



KaxaNuk
Sharing knowledge

Data Curation & Feature Engineering

Session 2



Content

- The Data Challenge in Investment Research
- Introduction to Data Curator
- Outlier Detection & Correction
- Feature Engineering
- Hands-On: Your First Data Pipeline

The Data Challenge in Investment Research

Bottlenecks



The Data Challenge

60–80%

of a quant researcher's time may be wasted on repetitive data wrangling tasks.

Reproducibility Crisis

- Undocumented workflows prevent reliable validation of results.

Scattered Sources

- Inconsistent provider formats require heavy normalization.

Manual Errors

- Human intervention introduces silent data failures.

Lost Velocity

- Time spent fixing pipelines reduces research productivity.



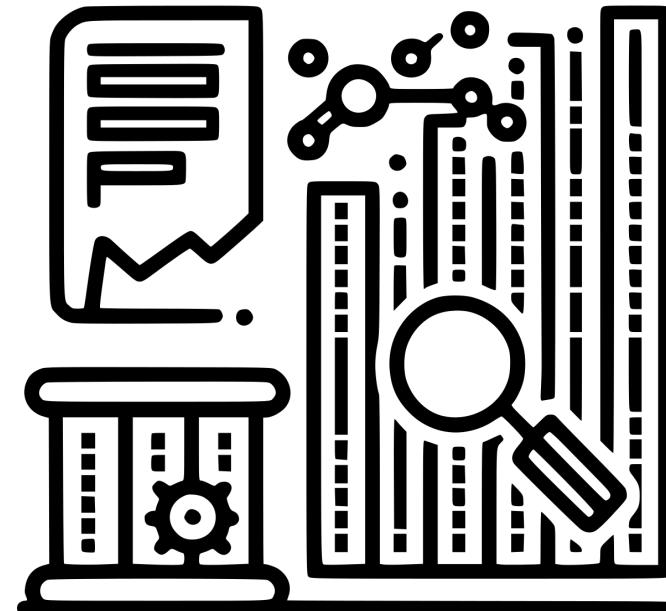
The Messy Middle Problem

Major bottlenecks:

- Data quality
- Pipeline consistency
- Governance gaps

In quant investing, this looks like:

- 40 Jupyter notebooks
- 15 slightly different datasets
- No reproducibility
- No audit trail



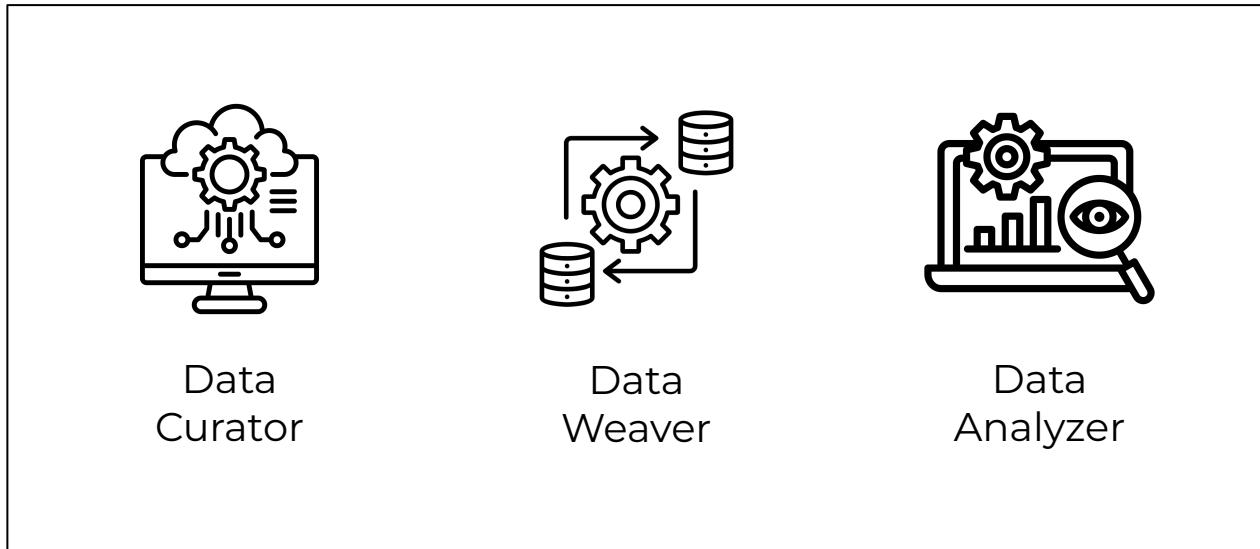
Professional Research Demands



Structured, versioned,
reproducible data systems.



The Tooling To Speed Up Feature Engineering



Introduction to Data Curator

The Solution to Improve Data Wrangling Tasks

Data Curator



A Python-based tool designed to **build structured databases** for market, fundamental, and alternative data, **transforming chaotic wrangling into a systematic workflow.**

Unified Data Access

Single interface for various data providers and custom in-house datasets, with homogenized tags.

Vendor Data Verification

Assess whether the provider's data is reliable or affected by serious issues (e.g., negative values, restated data).

Prebuilt Calculations

Ready-to-use functions for outlier adjustments, ratios, momentum, trend, volatility, volume, etc.

Custom Features

Easily define your own custom feature functions to extend the capabilities of the Data Curator.

Excel-Based Configuration: Specify tickers, dates, and features in a spreadsheet, **no complex coding required for setup.**



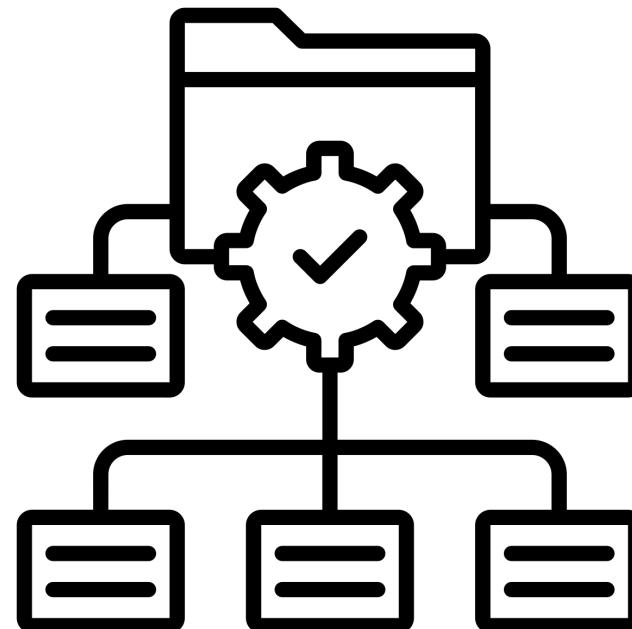
Structured Financial Data

Data is messy:

- Missing earnings dates
- Inconsistent ticker mapping
- Outliers
- Corporate actions
- Different frequencies

Structured data means:

- Aligned timestamps
- Point-in-time data
- Cleaned values
- Standardized schema
- Version-controlled datasets





Time Matters: Avoiding Look-Ahead Bias

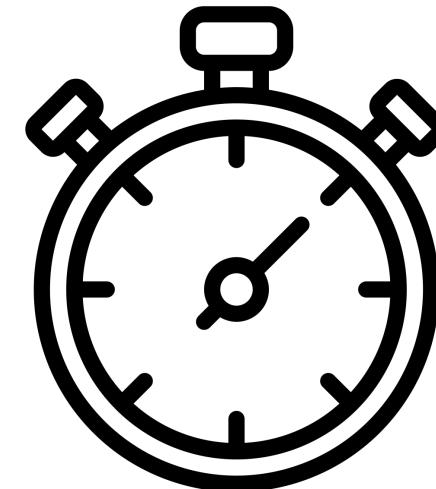
Every data point must respect:

- Release date**
- Availability date**
- Revision history**

Example:

- Q1 earnings released on May 5
- You cannot use them in March backtests.

Feature Engineering must be:



Chronologically honest.



Getting Structured Financial Data

The 3-Step Workflow

01 Install

Set up the environment. Requires Python 3.12+.

```
pip install kaxanuk.data_curator  
# Optional extension  
  
pip install  
kaxanuk.data_curator_extensions.  
yahoo_finance
```

02 Configure

Initialize and edit settings in Excel.

kaxanuk.data_curator init excel

Config/parameters_datacurator.xlsx
Define Tickers, Dates, Providers

Config/.env Store API Keys securely

03 Execute

Run the pipeline to fetch and process data.

kaxanuk.data_curator run

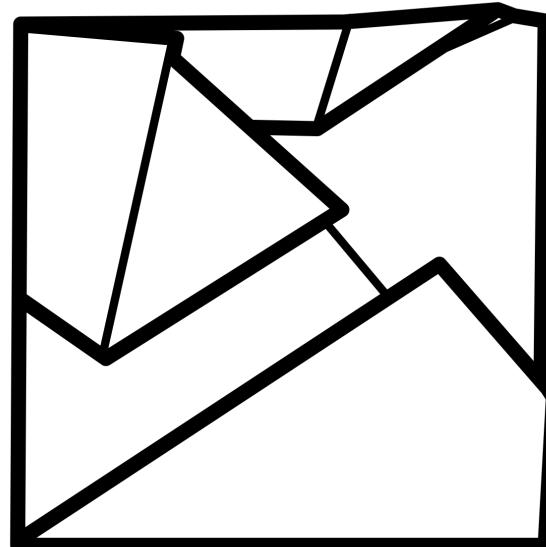
Output/ Structured CSV/Parquet files ready for analysis



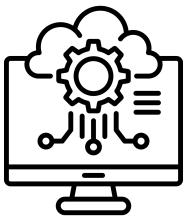
Before

Fragmented infrastructure:

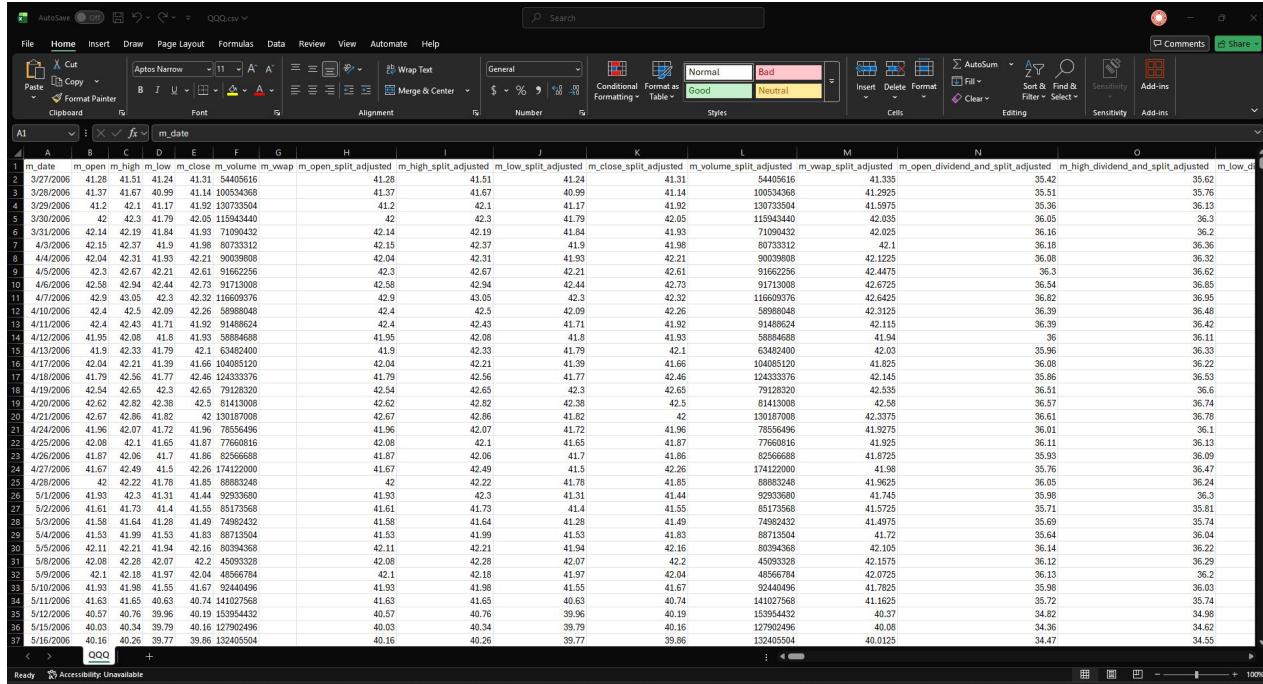
- Multiple APIs with heterogeneous formats
- Inconsistent field naming across providers
- Missing or partially available data
- Manual joins and ad-hoc transformations
- Point-in-time misalignment
- Non-reproducible research notebooks
- Difficult feature sharing across teams
- High debugging overhead
- ...



Unified Time Series Output Across All Tools



Data
Curator



	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	
1	m_date	m_open	m_high	m_low	m_close	m_volume	m_vwap	m_open_split_adjusted	m_high_split_adjusted	m_low_split_adjusted	m_close_split_adjusted	m_volume_split_adjusted	m_vwap_split_adjusted	m_open dividend	m_high dividend	m_low dividend
2	3/27/2006	41.28	41.51	41.24	41.31	54405616		41.28	41.51	41.24	41.31	54405616		35.42	35.62	
3	3/28/2006	41.37	41.67	40.99	41.14	100534368		41.37	41.67	40.99	41.14	100534368		35.51	35.76	
4	3/29/2006	41.2	42.1	41.17	41.92	130733504		41.2	42.1	41.17	41.92	130733504		35.36	36.13	
5	3/30/2006	42	42.3	41.79	42.05	115943440		42	42.3	41.79	42.05	115943440		36.05	36.3	
6	3/31/2006	42.14	42.19	41.81	41.93	71090432		42.14	42.19	41.84	41.93	71090432		36.16	36.2	
7	4/1/2006	42.04	42.37	41.9	41.98	80033312		42.15	42.37	41.9	41.98	80733312		36.18	36.55	
8	4/2/2006	42.01	42.31	41.81	42.05	80006868		42.04	42.31	41.93	42.21	90006868		36.00	36.32	
9	4/3/2006	42.3	42.67	42.26	42.61	91652256		42.3	42.67	42.27	42.21	91652256		36.3	36.62	
10	4/8/2006	42.58	42.94	42.44	42.73	91713008		42.58	42.94	42.44	42.73	91713008		36.54	36.85	
11	4/10/2006	42.9	42.05	42.3	42.22	116693976		42.9	42.05	42.3	42.32	116693976		36.82	36.95	
12	4/10/2006	42.4	42.8	42.09	42.26	8898848		42.4	42.5	42.09	42.26	8898848		36.89	36.98	
13	4/11/2006	42.4	42.49	41.71	41.92	91498624		42.4	42.43	41.71	41.92	91498624		36.39	36.42	
14	4/12/2006	41.95	42.08	41.8	41.93	58846488		41.95	42.08	41.8	41.93	58846488		36	36.11	
15	4/13/2006	41.9	42.33	41.79	42.1	63402400		41.9	42.33	41.79	42.1	63402400		35.99	36.33	
16	4/17/2006	42.04	42.21	41.39	41.66	104095120		42.04	42.21	41.39	41.66	104095120		36.08	36.22	
17	4/18/2006	41.79	42.56	41.77	42.46	124333376		41.79	42.56	41.77	42.46	124333376		35.86	35.53	
18	4/19/2006	42.54	42.65	42.3	42.65	79128320		42.54	42.65	42.3	42.65	79128320		36.51	36.6	
19	4/20/2006	42.62	42.82	42.38	42.5	81413008		42.62	42.82	42.38	42.5	81413008		36.57	36.74	
20	4/21/2006	42.67	42.86	41.82	42	130187008		42.67	42.86	41.82	42	130187008		36.61	36.78	
21	4/24/2006	41.96	42.07	41.72	41.96	78556496		41.96	42.07	41.72	41.96	78556496		36.01	36.1	
22	4/25/2006	42.08	42.1	41.65	41.87	77660816		42.08	42.1	41.65	41.87	77660816		36.11	36.13	
23	4/26/2006	41.87	42.06	41.7	41.86	82566688		41.87	42.06	41.7	41.86	82566688		35.93	36.09	
24	4/27/2006	41.67	42.49	41.5	42.26	174122000		41.67	42.49	41.5	42.26	174122000		35.76	35.47	
25	4/28/2006	42	42.22	41.78	41.85	88883248		42	42.22	41.78	41.85	88883248		36.05	36.24	
26	5/1/2006	41.93	42.3	41.31	41.44	92933680		41.93	42.3	41.31	41.44	92933680		35.98	36.3	
27	5/2/2006	41.61	41.73	41.4	41.55	85173568		41.61	41.73	41.4	41.55	85173568		35.71	35.81	
28	5/3/2006	41.58	41.64	41.28	41.49	74982432		41.58	41.64	41.28	41.49	74982432		35.69	35.74	
29	5/4/2006	41.62	41.71	41.33	41.55	80330564		41.53	41.62	41.33	41.55	80330564		35.64	35.64	
30	5/5/2006	42.11	42.21	41.43	42.15	80330568		42.11	42.21	41.54	42.15	80330568		36.14	36.22	
31	5/8/2006	42.08	42.26	42.07	42.2	45903328		42.08	42.26	42.07	42.2	45903328		36.12	36.29	
32	5/9/2006	42.1	42.18	41.57	42.04	48566784		42.1	42.18	41.97	42.04	48566784		36.13	36.2	
33	5/10/2006	41.93	41.98	41.55	41.67	23440496		41.93	41.98	41.55	41.67	92446496		35.98	36.03	
34	5/11/2006	41.63	41.65	40.63	40.74	141027568		41.63	41.65	40.63	40.74	141027568		35.72	35.74	
35	5/12/2006	40.57	40.76	39.93	40.19	153934432		40.57	40.76	39.96	40.19	153934432		34.82	34.98	
36	5/15/2006	40.03	40.34	39.79	40.16	127902496		40.03	40.34	39.79	40.16	127902496		34.36	34.62	
37	5/16/2006	40.16	40.26	39.77	39.86	132405904		40.16	40.26	39.77	39.86	132405904		34.47	34.55	

Outlier Detection & Correction

Ensuring Data Integrity



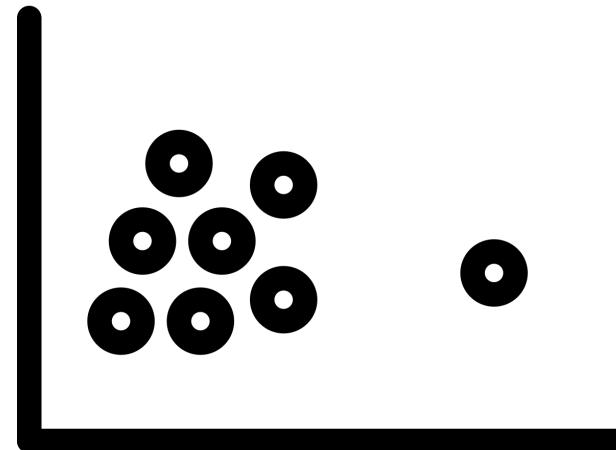
Why Outliers Destroy Signals

Outliers may come from:

- Data entry errors
- API glitches
- Stock splits not adjusted
- Extreme corporate events

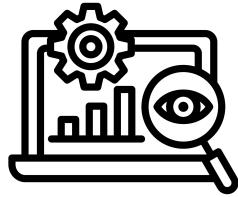
Consequences:

- Inflated z-scores
- False momentum signals
- Volatility distortion

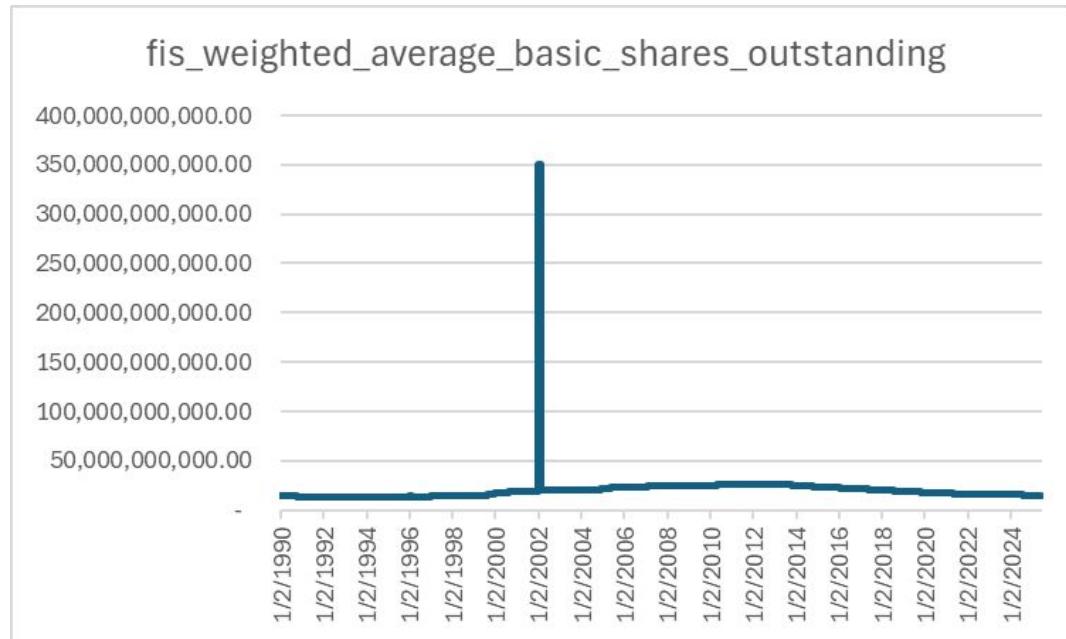




Data Entry Errors



Data
Analyzer

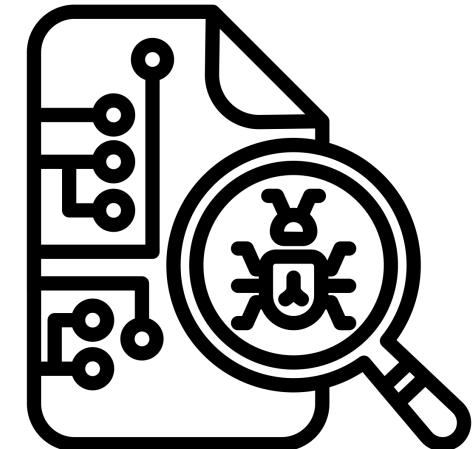




Methods for Outlier Detection

Some ideas we can test:

- Rolling Z-Score Thresholds
- Median Absolute Deviation (MAD)
- Volatility-Scaled Return Filters
- Abnormal price gaps vs rolling volatility
- Structural break detection
- Stale value detection
- Near-zero denominator flags
- Corporate action validation (splits, dividends)
- Ratio sanity checks (negative or impossible values)
- Extreme growth without event justification
- Accounting restatement detection





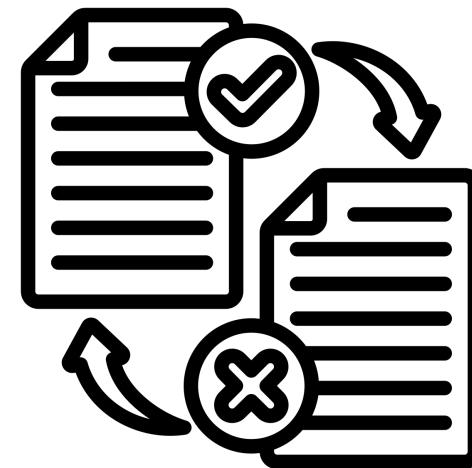
Methods for Outlier Correction

We can implement:

- Z-score filtering
- Rolling median deviation
- Winsorization
- Robust scaling
- Percentile clipping

Important:

Outlier correction must be rule-based, reproducible, and grounded in financial logic (e.g., corporate actions and fundamental events).





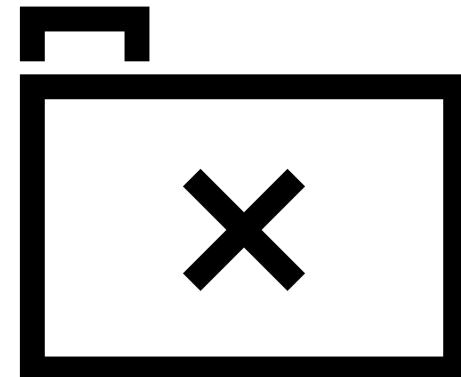
Correction vs Removal

Never “manually delete”.

Choose:

- Winsorize
- Cap values
- Replace with rolling median
- Flag for exclusion
- ...

The pipeline must explain itself.



Auditability > Convenience.



Outlier Detection and Correction

Data Curator enables **custom calculations** to diagnose, validate, and transform data into research-ready inputs.

We define explicit rules.

Those rules:

- Detect anomalies
- Correct distortions
- Standardize scaling
- Preserve chronology
- Remain reproducible

Outlier handling becomes part of the research system, **not a notebook hack**.

Custom Calculations

```
def c_test(m_open, m_close):
    """
    Example features calculation function.

    Receives the market open and market close columns, and returns a column with their difference.

    For this function to generate an output column, you need to:
    1. Place it in the Config/custom_calculations.py file (if it doesn't exist you can copy
       this file there).
    2. Add c_test to the Output_Columns sheet in the Config/parameters_datacurator.xlsx file.

    Parameters
    -----
    m_open : kaxanuk.data_curator.DataFrameColumn
    m_close : kaxanuk.data_curator.DataFrameColumn

    Returns
    -----
    kaxanuk.data_curator.DataFrameColumn
    """
    # we're just doing a subtraction here, but you can implement any logic
    # just remember to return the same number of rows in a single column!
    return m_close - m_open
```

Some Examples



Outlier Adjusted Data

- **Price Spike Filter** — Flag returns exceeding a multiple of rolling volatility and cap or replace when no corporate action justifies the move.
- **Corporate Action Adjustment** — Detect split- or dividend-driven price jumps and adjust the historical series to preserve economic continuity.
- **Ratio Explosion Control** — Nullify or redesign ratios when denominators approach zero to prevent artificial signal distortions.
- **Abnormal Growth Check** — Identify extreme revenue or earnings changes without event support and winsorize or smooth accordingly.
- **Extreme Z-Score Cap** — Bound cross-sectional z-scores at predefined limits to stabilize rankings and reduce signal instability.
- **Missing Data Guardrail** — Apply exclusion or forward-fill only under predefined, financially justified rules.
- **Negative or Impossible Values Filter** — Flag and correct financially inconsistent observations (e.g., negative shares outstanding).
- **Stale Price Detection** — Detect abnormal price stagnation and classify as illiquidity or potential data error.

Feature Engineering

From Data to Alpha Signals and Risk Factors

What Is a Feature?



A **feature** is a **structured variable** derived from data that captures information relevant for **prediction** or **risk modeling**.



Hypothesis-Driven Feature Engineering

Avoid: Data mining 200 features and picking the best.

Instead: Start from hypothesis.

Example

“Markets underreact to earnings surprises.”

Then engineer:

- Standardized earnings surprise
- Post-event drift
- Volume confirmation

Feature engineering follows theory.



Types of Features

Univariate ($1 \times N$) Features

- Derived from each asset's own history.
- Examples: Trend, rolling volatility, growth rates, historical z-scores.

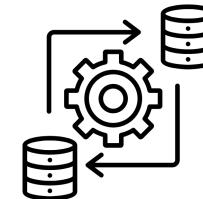


Data
Curator

Multivariate ($N \times N$) Features

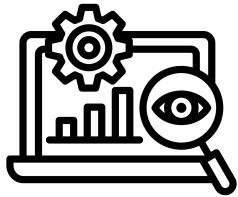
- Relative measures across assets at a given time.
- Examples: Sector-neutral ranks, cross-sectional z-scores, relative strength.

Data
Weaver

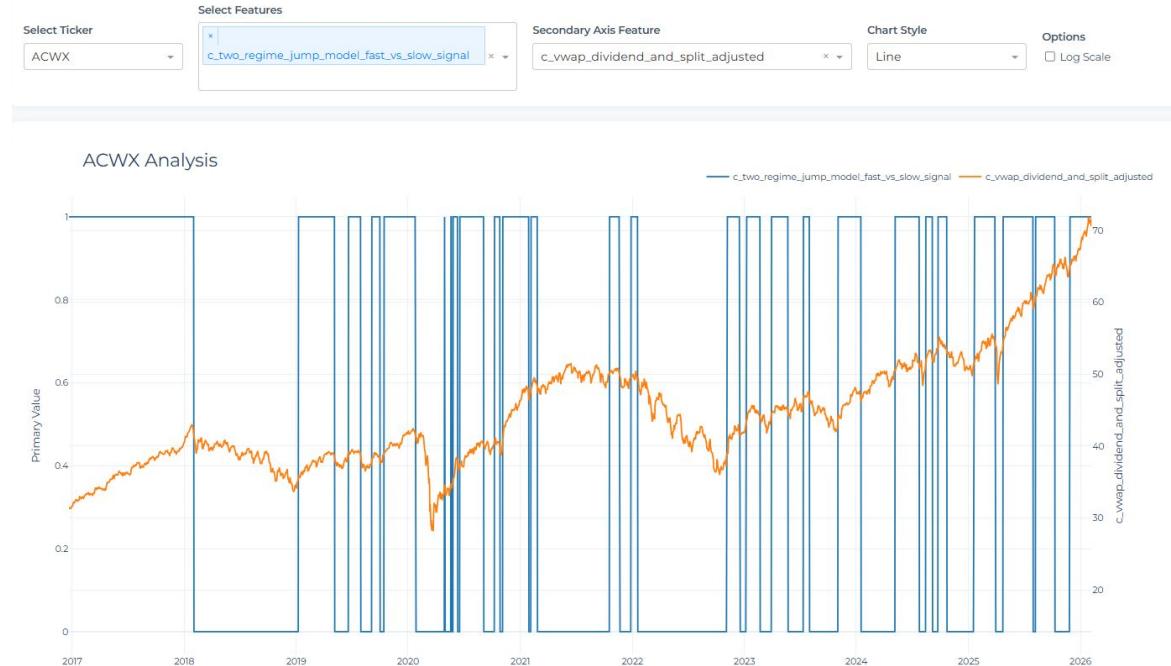




Feature Analysis



Data
Analyzer



Features Can Be Used as Alpha Signals or Risk Factors



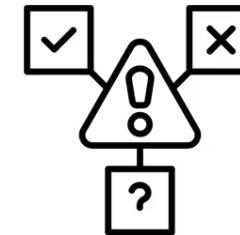
Alpha Signals

Idiosyncratic predictive features
designed to capture pricing inefficiencies
and generate **excess returns**.



Risk Factors

Systematic drivers of returns that are
common across assets and **explain risk premia**.





Some Examples

Alpha Signals

- **Earnings Surprise** — Abnormal returns following unexpected earnings results.
- **Short-Term Reversal** — Mean reversion after extreme short-term moves.
- **Residual Momentum** — Trend persistence unexplained by major risk factors.
- **Quality** — Excess returns linked to profitability and balance sheet strength.
- **Accruals** — Mispricing related to low earnings quality.
- **Sentiment** — Predictive signals derived from news or textual tone.
- **Dispersion** — Return opportunities arising from cross-sectional disagreement.
- **Event-Driven** — Temporary mispricing around corporate events (M&A, buybacks, guidance).

Risk Factors

- **Market** — Exposure to overall asset class return movements.
- **Sectors** — Exposure to industry-specific economic dynamics.
- **Beta** — Sensitivity of an asset's returns to the market benchmark.
- **Size** — Exposure linked to company market capitalization.
- **Value** — Exposure to relatively undervalued companies.
- **Momentum** — Exposure to persistent price trends.
- **Volatility** — Exposure to return variability and risk level.
- **Growth** — Exposure to companies with strong fundamental expansion.

Let's code!

Hands-On: Your First Data Pipeline

Disclaimers



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