# Quantum Neuroscience, Subjective Preferences, and Stock Market Dynamics: A Unified Framework

Heng-Fu Zou April 1, 2025

#### Abstract

This paper develops a unified framework linking subjective preferences, stock market behavior, and quantum neuroscience, arguing that financial decision-making originates in quantum cognitive processes rather than classical neural determinism. Preferences, judgments, and ideas are modeled as quantum states evolving within a cognitive Hilbert space, governed by a preference Hamiltonian. These quantum states—subject to superposition, tunneling, entanglement, and uncertainty—explain why investor behavior is inherently probabilistic, context-dependent, and often nonrational. Market prices emerge as observable outcomes of wavefunction collapse across interacting agents, while crashes and bubbles are modeled through quantum tunneling and collective decoherence. We derive a quantum uncertainty principle showing that evaluation volatility and risk perception are fundamentally bounded. Anomalies observed in behavioral and experimental economics, including framing effects and preference reversals, are explained through non-commuting cognitive operators and dynamic, operator-valued utilities. This framework reinterprets market irrationality as a natural consequence of quantum consciousness and provides a rigorous, empirically consistent theory of financial behavior.

# 1 Introduction

The traditional paradigms of financial economics, founded upon rational expectations and efficient market hypotheses, posit that stock market prices are primarily determined by the aggregation of objective information regarding firms' fundamentals, market conditions, and macroeconomic indicators. In these models, human agents are treated as rational calculators optimizing well-defined utility functions, and market prices are assumed to reflect the intrinsic value of assets. However, decades of empirical anomalies, speculative bubbles, excessive volatility, and experimental evidence have exposed the inadequacy of these classical approaches in capturing the true complexity of stock markets. Behavioral economics, pioneered by scholars such as Kahneman and Tversky,

has challenged the assumption of rationality, introducing biases, heuristics, and subjective evaluations as essential factors in decision-making. Experimental economics has further highlighted that real-world markets, even in controlled environments, deviate significantly from theoretical equilibria due to human subjectivity, framing effects, and social influences.

Yet, these behavioral and experimental insights, though profound, remain insufficient in explaining the fundamental source of subjectivity, indeterminacy, and preference formation in financial markets. This paper argues that the root cause of market dynamics lies deeper—in the neurocognitive processes governing human consciousness, preference evolution, and idea formation. Specifically, we propose that quantum neuroscience, which posits that cognitive processes at the synaptic and microtubule level are inherently probabilistic and governed by quantum principles, offers a robust explanatory framework. The stochastic, superposed, and entangled nature of neural activity leads to the emergence of subjective preferences, which in turn influence stock market prices. In essence, stock market dynamics are emergent properties of quantum-level cognitive processes distributed across millions of market participants.

By integrating quantum neuroscience with behavioral economics, experimental findings, and financial modeling, we develop a unified framework wherein stock prices are not fixed reflections of objective reality but are dynamically shaped by the fluctuating, superposed, and entangled subjective evaluations of market actors. In the sections that follow, we formalize this intuition mathematically, demonstrating how quantum concepts—such as superposition, tunneling, entanglement, and uncertainty—manifest in market phenomena like bubbles, crashes, volatility, and herding behavior. We argue that any comprehensive model of financial markets must incorporate the probabilistic and fundamentally subjective nature of human consciousness as articulated through quantum neuroscience.

# 2 Subjective Preferences and Stock Market Prices: Behavioral Foundations

At the core of traditional financial theory lies the assumption of the rational, utility-maximizing investor. Classical models such as the Capital Asset Pricing Model (CAPM) and the Efficient Market Hypothesis (EMH) treat preferences as exogenous, stable, and objectively measurable. Stock prices, within this framework, reflect the discounted present value of expected future cash flows, based on all available information. However, decades of empirical research have revealed persistent anomalies that cannot be reconciled within this framework: excessive volatility, momentum effects, market bubbles, and crashes, among others.

Behavioral economics emerged as a response to these deficiencies, positing that individuals' subjective evaluations, biases, and heuristics fundamentally shape their financial decisions. Pioneering work by Kahneman and Tversky introduced Prospect Theory, which demonstrated that individuals evaluate outcomes relative to a reference point and exhibit loss aversion—valuing losses more heavily than equivalent gains. Subjective preferences, under this view, are context-dependent and subject to framing effects. For instance, an investor's evaluation of a stock may change depending on how information is presented. leading to systematic deviations from "rational" pricing models.

To formalize this idea, consider a simple utility function under Prospect Theory:

$$U(x) = \begin{cases} (x-r)^{\alpha} & \text{if } x \ge r \\ -\lambda (r-x)^{\beta} & \text{if } x < r \end{cases}$$

 $U(x) = \begin{cases} (x-r)^{\alpha} & \text{if } x \geq r \\ -\lambda (r-x)^{\beta} & \text{if } x < r \end{cases}$  where x is the financial outcome, x is the reference point, x < 0 or x < 0. represent diminishing sensitivity, and  $\lambda > 1$  captures loss aversion. Unlike traditional concave utility functions, the kink at r embodies subjective preference distortions. This functional form inherently leads to price-setting behavior that diverges from objective valuation.

Moreover, experimental economics, through controlled laboratory settings, has confirmed that markets populated by human participants deviate significantly from theoretical equilibria. Vernon Smith's experimental markets have demonstrated price bubbles even in environments where fundamentals are common knowledge. The subjective preferences and bounded rationality of participants introduce systematic inefficiencies and volatility.

However, while behavioral and experimental economics expose the manifestations of subjective preferences, they often leave the source of subjectivity unexplained. Why do preferences shift dynamically? Why do agents evaluate the same asset differently under varying contexts? Behavioral economics describes the patterns, but it lacks a theory of how consciousness, preferences, and evaluations emerge at a fundamental level.

This leads us to a deeper ontological question: Where do subjective preferences originate? Why are they inherently indeterminate, fluctuating, and context-sensitive? The answer, we argue, lies in the probabilistic, non-deterministic nature of human cognition—specifically, the quantum processes underpinning neural activity. In the next section, we delve into experimental economics' insights and how they lay the groundwork for connecting subjective evaluations to quantum neural dynamics.

# 3 Experimental Economics and Market Anoma-

Experimental economics has provided compelling empirical evidence that realworld market behavior systematically deviates from the predictions of classical financial models. By simulating market environments in controlled laboratory settings, researchers have been able to isolate specific factors influencing human decision-making, particularly the role of subjective preferences, framing, and social dynamics. These experiments reveal that market anomalies—such as bubbles, crashes, and persistent inefficiencies—are not aberrations but are deeply rooted in the cognitive processes of human participants.

One of the most striking findings in experimental economics is the phenomenon of speculative bubbles in markets with known fundamentals. Vernon Smith's seminal asset market experiments in the 1980s demonstrated that even when all participants have complete information about the fundamental value of an asset, prices often deviate significantly from intrinsic values. Prices tend to rise well above fundamental levels, followed by abrupt crashes. Traditional theories cannot account for these dynamics; instead, they highlight the critical role of subjective beliefs, adaptive expectations, and social contagion.

Mathematically, we can formalize an experimental market scenario as follows. Suppose N traders participate in a market for a single asset with known fundamental value  $V_f$ . Each trader i forms a subjective belief about the price, denoted by  $P_i(t)$ , which evolves over discrete time steps t. The market price P(t) at time t is the aggregation of all individual beliefs:

$$P(t) = \frac{1}{N} \sum_{i=1}^{N} P_i(t)$$

 $P(t) = \frac{1}{N} \sum_{i=1}^{N} P_i(t)$ However, each trader's belief is influenced not only by the fundamental value but also by the observed behavior of others, personal risk preferences, and cognitive biases. We model each  $P_i(t)$  as:

$$P_i(t+1) = \gamma V_f + (1-\gamma) \left[ \eta_i(t) + \frac{1}{N-1} \sum_{j \neq i} P_j(t) \right]$$

where  $0 < \gamma < 1$  represents the weight given to fundamentals,  $\eta_i(t)$  captures trader i's idiosyncratic subjective evaluation (bias, overconfidence, framing), and the second term reflects social influence through averaging other traders' price expectations. This simple adaptive expectation model leads to complex dynamics: small shifts in  $\eta_i(t)$  can cascade through social influence, amplifying divergences from fundamentals and resulting in bubbles or crashes.

What is crucial here is that the individual subjective term  $\eta_i(t)$  is not fixed or purely stochastic noise—it reflects internal, cognitively-processed preferences and evaluations that vary over time. Experiments consistently show that these preferences are context-dependent, sensitive to framing, prior outcomes, and group behavior. This introduces indeterminacy in price formation: the same fundamental value  $V_f$  may lead to drastically different market outcomes depending on the collective subjective evaluations.

Yet experimental economics, while invaluable in demonstrating these effects, leaves the micro-foundations of preference formation underexplored. It describes how preferences shift and aggregate but not why subjective evaluations arise in probabilistic, fluctuating ways. To address this, we turn to quantum neuroscience, which provides a deeper, biologically grounded explanation of the inherent indeterminacy and context-sensitivity of human decision-making processes. In the next section, we will introduce the neural and quantum mechanisms underpinning consciousness and subjective preferences, laying the foundation for their role in financial markets.

# 4 Quantum Neuroscience: Consciousness, Ideas, and Subjectivity

To understand the origin of subjective preferences and their dynamic evolution, we must move beyond classical cognitive models, which treat the brain as a deterministic information processor. Quantum neuroscience offers a radical but increasingly plausible framework, suggesting that the fundamental processes governing human cognition and consciousness are quantum-mechanical in nature. This perspective holds profound implications for how subjective preferences—crucial determinants of stock market prices—are formed and fluctuate.

# 4.1 Quantum Brain Hypothesis: Theoretical Foundations

The Quantum Brain Hypothesis, advanced by researchers like Stuart Hameroff and Roger Penrose, posits that microtubules within neurons exhibit quantum coherence. These microtubules, cylindrical protein structures, form part of the cytoskeleton and are thought to support quantum superpositions and entanglement at the sub-neuronal level. Unlike classical neural network models, which assume definite firing states, the quantum model asserts that neuronal activity can exist in a superposed state, where multiple subjective evaluations or mental states coexist until a decision or observation collapses the cognitive wavefunction.

To formalize this, consider the state of an individual neuron's microtubule network as a quantum state vector:

$$|\psi_i\rangle = \alpha_i|0\rangle + \beta_i|1\rangle$$

where  $|0\rangle$  represents the neuron in a non-firing state,  $|1\rangle$  represents the firing state, and  $\alpha_i$ ,  $\beta_i$  in  $\mathbb{C}$  are complex probability amplitudes satisfying  $|\alpha_i|^2 + |\beta_i|^2 = 1$ .

The brain, comprising N such neurons, can be described by a collective tensor product state:

$$|\Psi_{\mathrm{brain}}\rangle = \bigotimes_{i=1}^{N} |\psi_i\rangle$$

Crucially, this state is non-factorizable if entanglement exists between neurons, meaning the global state cannot be decomposed into independent neuron states. Such entanglement leads to the formation of complex ideas, preferences, and subjective evaluations as emergent, non-local properties.

# 4.2 Subjective Preferences as Superposed Cognitive States

We posit that subjective preferences in financial decision-making correspond to superposed cognitive states. An investor evaluating whether to buy, hold, or sell a stock holds a mental superposition of multiple subjective evaluations. Mathematically, we model the investor's preference state as:

$$|\Psi_{\text{pref}}\rangle = c_B|B\rangle + c_H|H\rangle + c_S|S\rangle$$

where  $|B\rangle$ ,  $|H\rangle$ , and  $|S\rangle$  represent the cognitive states corresponding to "Buy", "Hold", and "Sell", respectively, and  $c_B$ ,  $c_H$ ,  $c_S$  are complex amplitudes with:

$$|c_B|^2 + |c_H|^2 + |c_S|^2 = 1$$

The actual decision (observable preference) is determined when the wavefunction collapses, typically triggered by an external event (e.g., news, market signals) or internal cognitive resolution. The probabilistic nature of the outcome reflects the inherent indeterminacy of subjective evaluation.

# 4.3 Quantum Hamiltonian for Preference Evolution

The evolution of subjective preferences over time can be described using a preference Hamiltonian operator, denoted  $\hat{H}_{\text{pref}}$ , governing the dynamics of the state vector:

 $i\hbar \frac{d}{dt} |\Psi_{\text{pref}}(t)\rangle = \hat{H} \text{pref} |\Psi_{\text{pref}}(t)\rangle$ 

A simple Hamiltonian model incorporating cognitive energy levels could be:

$$\hat{H}\text{pref} = \begin{pmatrix} E_B & \kappa BH & \kappa_{BS} \\ \kappa_{BH}^* & E_H & \kappa_{HS} \\ \kappa_{BS}^* & \kappa_{HS}^* & E_S \end{pmatrix}$$

where:

- $E_B$ ,  $E_H$ ,  $E_S$  represent cognitive energies associated with Buy, Hold, and Sell evaluations (subject to mood, framing, memory, and risk preference),
- $\kappa_{ij}$  terms capture the coupling and transition probabilities between states (reflecting susceptibility to external signals, internal biases, or social influence).

The time evolution solution is:

$$|\Psi_{\rm pref}(t)\rangle = e^{-\frac{i}{\hbar}\hat{H}{\rm pref}t}|\Psi_{\rm pref}(0)\rangle$$

This formulation reveals that subjective evaluations fluctuate probabilistically over time, even in the absence of new external information, due to intrinsic quantum cognitive dynamics.

# 4.4 Explicit Example: Preference Oscillation

Consider a simplified case where only Buy and Sell states are relevant, reducing the system to:

$$\begin{split} |\Psi_{\mathrm{pref}}\rangle &= c_B |B\rangle + c_S |S\rangle \\ \text{with Hamiltonian:} \\ \hat{H}_{\mathrm{pref}} &= \begin{pmatrix} E_B & \kappa \\ \kappa^* & E_S \end{pmatrix} \end{split}$$

Diagonalizing this Hamiltonian yields eigenstates corresponding to stable preference modes, while the off-diagonal  $\kappa$  terms induce oscillations between Buy and Sell preferences. The probability of switching preference oscillates over time, analogous to Rabi oscillations in quantum systems:

$$P_{\mathrm{Buy}}(t) = |c_B(t)|^2 = \frac{|\kappa|^2}{\Omega^2} \sin^2\left(\frac{\Omega t}{2\hbar}\right) + \text{constant terms}$$
  
where  $\Omega = \sqrt{(E_B - E_S)^2 + 4|\kappa|^2}$  is the generalized Rabi frequency.

Thus, even absent external information, the investor's subjective evaluation oscillates probabilistically, highlighting the inherent dynamism and indeterminacy of preferences: Quantum neuroscience provides the fundamental foundation for subjective preferences, demonstrating that preferences are not deterministic but evolve probabilistically due to underlying quantum neural mecha-

nisms. This explains why preferences fluctuate, aggregate non-linearly, and lead to emergent phenomena in stock markets.

### 5 Quantum Superposition of Preferences and Market States

Having established that individual subjective preferences are inherently quantum, existing in superposed cognitive states governed by probabilistic Hamiltonian dynamics, we now extend this framework to market-level price formation. Stock markets, composed of millions of individual investors, can be viewed as complex systems where aggregate market states emerge from the collective superposition and interactions of individual subjective evaluations. In this section, we construct a formal quantum model to capture how subjective superpositions influence stock prices.

#### 5.1Aggregate Market State as Superposed System

Let us denote the preference state of the i-th investor at time t as:

```
|\Psi_i(t)\rangle = c_{B_i}(t)|B\rangle + c_{H_i}(t)|H\rangle + c_{S_i}(t)|S\rangle
where |B\rangle, |H\rangle, |S\rangle represent the Buy, Hold, and Sell mental states, and:
|c_{B_i}(t)|^2 + |c_{H_i}(t)|^2 + |c_{S_i}(t)|^2 = 1
```

The entire market at time t is described by the tensor product of all Ninvestors' preference states:

$$|\Psi_{\text{market}}(t)\rangle = \bigotimes_{i=1}^{N} |\Psi_i(t)\rangle$$

However, due to social influence, communication, and psychological herding, investors' states are generally entangled, meaning the global market state is not separable. For simplicity, we will first proceed with independent states and later introduce entanglement.

#### 5.2 Price Formation Operator

Let the observable stock market price operator \hat{P} act on the global preference state, extracting the aggregate price level based on the proportion of Buy and Sell preferences.

```
Define \hat{P} as:

\hat{P} = P_0 + \lambda \sum_{i=1}^{N} (\hat{n}B_i - \hat{n}S_i)
```

- $P_0$  is the fundamental value baseline,
- $\lambda > 0$  is a scaling factor capturing price sensitivity to preferences,
- $\bullet \hat{n}B_i = |B\rangle\langle B|$  and  $\hat{n}S_i = |S\rangle\langle S|$  are projection operators onto Buy and Sell states of investor i.

Thus, the expected stock price at time 
$$t$$
 is:  $\langle \Psi_{\text{market}}(t)|\hat{P}|\Psi_{\text{market}}(t)\rangle = P_0 + \lambda \sum_{i=1}^{N} \left(|c_{B_i}(t)|^2 - |c_{S_i}(t)|^2\right)$ 

Thus, when more investors assign stronger probabilities to Buy states, prices tend to rise, whereas dominance of Sell preferences leads to price declines. The market price is inherently probabilistic, continually shaped by the evolving subjective evaluations and shifting preferences of all participants.

# 5.3 Example: Two-Investor Market

```
Consider a simple two-investor market. Each investor's state is:
```

$$|\Psi_1\rangle = \frac{1}{\sqrt{2}}|B\rangle + \frac{1}{\sqrt{2}}|S\rangle, \quad |\Psi_2\rangle = \sqrt{0.9}|B\rangle + \sqrt{0.1}|S\rangle$$

The global state is:

 $|\Psi_{\mathrm{market}}\rangle = |\Psi_1\rangle \otimes |\Psi_2\rangle$ 

Compute expected price:

 $\langle \Psi_{\text{market}}|\hat{P}|\Psi_{\text{market}}\rangle = P_0 + \lambda\left(\left(\frac{1}{2} - \frac{1}{2}\right) + (0.9 - 0.1)\right) = P_0 + 0.8\lambda$ 

Thus, despite one investor being indifferent (equal Buy/Sell probabilities), the second investor's strong Buy preference drives prices upward.

### 5.4 Time Evolution of Market State

The entire market's preference evolution is governed by the composite Hamiltonian:

$$\hat{H}$$
market =  $\sum (i = 1, ...N) \hat{H}_{pref,i}$ 

where each  $\hat{H}_{\text{pref},i}$  is as defined previously, potentially incorporating interactions between investors (off-diagonal entanglement terms).

The full market state evolves according to:

$$i\hbar \frac{d}{dt} |\Psi_{\text{market}}(t)\rangle = \hat{H} \text{market} |\Psi_{\text{market}}(t)\rangle$$

Given the probabilistic dynamics, the market price is not deterministic but fluctuates over time, reflecting the internal cognitive oscillations of investors' subjective preferences.

# 5.5 Explicit Example: Oscillating Market Price

Assume a single investor whose preference oscillates between Buy and Sell as:

$$|\Psi(t)\rangle = \cos\left(\frac{\Omega t}{2\hbar}\right)|B\rangle + i\sin\left(\frac{\Omega t}{2\hbar}\right)|S\rangle$$

Compute expected price:

$$\langle \Psi(t)|\hat{P}|\Psi(t)\rangle = P_0 + \lambda \left(\cos^2\left(\frac{\Omega t}{2\hbar}\right) - \sin^2\left(\frac{\Omega t}{2\hbar}\right)\right) = P_0 + \lambda \cos\left(\frac{\Omega t}{\hbar}\right)$$

Thus, market prices oscillate over time even without external news, purely due to subjective preference oscillations.

This quantum formalism provides crucial insights into stock market dynamics by showing that stock prices are not static reflections of objective fundamentals, but rather the aggregation of subjective, fluctuating, and probabilistic investor preferences. Prices evolve dynamically due to the internal oscillations of individual evaluations, shaped by psychological factors and constant reassessment. Furthermore, social influence, entanglement among investors, and external shocks amplify or collapse these superpositions of preferences, resulting in inherently volatile and non-linear market behavior. This framework naturally accounts for complex phenomena such as speculative bubbles—where collective preferences align toward mass Buy states—market crashes—where preferences

collapse rapidly toward Sell states—and persistent volatility, all without the need to invoke external, random shocks as primary drivers.

# 6 Quantum Tunneling and Sudden Market Shifts

One of the most striking and often unpredictable phenomena in financial markets is the occurrence of sudden regime shifts: abrupt market crashes, sharp rallies, and bubble bursts. Traditional financial models struggle to explain these discontinuities, typically attributing them to exogenous shocks or random noise. However, in our quantum framework, these events arise naturally through a well-understood quantum phenomenon—tunneling. In this section, we formalize the concept of tunneling in cognitive decision processes and demonstrate how it leads to sharp, unexpected transitions in stock market prices.

# 6.1 Cognitive Potential Barriers

Let us begin by conceptualizing an investor's subjective preference landscape as a potential energy profile over decision space. An investor may have strong cognitive inertia (loss aversion, sunk cost fallacy, confirmation bias), forming psychological barriers that prevent switching from one preference (e.g., Hold) to another (e.g., Sell).

Define a one-dimensional cognitive potential function V(x), where x represents the subjective evaluation continuum:

$$V(x) = \begin{cases} V_0 & \text{for } 0 < x < a \\ 0 & \text{otherwise} \end{cases}$$
 (barrier region)

Here:

- $V_0 > 0$  represents the cognitive resistance to preference change (strength of psychological bias or commitment).
  - a is the width of the cognitive barrier (depth of commitment).

The subjective preference wavefunction  $\protect\operatorname{psi}(x)$  evolves according to the time-independent Schrödinger equation:

$$-\frac{\hbar_{\text{cog}}^2}{2m}\frac{d^2\psi(x)}{dx^2} + V(x)\psi(x) = E\psi(x)$$
 where:

- m represents the cognitive "mass" or rigidity (related to stubbornness, resistance to change).
- $\bullet \hbar_{\rm cog}$  is the cognitive Planck constant, encoding the degree of subjective uncertainty,
- $\bullet$   $E < V_0$  is the cognitive energy level (strength of the internal drive to change preference).

# 6.2 Tunneling Probability Calculation

We seek to compute the tunneling probability, i.e., the likelihood that an investor's subjective preference "penetrates" the barrier, leading to a sudden shift (e.g., sudden selling after long holding).

The wavefunction solutions:

• Region I (before barrier, x < 0):

• Region I (before barrier, 
$$x < 0$$
).  
 $\psi_I(x) = Ae^{ikx} + Be^{-ikx}, \quad k = \frac{\sqrt{2mE}}{\hbar_{\text{cog}}}$   
• Region II (barrier,  $0 < x < a$ ):

• Region II (barrier, 
$$0 < x < a$$
):
$$\psi_{II}(x) = Ce^{\kappa x} + De^{-\kappa x}, \quad \kappa = \frac{\sqrt{2m(V_0 - E)}}{\hbar_{\text{cog}}}$$
• Region III (after barrier,  $x > a$ ):
$$\psi_{III}(x) = Fe^{ikx}$$

$$\psi_{III}(x) = Fe^{ikx}$$

Applying boundary conditions (continuity of  $\psi(x)$  and  $\frac{d\psi}{dx}$  at x=0 and x=a), we derive the transmission coefficient (tunneling probability):

$$T = \left| \frac{F}{A} \right|^2 = \frac{1}{1 + \frac{V_0^2 \sinh^2(\kappa a)}{4E(V_0 - E)}}$$

The limit of thick, strong barriers (\kappa a \gg 1), this simplifies to:

$$T \approx e^{-2\kappa a} = \exp\left(-2a\frac{\sqrt{2m(V_0 - E)}}{\hbar_{\cos}}\right)$$

Interpretation:

- Even if  $E < V_0$ , there is a non-zero probability that the investor "tunnels" through cognitive inertia, suddenly switching preferences.
- This leads to sharp sell-offs (crashes) or sudden buying sprees, depending on context.
  - 6.3 Explicit Example: Market Crash via Tunneling

Let's assign plausible cognitive parameters:

- Cognitive mass: m = 1,
- Barrier height:  $V_0 = 10$  (strong psychological commitment to holding
  - Cognitive energy: E=6 (moderate internal drive to sell),
  - Barrier width: a = 1,
  - Cognitive Planck constant:  $h_{cog} = 1$ .

Compute:

$$\kappa = \frac{\sqrt{2(10-6)}}{1} = 2.828$$

$$T \approx e^{-2 \times 2.828 \times 1} = e^{-5.656} \approx 0.0035$$

Thus, there's a 0.35% probability at each decision moment that the investor suddenly shifts from Hold to Sell, triggering a price impact. Importantly, if external events reduce  $V_0$  (e.g., bad news eroding confidence), tunneling probability increases exponentially, potentially aligning many investors to shift preferences simultaneously.

6.4 Collective Tunneling and Market Regime Shifts

In real markets, social networks, media, and communication channels couple the cognitive potentials of multiple investors. If investors' barriers become synchronized (herding), small shocks may induce collective tunneling—analogous to macroscopic quantum tunneling—resulting in sharp, systemic market crashes

Mathematically, we can model entangled cognitive barriers: 
$$V(x_1,x_2,\ldots,x_N)=\sum_{i=1}^N V_i(x_i)+\sum_{i\neq j}\epsilon_{ij}f(x_i,x_j)$$

where  $\epsilon_{ij}$  captures social coupling strength, and  $f(x_i, x_j)$  encodes herding tendencies.

As coupling increases, tunneling becomes correlated across investors, amplifying the magnitude of market shifts.

Therefore, Qquantum tunneling elegantly models sudden, large-scale preference shifts in financial markets, providing a rigorous mathematical explanation for phenomena such as: market crashes (simultaneous collapse of Hold to Sell preferences), bubble bursts (sudden Sell preference shifts overcoming psychological commitment to holding overvalued assets), and abrupt rallies (mass transition from Sell to Buy states after positive news).

These shifts need not be triggered by large external events; instead, small changes in cognitive potential or social coupling can unleash significant regime shifts, fully consistent with empirical observations.

# 7 Entanglement of Investor Decisions and Market Herding

While individual cognitive processes and subjective preferences exhibit probabilistic, superposed dynamics, as modeled in the previous sections, real-world financial markets are not merely the sum of independent investor evaluations. One of the most pervasive and empirically observable phenomena in stock markets is herding behavior—the tendency of investors to imitate each other, leading to correlated, collective decision-making. Traditional behavioral finance attributes herding to informational cascades, reputational concerns, or simple imitation. However, these explanations fail to capture the full systemic, instantaneous alignment often observed during bubbles, crashes, or market manias. In our quantum framework, herding behavior is elegantly and naturally formalized using the concept of quantum entanglement—non-separable correlations between the cognitive states of investors, leading to collective, synchronized preference shifts.

# 7.1 Defining Investor Entanglement

Let us denote the preference state of investor i as:

$$|\Psi_i\rangle = c_{B_i}|B\rangle + c_{H_i}|H\rangle + c_{S_i}|S\rangle$$

If investors' preferences are independent, the global market state is given by a separable tensor product:

$$|\Psi_{\text{market}}\rangle = |\Psi_1\rangle \otimes |\Psi_2\rangle \otimes \cdots \otimes |\Psi_N\rangle$$

However, empirical herding suggests that preferences become correlated—one investor's decision affects another's. This is modeled by constructing a non-factorizable entangled state, for instance, in the case of two investors:

$$|\Psi_{\rm ent}\rangle = \frac{1}{\sqrt{2}} \left( |B\rangle_1 |B\rangle_2 + |S\rangle_1 |S\rangle_2 \right)$$

This Bell-like entangled state implies that if one investor chooses Buy, the other instantaneously aligns to Buy, and similarly for Sell. The preference states

are not independent; measurement of one collapses the global wavefunction, determining the other's state.

#### 7.2Mathematical Representation of Entangled Market

For the general N-investor case, we can write:

$$|\Psi_{\text{market}}\rangle = \sum_{\{i\}} C_{i_1 i_2 \dots i_N} |i_1\rangle \otimes |i_2\rangle \otimes \dots \otimes |i_N\rangle$$

where  $i_k \in \{B, H, S\}$  and  $C_{i_1 i_2 \dots i_N}$  are complex coefficients encoding correlations. If the global coefficients cannot be factorized into products of individual coefficients, the system is entangled.

Consider an explicit 3-investor entangled GHZ state:

$$|\Psi_{\text{GHZ}}\rangle = \frac{1}{\sqrt{2}} (|B\rangle_1 |B\rangle_2 |B\rangle_3 + |S\rangle_1 |S\rangle_2 |S\rangle_3)$$

In this state, if any investor's preference is measured (e.g., choosing Buy), the entire system collapses deterministically, aligning all investors to Buy. This perfectly captures extreme herding: a single influential investor (large institutional trader, public figure) may trigger mass preference alignment.

#### 7.3Impact on Price Formation

Recall the price operator:  $\hat{P} = P_0 + \lambda \sum_{i=1}^{N} (\hat{n}B_i - \hat{n}S_i)$  For the entangled GHZ state:

$$|\Psi_{\rm GHZ}\rangle=\frac{1}{\sqrt{2}}\left(|B\rangle_1|B\rangle_2|B\rangle_3+|S\rangle_1|S\rangle_2|S\rangle_3\right)$$
 Compute the expected price:

$$\langle \Psi_{\text{GHZ}} | \hat{P} | \Psi_{\text{GHZ}} \rangle = P_0 + \lambda \times 3 \times \left(\frac{1}{2} \times 1 - \frac{1}{2} \times 1\right) = P_0$$

Initially, the expected price shows no net movement—Buy and Sell are balanced. However, upon measurement (real-world decision or news event), the entire system collapses into one extreme:

- All Buy: Price increases by  $3\lambda$ ,
- All Sell: Price decreases by  $3\lambda$ .

This sharp discontinuity in price behavior mirrors market stampedes—where a single informational trigger leads to synchronized buying or selling.

#### Explicit Example: Correlated Investment Funds 7.4

Consider two large investment funds, each representing thousands of investors, operating in an entangled state:

$$|\Psi_{\text{funds}}\rangle = \frac{1}{\sqrt{2}} (|Buy\rangle_A |Buy\rangle_B + |Sell\rangle_A |Sell\rangle_B)$$

Due to mutual observation, shared economic models, or regulatory synchronization, their preferences are entangled. Suppose Fund A's manager, reacting to a minor economic indicator, chooses Sell. Immediately, Fund B's state collapses into Sell. This coordinated behavior results in herding at scale, amplifying price movements well beyond fundamentals.

# 7.5 Modeling Dynamic Entanglement Strength

To model degrees of entanglement, we can parameterize the market state as:  $|\Psi_{\rm market}\rangle = \alpha |\Psi_{\rm sep}\rangle + \beta |\Psi_{\rm ent}\rangle$ 

where  $|\Psi_{\rm sep}\rangle$  is the separable (independent) state,  $|\Psi_{\rm ent}\rangle$  is the fully entangled state, and  $|\alpha|^2 + |\beta|^2 = 1$ .

- High entanglement  $(|\beta|^2 \to 1) \to \text{Extreme}$  herding, high volatility.
- Low entanglement  $(|\alpha|^2 \to 1) \to \text{More independent decision-making, less systemic risk.}$

Empirically, factors like media homogeneity, algorithmic trading, or synchronized policy decisions increase  $|\beta|^2$ , making markets more prone to collective collapses or booms.

Hence, quantum entanglement provides a rigorous, mathematical foundation for herding behavior in stock markets. Rather than treating investor preferences as isolated, we model them as non-separable, interdependent cognitive states, wherein measurement (decision) of one investor's preference instantaneously influences others. This naturally leads to phenomena such as:

- Coordinated buying/selling waves,
- Excessive volatility and bubble formation,
- Systemic risk amplification through entanglement coupling.

This formalism surpasses traditional behavioral finance by explaining not just observed herding patterns but the underlying quantum mechanisms driving preference synchronization.

# 8 Uncertainty Principle and Market Volatility

At the heart of quantum mechanics lies the celebrated Heisenberg Uncertainty Principle, which fundamentally limits the precision with which certain pairs of physical properties—such as position and momentum—can be simultaneously known. This inherent indeterminacy is not due to measurement error but reflects a deep ontological feature of reality. In our framework, where stock market dynamics are rooted in quantum cognitive processes, an analogous uncertainty relation governs the interplay between subjective evaluations (preferences) and market observables (prices, risk assessments). In this section, we derive a market uncertainty principle, providing a rigorous explanation for persistent market volatility and the limits of prediction, even in the absence of external shocks.

### 8.1 Subjective Evaluation and Risk Perception Operators

Let us begin by defining two key operators:

1. Subjective Evaluation Operator,  $\hat{E}$ :

Represents the investor's internal valuation or preference strength for an asset (e.g., tendency to Buy or Sell).

2. Risk Perception Operator,  $\hat{R}$ :

Represents the investor's perception of uncertainty, risk tolerance, or market unpredictability.

These operators, akin to non-commuting observables in quantum mechanics, satisfy the commutation relation:

$$[\hat{E}, \hat{R}] = i\hbar_{\rm econ}$$

where  $\hbar_{\rm econ}$  is a constant capturing the cognitive uncertainty intrinsic to financial decision-making (analogous to Planck's constant).

# 8.2 Market Uncertainty Relation

By standard quantum mechanical derivation, the uncertainty principle follows:  $\Delta E \, \Delta R \geq \frac{\hbar_{\rm econ}}{2}$ 

where:

- $\bullet$   $\Delta E$  is the standard deviation (volatility) of subjective evaluation (preference fluctuation),
  - $\Delta R$  is the uncertainty in risk perception.

This inequality implies that attempting to precisely fix one's subjective evaluation (high confidence in Buy/Sell decisions) inherently increases risk perception uncertainty, and vice versa.

# 8.3 Implications for Market Volatility

Consider a highly confident investor who has low  $\Delta E$ —a strongly fixed belief (e.g., overconfidence bias). According to the uncertainty principle, their risk perception  $\Delta R$  is intrinsically high. Such investors are prone to miscalculating actual market risk, potentially leading to destabilizing behavior (e.g., excessive leverage, speculative bubbles).

Conversely, when markets experience elevated uncertainty (e.g., during crises),  $\Delta R$  becomes large. To satisfy the inequality, subjective evaluations become highly volatile and indeterminate, increasing price fluctuations and contributing to market turbulence.

### 8.4 Explicit Example: Gaussian Preference Wave Packet

Let us illustrate with an explicit mathematical example. Model an investor's subjective evaluation as a Gaussian wave packet in evaluation space:

$$\psi(E) = \frac{1}{(\pi \sigma_E^2)^{1/4}} \exp\left(-\frac{(E - E_0)^2}{2\sigma_E^2}\right)$$
 where:

- E represents subjective preference strength (say, Buy inclination),
- $E_0$  is the mean evaluation,
- $\sigma_E = \Delta E$  is the evaluation volatility.

Applying Fourier transform, the corresponding risk perception distribution

is:

$$\begin{split} \tilde{\psi}(R) &= \frac{1}{(\pi \sigma_R^2)^{1/4}} \exp\left(-\frac{(R-R_0)^2}{2\sigma_R^2}\right) \\ \text{where } \sigma_R &= \Delta R \text{ satisfies:} \\ \sigma_E \sigma_R &= \frac{\hbar_{\text{econ}}}{2} \end{split}$$

Thus, narrow preference distribution (low  $\sigma_E$ ) leads to wide risk perception spread (high  $\sigma_R$ ), making precise predictions about market reactions impossible.

### 8.5 Market-Level Interpretation

Aggregating across all investors, market prices become a function of collective subjective evaluations:

$$P(t) = P_0 + \lambda \sum_{i=1}^{N} E_i(t)$$

Given the uncertainty relation at the individual level, aggregate price volatility is bounded below by the product of collective preference fluctuations and collective risk perception uncertainties. This naturally explains:

- Persistent volatility: Even in the absence of new information, subjective cognitive uncertainty injects randomness into prices.
- Limit to predictability: No model can simultaneously reduce preference volatility and risk perception uncertainty beyond the bound set by  $\hbar_{\text{econ}}$ .
- Amplification during crises: Elevated  $\Delta R$  (uncertainty panic) forces increased  $\Delta E$ , leading to wild price swings.

### 8.6 Parameter Estimation

To give concrete numbers, suppose empirically:

 $h_{\rm econ} \approx 0.1$ 

If during stable times:

$$\Delta E = 0.05 \quad \Rightarrow \quad \Delta R > 1$$

meaning investors are confident, but risk perception spreads moderately. During crises:

$$\Delta R = 5 \quad \Rightarrow \quad \Delta E > 1$$

indicating chaotic preference shifts, manifesting as market turmoil.

By extending the uncertainty principle to financial markets, we provide a rigorous, mathematical foundation for intrinsic volatility and unpredictability. Market prices are fundamentally shaped by subjective cognitive processes governed by quantum uncertainty, making certain forms of risk and price fluctuation irreducible. Behavioral anomalies such as overconfidence, panic selling, or irrational exuberance are no longer ad hoc; they are necessary consequences of the cognitive uncertainty relation.

# 9 Implications for Behavioral and Experimental Economics

This section elaborates on how our quantum neuroscience framework for modeling subjective preferences and market behaviors not only captures key empirical patterns in finance—such as indeterminate preferences, volatility, herding, and regime shifts—but also provides a rigorous foundation for long-standing findings in behavioral and experimental economics. Rather than viewing anomalies in human decision-making as deviations from classical rationality, we show that they arise naturally from the formal quantum mechanics of cognition: specifically, superposition, non-commutativity, entanglement, and measurement basis dependence.

In behavioral economics, classic anomalies include loss aversion, framing effects, status quo bias, anchoring, and overconfidence. These are typically accommodated through non-standard utility functions or heuristic-driven models. For example, Prospect Theory defines a kinked utility function:

$$U(x) = \begin{cases} (x-r)^{\alpha}, & x \ge r \\ -\lambda(r-x)^{\beta}, & x < r \end{cases}$$

with  $\alpha, \beta \in (0,1), \lambda > 1$ , and r as the reference point. While descriptive, such functions lack a generative explanation for why preferences are shaped this way. In contrast, within the quantum framework, these features emerge from first principles. Consider the evaluation of a decision involving potential gains and losses. We define two non-commuting cognitive operators:

 $\hat{E}$ gain,  $\hat{E}$ loss, with  $[\hat{E}$ gain,  $\hat{E}$ loss $] = i\hbar_{\text{bias}}$ .

This non-zero commutator reflects the impossibility of evaluating gains and losses simultaneously without cognitive interference. When a decision-maker processes both gain and loss frames, the resulting state is a quantum superposition, and the probability of selecting an option becomes:

$$P(\text{decision}) = |\psi_{\text{gain}} + \psi_{\text{loss}}|^2 = |\psi_{\text{gain}}|^2 + |\psi_{\text{loss}}|^2 + 2\text{Re}(\psi_{\text{gain}}^* \psi_{\text{loss}})$$

 $P(\text{decision}) = |\psi_{\text{gain}} + \psi_{\text{loss}}|^2 = |\psi_{\text{gain}}|^2 + |\psi_{\text{loss}}|^2 + 2\text{Re}(\psi_{\text{gain}}^* \psi_{\text{loss}}).$  The interference term,  $2\text{Re}(\psi_{\text{gain}}^* \psi_{\text{loss}})$ , introduces the observed framing effects and loss aversion, where the contextual evaluation skews choice probabilities even if objective payoffs are equivalent.

Turning to experimental economics, numerous findings show that human preferences are non-stable, non-additive, and contextually shaped by framing, history, and peer interaction. In our model, each individual's cognitive state is a superposition:

$$|\Psi_i\rangle = \sum_k c_k^{(i)} |k\rangle$$

 $|\Psi_i\rangle = \sum_k c_k^{(i)} |k\rangle$ , where each  $|k\rangle$  represents a discrete decision (e.g., cooperate, defect, bid high, bid low). These amplitudes  $c_k^{(i)}$  evolve dynamically and collapse only upon measurement—that is, when the individual is forced to choose. Thus, preferences are indeterminate until decision, and repeated choices may yield different outcomes due to wavefunction evolution between measurements. This explains why preferences shift even without changes in objective circumstances.

Moreover, social contexts induce quantum entanglement between agents. In a group decision experiment, the composite wavefunction becomes:

$$|\Psi_{\text{group}}\rangle = \sum_{k_1, k_2, \dots, k_N} C_{k_1, k_2, \dots, k_N} |k_1\rangle_1 \otimes |k_2\rangle_2 \otimes \dots \otimes |k_N\rangle_N$$

 $|\Psi_{\text{group}}\rangle = \sum_{k_1,k_2,...,k_N} C_{k_1,k_2,...,k_N} |k_1\rangle_1 \otimes |k_2\rangle_2 \otimes \cdots \otimes |k_N\rangle_N.$  This entangled state evolves collectively, such that the choice of one agent influences the state amplitudes of others, even in the absence of direct communication. Such a mechanism explains herding, peer effects, and convergence to social norms observed in lab experiments and real-world settings.

Consider a concrete example: Vernon Smith's double auction market. Classical theory assumes each trader i has a fixed valuation and strategy. Yet experiments show price bubbles and overvaluation even when fundamental information is common knowledge. In our quantum model, each trader's intention is in a superposition:

$$|\Psi_i\rangle = \alpha_i |H\rangle + \beta_i |L\rangle,$$

where  $|H\rangle$  and  $|L\rangle$  represent high and low bid states. As traders observe and react to others, their states become entangled. The full market state is:

$$|\Psi_{\text{market}}\rangle = \sum_{k_1,\dots,k_N} C_{k_1,\dots,k_N} |k_1\rangle \otimes \dots \otimes |k_N\rangle.$$

 $|\Psi_{\mathrm{market}}\rangle = \sum_{k_1,\dots,k_N} C_{k_1,\dots,k_N} |k_1\rangle \otimes \dots \otimes |k_N\rangle.$  A market clearing operation (measurement) collapses this state, yielding a realized price. Bubbles occur when amplitudes across traders temporarily align toward high-bid superpositions, amplified by entanglement. Collapse restores classical prices only after decoherence—a process rarely captured by traditional models.

Another striking implication is the role of measurement basis in shaping decisions. In quantum mechanics, outcome probabilities depend on the basis in which measurement is made. Similarly, in cognitive economics, framing effects arise from rotating the cognitive measurement basis. Evaluating choices in the  $\{|Gain\rangle, |Loss\rangle\}$  basis yields different outcomes than in the  $\{|Safe\rangle, |Risky\rangle\}$  basis. Mathematically, these are basis transformations in Hilbert space, and even simple linguistic reframing corresponds to a change in cognitive basis vectors.

Lastly, we generalize utility to a quantum observable. Instead of being a scalar function, utility becomes an operator:

$$\hat{U}(t) = \sum_{i,j} u_{ij}(t) |i\rangle\langle j|$$

 $\hat{U}(t) = \sum_{i,j} u_{ij}(t)|i\rangle\langle j|,$  where u\_{ij}(t) encodes the agent's evolving valuation of transitions between states | i\rangle and | j\rangle. The operator \hat{U}(t) evolves under a cognitive Hamiltonian:

$$i\hbar \frac{d}{dt}\hat{U}(t) = [\hat{H}_{\text{cog}}, \hat{U}(t)]$$

 $i\hbar \frac{d}{dt}\hat{U}(t) = [\hat{H}_{\rm cog}, \hat{U}(t)],$  which describes how attention, learning, mood, or social pressure dynamically alter preferences over time. This fully aligns with experimental observations that preferences are not fixed but fluctuate with attention, emotion, and context.

In conclusion, the quantum framework provides a unified mathematical foundation for phenomena traditionally viewed as anomalies in behavioral economics. Loss aversion, framing, unstable preferences, herding, and decision volatility all follow naturally from non-commuting cognitive operators, superposition, entanglement, and basis-dependent measurement. The apparent "irrationality" of human behavior is not a flaw—it is the reflection of a quantum cognitive architecture operating within a probabilistic and context-sensitive Hilbert space. This realization opens new possibilities for modeling, experimentation, and policy grounded in the physics of human thought.

#### Conclusion and Future Directions 10

This paper has developed a unified, mathematically rigorous framework connecting subjective preferences, stock market dynamics, and quantum neuroscience, arguing that stock prices and financial behaviors cannot be fully understood without grounding them in the quantum nature of human cognition, consciousness, and idea formation. We have demonstrated, section by section, that people's subjective preferences, ideas, and judgments—which underlie every financial decision—are not merely the result of classical, deterministic neural processes, but emerge from fundamental quantum processes occurring at the level of neural microstructures, specifically within synaptic transmissions and microtubule dynamics. These cognitive processes are inherently probabilistic, governed by principles such as superposition, tunneling, entanglement, and uncertainty, which manifest directly in the behavior of stock markets.

We began by challenging the classical models of financial economics that treat preferences as fixed, exogenous, and objectively measurable. Through insights from behavioral economics and experimental economics, it became clear that preferences are highly subjective, context-dependent, and unstable. However, these fields stopped short of explaining why such subjectivity and indeterminacy arise. Our core contribution fills this explanatory gap by showing that ideas, judgments, and preferences are fundamentally rooted in quantum neuroscience processes, which by their very nature introduce indeterminacy, context sensitivity, and probabilistic outcomes.

Mathematically, we formalized individual investors' cognitive states as superposed quantum states, governed by a preference Hamiltonian. Price formation was modeled as the observable outcome of aggregating these subjective superpositions, with market prices reflecting the collapse of collective preference wavefunctions. Sudden market shifts, such as crashes or bubble bursts, were rigorously explained through quantum tunneling, where investors overcome cognitive potential barriers due to inherent quantum fluctuations. Herding behavior was modeled via quantum entanglement, capturing the synchronized alignment of preferences across investors, amplifying volatility and systemic risk. Finally, we derived a market uncertainty principle, showing that the product of subjective evaluation volatility and risk perception uncertainty is fundamentally bounded, accounting for persistent market unpredictability.

Furthermore, we demonstrated that experimental anomalies—such as framing effects, preference reversals, and market inefficiencies—naturally arise from the non-commutative structure of cognitive operators and the dynamic, operator-valued nature of subjective utility functions, both deeply rooted in quantum cognitive dynamics.

Future Directions:

Several pathways emerge from this framework:

1. Empirical Calibration:

Experimental economics can be redesigned to explicitly test quantum features—such as superposition effects, entanglement correlations between participants, and tunneling behavior in decision-making—using controlled lab settings.

2. Neuroeconomic Experiments:

Directly linking neural measurements (e.g., EEG, fMRI) to observed financial choices can empirically validate the connection between quantum-level neural dynamics and subjective financial decisions.

3. Quantum-Informed Financial Models:

Practical stock pricing models incorporating dynamic, operator-valued preference terms and entanglement-based herding mechanisms could outperform classical models, particularly in volatile, high-frequency trading environments.

4. Policy and Systemic Risk:

Understanding systemic crashes as emergent quantum tunneling or entanglement collapse events opens new avenues for early warning indicators based on synchronization metrics of investor behavior.

### 5. Extension to Institutional and Macro-Level Dynamics:

Just as we linked micro-level cognitive quantum processes to individual financial behavior, future research can extend this framework to explore how collective ideas, institutions, and macroeconomic shifts arise from aggregated quantum cognitive dynamics.

In sum, people's subjective preferences, ideas, and judgments are not merely behavioral quirks or psychological heuristics—they are fundamentally quantum processes deeply embedded in the structure of human consciousness. Financial markets, as large-scale aggregations of these subjective evaluations, inherit all the complexity, uncertainty, and probabilistic behavior inherent in quantum cognitive systems. Therefore, any model that aspires to capture the full richness of stock market dynamics must necessarily incorporate the principles of quantum neuroscience, bridging the deepest layers of brain function with the most observable layers of economic behavior.