

# Basket Implied Correlation Analysis

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## Objective

This project computes the **implied correlation** of an equity basket using options-market data from the Bloomberg Terminal. It extracts the market's forward-looking estimate of average pairwise correlation between large-cap equities for use in **dispersion trading** signal generation and **correlation risk premium** analysis.

## Overview

Two Jupyter notebooks form the analytical core:

- `correl_matrix.ipynb`: Computes a 30-day rolling realized correlation matrix for the selected equity basket.
- `implied_correl.ipynb`: Calculates the basket-implied correlation time series using index and stock implied volatilities.

All data—including historical prices, implied volatilities, and basket weights—were extracted from the **Bloomberg Terminal**.

## Methodology

### 1. Portfolio Variance Decomposition

The total variance of a basket can be expressed as:

$$\sigma_{\text{basket}}^2 = \sum_i w_i^2 \sigma_i^2 + \rho \left[ \left( \sum_i w_i \sigma_i \right)^2 - \sum_i w_i^2 \sigma_i^2 \right]$$

where:

- $\sigma_{\text{basket}}$ : index or basket volatility (proxied by VIX)
- $\sigma_i$ : implied volatility of stock  $i$
- $w_i$ : weight of stock  $i$  in the basket
- $\rho$ : average pairwise correlation among constituents

## 2. Implied Correlation Extraction

Rearranging yields:

$$\rho_{\text{implied}} = \frac{\sigma_{\text{basket}}^2 - \sum_i w_i^2 \sigma_i^2}{\left( \sum_i w_i \sigma_i \right)^2 - \sum_i w_i^2 \sigma_i^2}$$

This isolates the implied correlation term from the variance equation, yielding the market's expectation of average pairwise correlation.

## 3. Realized Correlation

The realized correlation across  $N$  assets over a rolling window is given by:

$$\rho_{\text{realized}} = \frac{2}{N(N-1)} \sum_{i < j} \text{Corr}(r_i, r_j)$$

where  $r_i$  are the daily returns of each stock.

## Data

- **Source:** Bloomberg Terminal
- **Fields:** 30-day implied volatilities for S&P 500 constituents and index (VIX)
- **Frequency:** Daily

## Basket Composition

Top 10 S&P 500 constituents (approximately 38% index weight):

Ticker	Weight (%)
NVDA	7.38
AAPL	6.63
MSFT	6.13
AMZN	4.27
GOOGL	2.88
GOOG	2.70
META	2.55
TSLA	2.37
BRK.B	1.79
LLY	1.36

*Note: Static weights used. Expanding to top 50 names would improve representativeness (80% of index coverage).*

# Outputs

- Implied correlation time series
- Realized correlation series (30-day rolling)
- Comparative plots of implied vs realized correlation
- Rolling correlation matrices

# Applications

Area	Use Case
Dispersion Trading	Detect mispricing between index and basket options
Correlation Risk Premium	Measure implied-realized correlation spreads
Volatility Forecasting	Identify correlation regime shifts
Risk Management	Monitor systemic correlation clustering

# Interpretation

Correlation Range	Market Regime
0.0–0.3	Low correlation (diversified)
0.3–0.5	Normal
0.5–0.7	Elevated
0.7–1.0	Crisis co-movement

## Signal Interpretation:

- $\rho_{\text{implied}} > \rho_{\text{realized}}$ : index options rich → sell correlation / long dispersion
- $\rho_{\text{implied}} < \rho_{\text{realized}}$ : index options cheap → buy correlation / short dispersion

# Requirements

```
pandas >= 1.5.0
numpy >= 1.24.0
openpyxl >= 3.0.0
matplotlib >= 3.7.0
```

# Execution

1. Run `correl_matrix.ipynb` to compute realized correlations.

2. Run `implied_correl.ipynb` to calculate implied correlation.
3. Compare results for signal generation and visualization.

## Limitations

- Static weights (no rebalancing)
- Limited universe (10 stocks, 38% index coverage)
- Requires Bloomberg data access
- No backtesting framework

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## Citation

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