

# Grover's Algorithm

Hans Hohenfeld – June, 19<sup>th</sup> 2023

**QC for Software Engineers**

Universität Bremen & DFKI RIC

Robert-Hooke-Straße 1

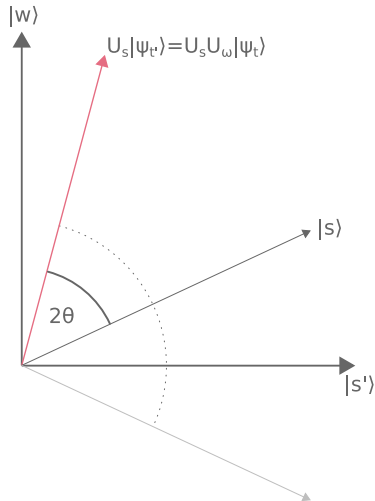
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Deutsches  
Forschungszentrum  
für Künstliche  
Intelligenz GmbH



# Searching and Finding...

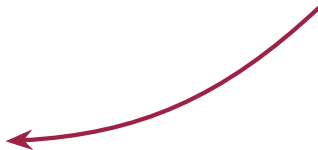
No.	Name	Phone No.
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⋮	⋮	

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What is **Mike**'s phone number?

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1	Alice	0123/45678
2	Anna	9876/654321
3	Arthur	999/654456
⋮	⋮	
<b>725</b>	<b>Mike</b>	<b>555/789789</b>
⋮	⋮	

What is **Mike**'s phone number?



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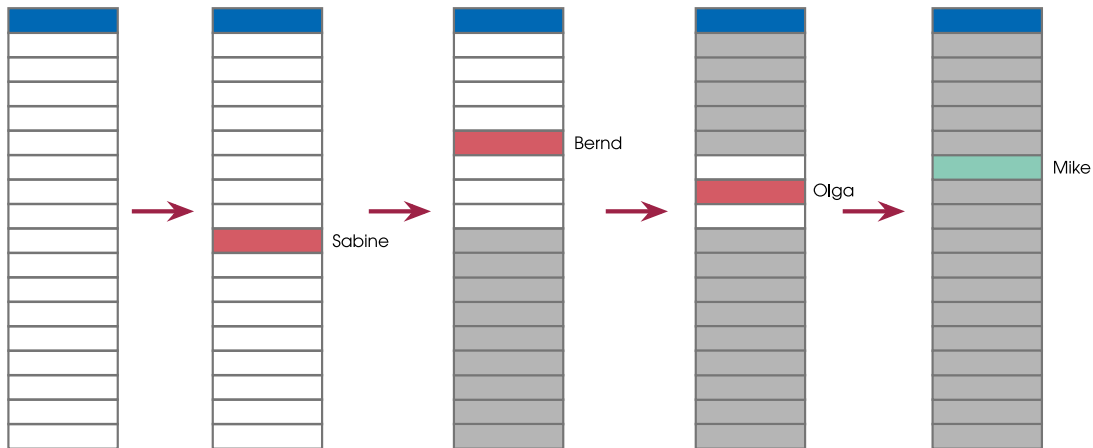
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- The list is **sorted**.
- We know the list's **length**.
- We can **read** every individual line.

# Assumptions

- **Order**: e. g. **John**  $\leq$  **Sabine**.
- The list is **sorted**.
- We know the list's **length**.
- We can **read** every individual line.
- Nothing else.

How do we find **Mike**'s phone number?

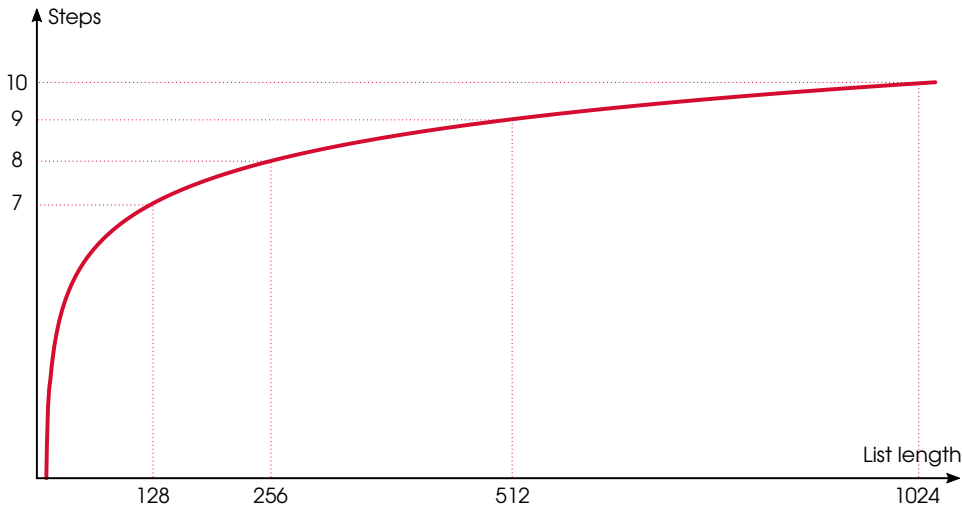
# Binary Search



# Time Complexity

Informally **time complexity** is a measure for how the computational effort in terms of time grows, relative to the problem size.

# Time complexity of Binary Search



# Time Complexity

More formally:



# Time Complexity

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The computational effort to find an element in a sorted list of length  $n$  does not grow faster than  $\log(n)$  (up to constant factors).

Or even more formally...

# Time Complexity

## Binary Search: Time Complexity

Let  $T_{BS} : \mathbb{N} \rightarrow \mathbb{N}$  be the **time step** function of Binary Search. It holds:

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We write:

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We say: Binary Search has (time) complexity  $O(\log(n))$ .

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**Big-O**

**Name**

**Problem**

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Big-O	Name	Problem
$O(1)$	Constant	Median of a sorted list



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$O(c^n)$	Exponential	Chess, Go, ...

We change the problem a bit...

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Who has the phone number  
**1234/999888?**



This type of problem is called  
**Unstructured Search.**

For reasons that become more clear later, we divide the problem  
between two instances.

# The Seeker

- Has access to the list ...

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- Has access to the list ...
- Can read arbitrary entries in constant time ...
- Decides which entry to read next ...

# The Oracle

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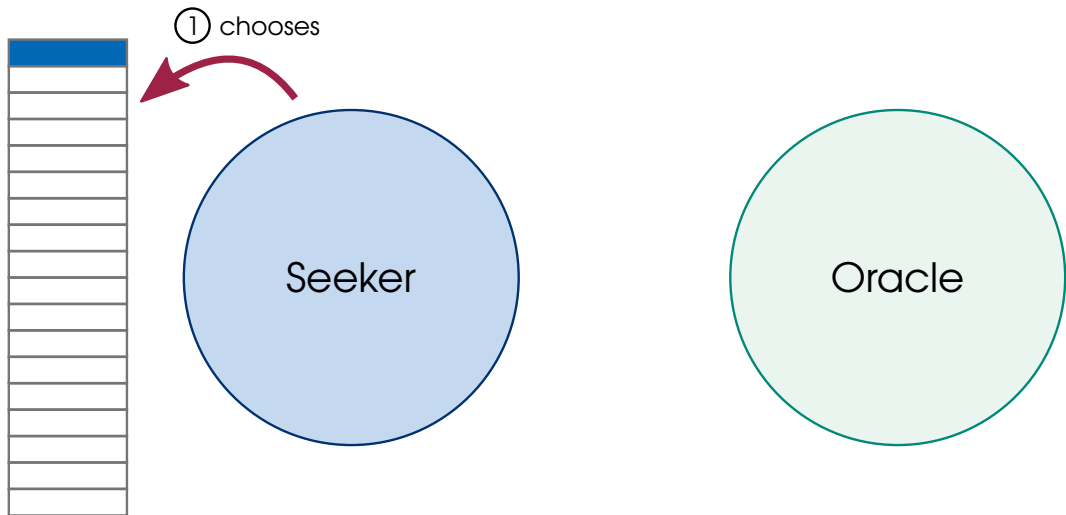
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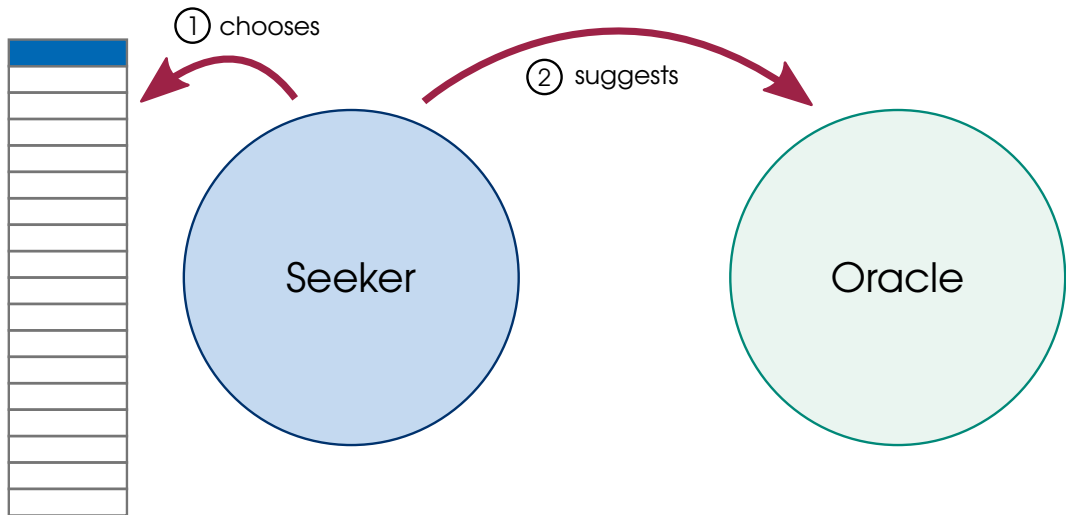
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- Takes constant time for an answer ...

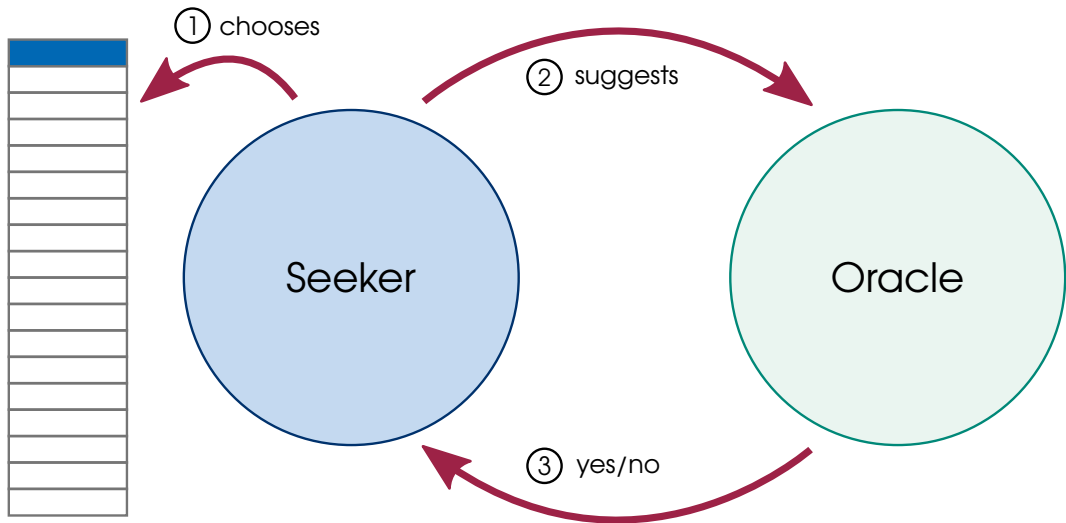


# The Oracle

- Knows the answer ...
- Gets entries suggested by the seeker and answers with **yes/no** ...
- Takes constant time for an answer ...
- Is never wrong ...







How often does the Seeker need to query the Oracle?

# Linear Search

- The only reasonable strategy is to query the Oracle for one entry after the other, starting from the first.

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- For a list of length  $n$  this requires  $n$  queries in the worst case.
- On average, the search is successful after  $\frac{n}{2}$  queries.
- **Important:** The Oracle does not change anything with regards to complexity.

# Linear Search

**Lineare Suche** has (query/time) complexity of  $O(n)$ .

# Questions?

# Searching and Finding ...

# Searching and Finding ... on Quantum Computers

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- Formulated 1996 by Lov K. Grover [4].
- The second quantum algorithm of practical relevance after Shor's factoring algorithm in 1994 [6].
- Implemented for 3 qubits on a programmable quantum computer in 2017 [3].

# Problem Statement

Given  $n$  qubits with  $N = 2^n$  basis states

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$$|0 \dots 0\rangle = |0\rangle$$

$$|0 \dots 1\rangle = |1\rangle$$

...

$$|1 \dots 1\rangle = |N - 1\rangle$$

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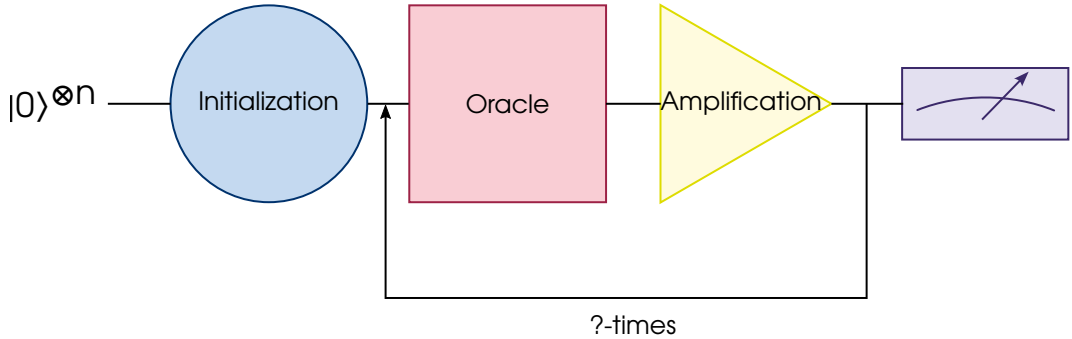
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$$|\omega\rangle$$

Let's first look at a schematic of the algorithm...



**Oracle** and **Amplitude Amplification** are executed several times



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How often will be the result of the following discussion.

First the details of each component...

# Initialization

At the start, we know nothing about the state we are searching.

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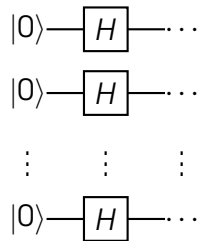
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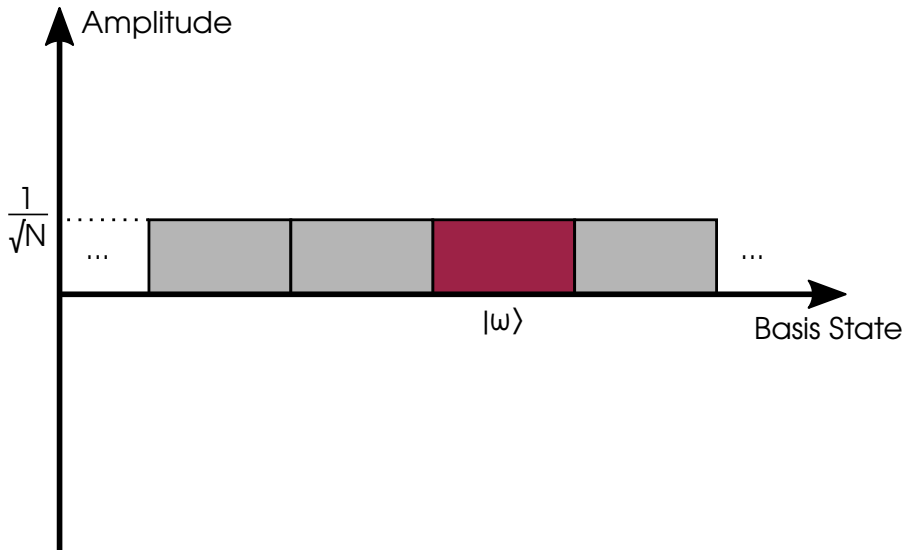
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After initializing the system, we have an equal probability of  $\frac{1}{N}$  to measure either of the  $N$  basis states.

# Oracle

The oracle  $U_\omega$  marks the state  $|\omega\rangle$  we are searching for. After applying the oracle, we want for each basis state:

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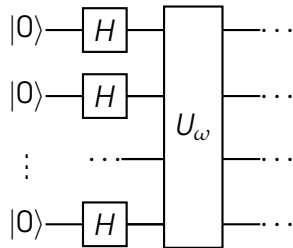
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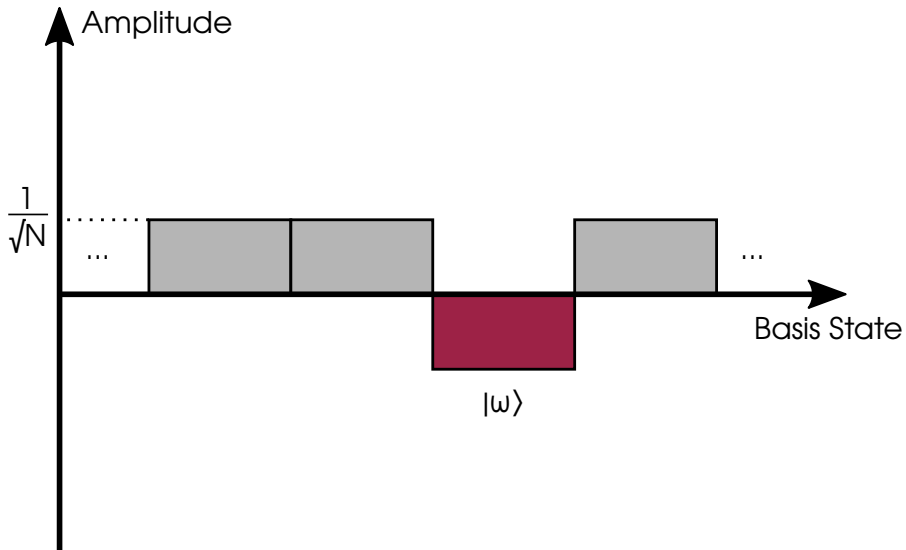
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Probabilities of measuring either basis state remain unchanged by this.  
The oracle only inverted the phase of the state we are searching for!

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- $U_\omega$  is a diagonal matrix.
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- In the line with index corresponding to the searched basis state, the entry is  $-1$ .

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Let  $|N - 1\rangle$  be the searched basis state. The corresponding oracle is then:

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$$U_{\omega} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & -1 \end{pmatrix}$$

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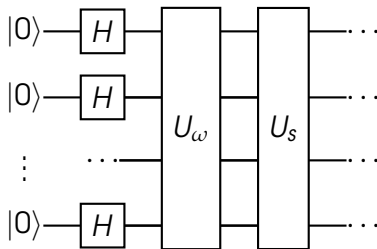
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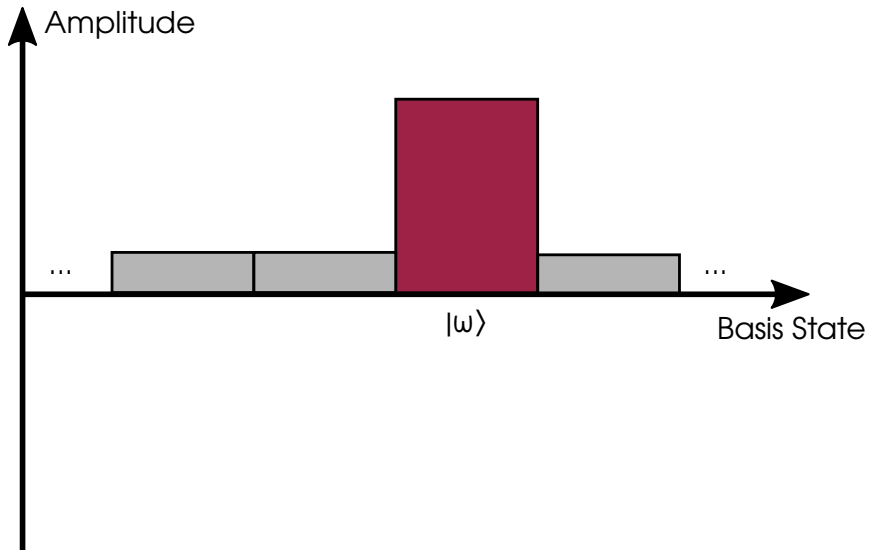


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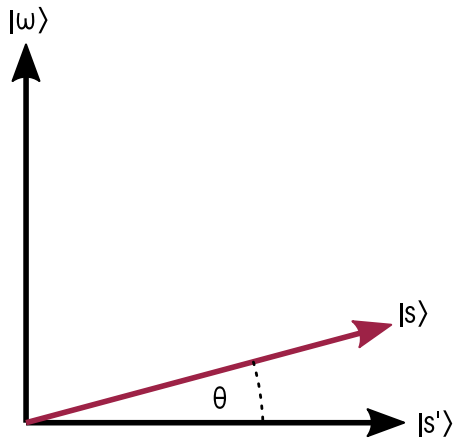
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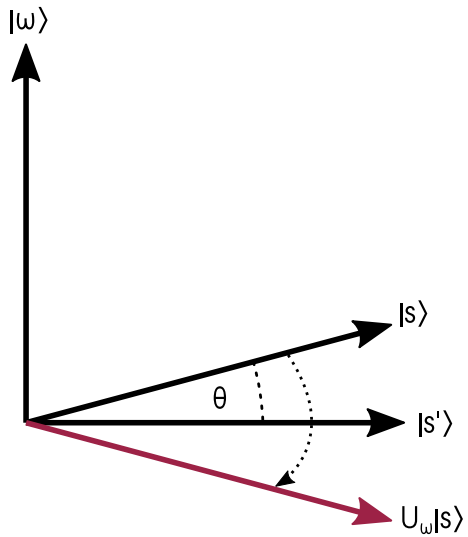
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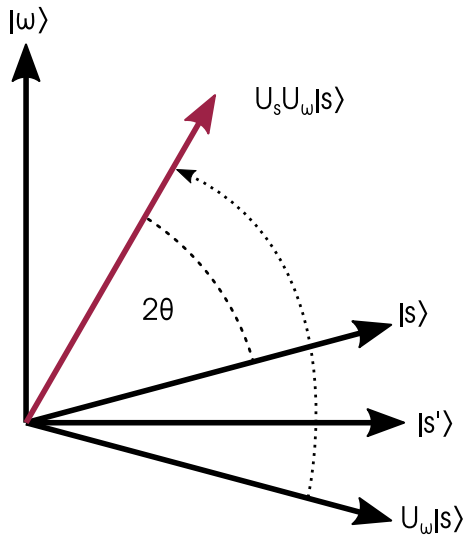




A geometric view on the algorithm...







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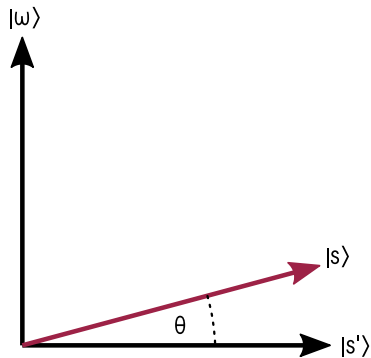
$$U_s = 2 |s\rangle \langle s| - I.$$

# Grover's Algorithm: Complexity

How often do we have to apply the oracle and amplitude amplification?  
Looking at  $\theta$  gives an intuition.

# Grover's Algorithm: Complexity

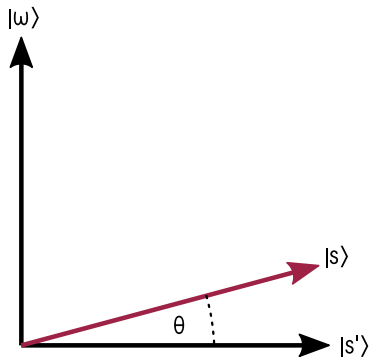
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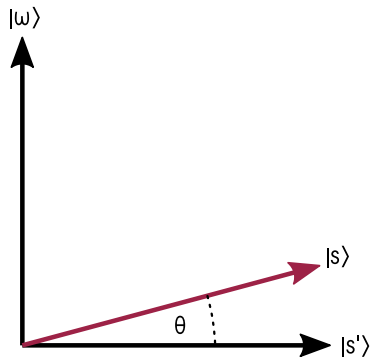
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We have:

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with

$$\theta = \arcsin \langle s|\omega\rangle = \frac{1}{\sqrt{N}}$$



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Hence, after approximately  $\sqrt{N}$  steps, the probability of measuring the correct state is very close to 100%.



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A quantum computer can solve this problem **quadratically faster**.

# Questions?

**Thank you.**

# Images

- Slide 29: Based on [3].
- Slides 34, 37, 42, 44-46 and title slide: Based on [1].

# References

- [1] Amira Abbas u. a. *Learn Quantum Computation Using Qiskit*. 2020. URL: <http://community.qiskit.org/textbook>.
- [2] Charles H. Bennett u. a. „Strengths and Weaknesses of Quantum Computing“. In: *SIAM Journal on Computing* 26.5 (1997), S. 1510–1523. DOI: 10.1137/S0097539796300933.
- [3] Caroline Figgatt u. a. „Complete 3-qubit Grover search on a programmable quantum computer“. In: *Nature communications* 8.1 (2017), S. 1–9. DOI: 10.1038/s41467-017-01904-7.



# References (cont.)

- [4] [Lov K. Grover](#). „A Fast Quantum Mechanical Algorithm for Database Search“. In: *Proceedings of the Twenty-Eighth Annual ACM Symposium on Theory of Computing*. STOC '96. Philadelphia, Pennsylvania, USA: Association for Computing Machinery, 1996, S. 212–219. DOI: 10.1145/237814.237866.
- [5] [Michael A. Nielsen und Isaac L. Chuang](#). *Quantum Computation and Quantum Information: 10th Anniversary Edition*. Cambridge Univ Press, 9. Dez. 2010.
- [6] [P.W. Shor](#). „Algorithms for quantum computation: discrete logarithms and factoring“. In: *Proceedings 35th Annual Symposium on Foundations of Computer Science*. 1994, S. 124–134. DOI: 10.1109/SFCS.1994.365700.