

Quantum Algorithms

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Exercises: Arjan Cornelissen

9th of September



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- Goal: Get an understanding of how quantum algorithms work and how they do not work.
- We will focus on the techniques and insight.
(The not so famous, but really cool stuff)

- Basics - Quantum states, gates and circuits.

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 - Shors factoring algorithm.
 - HHL for systems of linear equations.
 - Gradient computation.

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 - HHL for systems of linear equations.
 - Gradient computation.
- A quick quiz!

Basics

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- For a qubit:

$$|0\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad |1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

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- Z adds a -1 in front of $|1\rangle$.
- H changes $|0\rangle$ and $|1\rangle$ into $\frac{|0\rangle+|1\rangle}{\sqrt{2}}$ and $\frac{|0\rangle-|1\rangle}{\sqrt{2}}$.

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Claim: H is a basis transform

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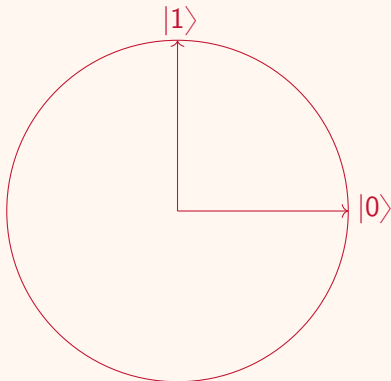
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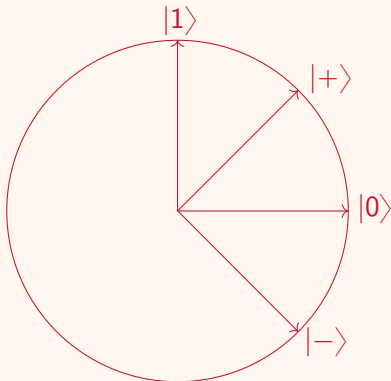


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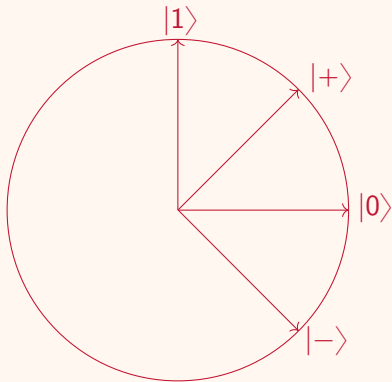
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This is a X gate! Z is just X in the $\{|+\rangle, |-\rangle\}$ basis (and vice versa).

Reflections

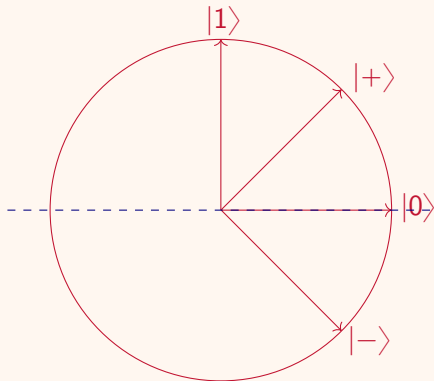
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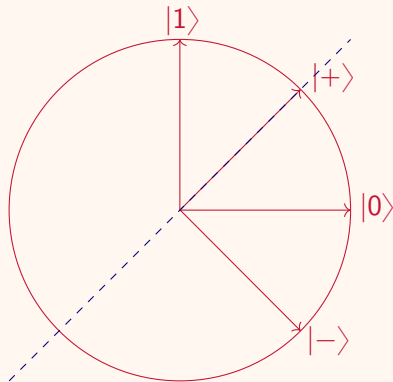


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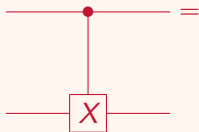
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- We often want a controlled phase oracle:

$$O_{x,\pm} |\hat{i}\rangle |0\rangle = |\hat{i}\rangle |0\rangle, \quad O_{x,\pm} |\hat{i}\rangle |1\rangle = (-1)^{x_i} |\hat{i}\rangle |1\rangle$$

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How would you convert between the two types of oracle?

Amplitude amplification

Randomized algorithms

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- Repeat

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times.

Example: prime finding

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$\Rightarrow \mathcal{O}(n)$ repetitions suffice.

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- If we just measure then the success probability is $p = |\alpha_G|^2$.

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$$\langle G|B\rangle \langle 1|0\rangle = \langle G|B\rangle \cdot 0 = 0$$

- So these two states are orthogonal!

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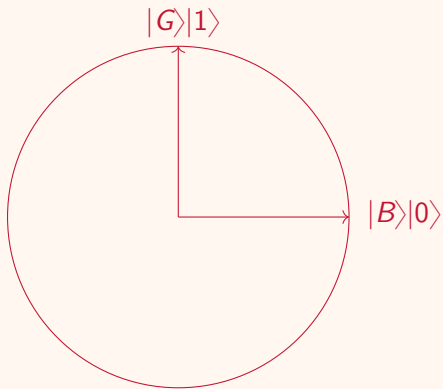
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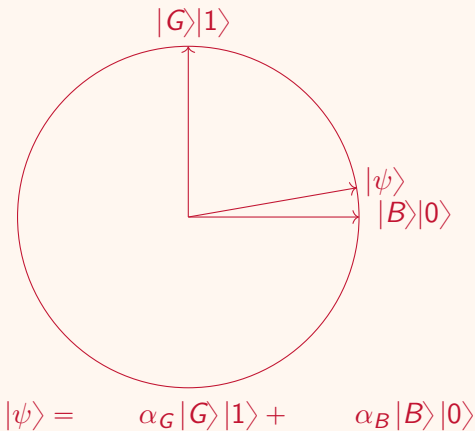
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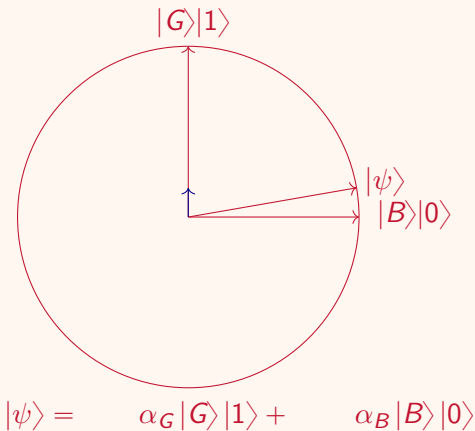
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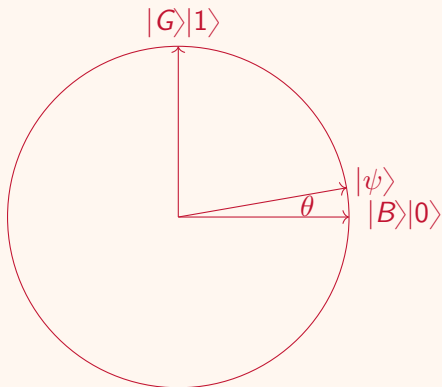
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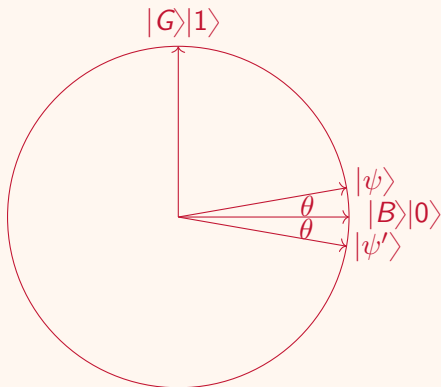


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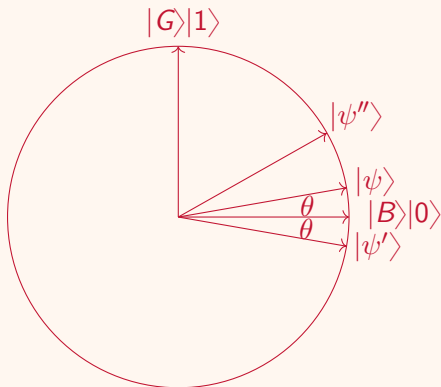
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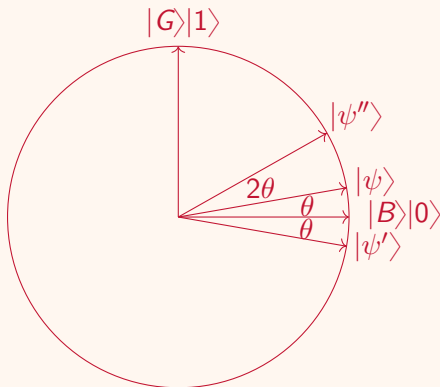
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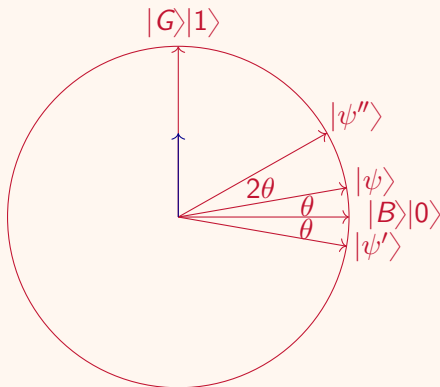


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- Nice, but can we actually implement these reflections?

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Use that $|\psi\rangle = U|0\rangle$:

1. Apply U^{-1} to map $|\psi\rangle$ to $|0\rangle$.
2. Reflect through $|0\rangle$.
3. Apply U to map $|0\rangle$ back to $|\psi\rangle$.

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With $\mathcal{O}(\sqrt{\frac{n}{k}})$ iterations we are done!

Phase estimation

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Eigenvectors and eigenvalues

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In general for a unitary U and eigenvector $|\psi\rangle$ of U :

$$U|\psi\rangle = e^{i2\pi\theta} |\psi\rangle$$

Remember: $e^{i\theta}$ is called a phase.

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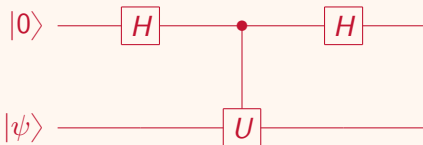
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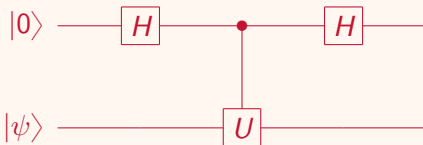
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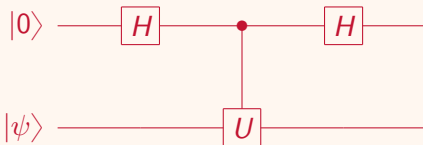
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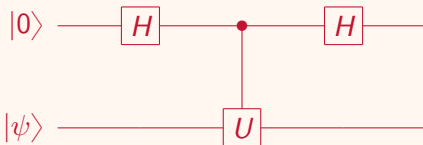


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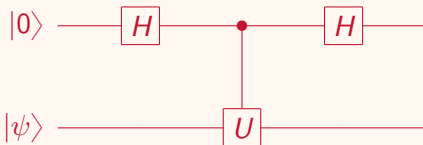


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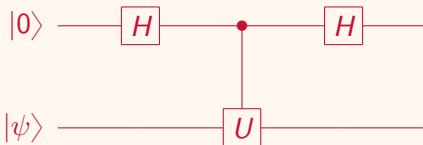


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- If this is only one bit ($\theta = 0.0$ or $\theta = 0.1 = \frac{1}{2}$) then we know what to do.

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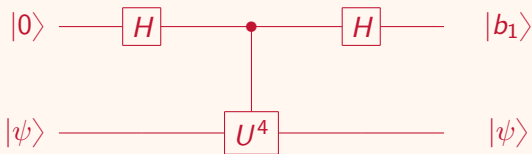
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- Simply use our previous circuit to get the last bit:



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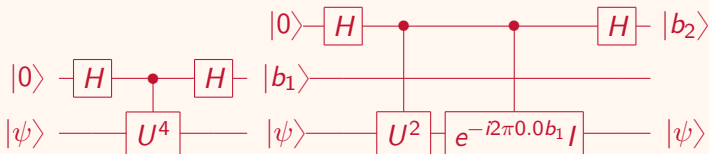
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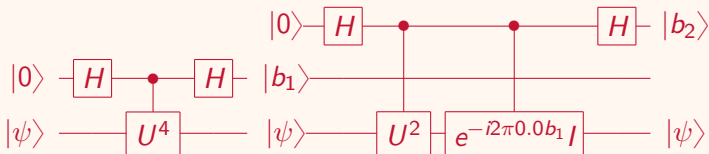
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- If $b_1 = 0$ then we can just use our algorithm. But what if $b_1 = 1$?
- Solution: calculate b_1 and then make a change to U if necessary:



Two bit phase estimation

Let $U|\psi\rangle = e^{i2\pi\theta}|\psi\rangle$ for $\theta = 0.b_3b_2b_1$, can we calculate b_2 as well?

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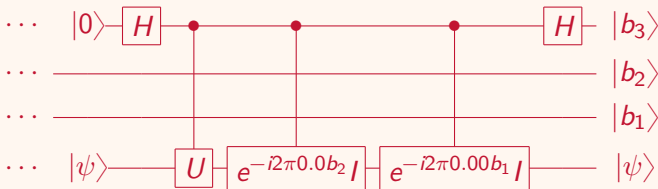
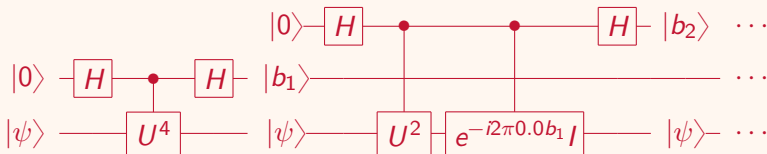
- This works because $U^2 e^{-i2\pi 0.b_1}$ adds a phase with $0.b_2b_1 - 0.0b_1 = 0.b_2$.

Three bit phase estimation

Let $U|\psi\rangle = e^{i2\pi\theta}|\psi\rangle$ for $\theta = 0.b_3b_2b_1$, can we calculate all the bits?

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Let $U|\psi\rangle = e^{i2\pi\theta}|\psi\rangle$ for $\theta = 0.b_3b_2b_1$, can we calculate all the bits?
 Yes! just keep going:

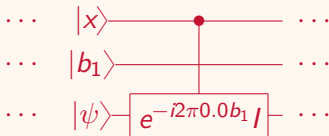


Clean up

Can we clean up this circuit?

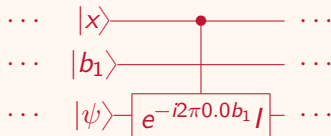
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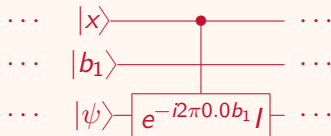
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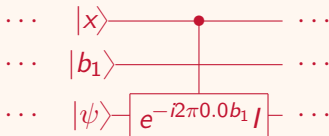


This applies a phase of $e^{-i2\pi 0.0 b_1}$ to the state if $x = 1$.

\Leftrightarrow Applies a phase of $e^{-i\pi/2}$ to the state if $x = 1$ and $b_1 = 1$.

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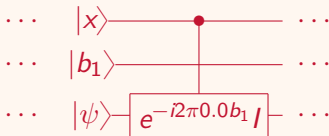
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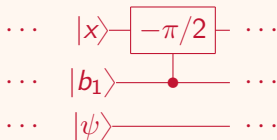
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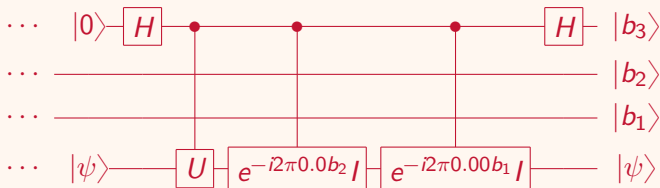
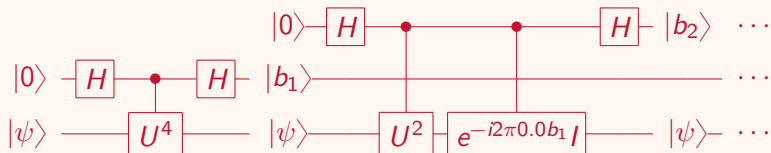
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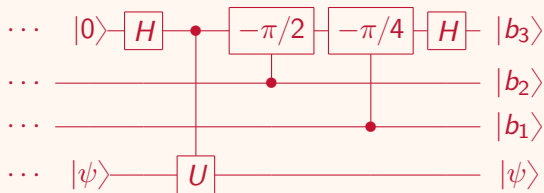
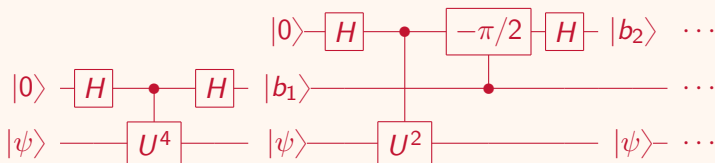
The inverse quantum Fourier transform

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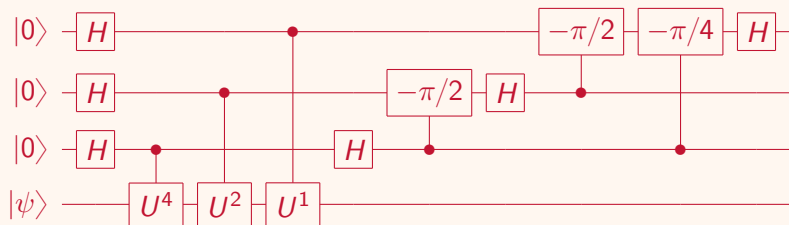
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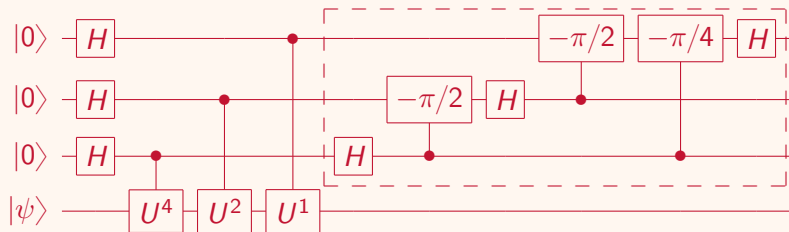
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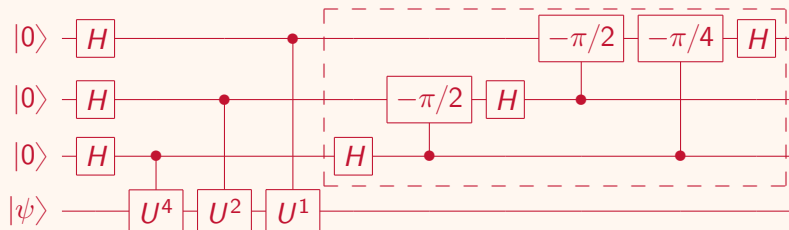
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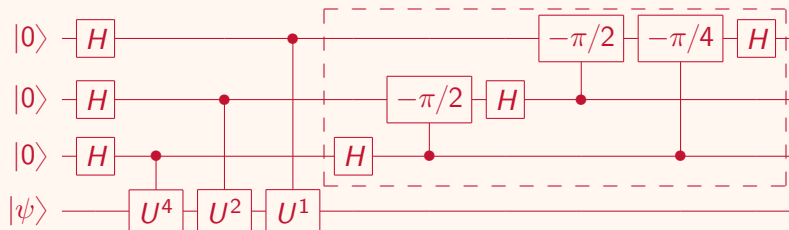


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In total we need to apply U^k for $k \sim \frac{1}{\varepsilon}$

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- Measuring the first register gives us a random θ_i , what happens to the second register?

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- For example for search: how many solutions are there?
- Too many iterations decreases the success probability again!

A closer look at the amplitude amplification iterate

In the basis $\{|G\rangle|1\rangle, |B\rangle|0\rangle\}$ the iterate is a rotation:

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The eigenvectors also lie in the 2-dimensional subspace.

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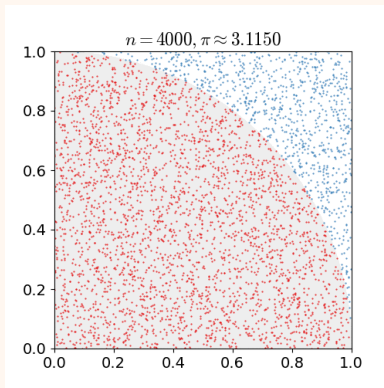
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- Both tell us θ and hence the probability up to error ε in

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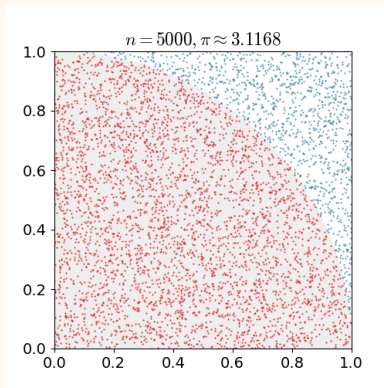
Example: Speeding up Monte Carlo methods

In classical computing a Monte Carlo method samples a lot of times to estimate a probability.



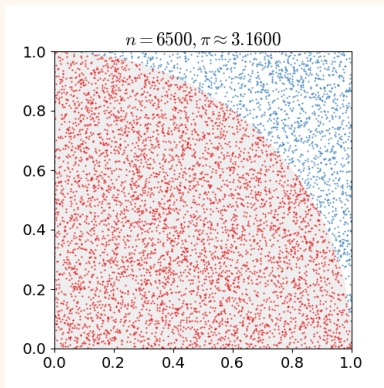
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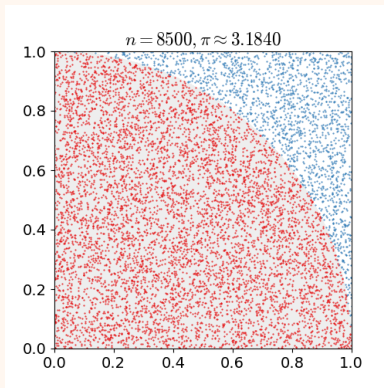
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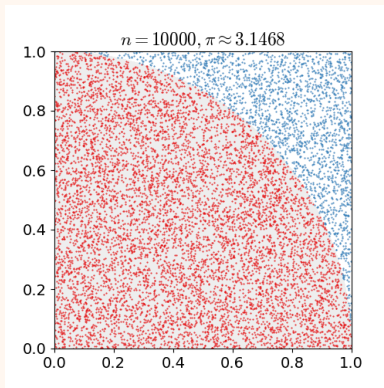
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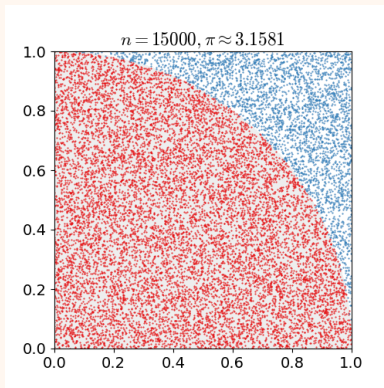
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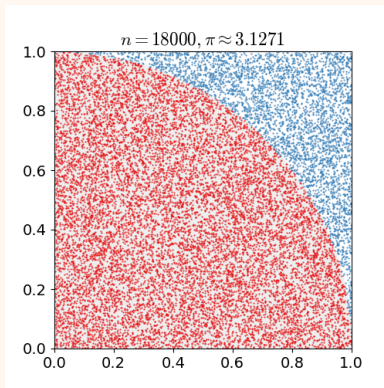
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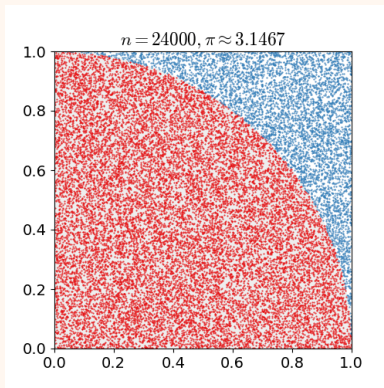
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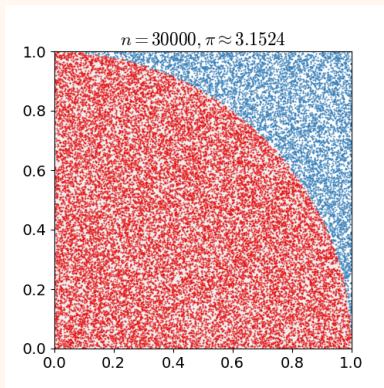
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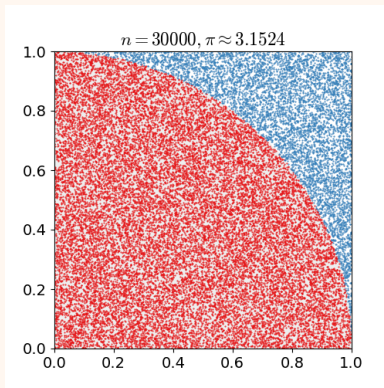
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A classical algorithm requires $\sim \frac{1}{\epsilon^2}$ samples to estimate, quantum $\frac{1}{\epsilon}$.

Overview of some other algorithms

Some other algorithms

There are many more quantum algorithms, often using the disced techniques. We will quickly discuss some other important algorithms:

- Shor's algorithm & factoring integers
- HHL for “solving” linear systems
- Faster gradient computation

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- Small detail: we can implement U^k faster than k repetition, why?

HHL algorithm for systems of linear equations

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Quantum linear system solving (HHL '09)

Given s -sparse matrix $A \in \mathbb{R}^{N \times N}$ and way to prepare $|b\rangle \in \mathbb{R}^N$, prepare a state $|x\rangle$ such that

$$A|x\rangle \propto |b\rangle$$

Takes only $\tilde{O}(s\kappa)$ time, but is not the same problem!

Gradient computation (Jordan '05)

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Given black-box access to a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$, estimate the gradient at 0.

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- Use phase estimation on all n coordinates at the same time!

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5. Quantum computers can solve linear systems of equations faster.
6. Quantum computers are wierd but cool.

That was it!