



Conjecture 2

The Central Pricing Problem

Tomorrow Capital Research

December 2025

Problem Description

We extend the Central Price Problem from a single asset to a system of interacting assets, with explicit microstructural and mean-field effects.

Setup

Fix an integer $N > 3$ denoting the number of traded assets. For each $i \in \{1, \dots, N\}$, let $\{P_t^{(i)}\}_{t \geq 0}$ denote the (mid-)price process of asset i , observed on a discrete time grid

$$0 = t_0 < t_1 < \dots < t_T < t_{T+1} < \dots,$$

with prices known for all $t_k \leq T$ and unknown for $t_k > T$. We write

$$\mathbf{P}_t := (P_t^{(1)}, \dots, P_t^{(N)})^\top$$

for the vector of prices at time t . For concreteness, one may think of:

- an underlying asset A , its option A^* and swap \tilde{A} ,
- another underlying B with derivatives B^* and \tilde{B} ,
- and, more generally, underlyings A, B, C, D, \dots with associated derivatives $A^*, \tilde{A}, B^*, \tilde{B}, \dots$

The assets may be related through payoff structure (underlying–derivative), sector or factor exposure, common fundamentals, or purely through flow and correlation.

Assumption 1 (Information set and observables). *At time T the modeller observes an information set \mathcal{I}_T that includes at least:*

1. *The joint price and volume history*

$$\{(\mathbf{P}_{t_k}, \mathbf{V}_{t_k}) : k = 0, \dots, T\},$$

where \mathbf{V}_{t_k} collects traded volumes or order-flow statistics for all N assets at time t_k .

2. *Cross-sectional and cross-asset features, e.g. empirical correlation matrices of returns, realised volatilities, liquidity measures, and regime indicators extracted from the joint history.*
3. *Fundamental data, such as financial statements, sector/industry classifications and macroeconomic indicators, at least for the issuers associated with a subset of the assets.*

4. Aggregate information about other market participants, encoded as a (possibly approximate) distribution over their states,

$$m_T \in \mathcal{P}(\mathcal{X}),$$

where $\mathcal{P}(\mathcal{X})$ denotes the set of probability measures on a suitable state space \mathcal{X} (e.g. inventory, risk limits, trading style, latency).

Assumption 2 (Immediate vs historical dependence). For each asset i and time T , the one-step-ahead price admits a decomposition of the form

$$P_{T+1}^{(i)} = \alpha_T^{(i)} I_T^{(i)} + (1 - \alpha_T^{(i)}) H_T^{(i)} + \varepsilon_{T+1}^{(i)}, \quad (1)$$

where:

- $I_T^{(i)}$ is an immediate impact term, measurable with respect to the short-horizon order-book state and flow around T ;
- $H_T^{(i)}$ is a historical dynamics term, measurable with respect to the longer-horizon price/volume/fundamental history up to time T ;
- $\alpha_T^{(i)} \in [0, 1]$ is a (possibly random) weight capturing the relative importance of immediate versus historical effects, allowed to vary in time and across assets;
- $\varepsilon_{T+1}^{(i)}$ is a noise term with mean zero (under an appropriate physical or risk-neutral measure), capturing residual unpredictable variation.

We assume that $\varepsilon_{T+1}^{(i)}$ is small in a sense to be made precise (e.g. sub-Gaussian, or with bounded conditional variance).

Assumption 3 (Structure of historical component). For each i , the historical component $H_T^{(i)}$ is a measurable functional of \mathcal{I}_T , and may depend on:

- past prices $\{P_{t_k}^{(j)} : 0 \leq k \leq T, 1 \leq j \leq N\}$,
- cross-asset and cross-derivative correlations,
- volatility and liquidity regimes (e.g. estimated from realised volatility, bid-ask spreads, depth, etc.),
- macro/fundamental signals extracted from financial statements and economic data,
- medium- to long-horizon features such as drift, mean reversion, or factor loadings.

In particular, $H_T^{(i)}$ may encode stylised facts such as:

- partial mean reversion after extreme returns or all-time highs/lows;
- reversion towards fundamentals (e.g. valuation ratios);
- persistent drift components over longer horizons.

Assumption 4 (Structure of immediate impact). The immediate impact term $I_T^{(i)}$ is a functional of the short-horizon order-book state and flow around T , including:

- incoming market and limit orders (sizes, sides, time priority),
- the current limit-order-book shape (depth at each price level, imbalance),
- observed behaviour of other agents (e.g. clustering of aggressive trades, herding).

When the modeller submits an order of price $p^{(i)}$ and size (or normalised volume) $q^{(i)}$ at time T , the mid-price of asset i is shifted according to an impact rule

$$mid_{T+}^{(i)} = mid_T^{(i)} + (p^{(i)} - mid_T^{(i)}) \phi^{(i)}(q^{(i)}, \mathcal{I}_T), \quad (2)$$

for some impact function $\phi^{(i)}$, which in a toy specification may take the form

$$\phi^{(i)}(q^{(i)}, \mathcal{I}_T) = \sqrt{q^{(i)}}.$$

More refined impact functions (e.g. concave in volume, state-dependent) are allowed and encouraged.

Assumption 5 (Mean-field interaction). *Other participants are modelled collectively via the distribution m_T over their states. Their joint behaviour feeds back into prices through both $H_T^{(i)}$ and $I_T^{(i)}$. In particular, for a given policy of the modeller, the evolution of \mathbf{P}_t is coupled with the evolution of m_t , leading to a mean-field type interaction in the sense of mean-field games or mean-field control.*

Central Price and Confidence Regions

Given the information set \mathcal{I}_T and the mean-field state m_T , define the *central price map*

$$F_T : (\mathcal{I}_T, m_T) \mapsto \widehat{\mathbf{P}}_{T+1} := (\widehat{P}_{T+1}^{(1)}, \dots, \widehat{P}_{T+1}^{(N)})^\top,$$

where $\widehat{P}_{T+1}^{(i)}$ is a model-based estimate of $P_{T+1}^{(i)}$ (e.g. the conditional expectation or a robust location functional).

For a fixed horizon $H \geq 5$, we also define a *confidence-region map*

$$\mathcal{C}_T : (\mathcal{I}_T, m_T) \mapsto \mathcal{R}_T \subset \mathbb{R}^{N \times H},$$

where \mathcal{R}_T is interpreted as a confidence set for the joint future price paths

$$\{\mathbf{P}_{T+1}, \mathbf{P}_{T+2}, \dots, \mathbf{P}_{T+H}\}.$$

Conjecture 1 (Central Price Problem, Multi-Asset Extension). *Under Assumptions 1–5 and under realistic market conditions, there exist:*

1. *A non-trivial, empirically calibratable family of central price maps*

$$F_T : (\mathcal{I}_T, m_T) \mapsto \widehat{\mathbf{P}}_{T+1},$$

such that for each asset i ,

$$P_{T+1}^{(i)} - \widehat{P}_{T+1}^{(i)}$$

is statistically small (e.g. sub-Gaussian with variance bounded uniformly in T) and the resulting central price estimates are practically useful for trading, market-making or risk management.

2. *A family of confidence-region maps*

$$\mathcal{C}_T : (\mathcal{I}_T, m_T) \mapsto \mathcal{R}_T \subset \mathbb{R}^{N \times H}, \quad H \geq 5,$$

such that the true joint future prices satisfy

$$(\mathbf{P}_{T+1}, \dots, \mathbf{P}_{T+H}) \in \mathcal{R}_T$$

with probability at least $1 - \delta$ for some prescribed level $0 < \delta < 1$, and such that this coverage property holds with empirical accuracy when tested on out-of-sample data from realistic markets.

Equivalently, there should exist an empirically testable mapping from:

- *historical dynamics (prices, volumes, correlations, volatility and liquidity regimes, fundamentals), and*
- *immediate impact (order-book state, flows, participant positions, the modeller's own trades),*

to:

- *short-term central prices $\widehat{\mathbf{P}}_{T+1}$, and*
- *medium-horizon confidence regions \mathcal{R}_T for a system of $N \geq 4$ interacting assets,*

with non-trivial predictive power beyond naive baselines (e.g. random walk or constant-volatility models).

Remark 1 (Unified pricing framework). Special interest is reserved for solutions to Conjecture 1 that embed the above construction into a unified theoretical framework covering simultaneously:

- *the market-maker's very short-horizon pricing and inventory-control problem,*

- the high-frequency trader's short- to medium-horizon alpha and execution problem, and
- the medium- to long-term asset-pricing and portfolio-allocation problem,

under a common set of assumptions linking:

- microstructure (order-book mechanics, impact, immediacy),
- participant behaviour and mean-field interactions, and
- longer-term price formation, drift, and risk premia.

Such a framework should make explicit how the decomposition (1) and the impact rule (2) (or their generalisations) propagate across time scales, from tick-by-tick dynamics to daily, weekly, or longer horizons.