

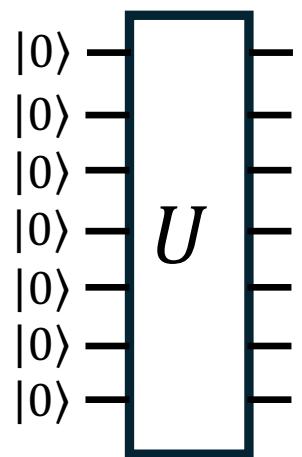
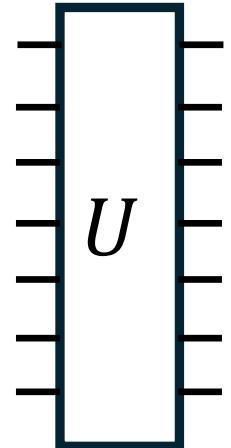
# Learning shallow quantum circuits and quantum states prepared by shallow circuits in polynomial time

Yunchao Liu (UC Berkeley)

Based on arxiv 2401.10095, and upcoming work with Zeph Landau

# Two fundamental problems

- Given access to an unknown constant depth quantum circuit  $U$ , learn a constant depth circuit that is close to  $U$
- Given copies of an unknown quantum state  $|\psi\rangle = U|0^n\rangle$  that is prepared by an unknown constant depth circuit  $U$ , learn a constant depth circuit that prepares  $|\psi\rangle$
- **This talk: polynomial time algorithms for both problems**
  - quasi-polynomial time, when depth of  $U$  is  $\text{polylog}(n)$



$||\psi\rangle$

# Quantum algorithms in NISQ

- NISQ computation can be modeled as shallow quantum circuits
  - Can generate probability distributions that are classically hard
- Key idea behind NISQ algorithms: try to discover a shallow circuit as a solution to an interesting problem (assuming the circuit exists)
  - Can be formulated as a learning problem
- Main challenge: how to develop efficient learning algorithms?
  - This talk: two new learning algorithms that provably work in simple settings
  - Primitives for new NISQ algorithms?

# Key challenge: efficient reconstruction

- Step 1: Learn local observables
  - Easy to do (using e.g. classical shadows)
  - Sufficient information
- **Step 2: efficiently reconstruct a quantum circuit from learned local observables**
  - This is a highly non-trivial problem
  - **Goal of this talk: demonstrate new and simple techniques to do this**

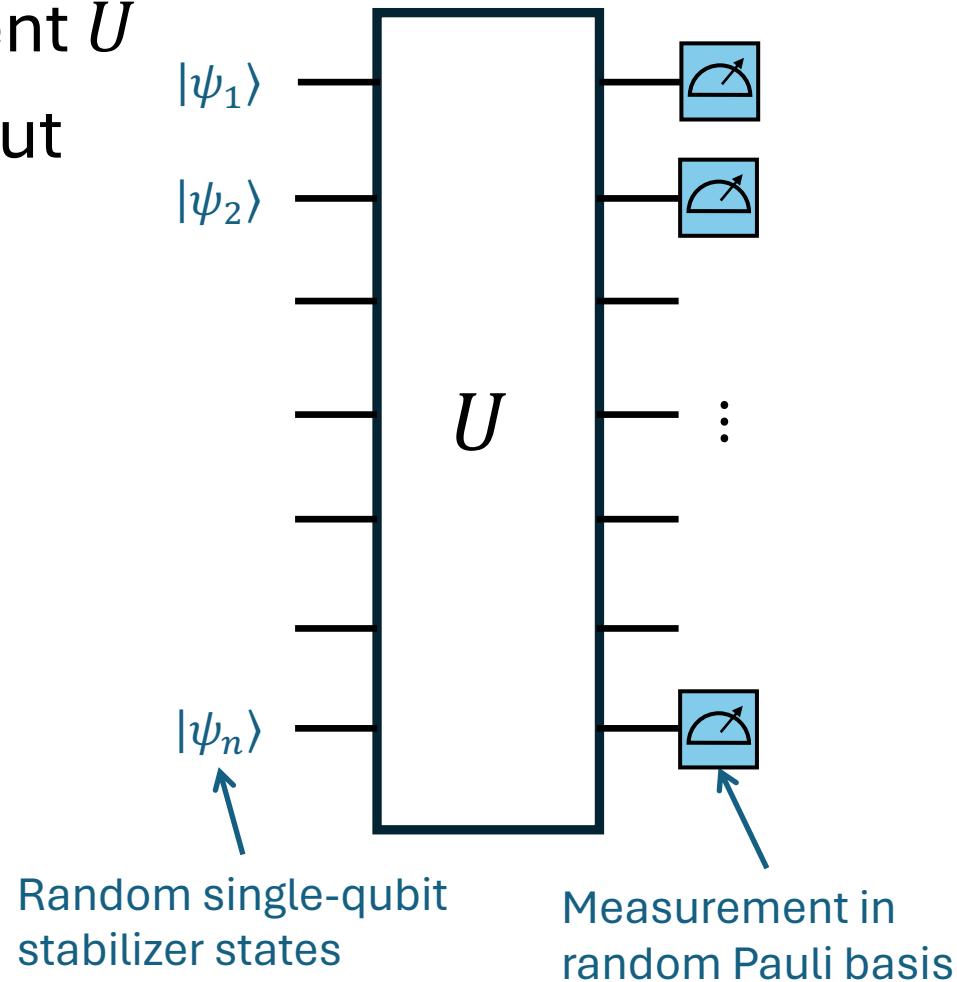
# Learning shallow quantum circuits

Based on arxiv 2401.10095 (QIP 2024, STOC 2024)

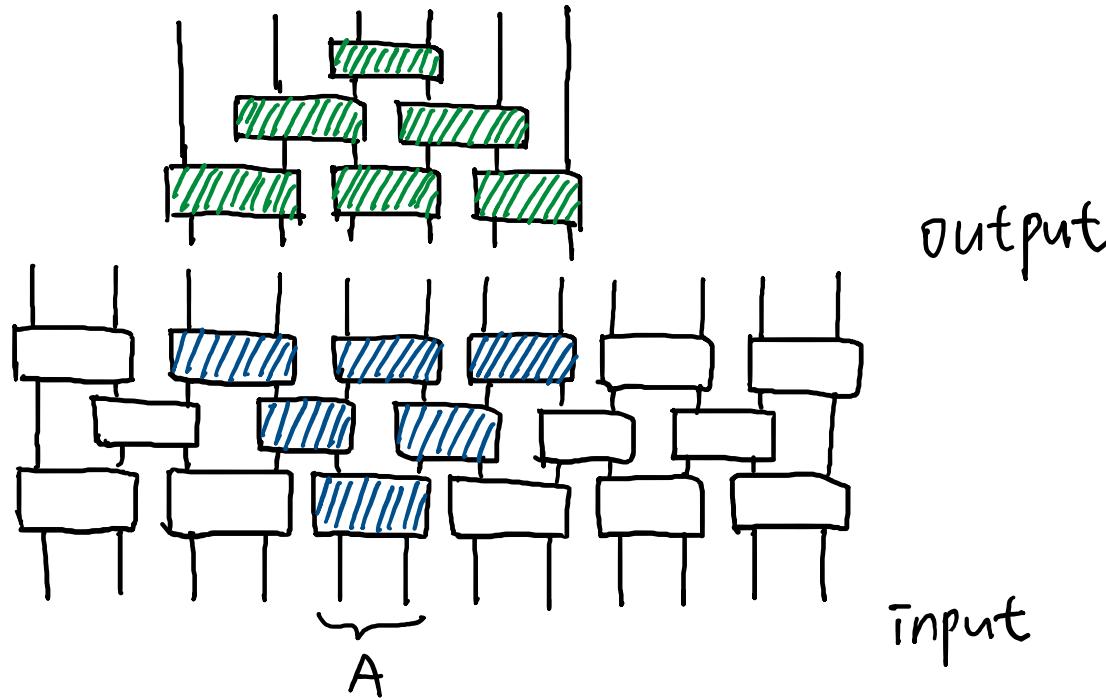
with Hsin-Yuan Huang, Michael Broughton, Isaac Kim, Anurag Anshu, Zeph Landau, Jarrod R. McClean

# Learning shallow quantum circuits

- Want to learn a shallow circuit to implement  $U$
- Only need single-qubit random input/output samples
- **Theorem.** Polynomial time algorithm for learning shallow quantum circuits from random input/output samples



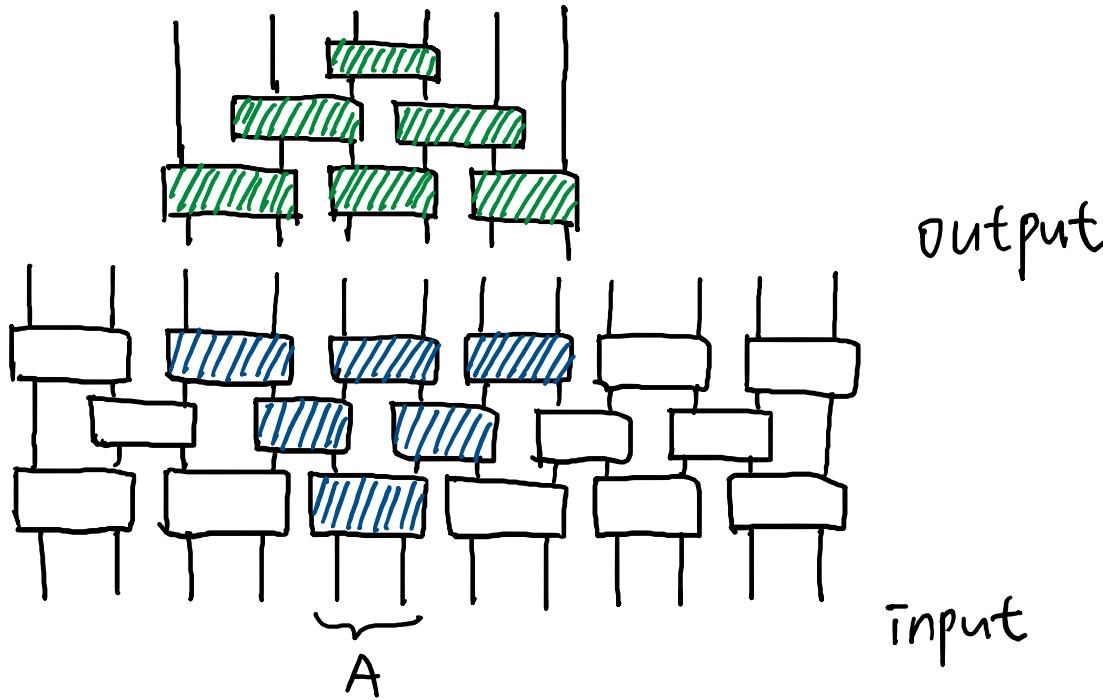
# Lightcone in shallow circuits



Basic idea: guess a **small circuit** to disentangle the region  $A$

- Consider a small region  $A$  (on a lattice), each input qubit in  $A$  only affects the output qubits in the **lightcone** of  $A$
- If we can **undo** the blue gates  $\rightarrow U$  acts as identity on  $A$

# Basic idea: local inversion

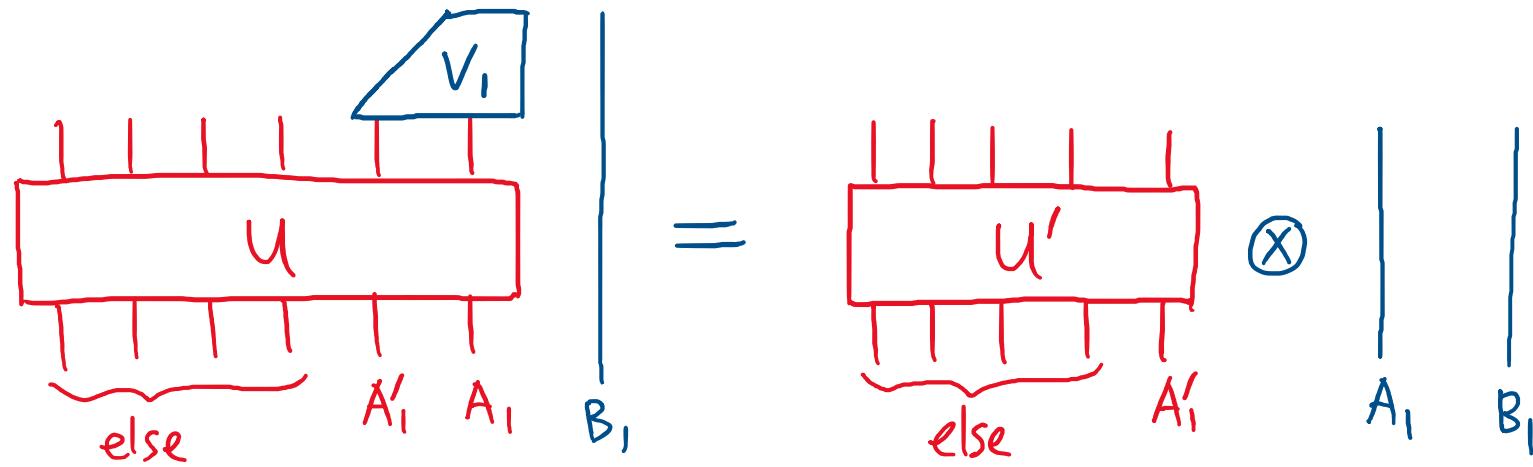


Local inversion: guess a **small circuit** to disentangle the region A

How? Enumerate **small circuits** and test if A is disentangled

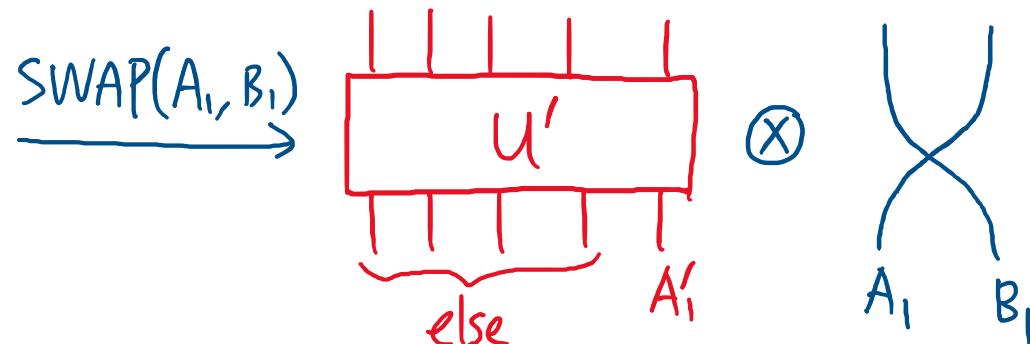
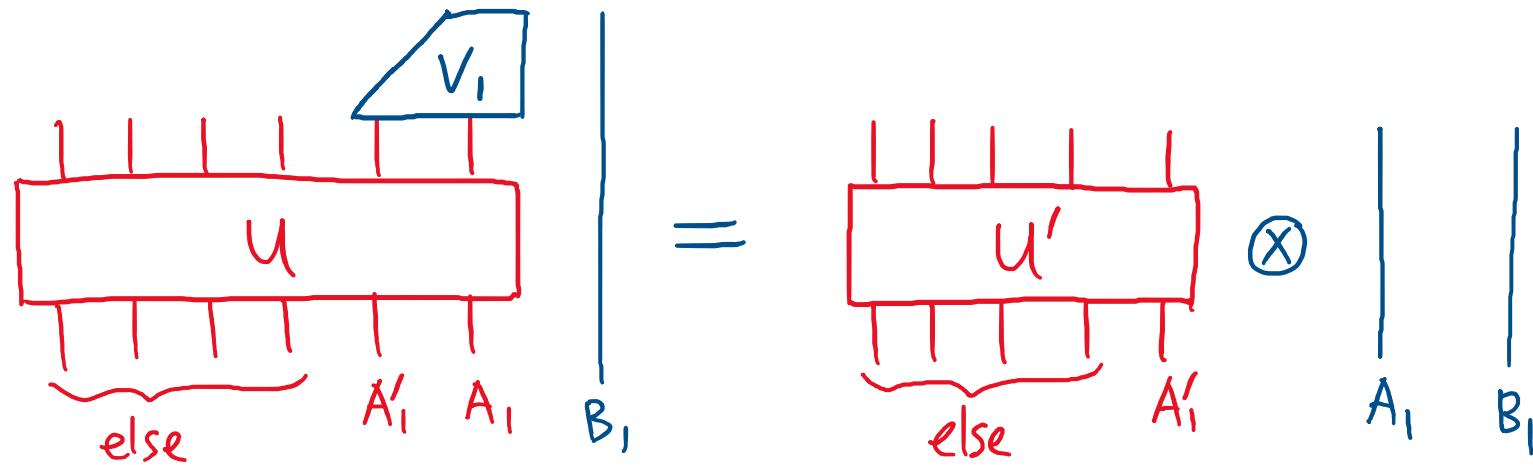
**Key issue: the small circuit we apply could be different from the actual lightcone; in this case it creates a mess on remaining qubits!**

# Key idea: disentangle, swap, *undo*



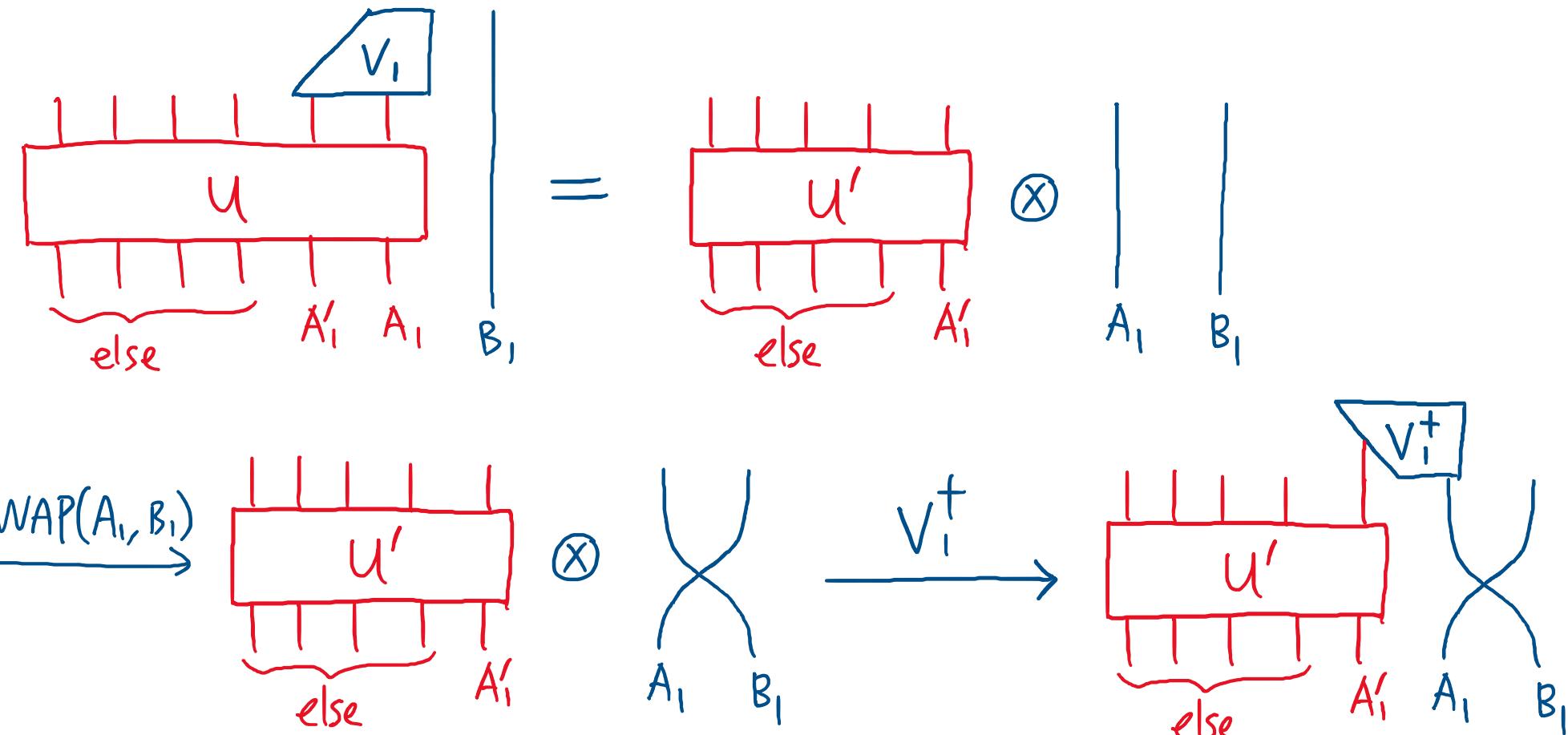
Key idea: introduce a new ancilla qubit, swap with  $A_1$ , then *undo*  $V_1$

# Key idea: disentangle, swap, *undo*



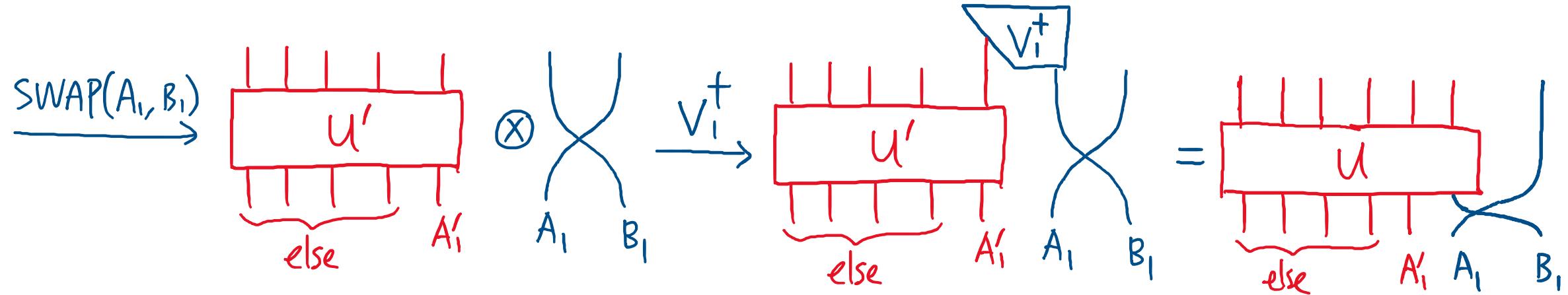
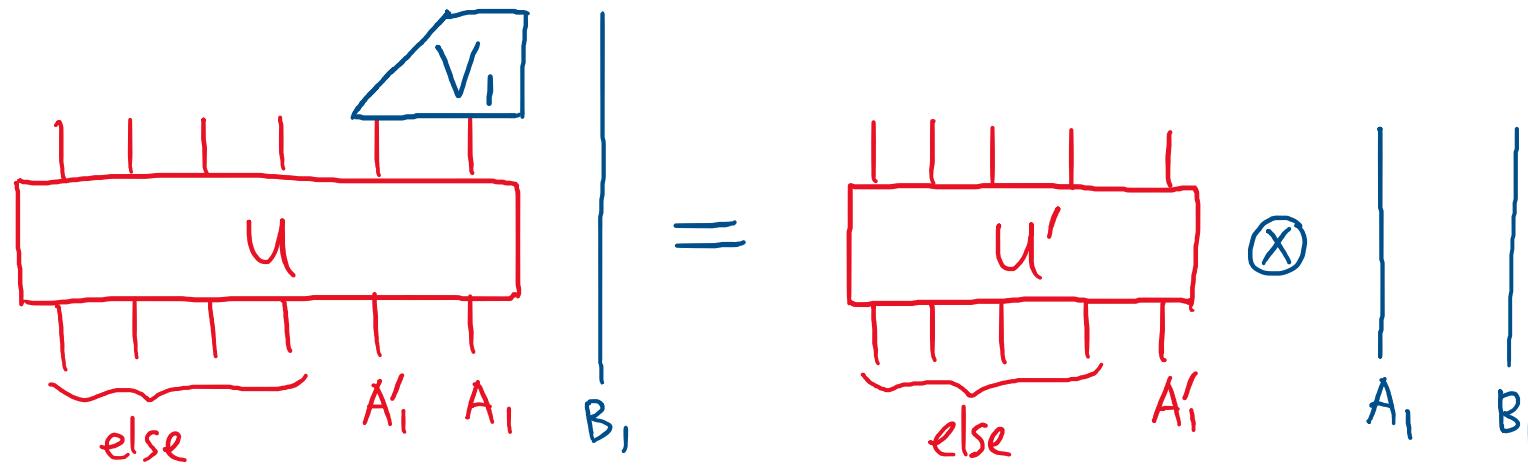
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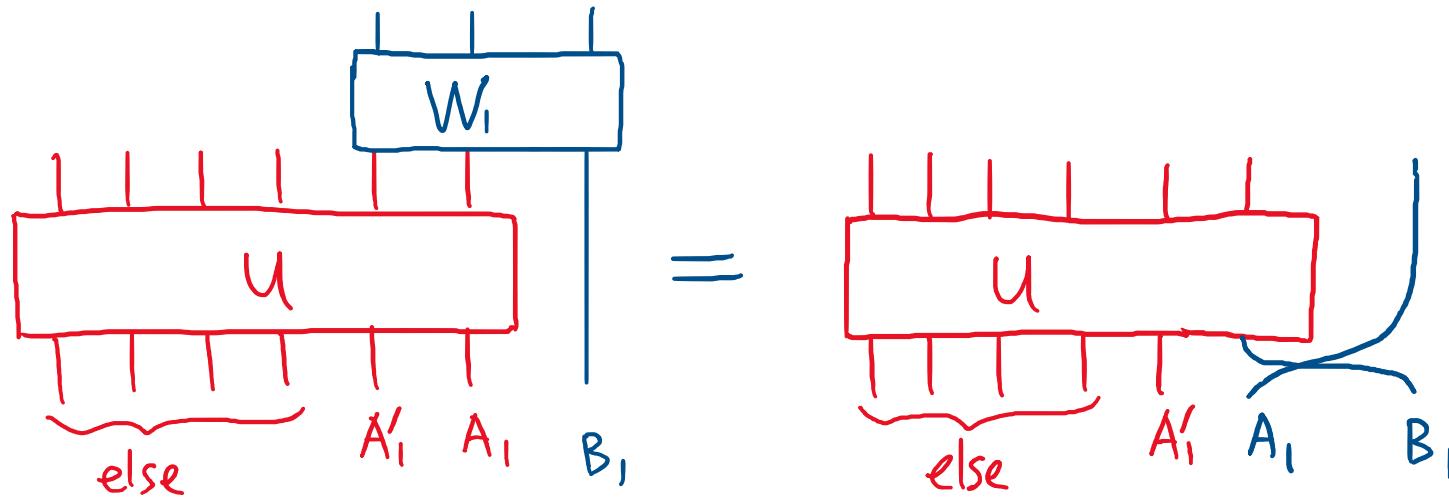
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Key idea: introduce a new ancilla qubit, swap with  $A_1$ , then **undo**  $V_1$

# Key idea: disentangle, swap, undo

- We have learned a unitary  $W_1$  such that

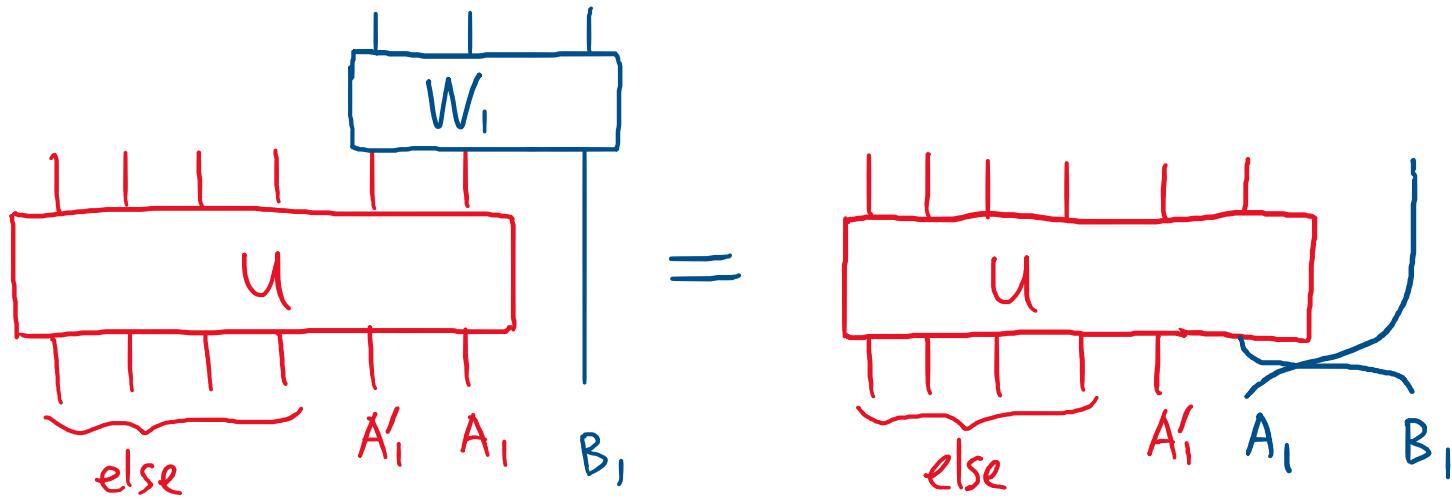


We have learned a circuit acting on top of  $U$  that achieves the effect of swapping an input wire

**Key observation:** the system is not disturbed; can repeat this for every qubit

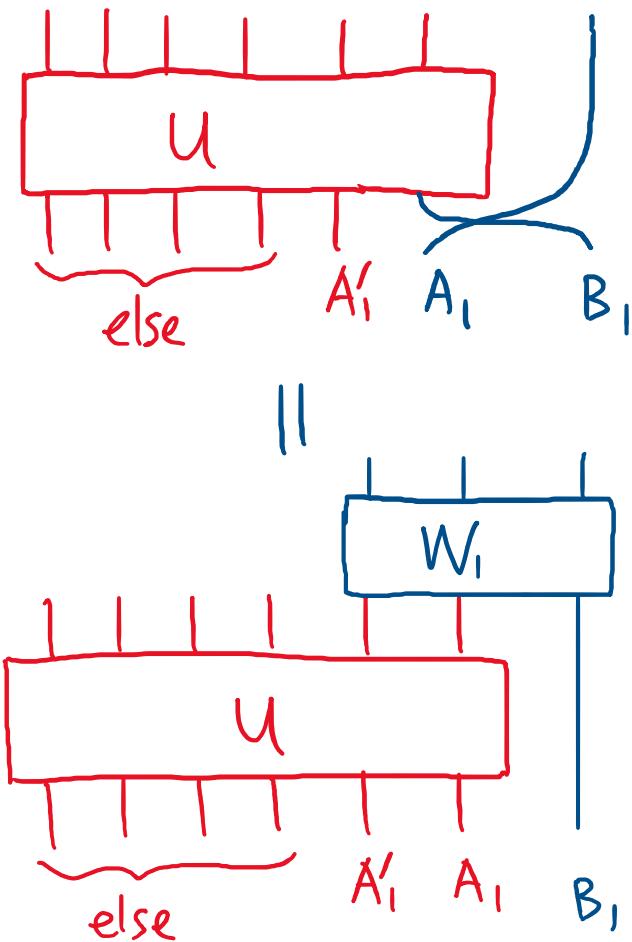
# Reconstructing the circuit

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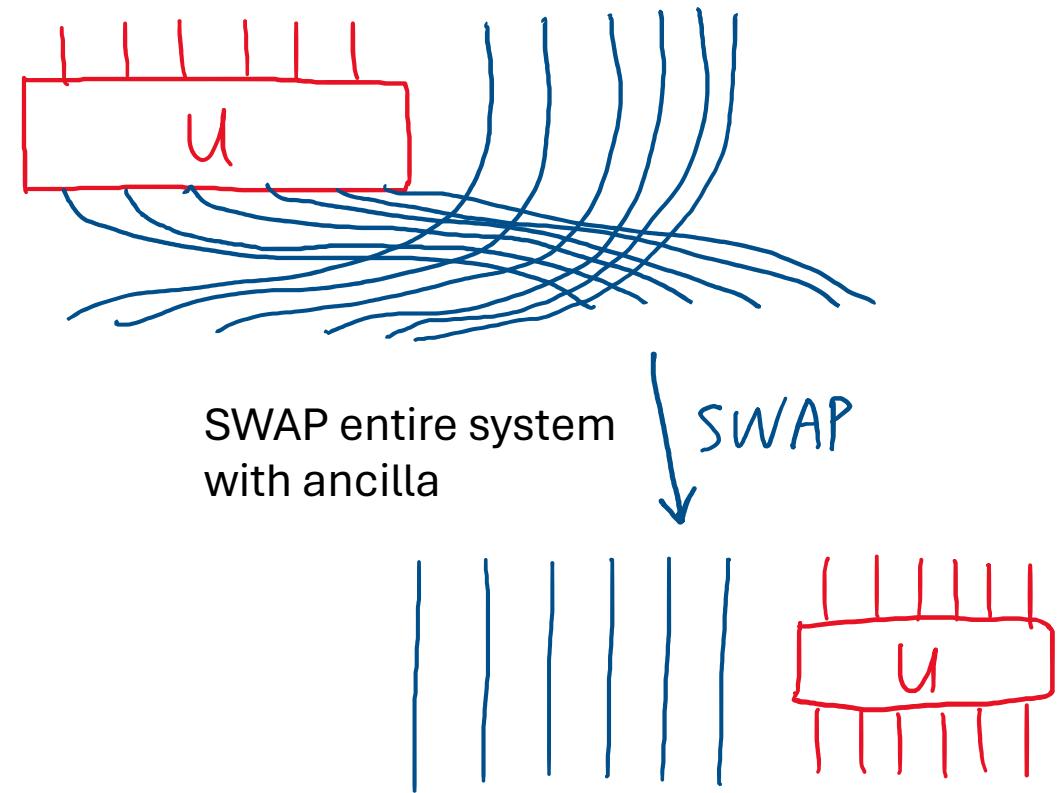


Claim: using this idea, can learn the description of a  $2n$ -qubit circuit  $W$  that satisfies  $W = U^\dagger \otimes U$

# Reconstructing the circuit

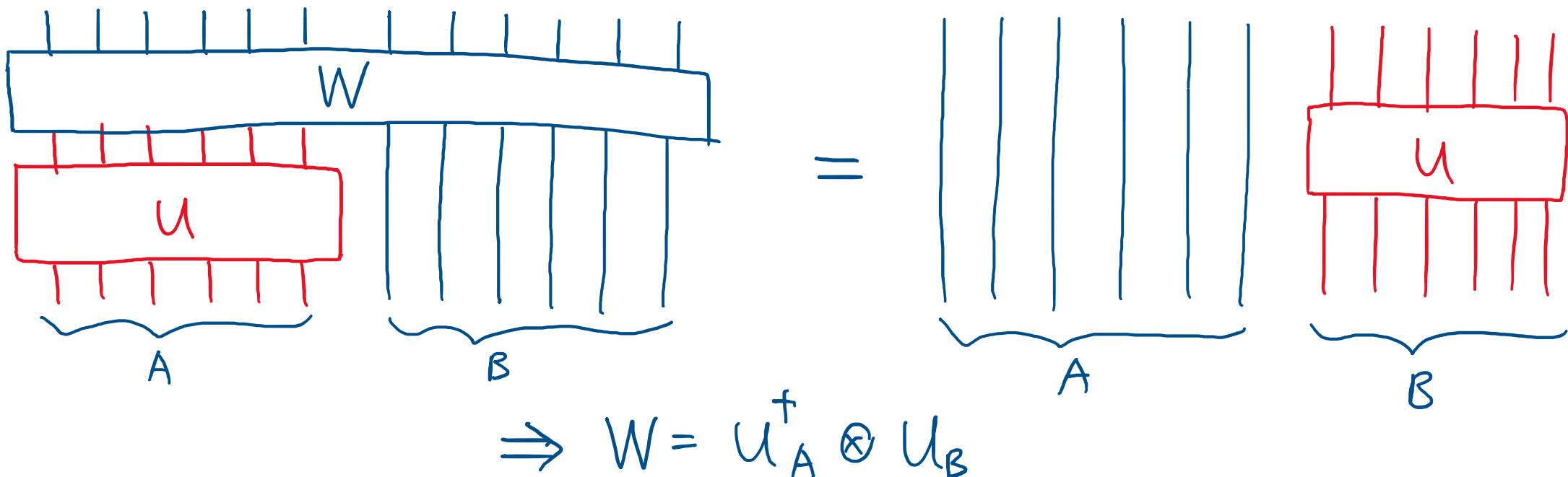


$W_2 W_3 \dots W_n$  →  
Do the same  
for every qubit



# Reconstructing the circuit

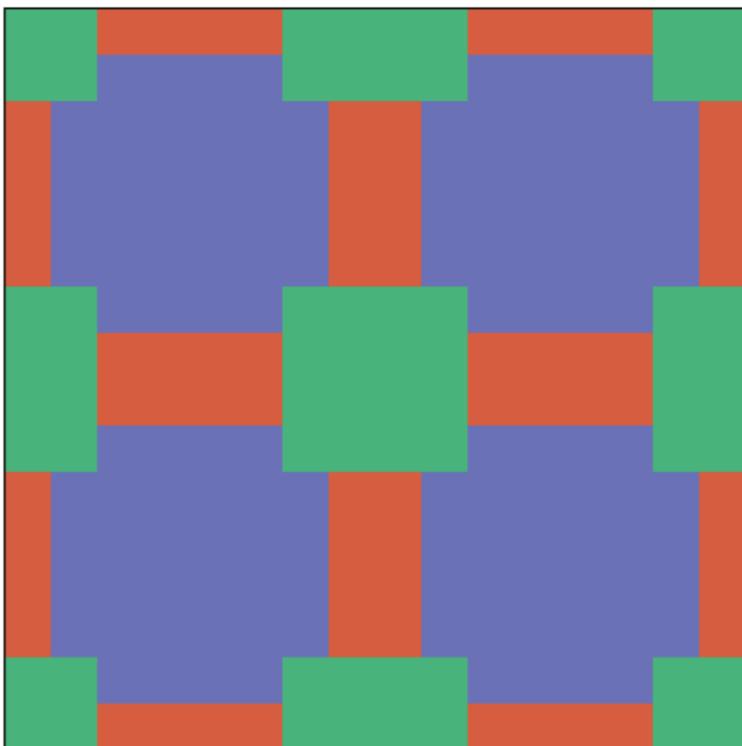
- We have learned a  $2n$ -qubit circuit  $W$  that satisfies



disentangle a local region without disturbing the system

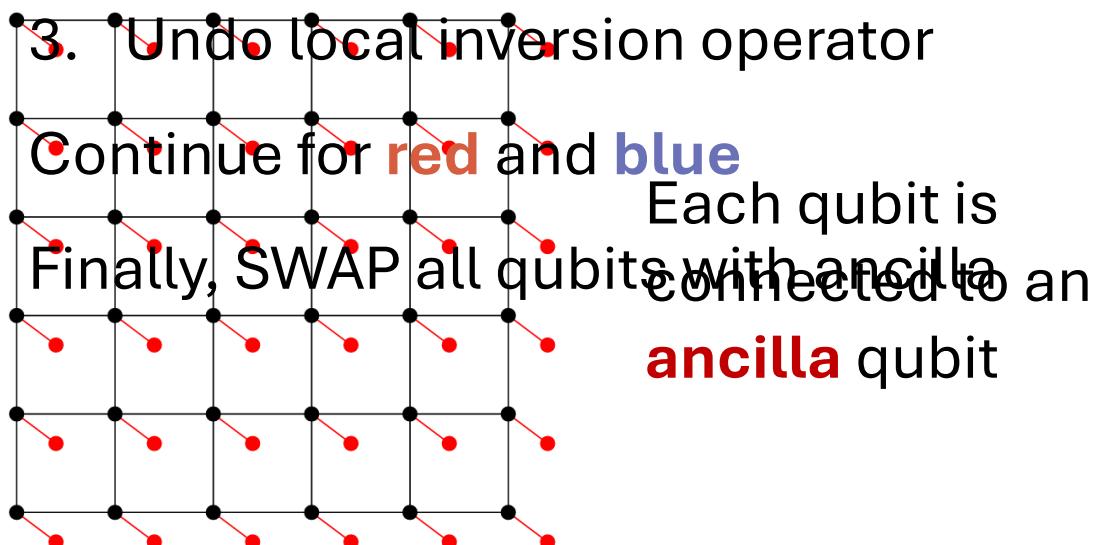
# Reconstructing the circuit in low depth

- 2D example: reconstruct the circuit in 3 layers



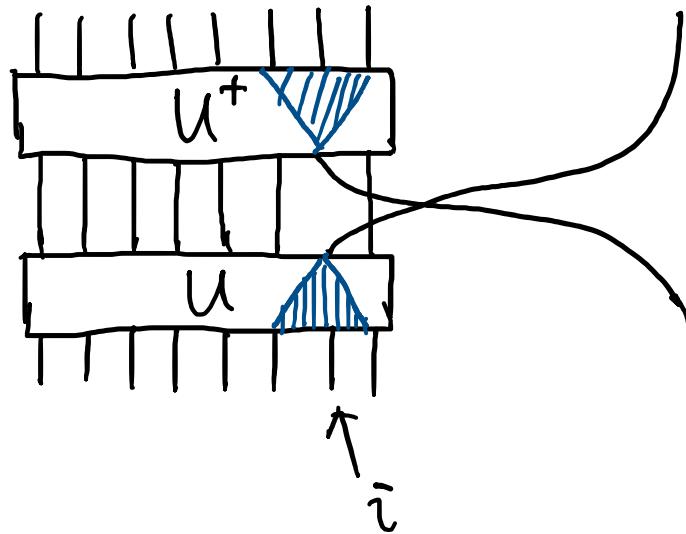
Do this in parallel for all **green** regions:

1. Apply local inversion operator
2. SWAP the entire region with ancilla wires



# Further simplifying the argument

- Key observation: steps 1-3 can be merged into one step, which is equivalent to implementing  $U^\dagger S_i U$  ( $S_i$ : SWAP  $i$ -th qubit and  $i$ -th ancilla)
  - Sanity check:  $U^\dagger S_i U$  is a local operator



# Further simplifying the argument

- Key observation: steps 1-3 can be merged into one step, which is equivalent to implementing  $U^\dagger S_i U$  ( $S_i$ : SWAP  $i$ -th qubit and  $i$ -th ancilla)
- Simpler algorithm: can directly learn these local operators  $U^\dagger S_i U$  and then combine them into a circuit that implements  $U \otimes U^\dagger$

One line proof: an identity for any unitary  $U$

$$U \otimes U^\dagger = \left( \prod_{i=1}^n S_i \right) \cdot \prod_{i=1}^n (U^\dagger S_i U)$$

- Extension: can learn **any unitary** that maps a local operator to a local operator (quantum cellular automata)

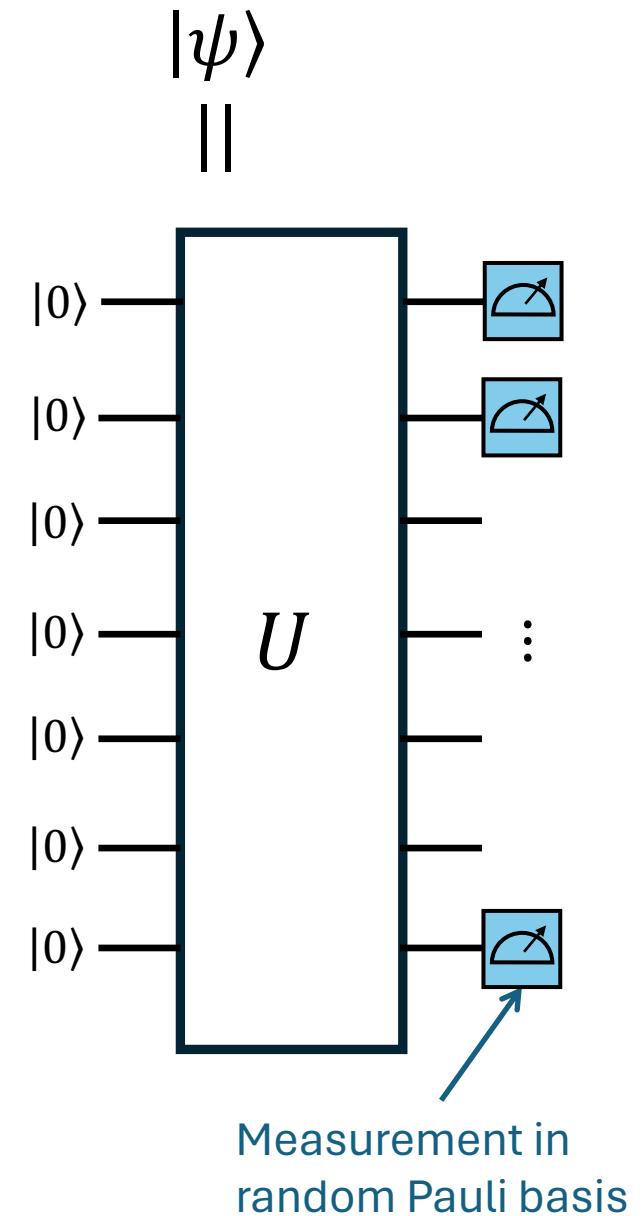
# Learning quantum states prepared by shallow circuits

Based on upcoming work with Zeph Landau

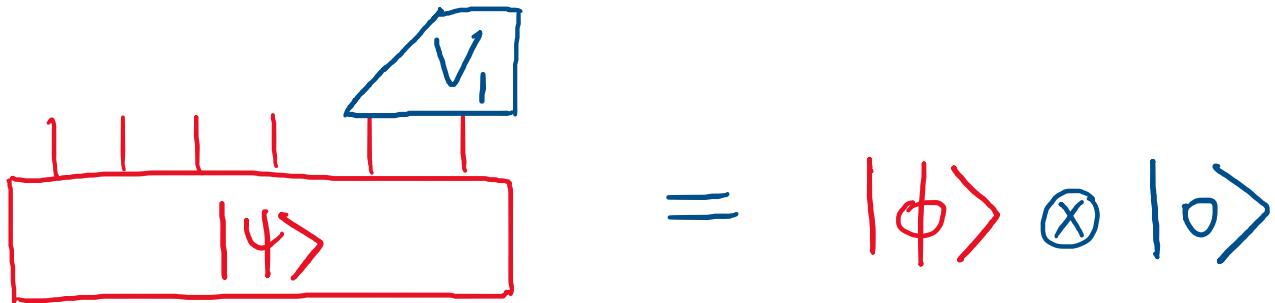
See related upcoming work of Hyun-Soo Kim, Isaac Kim, and Daniel Ranard for a different approach

# Learning quantum states

- Want to learn a shallow circuit to prepare  $|\psi\rangle$
- Only need single-qubit random measurement samples
- **Theorem.** Polynomial time algorithm for learning quantum states prepared by shallow circuits from random measurement samples



# Basic idea: local inversion



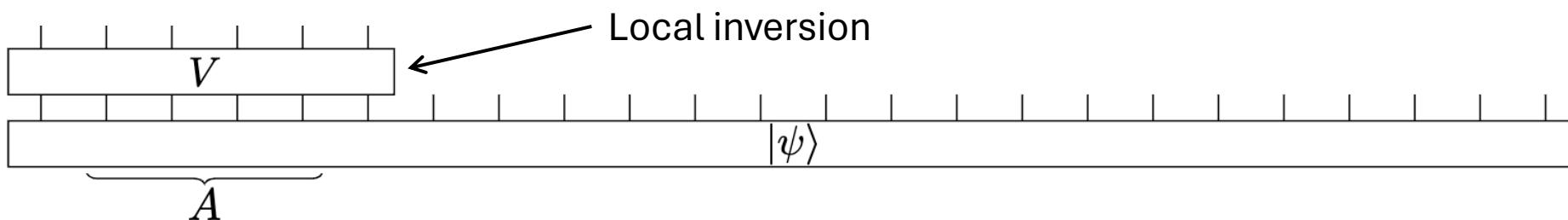
Local inversion: find a small circuit to invert a small region to  $|0\rangle$

What to do next?

Similar issues:  $|\phi\rangle$  could be a much more complicated state

# Key idea 1: local inversion, *undo*

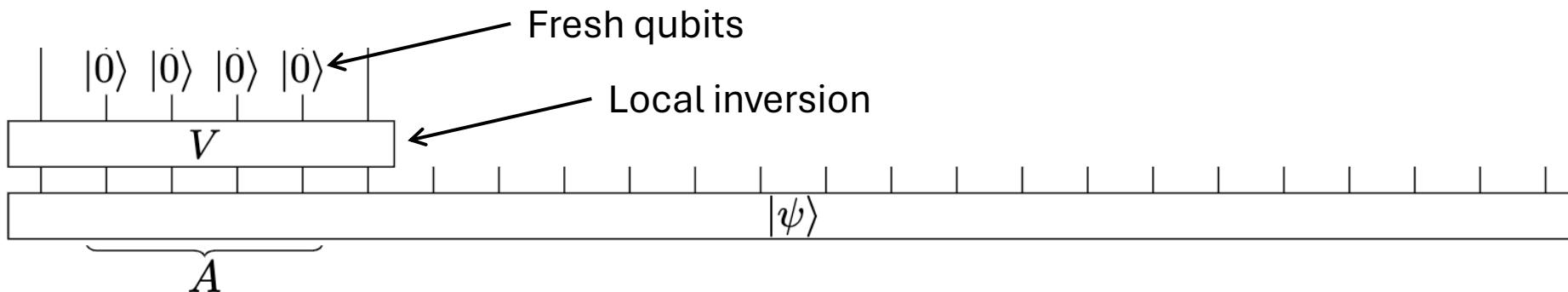
Step 1: Apply a local inversion  $V$  to invert the region  $A$ , get  $|\phi\rangle\otimes|0\rangle_A$



# Key idea 1: local inversion, *undo*

Step 1: Apply a local inversion  $V$  to invert the region  $A$ , get  $|\phi\rangle \otimes |0\rangle_A$

Step 2 (not doing anything): replace  $A$  with fresh qubits in state  $|0\rangle_A$

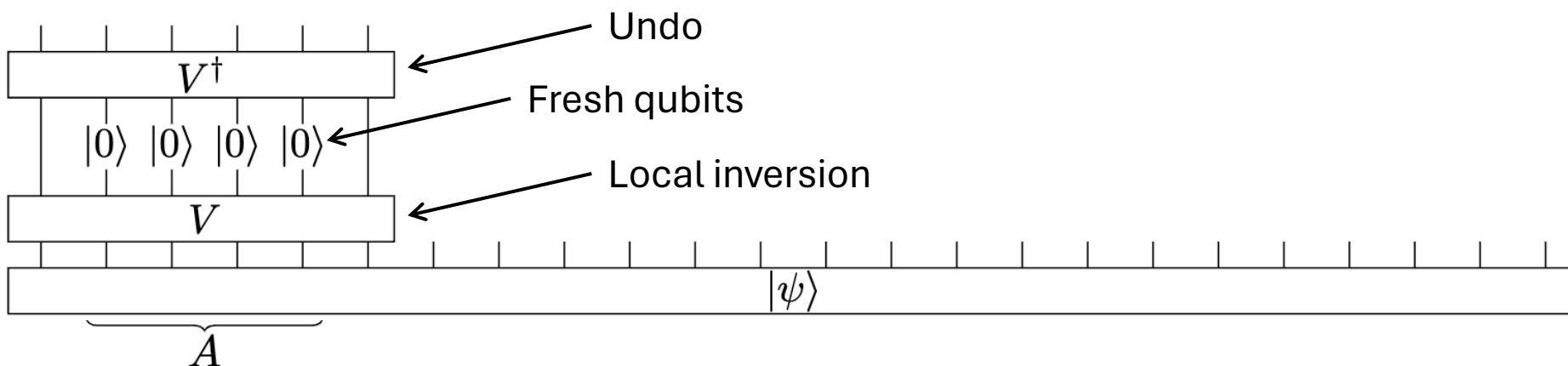


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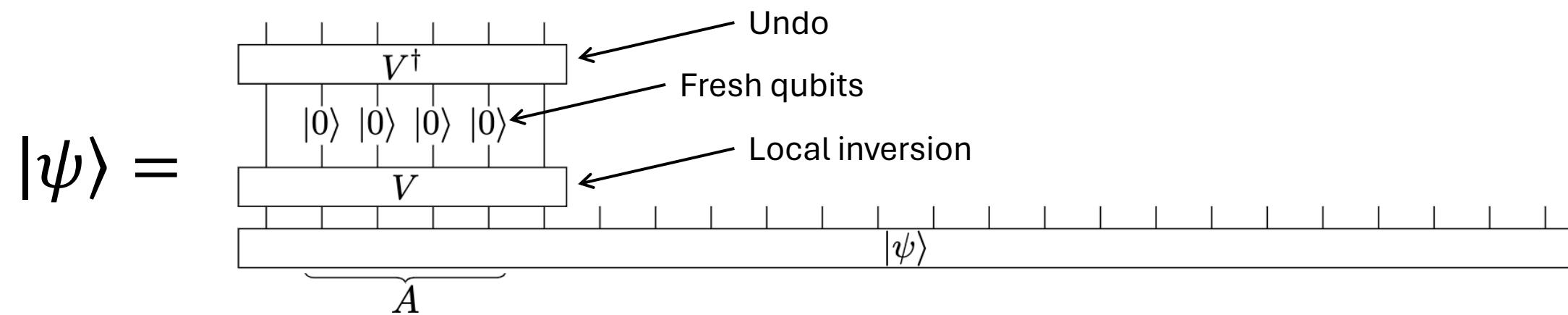
Step 2 (not doing anything): replace  $A$  with fresh qubits in state  $|0\rangle_A$

Step 3: Undo the local inversion, by applying  $V^\dagger$



# Key idea 1: local inversion, *undo*

**Observation: the state did not change**

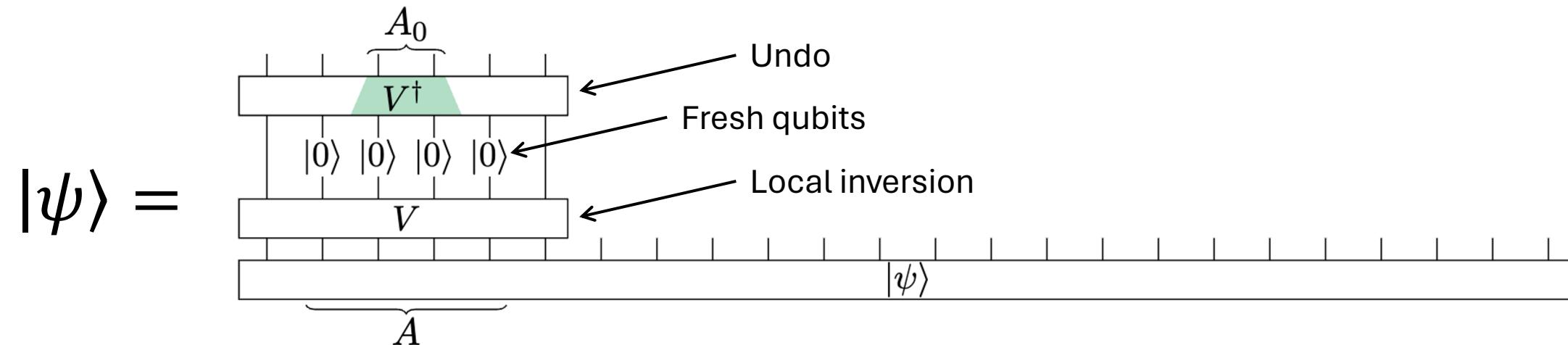


# Key idea 1: local inversion, *undo*

**Observation: the state did not change**

**Observation 2: we have learned a small part of the state**

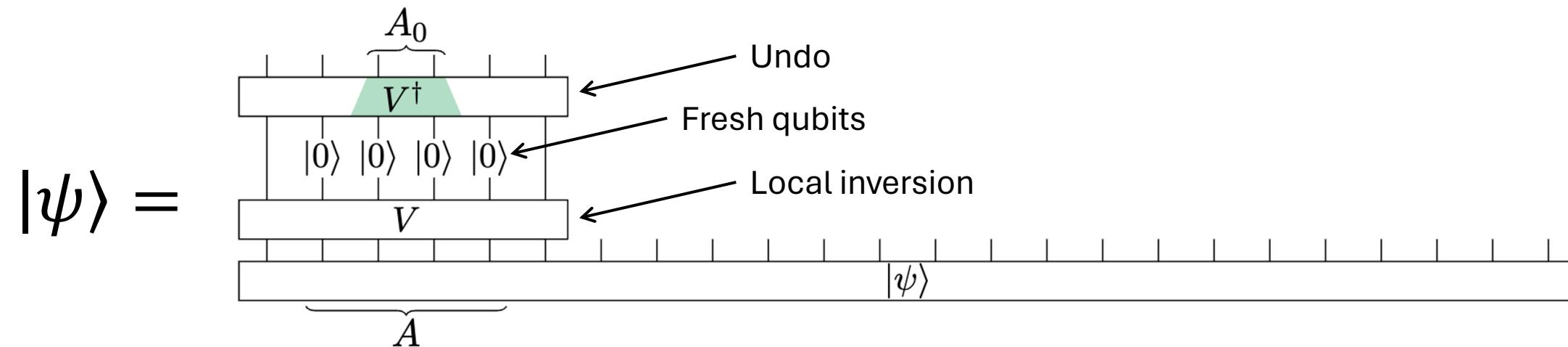
- We have learned a circuit for the reduced density matrix on  $A_0$
- Circuit = **backward lightcone**



# Key idea 1: local inversion, *undo*

**Observation: we have reconstructed a small part of the state, without changing the state at all**

**So, why not repeat?**

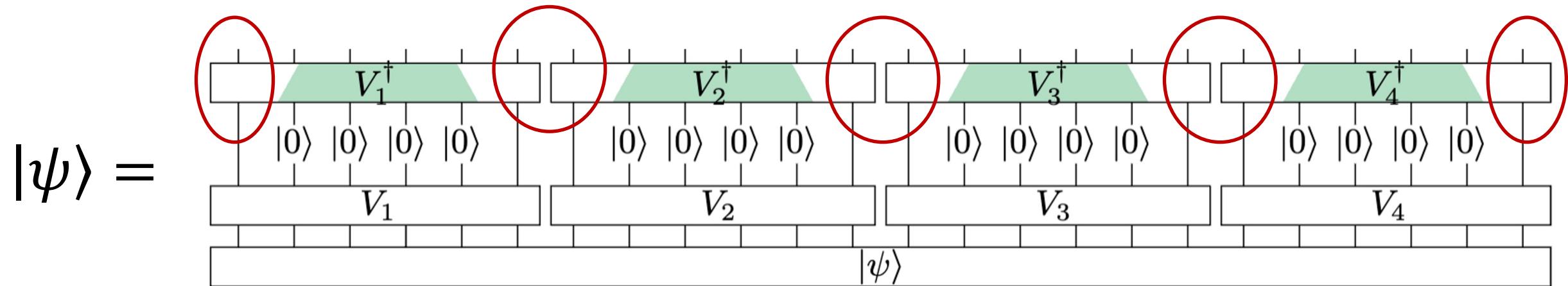


# Repeating what we just did

We have reconstructed part of the state

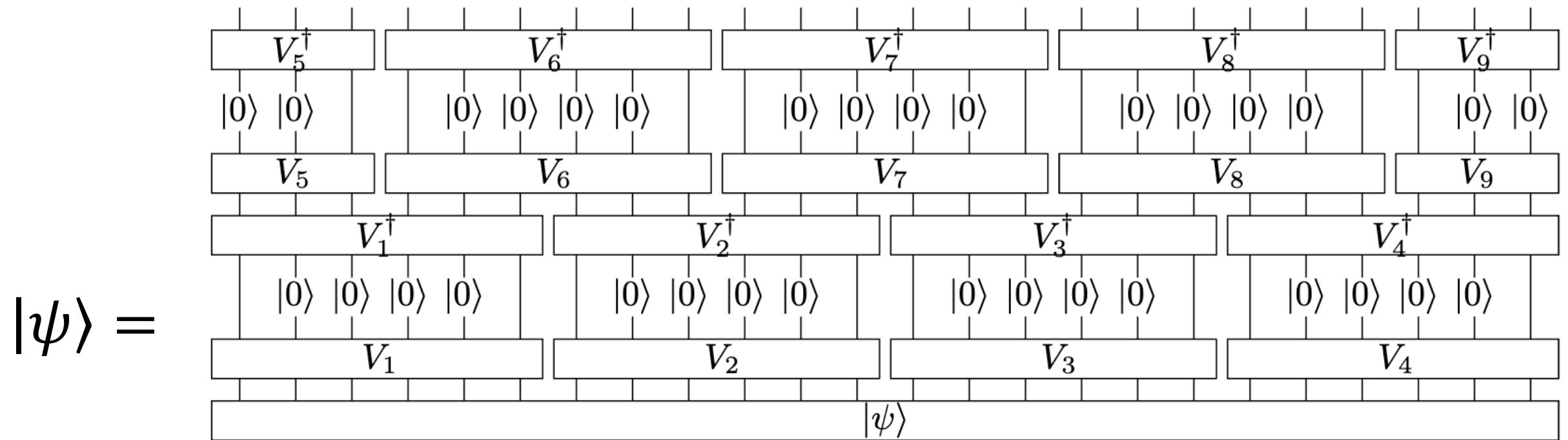
Not enough: what about **these regions**?

The state did not change! Just repeat by doing another layer



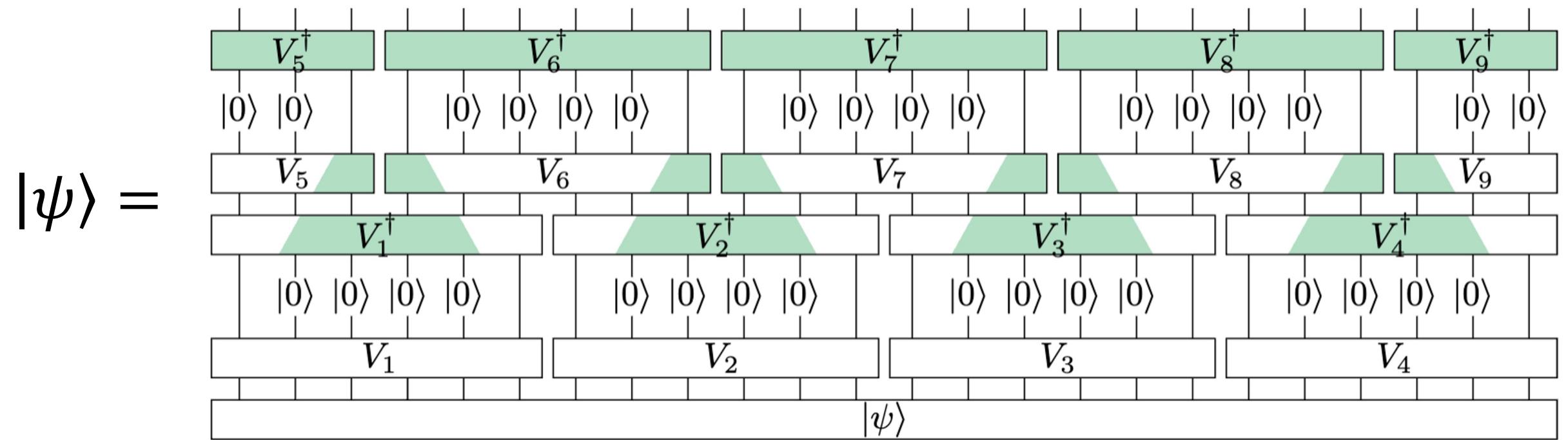
# Repeating what we just did

Potential issue: the new layer could mess up with what we had earlier  
Claim: problem solved; we already reconstructed a circuit for  $|\psi\rangle$



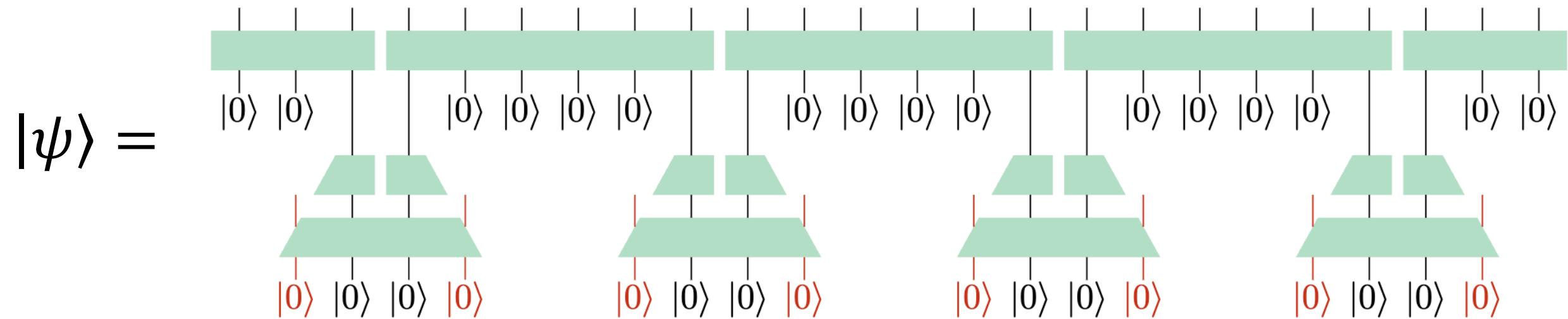
# Key idea: Reconstruction via backward lightcone

- Reconstruct the output state via **backward lightcone** of all wires
- **This works because backward lightcone stops entirely at fresh qubits**



# Reconstruction via backward lightcone

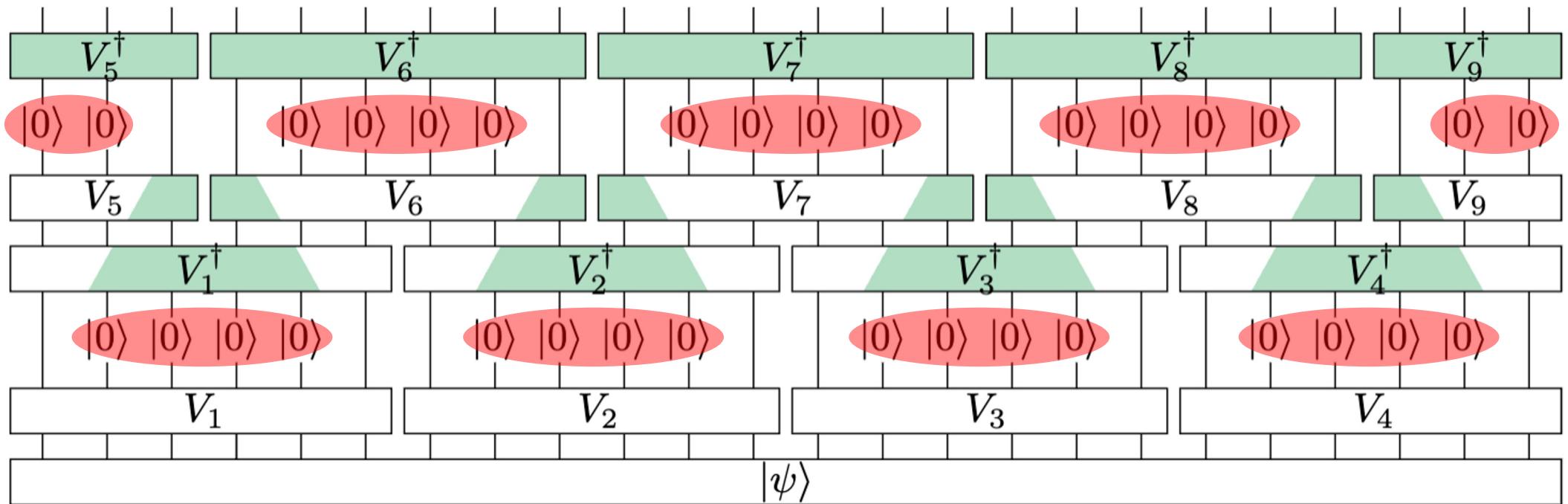
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- **This works because backward lightcone stops entirely at fresh qubits**



In general, this requires a careful geometric arrangement

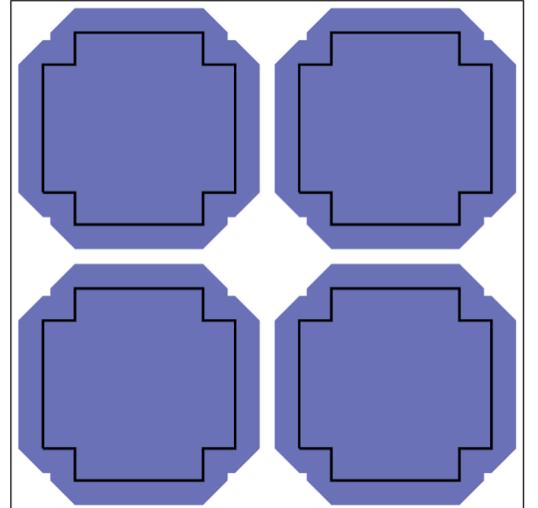
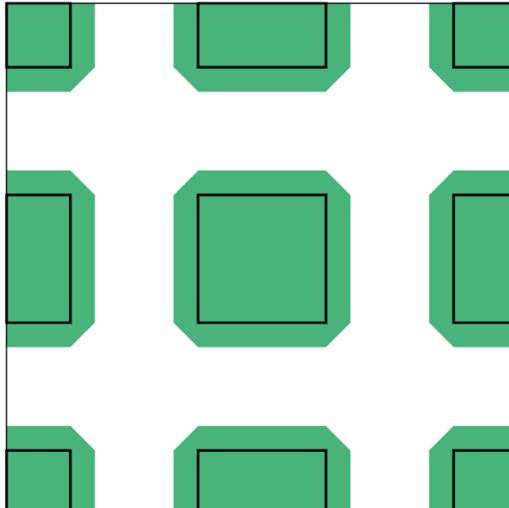
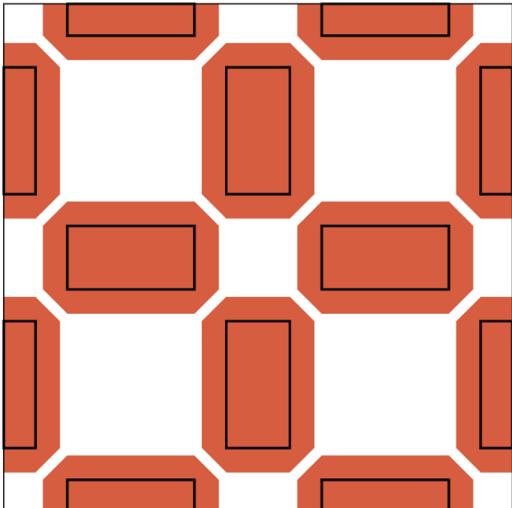
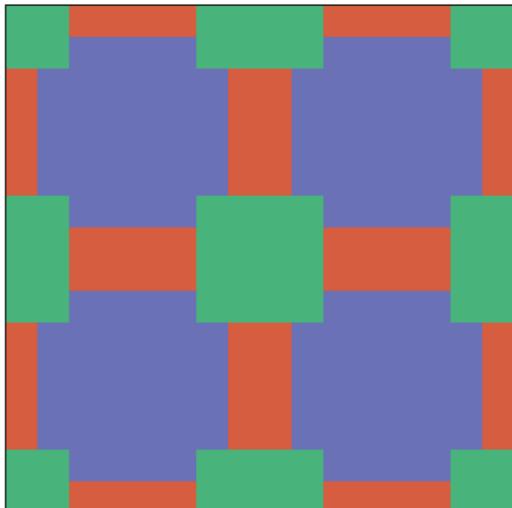
# Geometric arrangement: covering scheme

- Divide a lattice into layers of subsets
  - Subsets in the same layer are “sufficiently non-overlapping”
  - Subsets in different layers are “sufficiently overlapping”



# Geometric arrangement: covering scheme

- Divide a lattice into layers of subsets
  - Subsets in the same layer are “sufficiently non-overlapping”
  - Subsets in different layers are “sufficiently overlapping”
- Construction: take a lattice coloring, then make every subset larger



# Learning phases of matter

- “trivial phase”: quantum states prepared by constant depth circuits on finite dimensional lattice
- Quantum states in the “trivial phase” can be learned efficiently
- Corollary: given an arbitrary state, can efficiently test whether it is in the “trivial phase” (low circuit complexity),  
or “topological phase” (high circuit complexity)

# Discussion

# More general geometry?

- First algorithm (learning shallow quantum circuits) works for arbitrary or even unknown geometry
  - Recall one-line proof:  $U \otimes U^\dagger = (\prod_{i=1}^n S_i) \cdot \prod_{i=1}^n (U^\dagger S_i U)$
- Second algorithm (learning quantum states prepared shallow circuits) works for any geometry with a good covering scheme
  - We constructed good covering schemes for finite-dimensional lattices

# No ancilla qubits?

- Using ancilla qubits is essential in both learning algorithms
- Is it possible to reconstruct the circuit without using any ancilla qubits?
- In [arxiv 2401.10095] we give an algorithm specialized to learning quantum states in 2D, **no ancilla qubit assuming finite gate set**
  - **Key idea:** to learn a 2D state it suffices to solve a 1D CSP
  - Probably works for learning quantum circuit in 2D as well?

# Outlook

- Toward useful quantum advantage: learning shallow quantum circuits in NISQ algorithms
  - **Two new learning algorithms based on local inversion** that works in simple clean settings; very different from gradient descent
  - Can this be the basis of efficient learning algorithms in more general settings?
    - Need to deal with practical issues such as noise

# References

- Learning shallow quantum circuits: arxiv 2401.10095
- Learning quantum states prepared by shallow circuits in polynomial time:

