# **Problem 2: HHL Algorithm**

## **Problem & Analysis**

Solving a linear equation A \* x = b in a quantum computational manner, where A is an square matrix and b is an vector with matching dimension. For a simple example:

```
Solving Ax = b where A is \begin{bmatrix} 1, & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2}, & -1/\sqrt{2} \end{bmatrix} and b is \begin{bmatrix} 1/2 \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} \end{bmatrix}
The solution is unique: \begin{bmatrix} -1/4 \end{bmatrix} \begin{bmatrix} 3/4 \end{bmatrix}
```

From linear algebra basics, we know that the solution  $\times$  might be one from nothing, unique, or an ensemble, depending on the rank of the expanded matrix [A|b]. In **any of these cases** noted, we can apply the Gaussian Elimination method or Matrix Inversion to get the accurate answer, but naive implementations cost time of  $O(n^2)$ , while the most optimized classical algorithm reduces it to O(N\*sqrt(k)), where k describes the cost for inversing the matrix. However, in **case of unique solution**, we have another perspective view and thoughts upon form of the linear equation, where is also probably the HHL Alogrithm derives.

In quantum computation traditions, U denotes a quantum gate and  $|phi\rangle$  denotes a quantum state, when U is applied to  $|phi\rangle$ , it turns to be a new state, say U  $|phi\rangle$  ->  $|psi\rangle$ . Corresponding to the physical process that, a quantum system **evolves** its state from  $|phi\rangle$  to  $|psi\rangle$  under the environment influence U. Compared with the linear equation A \* x = b, it's easy to figure out the logical structure in parallel:

```
quantum evolution: U \mid phi \rangle \rightarrow |psi \rangle linear equation: A \mid x \mid = b where U and A are square matrices, others are all vectors in computational representation
```

i From this view, solving a linear equation is like to **finding the initial state of a quantum evolution process** where the final state is known to be |psi> while the evolution operator is U. Due to the **reversible computing** nature of quantum computating, this is not a magic in philosophical sense.

## **Solution**

Now it's clear enough to put an elephant into the refrigerator:

- Encode classical A and b to quantum U and |psi>
  - o ⚠ tricky: find ways to satisfy the required unitary & unit vector condition for A and b
- Find |phi> through matrix inversion of u, aka. computing the reversed |phi> = U+|psi>
  - • ∆ tricky: do not inverse u directly, but instead decompose |psi> to u 's eigen vectors, then simply inverse the eigen values
- Decode |phi> to get the solution x (in a probablistic approximated sense)

Assuming the toy case dim(A) = rand(A) = 2, we implement & explain the HHL alogrithm in both mathematic and programmatic view in following sections.

#### HHL in a mathematic view

Firstly, here's the mathematic formula outline sketch:

```
A|x\rangle = |b\rangle
                                       // assume A is an invertibe hermitian (no need be unitary), and |b> is a unit vector
 |x\rangle = A^{(-1)} |b\rangle
      = Σi 1/λi |vi><vi| |b>
                                                     // eigen decomposition of A^{(-1)} where \lambda i and |vi\rangle are the i-th eigen parameters
      = (\Sigma i \ 1/\lambda i \ | vi > \langle vi |) \ (\Sigma j \ \beta j \ | vj >) // decompose |b> to A's eigen basis |vi>; i,j \in [0, N-1], N = dim(A)
      = ΣiΣj βj/λi |vi><vi|vj>
      = \Sigma i \Sigma j \beta j / \lambda i | vi >
                                                     // <vi|vj> == 1 if i==j else 0
      = Σi (Σj βj) 1/λi |vi>
                                                    // aggregate inner const coeffs
      = Σi βi/λi |vi>
                                                      // rename, βi only depends on i
QPE 2(A, |00b>)
                                     // estimates eigen value of A over vector |b>, using two ancilla qubits inited as |
  = C_02(U(\theta 1)) * C_12(U(\theta 2)) * |++b>
                                                                    // | + \rangle = H | 0 \rangle, \theta k for q[k-1]
  = C_02(U(\theta 1)) * C_12(U(\theta 2)) * |+>(|\theta>+|1>)|b> // ignore global phase
  = C_02(U(\theta 1)) * |+>(|\theta b>+|1>U(\theta 2)|b>)
  = C_02(U(\theta 1)) * (|0>+|1>)(|0b>+|1>U(\theta 2)|b>)
  = C_02(U(\theta 1)) * (|00b\rangle + |01\rangle U(\theta 2)|b\rangle + |10b\rangle + |11\rangle U(\theta 2)|b\rangle)
  = |00b\rangle + |01\rangle U(\theta 2)|b\rangle + |10\rangle U(\theta 1)|b\rangle + |11\rangle U(\theta 1)*U(\theta 2)|b\rangle // <= how to choose U, then reduce U(\theta k)|b\rangle??
let U(t) = \exp(iAt) = \Sigma j \exp(i*\lambda j*t) |vj>< vj| where t is a tunable parameter (eg. 2*pi), now U is unitary since !
  = |00b\rangle + |01\rangle \exp(iAt2)|b\rangle + |10\rangle \exp(iAt1)|b\rangle + |11\rangle \exp(iAt1)^* \exp(iAt2)|b\rangle
  = |00b\rangle + |01\rangle \exp(iAt2)|b\rangle + |10\rangle \exp(iAt1)|b\rangle + |11\rangle \exp(iA(t1+t2))|b\rangle
  = |00b\rangle + \Sigma j \exp(i*\lambda j*t2) |vj\rangle \langle vj|b\rangle + \Sigma j \exp(i*\lambda j*t1) |vj\rangle \langle vj|b\rangle + \Sigma j \exp(i*\lambda j*(t1+t2)) |vj\rangle \langle vj|b\rangle
again decompse |b\rangle to \Sigma j \beta j |vj\rangle:
  = (|00>
     + \Sigma j \exp(i*\lambda j*t2) |01>
     + Σj exp(i*λj*t1) |10>
     + \Sigma j \exp(i*\lambda j*(t1+t2)) | 11>) \beta j | vj>
  = \Sigma j bj |\lambda j'\rangle |vj\rangle
where \lambda j' is the n-qubit binary approximation to 2^n*(\lambda j*t/2*pi), while \lambda j is the real eigvals, i.e.:
          \lambda j' = 2^n*(\lambda j*t/2*pi)
  \lambda j' / 2^n = \lambda j*t / 2*pi
U|b\rangle = U(\Sigma j \beta j | uj\rangle) // again decompose |b> to U's eigen basis |uj>
      = Σj βj U|uj>
      = Σj βj λj|vj>
                               // U|vj\rangle = \lambda j|vj\rangle if \lambda j and |vj\rangle are the i-th eigen pair of U
```

### HHL in a programmatic view

And here's the pseudo-code sketch for the procedure framework:

```
def HHL(A:Matrix, b:Vector) -> Vector:
    # Step 1: classical preprocess
    (Ah, b_n), stats = transform(A, b)  # transform to `Ah * y = b_n`

# Step 2: quantum computing
    cq = HHL_circuit(Ah, b_n)  # build qcircuit
    qstate = qvm.run(cq)  # run and get the final state vector
    y = project_q3(qstate)  # only need amplitude of |q3>

# Step 3: classical postprocess
    x = transform_inv(y, *stats)  # transform back to get `x`

return x
```

Components are explained in following sections.

Transform the equation

To make the quantum evolution framework work, it requires A to be **hermitian** while B to be a **normalized vector**.

However, A and b are **arbitarily** given from a linear equation.

Follow this to transform an arbitary linear equation to suit valid quantum gate and state:

It is easy to inverse the transformation back later, in order to get the final answer:

```
x / |A'*b| = y

x = y * |A'*b| # the final answer
```

Encode qunatum data

Find the circuits to encode the classical data to quantum facts  $U = \exp(iA\theta)$  and  $|b\rangle = b\theta|0\rangle + b1|1\rangle$ :

```
def encode_A(A:Matrix, θ:float) -> QGate:
    ''' encode a hermitian A to unitary exp(iAθ) '''
    assert is_hermitian(A)
    u = scipy.linalg.expm(1j*A*θ)  # matrix expotional, turing to be a unitry
    assert is_unitary(u)
    return QOracle(u)  # make an Oracle gate, or `matrix_decompose()` to U4 gates

def encode_b(b:Vector) -> QCircuit:
    ''' encode a unit vector b onto amplitude of |b> '''
    assert is_unit(b)
    θ = 2 * np.arccos(b[θ])
    cq = QCircuit() << RY(θ)  # rotate with RY gate
    if b[1] < θ: cq << Z()  # fix the sign of |1> part
    return cq
```

Main circuit routine

Construct the main circuit routine as HHL regires: QPE, controlled RY rotation and iQPE.

```
def HHL_circuit(A:Matrix, b:Vector, t0=2*pi, r=4) -> QCircuit:
  # alloc qubits
  q0, q1, q2, q3 = list(qvm.qAlloc_many(4))
  # Step 1: prepare |b>
  enc_b = encode_b(b, q3)
  # Step 2: apply QPE of A over |b>
  qpe = QCircuit() \
      << H([q1, q2]) \
      << encode_A(A, q3, theta=t0/4).control(q2) \</pre>
      << encode_A(A, q3, theta=t0/2).control(q1) \</pre>
      << SWAP(q1, q2) \
      << H(q2) \
      << S(q2).dagger().control(q1) \</pre>
      << H(q1) \
      SWAP(q1, q2)
  # Step 3: controlled rotate
  rc = QCircuit() \
     << RY(q0, 2*pi/2**r).control(q1) \</pre>
     << RY(q0, pi/2**r).control(q2)</pre>
  return enc_b << qpe << rc << qpe.dagger()</pre>
```

#### **Source Code**

Tested under pygpanda 3.7.12 + Python 3.8.15

#### Run demo:



### references

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