

# Dynamic Volatility Modeling for Option Hedging

## A Comparative Backtest of HV, GARCH, and GJR-GARCH Models

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# Introduction & Objective

- The standard Black-Scholes (BS) model's biggest weakness is its assumption of **constant volatility**.
- Real-world volatility is dynamic, possibly exhibiting:
  - **Clustering**: High-volatility periods follow high-volatility, low follows low.
  - **Leverage Effect**: Volatility increases more when prices fall.
- **Objective**: To test if dynamic volatility models (GARCH & GJR-GARCH) provide more effective delta-hedging performance than a simple Historical Volatility (HV) model.

# The Models: Three Contenders

## ① 30-Day Historical Vol (HV)

- *Benchmark.* Simple, adaptive, but naive. Ignores clustering and leverage effect.

## ② GARCH(1,1)

- *Symmetric Model.* Captures **volatility clustering**.
- Treats +5% and -5% shocks identically.

## ③ GJR-GARCH(1,1)

- *Asymmetric Model.* Captures clustering *and* the **leverage effect** via the gamma ( $\gamma$ ) parameter.

**Hypothesis:** The "smarter" GJR-GARCH model should be the best performer.

# Methodology: Rolling-Window Backtest

- **Assets:** AAPL, GOOGL, MSFT, NVDA, TSLA
- **Simulation Loop (Run Daily):**
  - "Sell" a new 30-day at-the-money (ATM) call option.
  - Forecast volatility using all 3 models (500-day rolling window for GARCH).
  - Calculate BS Delta using each of the 3 different volatility forecasts.
  - Re-hedge this position daily.
- **Key Metric: Variance of P&L (Hedge Error)**
  - A perfect hedge has a P&L variance of 0.
  - The model with the **lowest variance** is the most stable and effective hedge.

# Results 1: The "Textbook Case" (GOOGL & MSFT)

**Finding:** For GOOGL and MSFT, the hypothesis was **correct**. The asymmetric GJR-GARCH model was the clear champion.

<b>Ticker</b> <i>vs. HV</i>	<b>GARCH(1,1)</b> <i>Var. Reduction</i>	<b>GJR-GARCH</b> <i>Var. Reduction</i>
GOOGL	18.30%	<b>20.05%</b>
MSFT	14.53%	<b>18.49%</b>

**Table:** Variance Reduction vs. 30-Day Historical Vol. GJR-GARCH successfully reduced risk.

## Results 2: The "Whipsaw Case" (AAPL & NVDA)

**Finding:** For AAPL and NVDA, the "smarter" GJR-GARCH model **failed**, performing worse than the simpler GARCH(1,1).

<b>Ticker</b> <i>vs. HV</i>	<b>GARCH(1,1)</b> <i>Var. Reduction</i>	<b>GJR-GARCH</b> <i>Var. Reduction</i>
AAPL	-4.06%	<b>-12.65%</b>
NVDA	<b>12.48%</b>	7.21%

**Table:** Variance Reduction vs. 30-Day Historical Vol. GJR-GARCH *increased* risk for AAPL.

## Results 3: The "Overfitting Case" (TSLA)

**Finding:** For TSLA, the symmetric GARCH(1,1) was the clear winner. The GJR-GARCH model failed to provide a better hedge.

<b>Ticker</b> <i>vs. HV</i>	<b>GARCH(1,1)</b> <i>Var. Reduction</i>	<b>GJR-GARCH</b> <i>Var. Reduction</i>
TSLA	<b>11.73%</b>	10.43%

**Table:** Variance Reduction vs. 30-Day Historical Vol. The simpler model won.

# Analysis: The GJR Gamma ( $\gamma$ ) Parameter

The  $\gamma$  (**gamma**) parameter measures the leverage effect. A high, significant  $\gamma$  explains the results.

Ticker	Gamma ( $\gamma$ )	P-value	Finding
<i>Group 1</i>			
GOOGL	0.0477	0.000	<b>Moderate Effect</b>
MSFT	0.0989	0.008	<b>Moderate Effect</b>
<i>Group 2</i>			
AAPL	<b>0.1447</b>	0.000	<b>Extreme Effect</b>
NVDA	<b>0.1417</b>	0.022	<b>Extreme Effect</b>
<i>Group 3</i>			
TSLA	-0.0046	0.706	<b>Not Significant</b>

Table: GJR-GARCH(1,1) Parameter Fits (2017-2024 data)



# Grand Conclusion: "No One Size Fits All"

- **Case 1 (GOOGL, MSFT):** GJR-GARCH works perfectly. Their *moderate* and *significant* leverage effect is modeled well.
- **Case 2 (AAPL, NVDA):** GJR-GARCH fails. Their *extreme* leverage effect ( $\gamma$  is 1.5x-3x larger) creates a "whipsaw" when combined with a naive BS-delta. The model is **too reactive**.
- **Case 3 (TSLA):** GJR-GARCH fails. Its *gamma is not statistically significant* ( $p=0.7$ ). The model is **over-parameterized** for an asset with no leverage effect.

## Final Takeaway

A successful hedging strategy requires a model that is **correctly specified** for the unique volatility structure of the asset.

# Thankyou!