

Instrument Similarity Analysis

Stage 1 – Proposal (CMP223)

Enrico Dal Pizzol
UFRGS / INF / PPGC

September 3, 2025

Object: Methods for measuring similarity and correlation between financial time series.

Goal: Evaluate computational performance of different algorithms (execution time, CPU, memory).

Methods: Pearson, Spearman, Kendall, DTW (and optionally MI).

- Correlation influences risk, diversification, and trading decisions.
- Different methods \Rightarrow different assumptions & costs
- Pearson: linear, very fast
- Spearman/Kendall: monotonic, robust to outliers
- DTW: handles misalignment, but much heavier

Computational Object — Characteristics

- Depending on the chosen period and frequency, time series can become very large.
- Each algorithm has its own computational cost and statistical properties.
- The task is the comparison of two financial time series.
- Each correlation/similarity method has specific strengths:
 - Good for capturing linear trends (Pearson)
 - Robust to outliers and non-linearity (Spearman/Kendall)
 - Suitable for comparing series that are distorted or shifted in time (DTW)

Analysis Method (Measurement)

- Apply correlation methods on real or replayed financial time series
- Environment & tools: Python (NumPy, Pandas, SciPy) and Docker
- Metrics collected: correlation output, CPU usage (avg/peak), memory, and execution time

- Period: 1d, 7d, 1m, 6m, 1y
- Frequency: 1s, 1min, 1h, 1d
- Method: Pearson, Spearman, Kendall, DTW
- Instruments: X, Y, Z (pairs)
- Purpose: compare trade-offs: accuracy \times computational cost

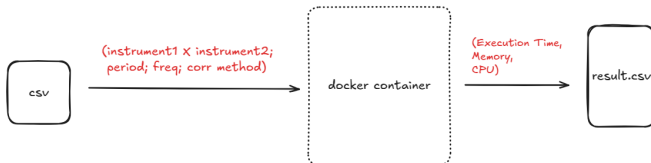
Metrics (What will be measured)

- Execution time
- CPU usage (%)
- Avg/Peak memory
- **Correlation variance**

Experiment Orchestration

- Input: CSV with all parameter combinations
- Runner: one container per row
- Inside container: calculate correlation
- Collector: monitor CPU & memory
- Aggregator: merge into `results.csv`

Experiment Pipeline




Example Result (Mock)

Instruments	Period	Freq	Method	Time	Mem	CPU
X×Y	1d	1min	Pearson	0.01s	5MB	2%
X×Y	1d	1min	Spearman	0.03s	6MB	3%
X×Y	7d	Daily	Kendall	0.02s	5MB	2%
X×Z	6m	1s	DTW	2h	700MB	45%

Next-Steps Schedule (8 Weeks)

- W1: Collect dataset and preprocess
- W2-3: Build test suite (Docker, scripts, ...)
- W4-6: Run experiments; record metrics
- W7: Analyze results; generate plots/tables
- W8: Write and submit final report

Thank you!

 `instrument-similarity-analysis`

References



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