

Unstructured Search

via Grover's Algorithm

Fernando Granha Jeronimo

Unstructured Search

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Any classical computer requires $\Omega(2^n)$ queries!

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Can a quantum computer beat brute-force search?

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- First, why should we care about this problem at all?

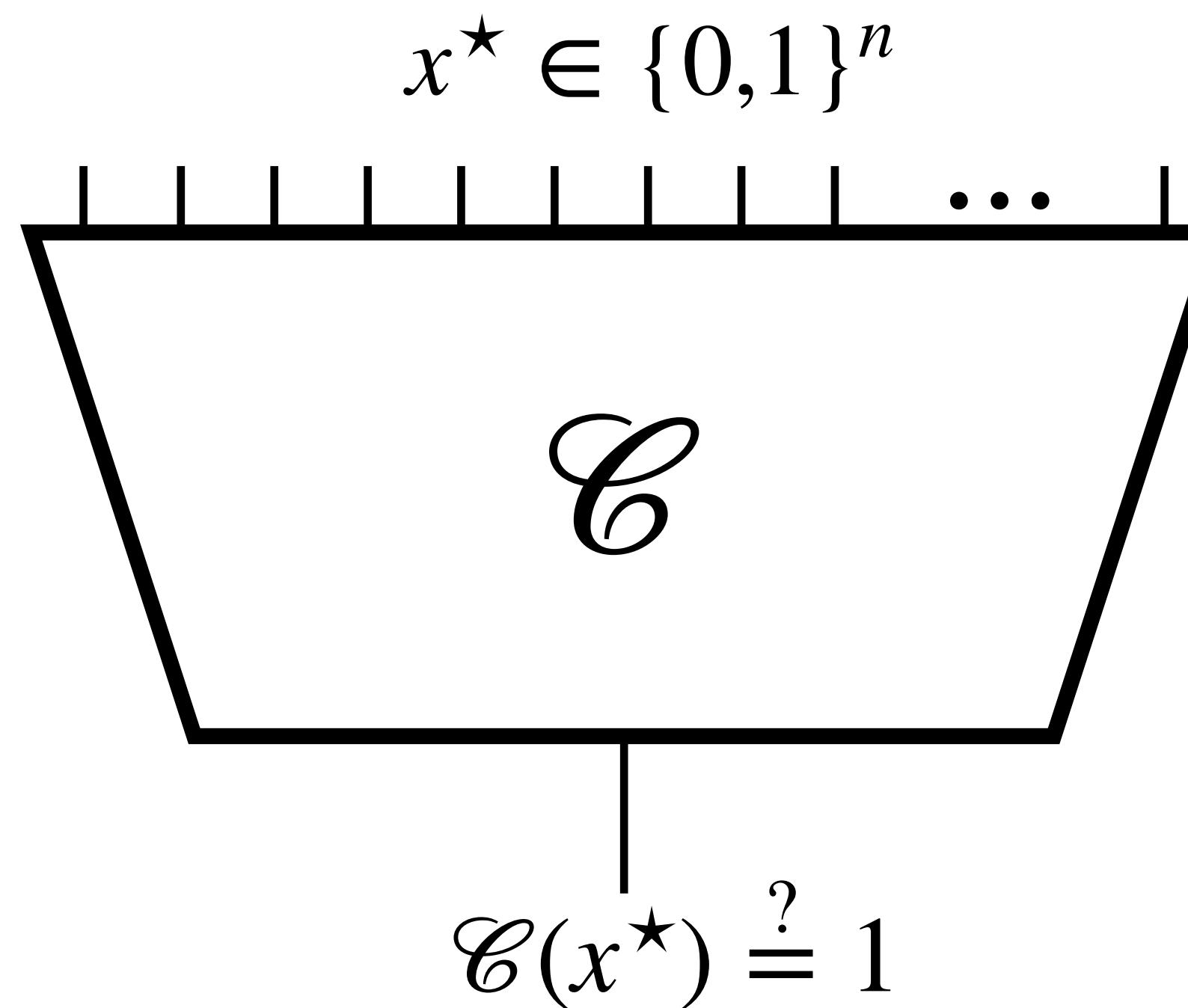
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In particular, f can encode the evaluation of a classical circuit \mathcal{C}

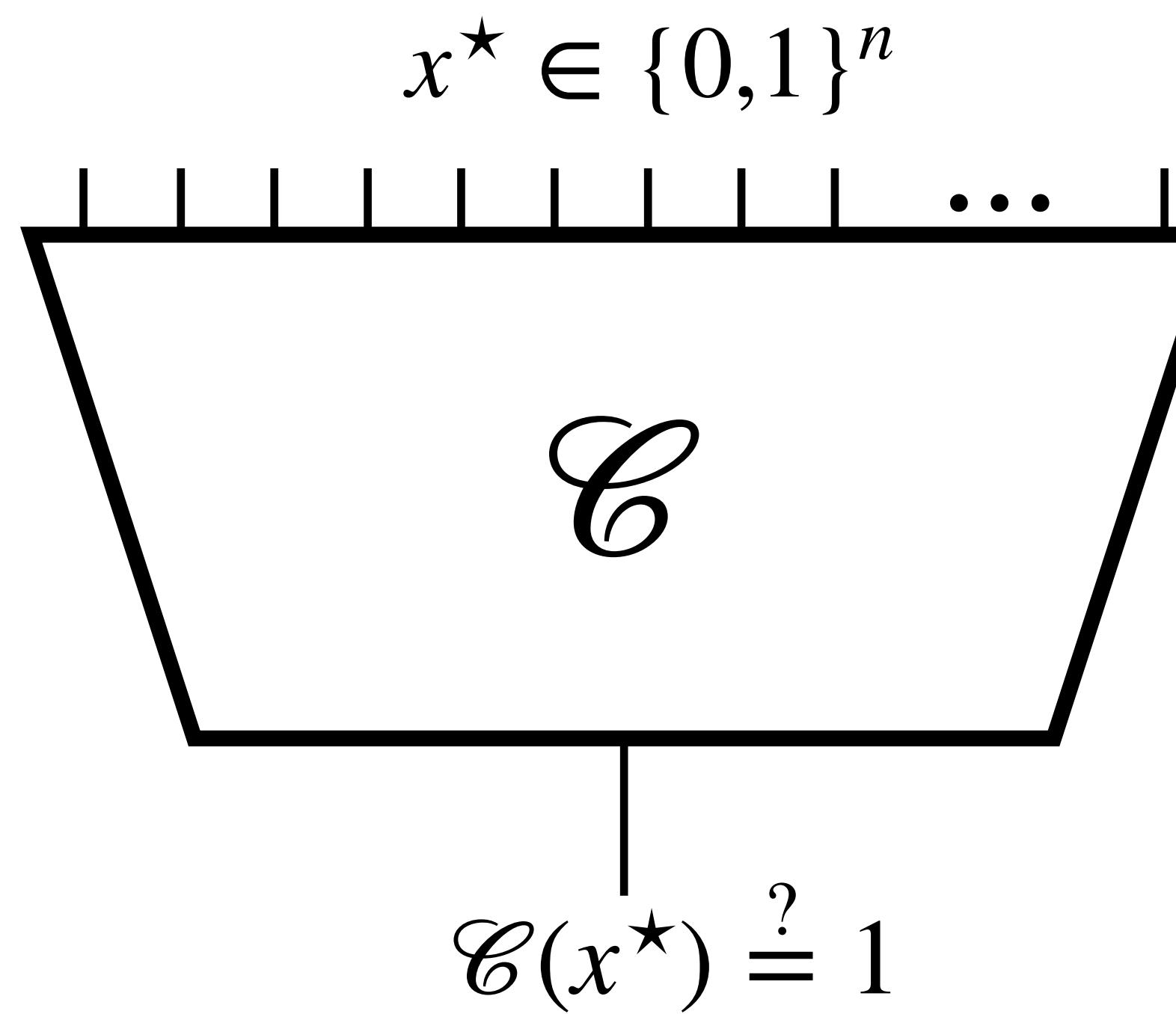
Unstructured Search

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Determining if \mathcal{C} has a satisfying input is NP-complete

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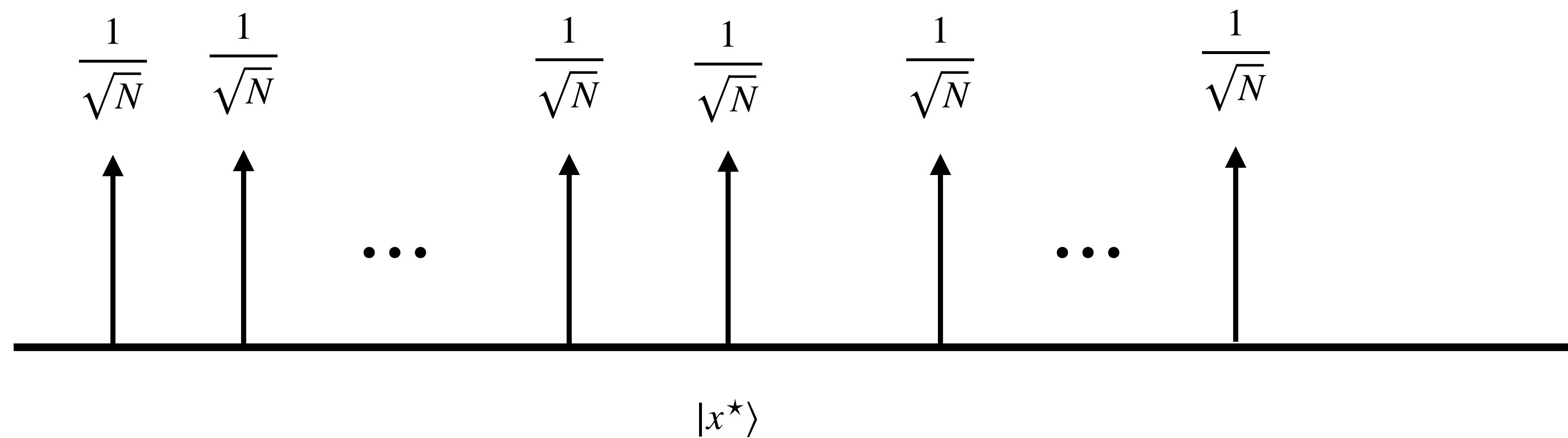
(Phase Oracle)



$$O_f|x\rangle|0\rangle = |x\rangle|f(x)\rangle$$

Some Intuition

$$N = 2^n$$

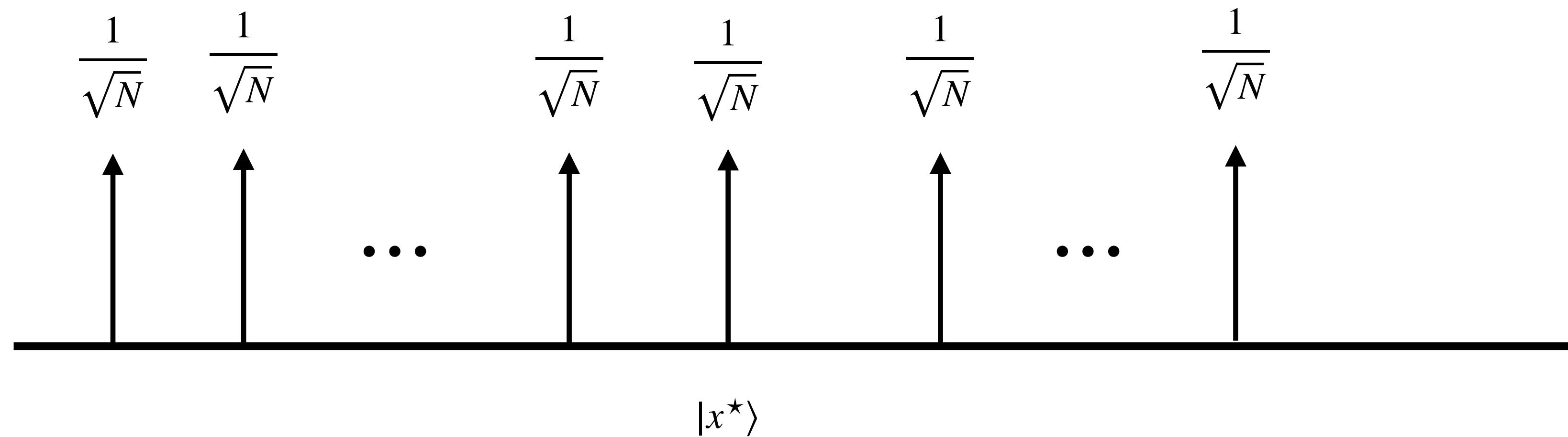


$$|u\rangle = \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} |x\rangle$$

Some Intuition

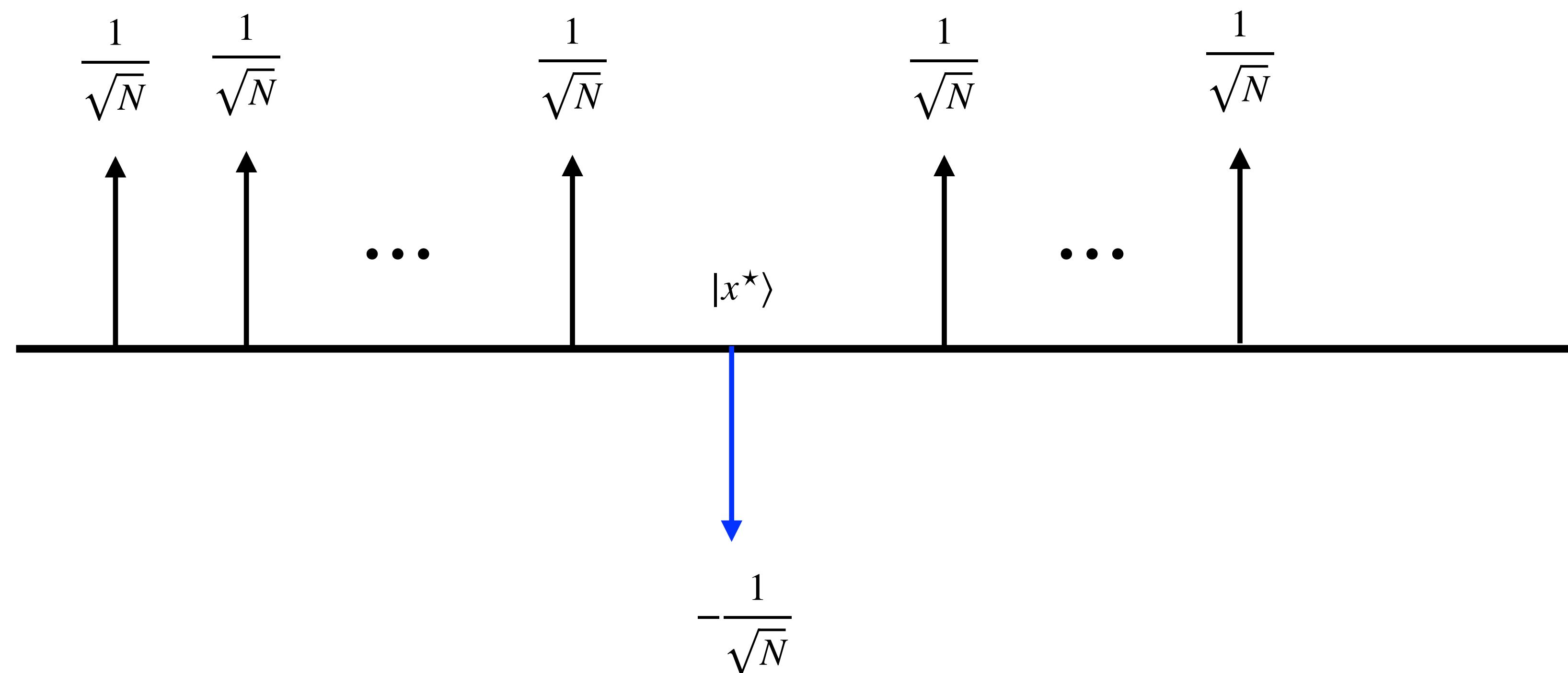
$$N = 2^n$$

For now, assume there is exactly one x^\star with $f(x^\star) = 1$



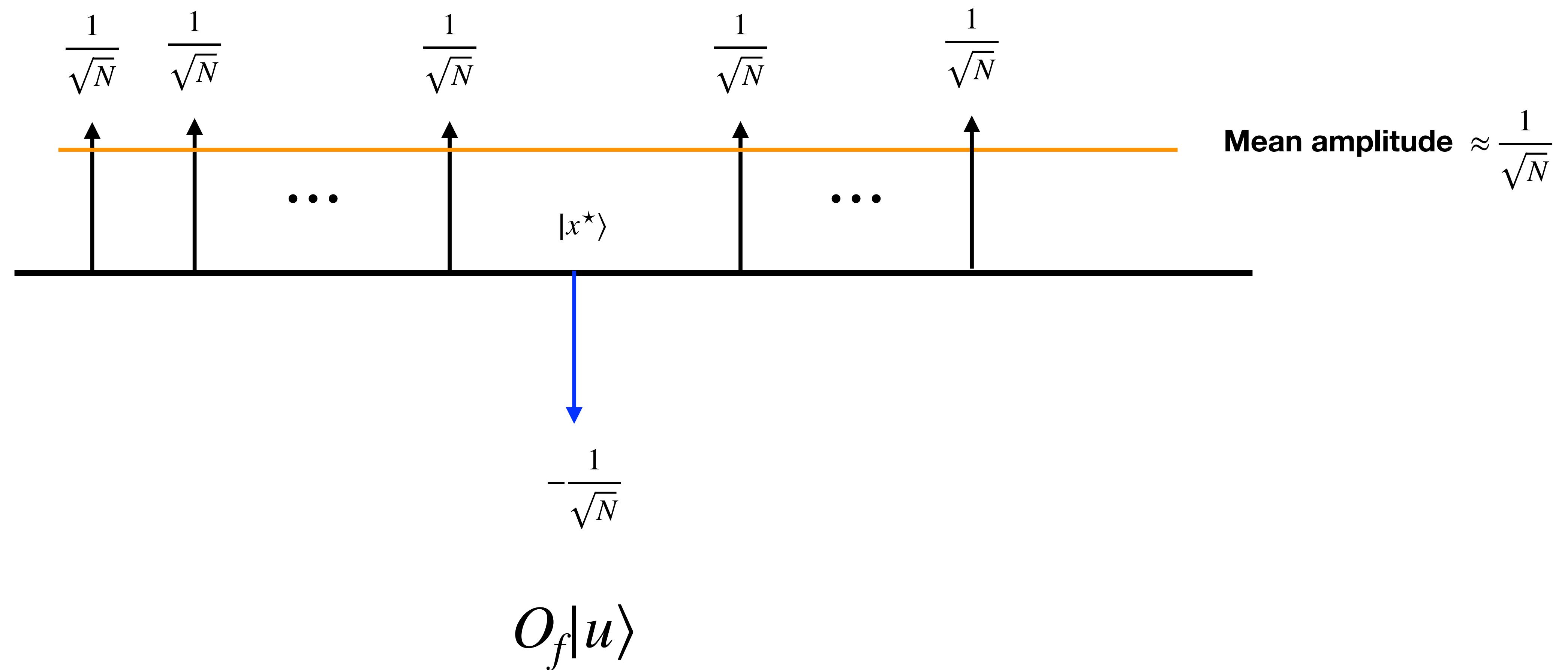
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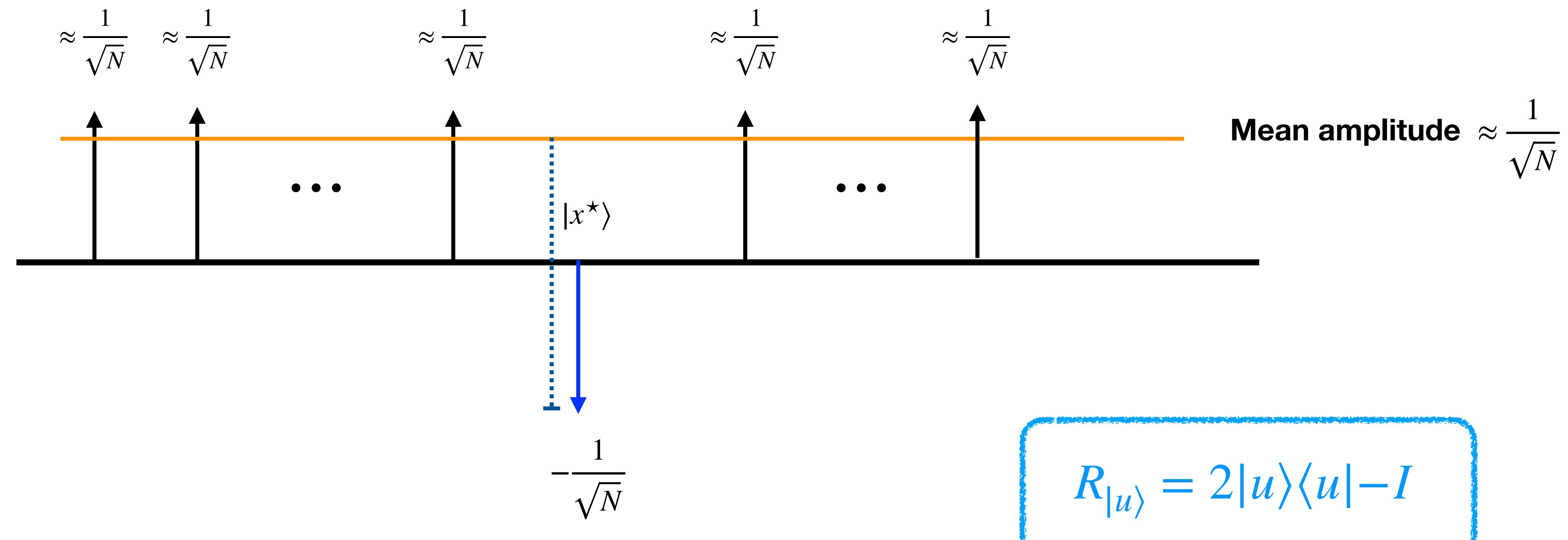


$$O_f|u\rangle$$

Some Intuition



Some Intuition

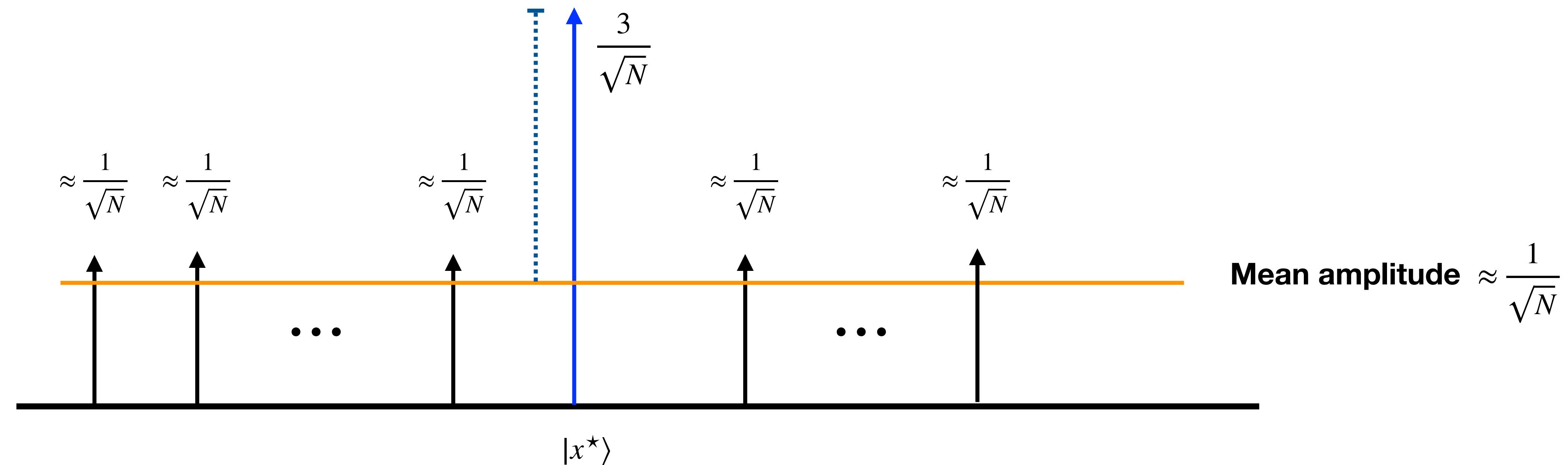


Reflect around the “mean”

$$R_{|u\rangle} O_f |u\rangle$$

$$R_{|u\rangle} = 2|u\rangle\langle u| - I$$

Some Intuition

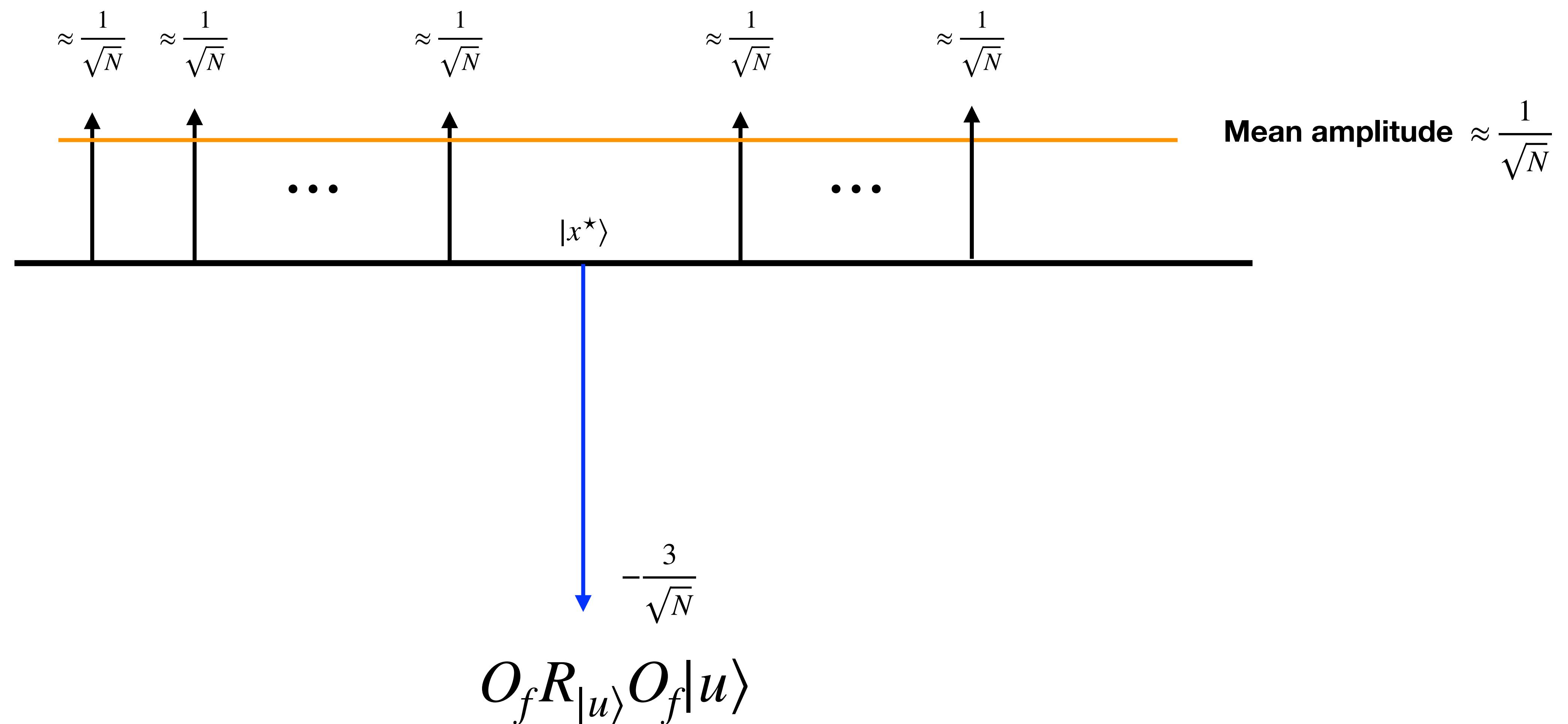


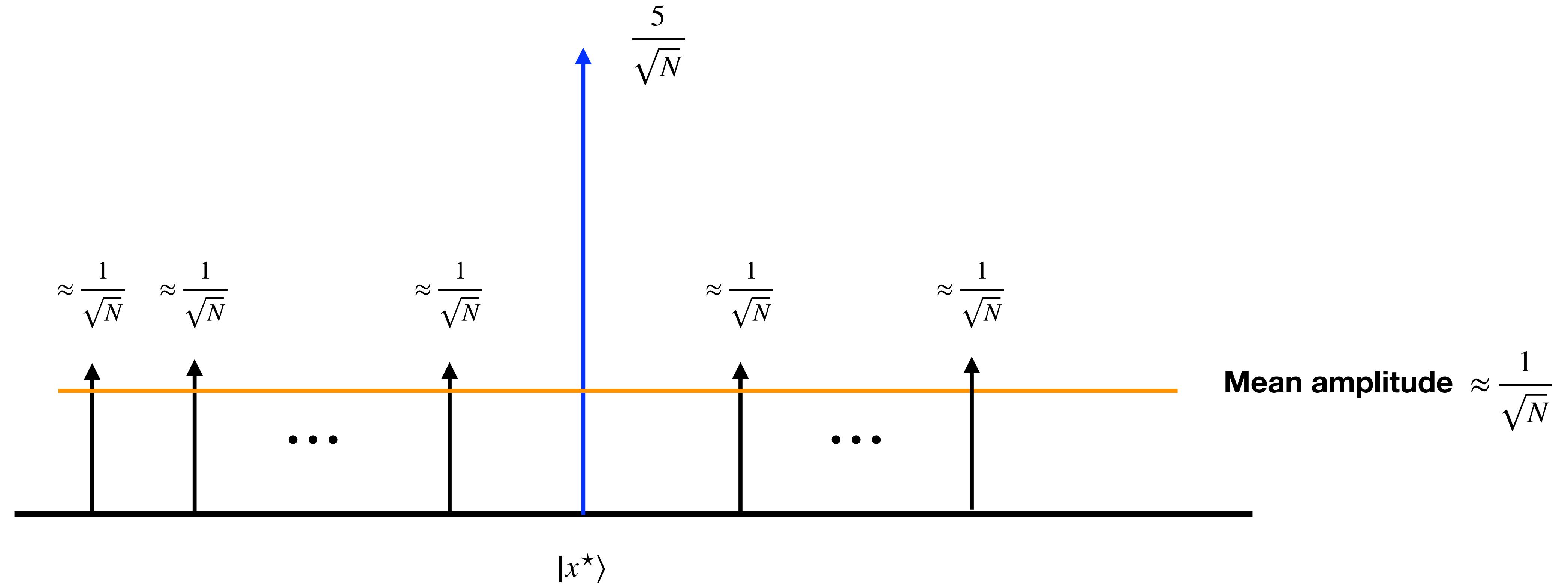
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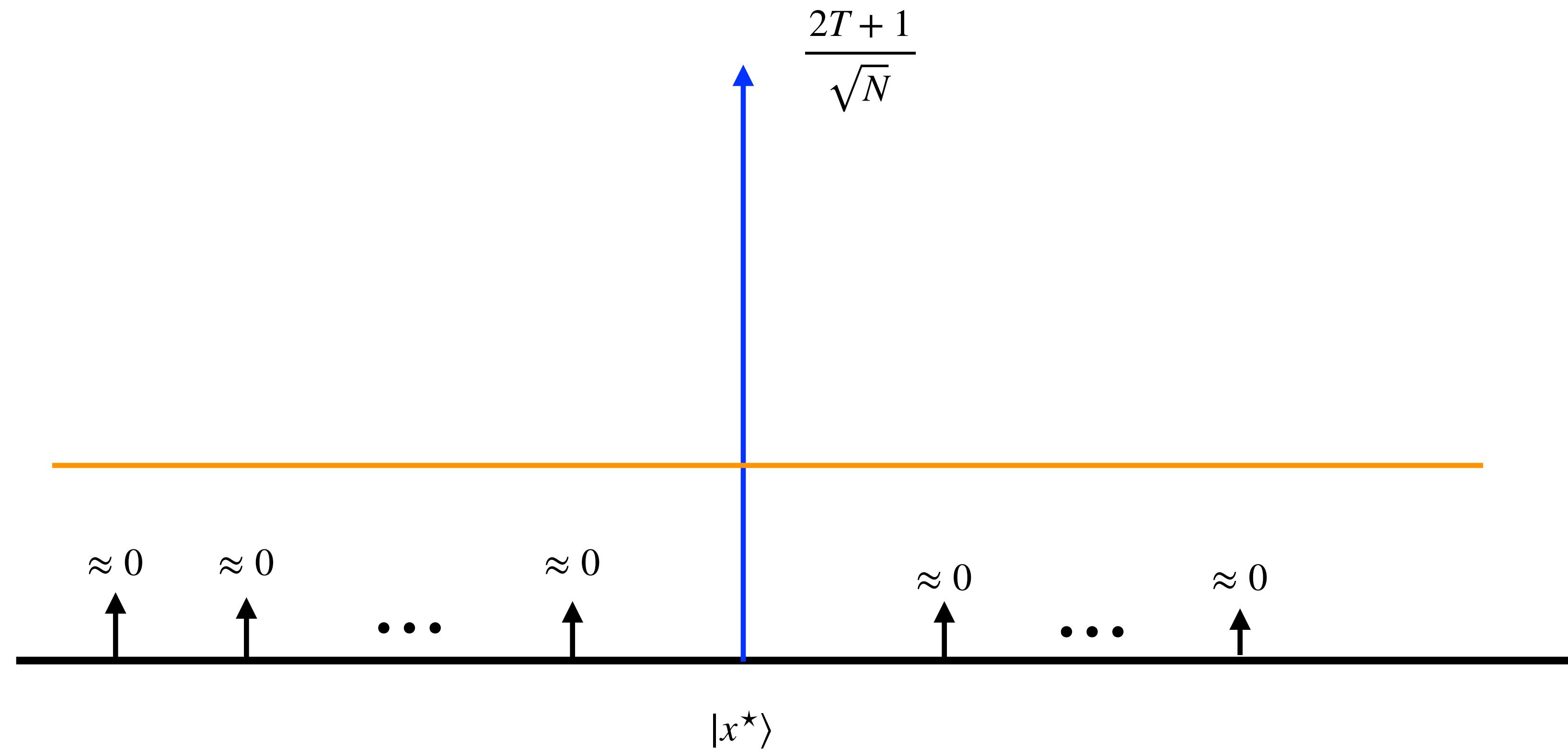




Reflect around the “mean”

$$R_{|u\rangle}O_fR_{|u\rangle}O_f|u\rangle$$

Extrapolating, we would guess...

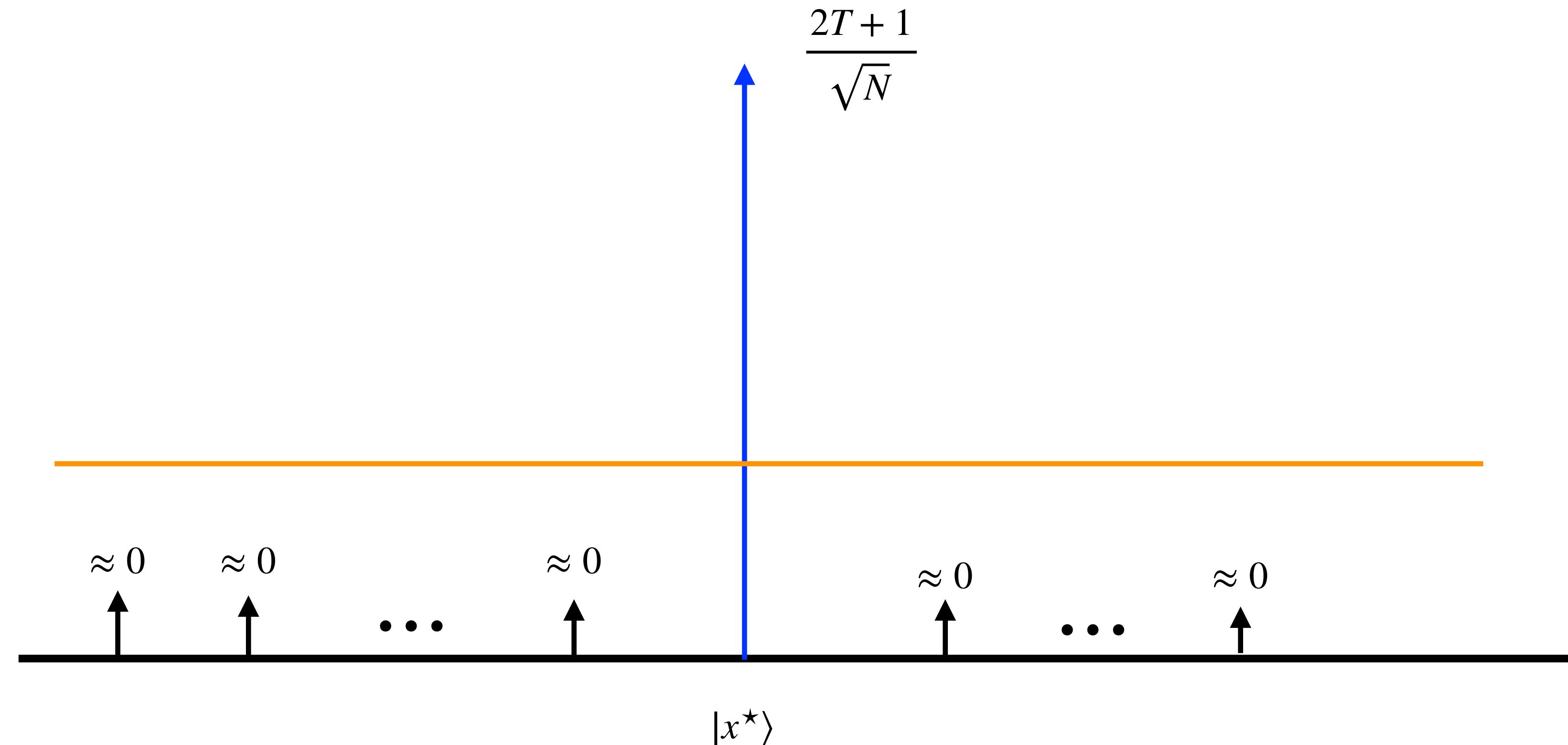


After T queries

$$(R_{|u\rangle}O_f)^T|u\rangle$$

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Suggests $T = O(\sqrt{N})$ suffice, beating brute-force search!

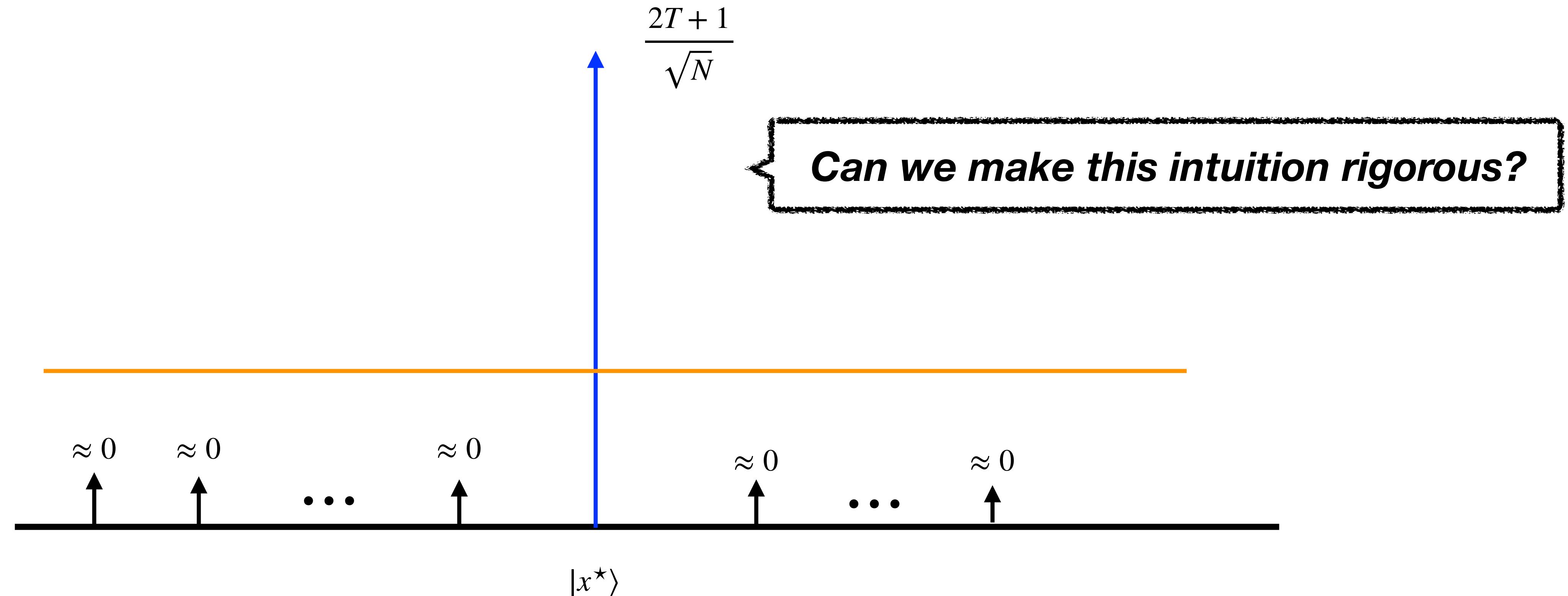


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Some Definitions

$$A = \{x \in \{0,1\}^n \mid f(x) = 1\}$$

Accepting Inputs

$$B = \{x \in \{0,1\}^n \mid f(x) = 0\}$$

Bad Inputs

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Bad Inputs

$$|A\rangle = \frac{1}{\sqrt{A}} \sum_{x \in A} |x\rangle$$

$$|B\rangle = \frac{1}{\sqrt{B}} \sum_{x \in B} |x\rangle$$

Understanding O_f

$$A = \{x \in \{0,1\}^n \mid f(x) = 1\} \quad \text{Accepting Inputs}$$

$$B = \{x \in \{0,1\}^n \mid f(x) = 0\} \quad \text{Bad Inputs}$$

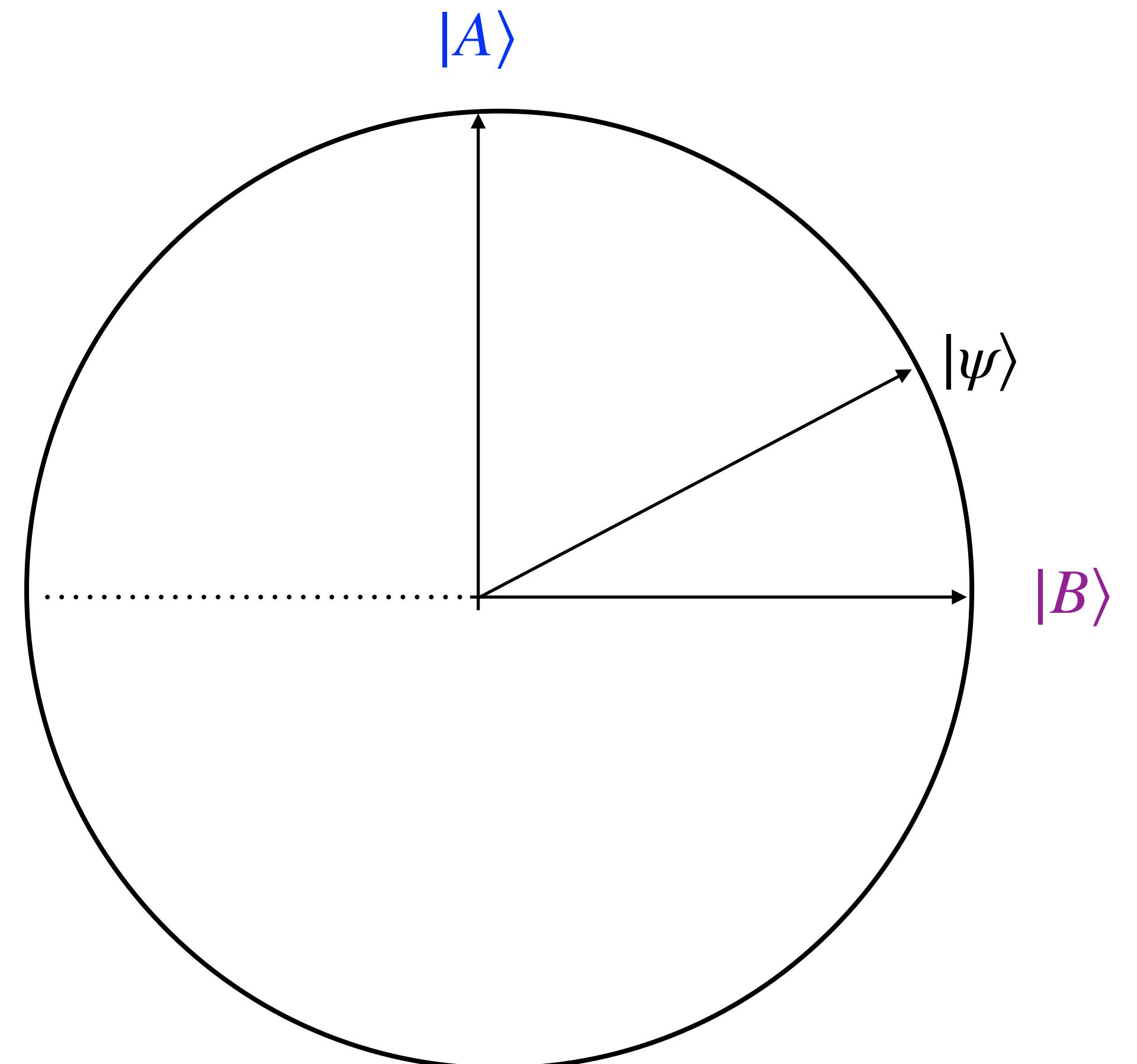
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$$O_f|B\rangle = |B\rangle$$

Understanding O_f

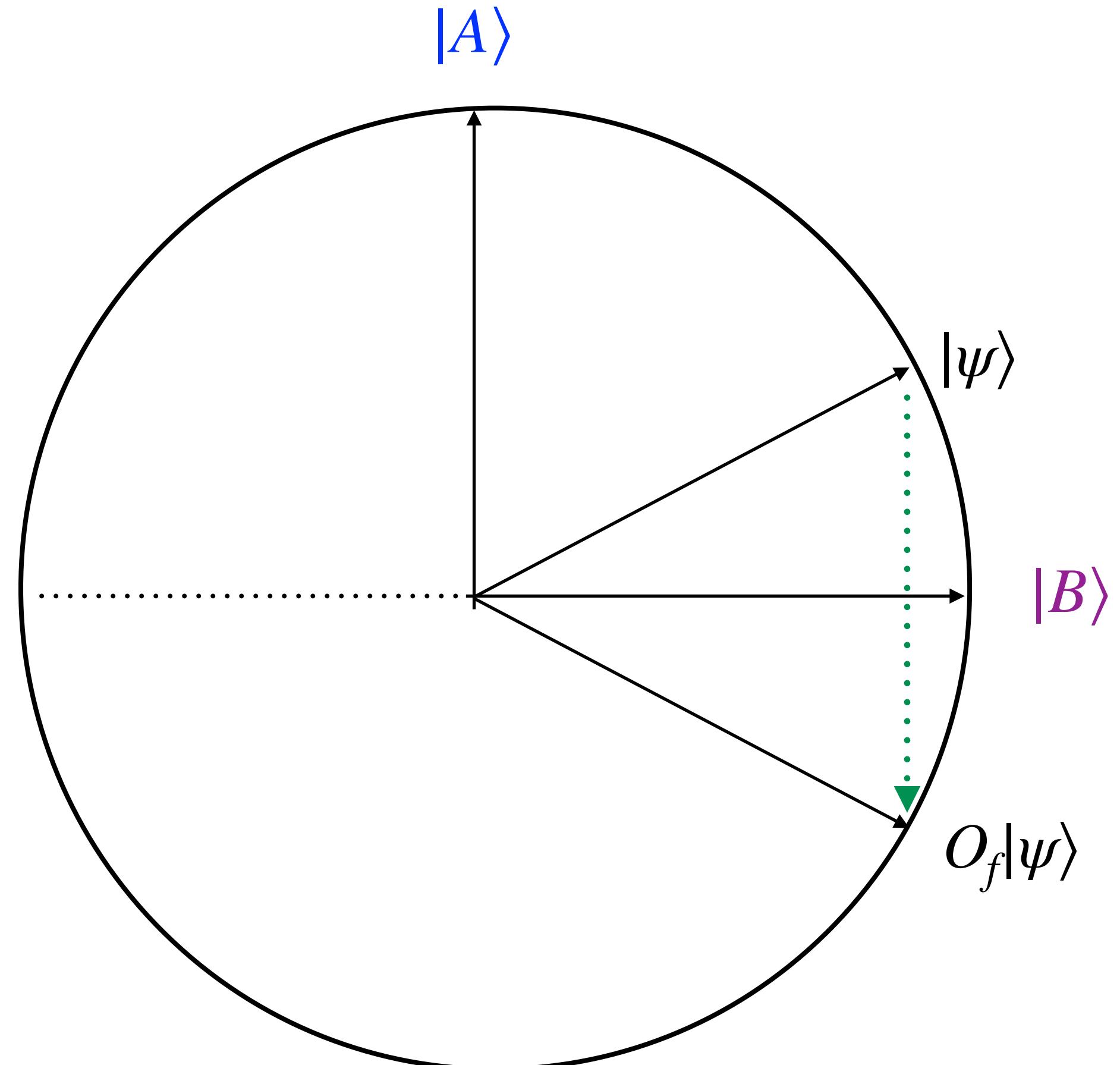


$$\langle A | B \rangle = 0$$

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Understanding O_f



$$\langle A | B \rangle = 0$$

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$$O_f|B\rangle = |B\rangle$$

O_f is a reflection around $|B\rangle$

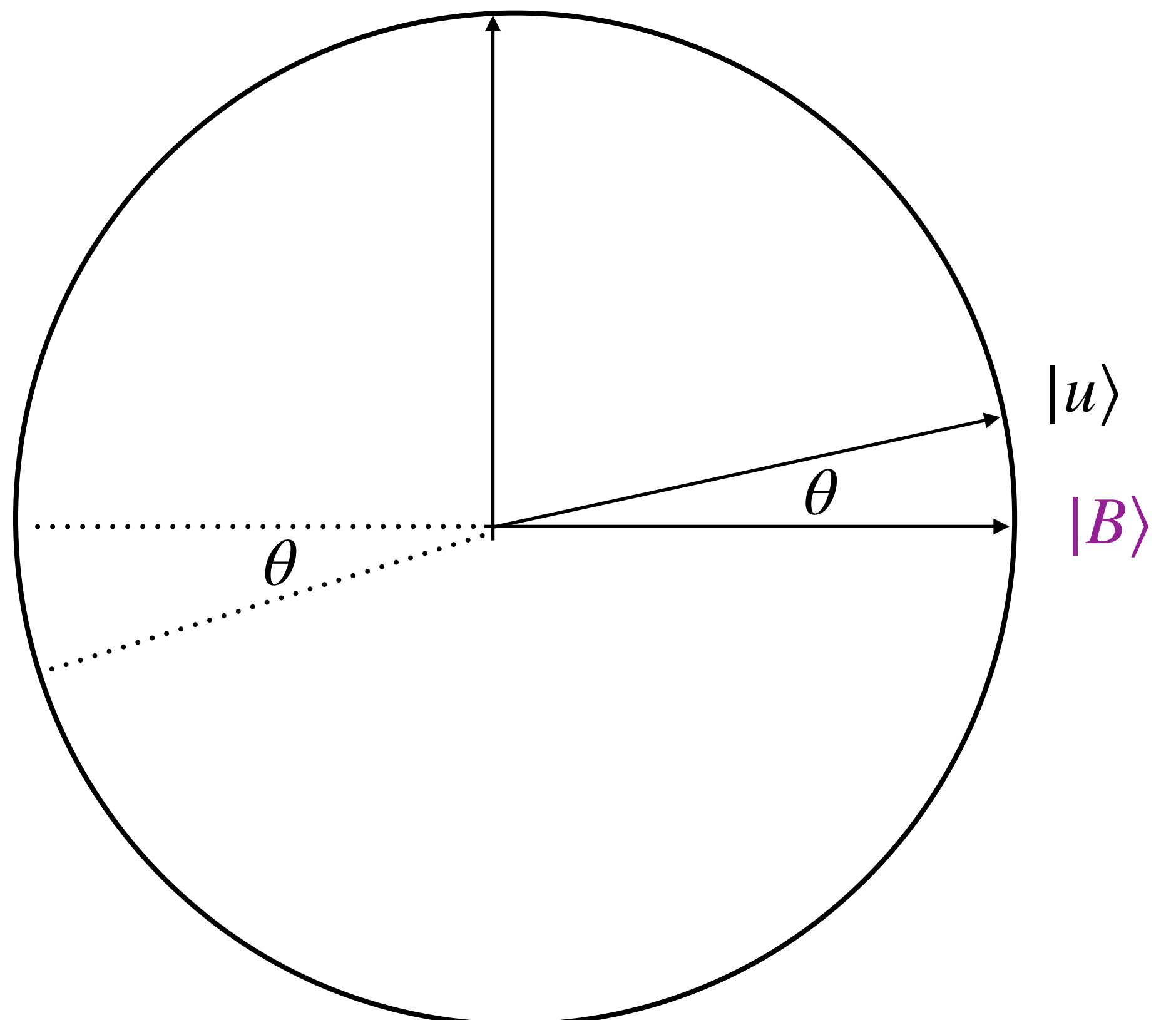
Composing two Reflections

$$|u\rangle = \sqrt{\frac{A}{N}}|A\rangle + \sqrt{\frac{B}{N}}|B\rangle$$

$$R_{|u\rangle} = 2|u\rangle\langle u| - I$$

$|A\rangle$

$$R_{|B\rangle} = O_f$$



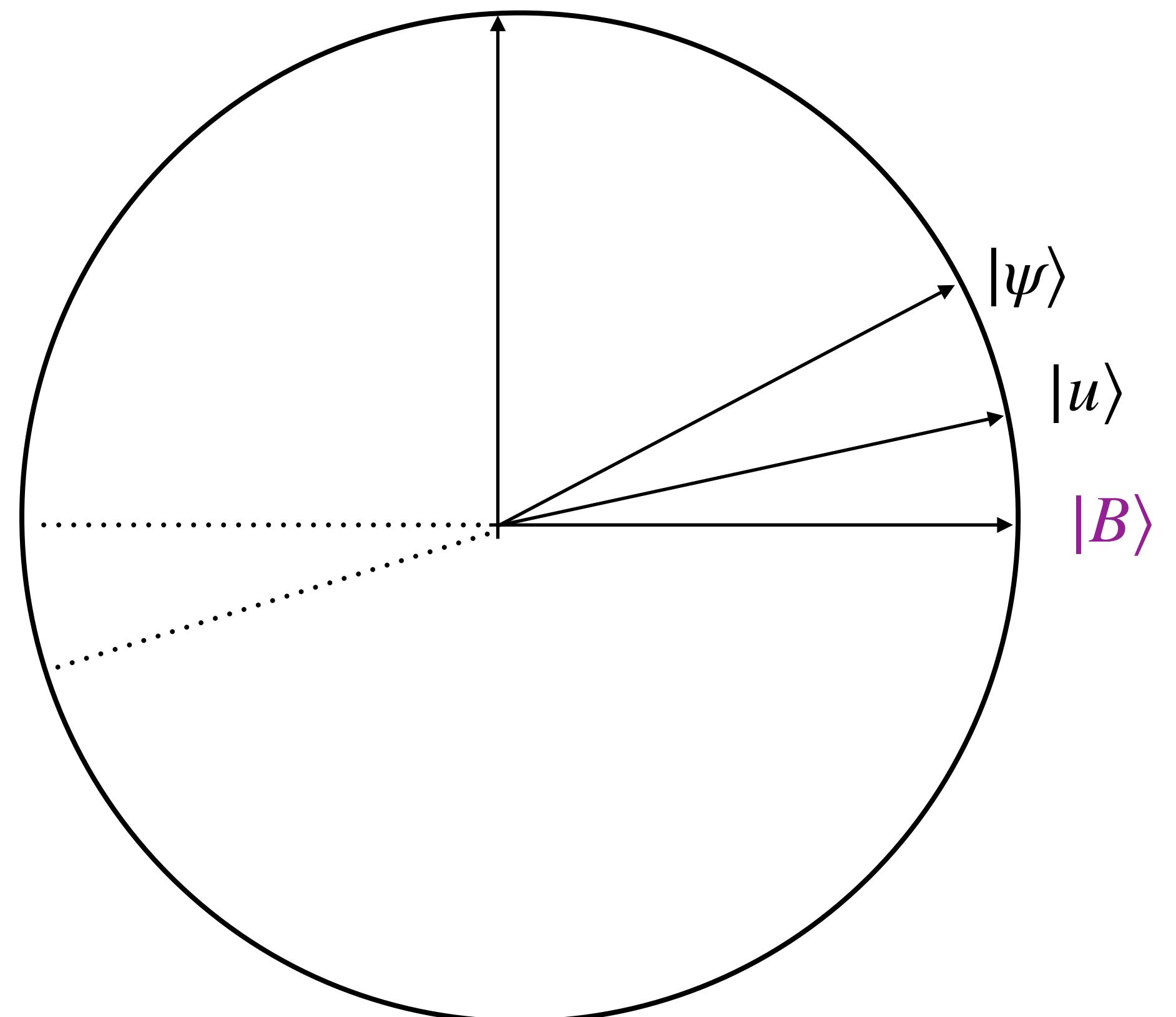
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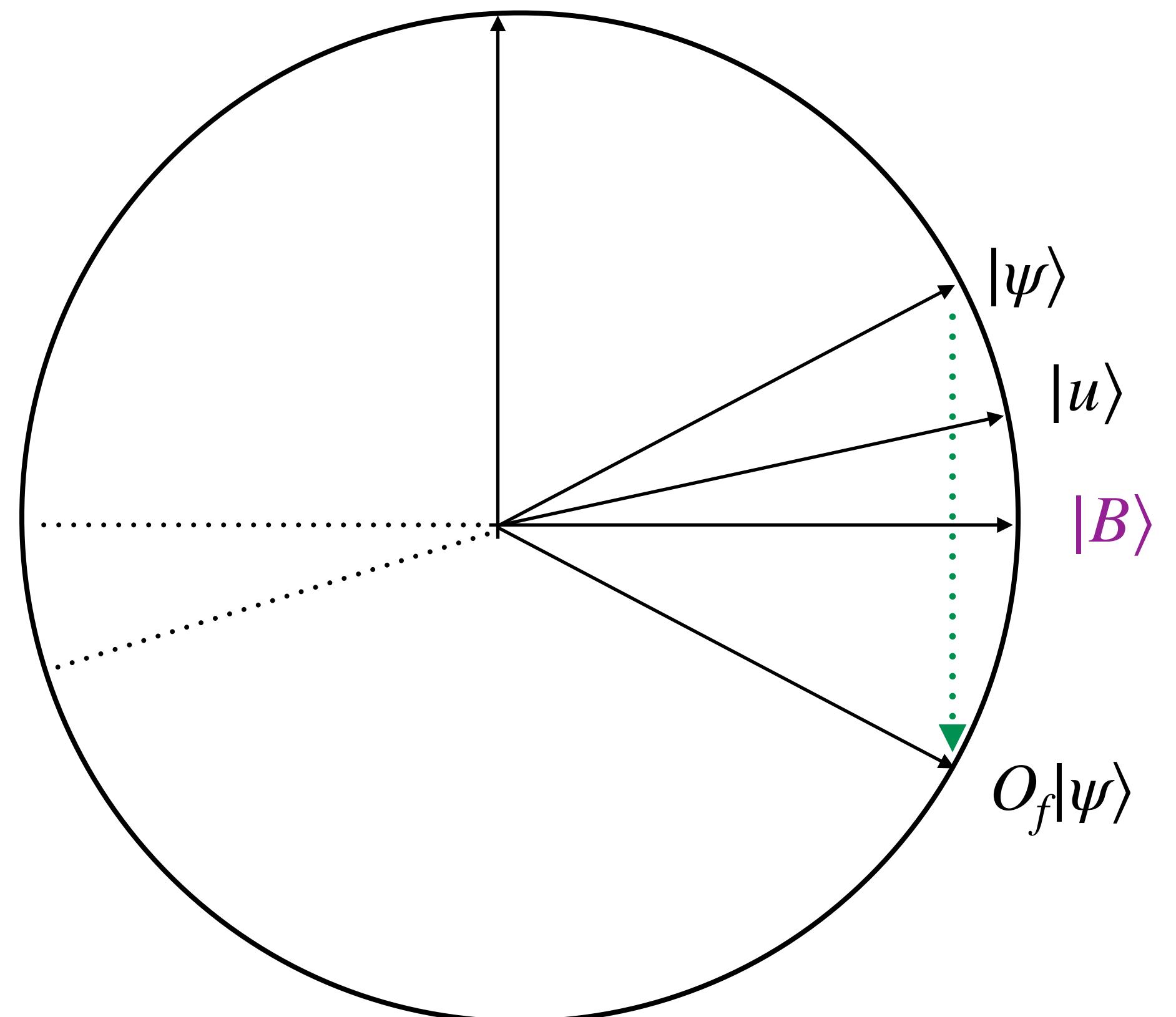
$|A\rangle$

$$R_{|B\rangle} = O_f$$

$|u\rangle$

$|B\rangle$

$O_f|\psi\rangle$



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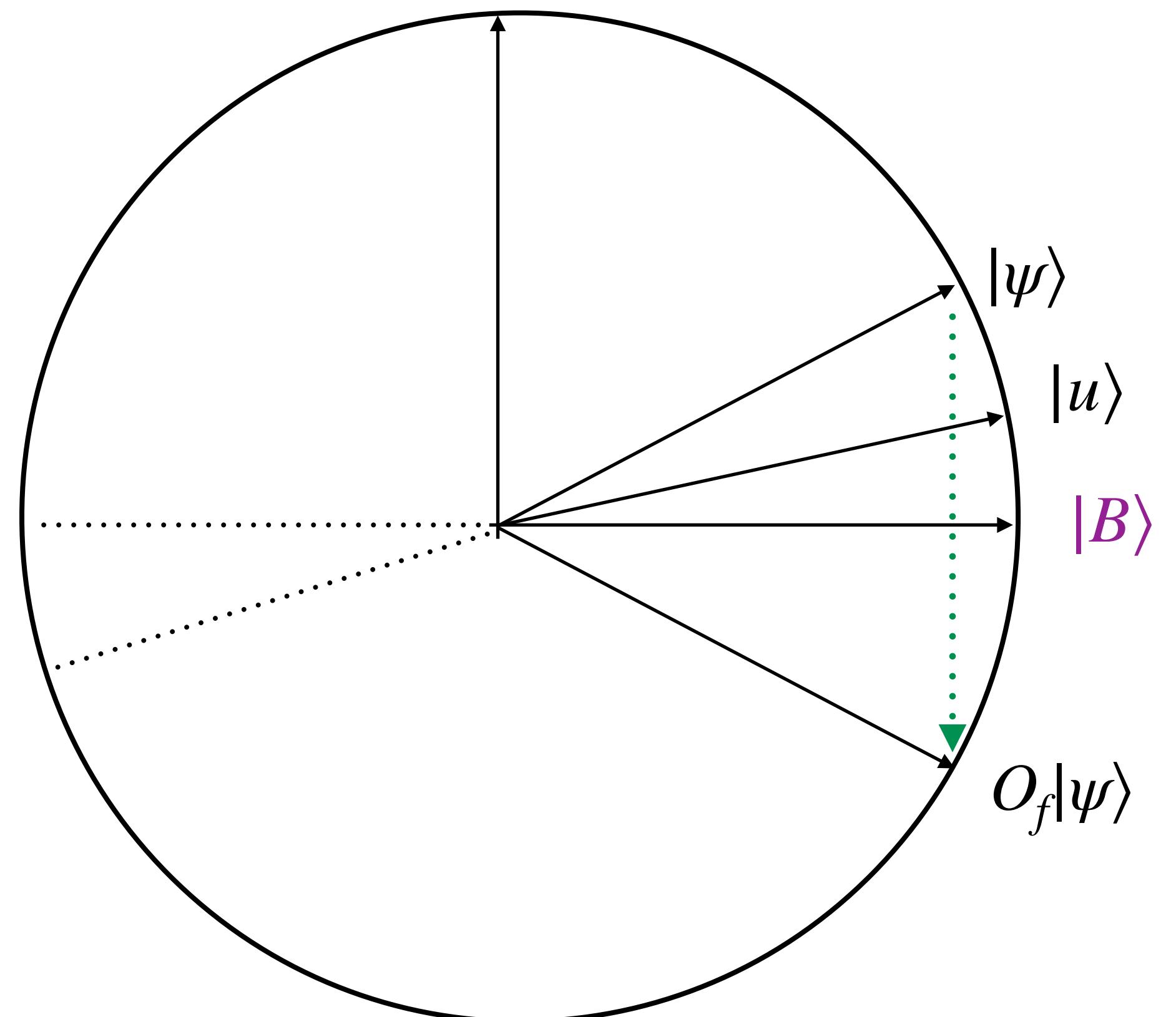
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$O_f|\psi\rangle$



$R_{|u\rangle}$ is a reflection around $|u\rangle$

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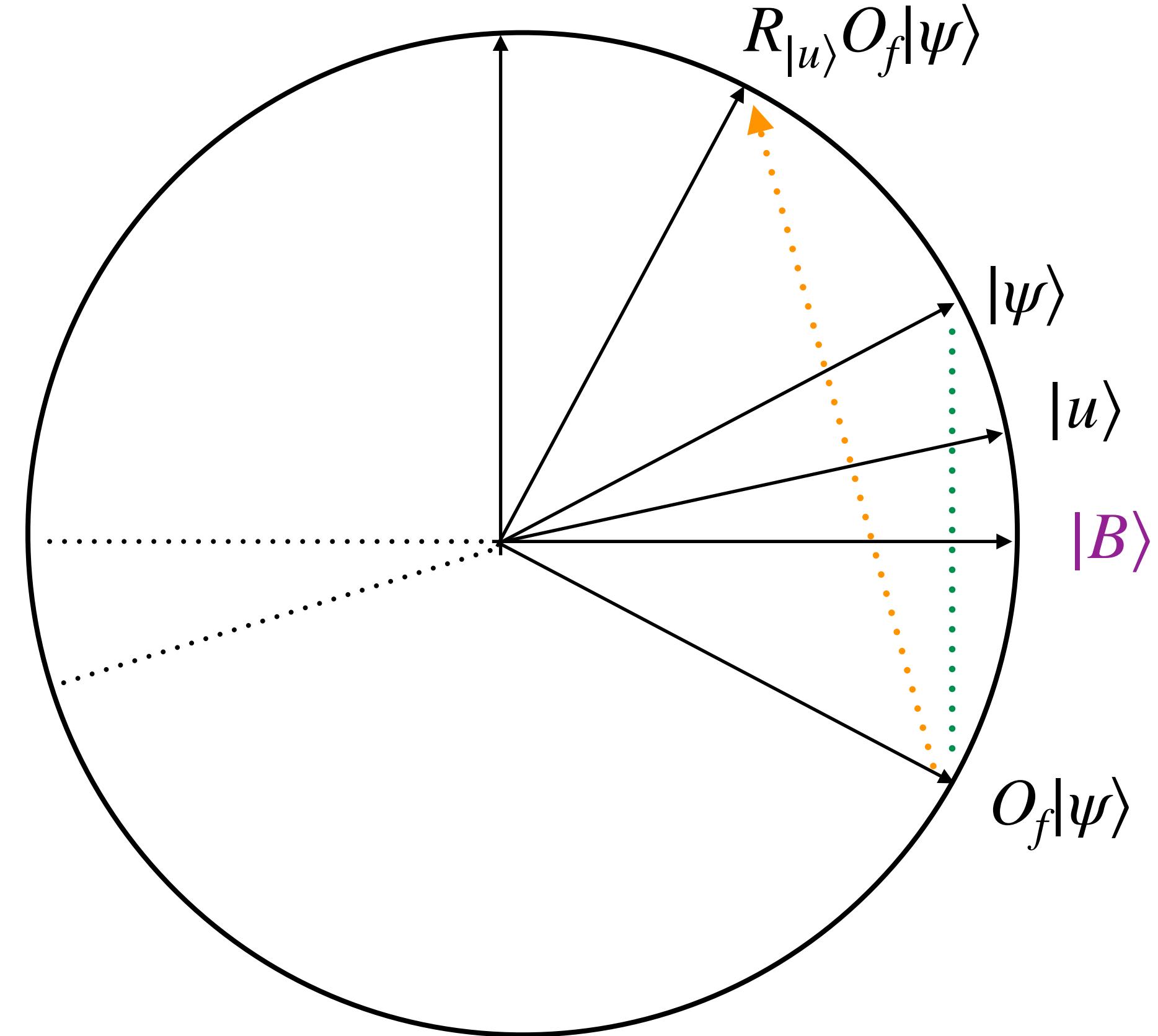
$R_{|u\rangle}O_f|\psi\rangle$

$|\psi\rangle$

$|u\rangle$

$|B\rangle$

$O_f|\psi\rangle$



$R_{|u\rangle}$ is a reflection around $|u\rangle$

Composing two Reflections gives a Rotation

$$|u\rangle = \sqrt{\frac{A}{N}}|A\rangle + \sqrt{\frac{B}{N}}|B\rangle$$

$$R_{|u\rangle} = 2|u\rangle\langle u| - I$$

$|A\rangle$

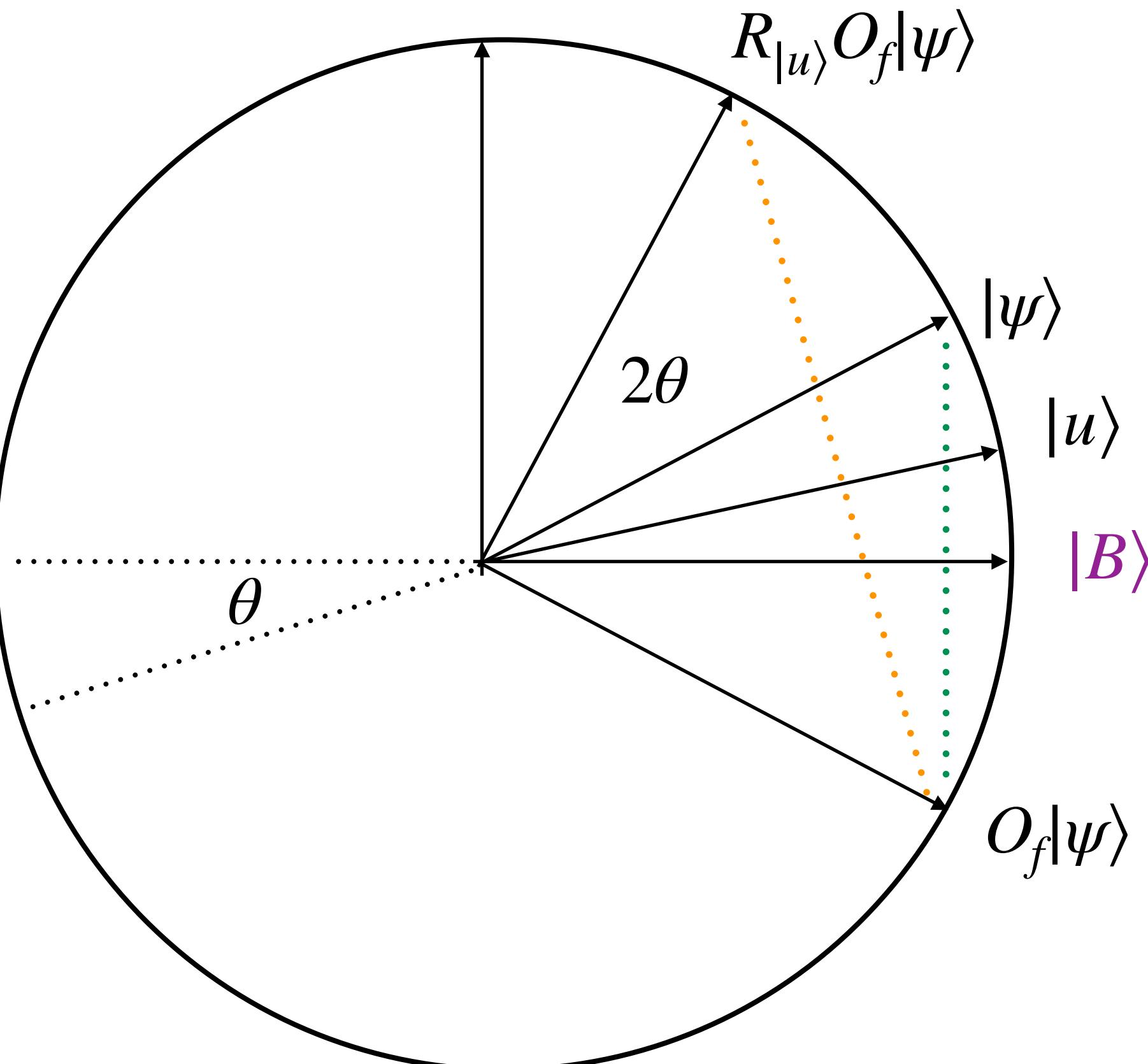
$$R_{|B\rangle} = O_f$$

$R_{|u\rangle}O_f|\psi\rangle$

$|\psi\rangle$
 $|u\rangle$

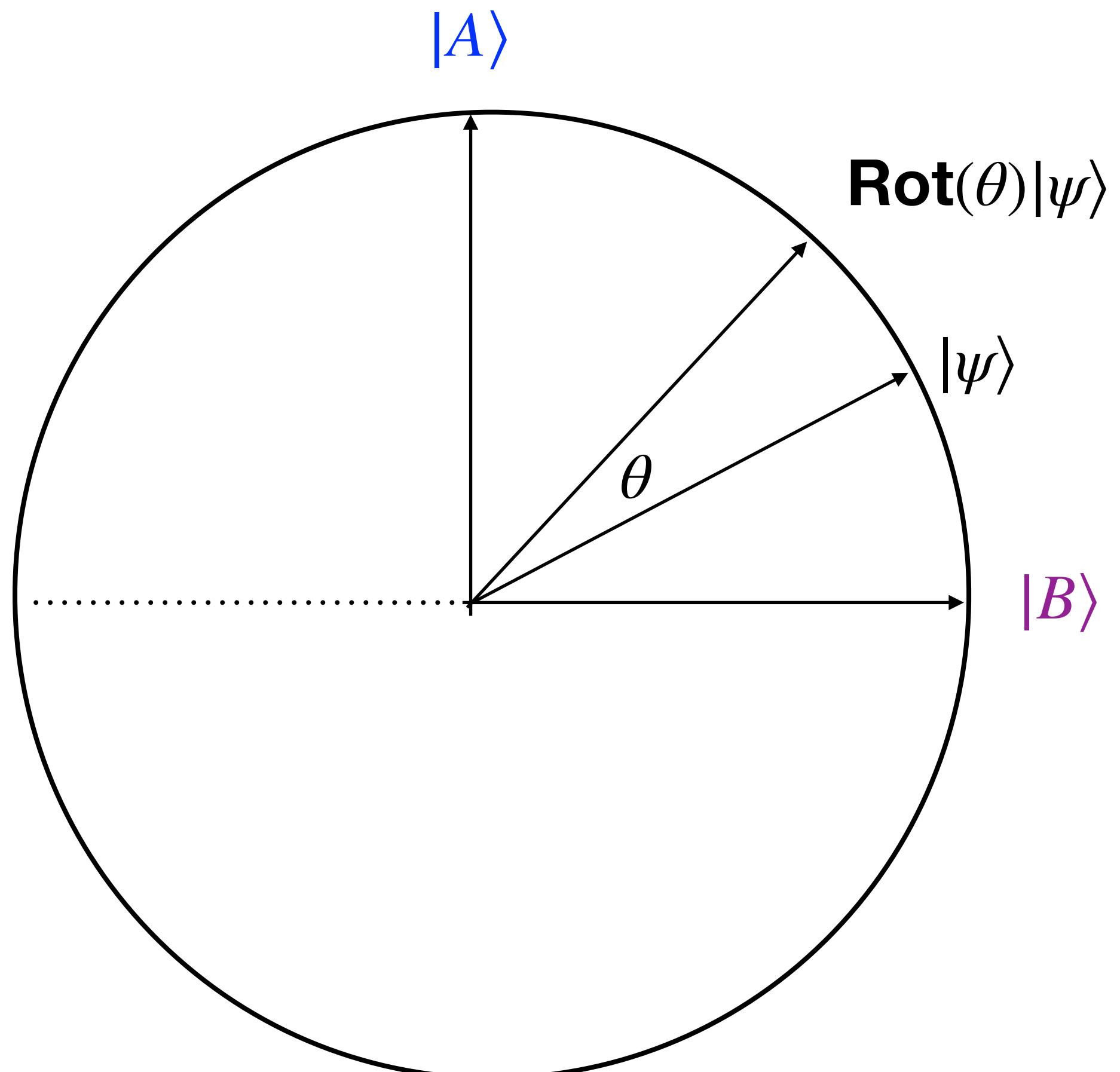
$|B\rangle$

2θ



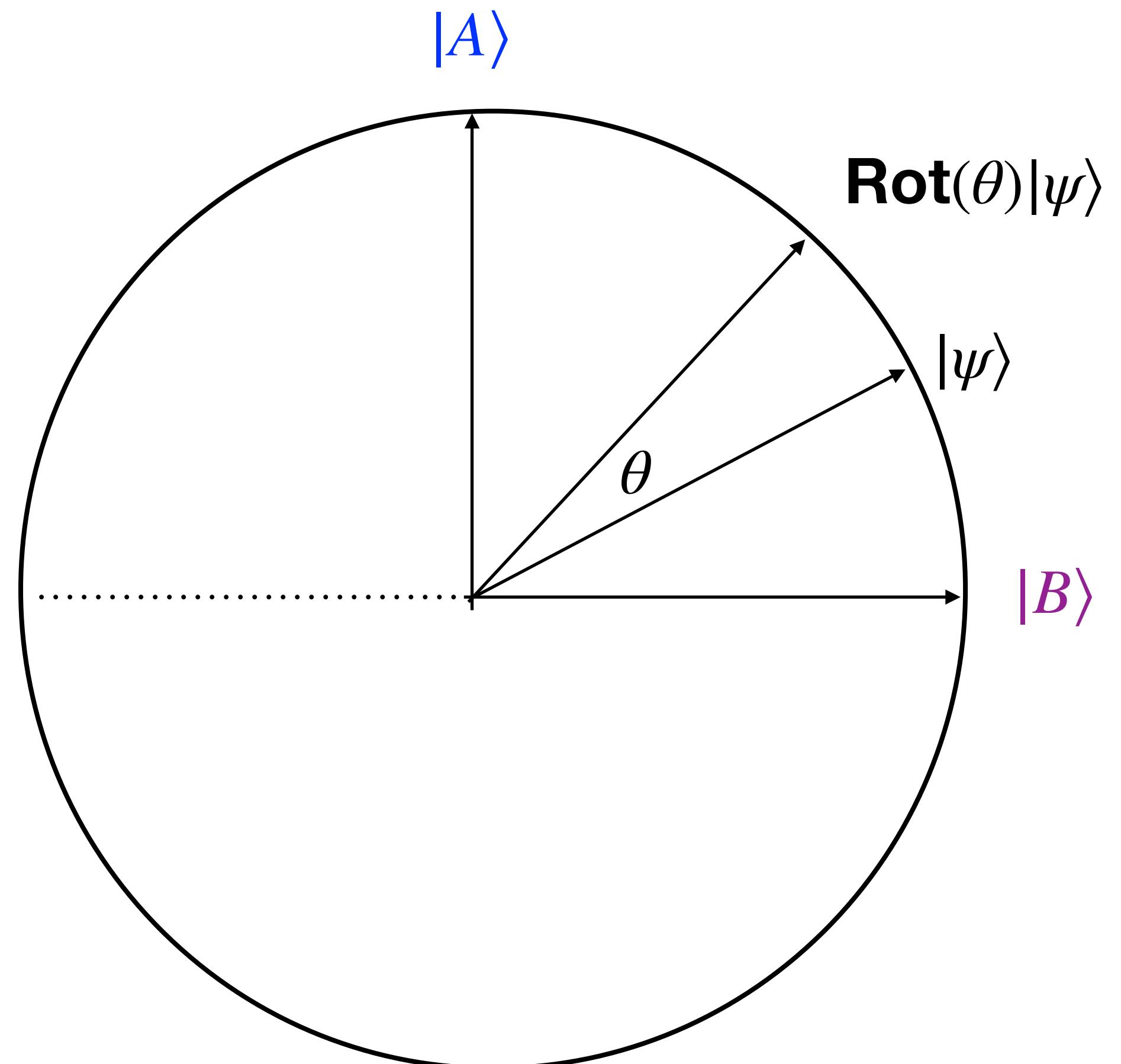
Fact: $R_{|u\rangle}R_{|B\rangle} = \text{Rot}(2\theta)$
 θ is the angle between $|u\rangle$ and $|B\rangle$

Refresher on Rotation Matrices



$$\text{Rot}(\theta) = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}$$

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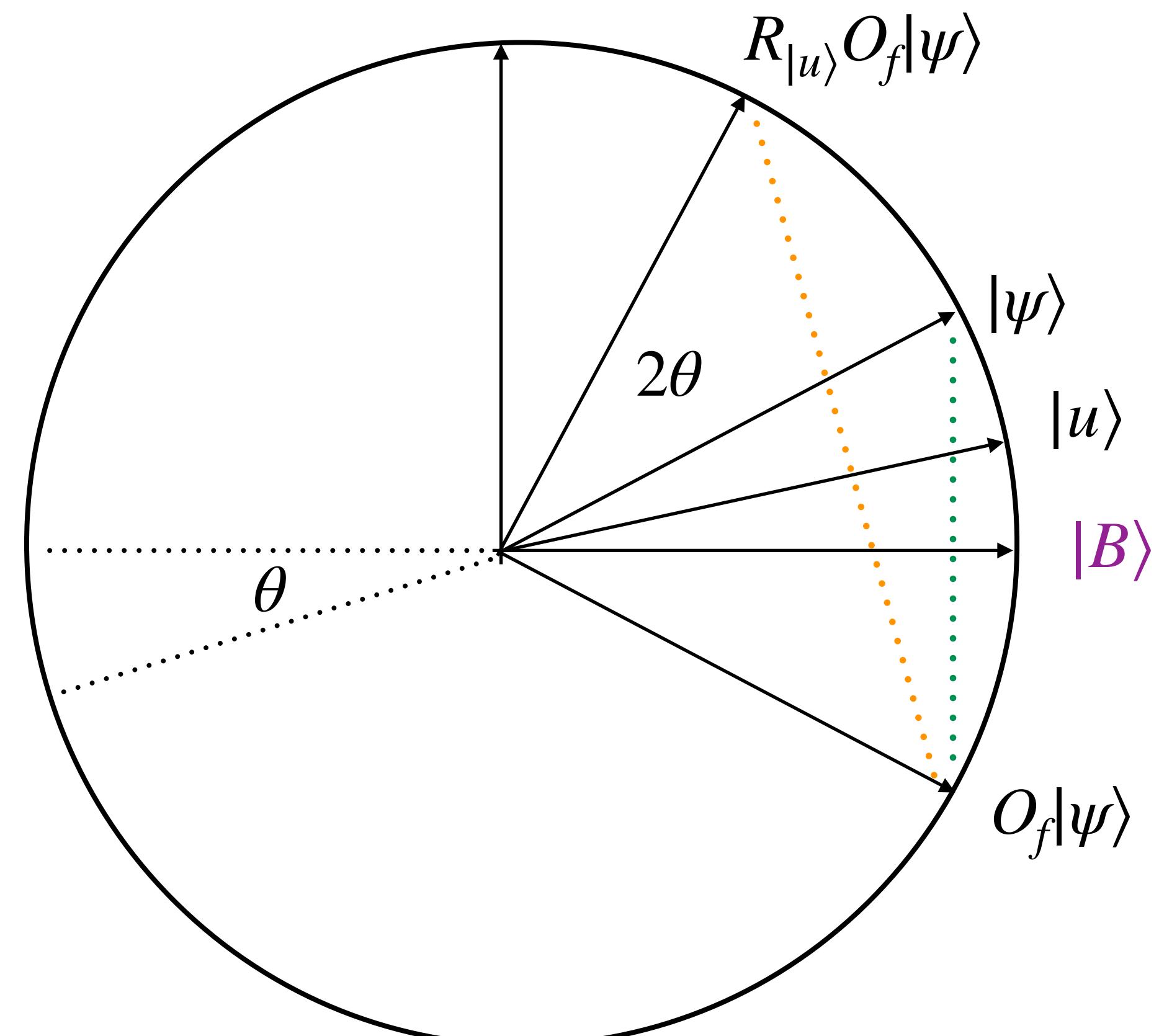
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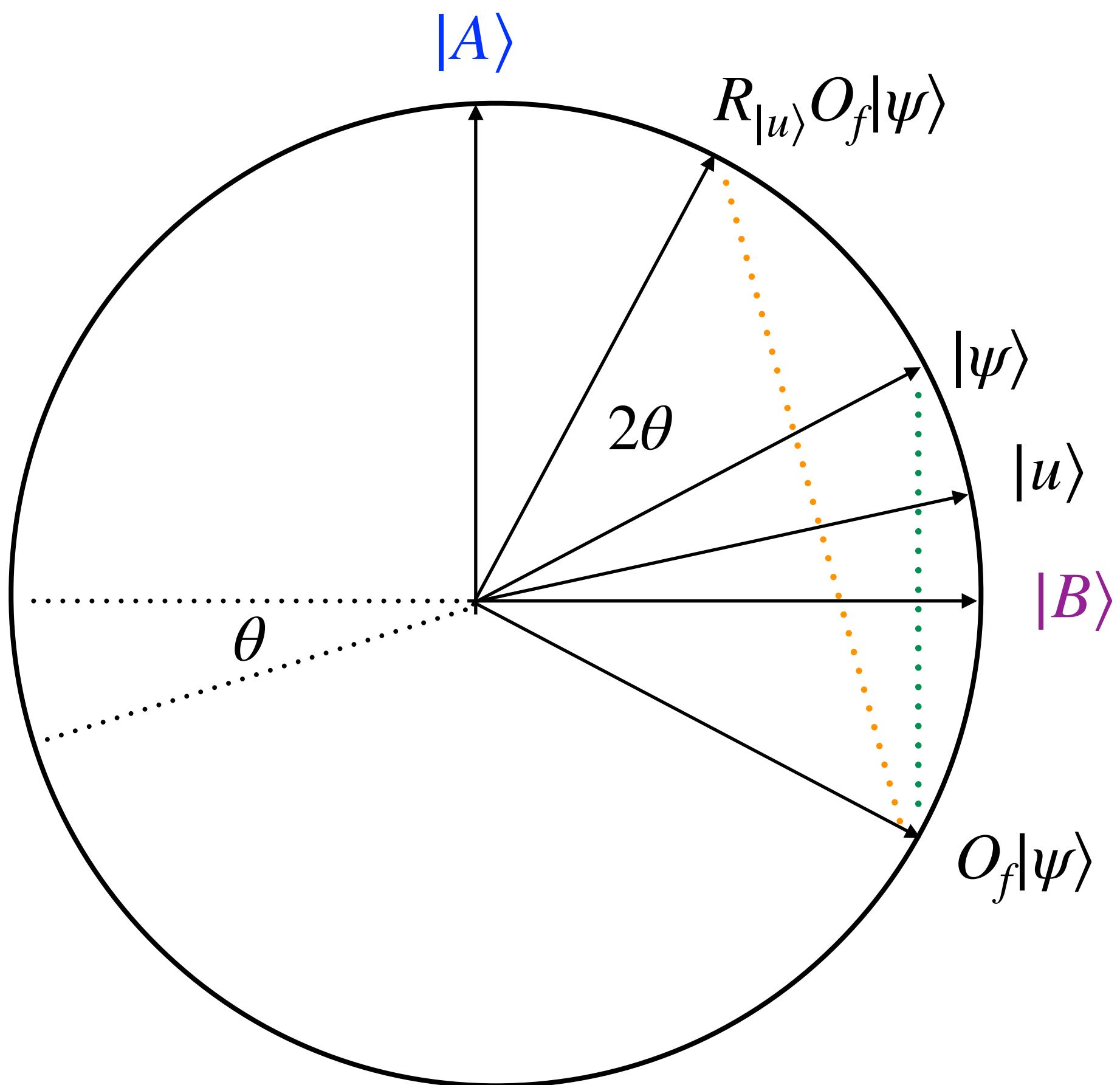
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Analyzing the Number of Queries

$$|u\rangle = \sqrt{\frac{A}{N}}|A\rangle + \sqrt{\frac{B}{N}}|B\rangle = \sin(\theta)|A\rangle + \cos(\theta)|B\rangle$$

Fact: $R_{|u\rangle}R_{|B\rangle} = \text{Rot}(2\theta)$
 $\theta = \sin^{-1}\left(\sqrt{\frac{A}{N}}\right)$

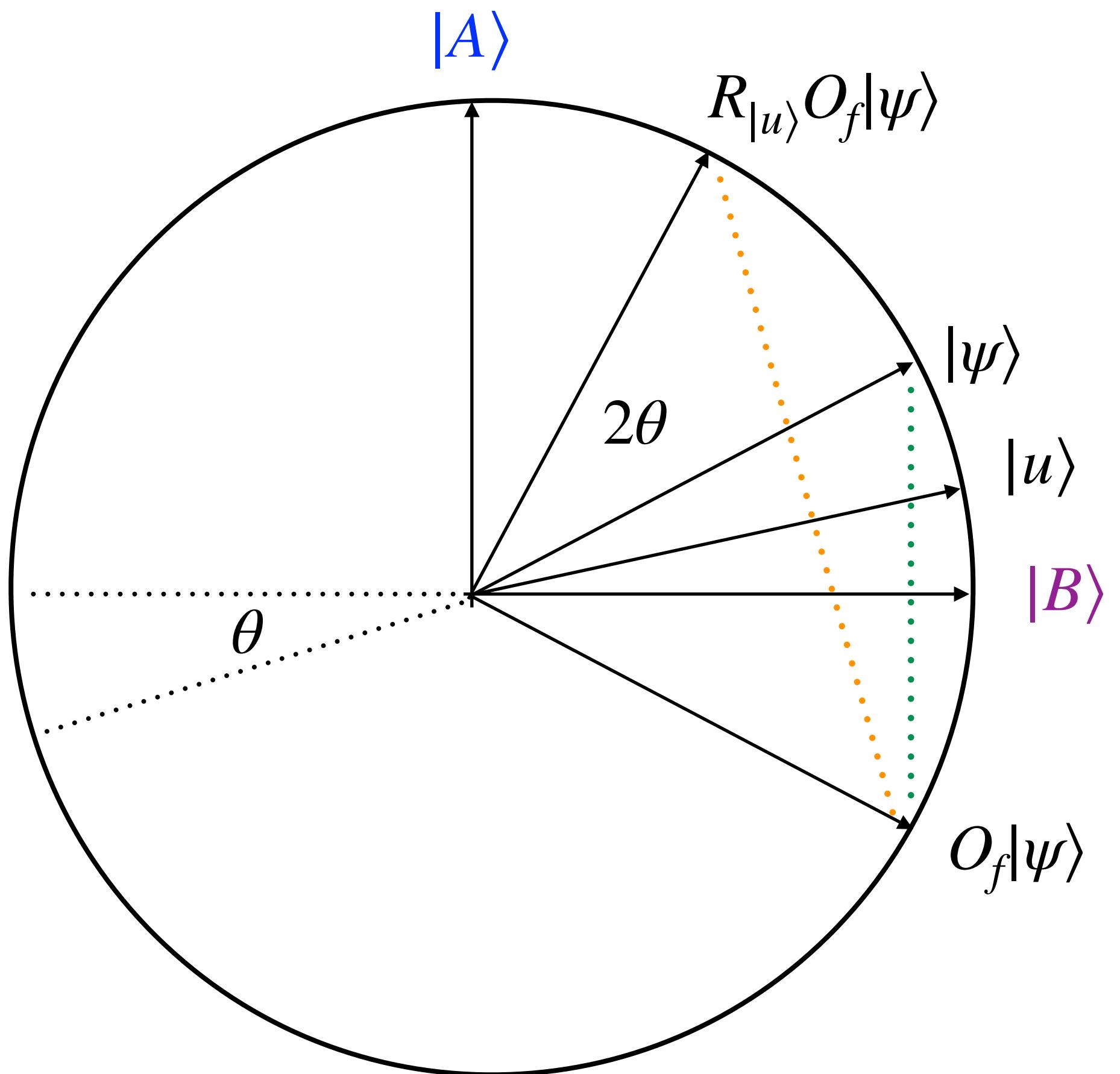


$$(R_{|u\rangle}R_{|B\rangle})^T|u\rangle = \text{Rot}(2T\theta)|u\rangle = \sin((2T+1)\theta)|A\rangle + \cos((2T+1)\theta)|B\rangle$$

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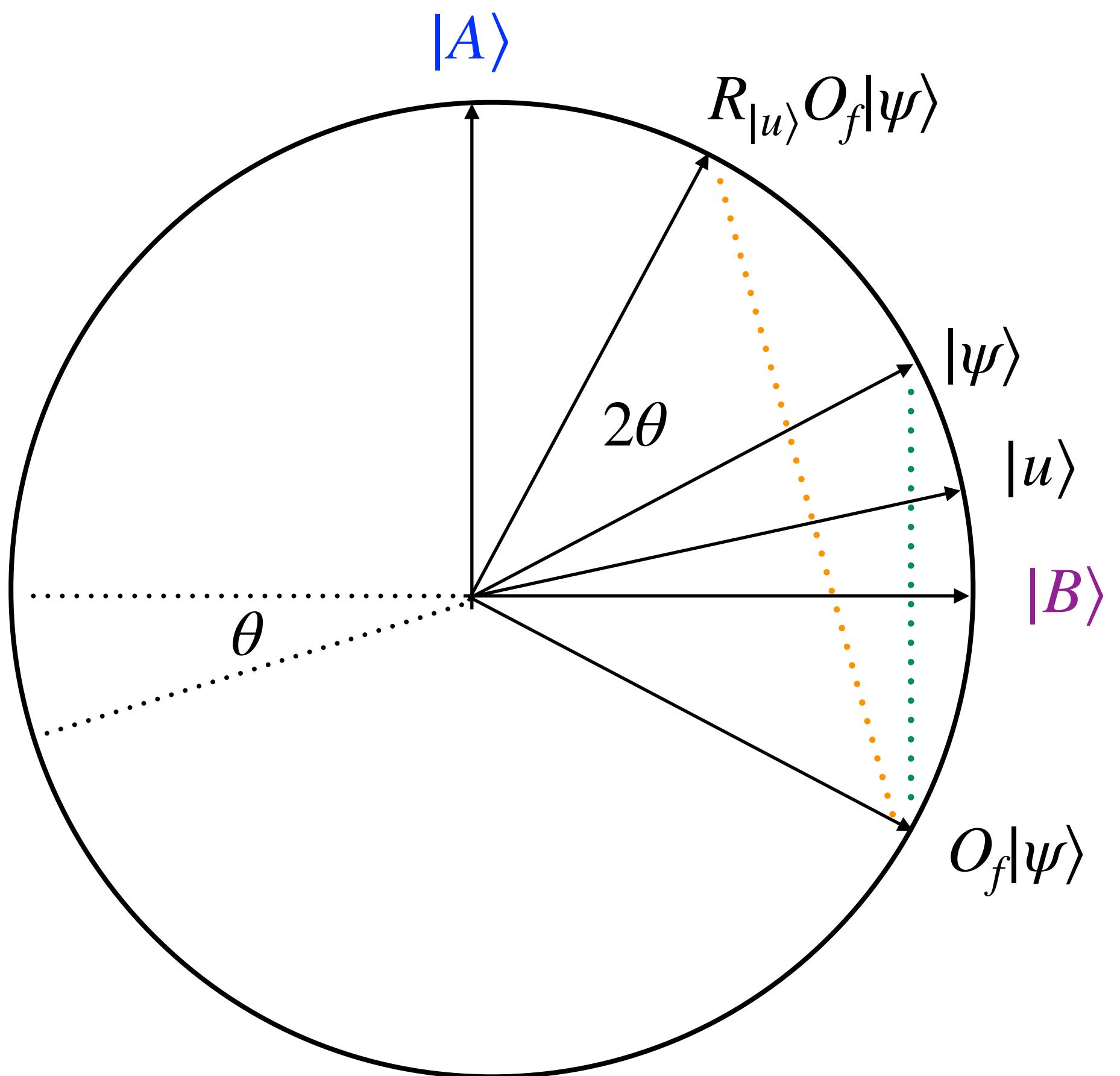


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Want: $(R_{|u\rangle}R_{|B\rangle})^T|u\rangle \approx |A\rangle \implies (2T+1)\theta \approx \frac{\pi}{2}$

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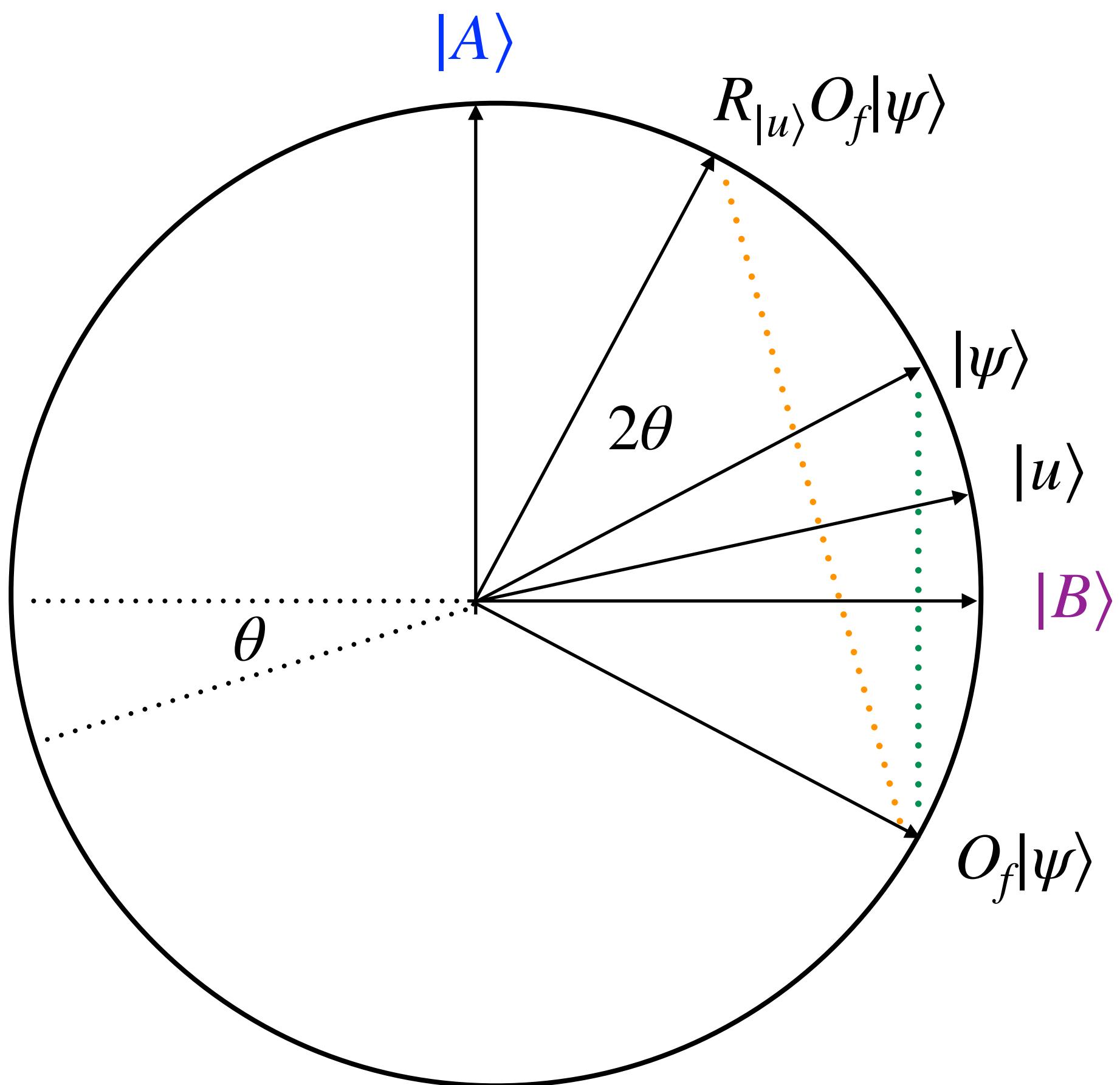
$$\theta \geq \sqrt{\frac{A}{N}} \text{ (Taylor series)}$$

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Number of Queries: $T \approx \frac{\pi}{4} \sqrt{\frac{N}{A}}$

In Summary: Grover's Algorithm gives a quadratic speedup

Grover'96

Unstructured Search can be solved with $O(\sqrt{N})$ quantum queries

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Widely used to obtain many other polynomial quantum speedups

Provably optimum in the black-box query model

Simple Technicalities to Handle

$$|u\rangle = \sqrt{\frac{A}{N}}|A\rangle + \sqrt{\frac{B}{N}}|B\rangle = \sin(\theta)|A\rangle + \cos(\theta)|B\rangle$$

Number of Queries: $T \approx \frac{\pi}{4} \sqrt{\frac{N}{A}}$

We (implicitly) assumed that $|A| \neq 0$ (the case $|A| = 0$ is easy to analyze, do you see this?)

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Can we implement the reflection $R_{|u\rangle}$ efficiently?

We do not know $|A|$ in advance, can think of strategies to deal with this?

Number of Queries in Grover's Algorithm

Can we do better (fewer queries) when $|A|$ is large?

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Grover's Search

Unstructured Search can be solved with $O\left(\sqrt{\frac{N}{A}}\right)$ quantum queries if $|A| \neq 0$

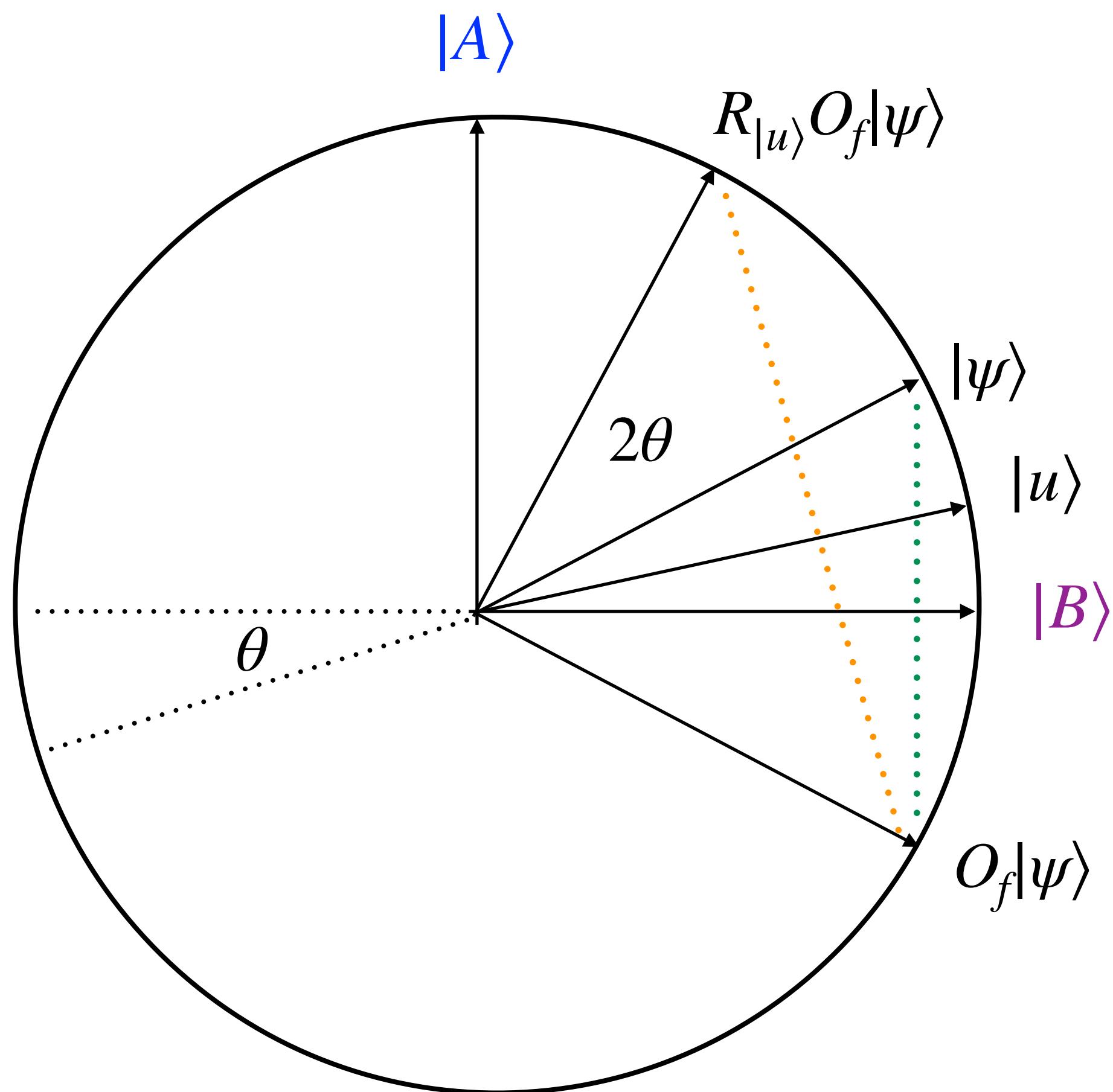
Completing the Proofs...

Proving the Composition Fact

$$|u\rangle = \sqrt{\frac{A}{N}}|A\rangle + \sqrt{\frac{B}{N}}|B\rangle$$

$$R_{|u\rangle} = 2|u\rangle\langle u| - I$$

$$R_{|B\rangle} = O_f$$



Fact: $R_{|u\rangle}R_{|B\rangle} = \text{Rot}(2\theta)$

θ is the angle between $|u\rangle$ and $|B\rangle$

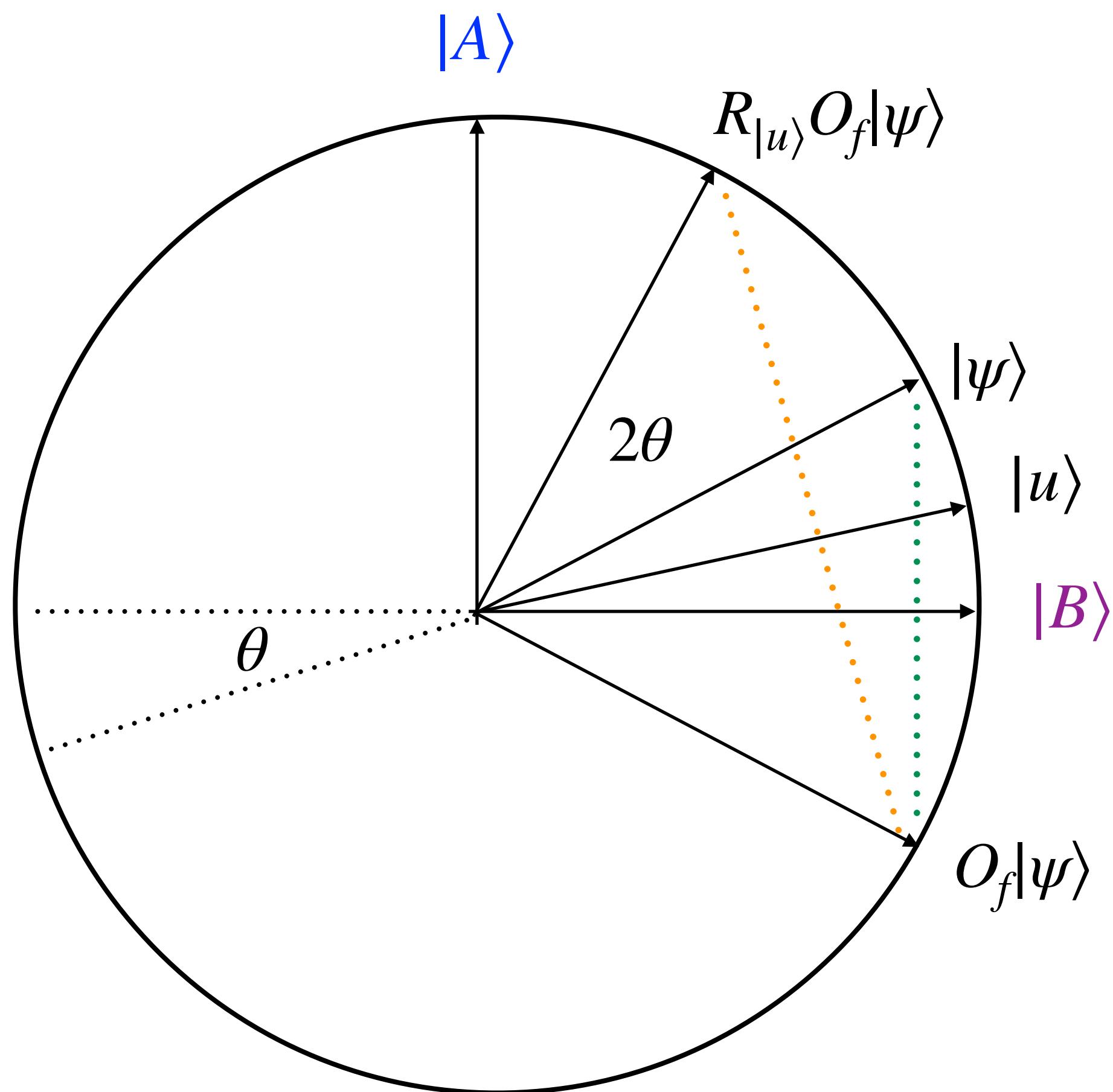
$$R_{|u\rangle}R_{|B\rangle}|A\rangle = -R_{|u\rangle}|A\rangle = -(2|u\rangle\langle u| - I)|A\rangle$$

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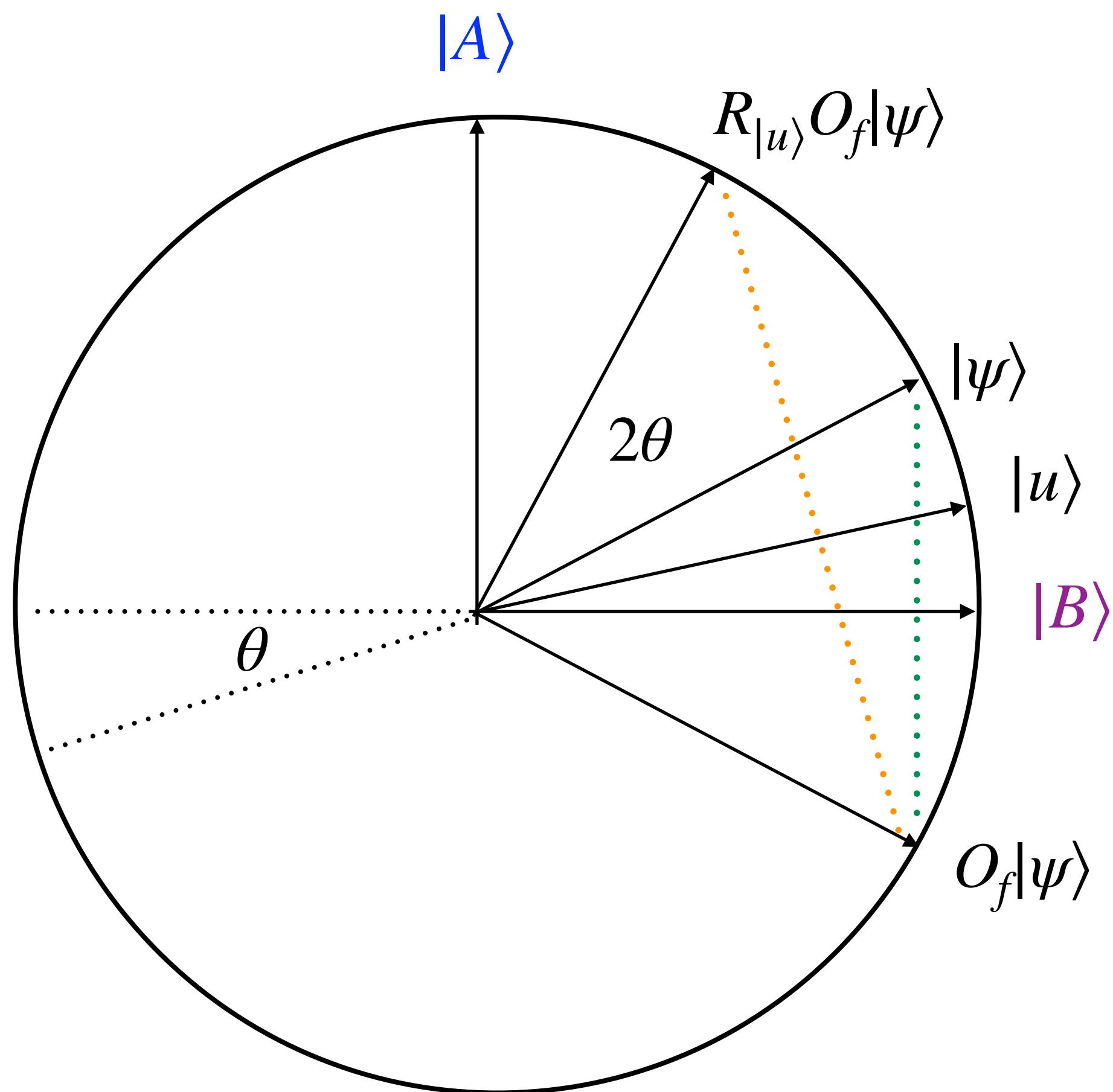
$$\begin{aligned} R_{|u\rangle}R_{|B\rangle}|A\rangle &= -R_{|u\rangle}|A\rangle = -(2|u\rangle\langle u| - I)|A\rangle \\ &= -2\sqrt{\frac{A}{N}}|u\rangle + |A\rangle \end{aligned}$$

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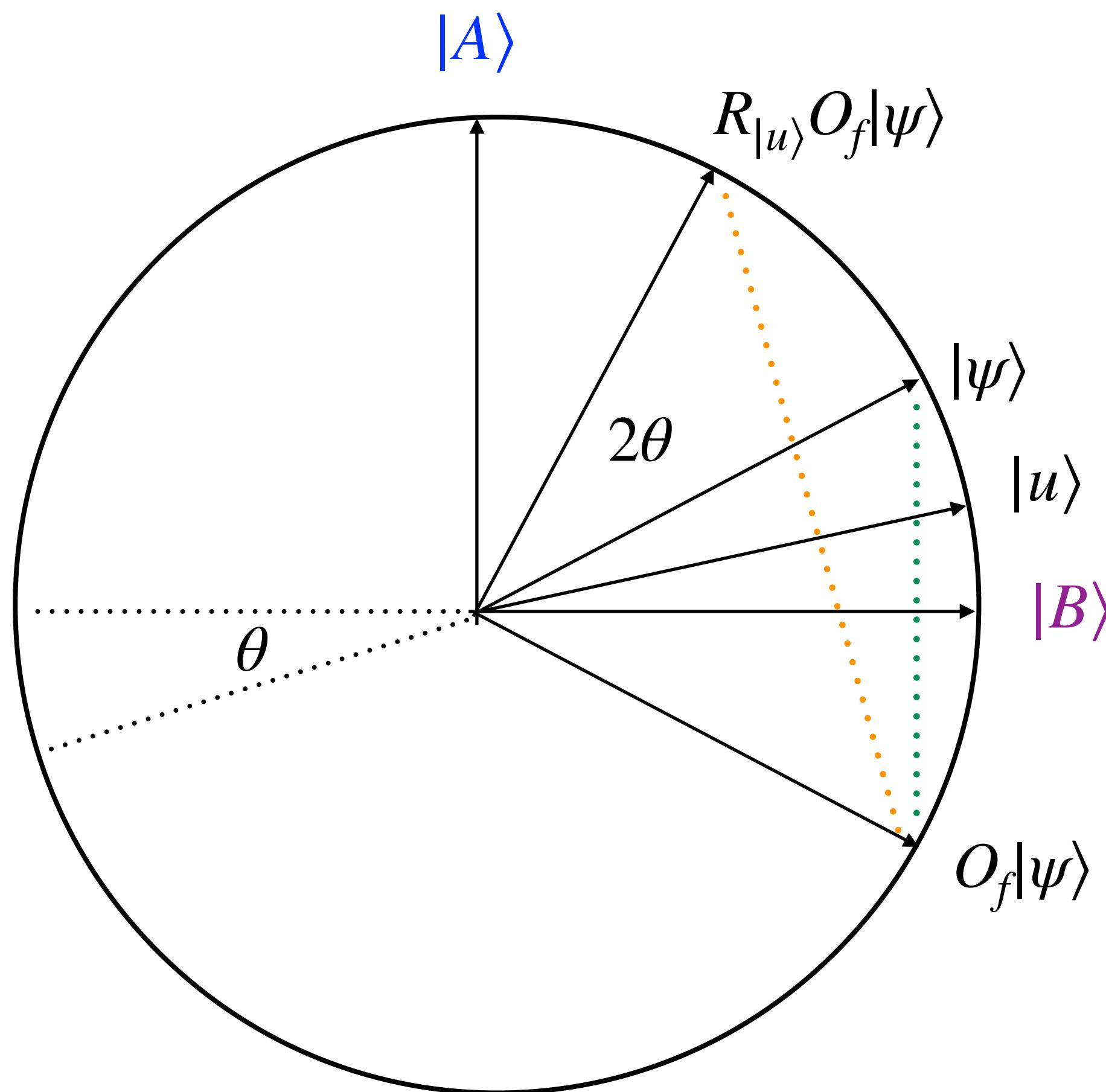
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Fact: $R_{|u\rangle}R_{|B\rangle} = \text{Rot}(2\theta)$

θ is the angle between $|u\rangle$ and $|B\rangle$

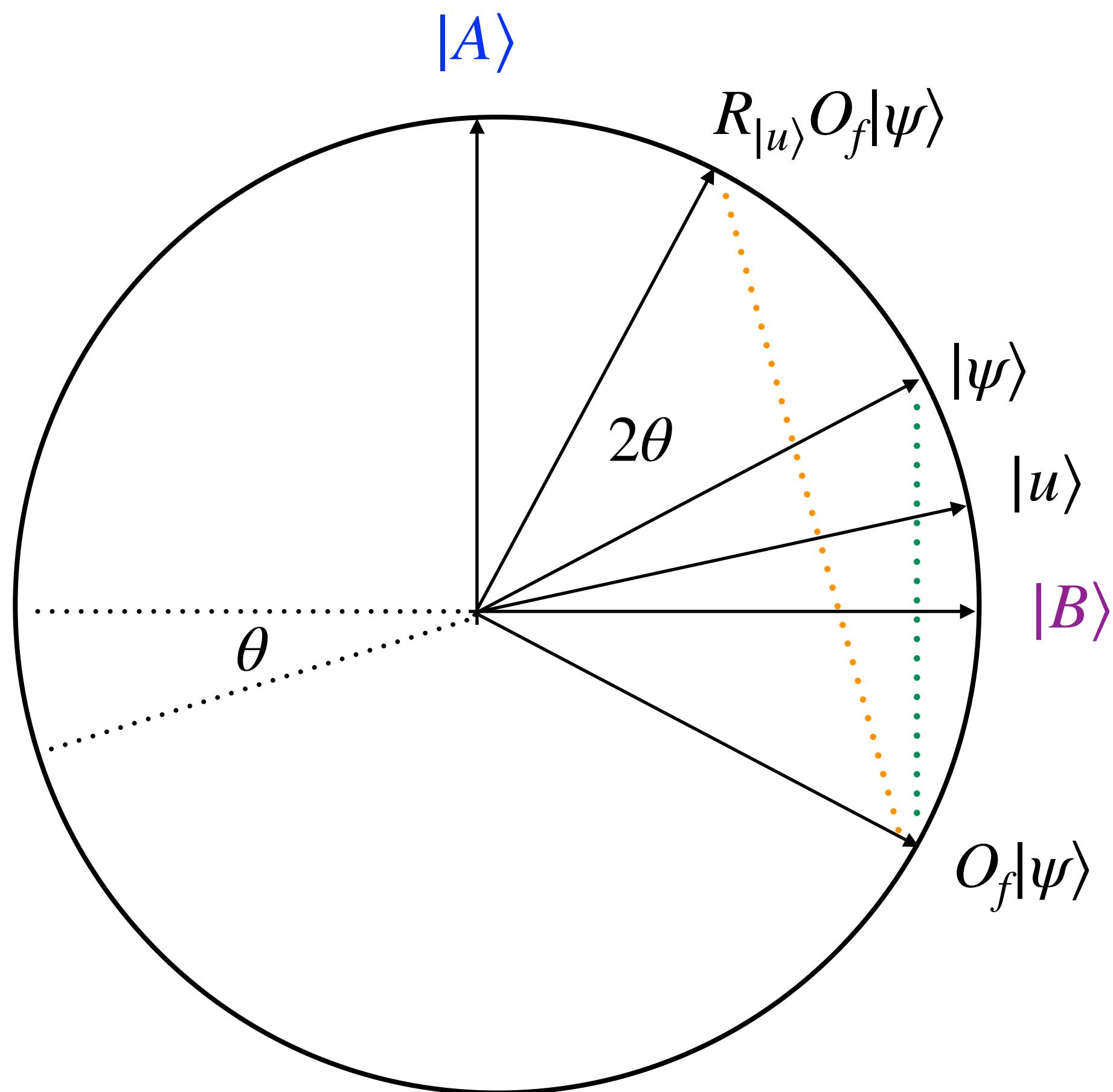
$$\begin{aligned} R_{|u\rangle}R_{|B\rangle}|A\rangle &= -R_{|u\rangle}|A\rangle = -(2|u\rangle\langle u| - I)|A\rangle \\ &= -2\sqrt{\frac{A}{N}}|u\rangle + |A\rangle \\ &= \frac{N-2A}{N}|A\rangle - \frac{2\sqrt{AB}}{N}|B\rangle \\ &= \frac{B-A}{N}|A\rangle - \frac{2\sqrt{AB}}{N}|B\rangle \end{aligned}$$

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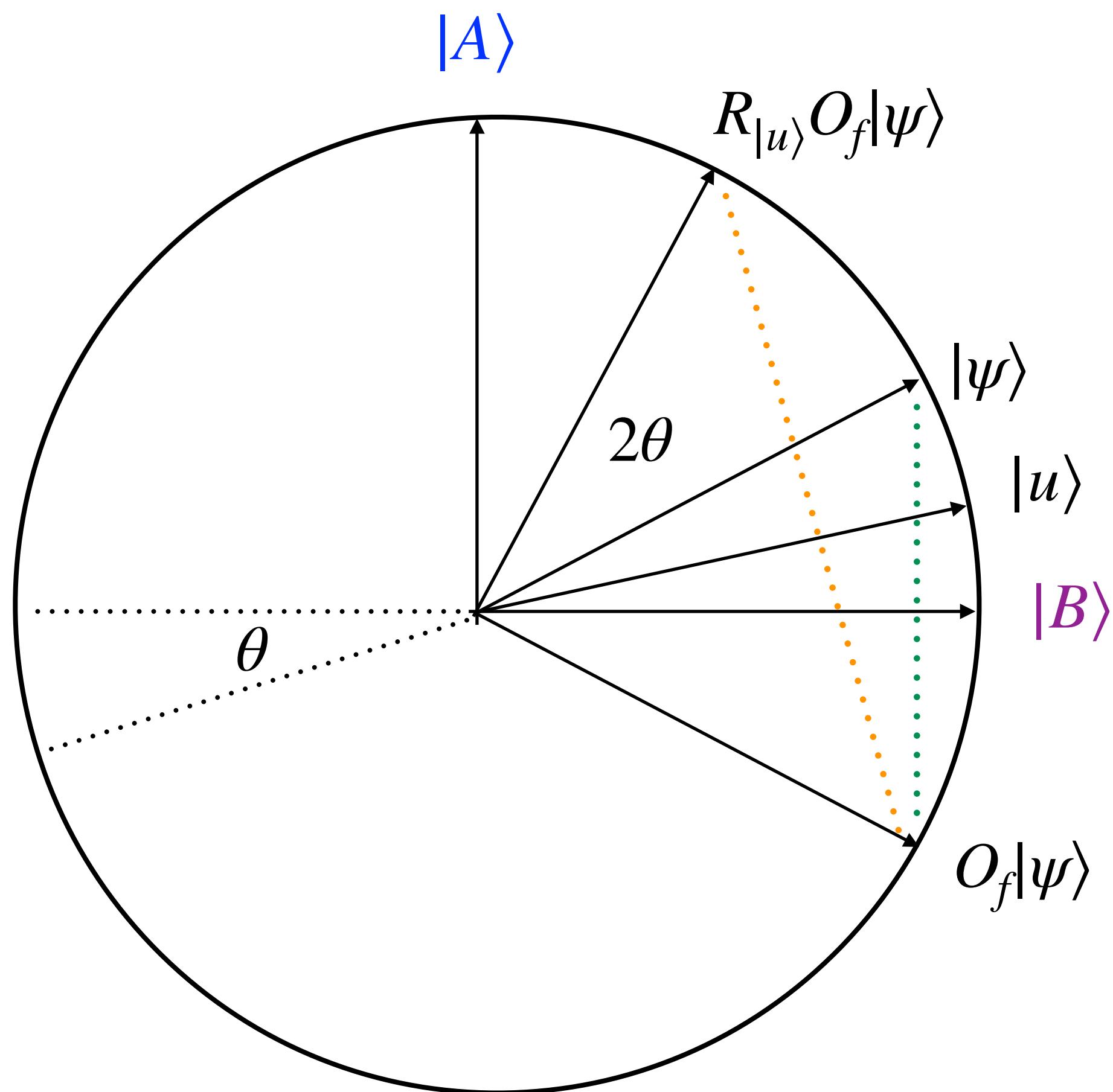
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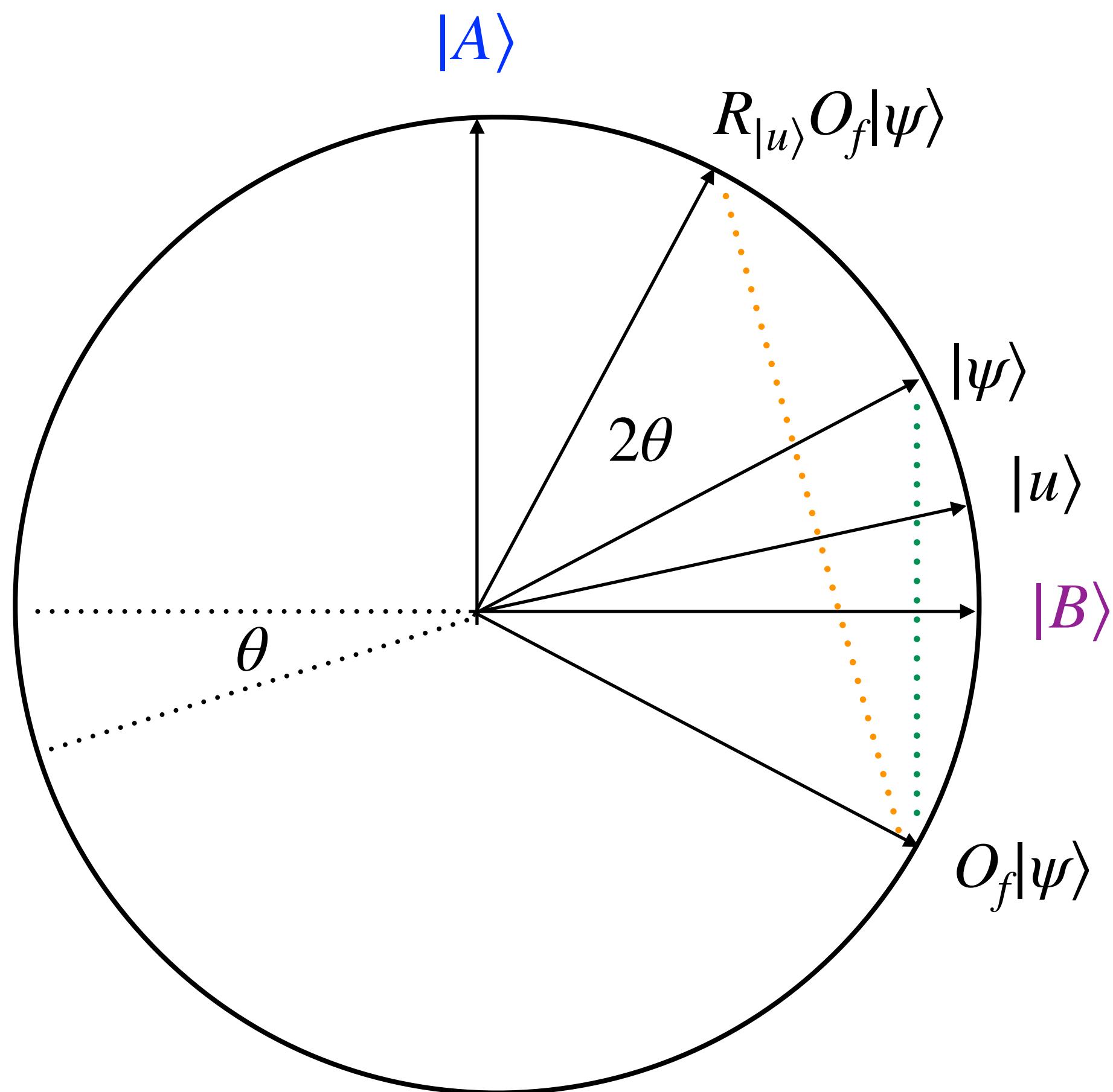
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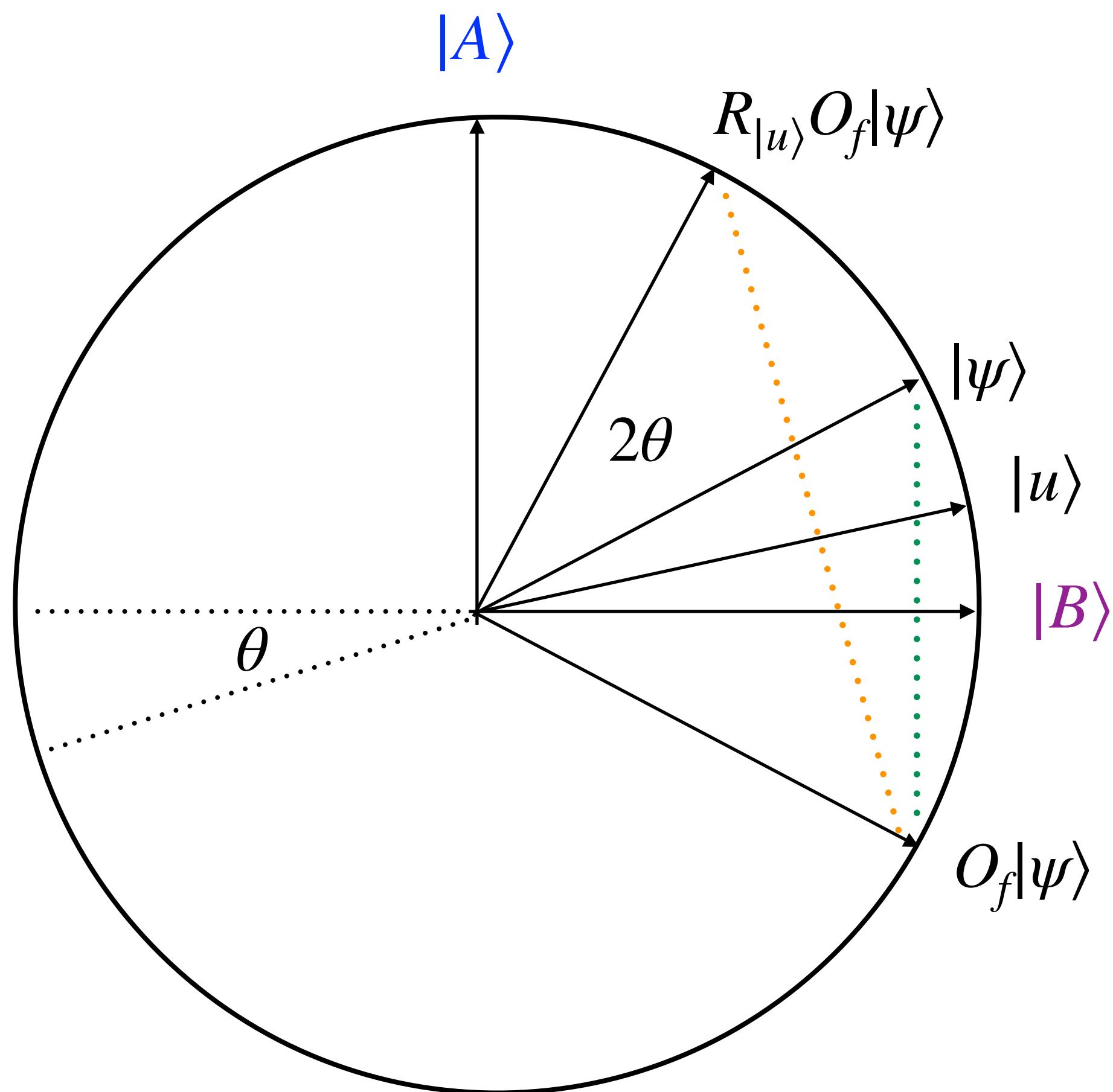
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$|B\rangle$ $|A\rangle$

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■

Can you implement other reflections?

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More Questions?