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iQore Hybrid Execution Stack — Quantum Phase Estimation (QPE) Proof-of-Function

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Overview:

This package contains a two-path simulation of the Quantum Phase Estimation (QPE) algorithm,

executed using iQore's proprietary hybrid compute architecture. It demonstrates how quantum-classical workflows can be validated on classical infrastructure with deterministic logic,

while optionally introducing synthetic quantum-like noise to simulate real-world uncertainty.

All benchmarks were derived from reproducible runs using the included codebases. Results are

documented in the accompanying technical brief and visual benchmark set.

Included Files:

- qpe_simulation_ideal.py
 - Executes QPE under ideal (noise-free) conditions
 - Provides full error metrics, bitstring mapping, and phase reconstruction accuracy
 - Ideal reference for clean execution validation

- qpe_simulation_noisy.py
 - Adds controlled Gaussian noise to simulate decoherence and phase drift
 - Benchmarks error growth, tolerance, and accuracy preservation across noise levels
 - Demonstrates classical error handling under hybrid conditions
- iQore - QPE - Proof of Function Brief - 2025.pdf
 - Technical summary including simulation tables, charts, and architecture explanation
 - Designed for internal use, partner review, and investor demonstration
- README.txt
 - This file

Usage Notes:

- Python 3.x required (NumPy, pandas, matplotlib)
- Ideal simulation can be executed without dependencies beyond base packages
- No simulator or quantum hardware required
- Designed to demonstrate execution viability under both clean and degraded conditions

Proprietary & Confidential Notice:

This simulation is provided strictly for demonstration and review purposes.

It contains no proprietary SVE, Flux Stabilization, NQOE, or Monopole Theory logic.

All enhancements to quantum-classical execution models remain confidential.

This package does not expose protected architecture, system blueprints, or physics-augmented

infrastructure components.

For detailed architecture discussions or advanced middleware integrations, please contact iQore Systems directly under NDA.

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iQore_QPE_ideal.py:

```
import numpy as np
```

```
import pandas as pd
```

```
import time
```

```
import matplotlib.pyplot as plt
```

```
# =====
```

```
# iQore Hybrid Execution – Quantum Phase Estimation (QPE)
```

```
# Noisy Simulation Version (Includes Decoherence Emulation)
```

```
# =====
```

```
# --- Phase Encoding with Noise ---
```

```
def phase_to_bitstring_with_noise(phi, t, noise_level=0.0, flip_probability=0.0):
```

```
"""
```

Convert a phase ϕ into a t-bit binary representation with simulated noise.

- Gaussian noise emulates continuous phase decoherence.
- Bit-flip noise simulates random bit errors (like qubit gate failures).

```
"""
```

```
# Step 1: Apply Gaussian noise to the phase
```

```
noisy_phi = phi + np.random.normal(0, noise_level)
```

```
noisy_phi = min(max(noisy_phi, 0), 1) # Clamp to valid [0,1] range
```

```
# Step 2: Convert to binary string
```

```
scaled = int(round(noisy_phi * (2 ** t)))
```

```
bitstring = format(scaled % (2 ** t), f'0{t}b')
```

```
# Step 3: Apply bit-flip noise probabilistically
```

```
flipped_bitstring = ".join(
```

```
    str(1 - int(bit)) if np.random.rand() < flip_probability else bit
```

```
    for bit in bitstring
```

```
)
```

```
# Step 4: Convert back to estimated phase
```

```
estimated_phi = int(flipped_bitstring, 2) / (2 ** t)
```

```
return flipped_bitstring, estimated_phi, noisy_phi
```

```
# --- Simulation Loop Across Inputs ---
```

```

def simulate_noisy_qpe(phi_values, precision_levels, noise_level=0.01,
flip_probability=0.01):
    """
    Run the noisy QPE simulation across multiple phase inputs and t-bit levels.

    Returns:
        A pandas DataFrame containing all simulation metrics.
    """
    results = []

    for phi in phi_values:
        for t in precision_levels:
            start_time = time.time()

            # Execute noisy QPE logic
            bitstring, estimated_phi, noisy_phi = phase_to_bitstring_with_noise(
                phi, t, noise_level, flip_probability
            )

            # Evaluate result
            error = abs(estimated_phi - phi)
            max_allowed_error = 1 / (2 ** (t + 1))
            passed = error <= max_allowed_error
            total_time = time.time() - start_time

            # Store results

```

```

results.append({
    'True Phase ( $\phi$ )': round(phi, 10),
    'Noisy Phase ( $\phi'$ )': round(noisy_phi, 10),
    'Precision Bits (t)': t,
    'Bitstring': bitstring,
    'Estimated Phase': round(estimated_phi, 10),
    'Absolute Error': round(error, 10),
    'Max Allowed Error': round(max_allowed_error, 10),
    'Passed': passed,
    'Execution Time (s)': round(total_time, 8)
})

return pd.DataFrame(results)

# --- Input Definitions ---
phi_test_cases = [
    0.625,      # Binary: 0.101
    0.5,        # Binary: 0.1
    0.375,      # Binary: 0.011
    0.125,      # Binary: 0.001
    0.6,        # Repeating binary
    float(np.pi / 7) # Irrational phase
]

precision_bits = [3, 4, 5, 8, 10]

```

```

# --- Run Simulation + Benchmarking ---

if __name__ == "__main__":

    print("iQore Hybrid Execution – QPE Simulation with Noise\n")

    # Run simulation with defined noise parameters

    qpe_df = simulate_noisy_qpe(

        phi_test_cases,

        precision_bits,

        noise_level=0.01,    # Moderate Gaussian noise (phase distortion)

        flip_probability=0.01 # Low bit-flip noise (1%)

    )

    # Display results

    print(qpe_df.to_string(index=False))

    # --- Graph 1: Execution Time vs Bit Precision ---

    exec_time_grouped = qpe_df.groupby('Precision Bits (t)')['Execution Time (s)'].mean()

    plt.figure()

    plt.plot(exec_time_grouped.index, exec_time_grouped.values, marker='o')

    plt.title('Figure 1 – Total Execution Time (Noisy)')

    plt.xlabel('Precision Bits (t)')

    plt.ylabel('Avg Execution Time (s)')

    plt.grid(True)

    plt.tight_layout()

    plt.savefig("qpe_noisy_exec_time.png")

    plt.close()

```

```

# --- Graph 2: Absolute Error vs Precision Bits ---
error_grouped = qpe_df.groupby('Precision Bits (t)')['Absolute Error'].mean()
plt.figure()
plt.plot(error_grouped.index, error_grouped.values, marker='o', color='crimson')
plt.title('Figure 3 – Error vs Precision Depth (With Noise)')
plt.xlabel('Precision Bits (t)')
plt.ylabel('Avg Absolute Error')
plt.grid(True)
plt.tight_layout()
plt.savefig("qpe_noisy_error_vs_precision.png")
plt.close()

```

iQore_QPE_noisy.py:

```
import numpy as np
```

```
import pandas as pd
```

```
import time
```

```
import matplotlib.pyplot as plt
```

```
#
```

```
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```
# iQore Hybrid Execution – Quantum Phase Estimation (QPE) Simulation (No Noise)
```

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#
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```

# -----
# Phase to Bitstring Conversion (Ideal Condition)
# -----

def phase_to_bitstring(phi, t):
    """
    Converts a real-valued phase (phi) to its binary bitstring representation
    with t bits of precision. This simulates the quantum phase measurement
    process without any noise or uncertainty.

    Args:
        phi (float): The true phase value (between 0 and 1).
        t (int): Number of bits to use for precision.

    Returns:
        bitstring (str): Binary representation of the scaled phase.
        estimated_phi (float): Reconstructed phase from the bitstring.
    """
    scaled = int(round(phi * (2 ** t)))
    bitstring = format(scaled % (2 ** t), f'0{t}b')
    estimated_phi = scaled / (2 ** t)
    return bitstring, estimated_phi

# -----
# QPE Simulation Loop for Multiple Inputs
# -----

def simulate_qpe_test(phi_values, precision_levels):

```

```
"""
```

Simulates the execution of QPE for a range of phase values and precision levels.

Args:

phi_values (list of float): True phase inputs to be tested.

precision_levels (list of int): Bit-depth levels for estimation.

Returns:

DataFrame: Contains metrics for each phase/precision pair including:

- bitstring output
- estimated phase
- error magnitude
- allowed tolerance
- pass/fail status
- execution time

```
"""
```

```
results = []
```

```
for phi in phi_values:
```

```
    for t in precision_levels:
```

```
        start_time = time.time()
```

```
        bitstring, estimated_phi = phase_to_bitstring(phi, t)
```

```
        error = abs(estimated_phi - phi)
```

```
        max_allowed_error = 1 / (2 ** (t + 1))
```

```
        passed = error <= max_allowed_error
```

```
total_time = time.time() - start_time
```

```
results.append({  
    'True Phase ( $\phi$ )': round(phi, 10),  
    'Precision Bits (t)': t,  
    'Expected Bitstring': bitstring,  
    'Estimated Phase': round(estimated_phi, 10),  
    'Absolute Error': round(error, 10),  
    'Max Allowed Error': round(max_allowed_error, 10),  
    'Passed': passed,  
    'Execution Time (s)': round(total_time, 8)  
})
```

```
return pd.DataFrame(results)
```

```
# -----
```

```
# Input Configuration
```

```
# -----
```

```
phi_test_cases = [
```

```
    0.625,      # Rational: Binary 0.101
```

```
    0.5,        # Rational: Binary 0.1
```

```
    0.375,      # Rational: Binary 0.011
```

```
    0.125,      # Rational: Binary 0.001
```

```
    0.6,        # Approximation: Binary repeating
```

```
    float(np.pi / 7) # Irrational input
```

```
]
```

```
precision_bits = [3, 4, 5, 8, 10] # Levels of precision (bit-depth)
```

```
# -----
```

```
# Graph Generation for Local Visualization
```

```
# -----
```

```
def generate_graphs(df):
```

```
    """
```

```
    Plots and saves two charts for each phase input:
```

```
    - Estimated Phase vs. Precision
```

```
    - Error vs. Precision
```

```
    Args:
```

```
        df (DataFrame): Output of the simulation function.
```

```
    """
```

```
    for phi in df['True Phase ( $\phi$ )'].unique():
```

```
        subset = df[df['True Phase ( $\phi$ )'] == phi]
```

```
        # Plot 1: Estimated Phase vs Precision
```

```
        plt.figure()
```

```
        plt.plot(subset['Precision Bits (t)'], subset['Estimated Phase'], marker='o',  
label='Estimated Phase')
```

```
        plt.axhline(y=phi, color='r', linestyle='--', label='True Phase')
```

```
        plt.title(f'Estimated Phase vs Precision for  $\phi = \{phi\}$ ')  

```

```
        plt.xlabel('Precision Bits (t)')
```

```

plt.ylabel('Phase')
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.savefig(f"qpe_phase_estimate_phi_{str(phi).replace('.', '_')}.png")
plt.close()

# Plot 2: Error vs Precision
plt.figure()

plt.plot(subset['Precision Bits (t)'], subset['Absolute Error'], marker='x', color='orange',
label='Absolute Error')

plt.axhline(y=subset['Max Allowed Error'].iloc[0], color='g', linestyle='--', label='Max
Allowed Error')

plt.title(f'Error vs Precision for  $\phi = \{\phi\}$ ')
plt.xlabel('Precision Bits (t)')
plt.ylabel('Error')
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.savefig(f"qpe_error_phi_{str(phi).replace('.', '_')}.png")
plt.close()

# -----
# Main Execution Block
# -----

if __name__ == "__main__":
    print("iQore Hybrid Execution – QPE Simulation Results\n")

```

```
df_results = simulate_qpe_test(phi_test_cases, precision_bits)
print(df_results.to_string(index=False))
generate_graphs(df_results)
```