

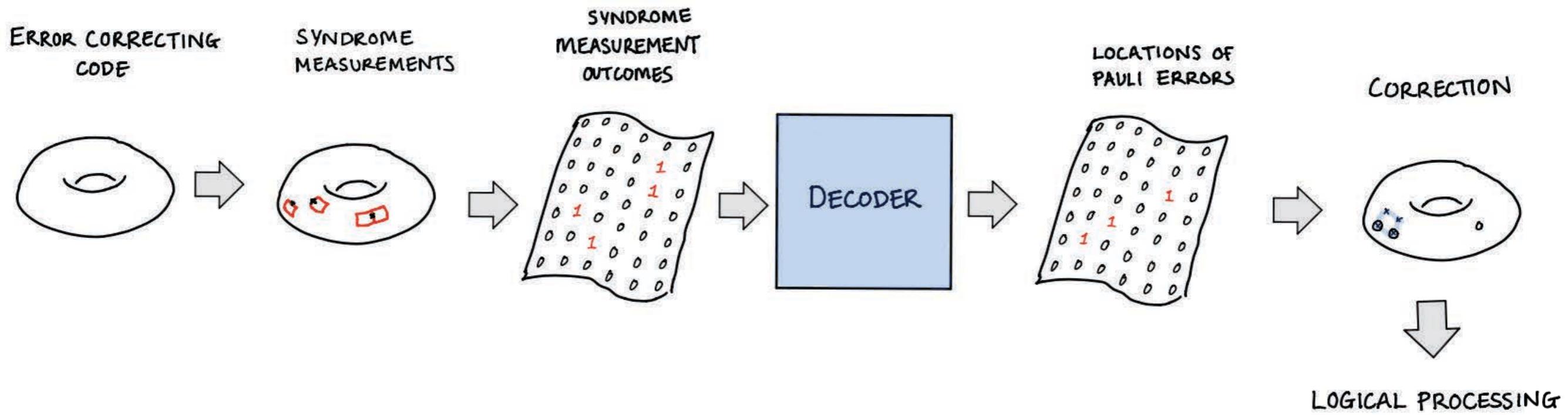
# Almost-linear time decoding algorithm for topological codes

Naomi Nickerson & Nicolas Delfosse

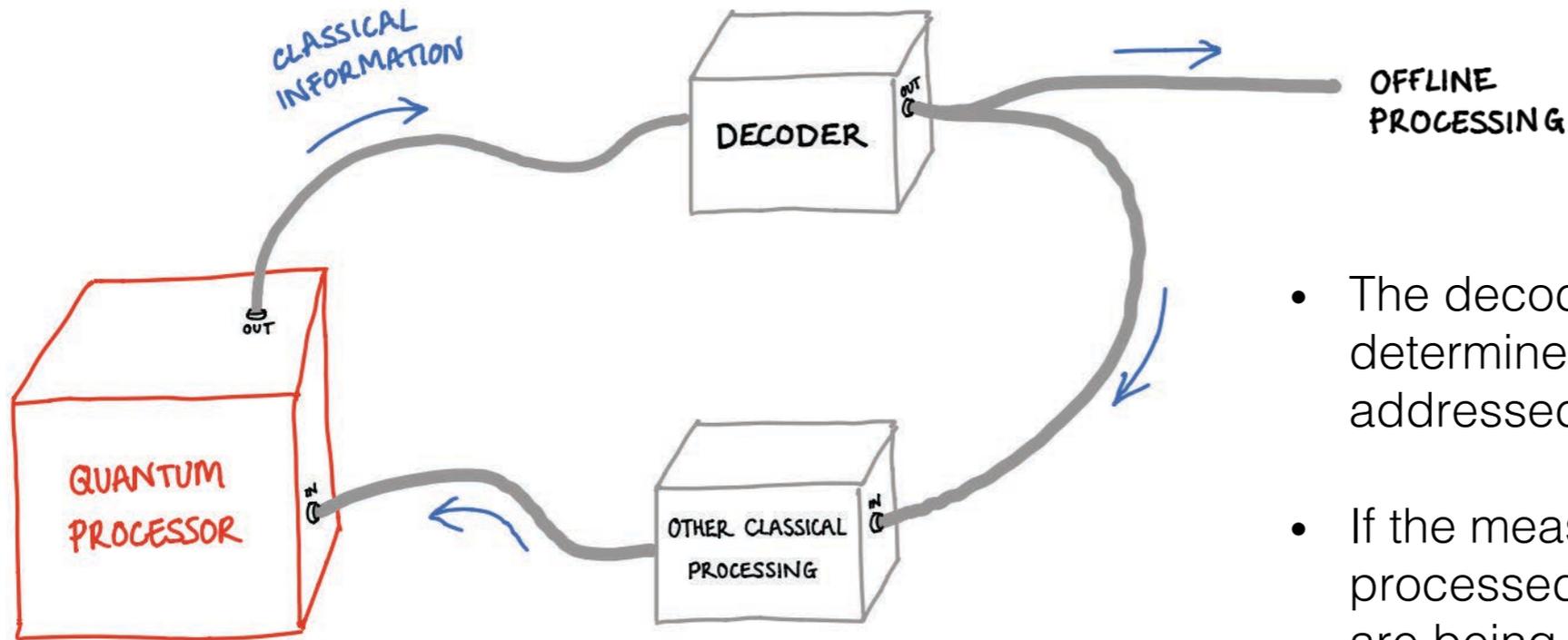


1. What is decoding and how fast does it need to be?
2. The Union Find decoder
3. Achieving almost-linear time

# What is a decoder?



# How fast?

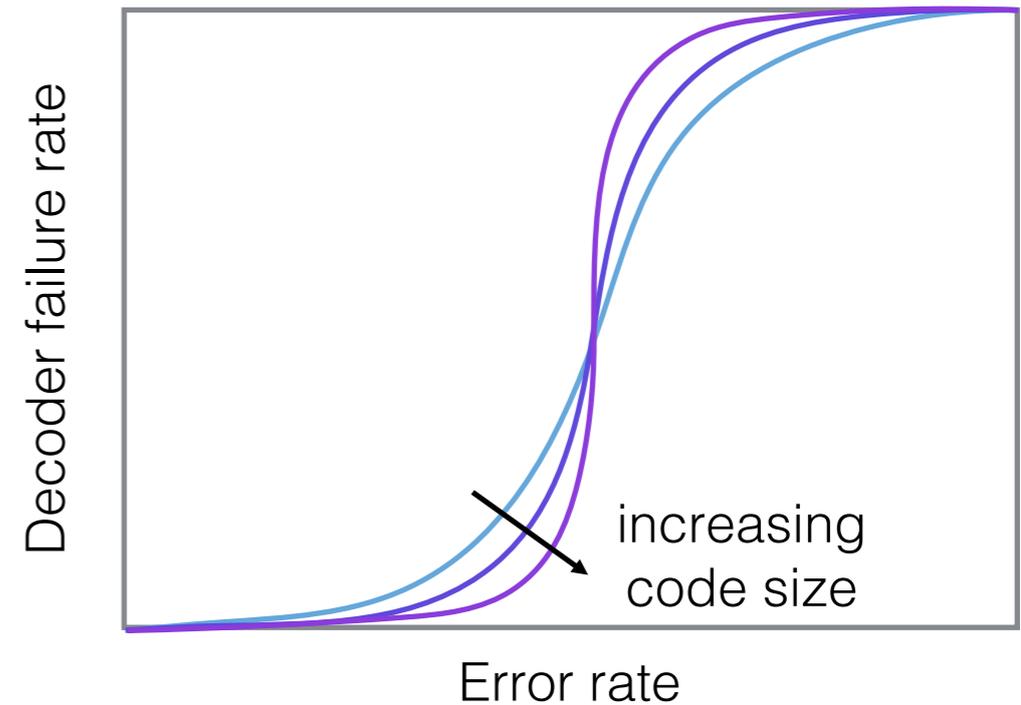


- The decoder is part of a feedback loop that determines how the quantum state should be addressed.
- If the measurements outcomes cannot be processed by the decoder as quickly as they are being produced, we end up with a backlog problem.

	<b>Clock speed</b>	Estimated decoding time (for a 20x20 lattice)	
Superconducting qubits	<b>20ns</b>	<b>400 ns</b>	1411.7403
Trapped ions	<b>480ns</b>	<b>10,000 ns</b>	1709.06952
Photons	<b>2 ns</b>	<b>40 ns</b>	
NV centres (electron)	<b>10 <math>\mu</math>s</b>	<b>200,000 ns</b>	Science 356, 634, 928-932
NV centres (nuclear spin)	<b>500 <math>\mu</math>s</b>	<b>10,000,000 ns</b>	Science 356, 634, 928-932

# What makes a good decoder?

- The optimal decoder cannot always correct errors, instead we see **threshold behavior**
- The optimal decoder has exponential complexity. Approximate algorithms trade off speed for lowering the threshold.



	Threshold (2d surface code)	Worst case Complexity	Works with any geometry?
Optimal decoder	11.0%	$e^N$	✓
MWPM (Harrington, Fowler)	10.3%	$O(N^3)$	x
RG decoder (Harrington, Duclos-Cianci, Poulin)	8.2%	$O(N \log N)$	x
HDRG (Wooton)	7.3%	$O(N^2)$	✓
<b>Union Find decoder</b>	<b>9.9%</b>	<b><math>O(\alpha(N) N)</math></b>	<b>✓</b>

1. What is decoding and how fast does it need to be?

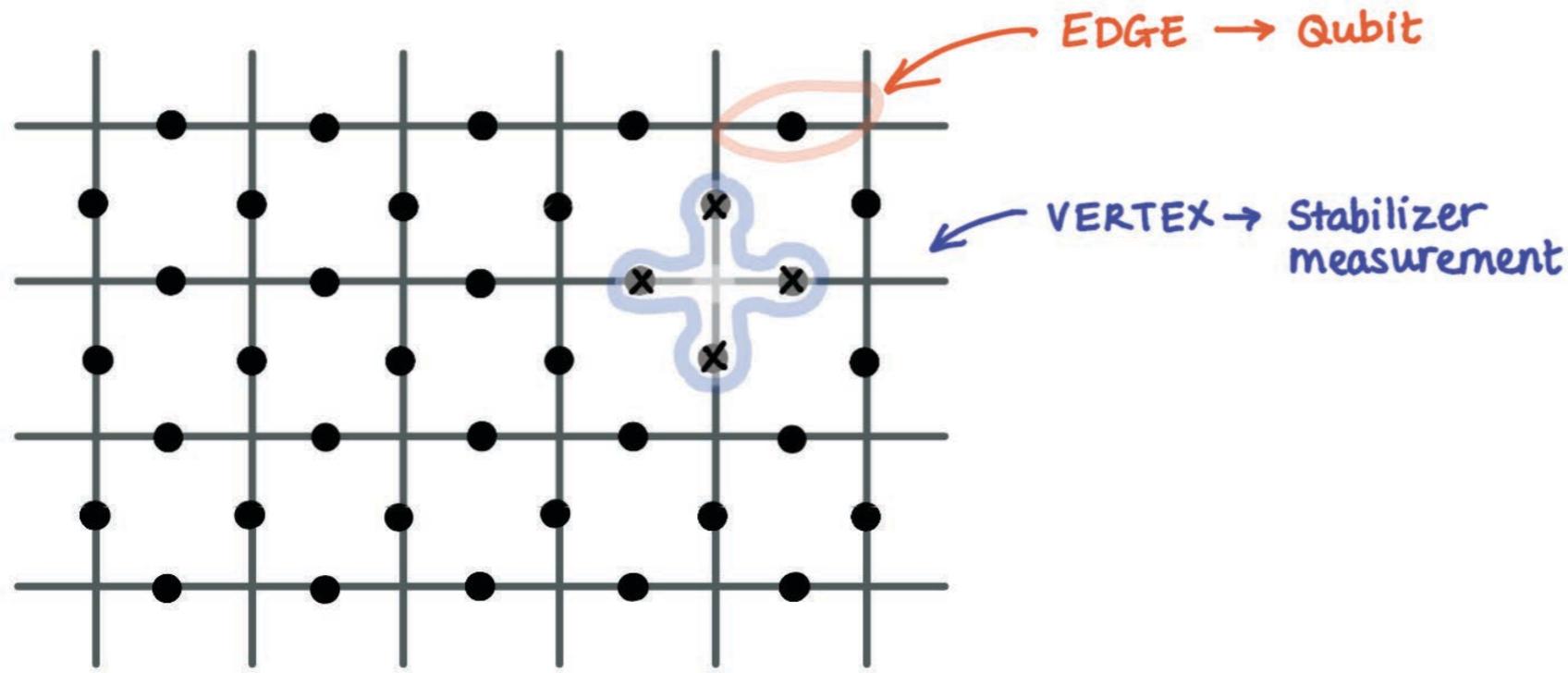


2. The Union Find decoder

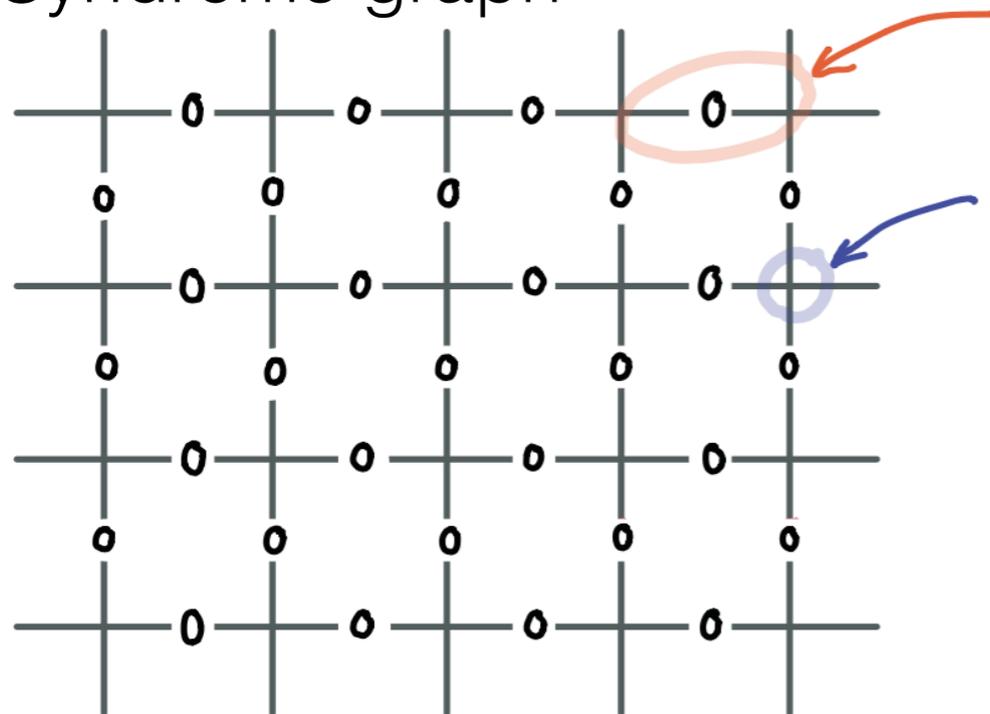
3. Achieving almost-linear time

# Syndrome graph

2D surface code



Syndrome graph



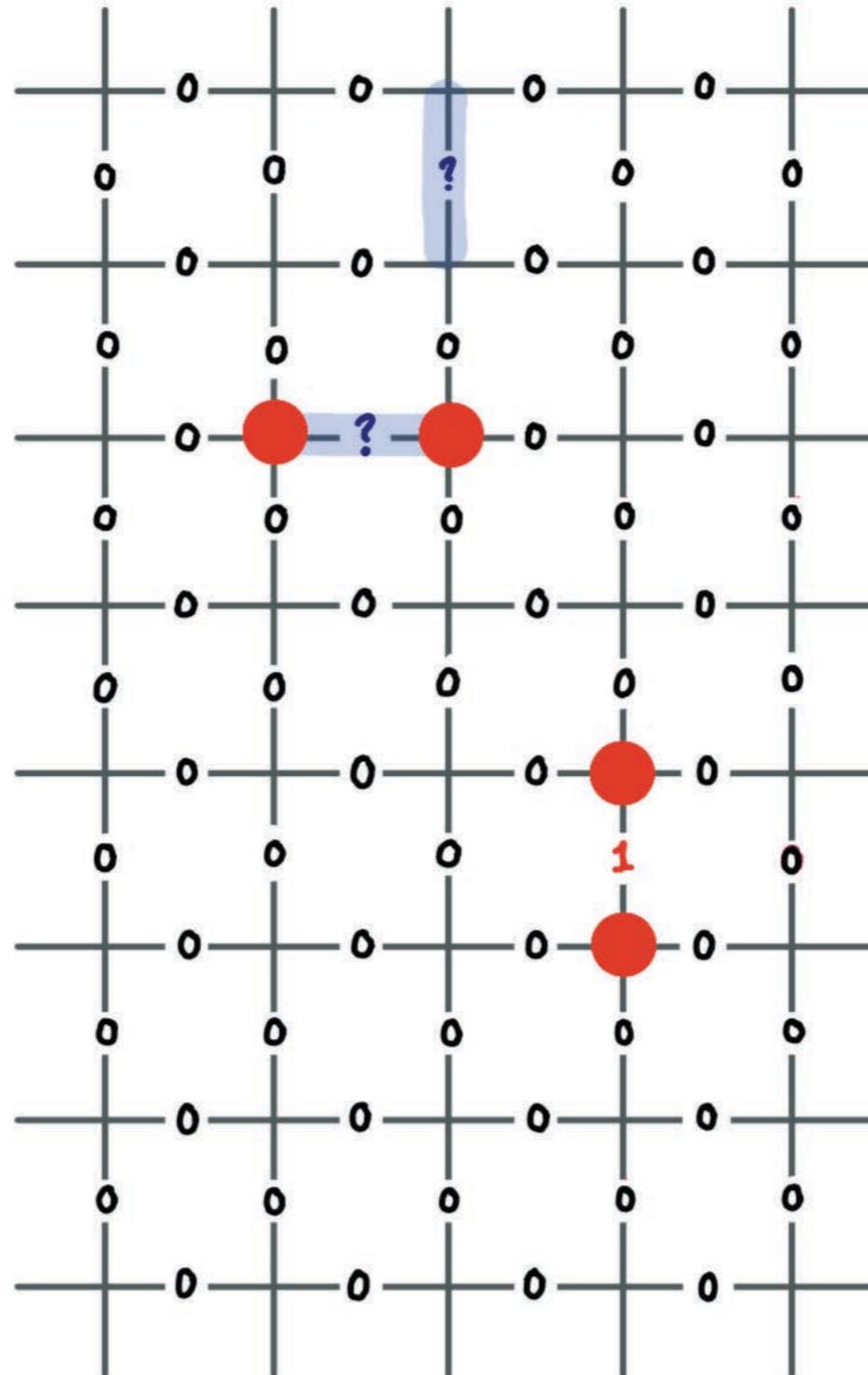
- The syndrome graph is a representation of (one basis of) the code.
- Each edge represents the error state of a qubit.
- Stabilizer measurement outcomes are associated with each vertex.
- The sum of the outcomes around each vertex should always be even.

# Identifying errors

## Erasure error

a.k.a loss

- A Pauli error occurs with 50% probability
- The location of the error is known.
- For example: we reinitialize a qubit into the maximally mixed state.



## Bit-flip error

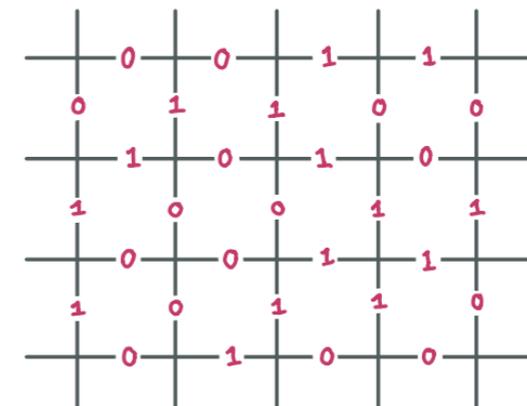
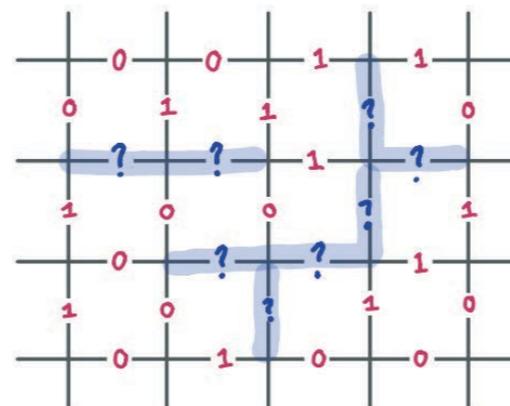
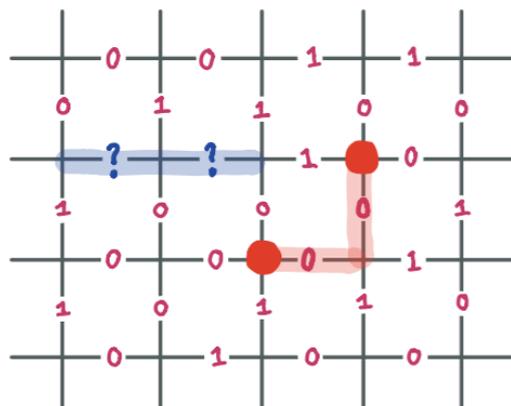
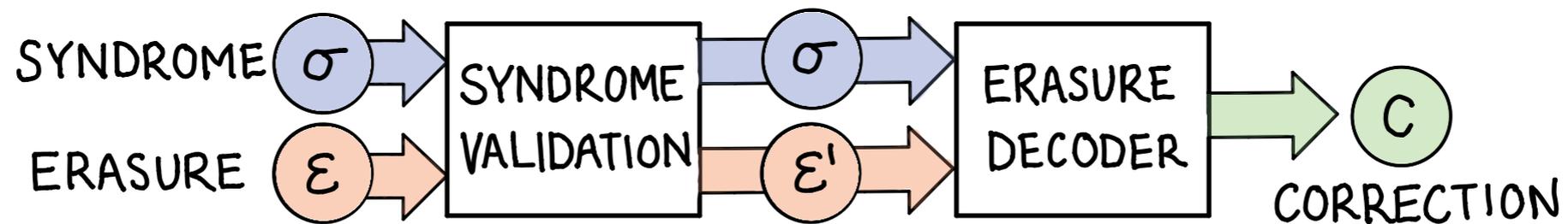
(Z-basis Pauli error)

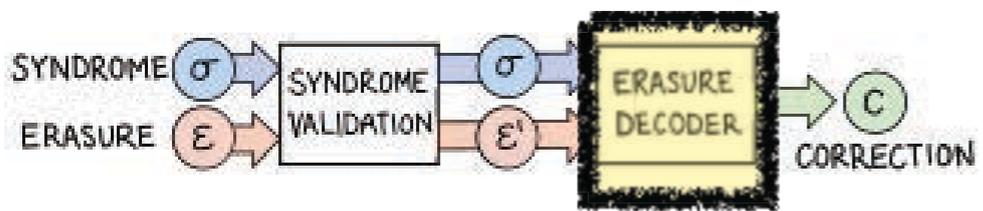
- The location of the error is unknown
- For example: the photon is in the 'wrong' rail before the final detection

# The Union Find Decoder

The Union Find decoding algorithm can be broken in two stages

1. Convert stochastic errors into erasures
2. Apply the erasure decoder





# Decoding Erasure

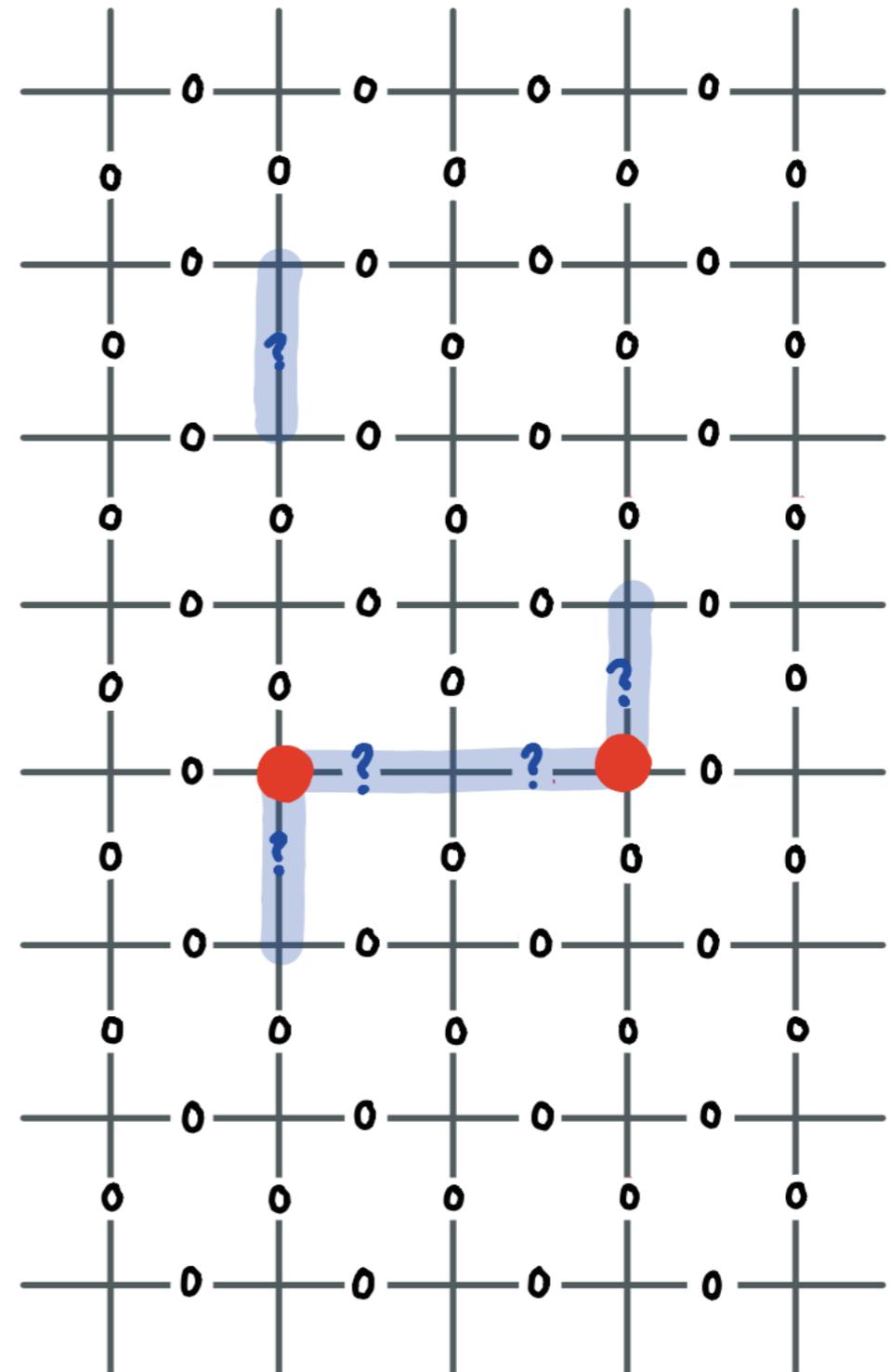
Measurement outcomes are represented as a graph where edges correspond to erased outcomes, and vertices are syndromes.

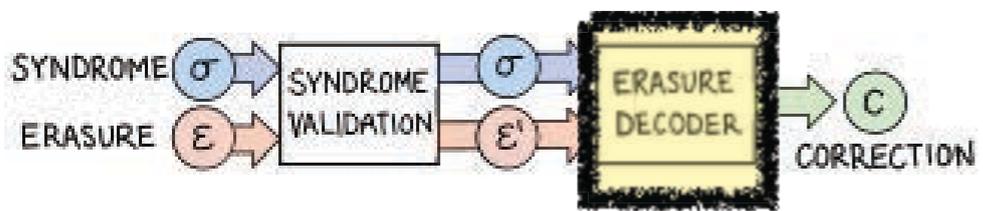
## Erasure Decoder

Grow a spanning forest to cover the erased edges.

while erased edges remain, do:

1. start at a leaf edge (u,v).
2. Remove the edge from the erasure, and apply the rules:
  - (R1) If the vertex u is odd, set the edge value to 1 and flip the value of v
  - (R2) If the vertex u is even, set the edge value to 0





# Decoding Erasure

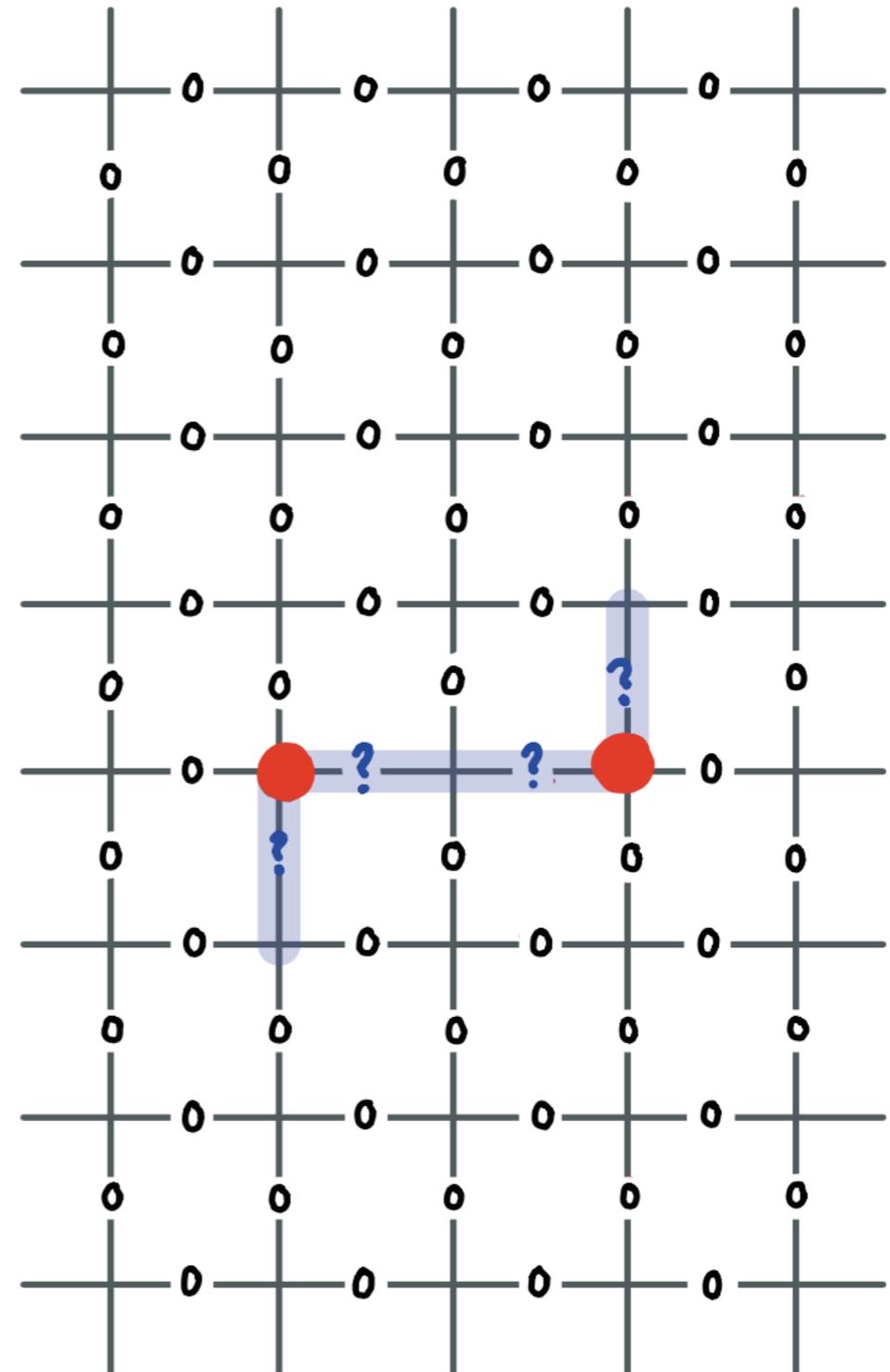
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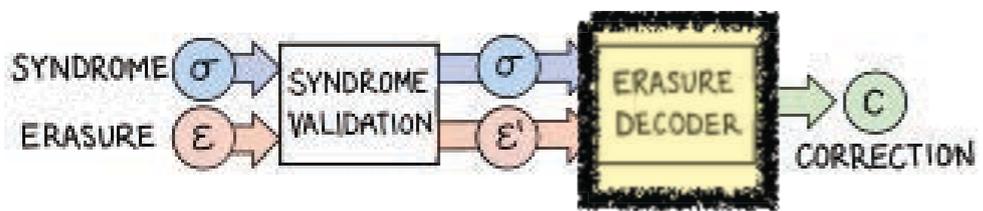
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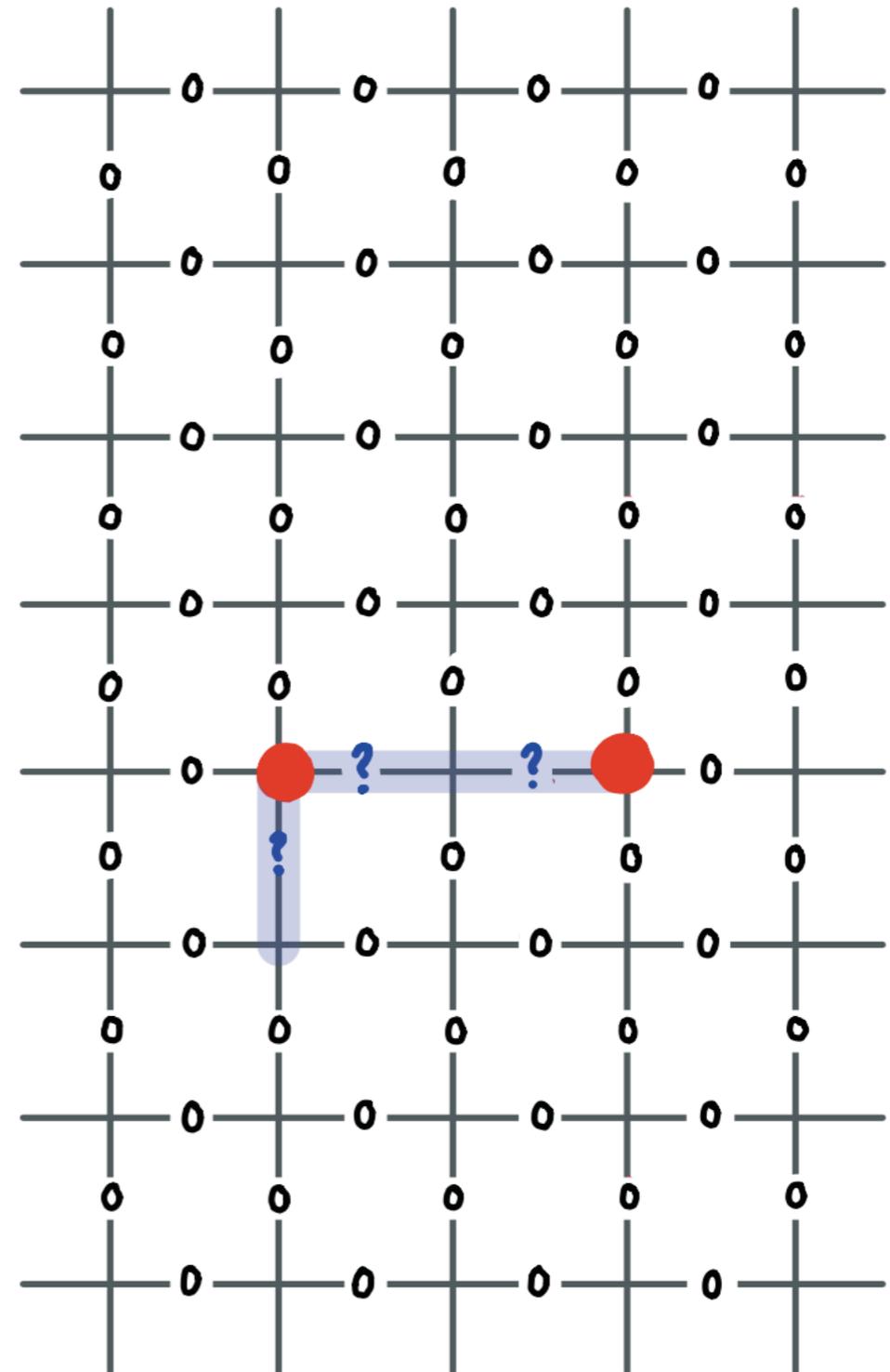
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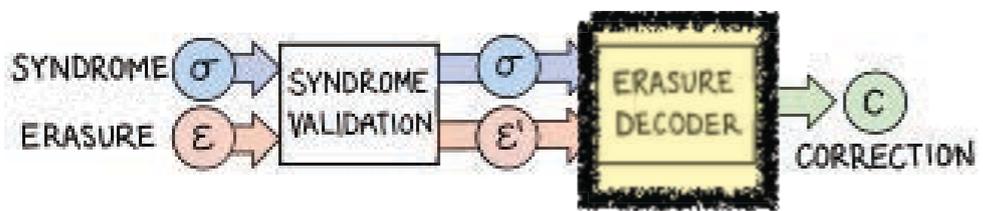
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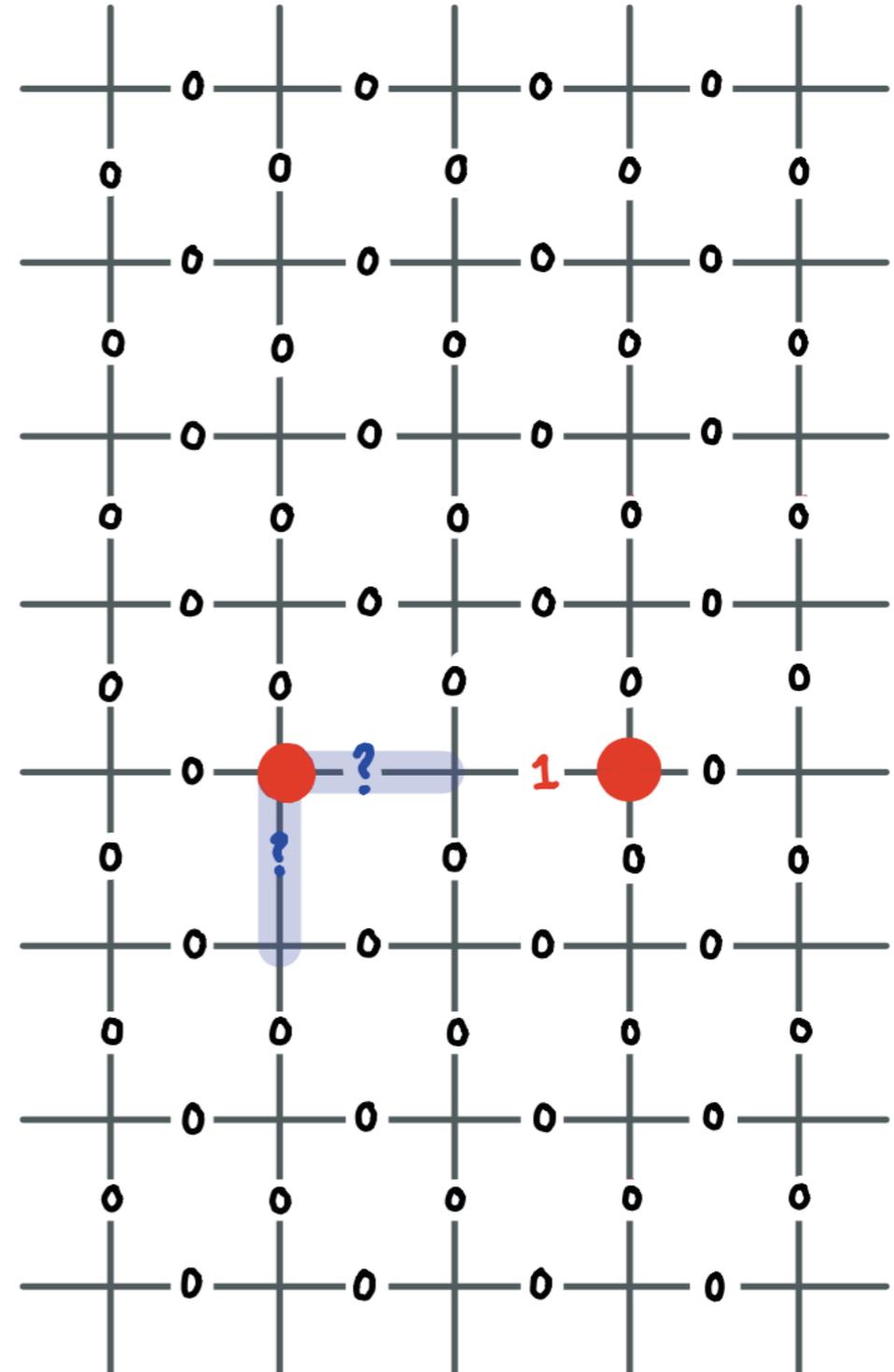
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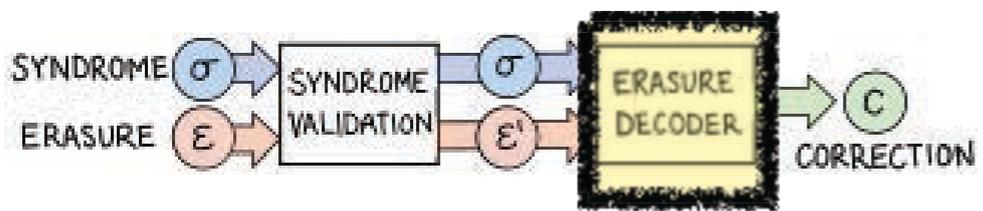
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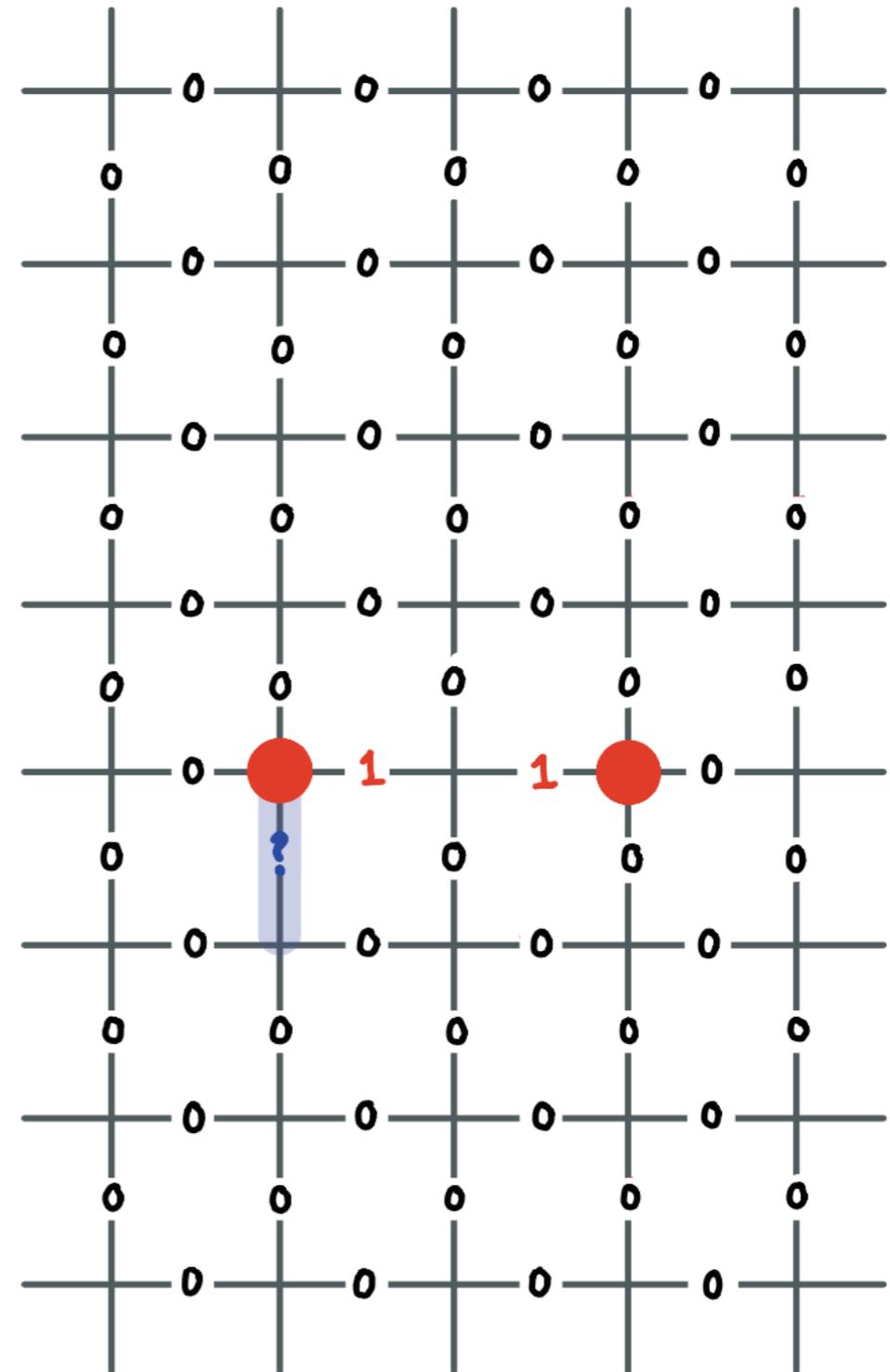
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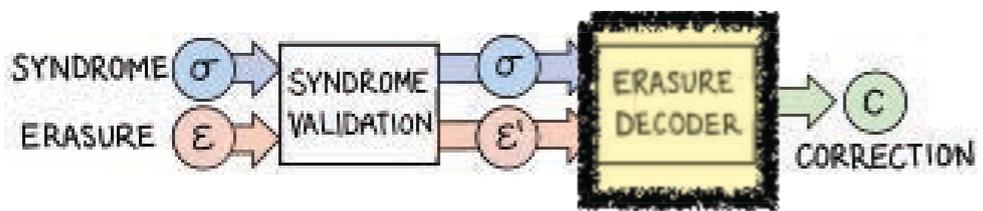
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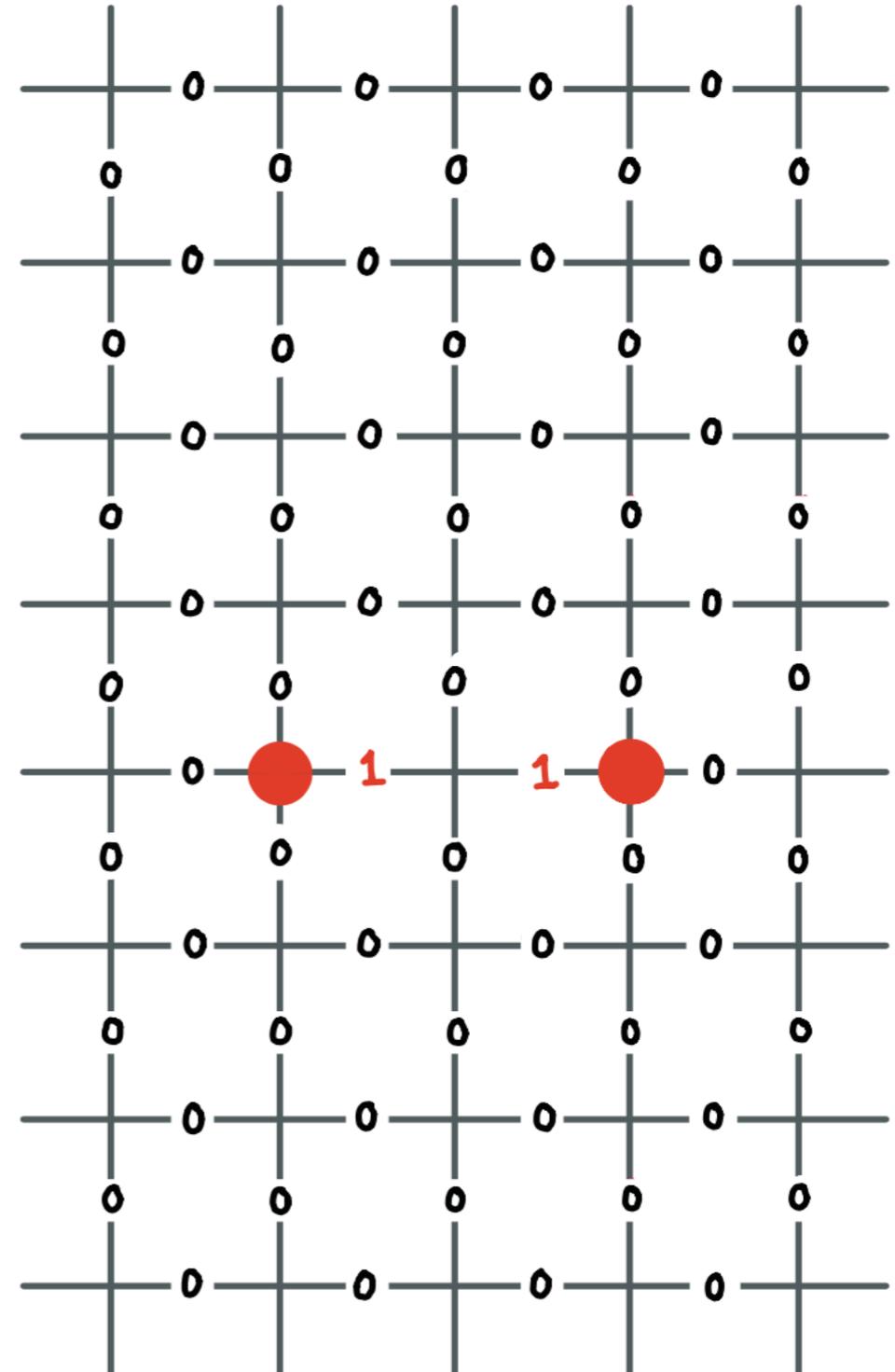
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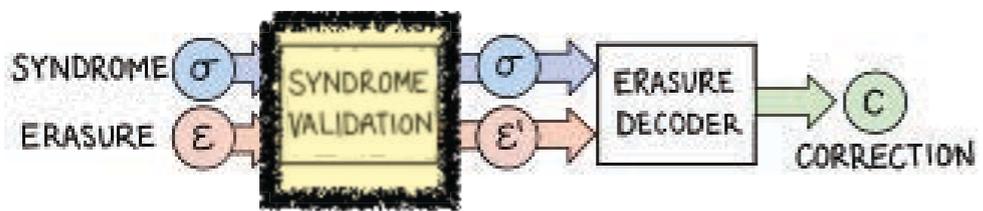
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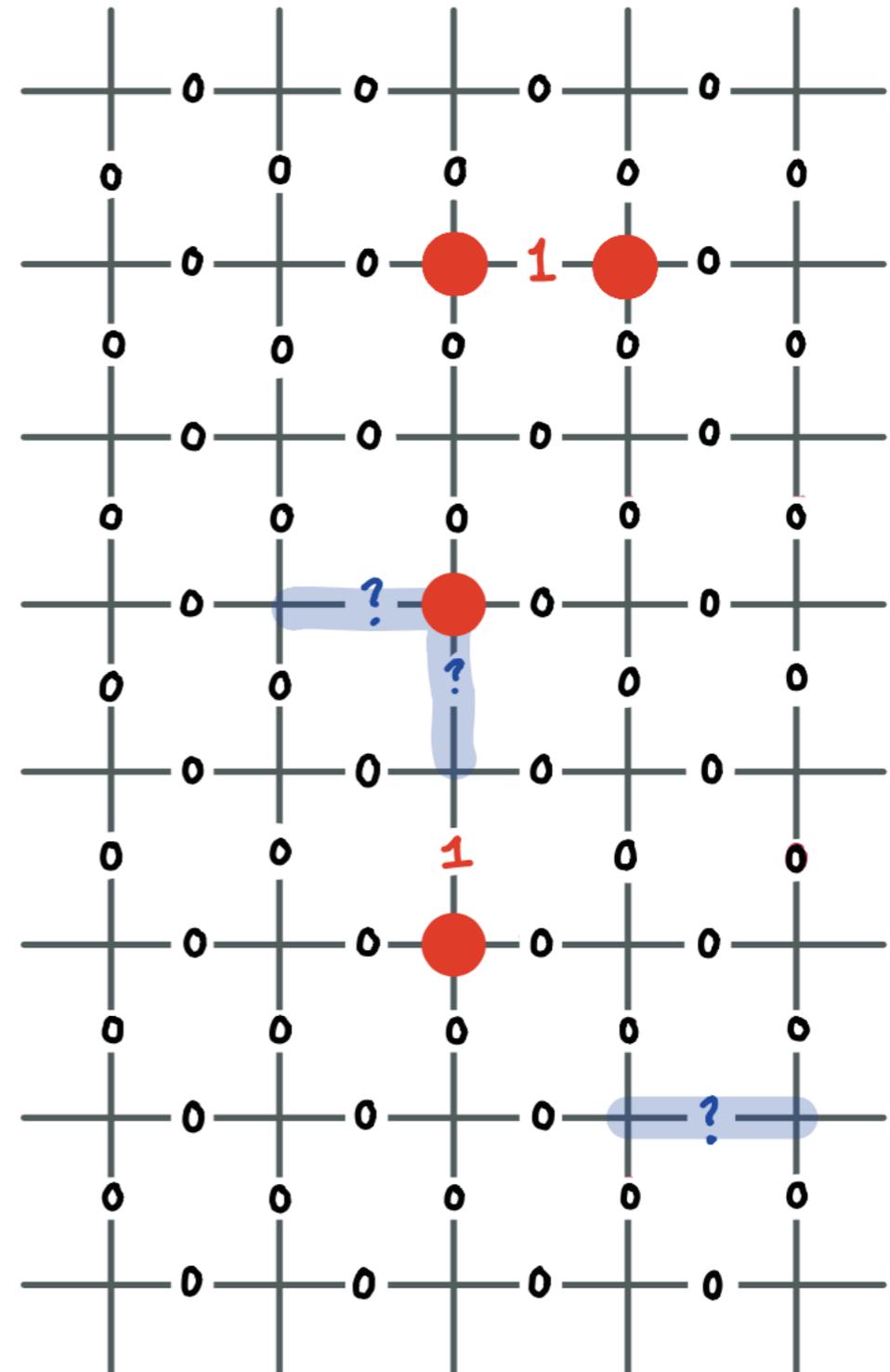
Create a list of all odd clusters

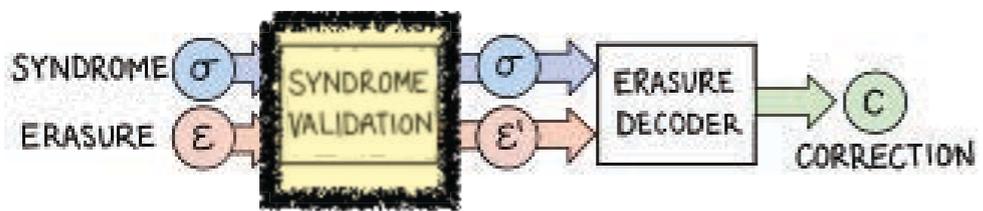
while there exists an odd cluster:

Iterate over all odd clusters:

1. Grow the cluster by a half-edge
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Add any edges that are full grown to the list of erased edges.





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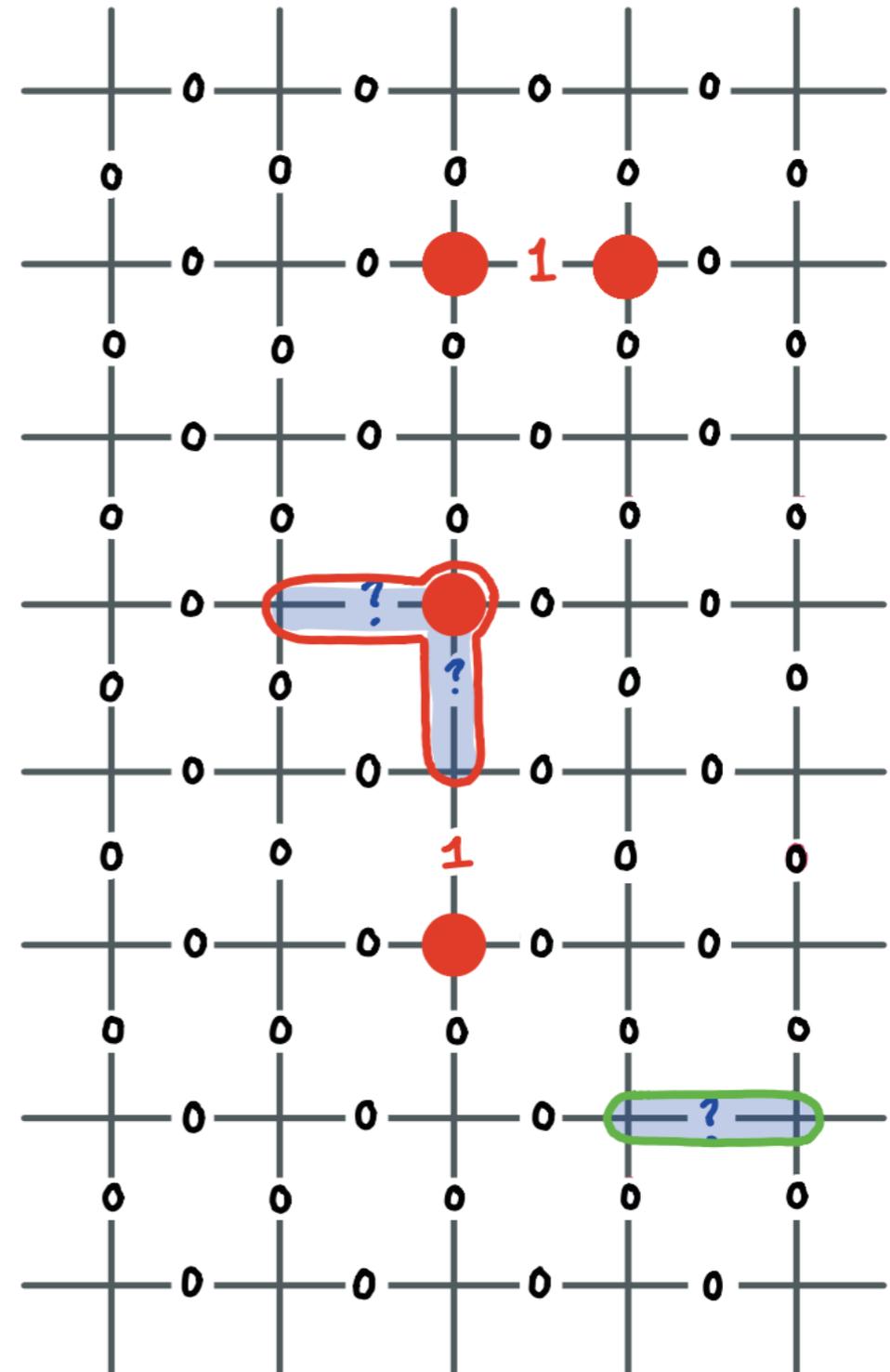
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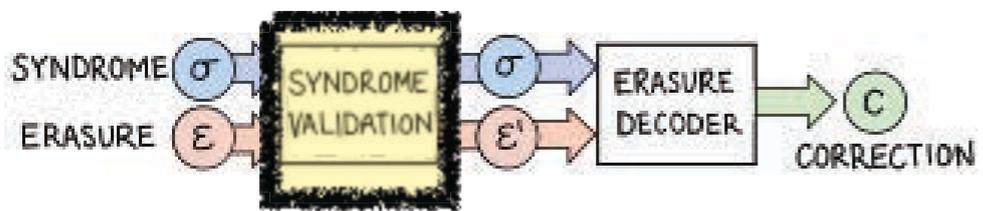
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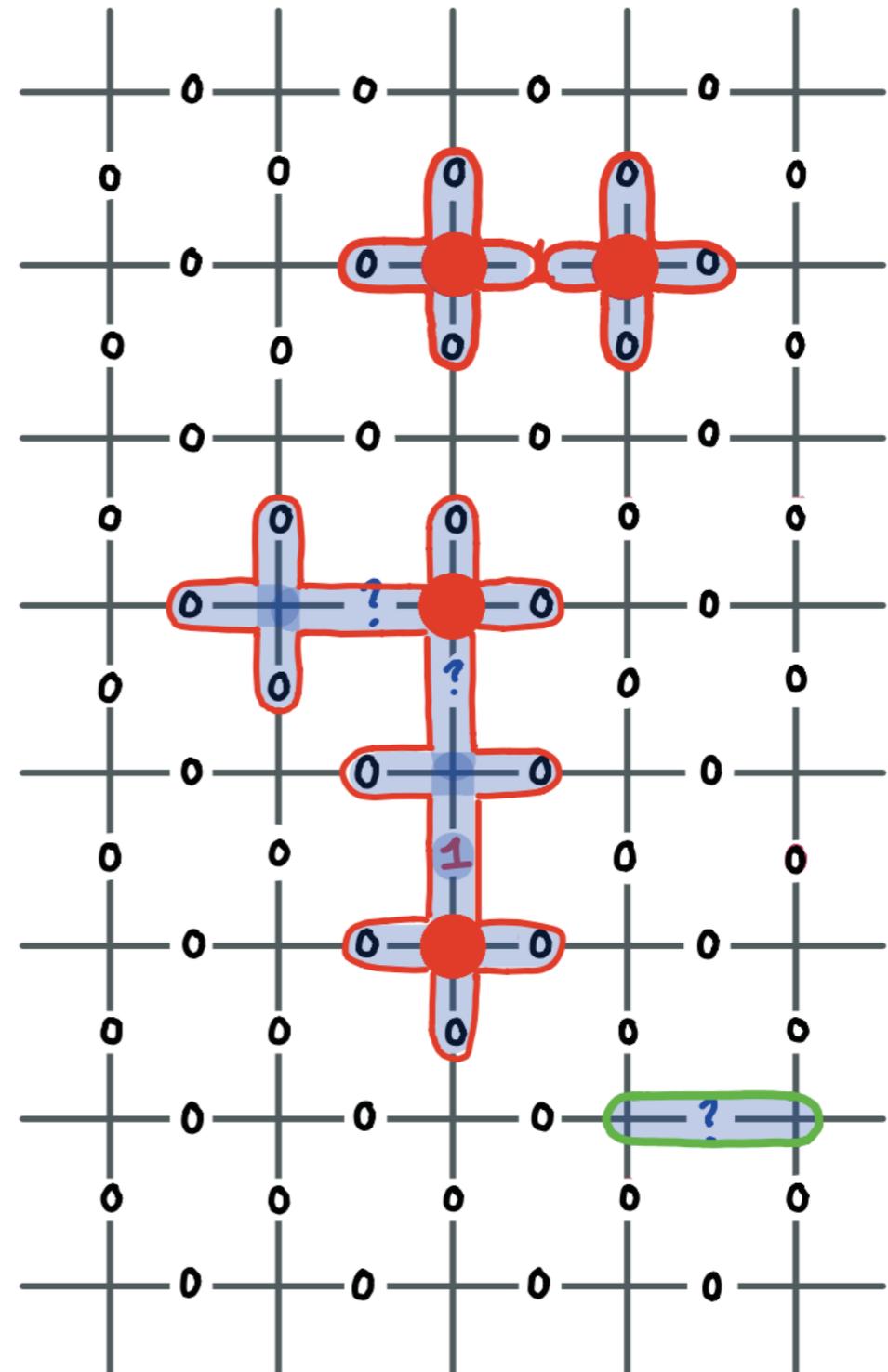
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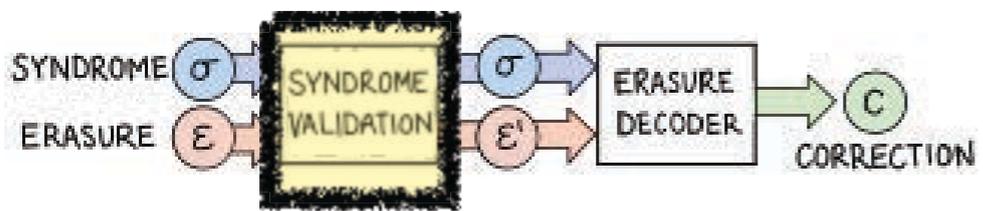
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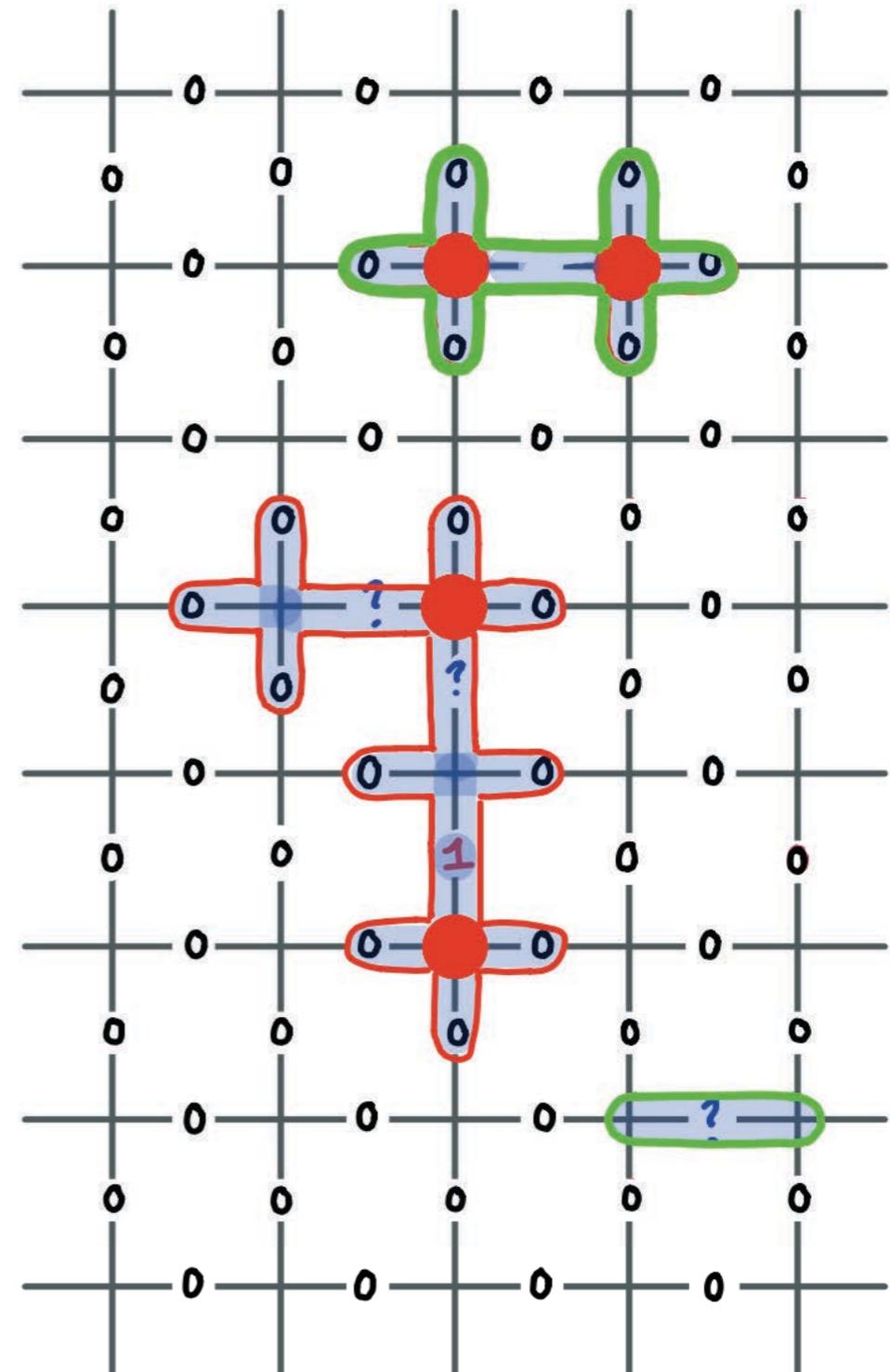
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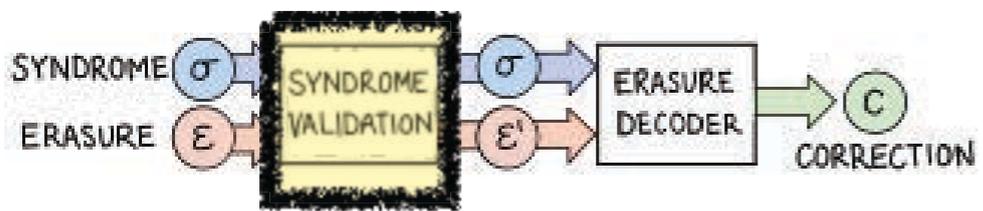
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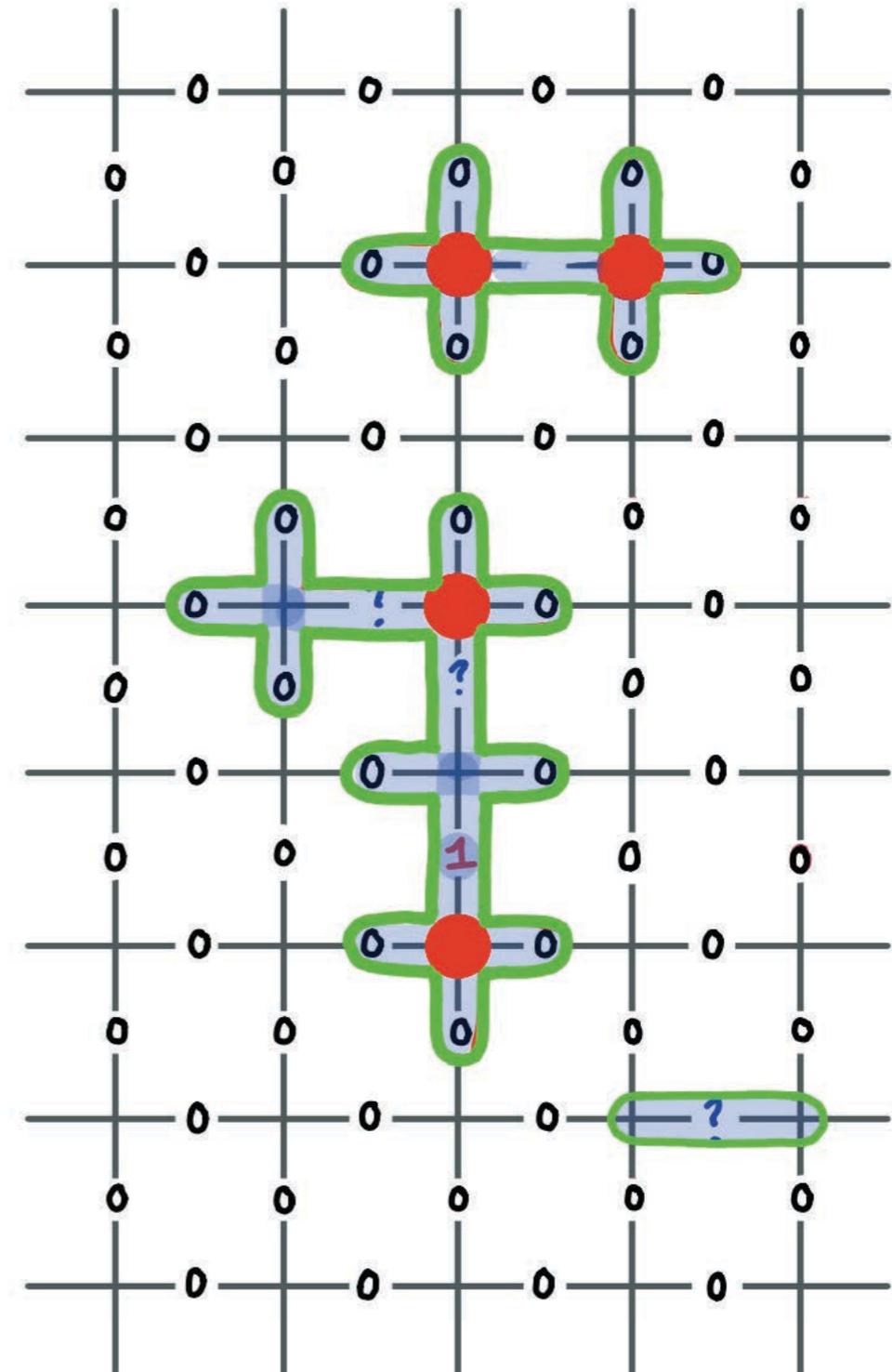
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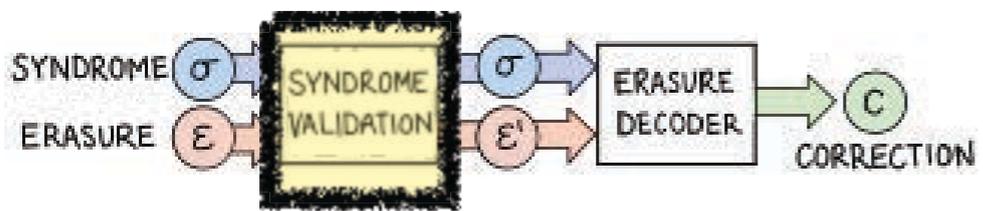
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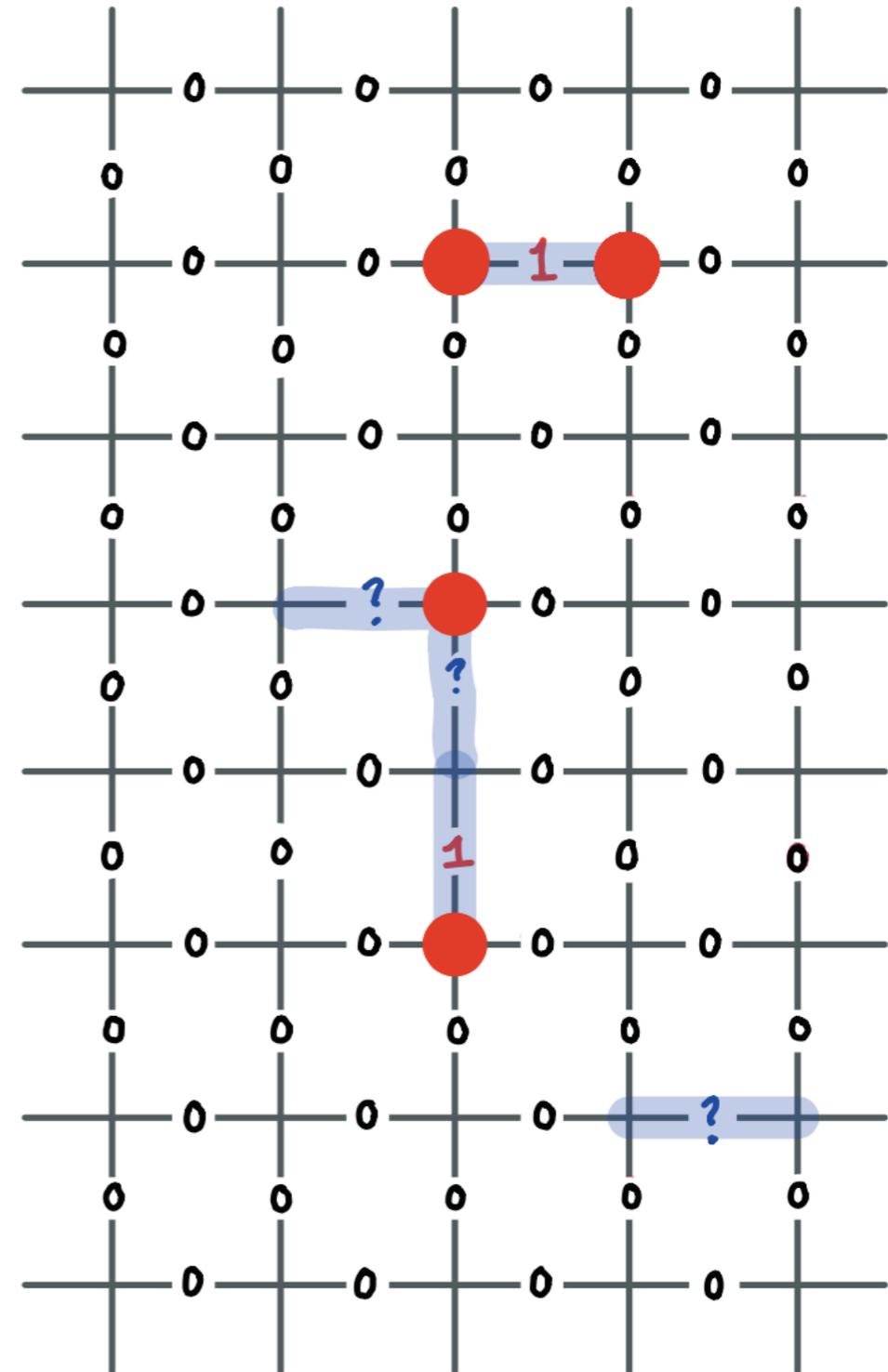
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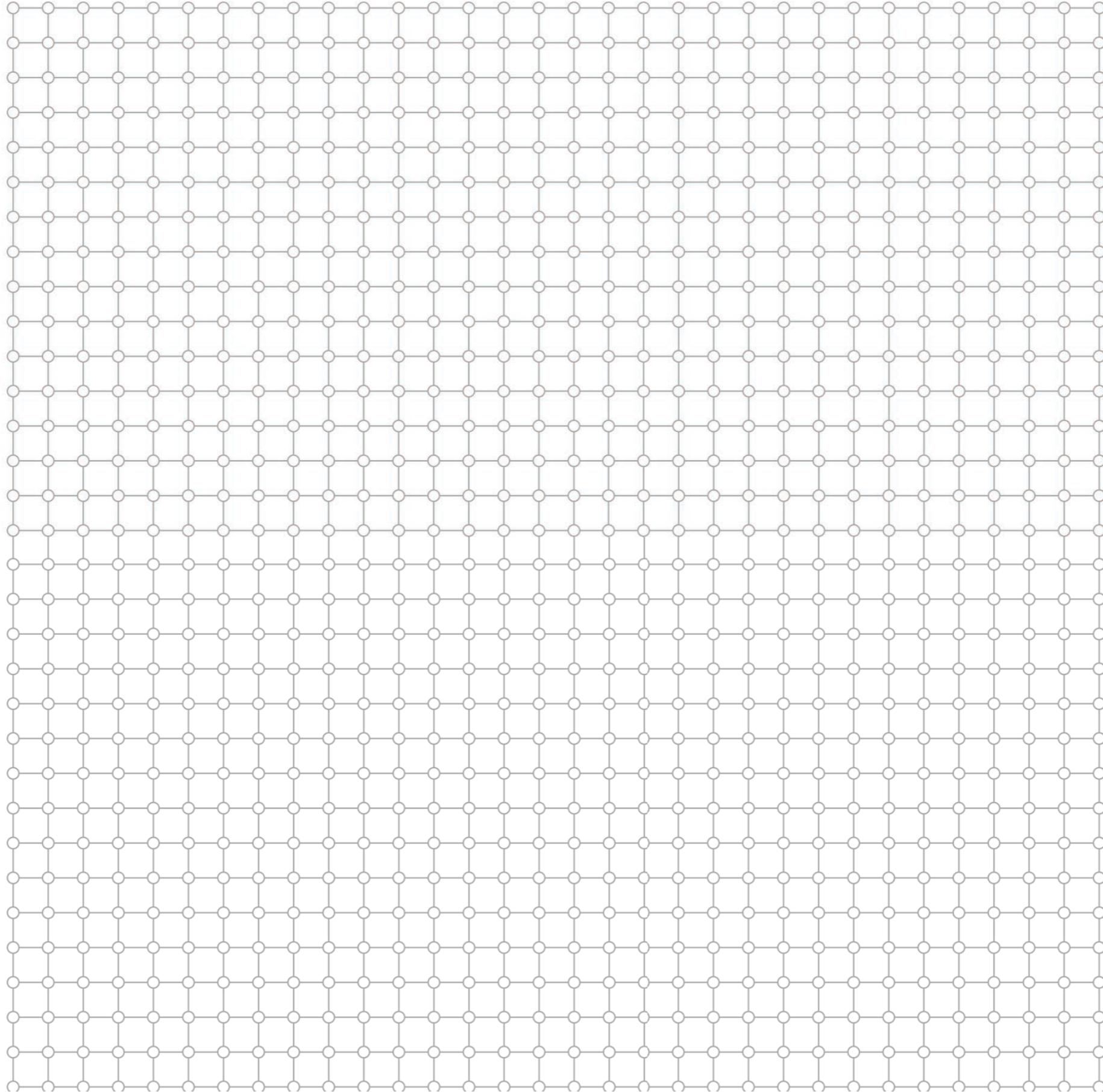
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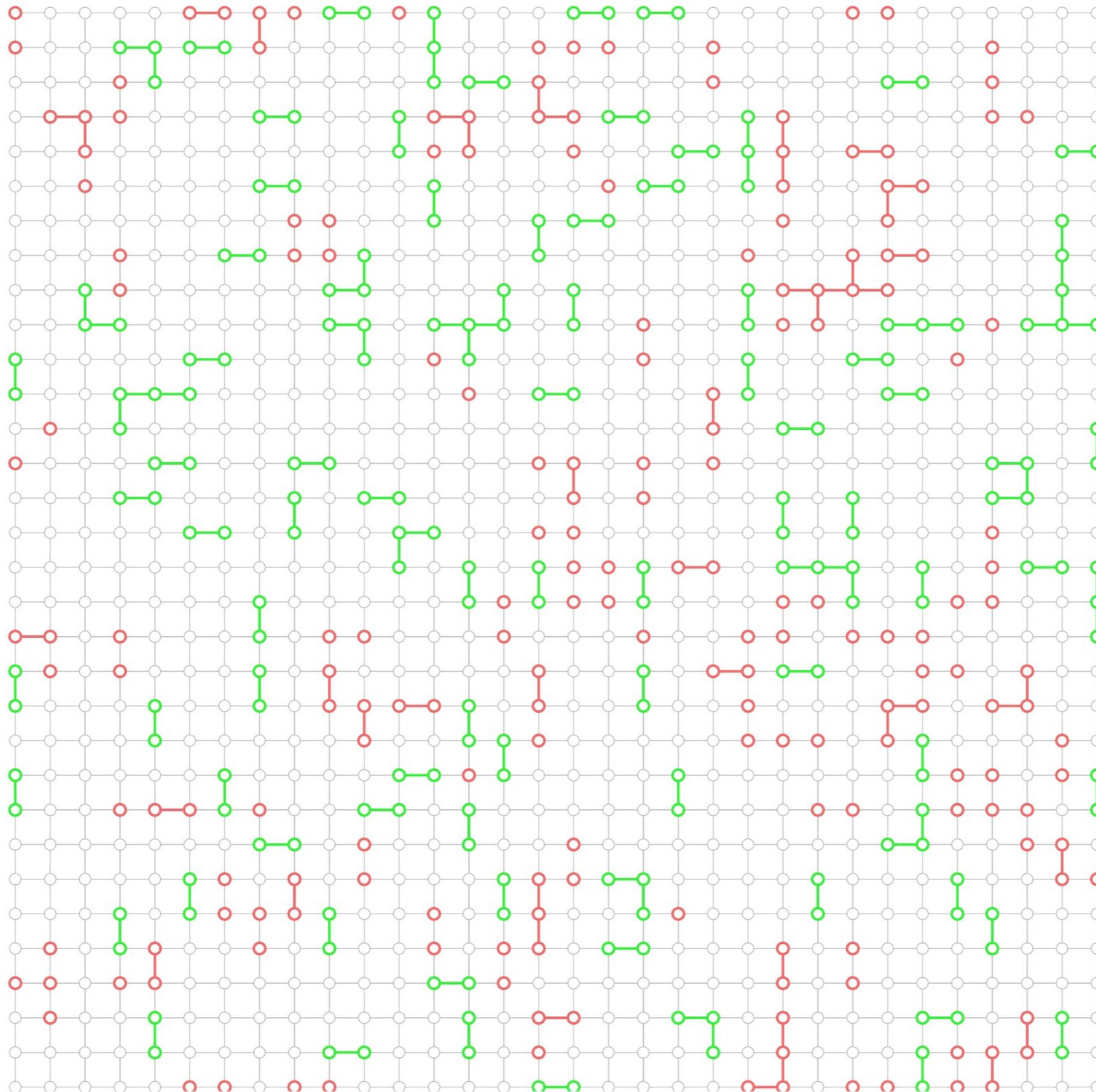


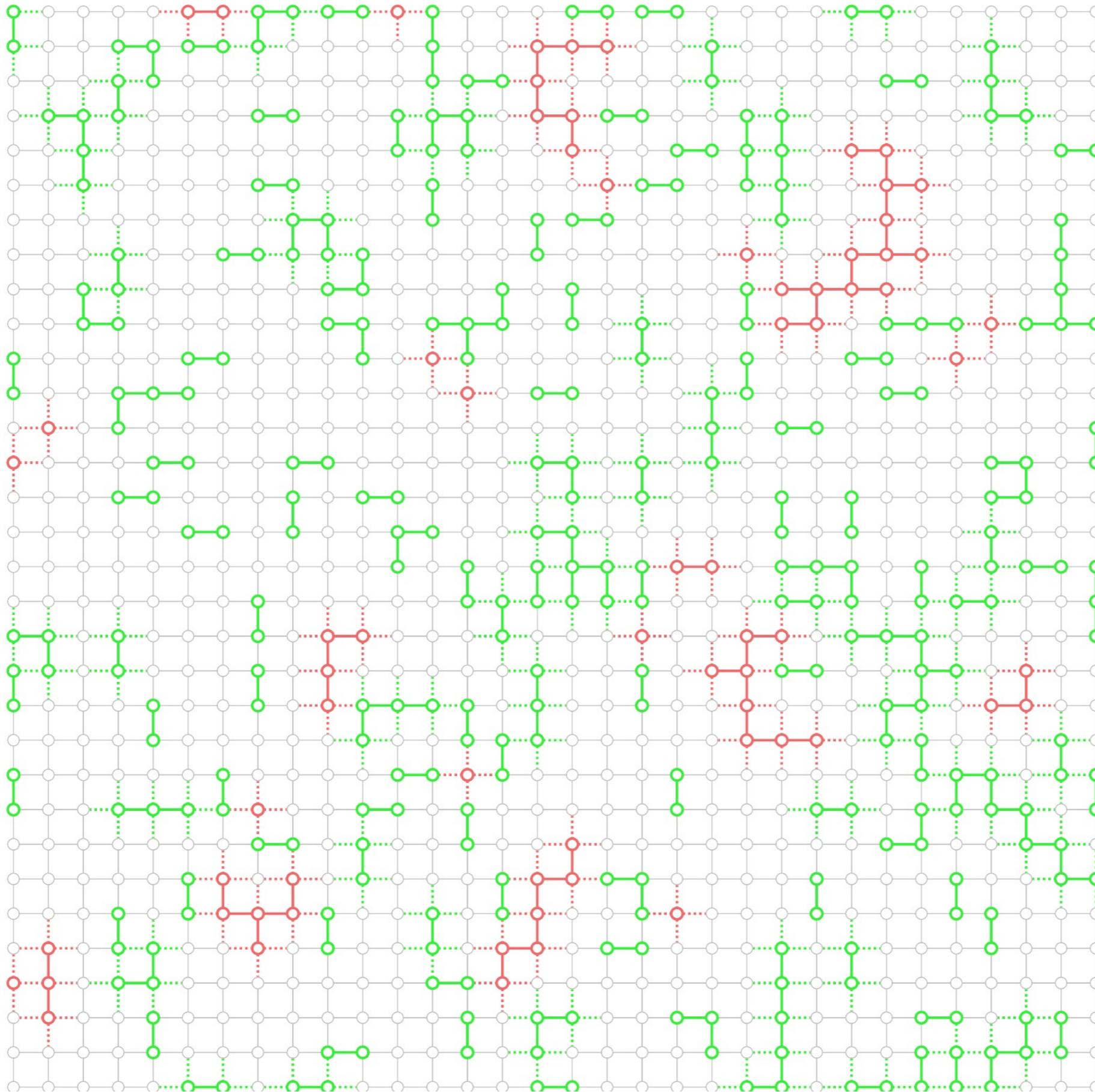
Example: decoding with 10% erasure, 5% Pauli error



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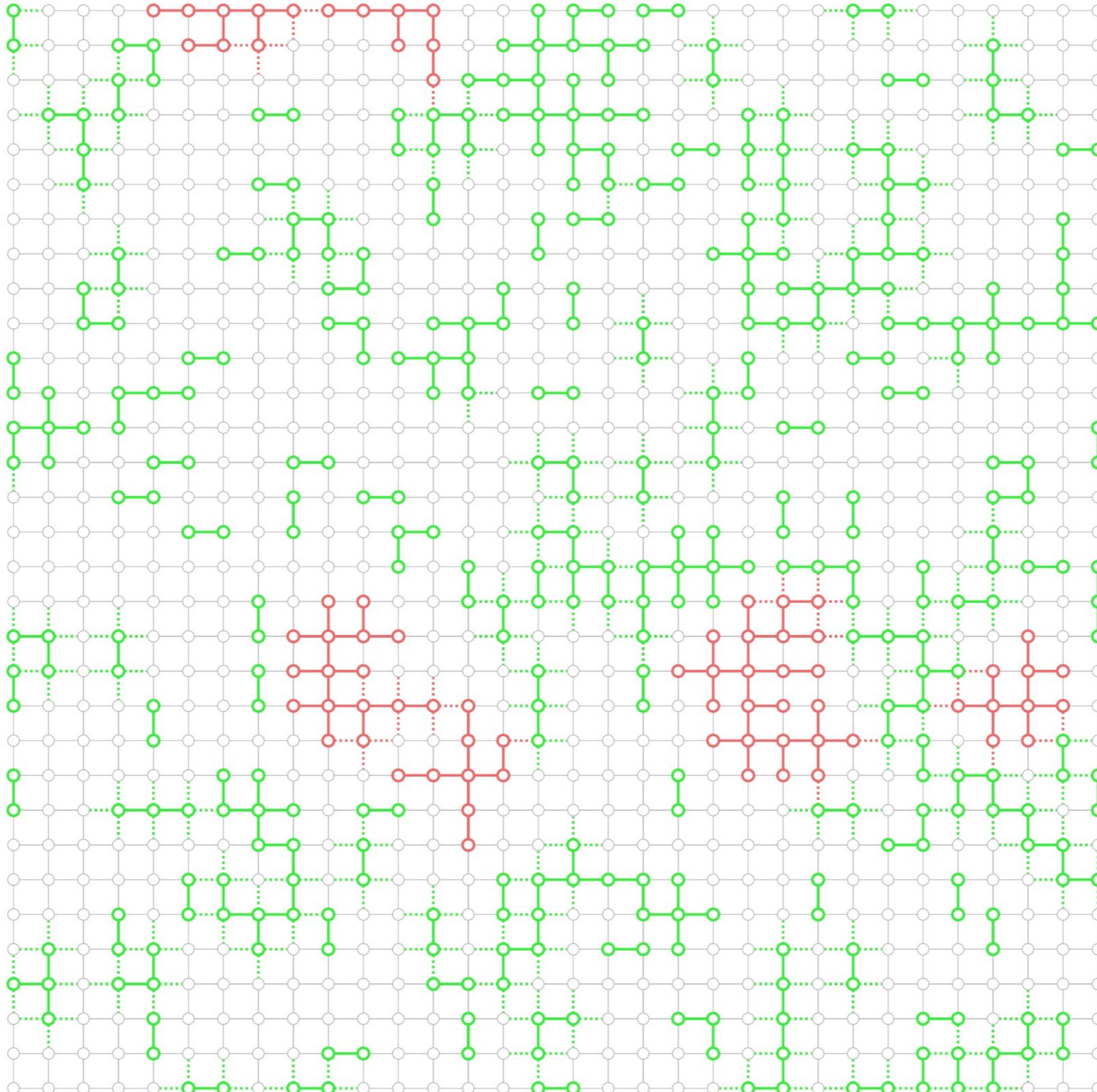
Measure Syndromes





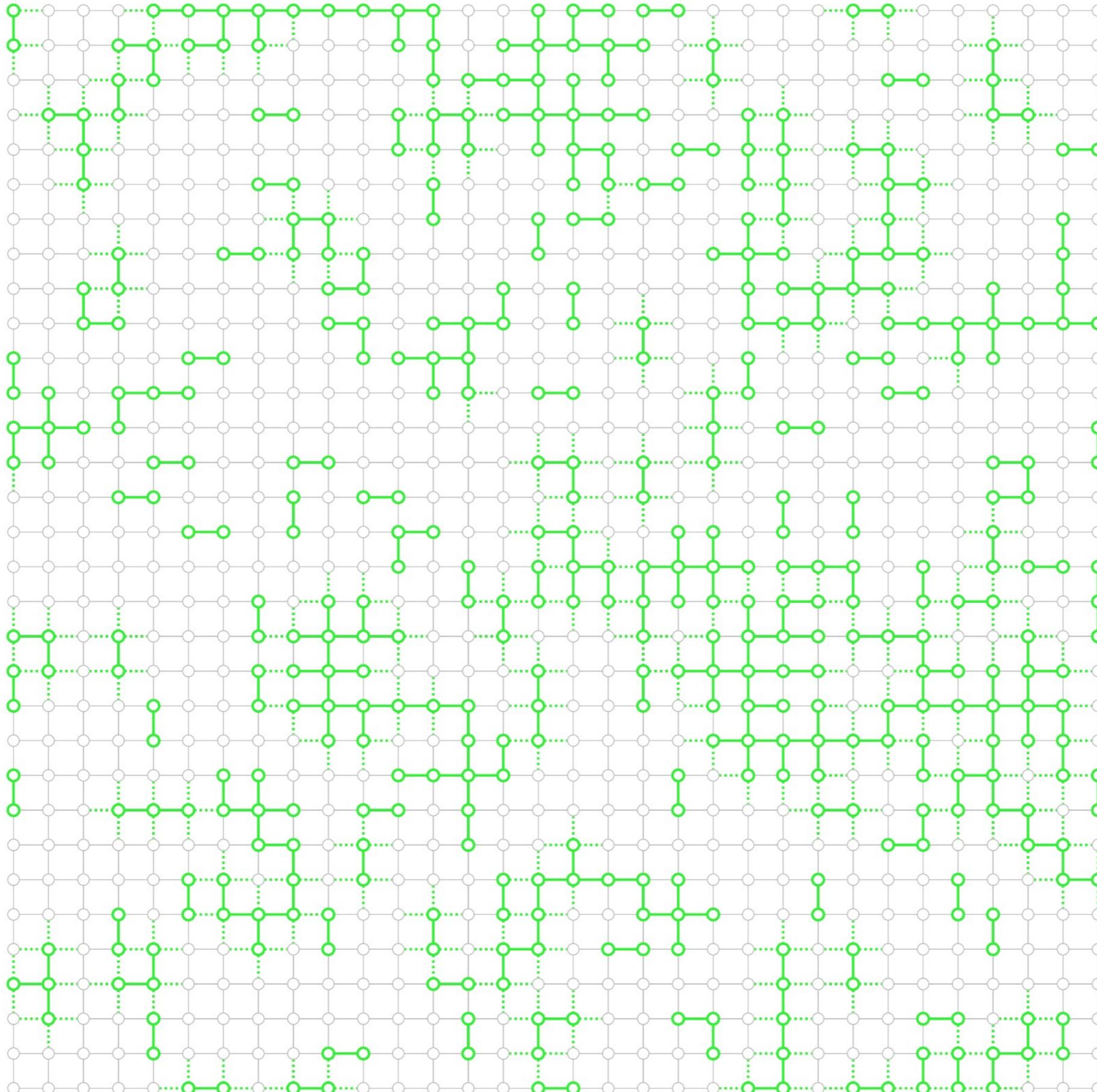
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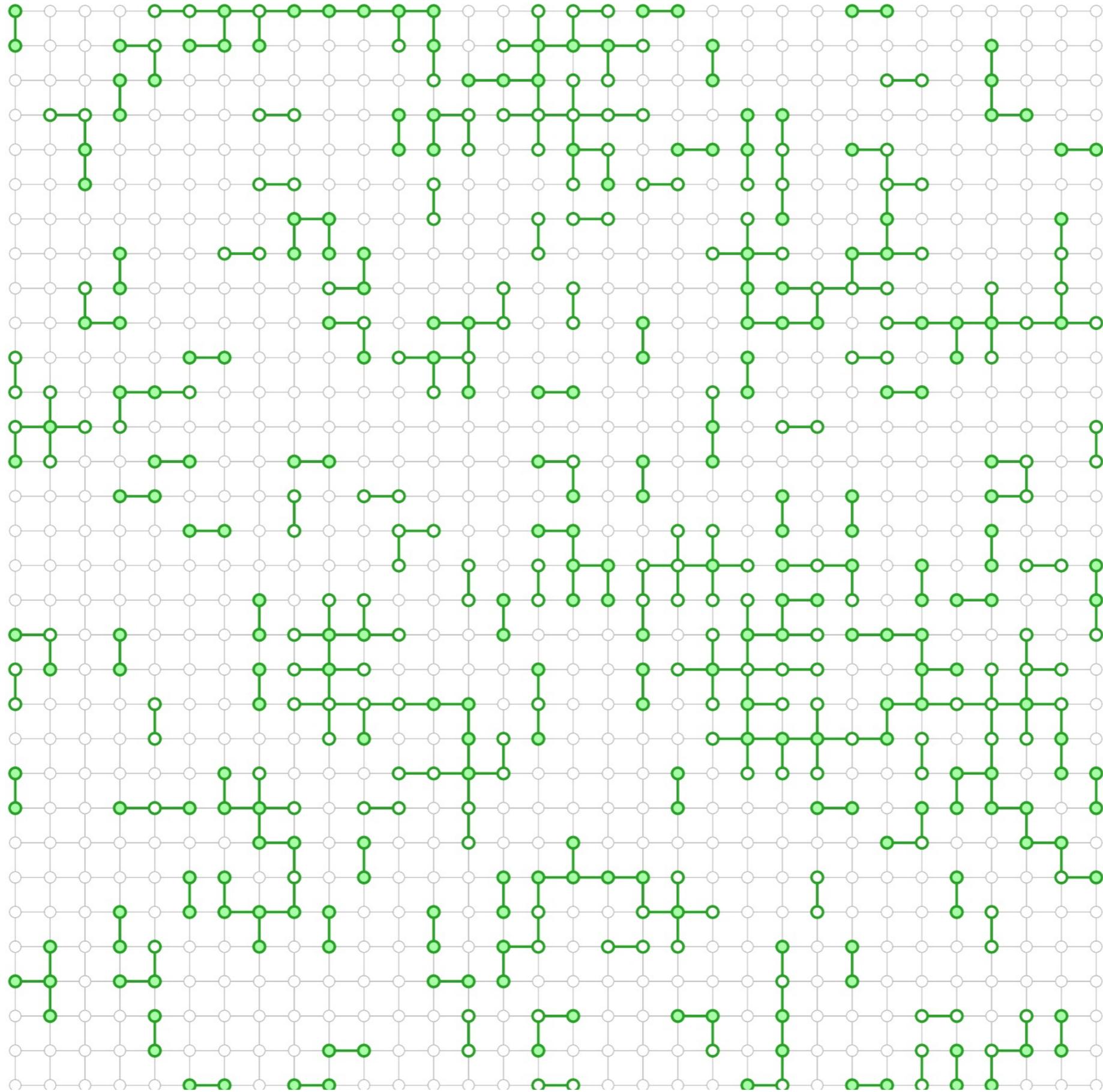
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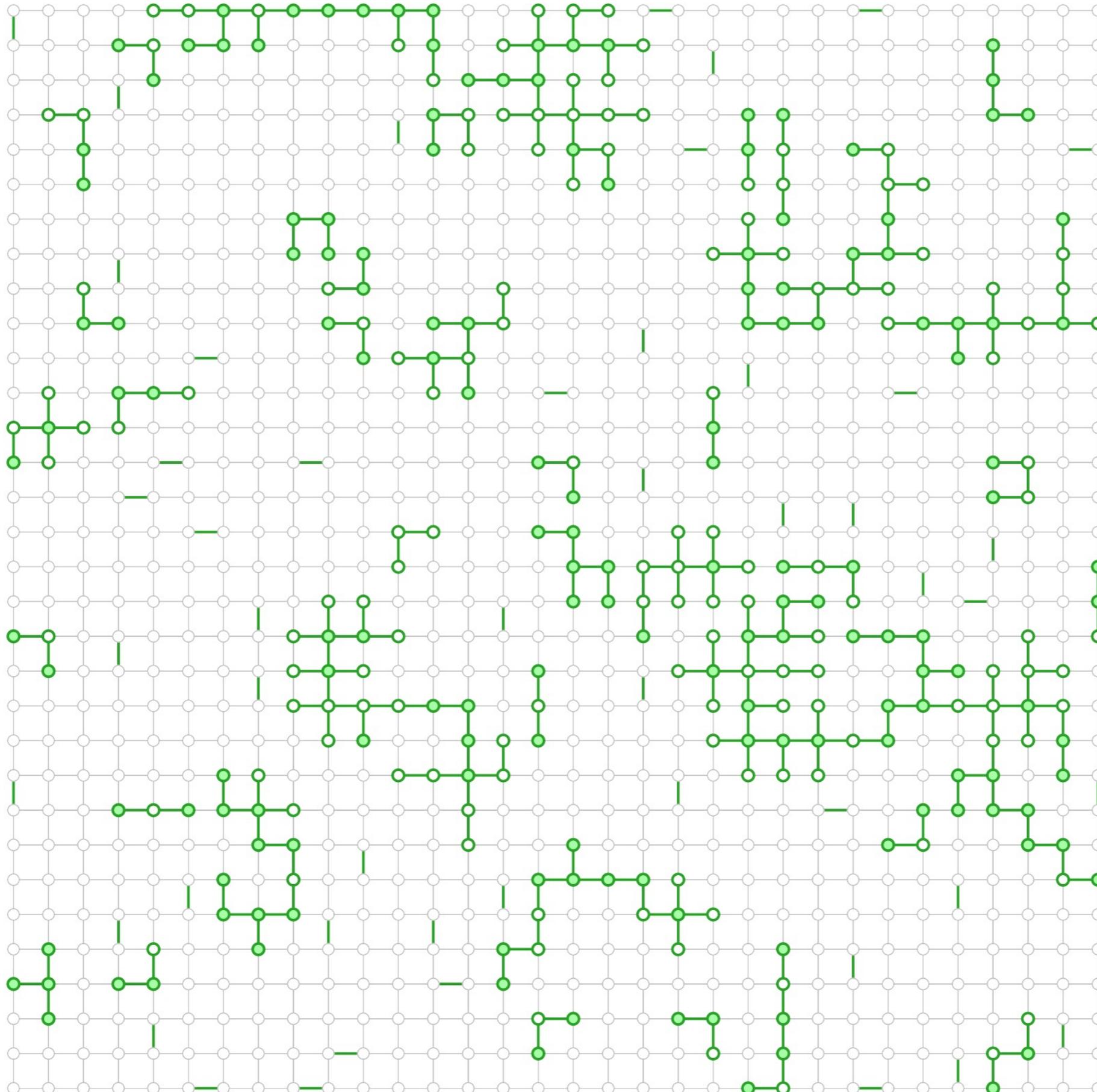
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Erasure Decoding



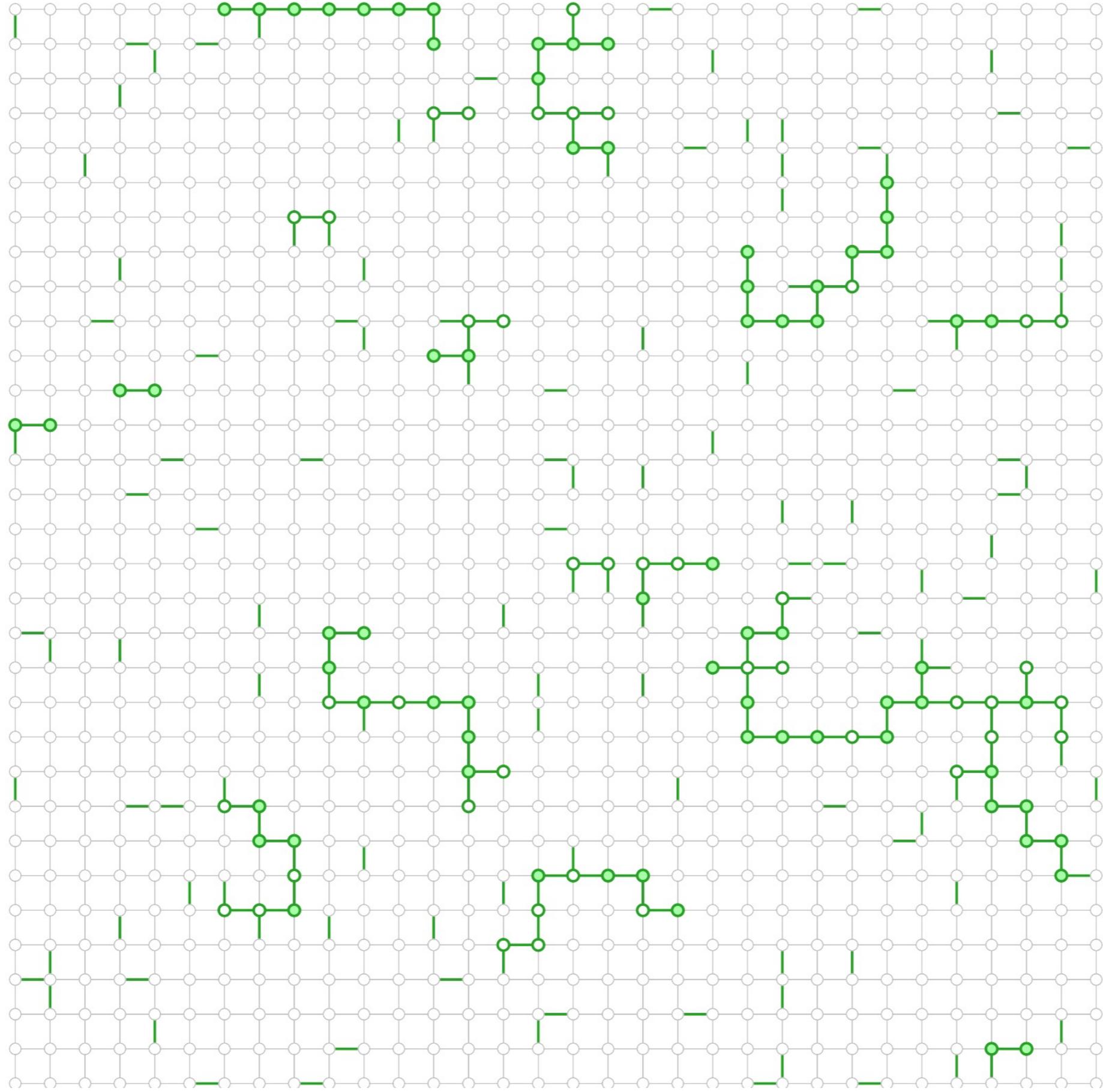
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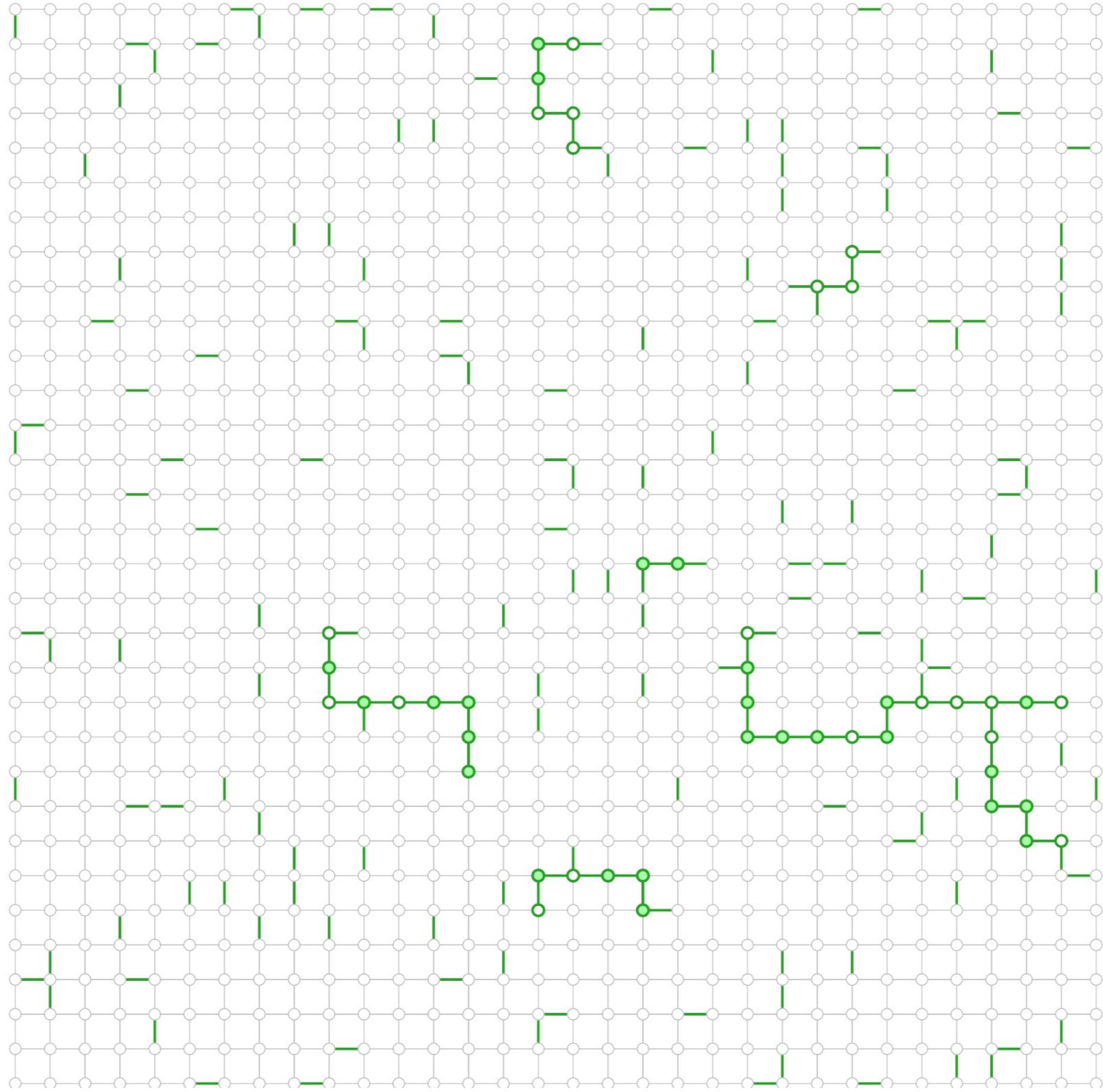
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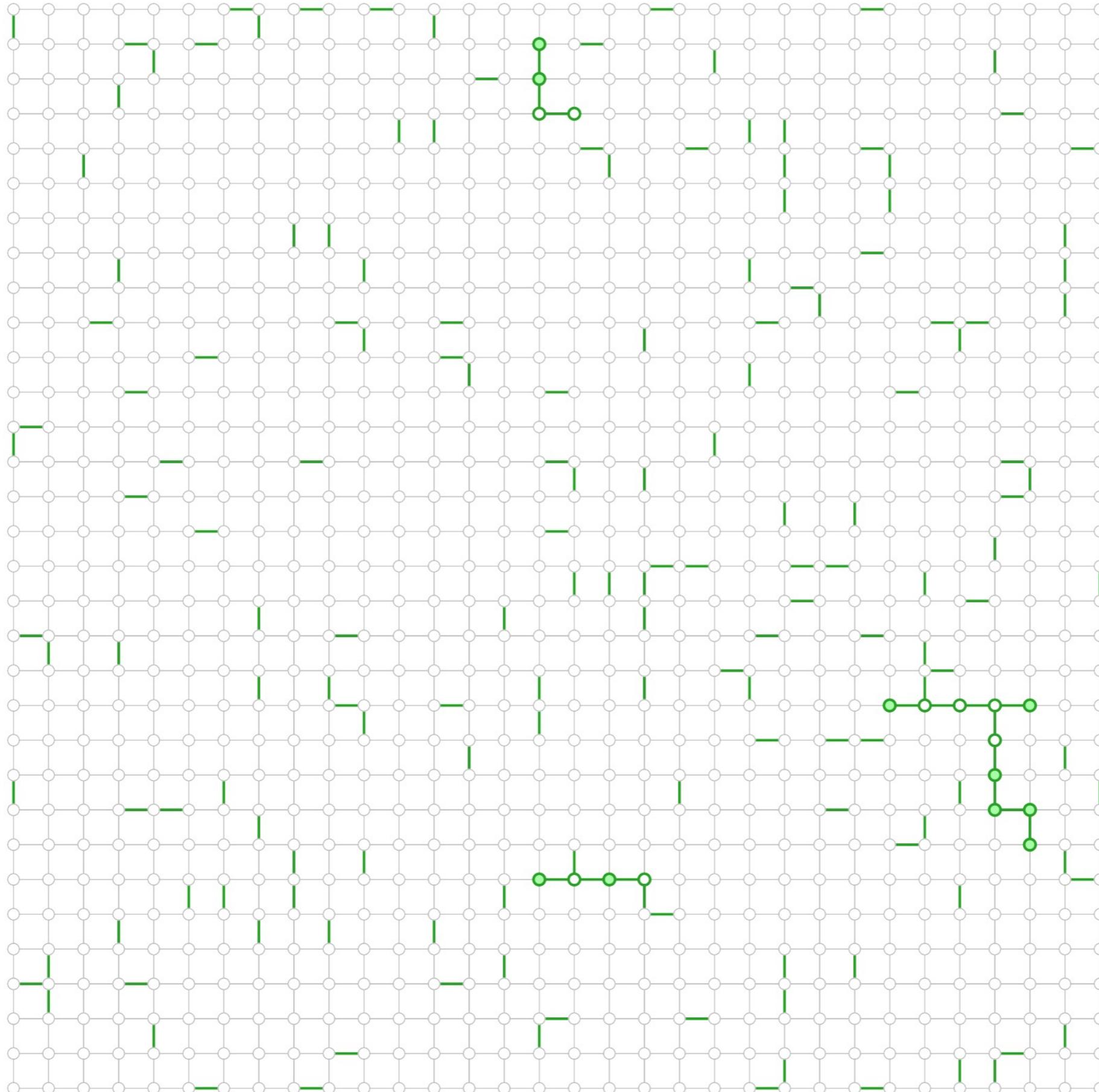
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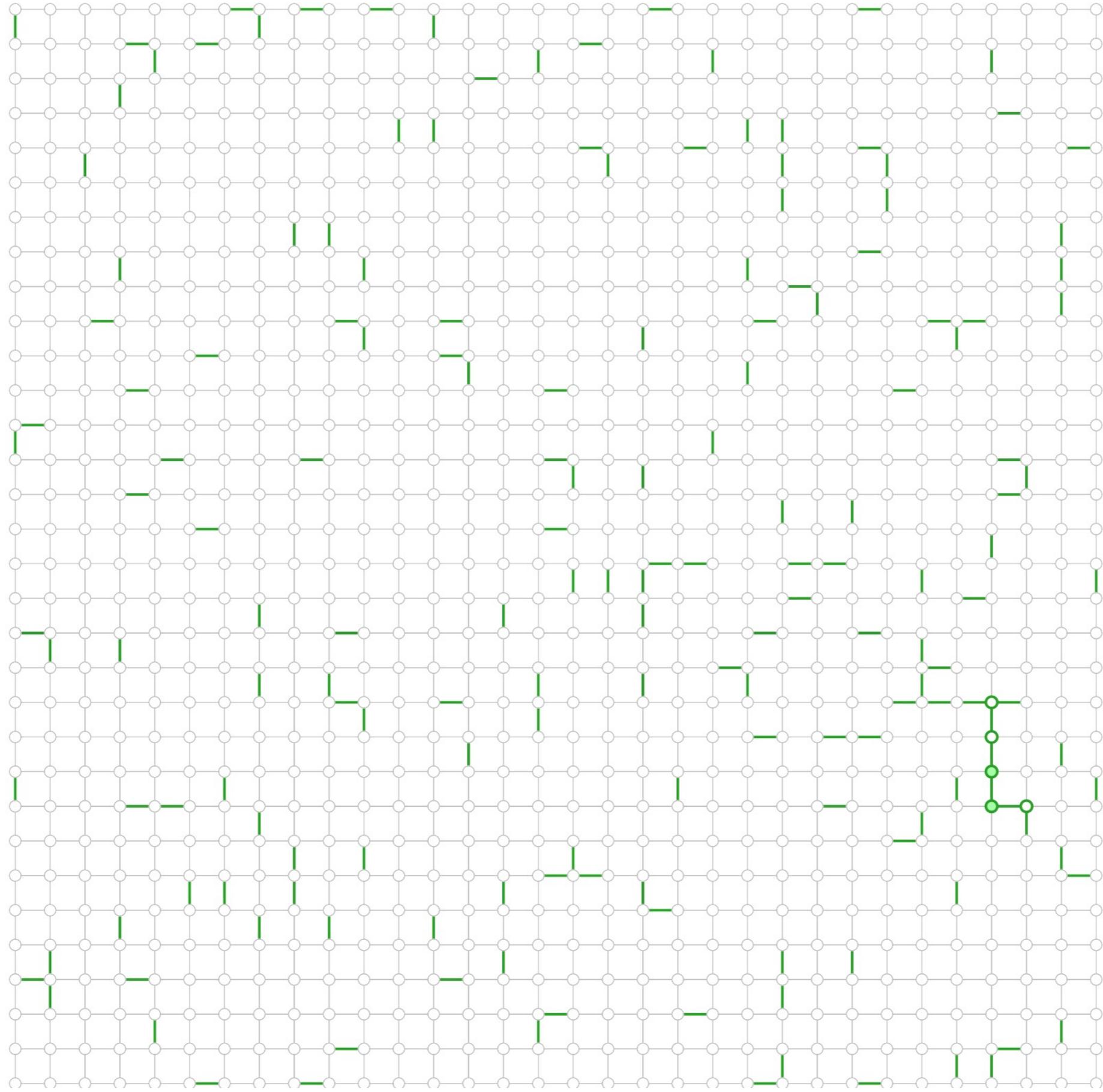
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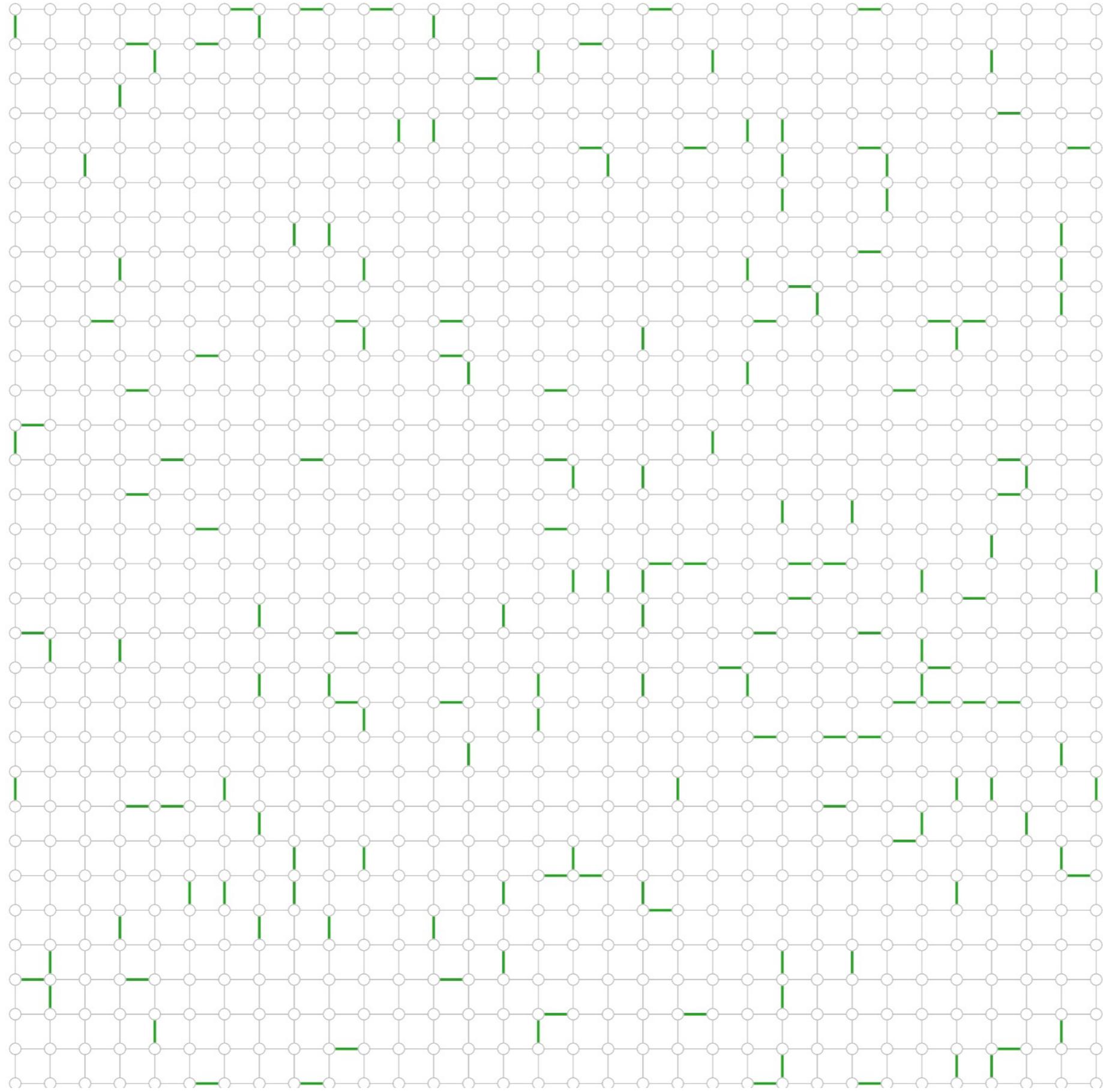
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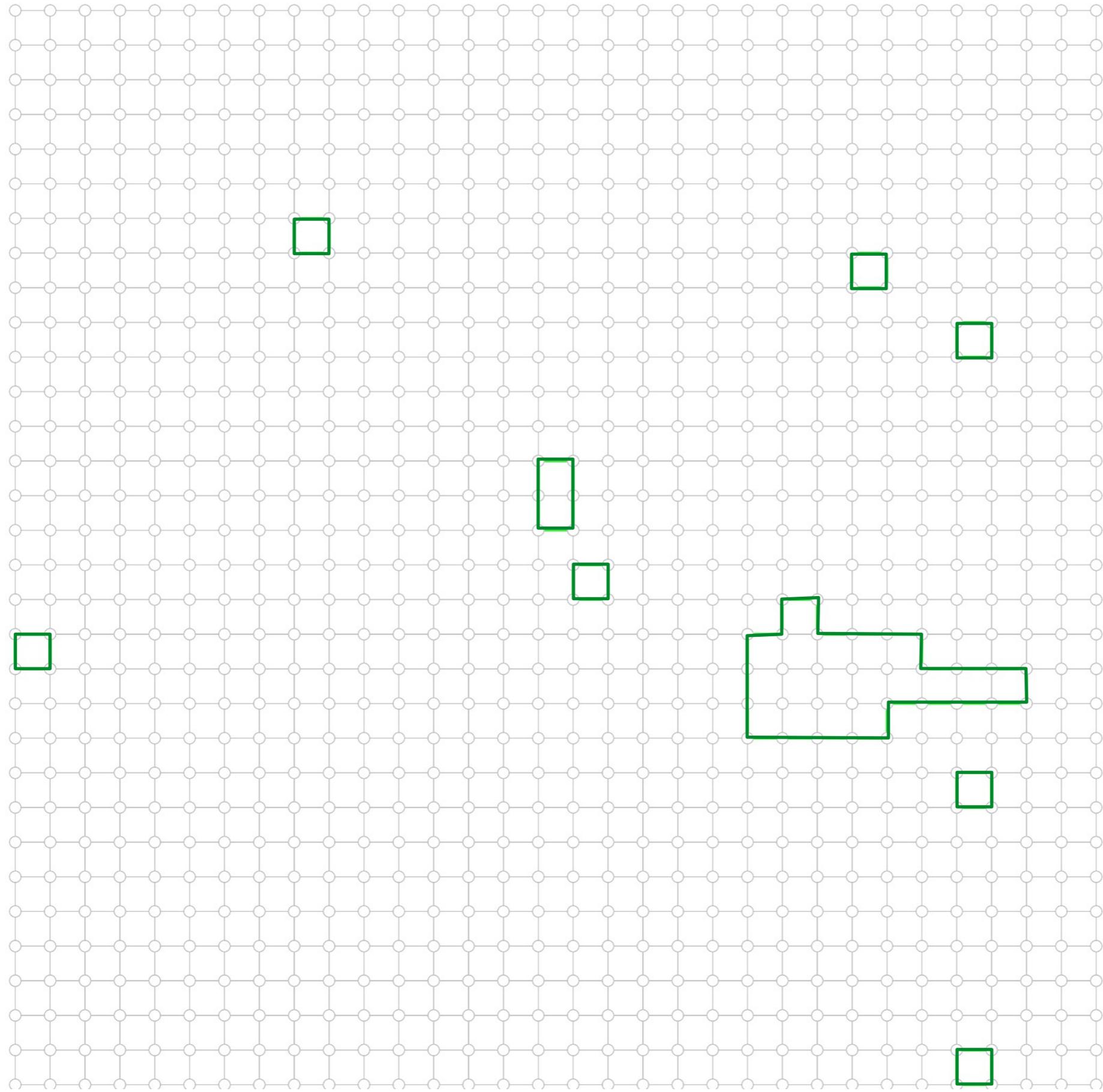
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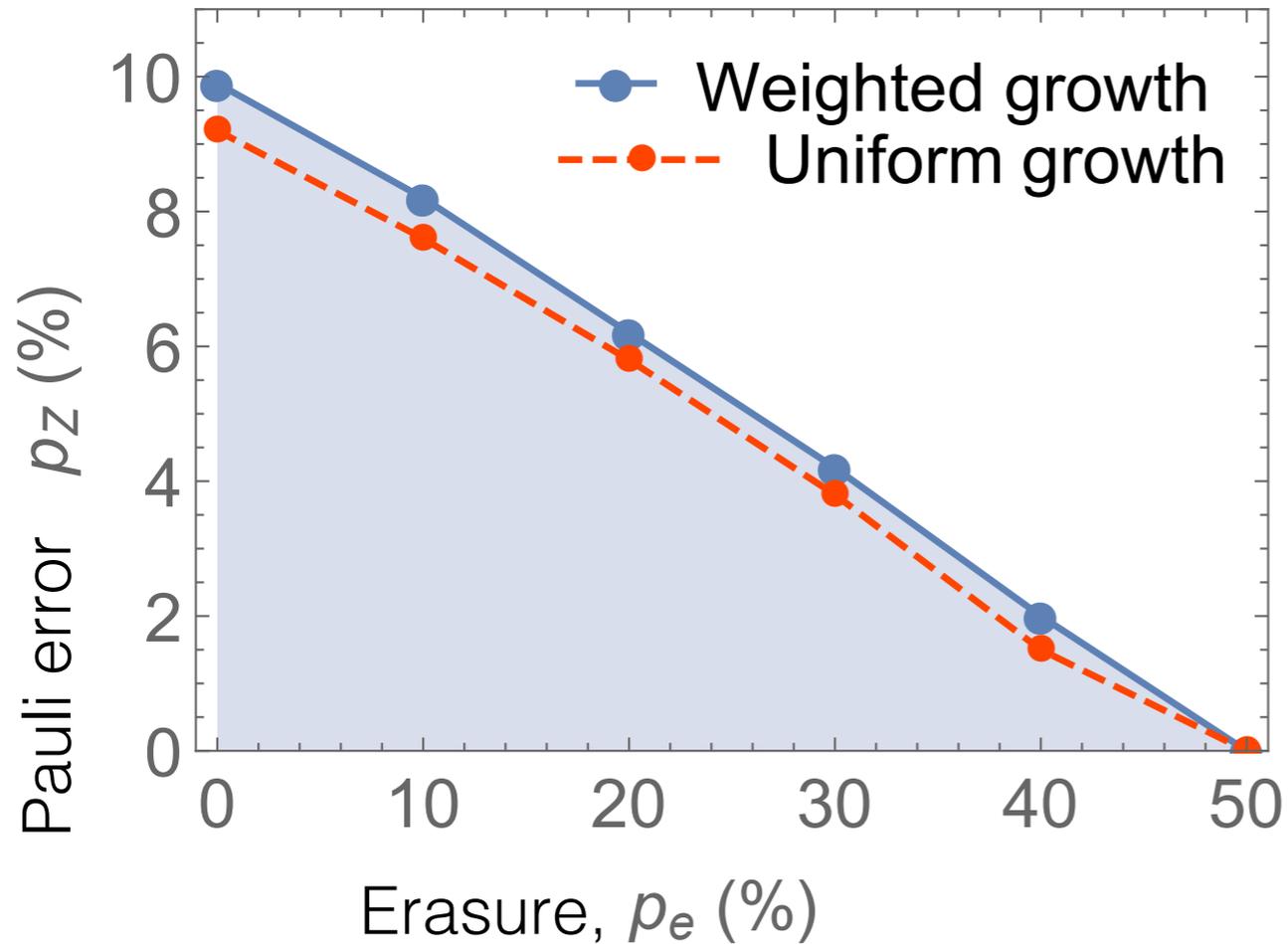
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Resulting error



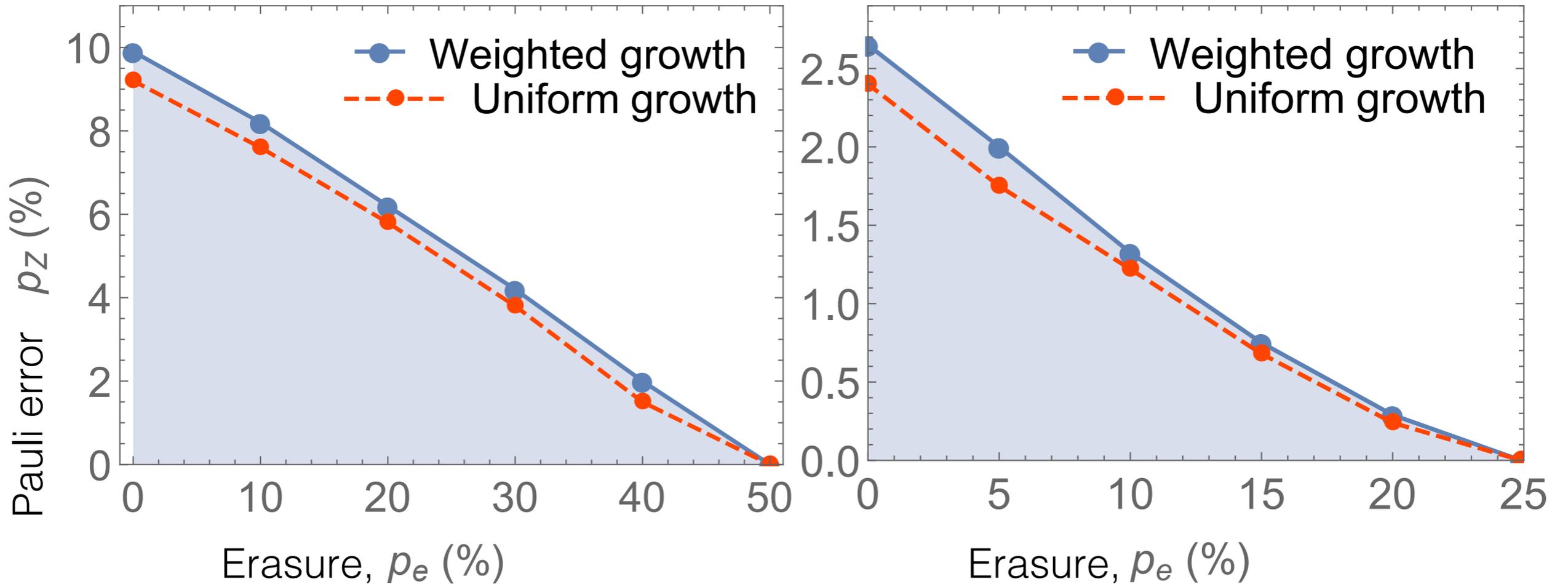
# Performance: Threshold

2D surface code with bit-flip errors



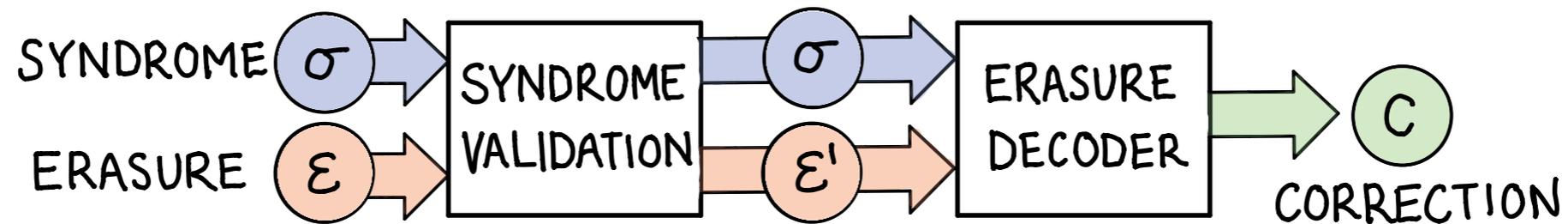
Pauli error threshold:  
9.9%

2+1D surface code, phenomenological noise



Phenomenological error threshold:  
2.6%

# Wider Applicability



- **Fault tolerance:** Exactly the same algorithm will decode the 2+1d surface code, which has a cubic syndrome graph.
- **Arbitrary surface code:** Method works for any structure of syndrome graph, and can be applied directly to any surface code, including those with unusual geometry such as hyperbolic codes.
- **Color code:** By projecting the surface code onto the color code, an arbitrary color code can be decoded.
- **Other codes?:** For any code for which there is an erasure decoder, and a notion of distance between syndromes, this approach can be used to create a decoder for Pauli error.

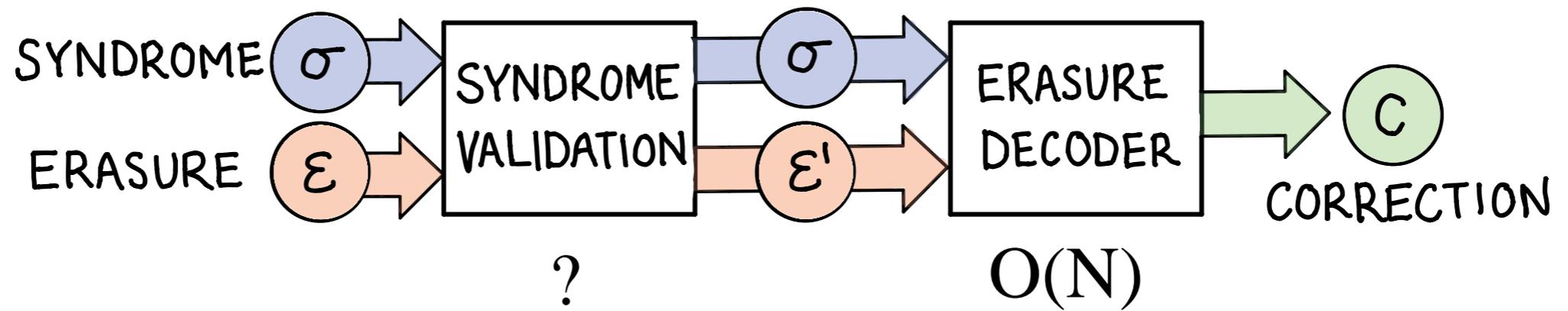
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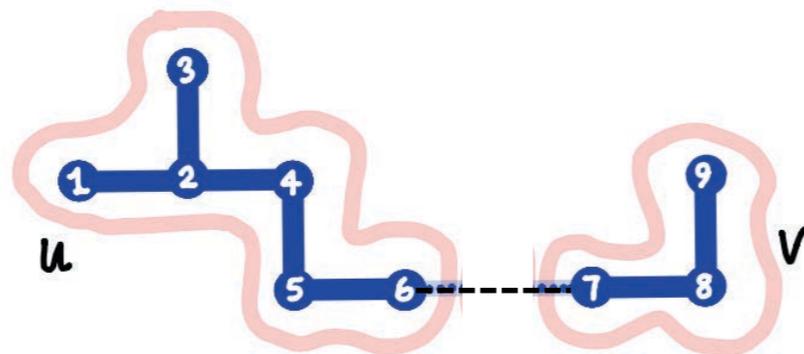
# Achieving almost linear complexity



?

UNION( $u, v$ )  
Merges clusters  $u$  and  $v$

FIND( $n$ )  
Returns the cluster to which  
node  $n$  belongs

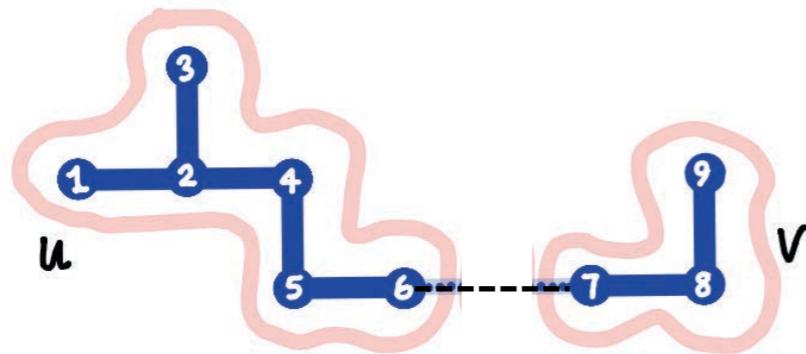


# Achieving almost linear complexity

## Naive algorithm

**Data structure:**

Lookup table for each node



node:	1	2	3	4	5	6		7	8	9
cluster	u	u	u	u	u	u		v	v	v

**FIND:**

Lookup node:  $O(1)$

**UNION:**

Relabel every element of one cluster:  $O(N)$

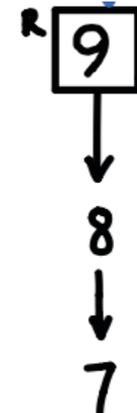
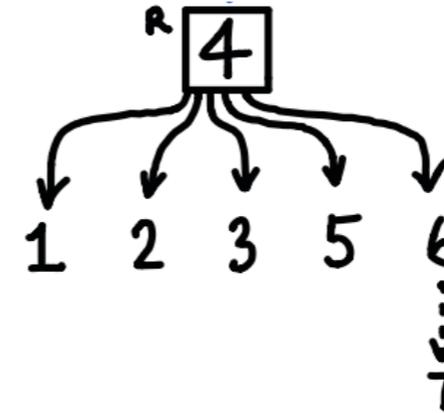
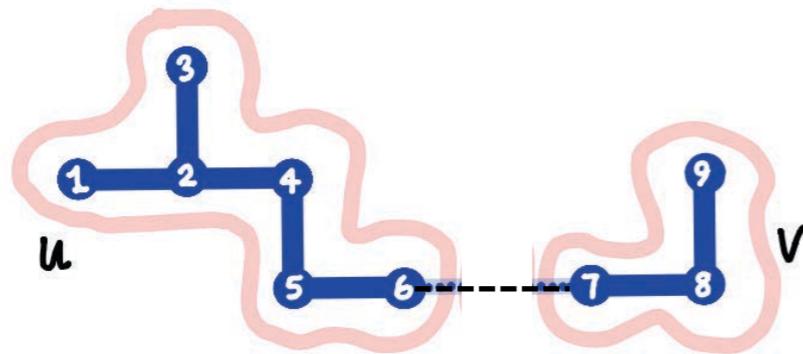
**Worst case complexity:**  $O(N^2)$

# Achieving almost linear complexity

## Better algorithm

### Data structure:

Tree, stored as a linked list  
root of tree identifies the cluster



### FIND:

Traverse tree to find root:  $O(\log N)$

### UNION:

Point root of one cluster to the other  $O(1)$

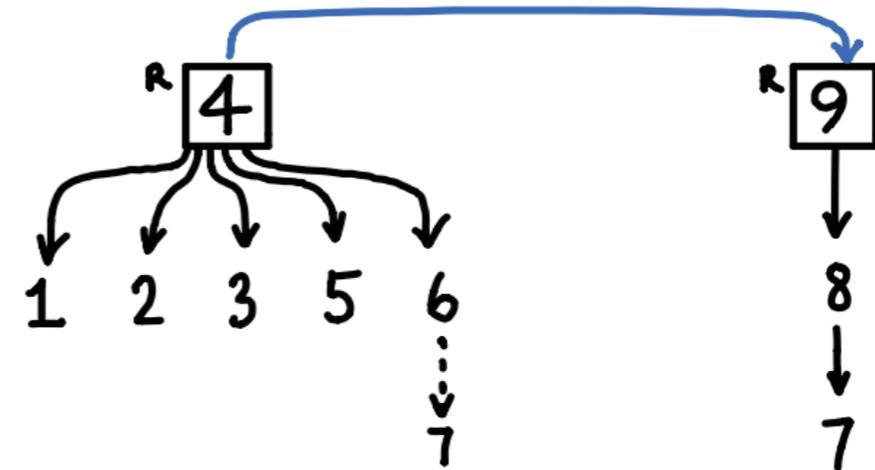
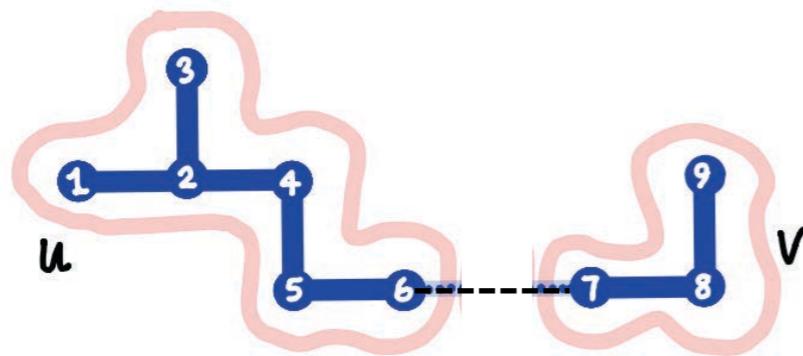
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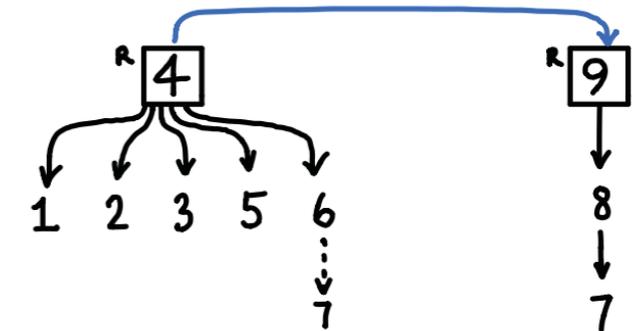
**Worst case complexity:**  $O(N \log N)$

# Achieving almost linear complexity

## Even better algorithm

### Data structure:

Tree, stored as a linked list  
root of tree identifies the cluster



### + Weighted Union

During UNION always updates the smallest of the two clusters.  
Size of smaller cluster at least doubles when UNION is called.

### + Path compression

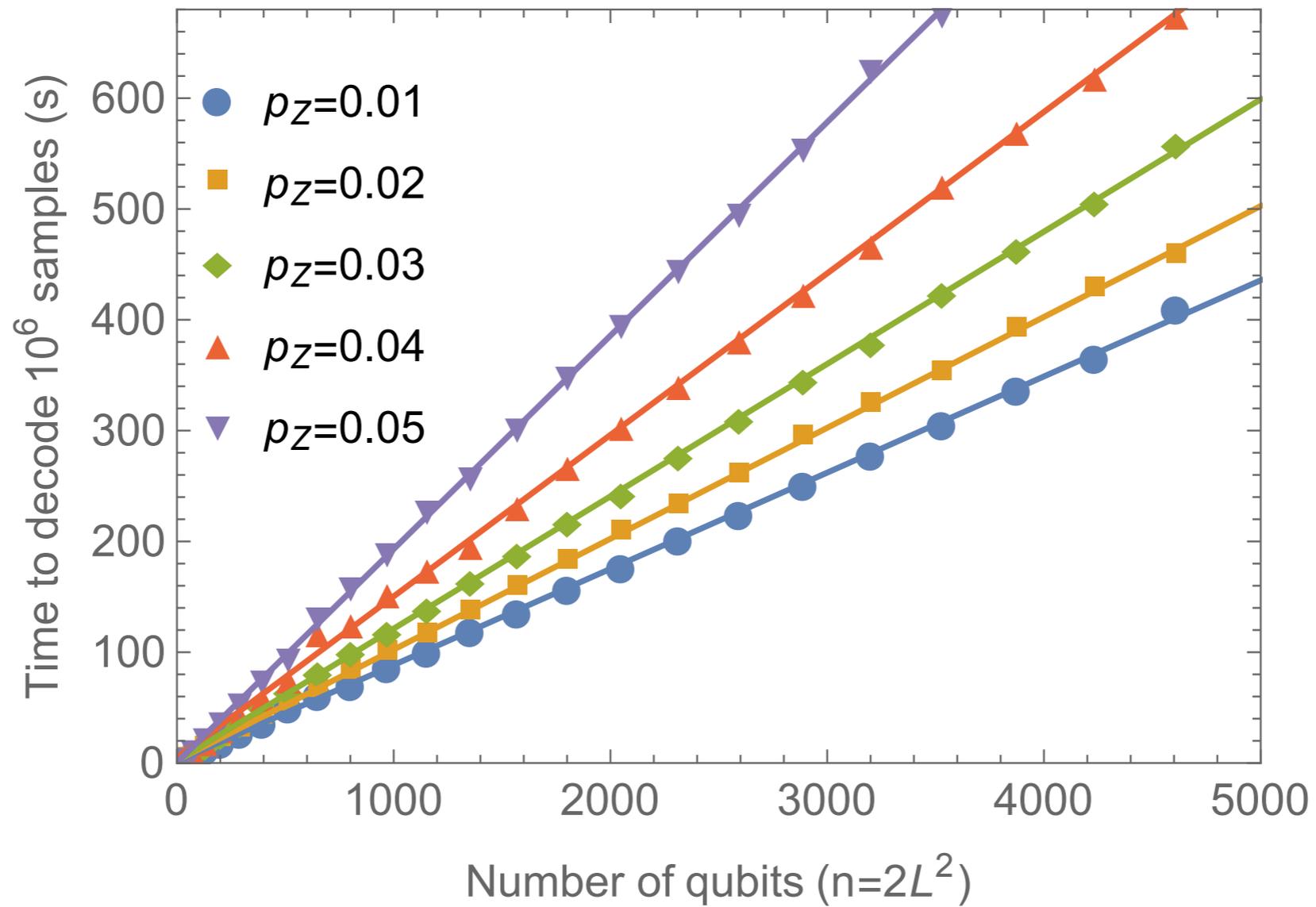
After FIND(u) is called, add a new edge pointing u directly to the root.  
If Find(u) is called again, it will take 1 step to return the root.

The analysis of these three things combined was first made by Tarjan:  
R. E. Tarjan, Journal of the ACM (JACM) 22, 215. (1975).

**Worst case complexity:**  $O(\alpha(N) N)$

$\alpha(N)$  is the inverse of Ackermann's function, and  $\alpha(N) \leq 3$  as long as  $N < 2^{2^{2^2}}$  with 65536 twos.

# Performance: Running time



## Why you should care about the Union Find Decoder if:

### **You want to build a quantum computer:**

- The UF decoder is very fast in practice, with effectively linear scaling and a small constant overhead
- Very simple algorithm, good for implementing in hardware.

### **You want to numerically study codes with unusual geometries:**

- The decoder can be applied to any surface code (2d or 2+1d), without any adaptation, to color codes, and potentially to wider classes of codes.

### **You want to understand the connection between erasure errors and Pauli errors:**

- The decoding algorithm provides one way of converting Pauli errors into erasures. Maybe this can help us better understand how they are related?

## Questions still to answer:

- How fast can it run in hardware?
- Compare directly to other decoders' below threshold performance
- What's the best way to parallelize the algorithm?
- Can the threshold be further improved?
- Can the algorithm be adapted to account for more types of errors?