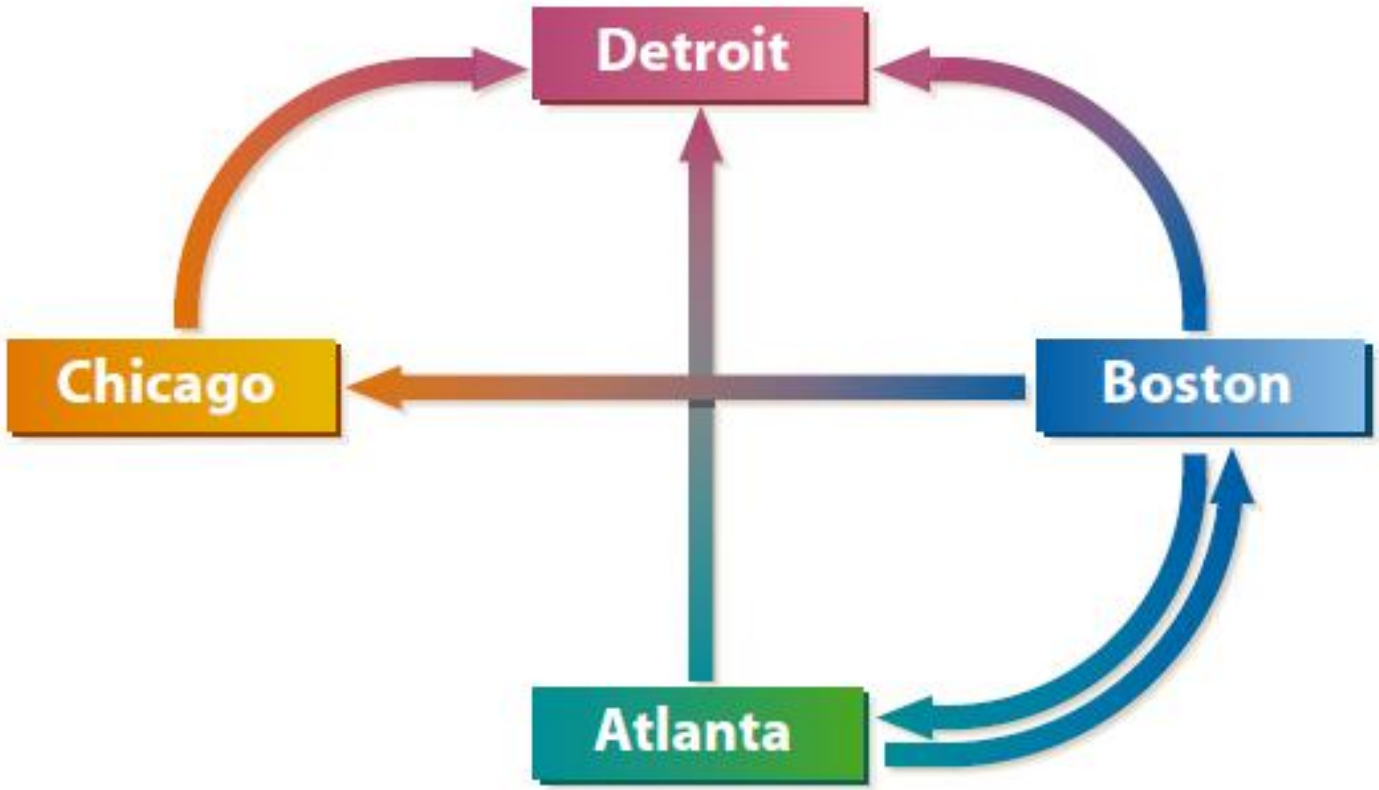
DNA Turing Machine



[Adl98]

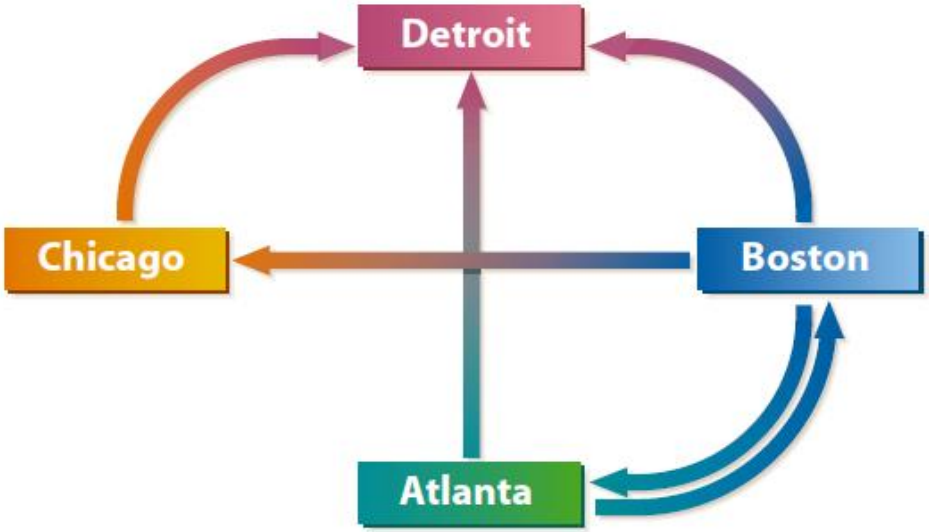
- A, T, C, and G can encode information
- A DNA strand is a data-storing tape
- Enzymes can encode states and rules for manipulating the tape

Hamiltonian Path Problem



[Adl98]

Hamiltonian Path Problem with DNA



CITY	DNA NAME	COMPLEMENT
ATLANTA	ACTTGCAG	TGAACGTC
BOSTON	TCGGACTG	AGCCTGAC
CHICAGO	GGCTATGT	CCGATACA
DETROIT	CCGAGCAA	GGCTCGTT

FLIGHT	DNA FLIGHT NUMBER
ATLANTA - BOSTON	GCAGTCGG
ATLANTA - DETROIT	GCAGCCGA
BOSTON - CHICAGO	ACTGGGCT
BOSTON - DETROIT	ACTGCCGA
BOSTON - ATLANTA	ACTGACTT
CHICAGO - DETROIT	ATGTCCGA

[Adl98]

Implementing the DNA Algorithm

	Conventional algorithm	**DNA algorithm**
1.	Generate a set of random paths	Mix city and flight strands
2.	Select paths that start and end at proper cities	PCR
3.	Select proper-length paths	Electrophoresis gel
4.	Select paths that visit each city	Watson & Crick pairing
5.	The remaining paths represent the solution	PCR, electrophoresis, and sequencing

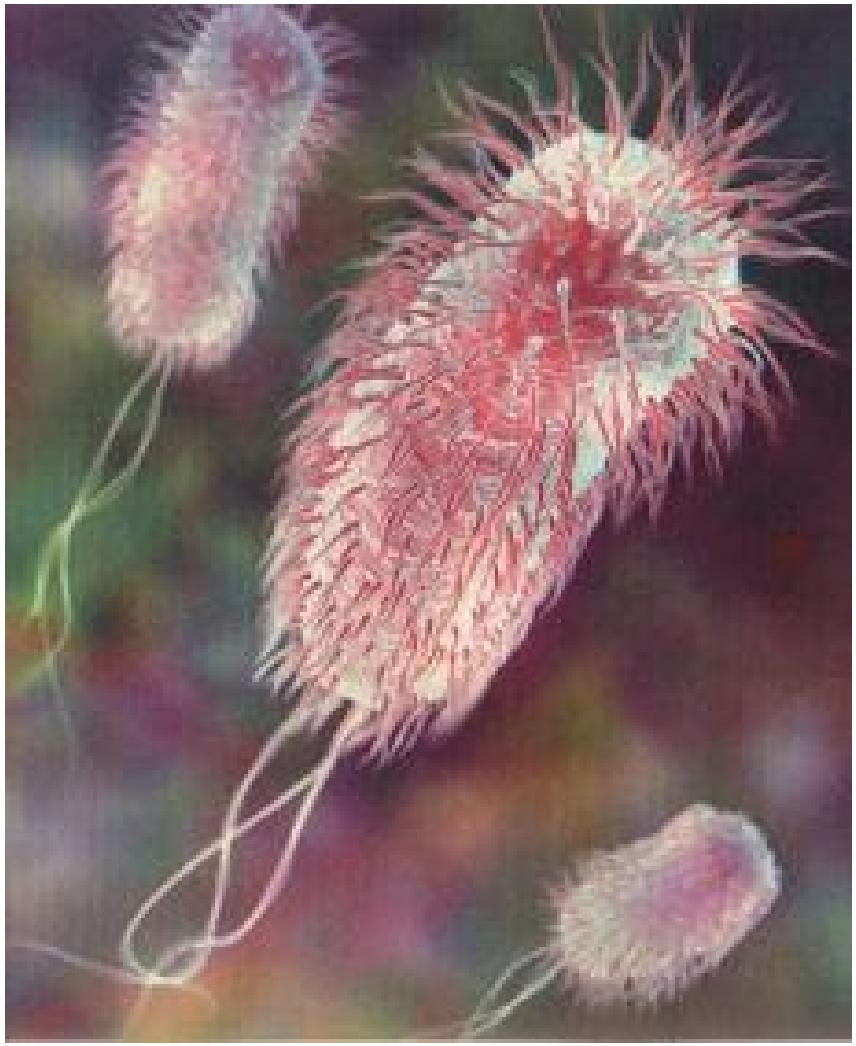
3-SAT With DNA

- In 2002, Braich et al. [BCJ+02] developed a DNA computer to solve 20-variable 3-SAT problems.
  - Worked most of time
  - Error rates grew proportionally to the number of variables
- A few other models emerged
  - Sticker model
  - Tile assembly model
    - more on this later...

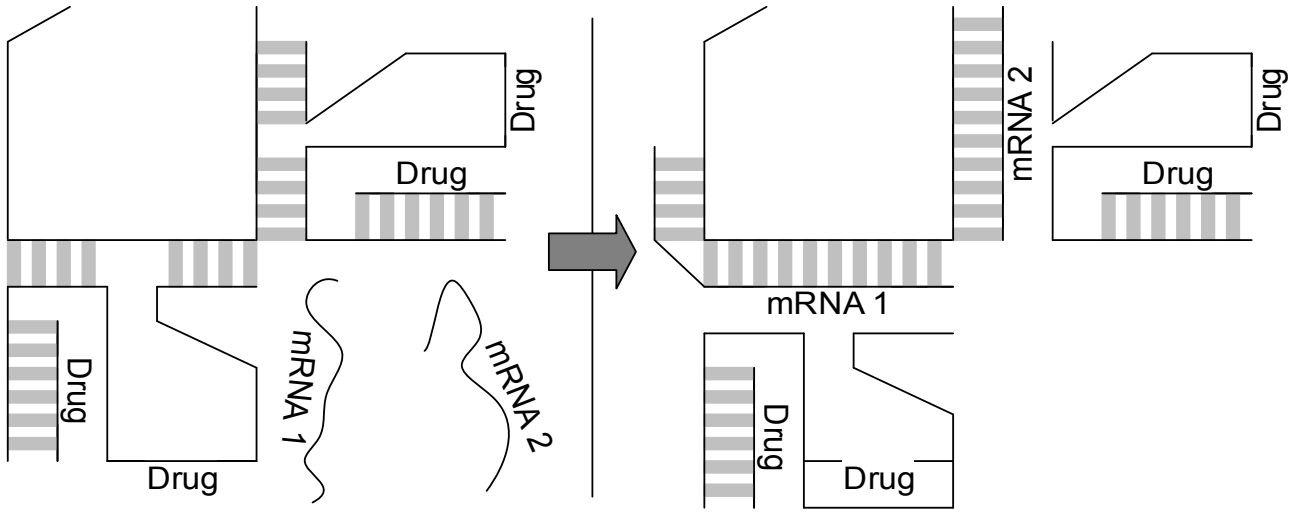
Protein Production Control

BioBricks [KS97]

- Controlling what proteins a cell produces
- Basis for the International Genetically Engineered Machine (iGEM) competition

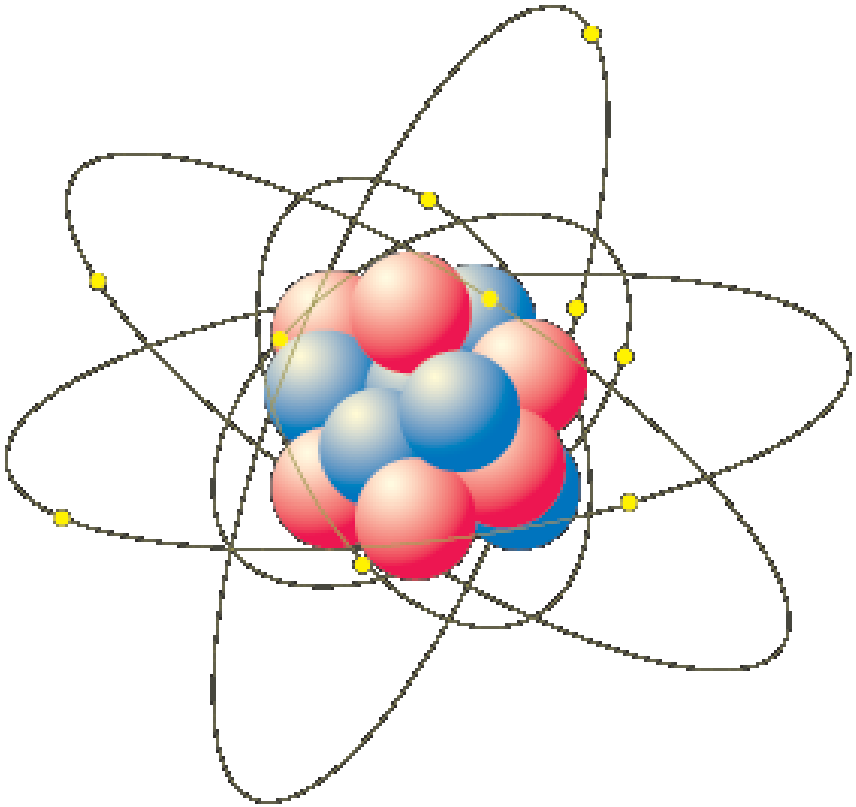


DNA Gates



- Binary gates that act on DNA-strand inputs [BG06]
- Previous work used enzymes [BGBD+04]
- Later work at Caltech improved the design [QW08]

Self-Assembly in Nature



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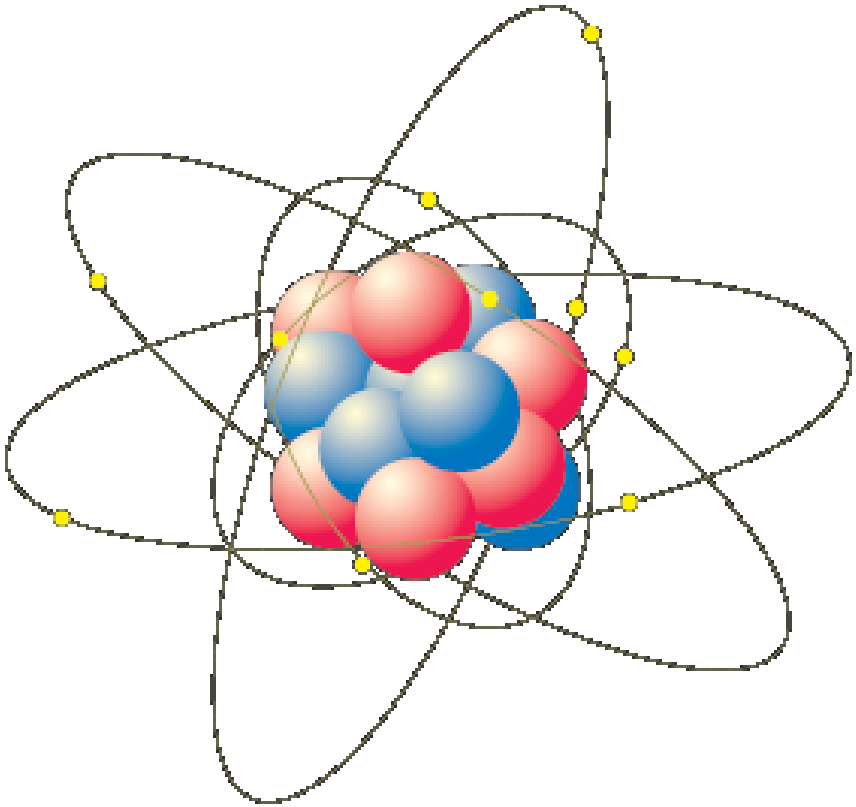

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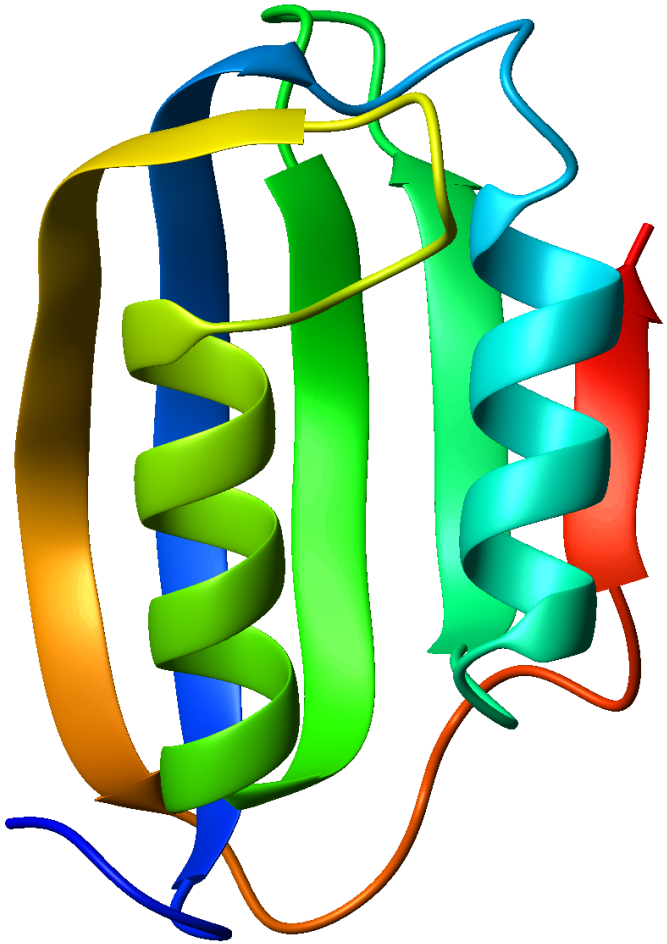
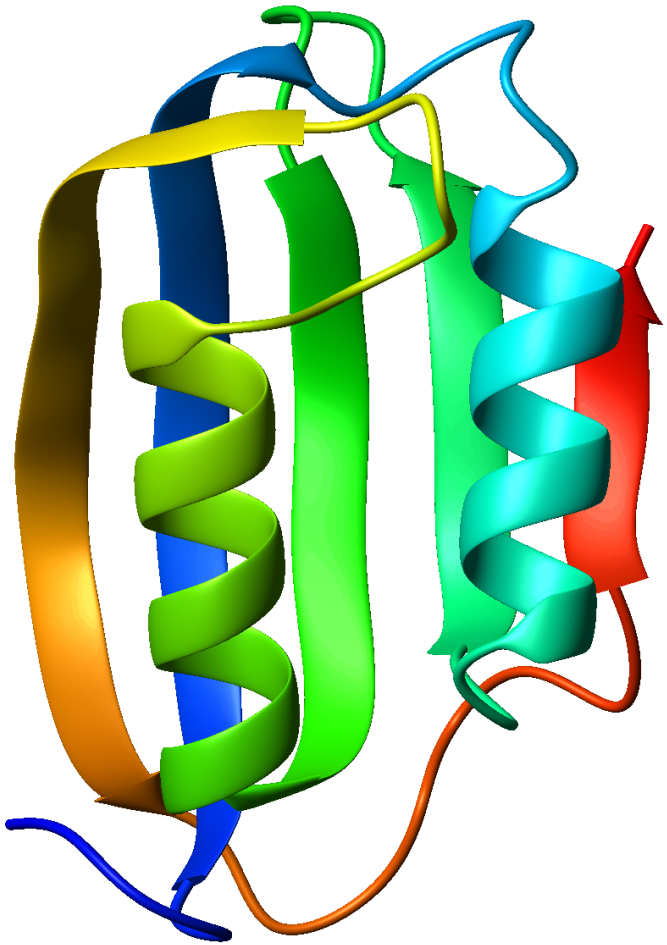
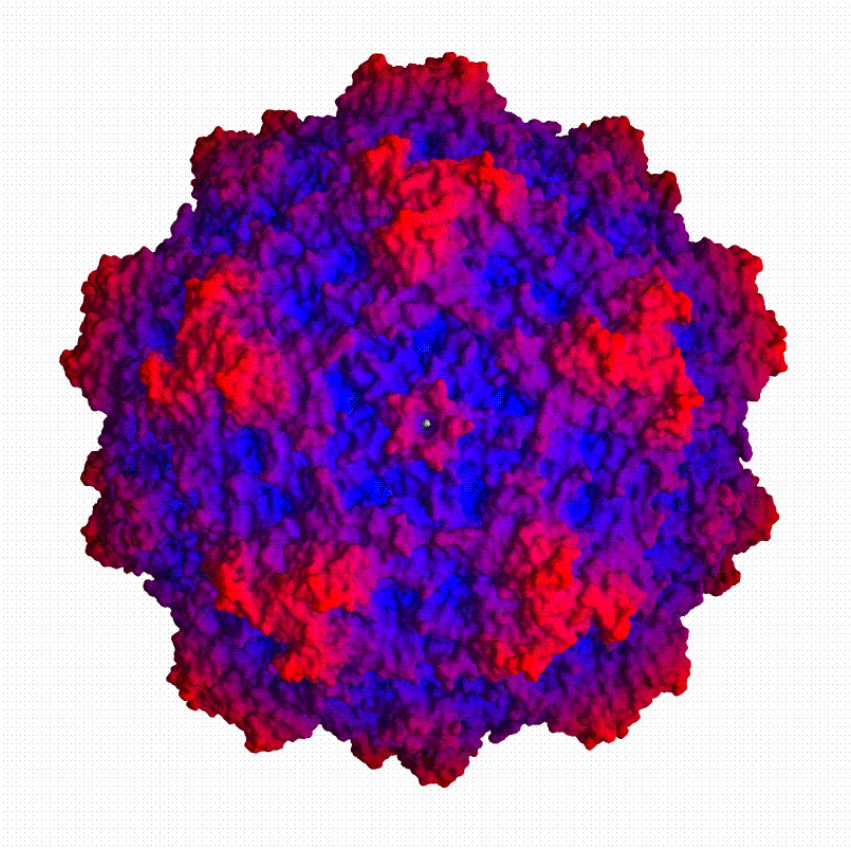


Self-Assembly in Nature





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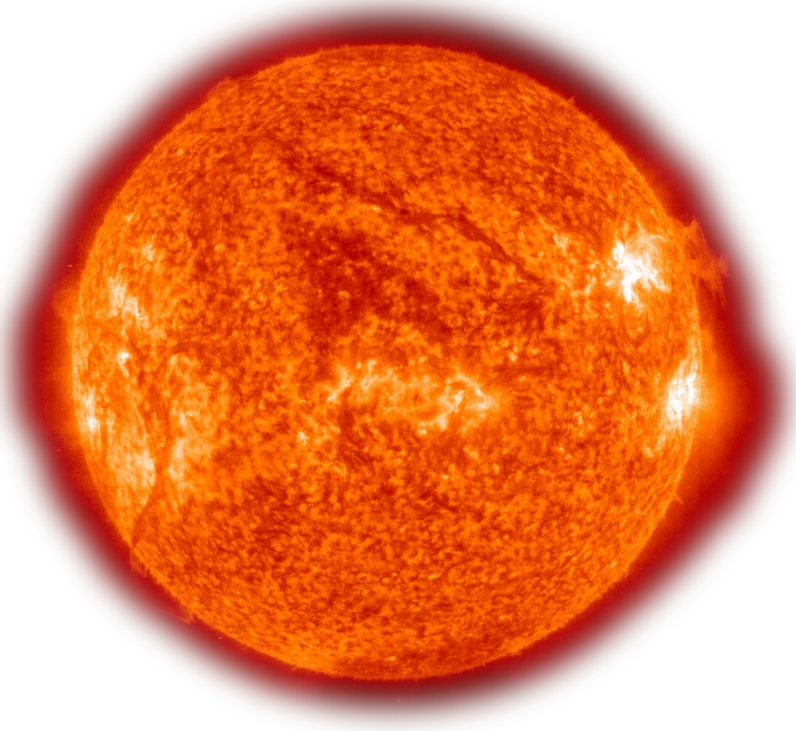

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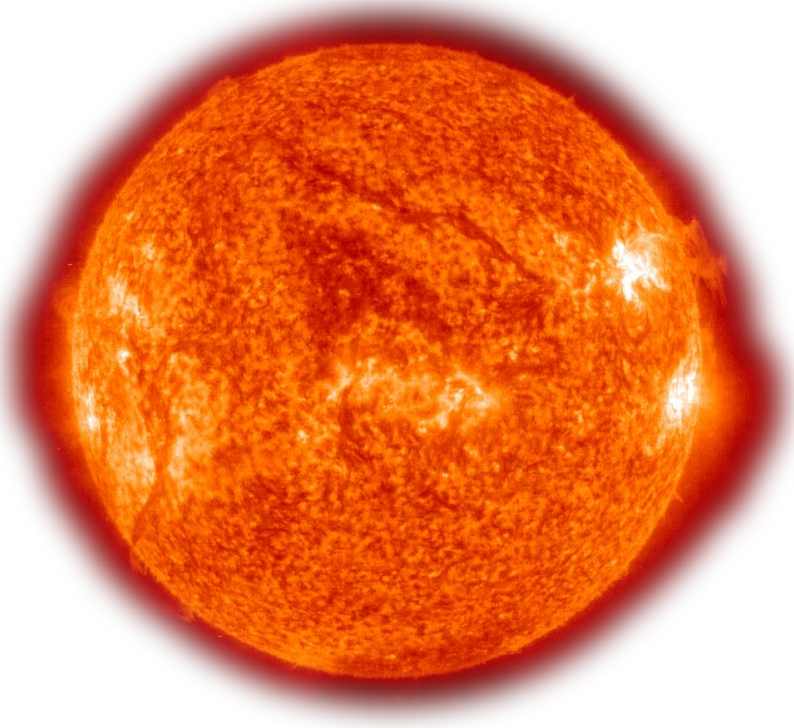
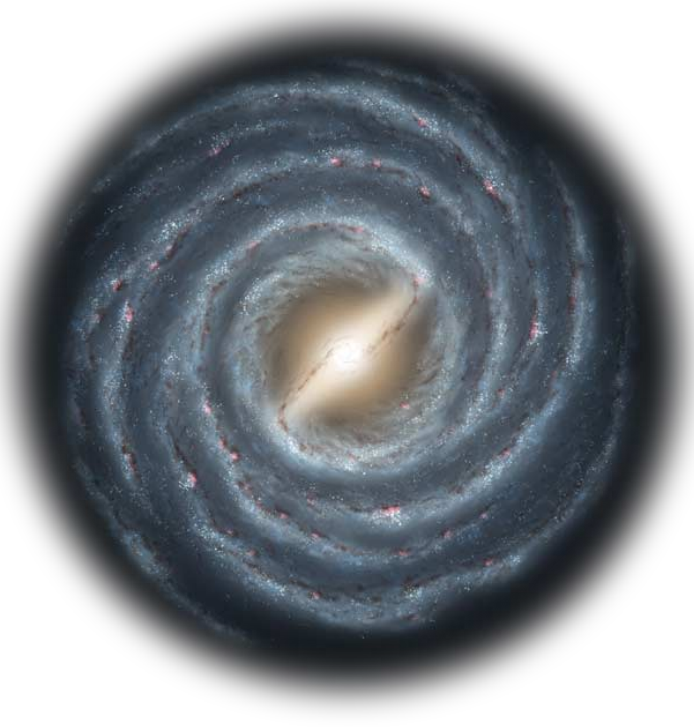
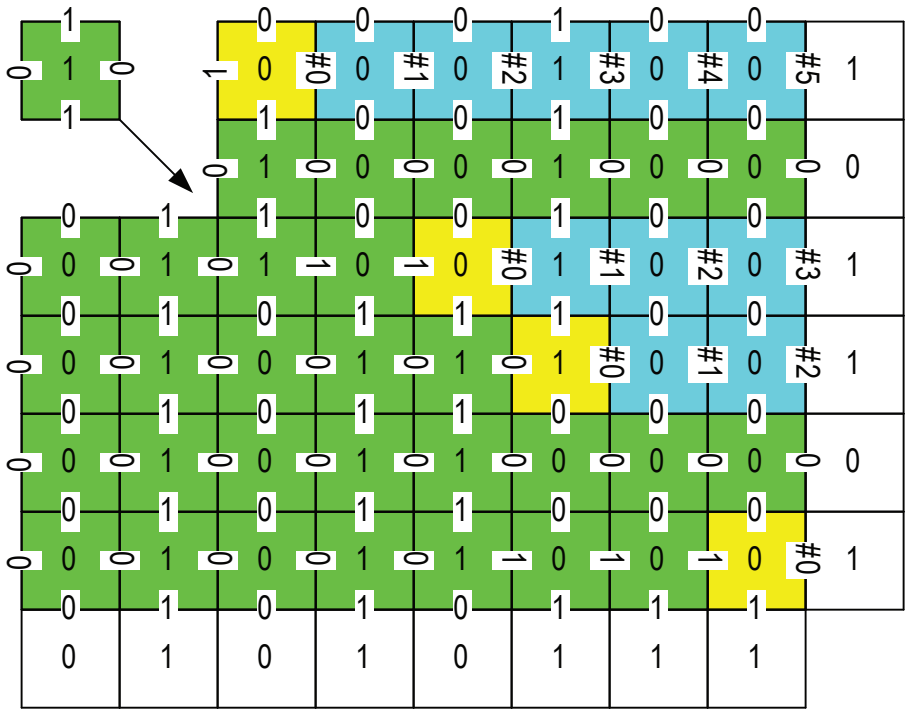




Tile Assembly Model [Win98b]

- Tile: a square with labels
- Each label has a strength
- Tiles attach if labels are strong enough



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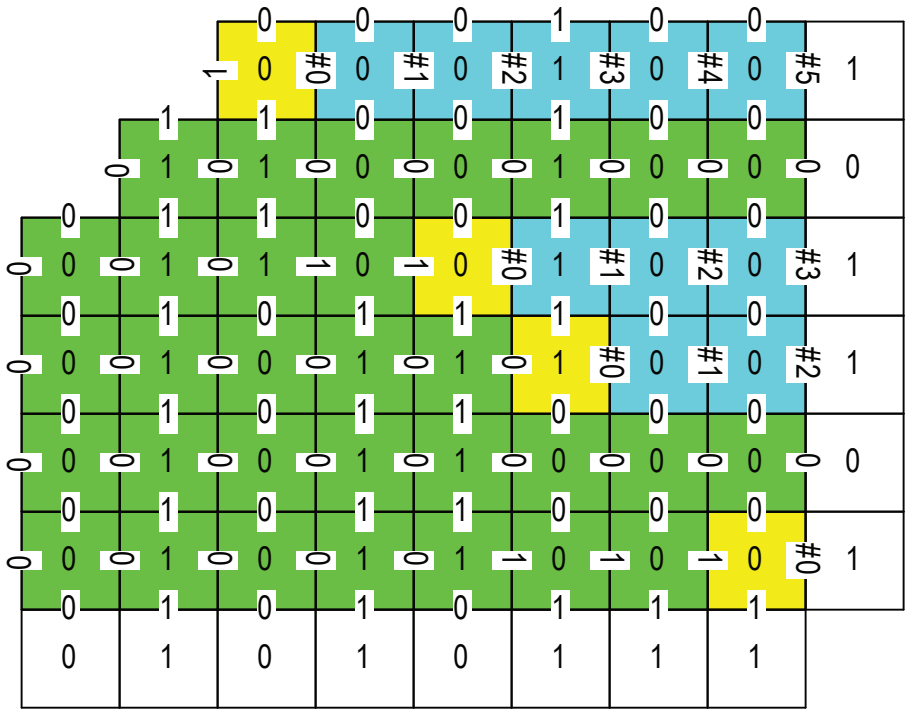
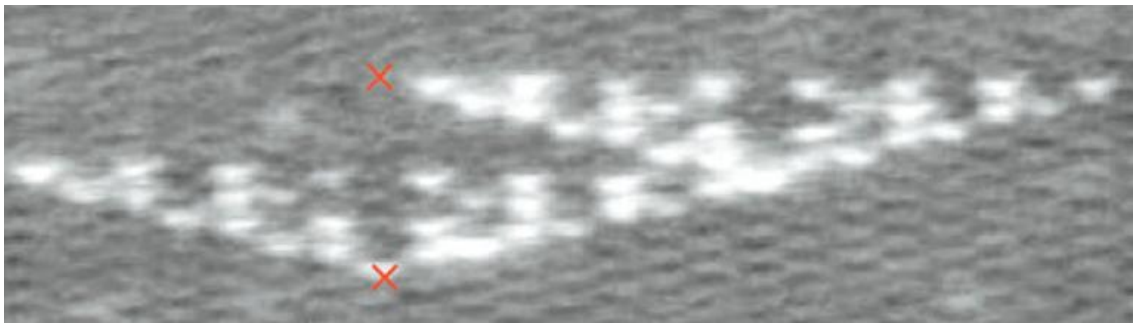


Tiles Can:

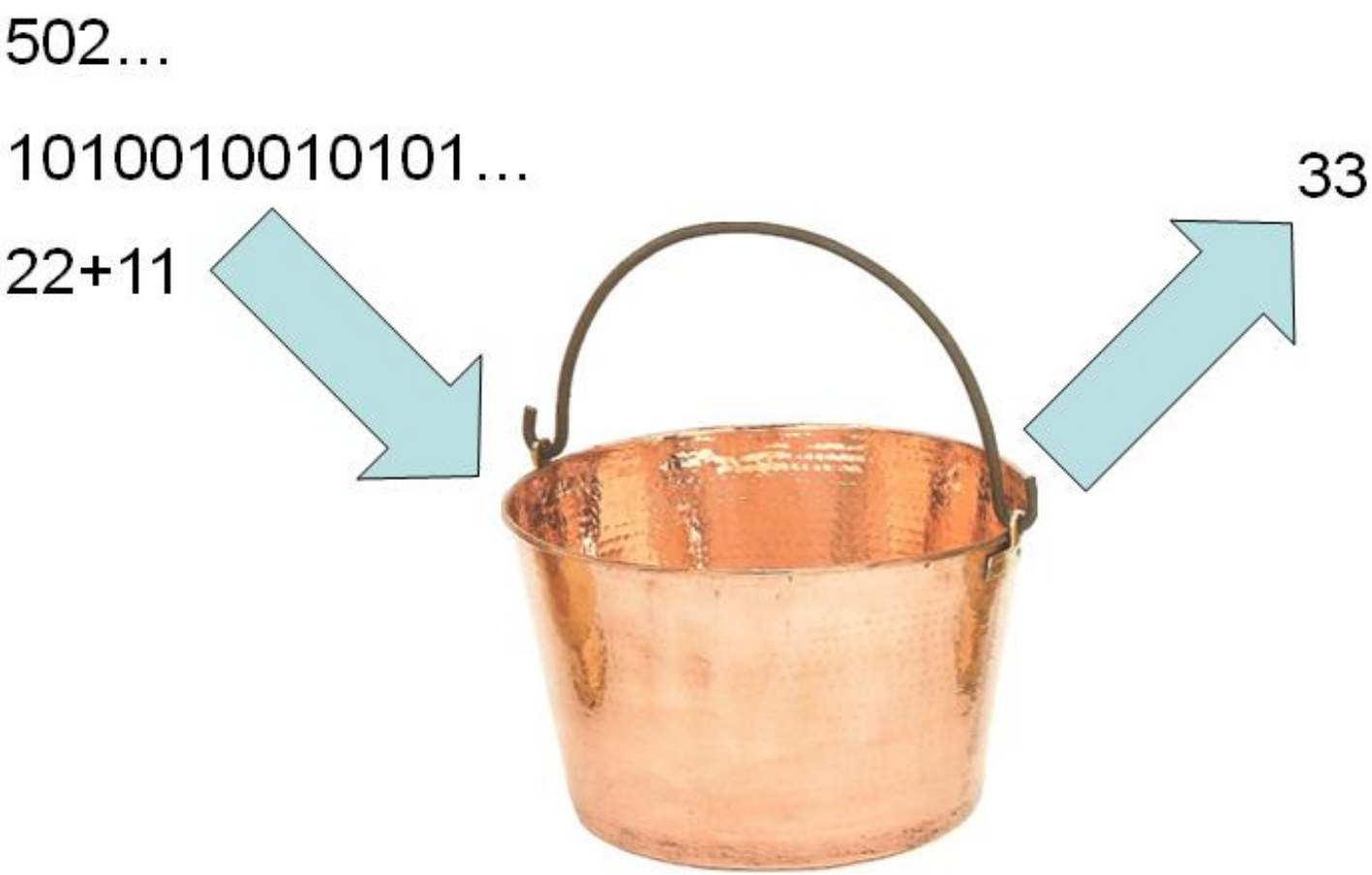
- Assemble
  - linear polymers [ACG+01]
  - squares [RW00, AGHM02, ACG+02]
  - computable shapes [SW07]
- Count [Win98a, Moi05, BRW05]
- Compute Binomial Coefficients [Win98a, RPW04]



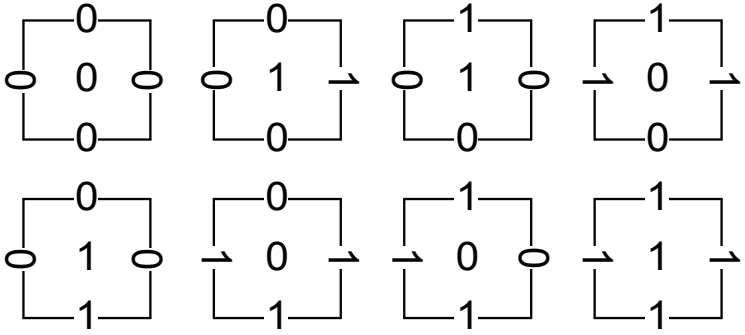
- Emulate Turing Machines [Win98b]

Outline

Computing with Tiles



Adding with Tiles [Bru07]



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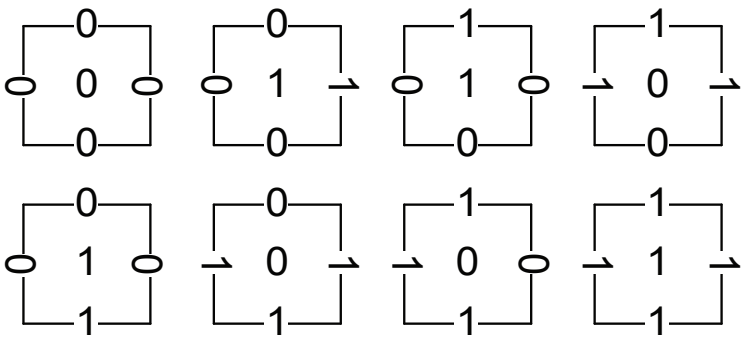
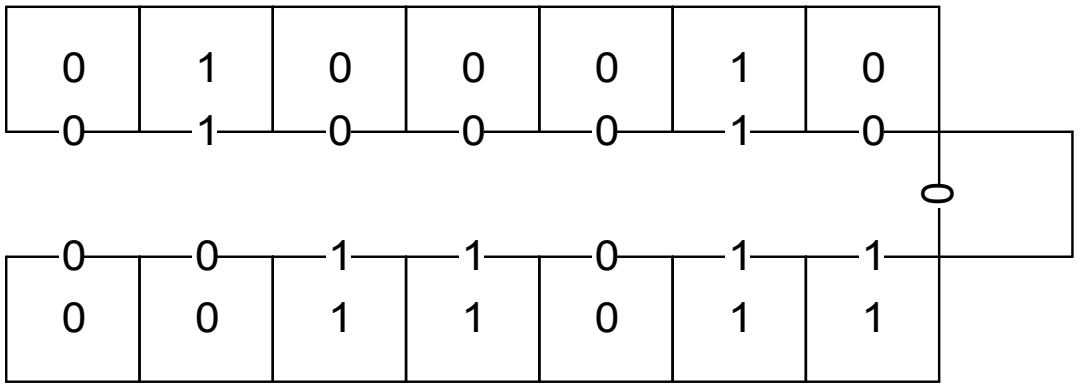


$$34 + 27$$



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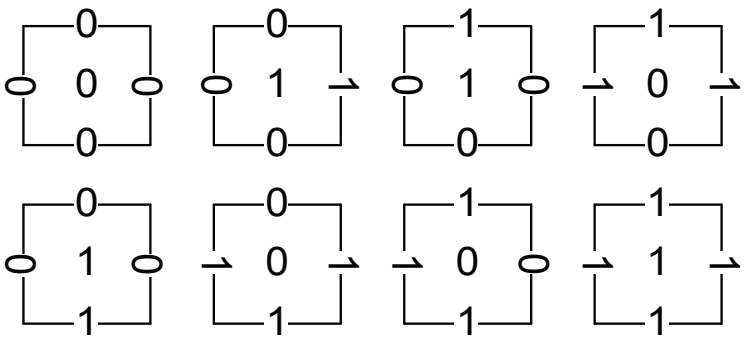
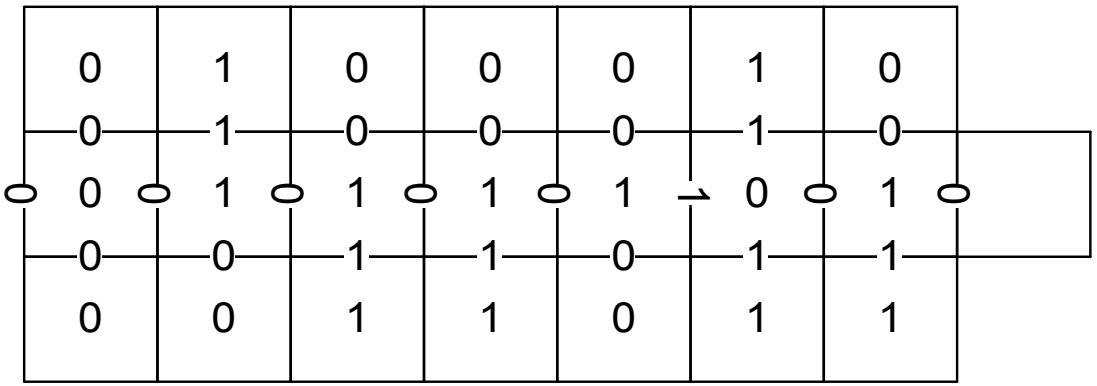
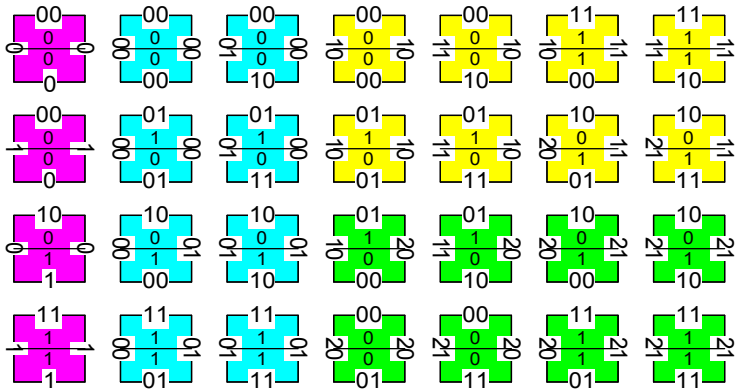


$$34 + 27 = 61$$



Multiplying with Tiles [Bru07]



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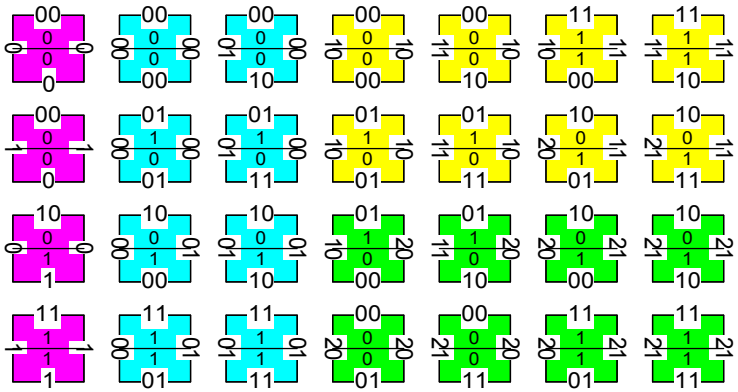
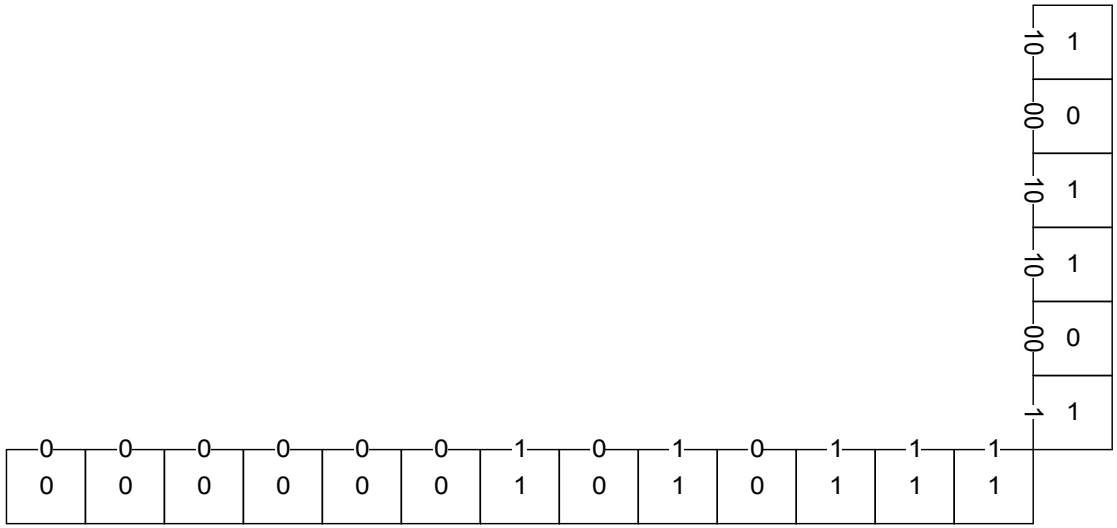


87×45



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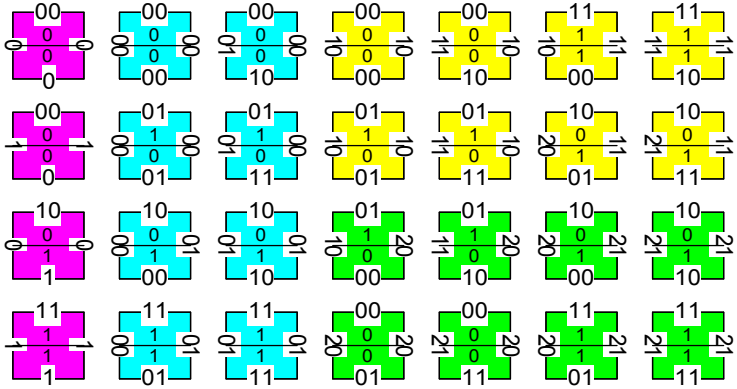
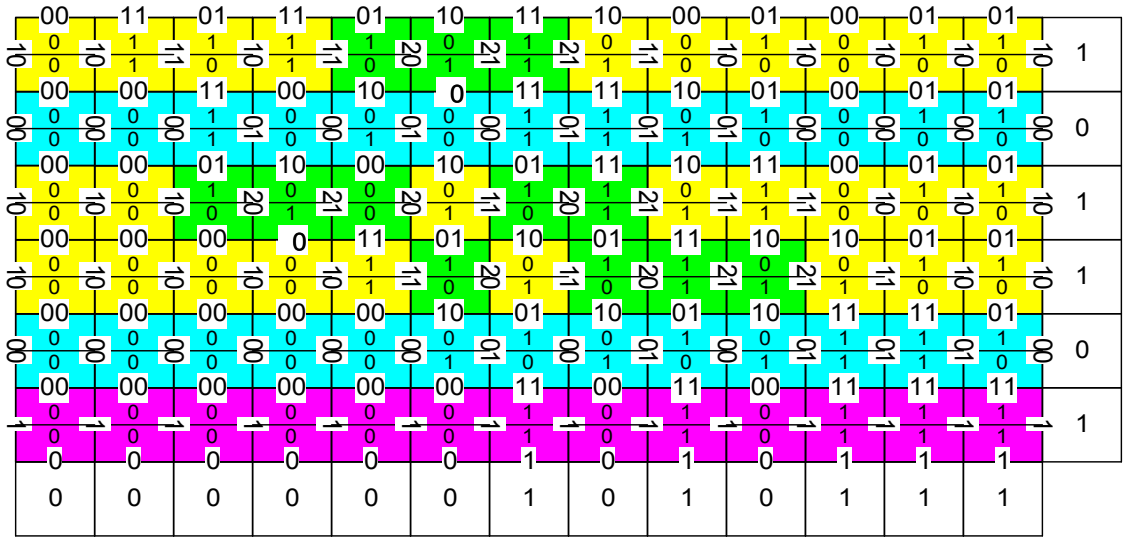


$$87 \times 45 = 3915$$



3-SAT

Variables: $x_0,\ x_1,\ x_2\ \ldots$

Literals: $x_0,\ \neg x_0,\ x_1,\ \neg x_1,\ \ldots$

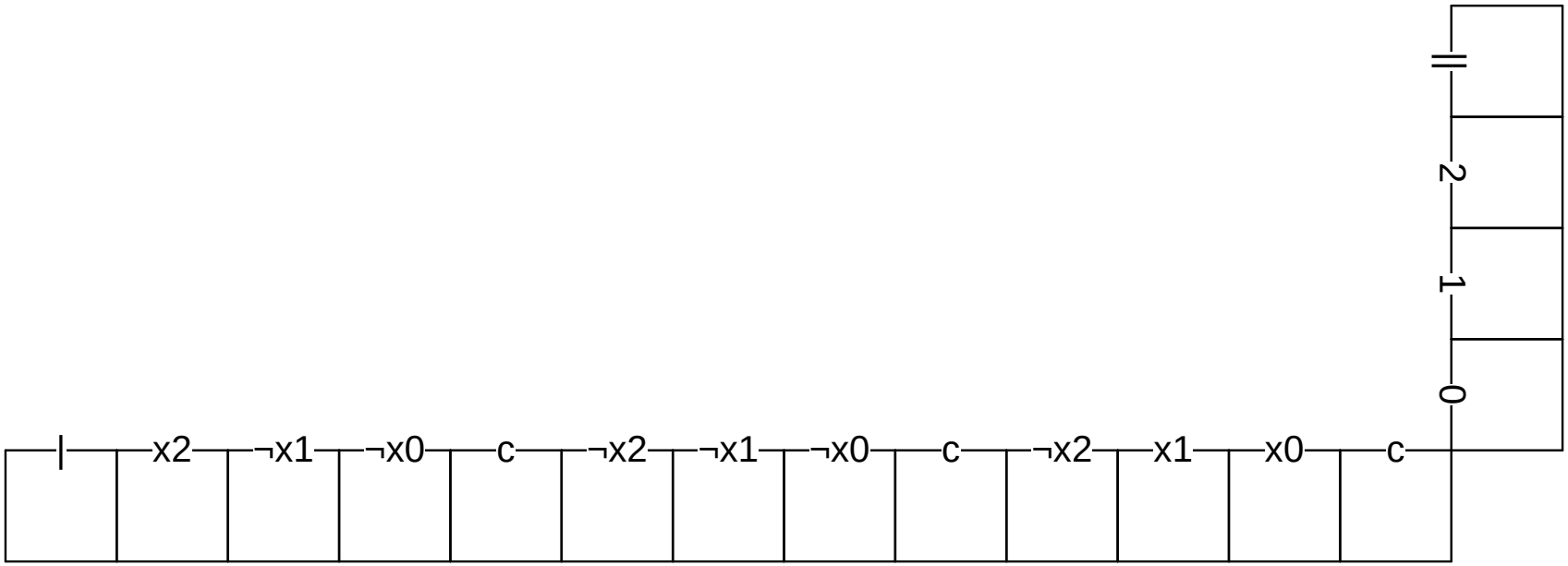
Clauses: $(x_2 \vee \neg x_1 \vee \neg x_0)$

Formula: $(x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee \neg x_1 \vee \neg x_0)$

The question: Does there exist an assignment of $TRUE$ / $FALSE$ values to the variables that makes the formula $TRUE$?

$\Theta(n^2)$-Tileset Approach [LL99]

$$(x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee x_1 \vee x_0)$$



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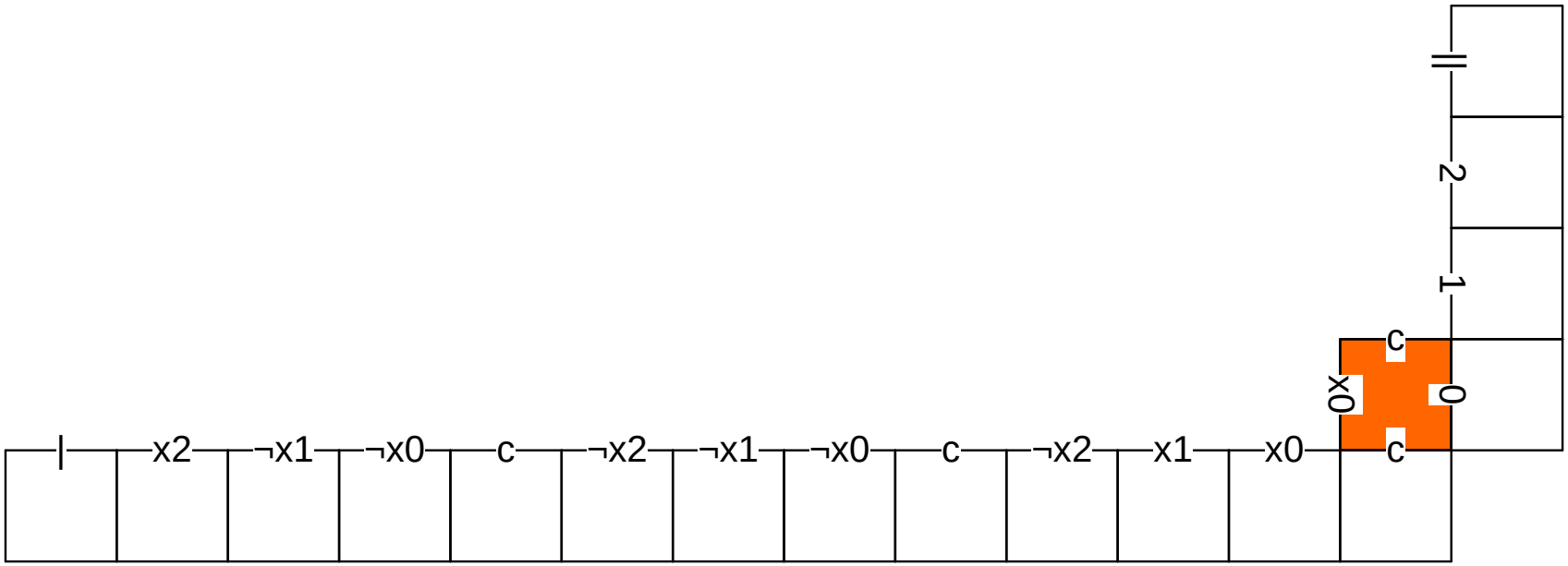

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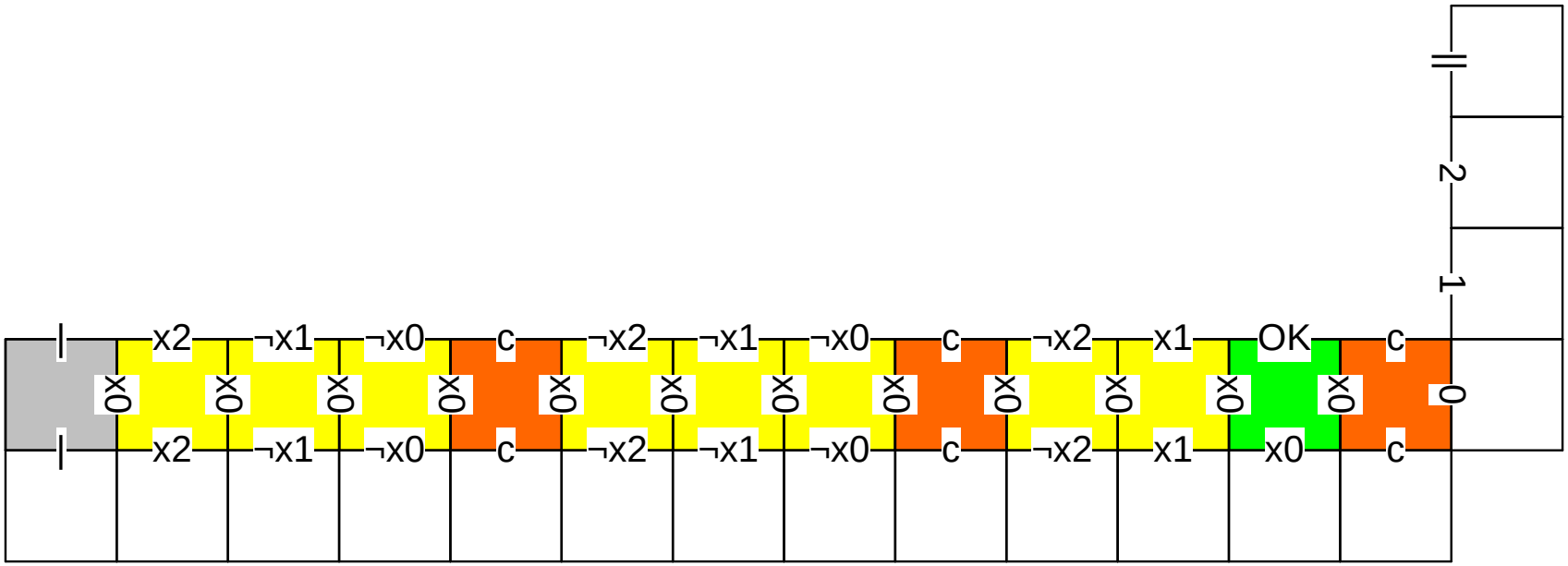

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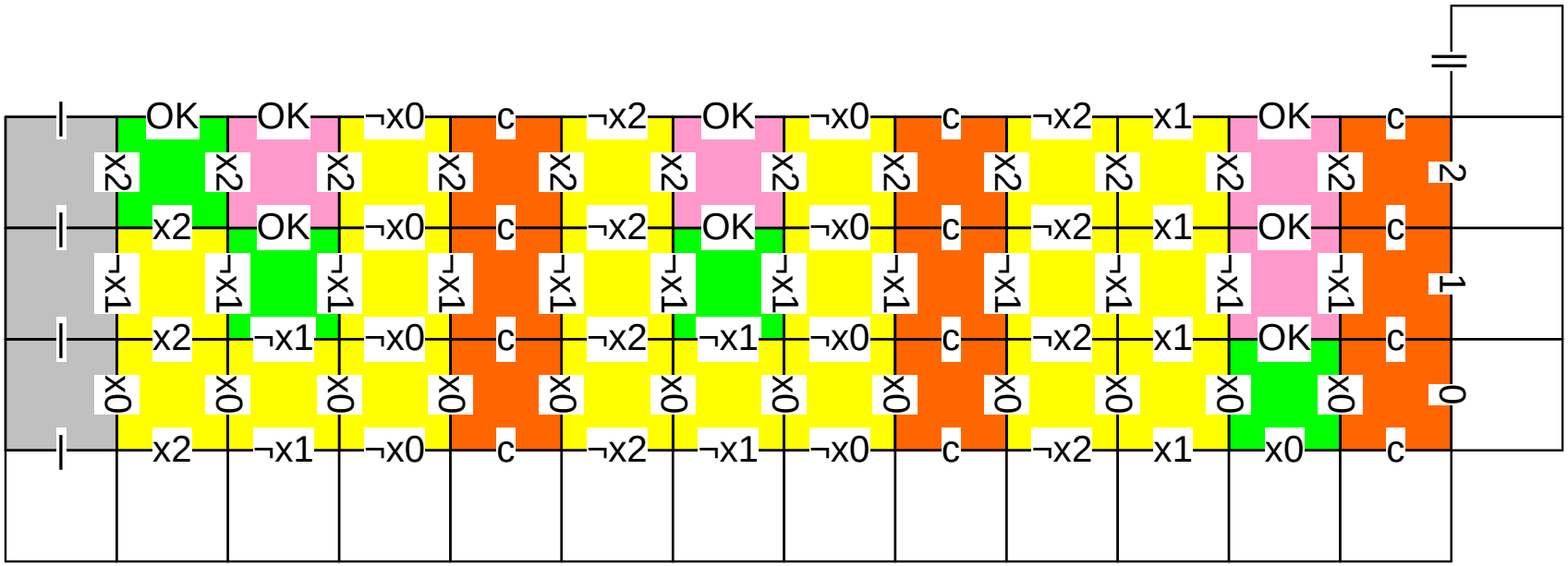

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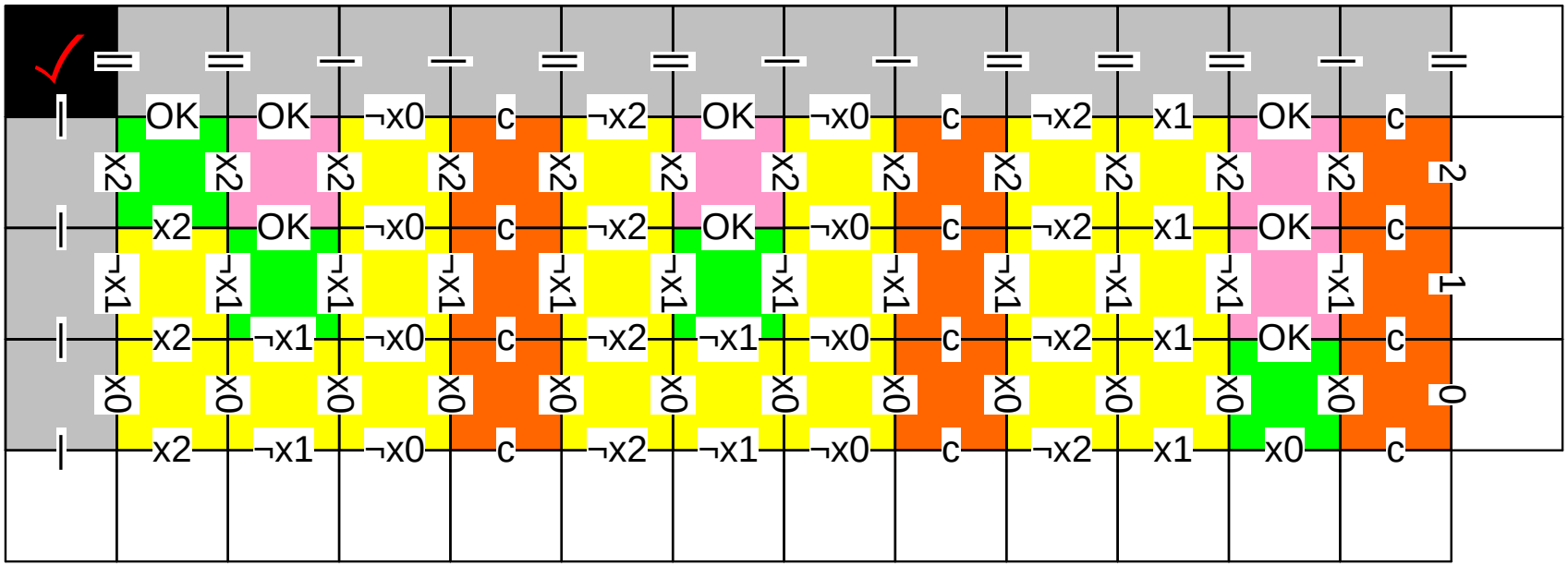

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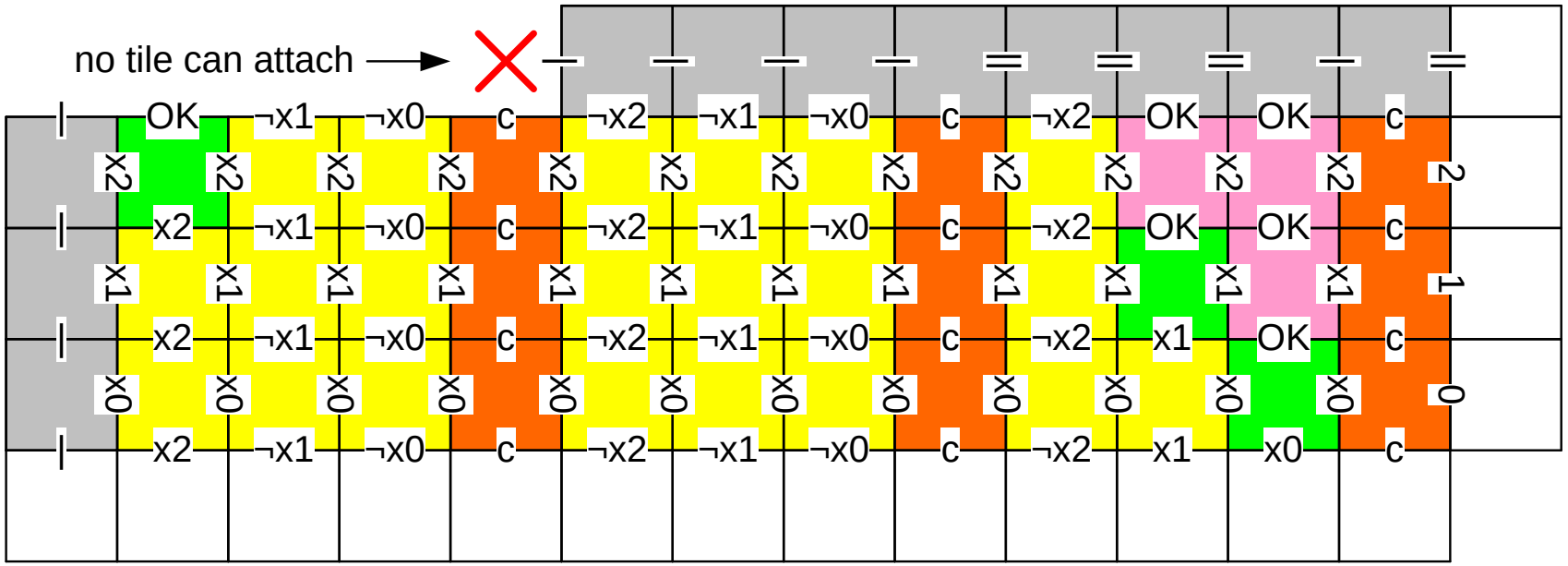
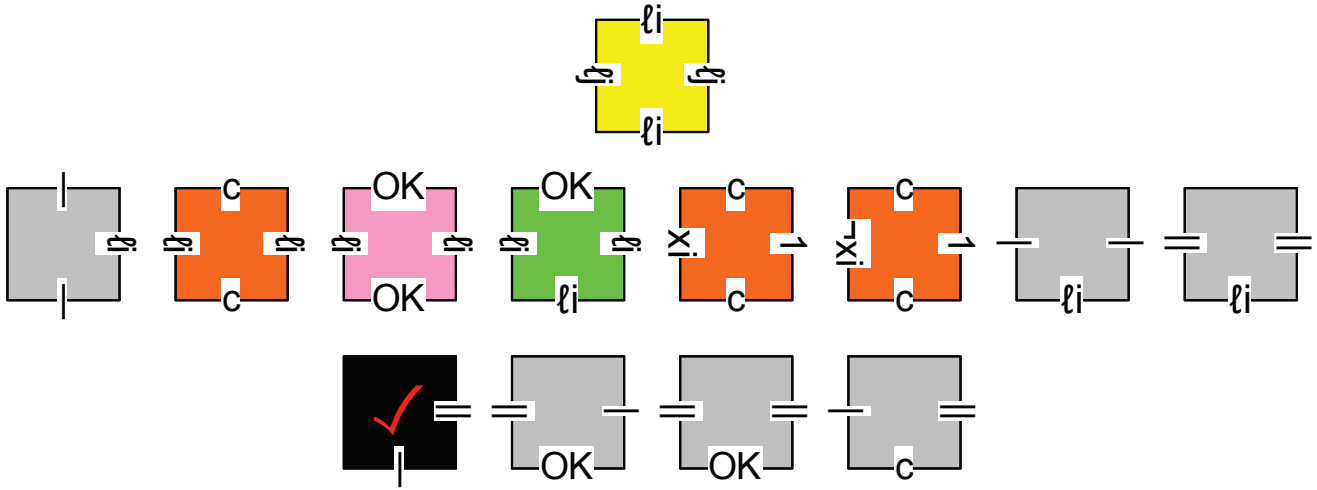


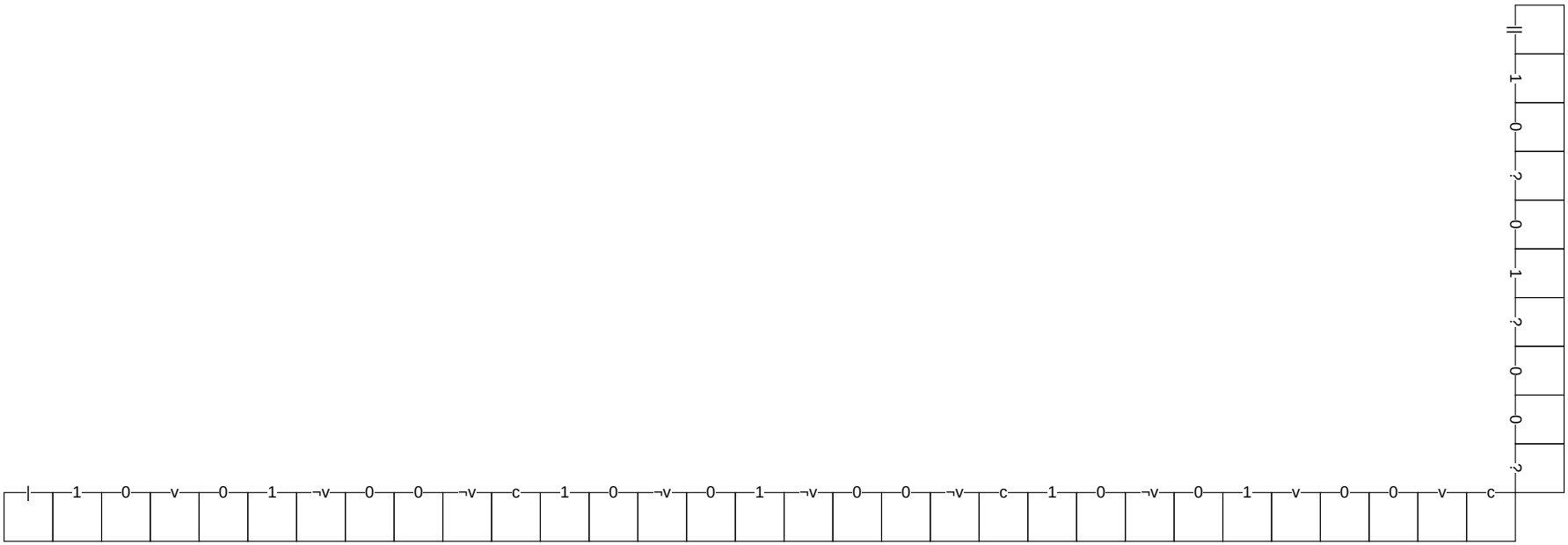
The $\Theta(n^2)$ Tileset 3-SAT Solution [LL99]

- $\Theta(n^2)$ tile types
- Probability of success $\geq \left(\frac{1}{2}\right)^n$



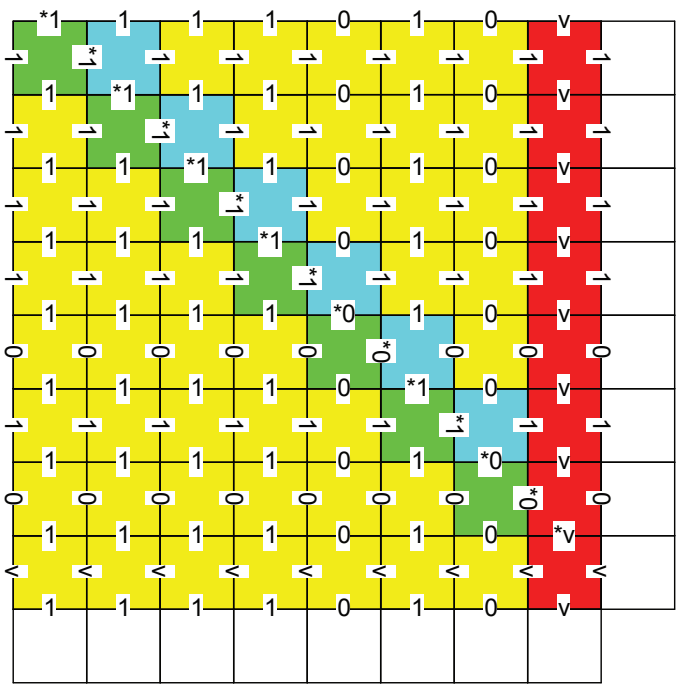
Encoding Formulae with a $\Theta(1)$ Tileset

$$(x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee x_1 \vee x_0)$$

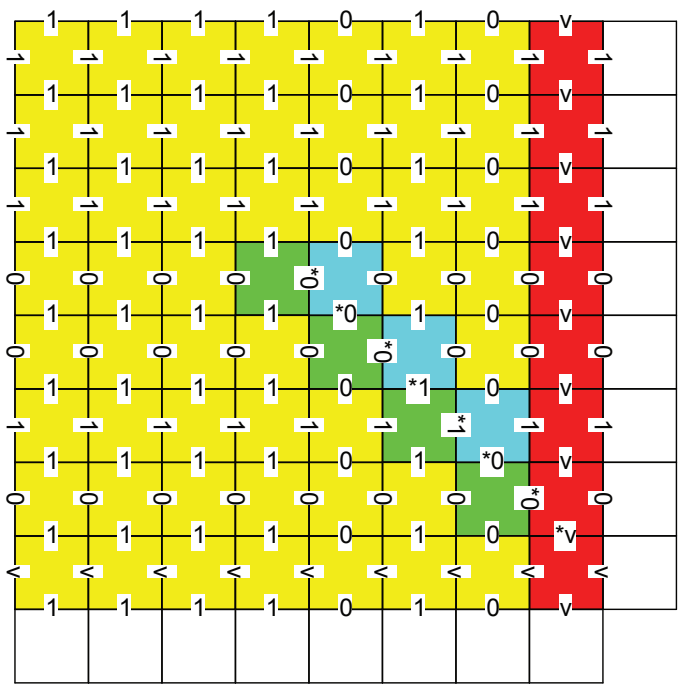


Comparing Literals

x_{122} vs. x_{122}

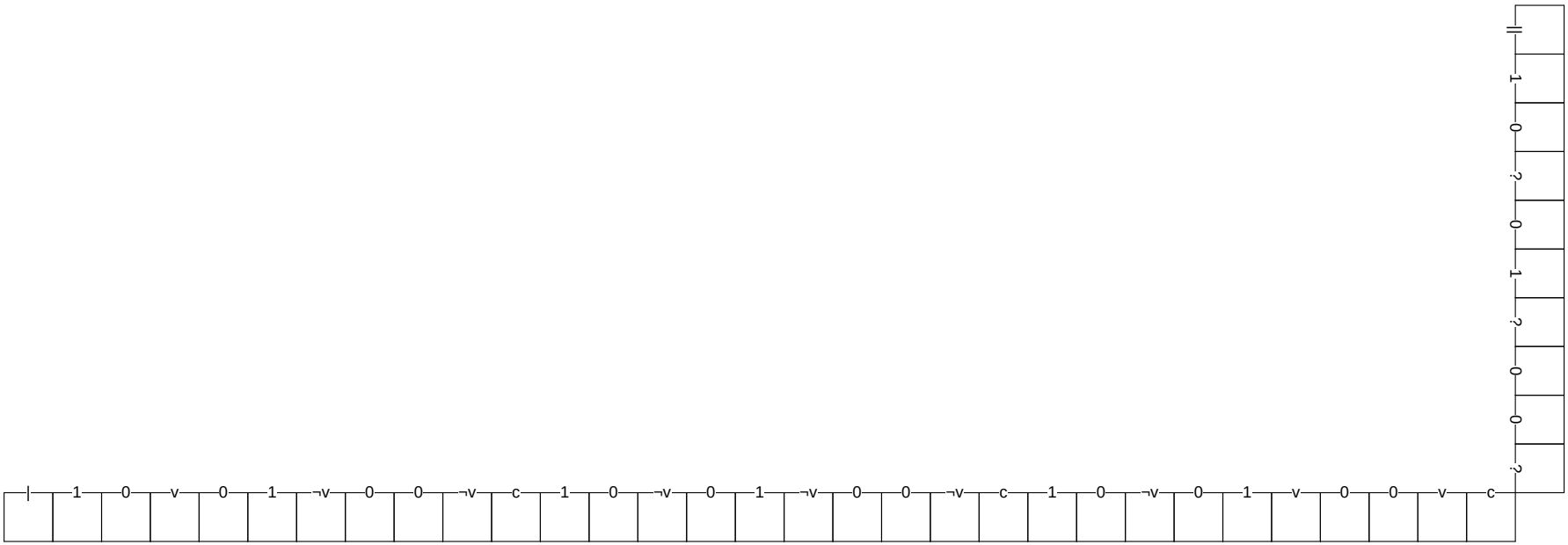


x_{122} vs. x_{114}



Solving 3-SAT

$$(x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee x_1 \vee x_0)$$



Solving 3-SAT

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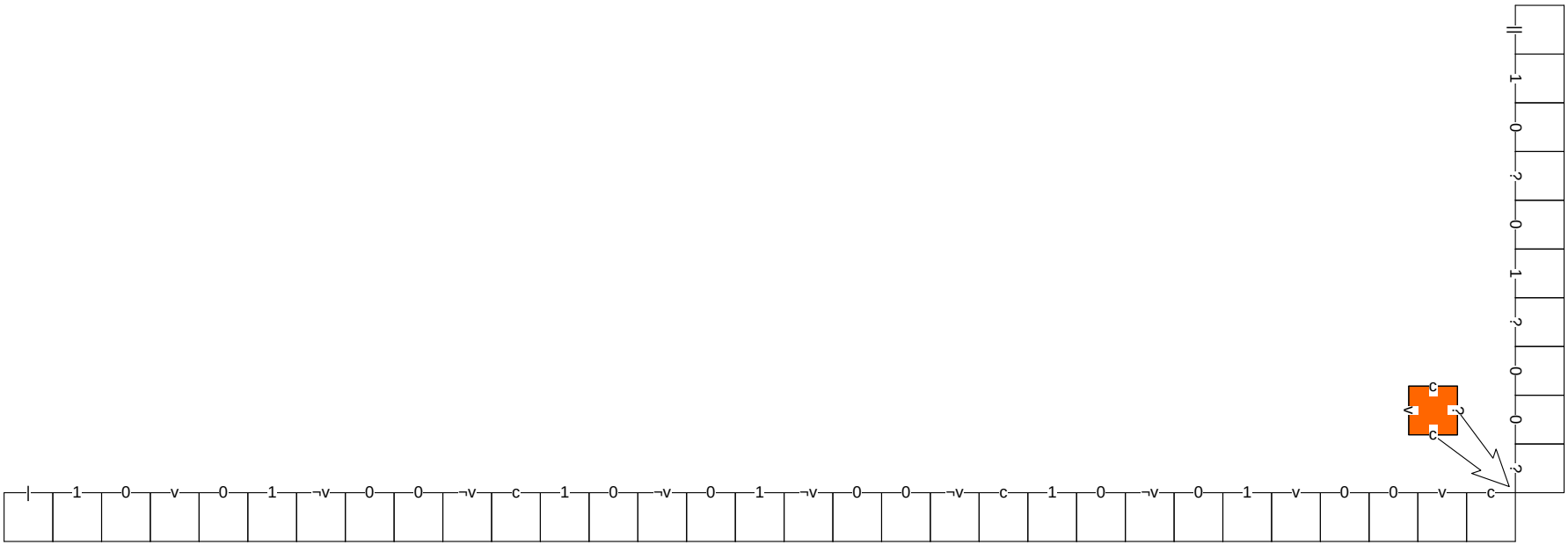

Solving 3-SAT

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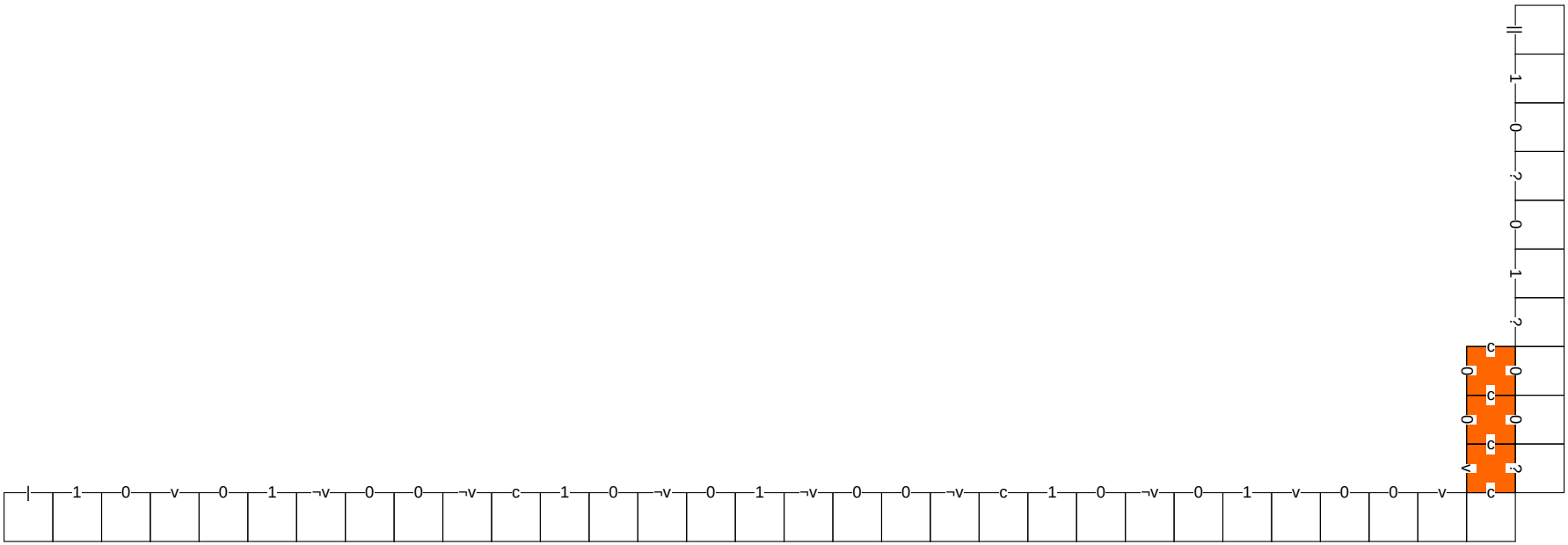

Solving 3-SAT

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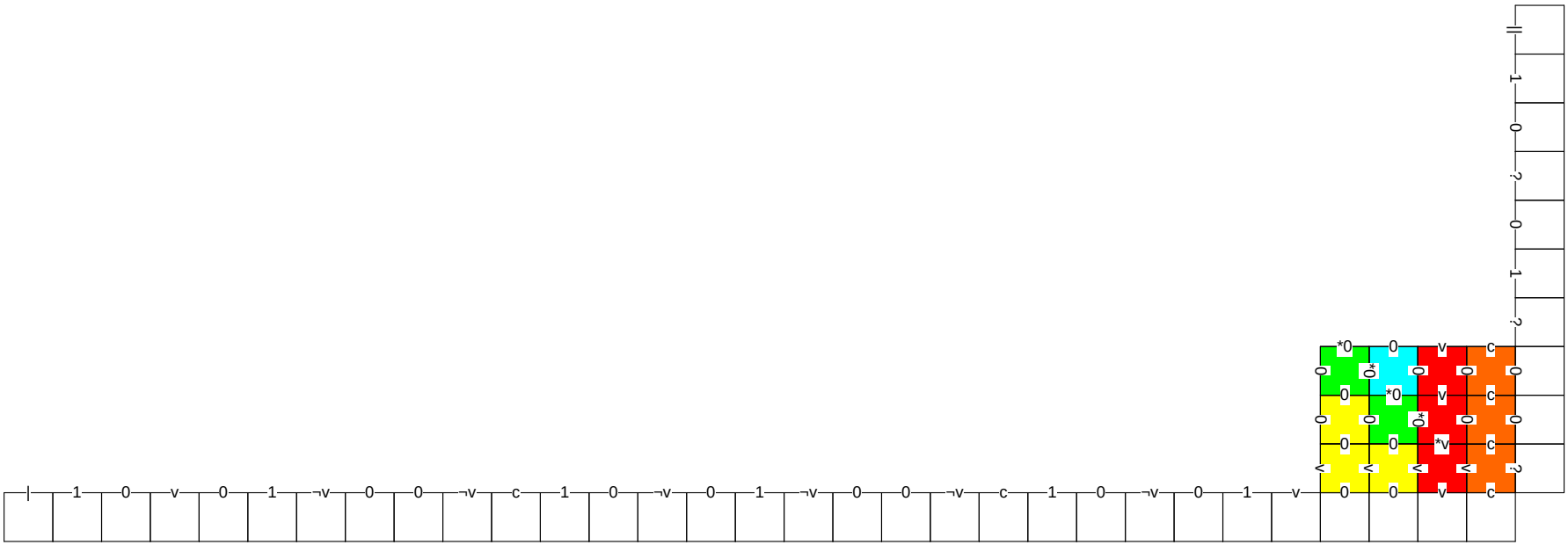
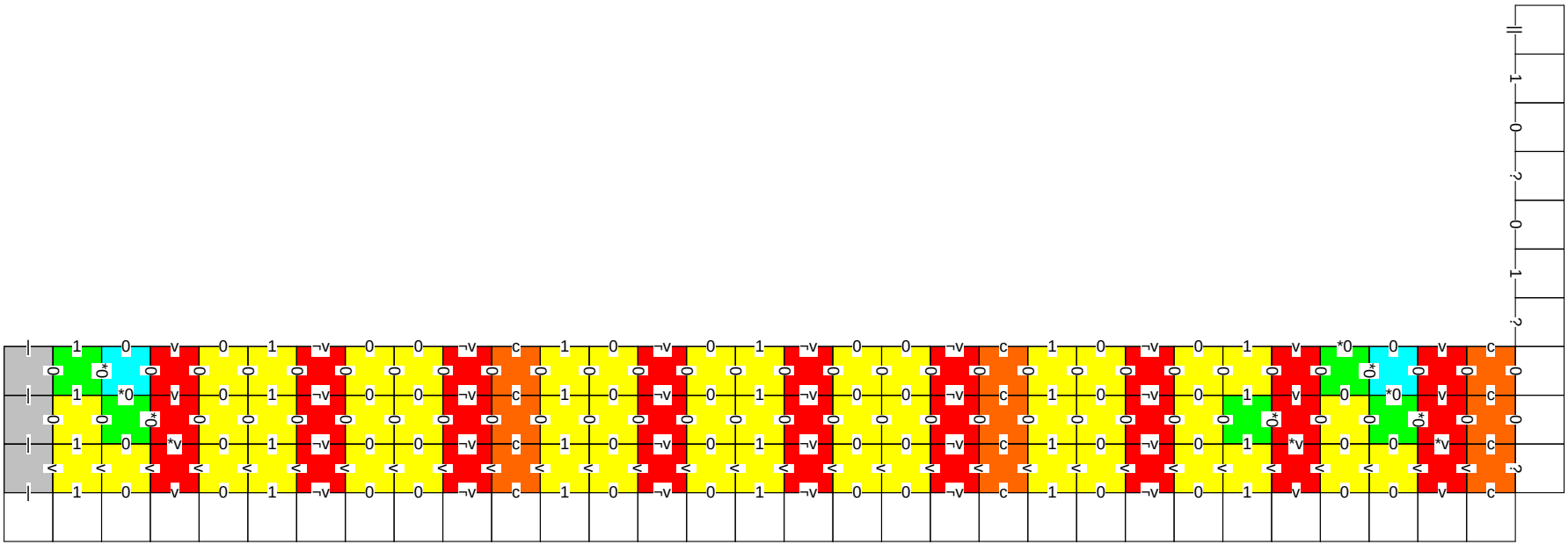


Solving 3-SAT

$$(x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee \neg x_1 \vee \neg x_0) \wedge (\neg x_2 \vee x_1 \vee x_0)$$



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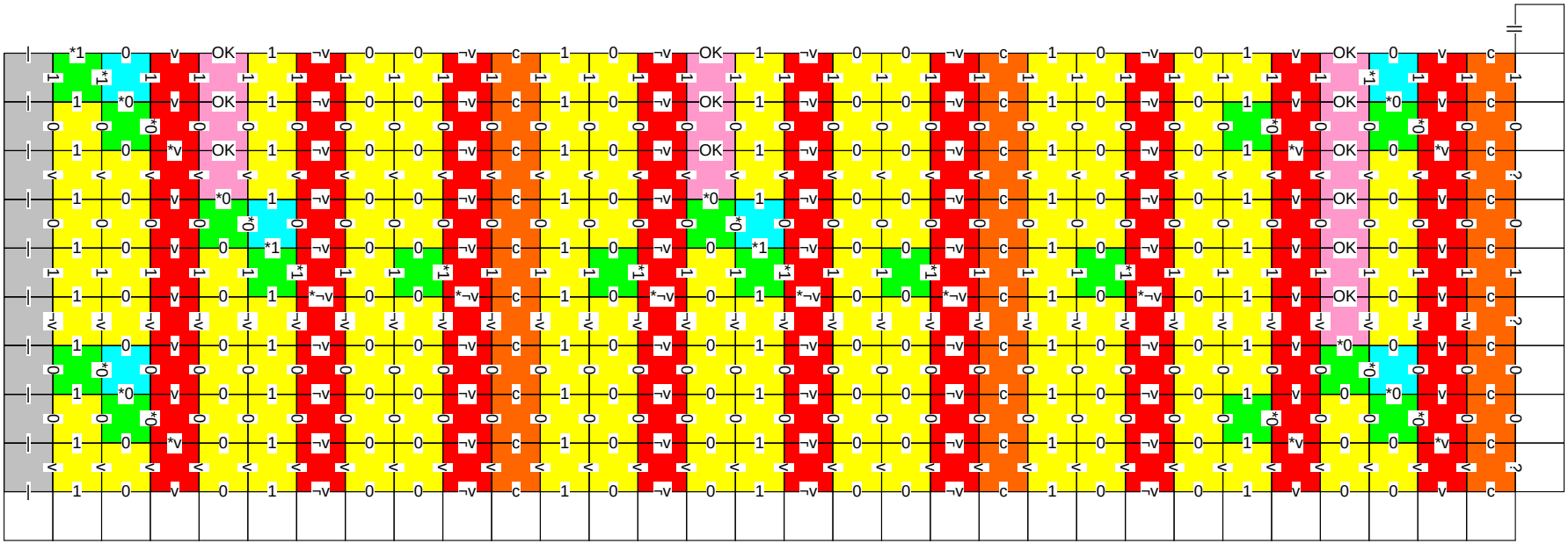

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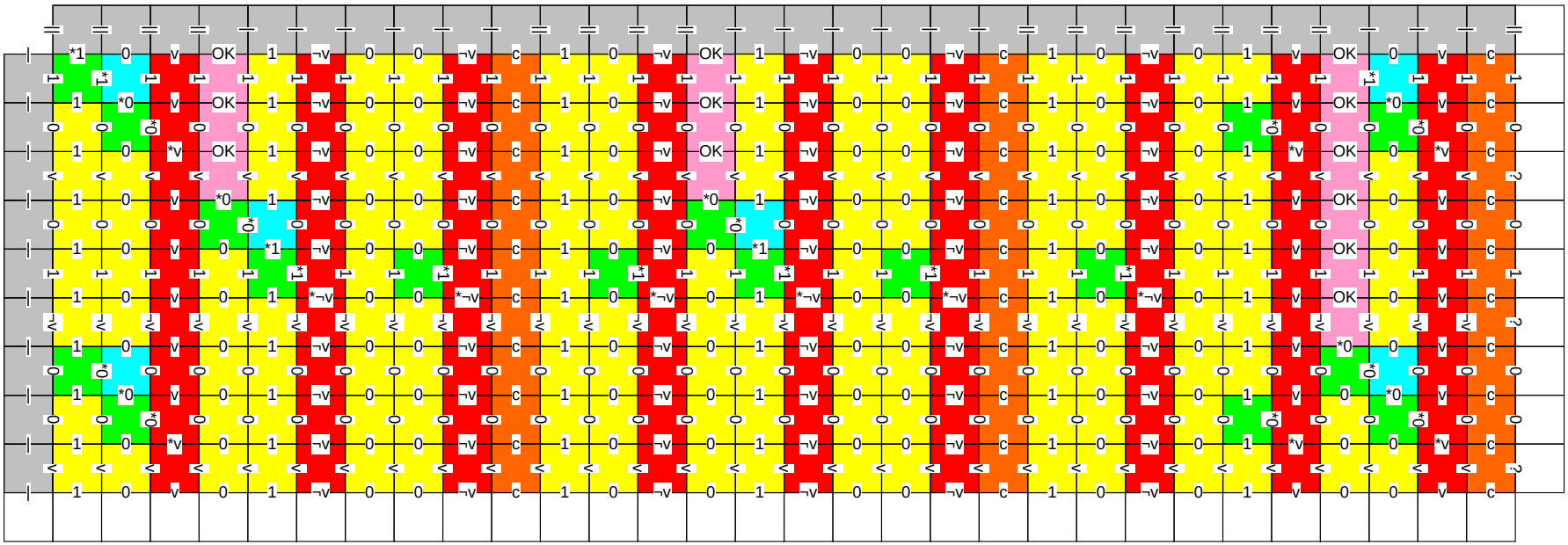

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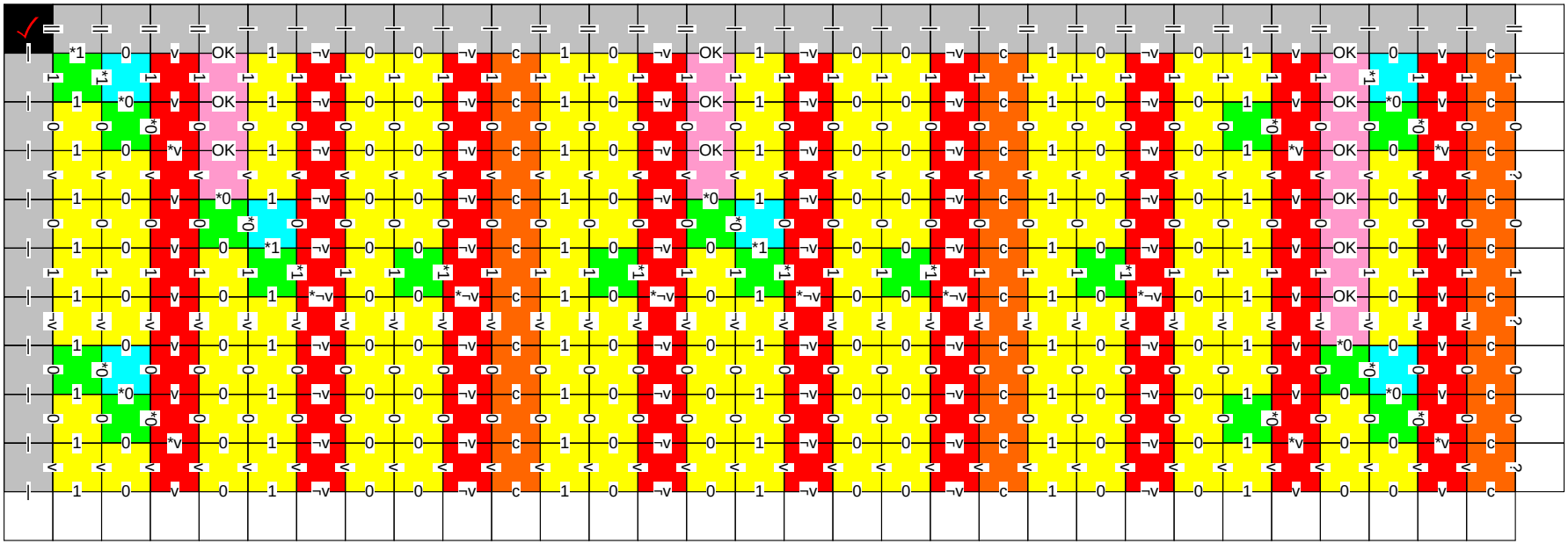
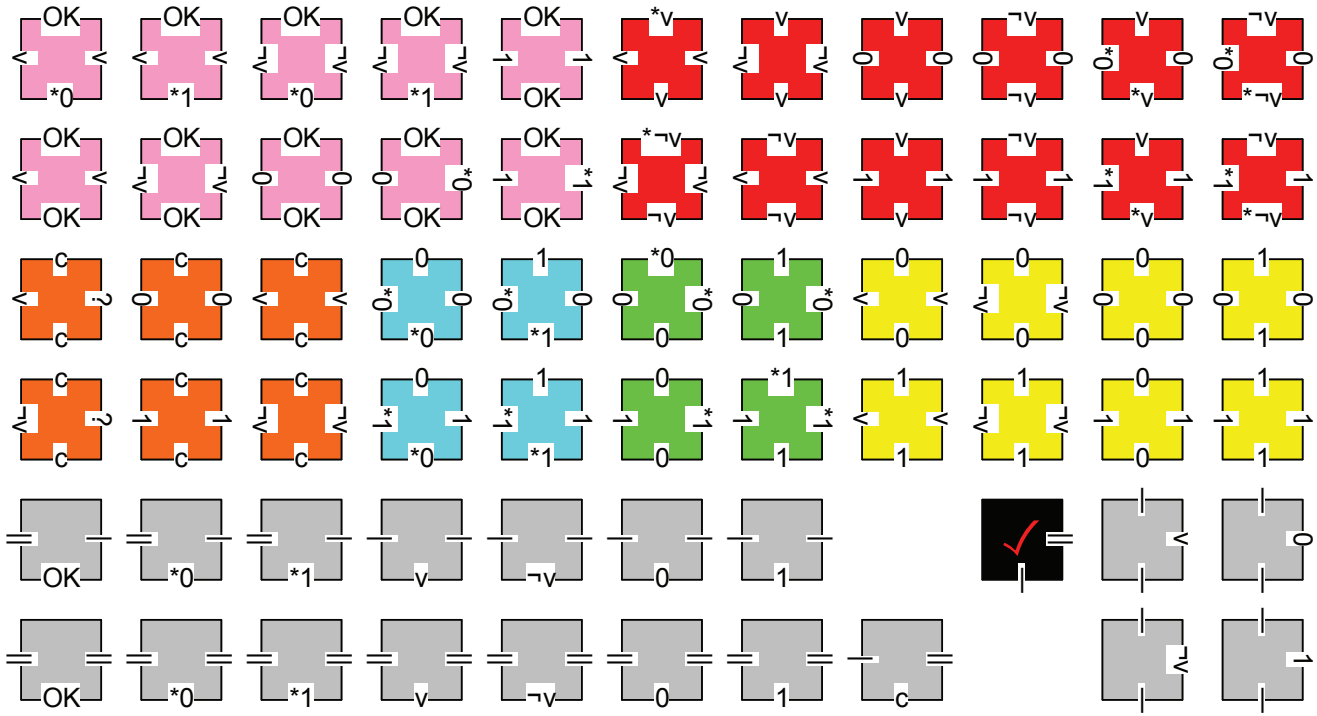


$\Theta(1)$-Tileset 3-SAT Solution [Bru08c]

- 64 tile types
- Probability of success $\geq \left(\frac{1}{2}\right)^n$



Improving the 3-SAT Algorithm Runtime

Some algorithms reduce the base of the exponent

Fastest known: $O^\star(1.3333^n)$ [Woe03].

An $O^\star(1.8393^n)$ algorithm [Woe03]

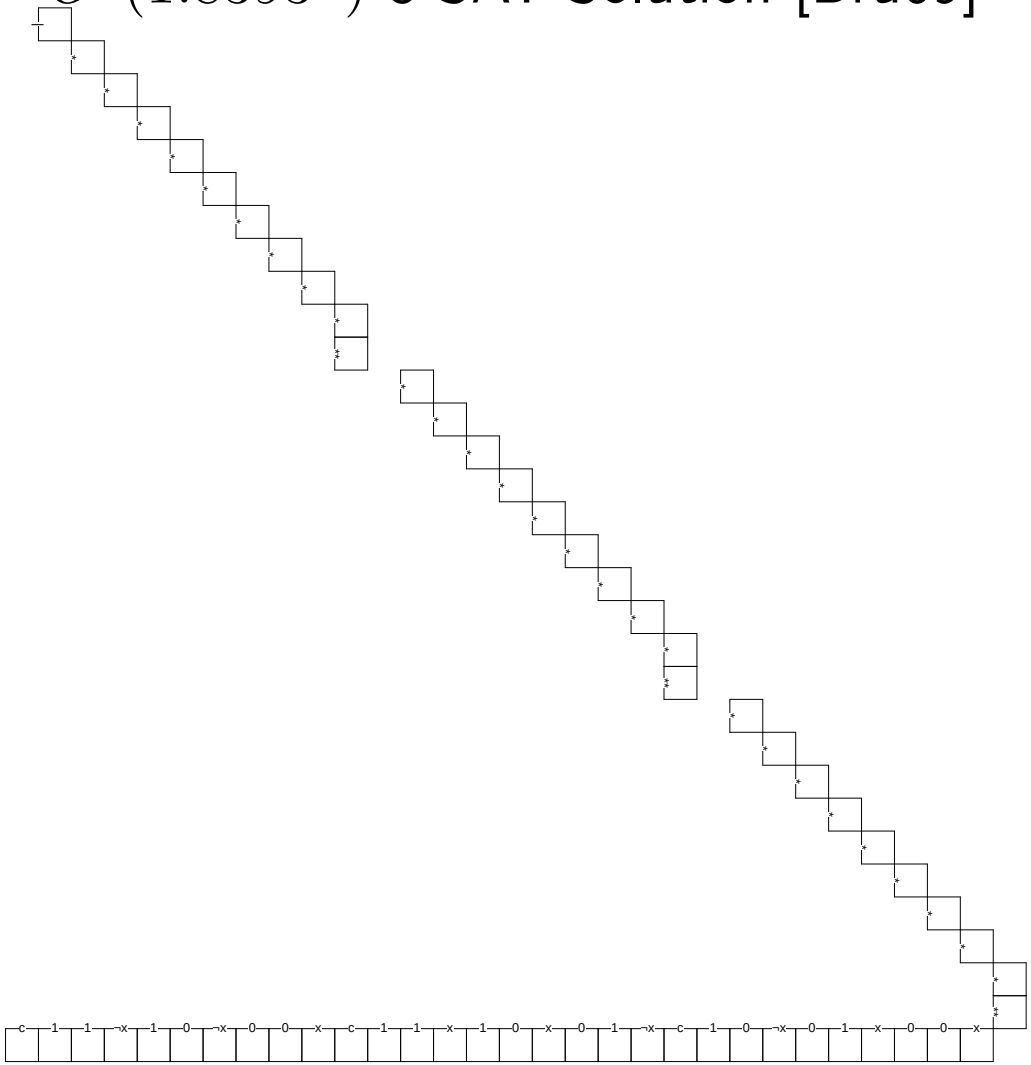
Suppose $\phi = (x_1 \vee \neg x_2 \vee x_3) \cdots$.
There are 3 relevant possibilities: either,

- first literal is $TRUE$, or
- first literal is $FALSE$ and second literal is $TRUE$, or
- first two literals are $FALSE$ and third literal is $TRUE$.

$$T(n,m) = c + \sum_{i=1}^{3} T(n-i, m-1) = O^\star(1.8393^n m).$$

Can Tiles Implement More-Efficient Algorithms?

$O^\star(1.8393^n)$ 3-SAT Solution [Bru09]



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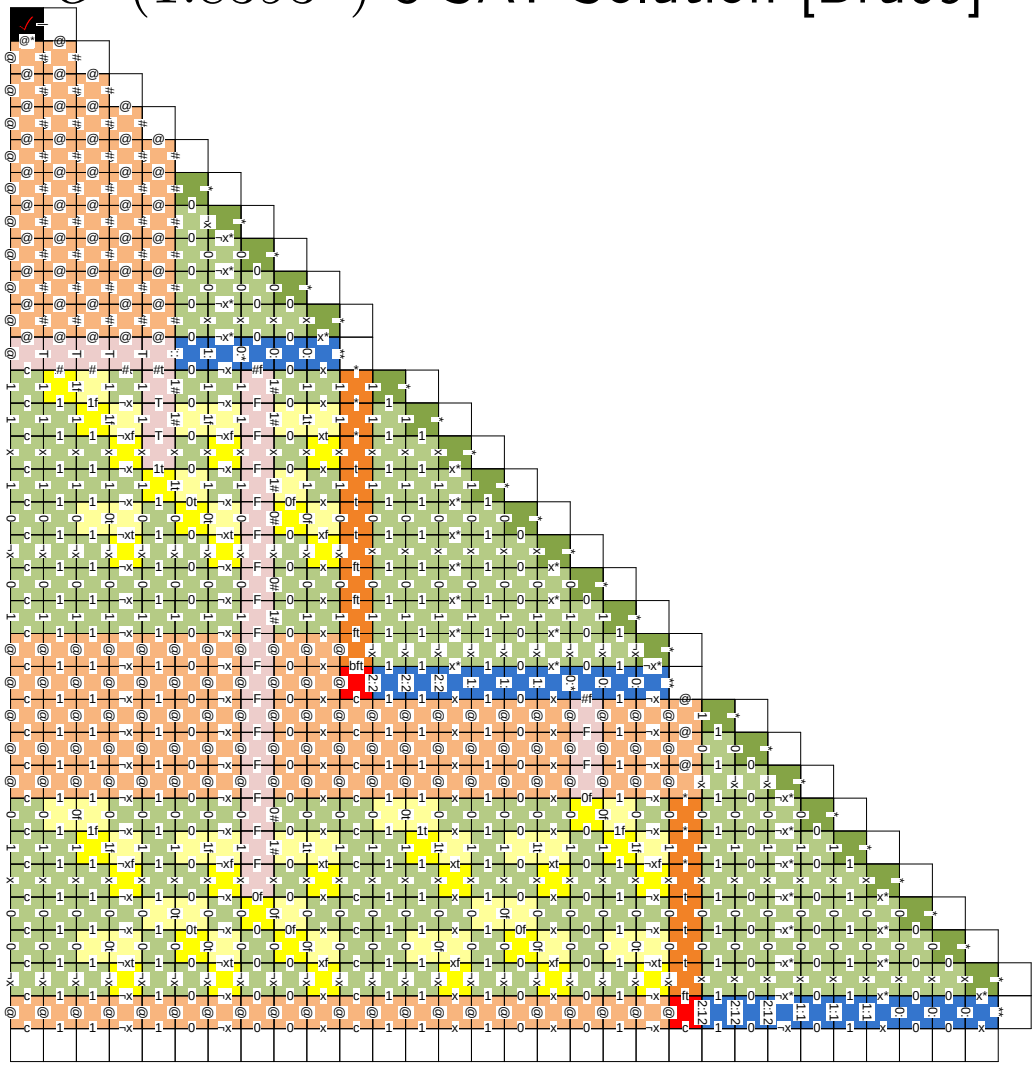


$O^\star(1.8393^n)$ 3-SAT Solution [Bru09]

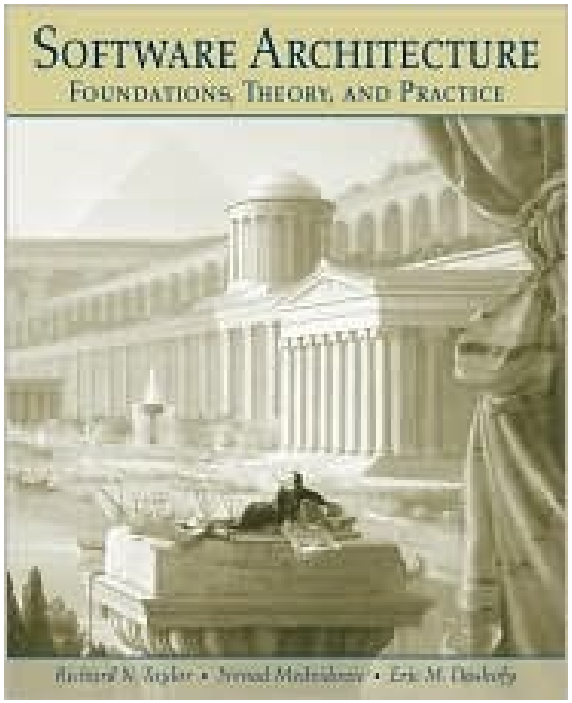
- 150 tile types
- Probability of success $\geq \left(\frac{1}{1.8393}\right)^n$

Efficient Tile Systems

- Add [Bru07]
- Multiply [Bru07]
- Factor [Bru08a]
- Solve SubsetSum [Bru08b]
- Solve k-SAT [Bru08c]

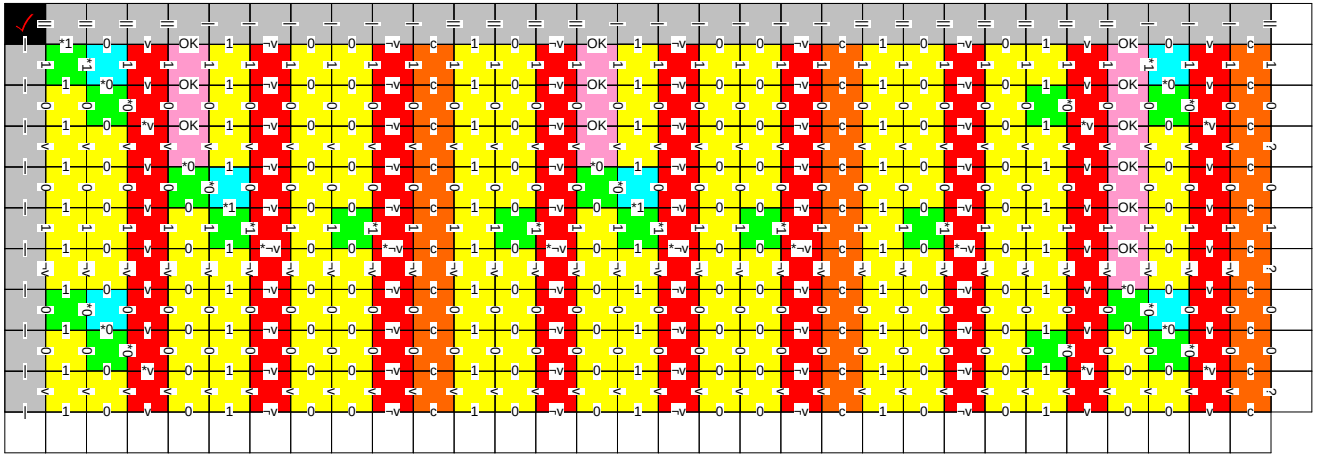
Outline

Software Architecture



"Software architecture: the set of principal design decisions made about a system." [TMD09]

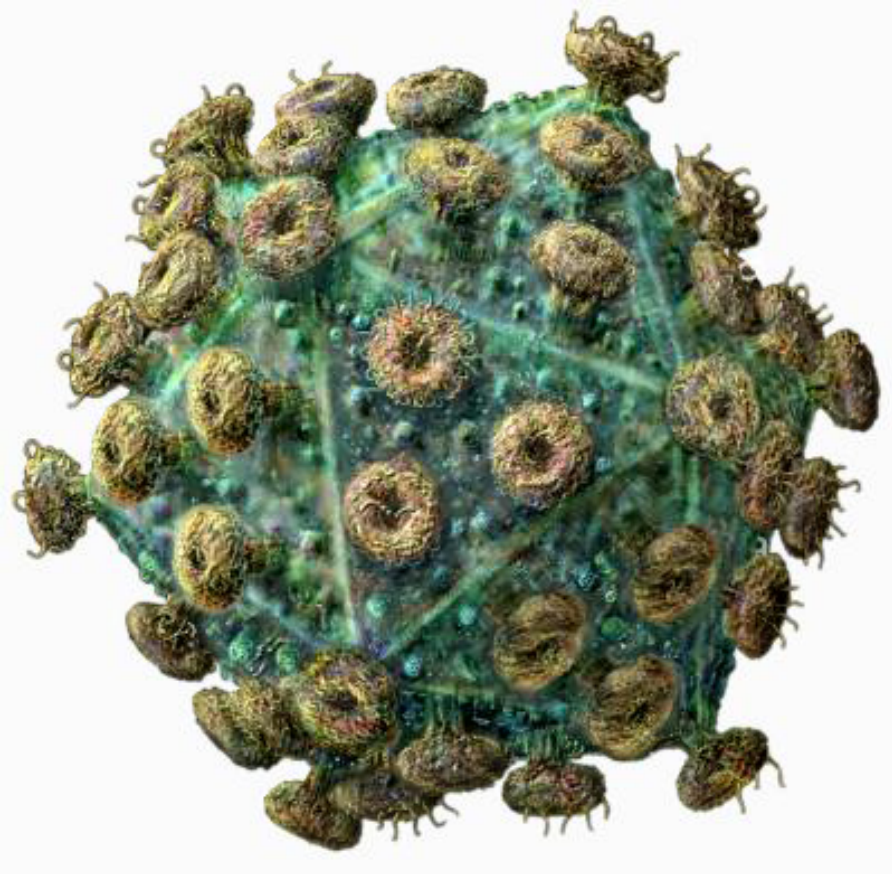
Converting the Model to an Architecture



Architectural Elements [MRMM02]

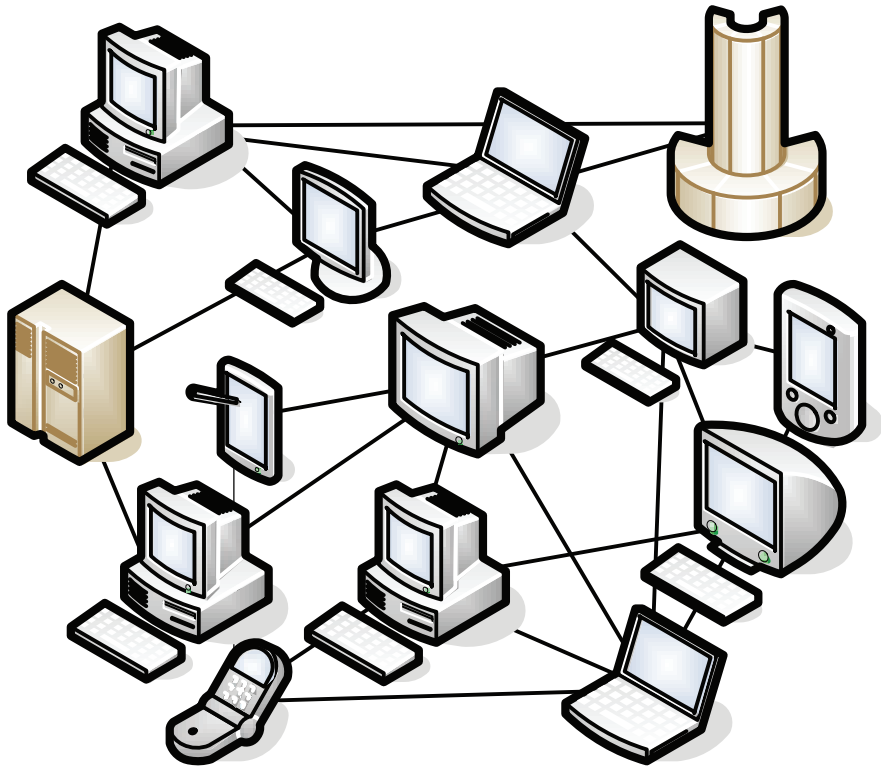
- Components: tiles
- Interfaces: side labels
- Topology: 2-D grid
- Behaviors: identifying nodes, recruiting attachments, replicating, and reporting the solution
- Interaction: recruitment data exchange

Computationally Intensive Problems



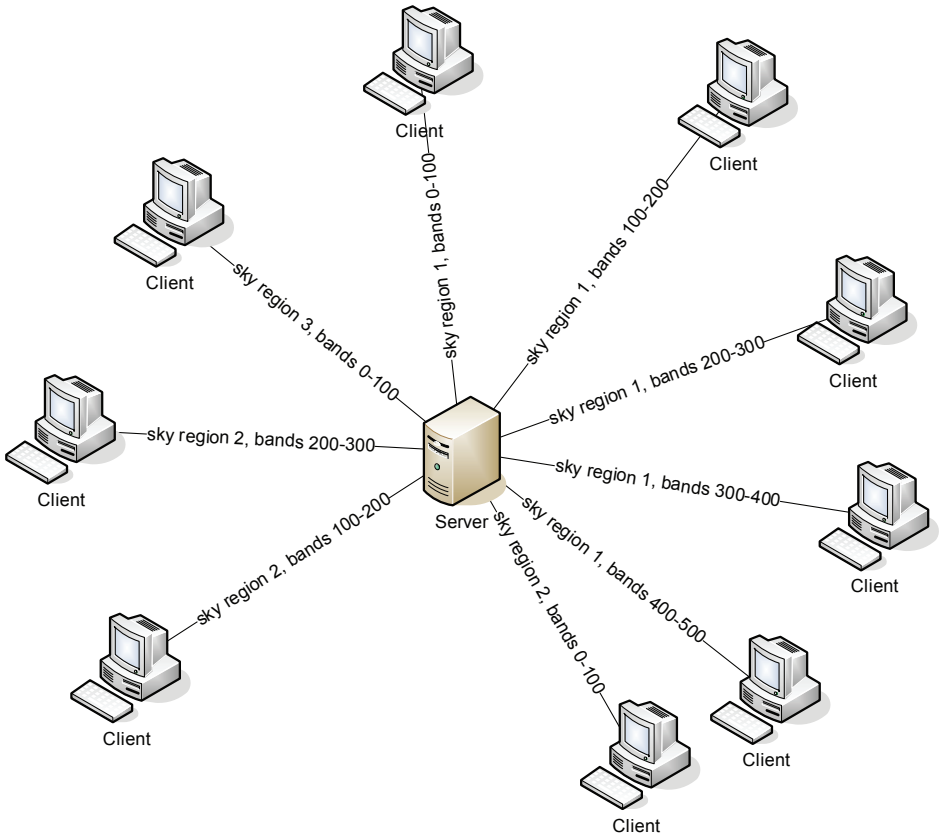
Internet as a Computing Medium

- Billion machines
- Mostly idle
- Insecure



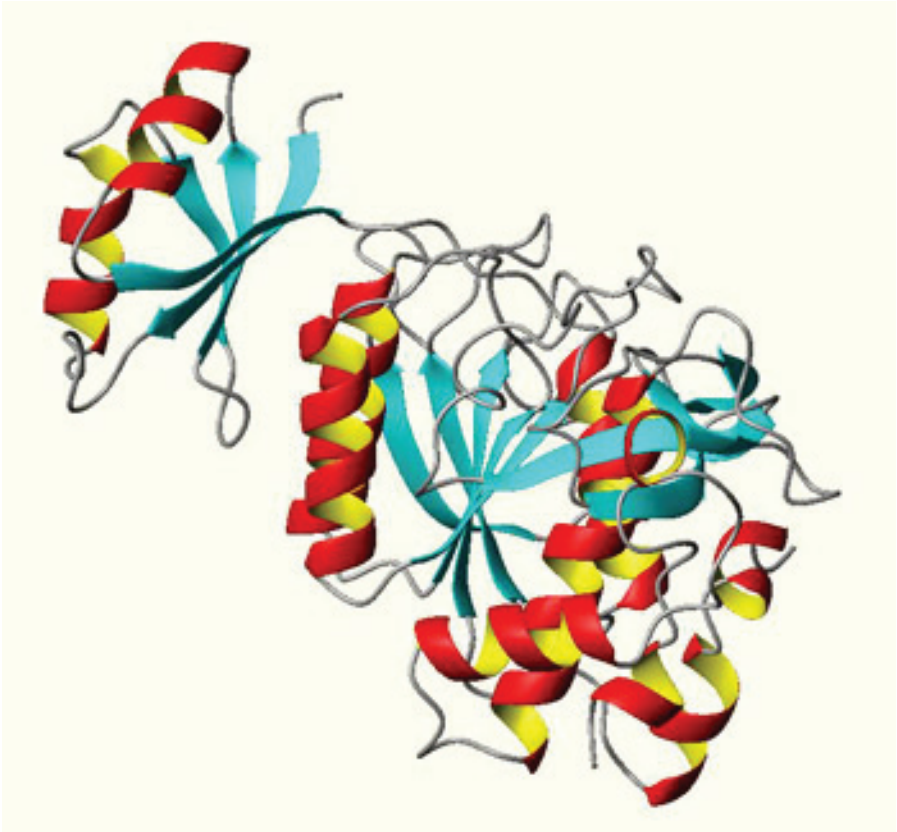
Distributed Computation

- Computation on the Internet
  - SETI@home [KWA+96]
  - Folding@Home [LSSP02]
  - Rosetta@home [Ros07]
- Grid Computing & Clouds
  - MapReduce [DG04]
  - OrganicGrid [CB04]
- Do not preserve privacy

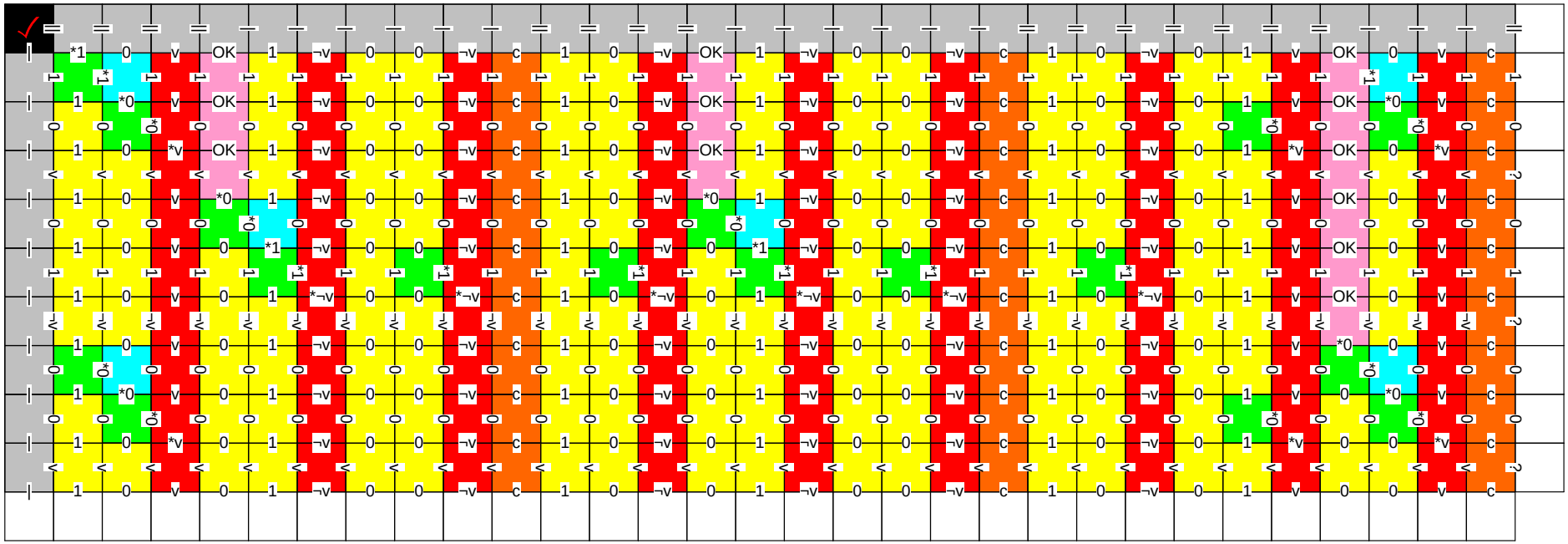


Example Scenario

- Possible cancer cure
- Find minimal-free-energy configuration
- Keep amino acid sequence private

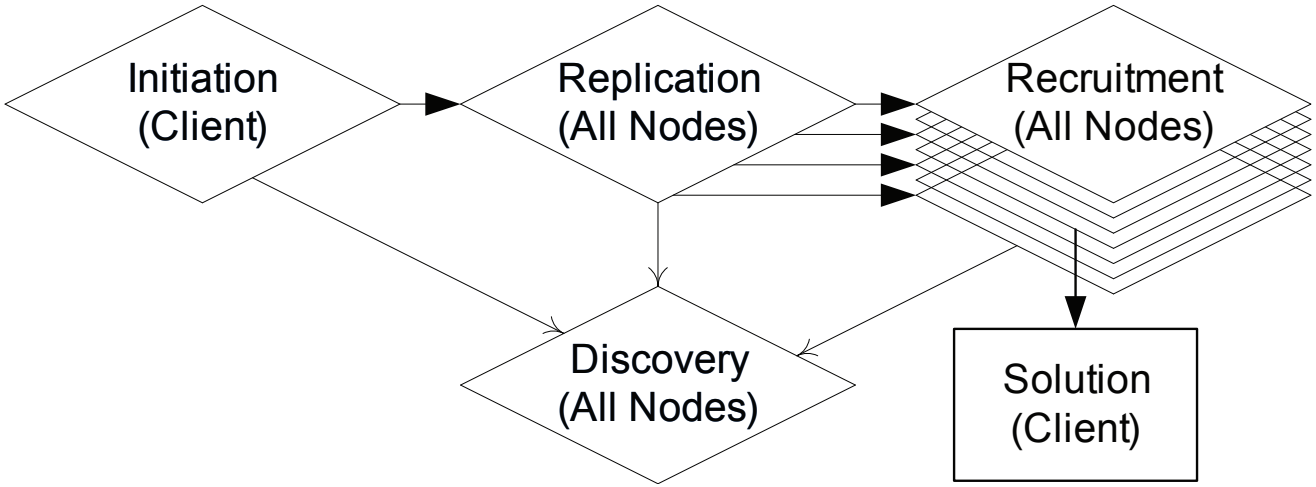


Tile Style Intuition



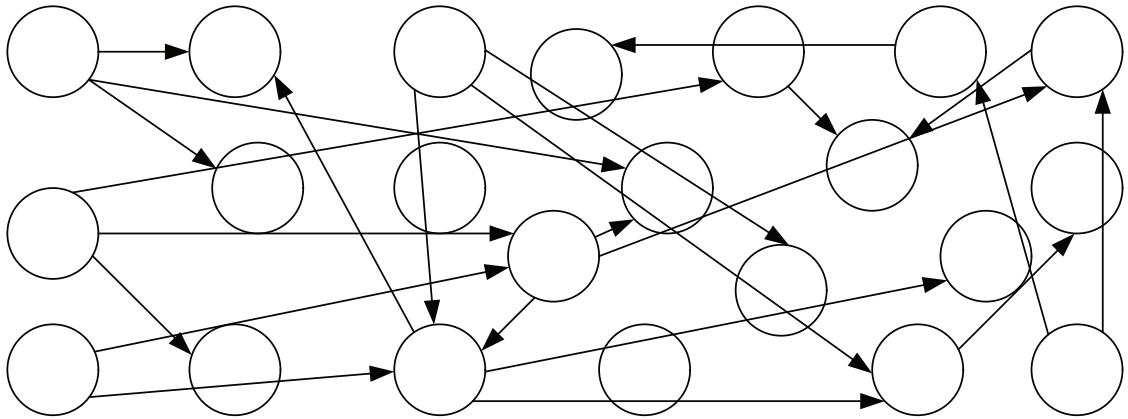
Node Operations [BM07a]

- Initiation (by the client)
- Node Discovery
- Replication
- Recruitment



Node Discovery

- Each node can return a randomly-uniform node of each tile component type



- Each node, for each tile type, keeps a list of 3 nodes that deploy that type
- When queried, a node returns one of the 3 elements at random, and replaces its list with that nodes list of 3
- Result: the algorithm returns a uniformly-random IP after only $\Theta(logN)$ requests [MR95]

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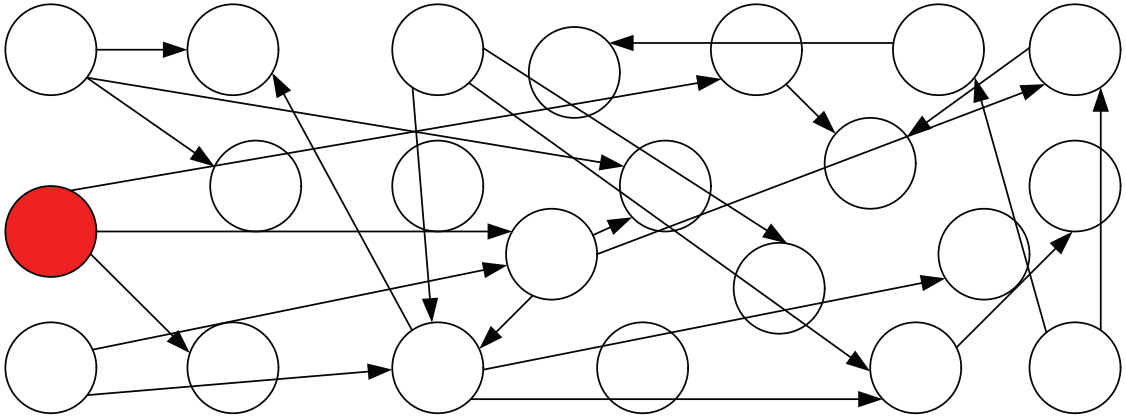

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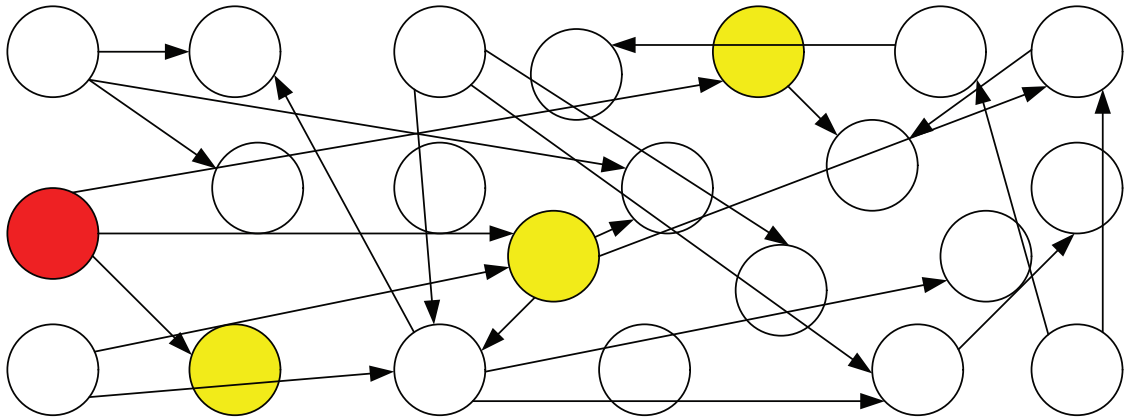

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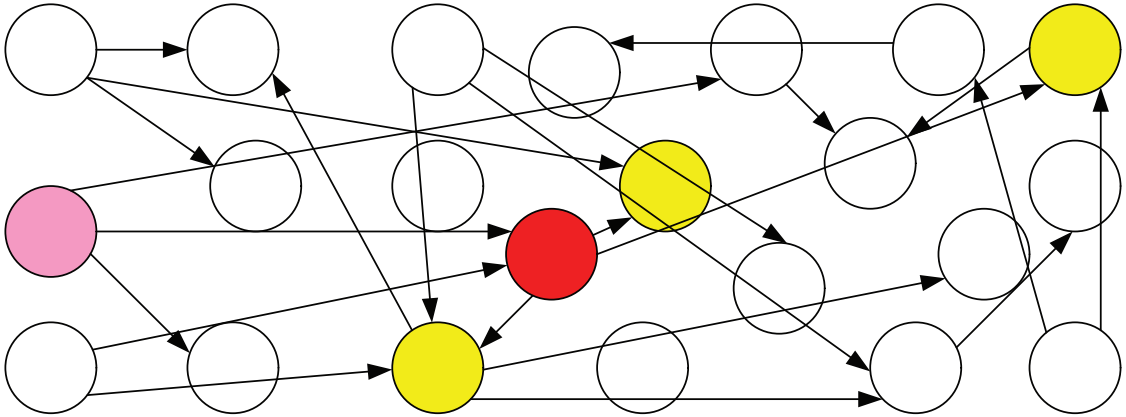

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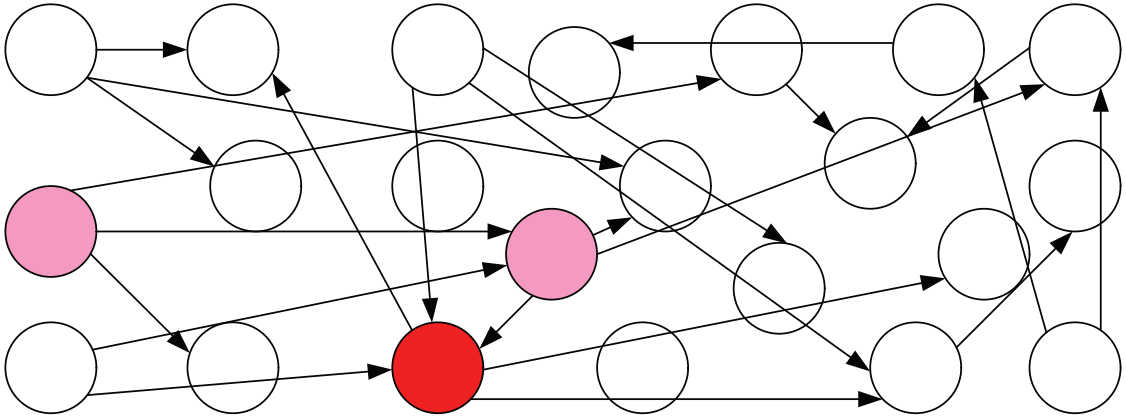

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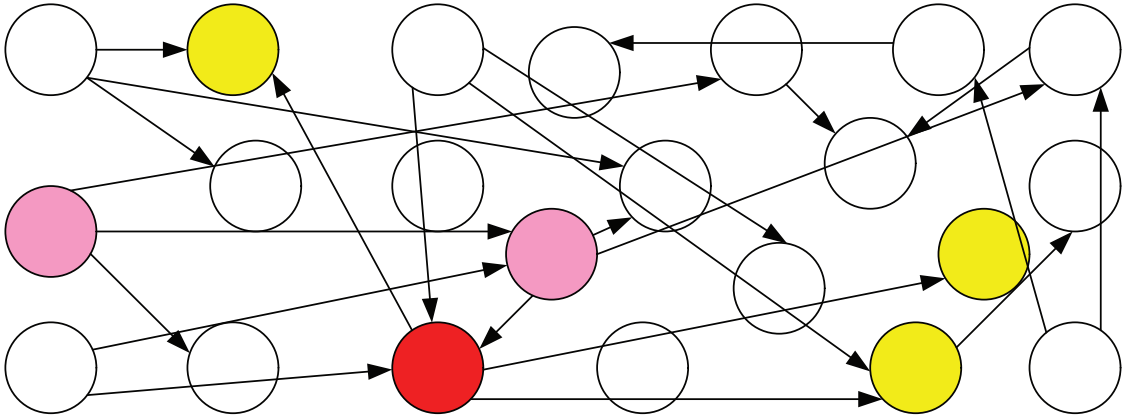

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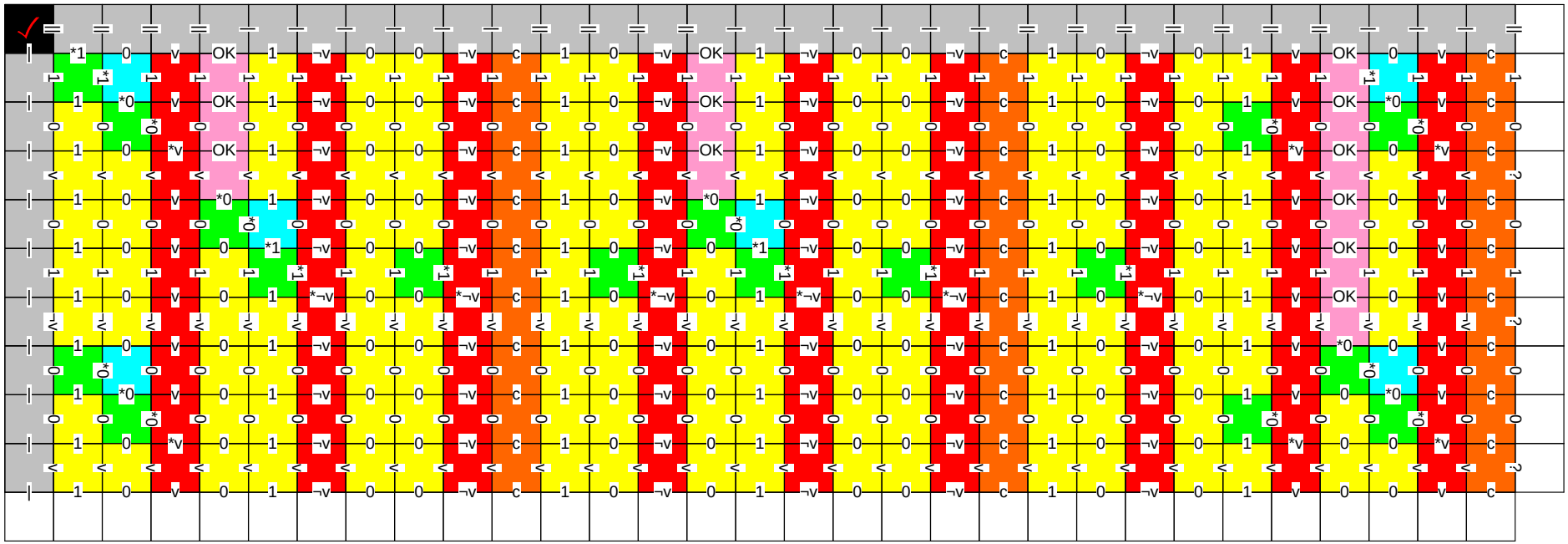
Privacy Preservation

- Data
  1. Each node knows very little
  2. It is hard to control the entire input
- Algorithm
  3. One tile type implies nothing
  4. It is hard to learn all the tile types
  5. Knowing the tile types does not reveal the algorithm

Data: Each Node Knows Very Little



Less than 1 bit of information per tile

Data: It Is Hard to Control the Entire Input

$$\boxed{1 - (1 - c^n)^s}$$

n — bits in input

c — compromised fraction

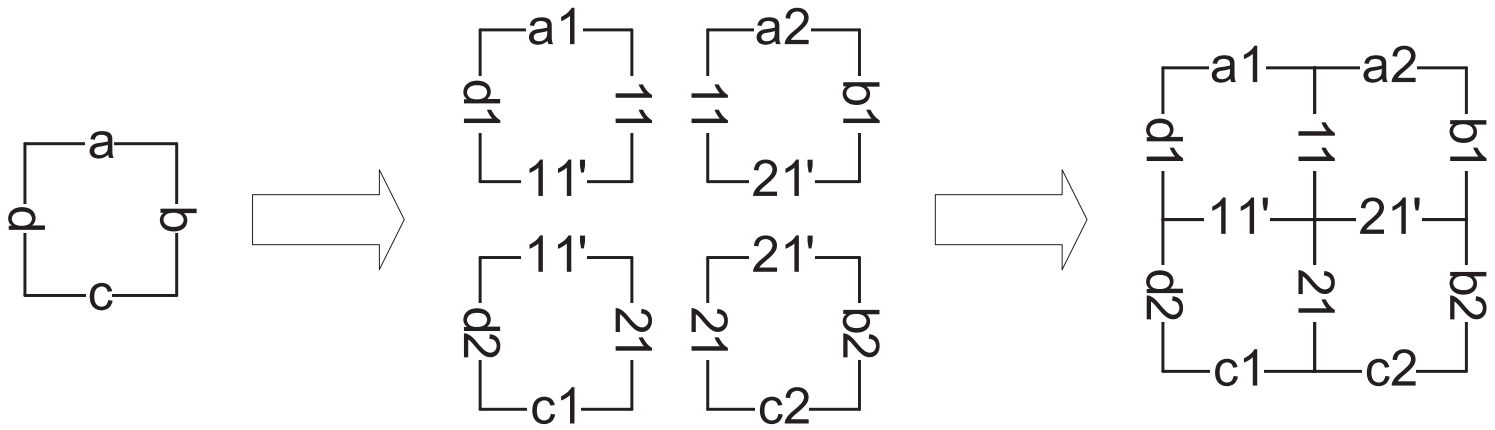
s — number of seeds

- TeraGrid ($100,000$ machines)
- 17-variable 100-clause 3-SAT problem

Compromised Fraction	Confidence Level
$\frac{1}{8}$	$1 - 10^{-10}$
$\frac{1}{4}$	$1 - 10^{-5}$
$\frac{1}{3}$	$1 - 10^{-3}$

Fault-Tolerant Tile Style [BM07b]

- Tile systems can be designed to be tolerant to misbehaving tiles
- For example, [WB03]



Provably Correctable Errors

- Failing tiles
- Misbehaving tiles
- Byzantine tiles
- Service attacks
- Privacy attacks
- . . . probably many more

Tile Style Hypotheses

1. Speed $\propto$ network size
2. Robust to network delay
3. Can solve real-world-sized problems

Experimental Setup

- Mahjong: tile style implementation
  - Java, 3K LoC
  - Leverages Prism-MW [MMRM05]
  - Download: http://csse.usc.edu/~ybrun/Mahjong

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- Networks
  - 11-node private cluster (P4 1.5GHz, 512MiB, WinXP/2000)
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- Sample problems:
  $\mathfrak{A}$: 5-number 21-bit $SubsetSum$
  $\mathfrak{B}$: 11-number 28-bit $SubsetSum$
  $\mathfrak{C}$: 20-variable 20-clause 3-SAT
  $\mathfrak{D}$: 33-variable 100-clause 3-SAT

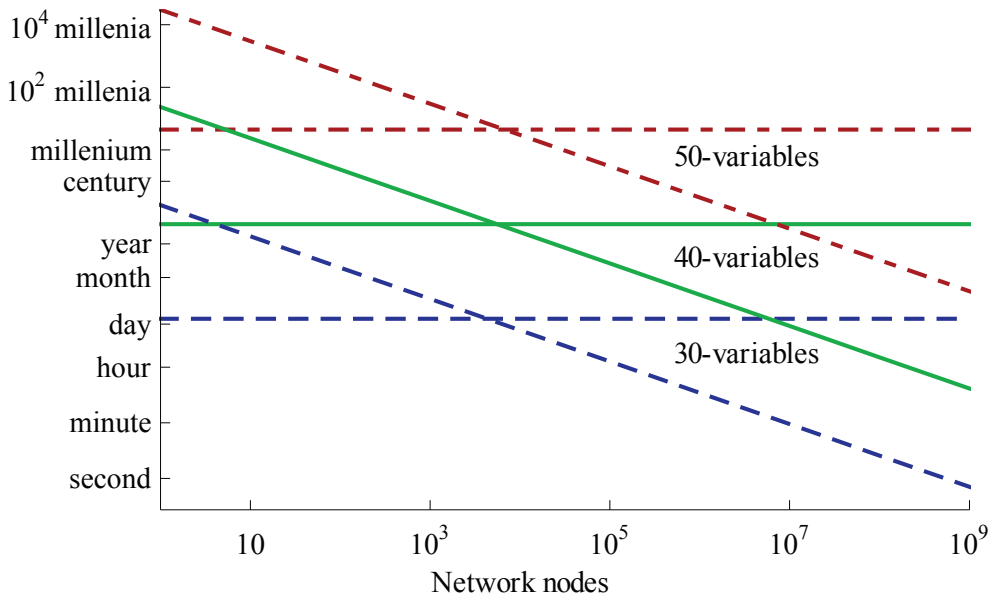
Scalability: Speed $\propto$ Network Size

Network & Problem	# of Nodes	Execution Time	Speed-up Ratio
Private Cluster	5	43.2 sec.	
$\mathfrak{A}$	10	22.9 sec.	1.89
HPCC	93	220 min.	
$\mathfrak{C}$	186	116 min.	1.90
PlanetLab	50	9.2 min.	
$\mathfrak{B}$	100	4.8 min.	1.92
Simjong	125,000	8.7 hours	
	250,000	4.5 hours	1.93
$\mathfrak{D}$	500,000	2.1 hours	2.14
	1,000,000	64 min.	1.97

Robustness to Network Delay

Problem	# of Nodes	Network Delay	Execution Time
Mahjong			
$\mathfrak{A}$	11	Private Cluster	20.1 sec.
		HPCC	19.3 sec.
		PlanetLab	18.5 sec.
$\mathfrak{B}$	11	Private Cluster	41.6 min.
		HPCC	41.2 min.
		PlanetLab	43.9 min.
Simjong			
$\mathfrak{D}$	1,000,000	0ms	65 min.
		10ms	57 min.
		100ms	64 min.
		500ms	60 min.
		Gaussian	68 min.
		Distance-based	59 min.

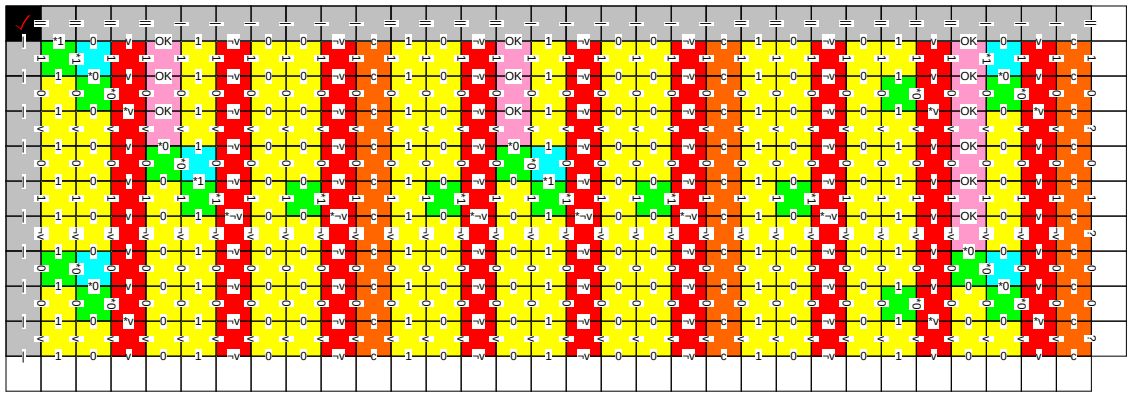
Efficiency: Solving Real-World-Sized Problems



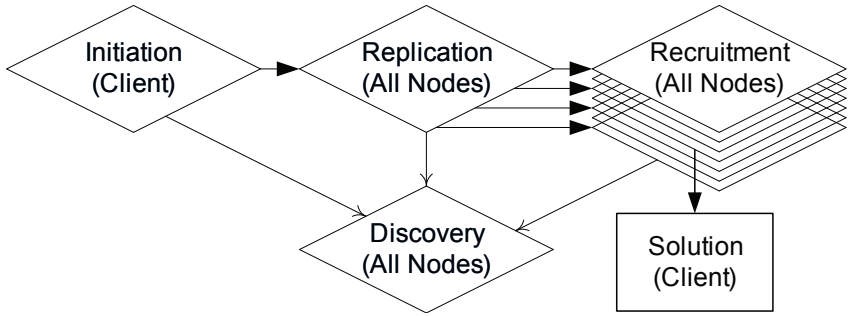
# of Nodes	Execution Time	
	Simjong	Theoretical Calculation
125,000	8.7 hours	9.1 hours
250,000	4.5 hours	4.5 hours
500,000	2.1 hours	2.3 hours
1,000,000	64 min	68 min.

Tile Style

- Developed self-assembling systems to solve complex computational problems



- Designed the tile architectural style for deploying tile systems on large networks



The Big Picture

Nature

- Bring forward novel, well-tested, well-scaling, robust mechanisms
- Present outside-the-box solutions

Software Architecture

- Facilitate translation of a nature-inspired model to software
- Aid design, implementation, and evaluation

Software Architecture III
Leveraging Nature to Build Better Systems

Yuriy Brun

http://www.cs.washington.edu/homes/brun/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