

# The Subsidy Immunity Formula (SIF)

## An Optimal Stopping Framework for Retail Mean Reversion Strategies

### Technical Whitepaper

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#### Abstract

Retail investors frequently suffer from the Disposition Effect, holding losing positions due to a lack of explicit opportunity cost quantification. This paper formalizes the **Subsidy Immunity Formula (SIF)**, derived from the *Principle of Indifference*. By equating the value of immediate sale against the expected value of waiting, and algebraically isolating the volatility time horizon from the *Reflection Principle*, we define two novel metrics: **Intrinsic Duration** ( $D_i$ ) and **Dynamic Convexity** ( $C$ ). The interaction of these terms determines the **Strategic Shielding Time (SST)**, identifying the critical “Log-Moneyness” barrier ( $\delta_{crit}$ ) where a position becomes mathematically insolvent.

## 1 Introduction

The core dilemma in managing a losing position is distinguishing between a temporary drawdown and an inefficient use of capital. We approach this through **Revenue Management**: capital is perishable inventory. Holding a losing stock “subsidizes” the position using the risk-free rate ( $r$ ). We aim to find the Time-Based Efficiency limit.

## 2 Model Definitions

The asset price  $S_t$  follows a Geometric Brownian Motion (GBM). We define:

- $S_t$ : Current Spot Price ( $x$ ).
- $K$ : Target Price ( $T$ ).
- $\delta$ : Log-Moneyness,  $\delta = \ln(K/S_t)$ .
- $\sigma$ : Annualized volatility.
- $\tau$ : Volatility time horizon.

## 3 Mathematical Derivation

The SIF model resolves the conflict between the *Funding Time* (linear) and the *Volatility Time* (quadratic).

### 3.1 Phase 1: The Indifference Principle

In Revenue Management, we establish a “Protection Level” where the investor is indifferent between realizing a guaranteed loss now or waiting for an uncertain recovery. We define two value functions:

**1. Value of Immediate Sale ( $V_{now}$ ):** If we sell the asset at current price  $S_t$  today and invest the recovered capital at the risk-free rate  $r$  for a duration  $t$ , the future value is:

$$V_{now}(t) = S_t \cdot e^{rt} \quad (1)$$

**2. Value of Waiting ( $V_{wait}$ ):** The expected value of holding the position until it hits the target  $K$ . This is the target value adjusted by the probability of success  $P$ :

$$V_{wait} = K \cdot P \quad (2)$$

**3. The Equilibrium Condition:** To find the maximum rational holding time, we equate the certainty of the bank deposit with the expectation of the trade:

$$V_{wait} = V_{now}(t) \implies K \cdot P = S_t \cdot e^{r \cdot t} \quad (3)$$

**4. Solving for Funding Time ( $t_{fund}$ ):** Isolating the time variable  $t$  purely from the funding perspective (assuming  $P \rightarrow 1$  for the intrinsic component), we obtain the linear funding limit:

$$t_{fund} = \frac{\ln(K/S_t)}{r} = \frac{\delta}{r} \quad (4)$$

### 3.2 Phase 2: The Statistical Constraint

We cannot simply assume  $P = 1$ . Under the **Reflection Principle** for Brownian Motion, the probability of hitting barrier  $K$  depends on the observation time  $\tau$ :

$$P = 2 \cdot \Phi(-z) \quad \text{where} \quad z = \frac{\delta}{\sigma\sqrt{\tau}} \quad (5)$$

Substituting this into the equilibrium logic, we equate the probability of success to the inverse of the moneyness ratio ( $S_t/K$ ) required to break even:

$$2 \cdot \Phi\left(-\frac{\delta}{\sigma\sqrt{\tau}}\right) = \frac{S_t}{K} \quad (6)$$

### 3.3 Phase 3: Algebraic Isolation of $\tau$

Equation 6 defines  $\tau$  implicitly. To obtain a closed-form solution, we isolate  $\tau$  (Volatility Latency) in sequential steps:

**Step 1:** Isolate the CDF  $\Phi$ :

$$\Phi\left(-\frac{\delta}{\sigma\sqrt{\tau}}\right) = \frac{S_t}{2K} \quad (7)$$

**Step 2:** Apply the Inverse Normal CDF ( $\Phi^{-1}$ ):

$$-\frac{\delta}{\sigma\sqrt{\tau}} = \Phi^{-1}\left(\frac{S_t}{2K}\right) \quad (8)$$

**Step 3:** Solve for  $\tau_{vol}$ : Squaring both sides removes the negative sign and the root, yielding the volatility time penalty:

$$\tau_{vol} = \left[ \frac{\delta}{\sigma \cdot \Phi^{-1}\left(\frac{S_t}{2K}\right)} \right]^2 \quad (9)$$

We define the denominator term as the **Exigence Coefficient (EC)**, which captures the exponential decay of probability as  $S_t$  drops ( $S_t/K \approx e^{-\delta}$ ):

$$EC(\delta) = \left| \Phi^{-1}\left(\frac{1}{2e^\delta}\right) \right| \quad (10)$$

### 3.4 Phase 4: Synthesis and Definitions

The **Strategic Shielding Time (SST)** is the net strategic advantage:  $SST = t_{fund} - \tau_{vol}$ . Substituting the terms, we formally define the components of the SIF model:

**Definition 1: Intrinsic Duration ( $D_i$ )** Represents the linear sensitivity of the funding horizon to displacement.

$$D_i = \frac{1}{r} \quad (11)$$

**Definition 2: Shielding Convexity ( $C(\delta)$ )** Represents the quadratic penalty imposed by

volatility and market exigence. Unlike bond convexity, this is dynamic:

$$C(\delta) = \frac{1}{[\sigma \cdot EC(\delta)]^2} \quad (12)$$

**The SIF Master Equation:** Combining definitions, we obtain the canonical form:

$$SST = D_i \cdot \delta - C(\delta) \cdot \delta^2 \quad (13)$$

## 4 Critical Liquidation Analysis

The position becomes insolvent when the cost of time outweighs the probability of recovery, i.e.,  $SST \leq 0$ .

### 4.1 Critical Log-Moneyness ( $\delta_{crit}$ )

Solving  $(D_i \cdot \delta) - (C(\delta) \cdot \delta^2) = 0$ :

$$D_i = C(\delta) \cdot \delta \implies \frac{1}{r} = \frac{\delta}{(\sigma \cdot EC)^2} \quad (14)$$

Rearranging for  $\delta$ , we find the structural limit:

$$\delta_{crit} = \frac{(\sigma \cdot EC(\delta_{crit}))^2}{r} \quad (15)$$

*Interpretation:* The maximum sustainable distance is proportional to the variance ( $\sigma^2$ ) scaled by exigence, and inversely proportional to the cost of capital ( $r$ ).

## 5 The Greeks of Shielding

To manage the strategy dynamically, we derive the sensitivities.

### 5.1 Shielding Vega ( $\nu_{sif}$ )

Measures sensitivity to volatility.

$$\nu_{sif} = \frac{\partial SST}{\partial \sigma} = \frac{2\delta^2}{\sigma^3 \cdot EC^2} > 0 \quad (16)$$

Higher volatility reduces the convexity penalty  $C(\delta)$ , extending the shielding time.

### 5.2 Shielding Rho ( $\rho_{sif}$ )

Measures sensitivity to interest rates.

$$\rho_{sif} = \frac{\partial SST}{\partial r} = -\frac{\delta}{r^2} < 0 \quad (17)$$

Higher rates reduce  $D_i$ , accelerating the collapse of the shield.

## 6 Numerical Illustration

We simulate the SST collapse for two distinct risk profiles. The **Blue curve** represents a standard asset ( $r = 5\%, \sigma = 20\%$ ), while the **Red curve** illustrates a high-volatility asset ( $r = 5\%, \sigma = 50\%$ ).

Consistent with our derivative analysis ( $\nu_{sif} > 0$ ), the graph confirms that higher volatility reduces the convexity penalty  $C(\delta)$ , effectively extending the strategic shielding time before insolvency.

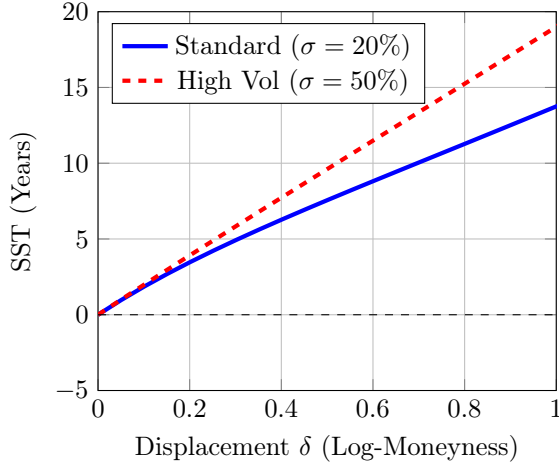


Figure 1: SST Comparison: Volatility buys time.

## 7 Sensitivity Analysis

To validate the model's behavior, we perform a comparative statics analysis, isolating the effects of Intrinsic Duration and Shielding Convexity.

### 7.1 Impact of Rates (Duration Effect)

We vary the risk-free rate  $r$  while holding volatility constant ( $\sigma = 20\%$ ).

- **Low Rates** ( $r = 2\%$ ): The cost of capital is negligible, extending the Intrinsic Duration ( $D_i = 50$ ). The SST remains positive for large displacements.
- **High Rates** ( $r = 10\%$ ): The opportunity cost is punitive ( $D_i = 10$ ). The SST collapses rapidly, forcing early liquidation.

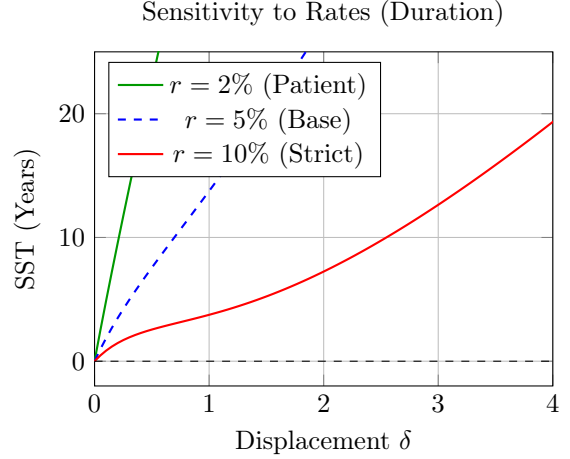


Figure 2: Higher rates accelerate shield collapse.

### 7.2 Impact of Volatility (Convexity Effect)

We vary volatility  $\sigma$  while holding rates constant ( $r = 5\%$ ). This illustrates the "Option Value" of waiting.

- **Low Volatility** ( $\sigma = 10\%$ ): Recovery is statistically unlikely. The Convexity penalty is massive ( $C \uparrow$ ), killing the position immediately.
- **High Volatility** ( $\sigma = 40\%$ ): The distribution tails are fat. The penalty shrinks ( $C \downarrow$ ), granting the investor "survival time".

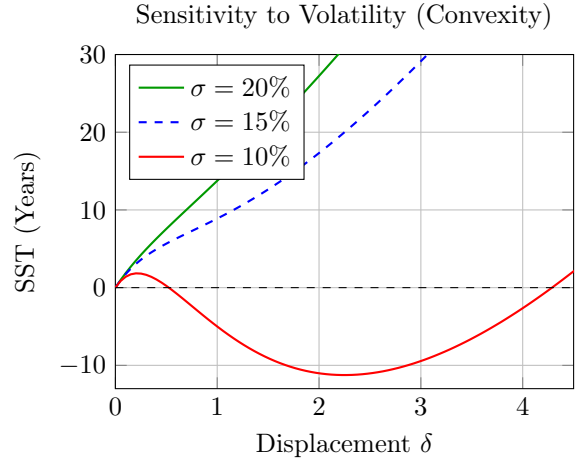


Figure 3: Volatility buys time (reduces Convexity).

## 8 Scenario Analysis: The Greeks in Action

To bridge theory and practice, we analyze the SIF model's behavior under two common market regimes, illustrating the practical implications of  $\rho_{sif}$  and  $\nu_{sif}$ .

## 8.1 The “Hawk” Regime (Rho Risk)

Consider a mean-reversion strategy in a rising rate environment, such as the 2022 tightening cycle. As the central bank raises the risk-free rate  $r$  from 0% to 5%:

- **Impact:** The Intrinsic Duration ( $D_i = 1/r$ ) contracts sharply. Since  $\rho_{sif} < 0$ , the "patience" of the capital evaporates.
- **Result:** A position that was solvent at a displacement of  $\delta = 1$  under a ZIRP regime ( $r \approx 0\%$ ) may instantly become insolvent ( $SST < 0$ ). The SIF model forces immediate liquidation to prevent the high opportunity cost from eroding equity.

## 8.2 The “Panic” Regime (Vega Shield)

Conversely, consider a liquidity crunch where the market crashes and the VIX index spikes (e.g.,  $\sigma$  jumps from 20% to 50%).

- **Impact:** Since  $\nu_{sif} > 0$ , the surge in volatility reduces the Shielding Convexity ( $C$ ). The model recognizes that in high-volatility regimes, large displacements are likely "noise" rather than structural breaks.
- **Result:** The SST expands, effectively advising the trader to hold through the turbulence. This prevents the "puking" of positions at the bottom of a volatility spike, aligning with the Reflection Principle's prediction of probable reversion.

## 9 Conclusion

The SIF framework, developed by Emmanuel Normabuena, transforms optimal stopping into a deterministic calculation. By formally defining  $D_i$  and  $C(\delta)$ , we prove that every mean-reversion trade has a mathematical event horizon  $\delta_{crit}$ , beyond which the position is purely speculative.

## References

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