

The Concept of Duality

From Wave-Particle Physics to Generative AI Architectures

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1. The Origin: Wave-Particle Duality

The Fundamental Paradox of Quantum Mechanics

- **Particle View:** Matter exists as discrete points (e.g., electrons, photons hitting a screen). Localized in space.
- **Wave View:** Matter exhibits interference and diffraction. Spread out over space.

The Core Duality

An entity is not exclusively a wave or a particle, but behaves as one or the other depending on the measurement apparatus.

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- **No Definite Trajectory:** It does not follow a single path (Point A → Point B).
- **Interference:** It can interact with itself, canceling out or amplifying its own probability of existing in certain locations.

2. The Double Slit Experiment Setup

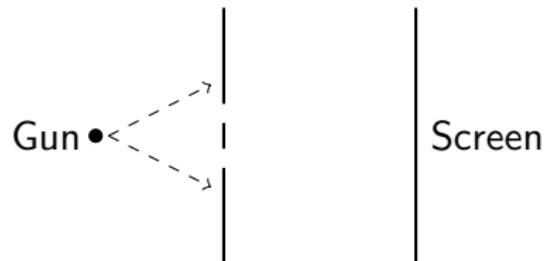
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The Apparatus:

- ① **Source:** An electron gun firing one electron at a time.
- ② **Barrier:** A wall with two slits (Slit A and Slit B).
- ③ **Detector:** A back screen that records where the electron lands.



3. Scenario A: Observed (Particle Behavior)

The Condition: We place a sensor at the slits to track which hole the electron goes through.

- **Observation causes Collapse:** The act of measuring forces the wave function to "choose" a state.
- **The Result:** The electron acts like a bullet. It goes through **either** Slit A **or** Slit B.

The Pattern on the Wall: Two distinct bands of hits. (No interference).

4. Scenario B: Unobserved (Wave Behavior)

The Condition: We do **not** check which slit the electron uses. We only check where it lands at the end.

- **Superposition:** The electron passes through **both** Slit A and Slit B simultaneously as a wave.
- **Self-Interference:** The "ripple" from Slit A crashes into the "ripple" from Slit B.

The Pattern on the Wall: An **Interference Pattern**. Alternating bright stripes (many hits) and dark stripes (no hits).

5. The Big Question

The Paradox

"If observing destroys the wave, how do we know it was ever a wave?"

Answer: We don't see the wave itself; we see the **footprints** it leaves behind.

We deduce the journey (Wave) from the destination (Interference Pattern).

6. The Proof: The Dark Spots

The strongest evidence is not where the electrons *are*, but where they *are not*.

- **Particle Logic:** Opening a second door should increase the chance of hitting any spot on the wall. ($P_{Total} = P_A + P_B$).
- **Wave Logic:** Opening a second door creates interference. Peaks meet troughs and cancel out. ($P_{Total} = |\Psi_A + \Psi_B|^2$).

Conclusion: Because we see "Dark Spots" (areas that get hit *less* when the second slit is open), we know the particle must have traveled through both slits to cancel itself out.

Summary of Duality

Feature	Observed	Unobserved
Nature	Particle (Bullet)	Wave (Probability Cloud)
Path	Definite (A or B)	Indefinite (A and B)
Result	Two Bands	Interference Pattern
Mathematics	Probability Sum	Wave Function Collapse

2. The Heisenberg Uncertainty Principle

This physical duality leads to a fundamental mathematical limit in precision.

$$\sigma_x \sigma_p \geq \frac{\hbar}{2}$$

Translated to Signal Processing (Time-Frequency):

- You cannot perfectly localize a signal in both **Time** and **Frequency** simultaneously.
- **Narrow Time Window** → Wide Frequency Band (Impulse).
- **Narrow Frequency Band** → Wide Time Window (Pure Sine Wave).

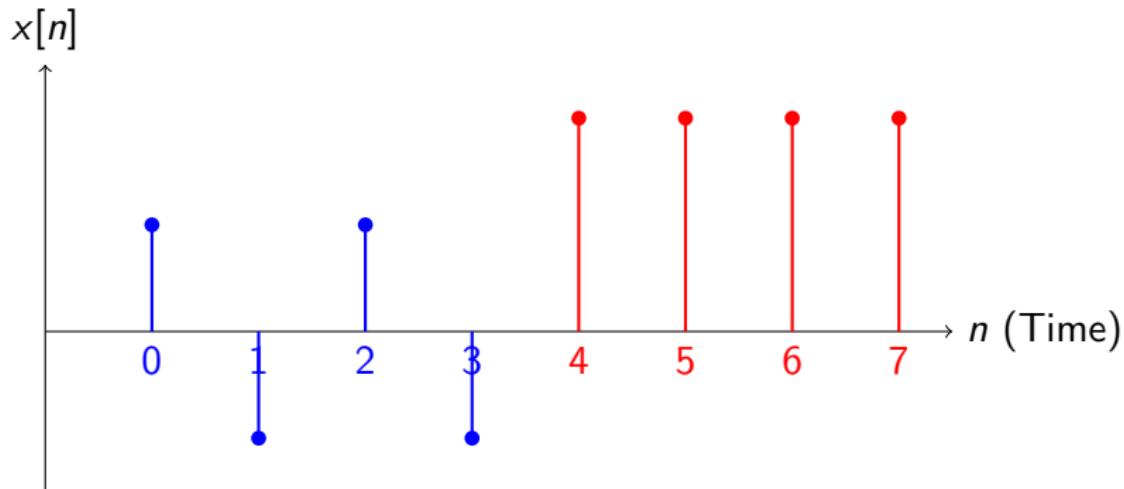
*"To know **when** a note happens, you lose precision on exactly **what note it is**."*

3. Example: Discrete Data (Time Domain)

Let us visualize a simple discrete signal $x[n]$ with 8 points.

Signal: A high-frequency burst followed by a static value.

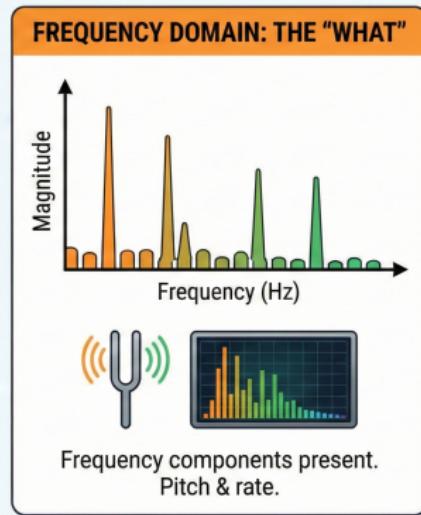
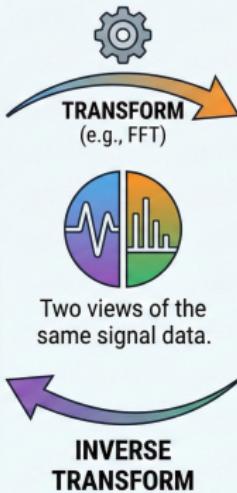
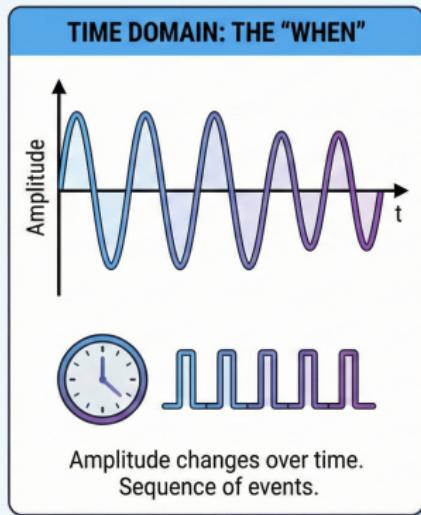
$$x = [1, -1, 1, -1, 2, 2, 2, 2]$$



Distinct "events" happen at distinct indices.

Visualization

TIME & FREQUENCY DOMAIN TRANSFORM: THE DUALITY BRIDGE



4. Example: Discrete Data (Frequency Domain)

If we switch to the Frequency Domain (Spectral View), we stop looking at "when" and look at "how fast".

Component A (First Half):

- Alternating $[1, -1 \dots]$
- Max Frequency (Nyquist).

Component B (Second Half):

- Constant $[2, 2 \dots]$
- Zero Frequency (DC Bias).

The Trade-off: A global Frequency Transform would just say: "*This signal contains both High Frequencies and DC bias.*" It obscures the fact that they occurred at different times.

5. The Mapping Mechanism: FFT & IFFT

The **Fast Fourier Transform (FFT)** is the operator that maps us between these dual worlds.

Forward Transform (Time → Frequency)

Decomposes the signal into a sum of spinning sinusoids.

$$X[k] = \sum_{n=0}^{N-1} x[n] \cdot e^{-i2\pi \frac{kn}{N}}$$

Inverse Transform (Frequency → Time)

Reconstructs the time signal by summing the frequencies.

$$x[n] = \frac{1}{N} \sum_{k=0}^{N-1} X[k] \cdot e^{i2\pi \frac{kn}{N}}$$

This is a lossless, reversible change of basis.

6. Hypothesis: Duality in Generative AI

We can extend this Time-Frequency duality to understand modern Generative Models.

LLMs
(Auto-regressive on Tokens)
"The Time Domain"

Diffusion Models
(Iterative Refinement)
"The Frequency Domain"

- **LLMs** generate sequentially ($t_1 \rightarrow t_2 \rightarrow t_3$).
- **Diffusion** generates structurally (Coarse → Fine).

7. LLMs: The Time Domain Perspective

Large Language Models operate like a time-series signal.

$$P(x) = \prod_{t=1}^T P(x_t | x_{<t})$$

- **Sequential Causality:** The generation of token x_t depends strictly on the discrete history $x_1 \dots x_{t-1}$.
- **Local Dependency:** Like a time-domain signal, an error at t propagates forward in time.
- **Discrete Nature:** The output is a sequence of discrete "impulses" (tokens) placed one after another on the time axis.

8. Diffusion: The Frequency Domain Perspective

Diffusion models can be viewed as **Autoregressive in the Frequency Domain.**

- **Spectral Evolution:** Diffusion generation starts from pure noise (high entropy, all frequencies) and slowly resolves the signal.
- **Coarse-to-Fine:**
 - ➊ Early steps resolve global structure (Low Frequencies).
 - ➋ Later steps resolve texture and edges (High Frequencies).
- **The Duality:** Just as FFT builds a signal by stacking sine waves, Diffusion builds an image by stacking "detail levels."

Hypothesis: While LLMs predict the "next moment in time", Diffusion predicts the "next level of clarity" (Frequency).

Conclusion

Summary of Duality

Concept	Domain A (Localized)	Domain B (Distributed)
Physics	Particle	Wave
Signal	Time (t)	Frequency (ω)
Gen AI	LLM (Token-wise)	Diffusion (Spectral-wise)

Understanding this duality allows us to apply signal processing techniques (like spectral filtering) to improve Diffusion models, or causal masking techniques to improve LLMs.

The Game: "Guess the Function"

The Rules:

- ① I will give you a set of input points: $x \in [-2, 2]$.
- ② I will give you the **slope** (gradient) of an unknown function $f(x)$ at those points.
- ③ Your job is to identify the function $f(x)$.

Hypothesis: If we know all the slopes, do we know the function?

Round 1: The Easy One

Data Provided:

Point (x)	Slope (y)
-2	-4
-1	-2
0	0
1	2
2	4

Student Reasoning:

- The slope y is directly proportional to x .
- Equation: $f'(x) = 2x$.
- Integrate: $f(x) = x^2$ (ignoring constant C).

Answer: $f(x) = x^2$

Status: Unique mapping between x and slope y .

Round 2: The Curve Ball

Data Provided:

Point (x)	Slope (y)
0.5	-2
1	-1
2	-0.5
4	-0.25

Student Reasoning:

- As x increases, slope becomes less negative.
- Relationship: $y = -1/x$.
- Equation: $f'(x) = -\frac{1}{x}$.
- Integrate: $f(x) = -\ln(x)$.

Answer: $f(x) = -\log(x)$

Status: Still a unique, one-to-one mapping.

Round 3: The Subgradient Hint

Data Provided:

Point (x)	Slope (y)
-2	-1
-0.5	-1
0.001	1
2	1

Student Reasoning:

- The slope is constant on the left (-1).
- The slope is constant on the right (+1).
- Sudden jump at 0.

Answer: $f(x) = |x|$

Status: *Mapping is unique almost everywhere.*

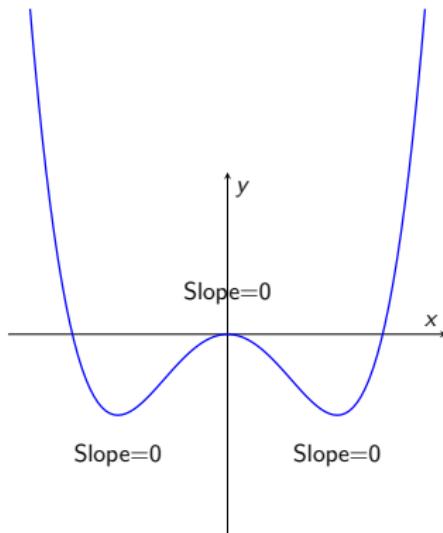
Round 4: The Problem (Non-Convexity)

Data Provided:

"I have a function where the slope is 0."

Possible Candidates for x :

- If $f(x) = x^2$, then $x = 0$.
(Unique!)
- If $f(x) = \cos(x)$, could be
 $x = 0, 2\pi, -2\pi \dots$
- If $f(x) = x^4 - 2x^2$ (Double Well), could be $x = 0, 1, -1$.



The Failure: Knowing the slope y does NOT uniquely identify the state x if the function is non-convex.

Motivation for Conjugate Functions

We saw that for nicely behaved (convex) functions, specifying the **slopes** is just another way of describing the function.

The Dual Perspective

Instead of describing a function as a set of points $(x, f(x))$, can we describe it as a set of tangent lines (slopes and intercepts)?

This "description by slopes" is exactly what the **Conjugate Function** $f^*(y)$ provides.

$$f^*(y) = \sup_x (yx - f(x))$$

It tries to reconstruct the function from its slopes. Visual:

<https://misterpawan.github.io/learn/optimization-methods/convex-functions.html>

The "Recovery" Gap

What happens if we try to recover the original function from its slope description?

Convex Case (x^2)

- Slopes map uniquely to points.
- $f^{**} = f$
- Perfect recovery.

Non-Convex Case (Double Well)

- Multiple points share slopes.
- The conjugate "fills in" the bumps.
- $f^{**} = \text{ConvexEnvelope}(f) \neq f$
- Information is lost!

Conclusion: Conjugate functions naturally "convexify" problems.