

# Quantum Algorithm for Solving LSE

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#### Desiderata

- Problem Definition
- Classical Solutions
- Problem Re-Definition
- Quantum Solution
  - Insight
  - Algorithm
  - Subtleties
  - Exceptions to the Rule
- Classical Simulation
  - Old Faithful
  - Results



#### **Problem Definition**

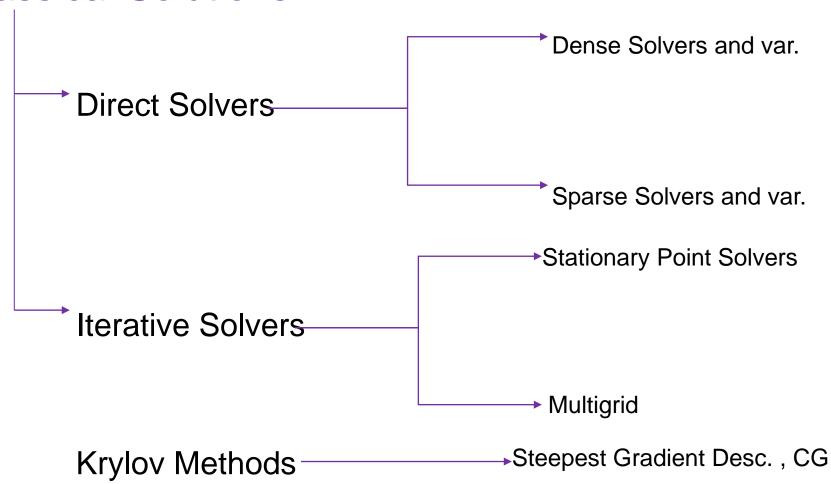
$$A x = b$$

$$A \epsilon R^{N \times N} \qquad x \epsilon R^{N} \qquad b \epsilon R^{N}$$

- Most ubiquitous equation
- Does it make sense to talk about this? Yes
  - Big Data in Science LHC
  - Climate Modelling
  - Deep Learning



#### **Classical Solutions**





#### **Quantum Solution**

- HHL Algorithm 2009
  - Harrow Hassidim Lloyd.

- Exponential Speedups
  - Quantum Simulation RPF
  - Shor's Algorithm
  - Grover's Search
  - HHL

$$\tilde{O}\left(\log(N)s^2\kappa^2/\epsilon\right)$$



# HHL: Insight

Any Matrix A and its inverse has reciprocating eigenvalues.

$$b := \sum b_i u_i$$

$$Ab := \lambda \sum b_i u_i$$

$$A^{-1}b := \frac{1}{\lambda} \sum b_i u_i$$

If there exists a quantum algorithm to compute  $\lambda$  and in turn  $1/\lambda$ , we could compute x.



#### Problem Re-definition

1) How to determine eigenvalues of an arbitrary matrix A?

2) Given a register of bitwise representations of a decimal number, how an we compute its reciprocal?

3) Given a bitwise representation of a given normalized angle, how can we compute the sine of the angle?



#### Phase Estimation

Assuming that  $|v\rangle$  is an eigenvector of U and  $\phi$  the corresponding phase angle.

$$U|v\rangle = exp^{2i\pi\phi}|v\rangle$$

In the real world, Unitary Matrices (Orthogonal Matrices) are hard to come by.

A is rarely unitary. =>?

Enter: Quantum Simulation ... aka Hamiltonian Simulation.



#### Hamiltonian Simulation

The moral we draw is that "information is physical."



#### Idea:

- 1) Convert your matrix A to a valid Hamiltonian H.
- 2) Allow a closed / open quantum system to evolve regulated by H.

#### Realization:

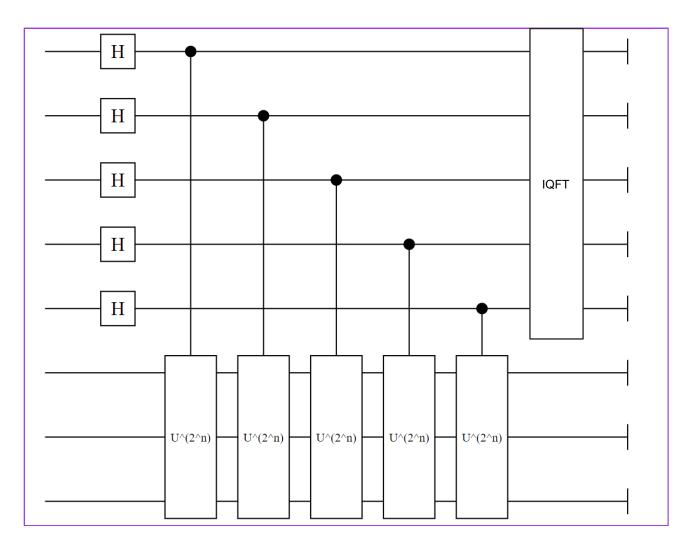
$$\tilde{O}(\log(N)s^2t)$$

- 1) HARD!: Sparse and Dense Hamiltonians -> Quantum Simulations.
- 2) Smaller ε and Faster τ are focal points.

Lie Trotter decomp / recomb ; Galaxy decomp ; Forrest decomp .....



Phase Estimation: Successive Controlled Hamiltonian Simulations





#### Phase Estimation – Crash Course

$$|0\rangle^{\otimes t}\otimes|u\rangle$$

$$\frac{|0\rangle+|1\rangle}{\sqrt{2}}^{\otimes t}\otimes|u\rangle$$

$$\frac{|0\rangle+|1\rangle}{\sqrt{2}}^{\otimes t}\otimes U|u\rangle$$

$$\frac{|0
angle + exp^{2i\pi\phi}|1
angle}{\sqrt{2}} \otimes U|u
angle$$

$$\bigotimes_{j=0}^{t-1} \left[ \frac{|0\rangle + exp^{2i\pi\phi*2^j}|1\rangle}{\sqrt{2}} \otimes U^{2^j}|u\rangle \right]$$

$$\sum_{k=0}^{2^t-1} exp^{2ij\pi\phi} |k\rangle$$

Eigenvalues of U take the form  $exp(2i\pi\phi)$ .

$$U = exp(2\pi iA).$$

Eigenvalues of U are now  $exp(2i\pi\lambda)$  where  $\lambda$  is the eigenvalue of A

At the end of PE, the register contains  $\phi$  for the normal case.

In the QS case PE produces  $\lambda$  in the register.

## 3) Sine of a bitstring – Controlled Rotation



$$|\theta\rangle|0\rangle - > |\theta\rangle(\cos\theta|0\rangle + \sin\theta|1\rangle)$$

$$(|\theta\rangle\theta|\otimes R_y)(|\theta\rangle\otimes|0\rangle) = |\theta\rangle\otimes R_y|0\rangle$$

$$R_y = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

$$R_y|0\rangle = \cos\theta|0\rangle + \sin\theta|1\rangle$$

Let  $\theta = \theta \cdot \theta_1 \theta_2 \theta_3 \dots \theta_f$  be decimal representation

To transform  $|\theta\rangle \rightarrow \sin |\theta|$ , we use the following idea.

If 
$$m{ heta} = \sum_{i=1}^f m{ heta}_i * 2^{-i}$$
 Then  $e^{m{ heta}} = \prod_{i=1}^f e^{m{ heta}_i * 2^{-i}}$ 

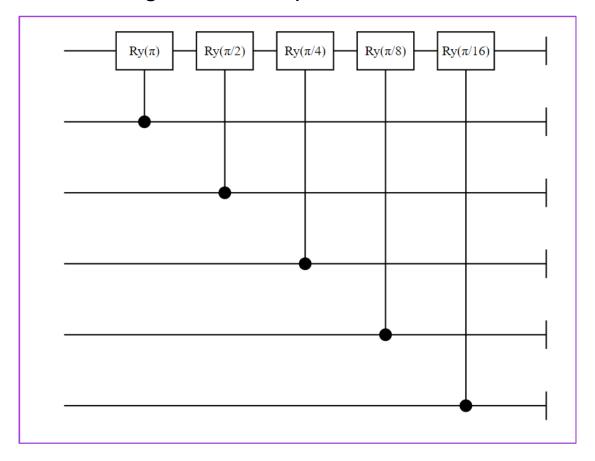
WKT  $|sin\theta| < 1$ .

We normalize theta using a C such that  $sin(C*\theta) \approx C*\theta$ 



## Sine of a Bitstring

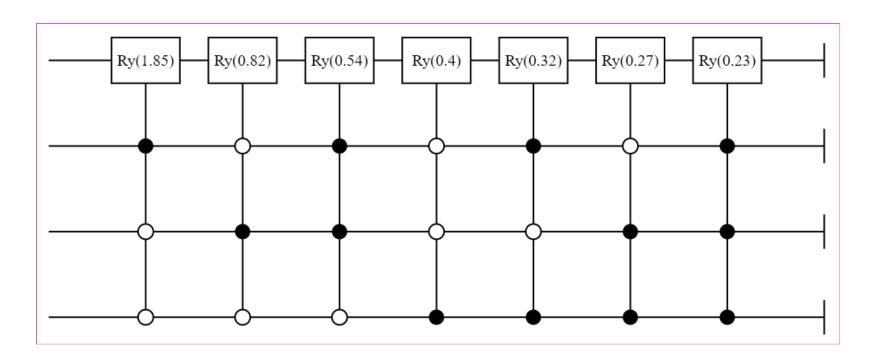
Controlled Rotation: Eigenvalues a priori.





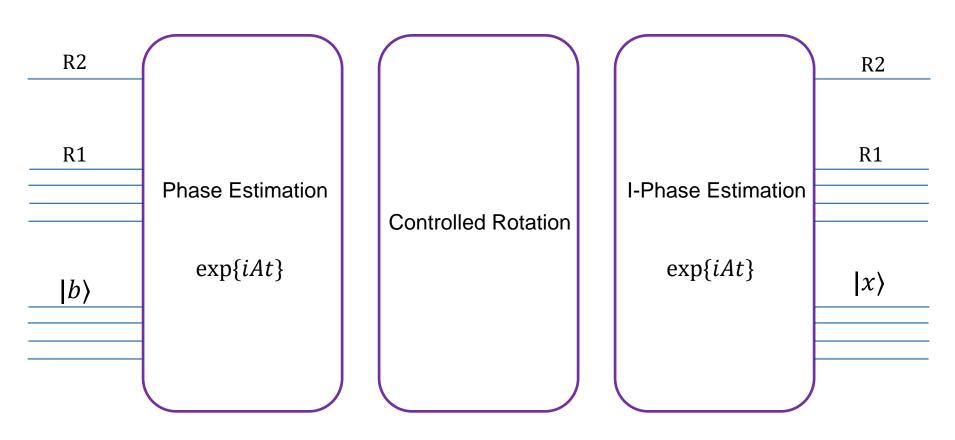
# Sine of a Bitstring

#### Controlled Rotation : More robust





# Algorithm – Components



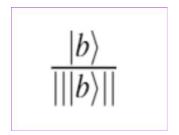
A – Hermitian , sparse , well conditioned.



# Brass Tacks – (i)

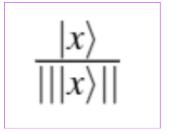
Encode  $|b\rangle$  using Amplitude encoding.

• Amplitude embedding: roof(log(2,length(b))



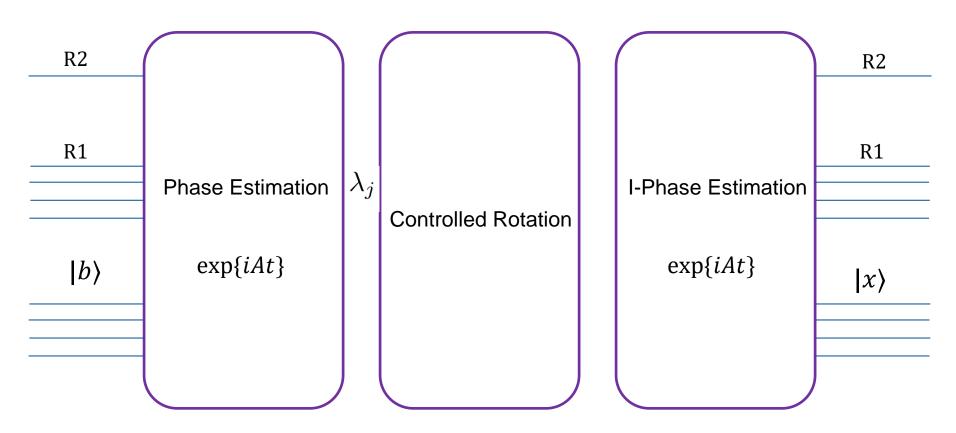
Caveat : Solution is also amplitude embedded

`





# Algorithm – Components



A – Hermitian, sparse, well conditioned.



# Brass Tacks – (v)

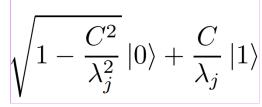
#### **Controlled Rotation**

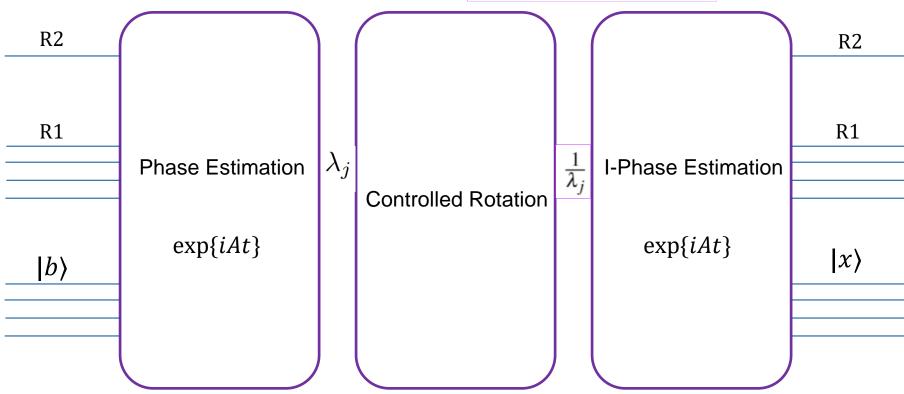
- $|\lambda\rangle$  is housed in Reg. A in a bitwise manner.
- Add a ancilla qubit R2.
- Prior to controlled rotation we reciprocate. i.e, transform  $|\lambda\rangle \rightarrow \left|\frac{1}{\lambda}\right\rangle$
- If  $|\lambda\rangle$  is not normalized, choose C such that  $C < \min \lambda_i$ .
- State of R2 after CRot :

$$\sqrt{1 - \frac{C^2}{\lambda_j^2}} |0\rangle + \frac{C}{\lambda_j} |1\rangle$$



### Algorithm – Components





A – Hermitian, sparse, well conditioned.



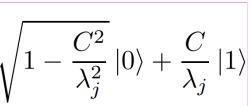
# Brass Tacks – (vi)

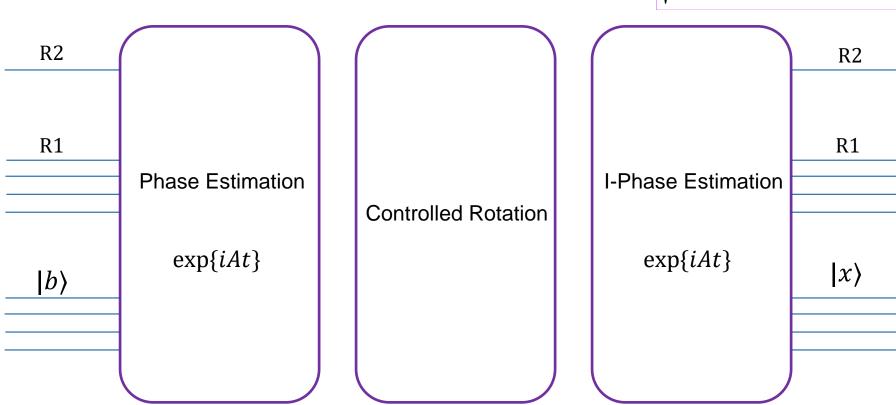
#### Un-compute PE

- Throw away garbage in R1
- Use Inverse Phase estimation
- Runtime Significance
- Error Significance



## Algorithm – Components





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### Brass Tacks – (vii)

#### Measure ancilla qubit

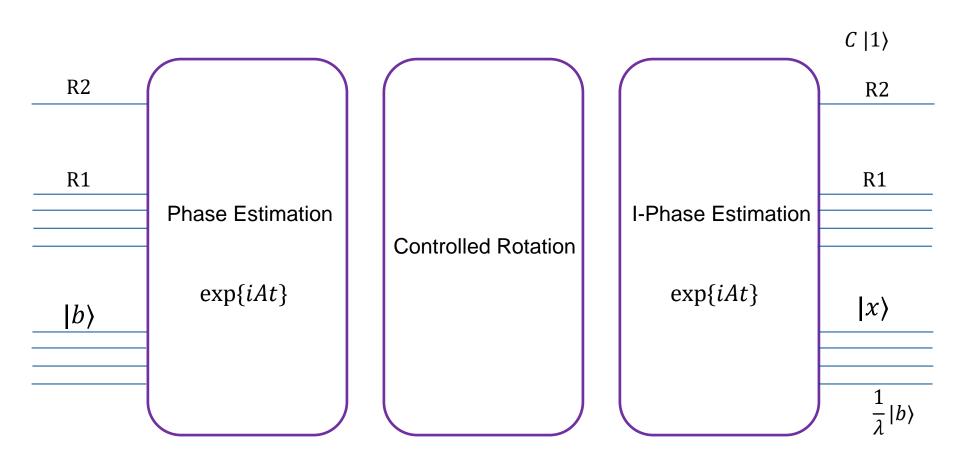
- Most boring yet crucial part of the algorithm
- Measure ancilla
- If we get 1 => success, solution at the base register
- Else Back to square one

What could we measure with respect to?

- Projectors
- Compute Expectation



# Algorithm – Components



A – Hermitian , sparse , well conditioned.



#### Caveats

- norm in  $b \rightarrow |b\rangle$
- t in Quantum Simulation
- $n_A$  number of qubits in Register A.
- C in Controlled Rotation



# Difficulties in Actual Q.Computation

- Preparation of |b⟩ efficiently
- Availability of qubits for a practical problem.
  - If  $\lambda \approx \epsilon$ , we require as many as 50 qubits for Register A alone for  $A \in \mathbb{R}^{32x32}$ .
- The final measurement is a hit and miss. Need to perform several experiments.
- Not realizable! However (see ref.)



#### Hamiltonian Simulation

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# Exceptions to the Rule

Non Hermitian Matrices

Non Square Matrices

Dense Matrices

Non Invertible Matrices



#### Classical Simulation Frameworks

- Yao.jl
- Quaintum.jl
- Qiskit Terra



## Old Faithful – Poisson Equation

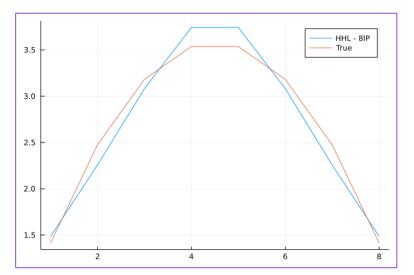
- $\Delta \phi = f$  ;  $x \in [a, b]$
- $\delta\Omega:\phi(a)=0$ ;  $\phi(b)=0$
- f = ones(N)
- Finite Difference Method. HHL vs Julia "\"

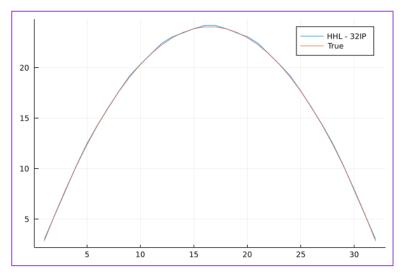
$$\begin{bmatrix} -2 & 1 & 0 & 0 & 0 & \cdots & \cdots & 0 \\ 1 & -2 & 1 & 0 & 0 & \cdots & \cdots & 0 \\ 0 & 1 & -2 & 1 & 0 & \cdots & \cdots & 0 \\ 0 & 0 & 1 & -2 & 1 & \ddots & \cdots & 0 \\ 0 & 0 & 0 & 1 & -2 & \ddots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \ddots & \ddots & \ddots & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & -2 \\ \end{bmatrix} \times \begin{bmatrix} \Phi_1 \\ \Phi_2 \\ \Phi_3 \\ \Phi_4 \\ \Phi_5 \\ \vdots \\ \Phi_{N-1} \\ \Phi_N \end{bmatrix} = \begin{bmatrix} h^2 f_1 - \Phi_a \\ h^2 f_2 \\ h^2 f_3 \\ h^2 f_4 \\ h^2 f_5 \\ \vdots \\ h^2 f_{N-1} \\ h^2 f_N - \Phi_b \end{bmatrix}$$

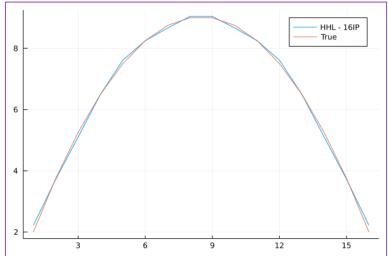
# Results $n_{\lambda} = 12$



1D



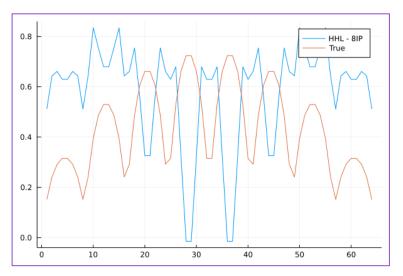


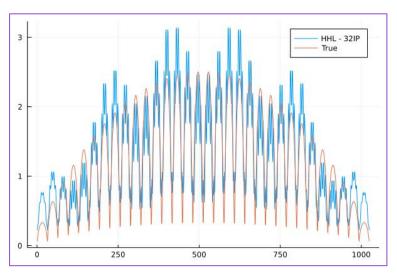


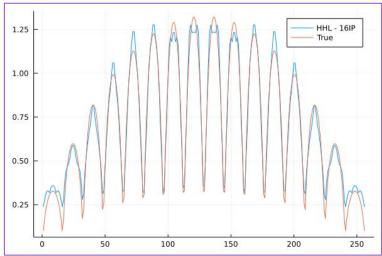
# Results $n_{\lambda} = 12$



#### 2D



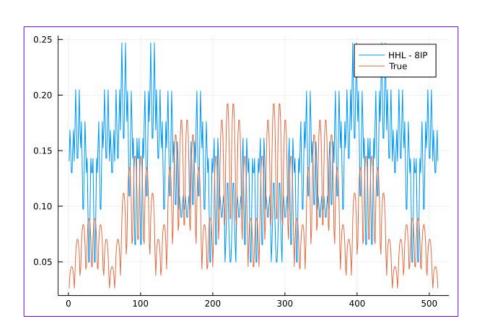






# Results $n_{\lambda} = 12$

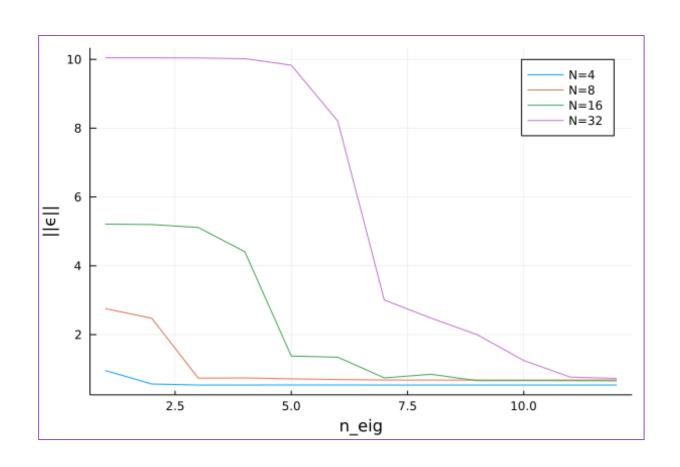
3D



#### Results



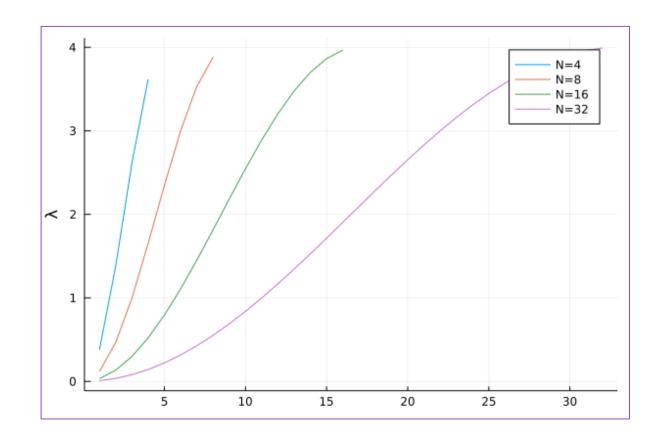
#### Error Analysis – 1D



#### Results



Error Analysis – 1D – Inference : Relation to eigenvalues





#### More on HHL

- Baseline Implementation
- Rahul Manavalan / Solving the Heat Equation with the HHL Algorithm GitLab (Irz.de)
- Elaborations into the claims made in the HHL paper.
  - The Why of HHL.
  - <a href="https://dynamic-queries.medium.com/chapter-1-1-solutions-to-linear-systems-and-all-that-69dc73e02739">https://dynamic-queries.medium.com/chapter-1-1-solutions-to-linear-systems-and-all-that-69dc73e02739</a>



#### References

- Read the fine print (nature.com) : Prelude
- <u>1301.2340.pdf (arxiv.org)</u>: Box to Box Solution
- 1302.1210.pdf (arxiv.org): Experimental Implementation albeit for a small system.
- <u>qiskit-terra/exact\_reciprocal.py at a9289c085036002fa864b30edcd0b8ce066ea858 Qiskit/qiskit-terra (github.com)</u>: Exact reciprocal.
- <u>1802.08227.pdf (arxiv.org)</u>: Primer
- <u>arXiv:0811.3171v3 [quant-ph] 30 Sep 2009</u>: Base Paper: Hard Read
- 1110.2232.pdf (arxiv.org) : Saved hours , quite comprehensive