

2026 QuEra Technical Challenge Presentation



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Today's quantum computers face issues with **noise** and **decoherence**.

We aim use the following to **benchmark** and **improve performance** of the logical qubit as **a memory**.

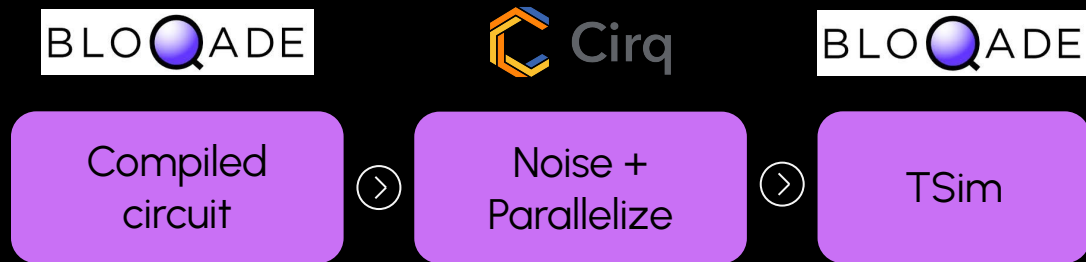
Steane QEC with
[7-1-3] color code and
magic state

Steane QEC with
[7-1-3] color code and
non-magic state

Steane QEC with
[[7-1-5] color code and
magic state

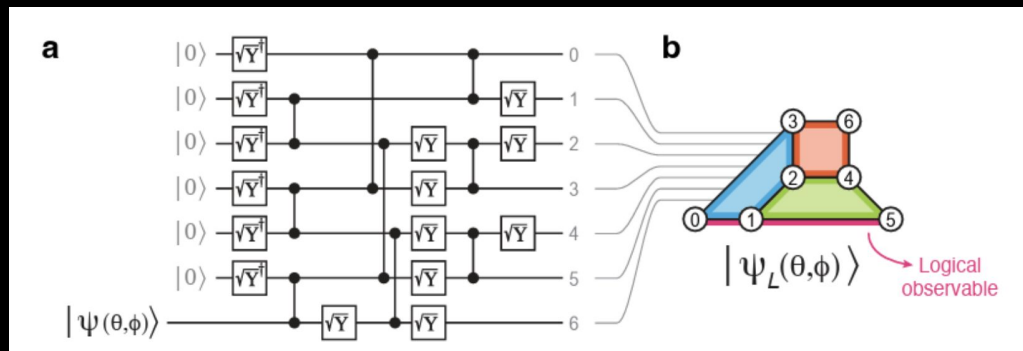
Steane QEC with
[[7-1-5] color code and
non-magic state

We compiled our static circuit into a Squin circuit, converted to Cirq, applied the GeminiOneZoneNoiseModel with custom noise modulation, parallelized the circuit, and re-converted to Squin for TSim



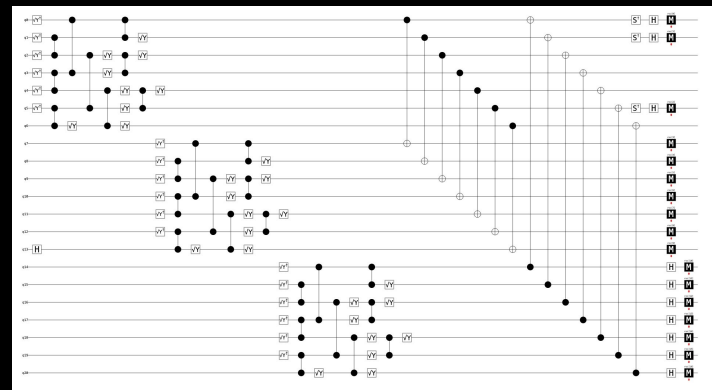
```
def embed_mover_noise(k, noise_param, dnoise):  
    """  
    Apply Gemini noise to a Squin kernel.  
    """  
    # Convert Squin kernel to Cirq circuit  
    cirq_circuit = utils.emit_circuit(k)  
  
    # Set default noise for each parameter  
    default_noises = {param: dnoise for param in noise_param}  
  
    # Initialize Gemini noise model  
    noise_model = utils.noise.GeminiOneZoneNoiseModel(**default_noises)  
  
    # Apply noise and convert to native gates, parallelizing if possible  
    noisy_cirq_circuit = utils.noise.transform_circuit(  
        cirq_circuit, to_native_gateset=True, model=noise_model, parallelize_circuit=True  
    )  
  
    # Load back into Squin  
    squin_circuit = utils.load_circuit(noisy_cirq_circuit, kernel_name="main_loaded")  
  
    return squin_circuit
```

We first prepared the T state, entangled it with 6 physical qubits, and injected the resulting logical qubit into a distance-3 color code [[7-1-3]]



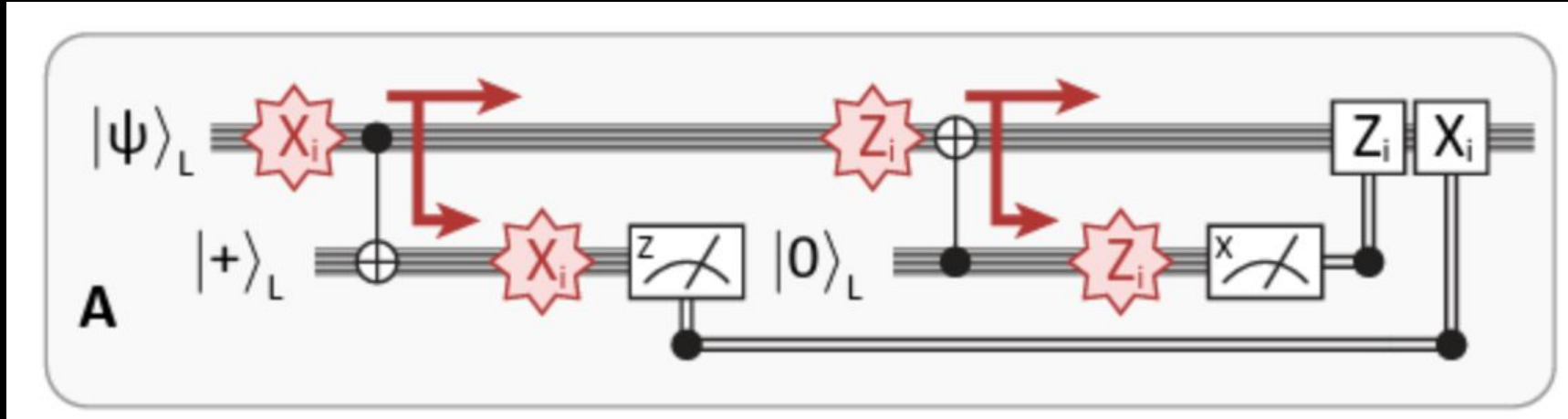
T state

Reference circuit



Code visualization of circuit

We utilized ancilla qubits to capture 2 types of unwanted noise:
X (bit) flips and **Z (phase) flips**



Prepare ancilla
qubit

Initialize to $|+\rangle$
and CNOT
target ancilla

Measure in Z
basis to detect X
flips

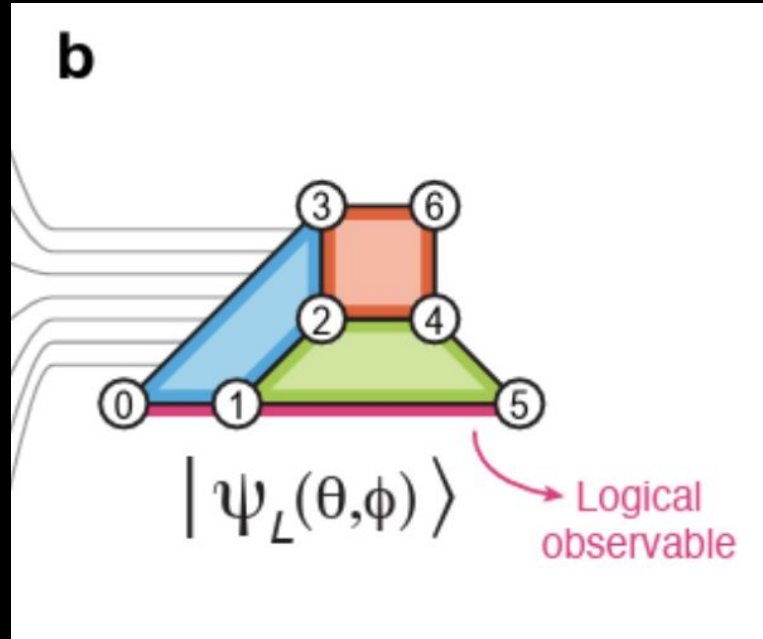


Prepare ancilla
qubit

Initialize to $|+\rangle$
and CNOT
control ancilla

Measure in X
basis to detect Z
flips

Without collapsing the logical qubit, we obtained the stabilizers and syndromes for both ancilla to determine post-selection

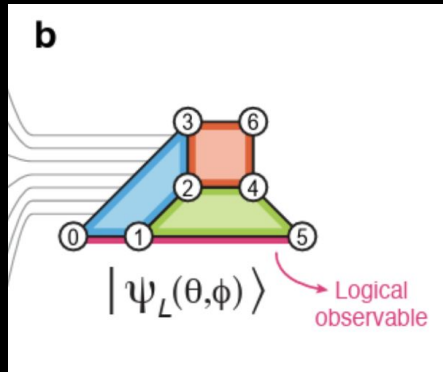
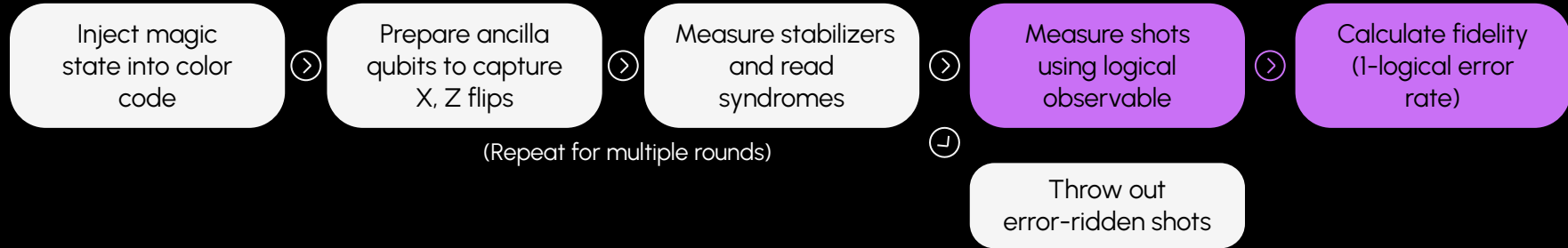


Stabilizer	Covered Qubits
Stabilizer Blue	Qubits 0, 1, 2, 3
Stabilizer Red	Qubits 2, 3, 4, 6
Stabilizer Green	Qubits 1, 2, 4, 5

If the product of the eigenvalues of the covered qubits is -1, then an error has occurred in that stabilizer color zone.

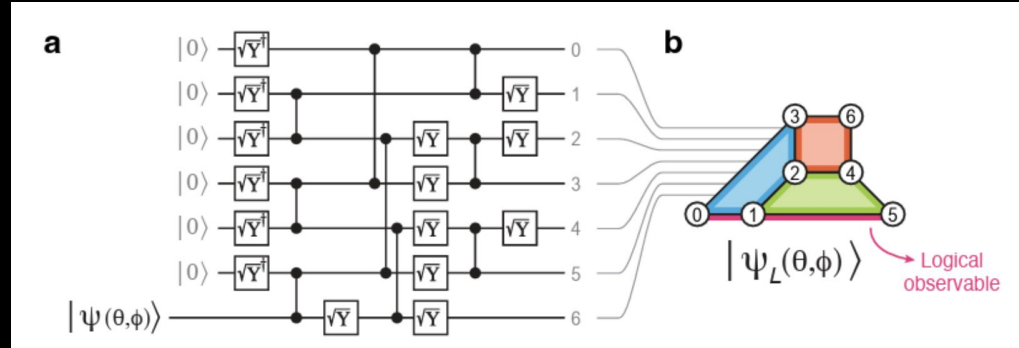
If even one stabilizer has eigenvalue -1 (syndrome indicates error) we throw out the entire run

We measure the shots that made it to the end using the logical observable, quantum tomography, and fidelity calculations



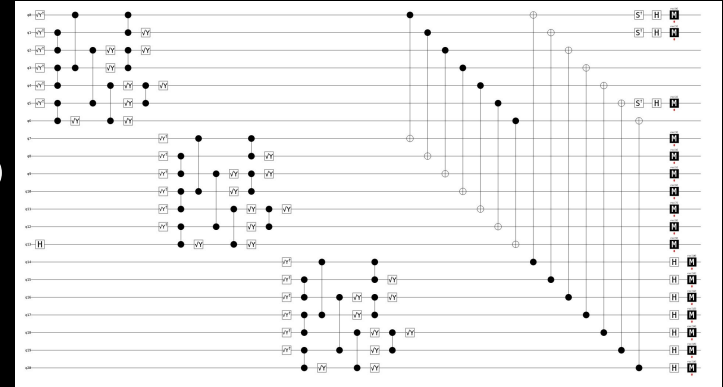
Calculating the product of the eigenvalues of the **logical observable** qubits gives us information to perform quantum tomography and obtain fidelity

We also injected the 0 state into a distance-3 color code to compare its logical error rates vs the previous magic T state



$|0\rangle$ state

Reference circuit

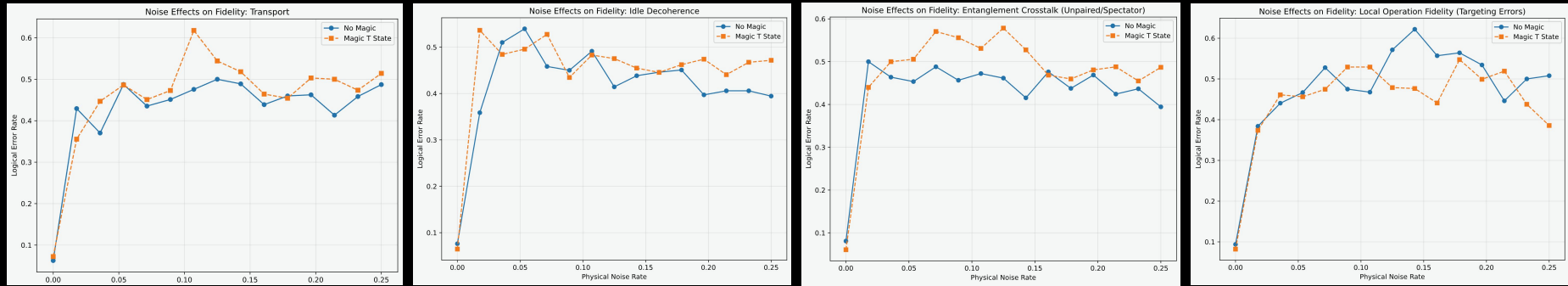


Code visualization of circuit

We prepared 4 relevant noise channels, grouping together related parameters presented by GeminiOneZoneNoiseModel

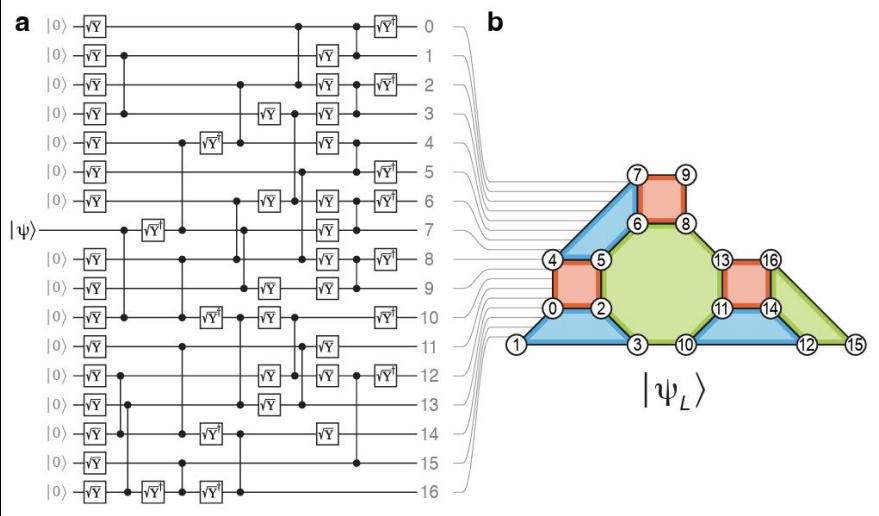
Noise Group	Significance	Parameters
Transport	Decoherence errors accumulated while atom is physically moved by optical tweezers	<ul style="list-style-type: none">mover_pxmover_pymover_pz
Idle Decoherence	Decoherence accumulated by stationary "sitter" atoms that wait while other atoms are being moved around them	<ul style="list-style-type: none">sitter_pxsitter_pysitter_pz
Entangled Crosstalk	Error accumulated by "spectator" atoms that sit in global laser beam without a partner during two-qubit gates	<ul style="list-style-type: none">cz_unpaired_gate_pxcz_unpaired_gate_pycz_unpaired_gate_pz
Local Operation Fidelity	baseline error rate for single-qubit gates that require precise individual targeting (~10 times higher than global operations)	<ul style="list-style-type: none">local_pxlocal_pylocal_pz

As expected, using our fidelity calculations from QEC, we were able to plot a roughly quadratic relationship between fidelity (1-logical error rate) as a function of physical error rate



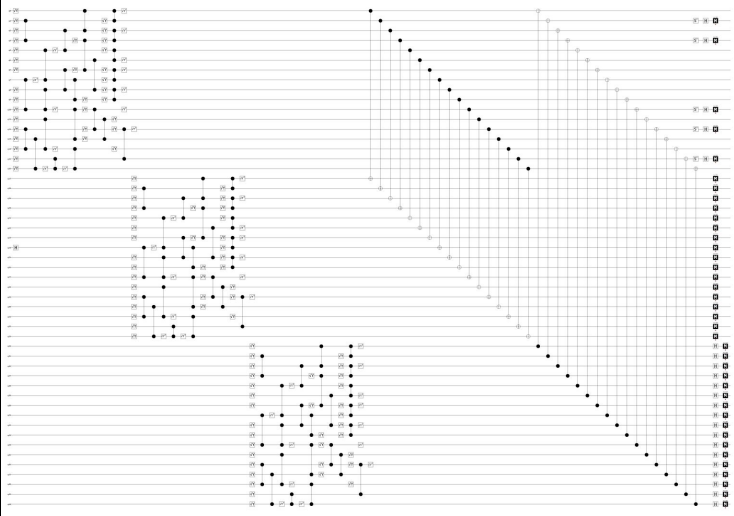
Across the 4 noise models, the **non-magic state** generally performed **similar** to the **magic state** (comparable logical error rate)

We injected the T state into a distance-5 color code with 16 physical qubits to observe its effect on logical error rates



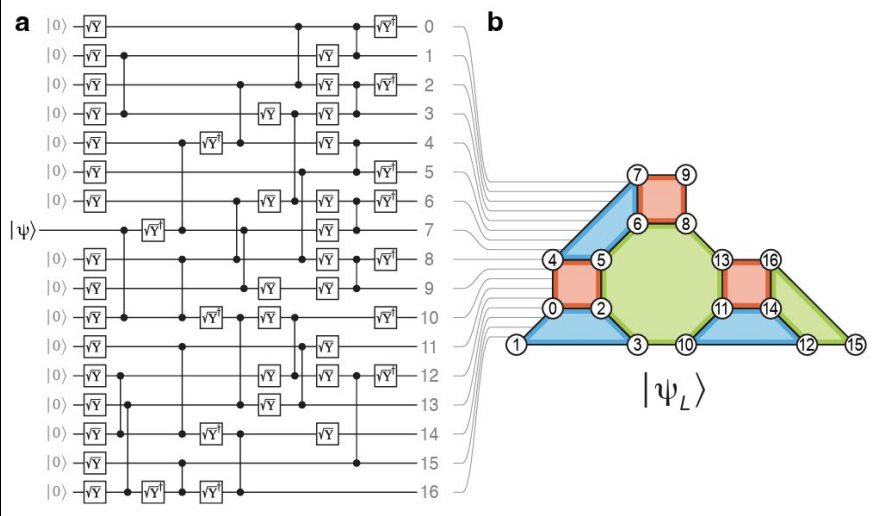
T state

Reference circuit



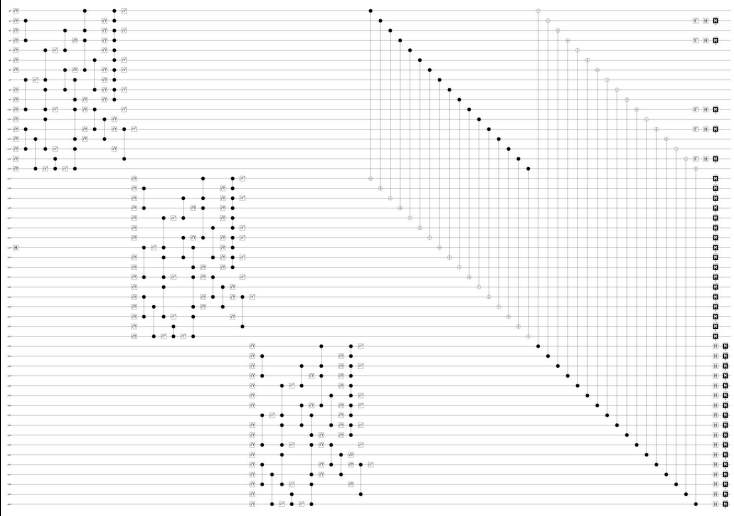
Code visualization of circuit

We also injected the 0 state into a distance-5 color code to observe its effect on logical error rates



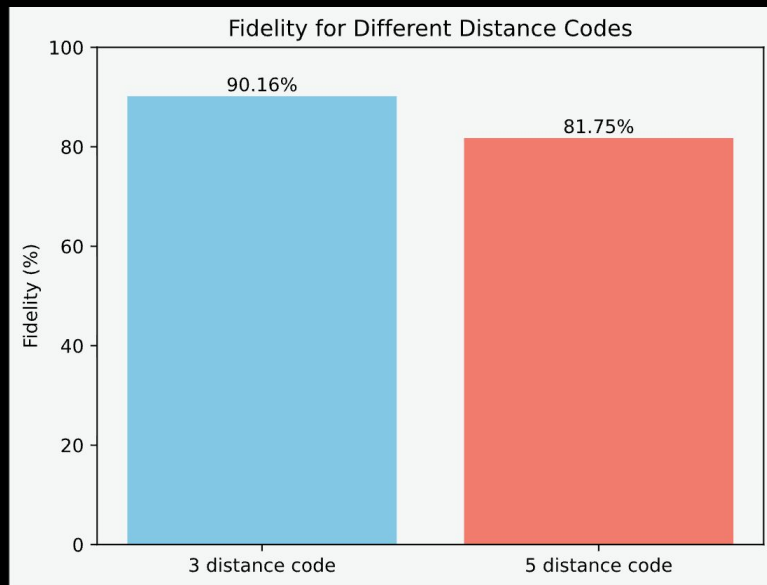
$|0\rangle$ state

Reference circuit

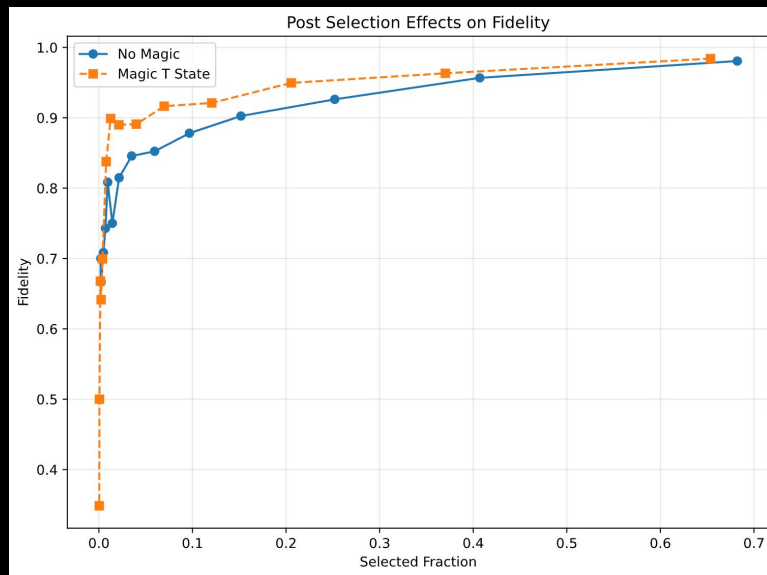


Code visualization of circuit

We found an average higher fidelity for distance-3 code vs distance-5 code



Comparing our results from magic and non-magic state, we saw that higher post-selection would have the expected positive effect on fidelity for distance-5 code



While we understood the feedforward logic, we ran into a bottleneck when attempting to control quantum gates using classical bits

```
def get_bit_and_stabilizers(kernel):  
    """  
    Run the kernel and extract logical measurement outcomes and stabilizers.  
    """  
    # Sample measurement outcomes  
    samples = kernel.compile_sampler(seed=0).sample(shots=5000, batch_size=10000)  
  
    # Split samples into logical qubits, plus block, and zero block  
    logical_combination = samples[:, :3]  
    plus_syndromes = samples[:, 3:10]  
    zero_syndromes = samples[:, 10:17]  
  
    # Compute logical bit values  
    logical_bit = eigen_calc_prod(logical_combination)  
  
    # Compute X-type stabilizers from + block  
    plus_stabilizers = np.stack([  
        eigen_calc_prod(plus_syndromes[:, [0, 1, 2, 3]]),  
        eigen_calc_prod(plus_syndromes[:, [1, 2, 4, 5]]),  
        eigen_calc_prod(plus_syndromes[:, [2, 4, 6, 3]]),  
    ], axis=1)  
  
    # Compute Z-type stabilizers from 0 block  
    zero_stabilizers = np.stack([  
        eigen_calc_prod(zero_syndromes[:, [0, 1, 2, 3]]),  
        eigen_calc_prod(zero_syndromes[:, [1, 2, 4, 5]]),  
        eigen_calc_prod(zero_syndromes[:, [2, 4, 6, 3]]),  
    ], axis=1)  
  
    # Combine X and Z stabilizers  
    stabilizers = np.concatenate([plus_stabilizers, zero_stabilizers], axis=1)  
  
    return logical_bit, stabilizers
```

Syndrome Extraction



```
@kernel  
def circuit():  
    # Allocate qubits for logical, plus, and zero blocks  
    q_logical = squin.qalloc(7)  
    q_plus = squin.qalloc(7)  
    q_zero = squin.qalloc(7)  
  
    # Inject magic state if requested  
    if inject_magic:  
        squin.broadcast.h(q_logical[-1])  
        squin.broadcast.t_adj(q_logical[-1])  
  
    # Prepare auxiliary + state  
    for _ in range(n_iterations):  
        squin.broadcast.h(q_plus[-1])  
        encode_block(q_logical)  
        encode_block(q_plus)  
        encode_block(q_zero)  
  
    # Entangle logical, plus, and zero blocks  
    squin.broadcast.cx(q_logical, q_plus)  
    squin.broadcast.cx(q_zero, q_logical)  
  
    # Measure blocks  
    measure_all(q_plus)  
    squin.broadcast.h(q_zero)  
    measure_all(q_zero)  
  
    # Classical logic involving if statements  
    # <==== Here =====>  
  
    measure_subset(q_logical)  
  
    return circuit
```

Quantum Circuit Classical Logic Integration