

# Mathematical Foundations of Quantum Mechanics

From Functional Analysis to Quantum Computing

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Boston University Directed Reading Program

# Introduction

- Sophomore undergraduate studying math and computer science at Boston University
- Interested in quantum computing and mechanics as well as a multitude of different mathematical subjects
- Wanted to explore the mathematical foundations of certain quantum formulas & algorithms normally glossed over in other classes



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If  $\mathcal{X}$  is a vector space in the complex field  $\mathbb{C}$ , an inner product (denoted  $\langle \cdot, \cdot \rangle$  or  $\langle \cdot | \cdot \rangle$ ) for all  $\alpha, \beta \in \mathbb{C}$  and  $x, y, z \in \mathcal{X}$  satisfies the following:

- $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$
- $\langle x, \alpha y + \beta z \rangle = \bar{\alpha} \langle x, y \rangle + \bar{\beta} \langle x, z \rangle$
- $\langle x, x \rangle \geq 0$
- $\langle x, y \rangle = \overline{\langle y, x \rangle}$
- $\langle x, 0 \rangle = \langle 0, y \rangle = 0$
- If  $\langle x, x \rangle = 0$ , then  $x = 0$



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The key thing to note is that these properties all apply to the dot product for finite dimensions.

## Basics of Functional Analysis: Introducing $L^2$

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We use  $L^2$  since it has a lot of nice mathematical and physical properties, one of which is that it behaves a lot like  $\mathbb{R}^n$  in infinite dimensions.

# Schrödinger Equation: Infinite Square Well

The Schrödinger equation states that

$$\begin{aligned}i\hbar \frac{\partial \Psi}{\partial t}(t, x) &= -\frac{\hbar^2}{2m} \frac{\partial^2 \Psi}{\partial x^2}(t, x) + V(x)\Psi(t, x) \\ &= H\Psi(t, x)\end{aligned}$$

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If we assume that time is independent, i.e.  $\Psi(t, x) = \varphi(t)\psi(x)$ , then we can separate this equation and solve

$$\begin{cases} i\hbar \varphi'(t) = E\varphi(t) \\ H\psi(x) = E\psi(x) \end{cases}$$

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The potential of the infinite square well is described by

$$V(x) = \begin{cases} 0 & 0 \leq x \leq a \\ \infty & \text{otherwise} \end{cases}$$

# Schrödinger Equation: Infinite Square Well

We can let  $H : L^2[0, a] \rightarrow L^2[0, a]$  with  $\psi(0) = \psi(a) = 0$ . In this case, we know that  $H = -\frac{\hbar^2}{2m}\partial_{xx}$  in  $[0, a]$ . If we solve the ODE  $-\frac{\hbar^2}{2m}\psi_{xx} = E\psi$ , we find that the eigenvalues and eigenfunctions are

$$E_n = \frac{n^2\pi^2\hbar^2}{2ma^2}, \psi_n(x) = \sqrt{\frac{2}{a}} \sin\left(\frac{n\pi}{a}x\right) \text{ for } n = 1, 2, 3, \dots$$



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These  $\{\psi_n\}_{n=1}^{\infty}$  form an orthonormal basis of  $L^2[0, a]$ , so

$$\Psi(t, x) = \sum_{n=1}^{\infty} \langle \Psi(0, \cdot), \psi_n \rangle e^{-i\frac{E_n}{\hbar}t} \psi_n(x)$$

Therefore, we have solved the Schrödinger equation for the infinite square well case.

## Schrödinger Equation: Free Particle

Let's now suppose that  $V(x) = 0$ . Our previous strategy was to solve the ODE  $-\frac{\hbar^2}{2m}\psi_{xx} = E\psi$  and find an orthonormal eigenbasis  $\{\psi_n\}$  with eigenvalues  $\{E_n\}$  on  $L^2([0, a])$ , so we could write a solution as a linear combination.

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However, with no boundary, any  $E$  will work here. We want to ultimately write

$$\Psi(t, x) = \sum_{n=1}^{\infty} \langle \Psi(0, \cdot), \psi_n \rangle e^{-i\frac{E_n}{\hbar}t} \psi_n(t)$$

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How can we do this? We can use the continuous version of summation  $\implies$  integration!

## Schrödinger Equation: Free Particle

Let  $f \in L^2[0, 2\pi]$  and  $e_k(x) = \frac{1}{\sqrt{2\pi}} e^{ikx}$ , which can be thought of as a period  $\frac{2\pi}{k}$  “frequency”. We can see that

$$f_n = \int_0^{2\pi} f(x) e_k(x) dx \implies f(x) = \sum_{n=0}^{\infty} f_n e_n(x) = \sum_{n=0}^{\infty} \langle f, e_n \rangle e_n(x)$$

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Let  $k = \frac{\sqrt{2mE}}{\hbar}$  and  $\phi(k) = \langle \Psi(0, \cdot), e_k \rangle$ . We want to write over infinite domain

$$\Psi(t, x) = \sum_{k=-\infty}^{\infty} \phi(k) e^{-i \frac{\hbar k^2}{2m} t} \frac{1}{\sqrt{2\pi}} e^{ikx}$$

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Since  $k$  can be any real number, we write this as an integral:

$$\Psi(t, x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \phi(k) e^{i(kx - \frac{\hbar}{2m} k^2 t)} dk$$



# Continuous Fourier Transform

The Fourier transform is defined as

$$\hat{f}(\xi) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-i\xi x} dx \text{ for } \xi \in \mathbb{R}$$

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We can invert the Fourier transform very easily. With

$\mathcal{F} : L^2(\mathbb{R}) \rightarrow L^2(\mathbb{R})$ , we have an inversion formula

$$f(x) = \mathcal{F}^{-1}(\widehat{f})(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \widehat{f}(\xi) e^{i\xi x} d\xi$$

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A few other nice properties:

- Maps derivatives to products:  $\widehat{f^{(n)}}(\xi) = (i\xi)^n \widehat{f}(\xi)$
- Maps products to convolutions:  $\mathcal{F}(fg)(\xi) = (\widehat{f} * \widehat{g})(\xi)$
- Maps convolutions to products:  $\mathcal{F}(f * g)(\xi) = \widehat{f}(\xi) \widehat{g}(\xi)$

# Discrete Fourier Transform

Given data  $(x_0, x_1, \dots, x_{N-1}) \in \mathbb{C}^{N-1}$ , we want to form  $(y_0, y_1, \dots, y_{N-1}) \in \mathbb{C}^{N-1}$  where  $y_k = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} x_j \omega_N^{-jk}$  and  $\omega_N := e^{i\frac{2\pi}{N}}$ . This is a discrete variation of the Fourier transform we just defined.

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In quantum computing, the inverse of this operation is known as the quantum Fourier transform. If we have a quantum state  $|x\rangle = \sum_{j=0}^{N-1} x_j |j\rangle$ , the QFT maps  $|x\rangle \mapsto |y\rangle = \sum_{k=0}^{N-1} y_k |k\rangle$ , where  $y_k = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} x_j \omega_N^{jk}$ .

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The QFT can be thought of as a unitary matrix mapping

$$F_N = \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega_N & \omega_N^2 & \dots & \omega_N^{N-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega_N^{N-1} & \omega_N^{2(N-1)} & \dots & \omega_N^{(N-1)(N-1)} \end{bmatrix}$$

# Quantum Phase Estimation

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Assume that  $\theta = 2\pi j/2^m$  for some integer  $j$ . Then, we know the inverse QFT maps

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In fact, regardless of the value of  $\theta$ , the output  $|j\rangle$  turns out to be a very good approximation.

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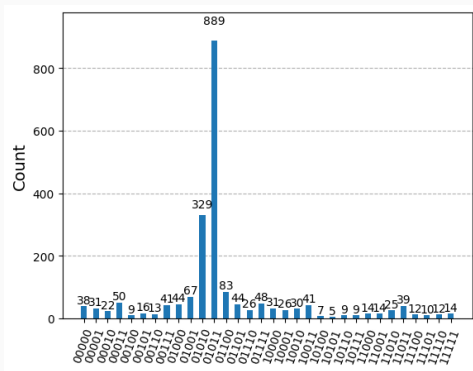
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We then discard the  $|v_\theta\rangle$  register. We can then simply apply the inverse QFT to get an output  $|j\rangle$  such that  $\theta \approx 2\pi j/2^m$ .

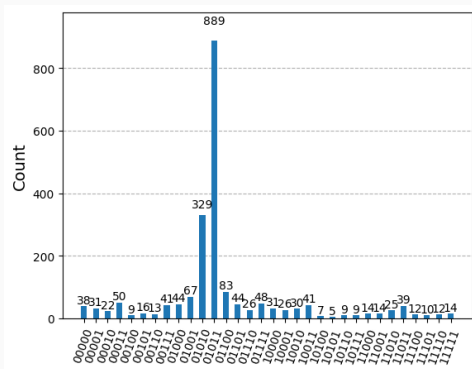
# QPE Example

Given this real quantum output plot, what can we say about  $\theta$ ?



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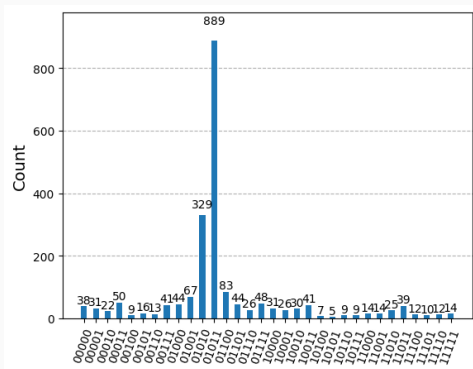
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The largest  $|j\rangle$  is  $|01011\rangle = 11$  so  $\theta \approx 22\pi/32$ . This is probably a bit of an overestimation since  $|01010\rangle$  is second largest.

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In fact, the actual  $\theta = 2\pi/3$ .



**Thank you for listening!**

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Let me know if you have any questions!