

# Enhanced Institutional Rotation Detection

## Methodology for Early Signals

### Overview

Detecting institutional accumulation or distribution **earlier** than traditional filings requires expanding beyond quarterly reports. The enhanced methodology builds on the existing rotation detection framework <sup>1</sup> by integrating *real-time SEC data, nuanced insider clues, and advanced analytical tools*. This approach aims to catch “quiet” institutional moves – such as stealth buying in dark pools or pre-announced insider trades – that precede public awareness. By focusing on primary sources (SEC filings, ownership disclosures) and **overlooked signals** (derivative footnotes, special insider arrangements), we can reduce data lag and improve early-warning capabilities. The result is a more proactive system that identifies rotations like those in GameStop (GME) *ahead* of the traditional 13F/quarterly news cycle.

### Primary Data Integration (Beyond 13F)

Traditional rotation detection relied heavily on Form 13F (quarterly fund holdings) and occasional Schedule 13D/13G filings <sup>2</sup>. However, **Form 13F data is delayed 45 days after quarter-end, making it stale upon release** <sup>3</sup>. To get ahead of these lags, the enhanced methodology incorporates a broader set of timely SEC filings and data sources:

- **Schedule 13D & 13G (Beneficial Ownership):** These must be filed when an investor crosses 5% ownership (13D for activists within 10 days; 13G for passive holders, often annually or within 10 days if >10%) <sup>4</sup> <sup>5</sup>. Monitoring 13D/13G in near-real-time is critical – a new 13D can reveal a **large stake accumulation mid-quarter** (well before 13F reports). Early detection comes from polling EDGAR’s feeds continuously so that as soon as a 13D/G is disseminated, the system flags the stake. For example, an activist 13D filed today signals accumulation over recent days, providing an “*early warning*” versus waiting for end-of-quarter reports <sup>6</sup>.
- **Form 4 and 3 (Insider Transactions):** These Section 16 filings report insider trades within **2 business days** of the transaction. Incorporating Form 4 data means capturing *insider buying or selling* activity almost immediately. **Insider buying is particularly telling** – insiders only buy stock for one reason (they expect it to rise) <sup>7</sup> <sup>8</sup>, whereas sales can be for diversification or taxes. Clusters of insider buys (or unusual one-off large purchases by key executives) can foreshadow positive moves, so the pipeline will parse Form 4s nightly. (Insider selling, while more common, is less predictive on its own <sup>8</sup>, but a wave of insider sales combined with other signals might reinforce a distribution alert.) Form 3 (initial ownership) is also parsed, as it can reveal when a new officer or major holder joins – e.g. if a renowned investor is appointed to the board and files a Form 3 showing a stake, it hints at confidence or planned changes.
- **Form 8-K and 10-K/10-Q Disclosures:** Though not primarily ownership filings, these can contain valuable clues. Item 5 of an 8-K may announce if a significant investor has entered a cooperation agreement or if there’s a change in control or new investment. **10-K/10-Q** annual and quarterly reports often list **5% beneficial owners** and can reveal changes in top holders or share repurchase plans. Proxies (DEF 14A) list large shareholders and insider equity awards. The

enhanced system cross-references each 10-K/Q and proxy for changes in the roster of top owners or any mention of new significant holdings (sometimes an activist accumulates <5% and thus no 13D, but the company's proxy might list them if above 5% by record date).

- **Dark Pool Trade Reports:** Traditional methods largely ignore *off-exchange trading*, but these can signal quiet accumulation. Institutions often use **Alternative Trading Systems (dark pools)** to buy/sell large blocks *without alerting the public market* <sup>9</sup>. FINRA now publishes weekly OTC (ATS & non-ATS) volume data by security <sup>10</sup>. By integrating these reports, the methodology can detect unusual spikes in off-exchange volume or block trade prints in GME and similar stocks. **Example:** If 40%+ of GME's trading volume for several weeks is consistently off-exchange (above normal levels), it may indicate "*dark pool quiet accumulation*" – large players stealthily building a position <sup>9</sup>. Monitoring ATS data gives an earlier clue to such rotation, which might only show up in ownership filings much later.
- **Ownership Schedule Parsing:** All relevant SEC ownership schedules are monitored, including any Schedule 13F amendments and Form 13H (large trader registrations) if applicable. While 13H doesn't disclose holdings, it identifies *who* the big traders are – our system flags any new 13H filers in GME's ecosystem to know which entities might be executing huge volumes. Additionally, **Form 144** (notice of proposed insider sale) is now electronically available. A surge in Form 144 filings for a particular stock could presage significant insider selling (distribution) *before* the actual sales occur. These primary filings, accessible via EDGAR, form the backbone of the enhanced data ingestion pipeline <sup>11</sup>.

**Data Pipeline Improvements:** Instead of relying on manual quarterly updates, the system uses EDGAR's new **JSON API** to pull filings continuously. EDGAR's API provides real-time JSON updates of company submissions throughout the day <sup>12</sup>, enabling a "live feed" of ownership events. Every few minutes, the pipeline queries for any new filings of interest (13D, 4, etc.) for target companies or investors. This real-time ingestion dramatically cuts detection latency – we no longer wait for data vendors or end-of-day scraping. Each filing is parsed (via SGML/XBRL or JSON) into an **ownership delta pipeline** that updates positions immediately and calculates changes. For instance, if a 13G shows Vanguard increased its GME stake from 4% to 8%, the system computes that delta on the fly and evaluates it as an accumulation signal **on the filing day**, rather than waiting for quarter-end comparisons.

## Lesser-Known Signals & Footnote Analysis

Institutional moves often leave subtle traces in **footnotes and ancillary disclosures** that traditional screens overlook. The enhanced methodology adds text-mining and targeted analysis to surface these hidden signals:

- **Special Insider Arrangements:** We scan 8-Ks, proxies, and employment agreements for hints of *unusual insider contracts* that might presage trading. For example, if a CEO's 8-K employment contract includes a provision for **automatic stock sales or transfers** under certain conditions (or the adoption of a 10b5-1 selling plan), our system flags it. Similarly, if the proxy discloses a new **golden parachute or change-in-control payout** heavily tied to stock price, it could incentivize certain insider actions (like facilitating a takeover or buyout). These arrangements can foreshadow **strategic shifts** or exits that coincide with institutional rotation (e.g. an LBO possibility might cause some institutions to accumulate).

- **Vesting Schedules & Option Exercises:** Equity vesting events can lead to sales (insiders often sell shares when their stock awards vest). The methodology ingests the **vesting schedules** disclosed in proxies (e.g. large blocks of RSUs vesting on a certain date) and sets watches around those dates. Unusual patterns – say, a one-time mega grant that vests all at once next quarter – can lead to *significant insider liquidity events* (distribution). By tracking this, we anticipate when insider selling pressure might occur in the market **before** the Form 4s hit. We also monitor Form 4 footnotes for indications of **option exercises** or conversions (e.g. an insider exercising options and simultaneously selling shares in a planned manner). If multiple executives are exercising options early (perhaps to cash out), it may indicate an expected peak in stock price or some planned rotation.
- **Unusual Option Grants or Derivative Awards:** The presence of *structured or large option grants* can hint at insider sentiment or forthcoming events. For instance, if a company grants the CFO a large batch of performance-based stock options with a short expiration or at-the-money strike, it suggests an expectation of stock movement soon. Our system parses DEF 14A and 8-Ks for any such **one-off grants** or amendments to equity incentive plans. These are catalogued as potential “insider confidence” signals (if new grants) or “insider caution” (if an insider hastily exercises options to lock gains). Additionally, **footnotes about derivatives** in ownership filings are key. Activist investors sometimes use cash-settled equity swaps or forward contracts to build stakes under the radar (since these might not immediately count as “beneficial ownership”). We employ NLP on Schedule 13D filings and their exhibits to catch telltale language like “*total return swap*”, “*cash-settled forward*”, or “*options agreement*”. Any disclosure that a filer **holds derivatives referencing the stock** is noted. Such footnotes often reveal the true economic exposure – e.g. an activist might own 4.9% in shares (just below the 5% threshold) but have economic exposure to another 5% via swaps; this would be flagged as a *hidden accumulation* strategy. While historically these tactics evaded prompt disclosure, regulators are moving to tighten rules (e.g. proposed Schedule 10B for swap positions) to prevent stealth builds <sup>6</sup> <sup>13</sup>. Until then, our methodology treats derivative position footnotes as red flags for early accumulation.
- **Structured Trades & 10b5-1 Plans:** We also capture indicators of *planned trading* which might dampen or enhance signal interpretation. A Form 4 often indicates if a sale was under a **Rule 10b5-1 trading plan** (sometimes via a footnote or a checkbox in newer forms). Sales under 10b5-1 are pre-scheduled, so a wave of such filings is less an *informational* sell signal and more a routine sell-off. Conversely, **termination of a 10b5-1 plan** can be a signal – if an insider ends their plan or dramatically alters it (info sometimes found in footnotes or subsequent filings), it might mean they intend to sell outside the plan (potentially more opportunistically, which could align with negative news or rotations). The enhanced system logs which insider trades are automated vs discretionary, adjusting their weight in the rotation score.

In summary, by reading the fine print of filings and exhibits, the methodology captures *nuanced early signals*: a derivative hiding true ownership, an insider’s pre-planned sale, or an atypical equity award. These factors enrich the quantitative model with qualitative alerts that something unusual is afoot, allowing analysts to investigate further even before any price move or major filing occurs.

## Real-Time Monitoring Tools & Automation

Achieving timely detection across these diverse data streams requires a **robust, automated monitoring infrastructure**. We leverage resilient HTTP and API techniques to continuously harvest data without gaps:

- **EDGAR XML/JSON Feeds:** We subscribe to EDGAR's real-time feeds via the official APIs. As noted, the SEC's JSON submission feed updates throughout the day as filings are received <sup>12</sup>. Our tool queries the feed for target companies *and* for key investors (CIKs of major funds) to catch any filing that might indicate rotation. For instance, we monitor not just GME's filings, but also filings by large holders of GME (if known) – e.g. if BlackRock files a 13G increasing GME holdings, we catch it, and if Ryan Cohen (individual) files a Form 4 or 13D, we catch that too. This ensures coverage of both *issuer* and *investor* filings relevant to the stock.
- **Ownership Delta Pipeline:** All new data points flow into an “ownership delta” computation engine. This pipeline instantly recalculates changes in ownership percentages *at the moment of filing*. By mapping each filing's reported positions to previous known positions, we generate **delta events** (e.g. Fund A +2 million shares, Fund B –5 million shares). These events are timestamped by transaction date (if available) and filing date. The result is a high-frequency time series of who's buying/selling a stock over time. We maintain a rolling window (e.g. quarter-to-date) of accumulated buys and sells from filings, so we can tell, say, halfway through a quarter that “known institutions have collectively bought X shares and sold Y shares” so far. This provides a *live* rotation score update rather than waiting until quarter-end. If the net balance flips significantly to buying early in the quarter, that could signal emerging accumulation, prompting an alert.
- **Agent-Based Pattern Detection:** The system employs *agent processes* (or bots) that look for *repeatable patterns* in the data. For example, one agent might track if **multiple insiders and one 5%+ holder file sales within days of each other** – a pattern suggesting a coordinated distribution or negative development known internally. Another agent could monitor if a particular fund exhibits **copycat behavior** (e.g. Fund X often buys right after Fund Y dumps – indicating an uptake pattern). These agents use historical rotation data and machine learning to flag when current activity matches a past pattern associated with big moves. By encoding these patterns, the system becomes proactive: it doesn't just wait for a threshold breach, but can say “this configuration of insider and institutional actions looks like the early stage of a past rotation event.” In effect, the algorithm learns the *signature* of early rotation.
- **Resilient HTTP Infrastructure:** Scraping EDGAR and other sources at high frequency demands a fault-tolerant HTTP system. We use the **Resilient HTTP** library (AirNub's tooling) with features like automatic pagination, retry policies, and request interceptors to handle the data fetching at scale. This means our scrapers can traverse multi-page results, respect SEC's rate limits, and gracefully recover from errors. For instance, when pulling a large list of Form 4 filings for all insiders of GME, the library's pagination utility iterates through pages of results in a robust way (offset or cursor-based) without missing any items <sup>14</sup> <sup>15</sup>. *Interceptors* enforce backoff if the SEC site responds with throttling, and ensure all HTTP calls conform to a central “policy brain” for rate limits <sup>16</sup> <sup>17</sup>. This prevents overload (e.g. pausing requests if we approach EDGAR's threshold). The resiliency setup also includes logging and telemetry – if a scraping agent fails or a page is skipped, it's recorded and retried, so the monitoring never silently fails. In short, a “**satellite**” **HTTP system** orbits the data sources, continuously pulling updates with high reliability.

- **Alerting and Latency Modeling:** Every incoming filing or data update is evaluated not just for content but also for timing. We map the **transaction date vs. filing date** for each event to understand latency. This creates a model of how delayed each type of information is. For example, if a big insider sale on June 1 hits the tape (market) and the Form 4 is filed on June 3, that 2-day gap is latency. Over time, we learn which insiders or institutions tend to file last-minute versus promptly. The system can then account for these in alerts: if we detect an especially *large* transaction in market data (say, an unexplained block trade) and we know certain institutions often file 13D amendments a week late, we might proactively flag: “Unusual \$100M block trade – could be a late-reporting 13D filer accumulating, keep watch for a filing in coming days.” By mapping typical delays, we can issue *pre-filings alerts* when market activity suggests a likely undisclosed rotation in progress. This bridges the gap between **market surveillance** and **filing data**.

## Alternative Data from Investor Communications

In addition to official filings, the methodology taps **web-based investor communications** to glean early hints of rotations:

- **Hedge Fund Letters & Presentations:** Many institutional investors discuss their portfolio moves in quarterly letters or investor deck presentations (often published on their websites or discussed on earnings calls). For example, a hedge fund might mention “we initiated a position in GME this quarter” in an investor letter *before* the 13F is filed. Our system uses web crawlers and RSS monitors to collect such letters and presentations from known large holders or activists. We then apply text analytics to identify mentions of specific stocks or changes in tone (e.g. a fund that was negative on GME last quarter is suddenly praising its prospects – implying they may have rotated into it). By correlating these narratives with subsequent filings, we often find the **letters foreshadow the filings**. Incorporating this source can give us a heads-up on accumulation: if two prominent funds both write bullish commentary on GME in July, we anticipate increased 13F positions when the Q3 filings arrive – an early accumulation signal in July rather than mid-November.
- **Schedule 13D Exhibits & Activist Websites:** Activist investors, when filing a Schedule 13D, sometimes attach letters to management or detailed presentations as exhibits. These can contain **forward-looking plans** (e.g. push for a sale, board changes) that signal potential upcoming rotations by others (other institutions might buy in anticipation, shorts might cover, etc.). We automatically retrieve and scan 13D exhibits for such content. Also, activists often simultaneously release their letters on websites or social media. By monitoring these channels (e.g. press releases on the activist’s site or their Twitter feed), we ensure the moment an activist position is public (even slightly before EDGAR updates), we capture it. This is especially important for early detection because activists often coordinate media with filings to amplify impact.
- **Insider Interviews and Web Disclosures:** Occasionally, executives might disclose intentions in interviews or at conferences (e.g. “our CEO plans to buy more shares” or “Founder hints he may reduce his stake for philanthropy”). While not official filings, these hints can precede actual Form 4 transactions. Our methodology includes monitoring news and transcripts via financial news APIs for any **insider ownership commentary**. If, say, a director in an interview mentions they think the stock is undervalued (often a prelude to them buying shares), we take note as a potential accumulation signal. Likewise, if a large investor publicly talks down the stock or hints at profit-taking, it foreshadows distribution.

- **Correlation of Public Statements with Filing Timeline:** We maintain a mapping of when such communications happen relative to filings. For instance, if a fund letter in early May mentioned a new stake and we then see a 13G filed mid-May, that sets a precedent. In the future, the moment we see a similar mention, we won't wait – we'll treat it as if a filing is coming. This *accelerates our reaction time* relative to the market, which might not price in a position until the filing hits the wire or a 13F is out. Essentially, web disclosures act as **file-early indicators** in our model.

By combining these alternative sources with the core data, the methodology casts a wider net. It recognizes that not all informative signals come through EDGAR alone – some arrive in **investor discourse** or media before official channels. Tapping into that gives us a crucial edge in identifying rotations at the incubation stage, not just after they've hatched.

## Latency-Aware Analysis & Pattern Recognition

A critical innovation in this enhanced approach is explicitly accounting for **time lags** and using pattern analysis to infer flows that are not directly observable:

- **Transaction Date vs. Filing Date Mapping:** For each ownership change detected, we log the **transaction date** (when the trade actually occurred, if disclosed) against the **filing date** (when it became public). This allows us to construct a “latency profile” for different forms and filers. For example, insiders (Form 4) generally have ~2 days latency, 13D activists up to 10 days, and 13F investors up to 45 days. By mapping these, we create a timeline of what *could* have been happening before the public knew. This is invaluable for understanding market moves. If GME rose 30% in early January and only later we see a 13G from a big fund dated January 5, we learn that the accumulation *preceded* the spike. Repeating this across events, we can often deduce patterns like “stock XYZ often rallies before Fund ABC's buys are disclosed – perhaps traders anticipate that fund's moves.” Such insights feed back into an **AI model** that tries to predict ongoing accumulation from price/volume anomalies combined with knowledge of who hasn't filed yet. In effect, we become *latency-aware*: the model can say “we have evidence of buying that likely happened last week but isn't public – treat that as a current accumulation signal with X% confidence.”
- **Rotation Pattern Library:** We enrich the system with an expanding library of *historical rotation cases*, including known patterns of accumulation/distribution. For instance, a “coordinated rotation” pattern might be: one top holder sells a huge stake while 3–4 smaller institutions all increase stakes the same quarter <sup>18</sup> <sup>19</sup>. A “stealth accumulation” pattern might involve steadily rising off-exchange volume (dark pool buys) and modest price upticks with no news, followed by a 13D weeks later. By cataloguing such patterns (using Graph analysis and machine learning), the system can **match** current data against them. If it finds a match – say current activity in GME closely resembles a past rotation pattern that led to a big rally – it will raise the alert level and even provide an explanation like “*Pattern match: quiet accumulation resembling the Q4 last year scenario by institutions.*” This pattern recognition is augmented by the **GraphRAG** approach: a graph-based AI analysis that infers likely buyer-seller relationships <sup>20</sup>. The graph component maps institutions and stocks as nodes, linking “who sold” to “who bought” insofar as the data suggests. While we can't know exact transfers, graph algorithms help **infer flows** by looking at overlaps (e.g. if Fund X's sell coincides with Fund Y and Z's buys, the graph links  $X \rightarrow \{Y, Z\}$  as a probable flow) <sup>21</sup>. Over time, communities of frequent co-traders emerge. This allows detection of *coordinated behavior*: if a known cluster of hedge funds all start buying the same small-cap within a month, it might indicate a shared thesis or information – a strong bullish rotation signal.

- **Short Interest and Market Impact Integration:** Our model also considers **short interest changes** and price impact around rotation events. The existing methodology already uses a short interest “relief” metric <sup>22</sup>. We enhance this by mapping *when* short interest drops relative to rotation signals. If we see that shortly after a suspected accumulation, short interest substantially declines (covering) <sup>23</sup>, it corroborates that institutions are absorbing shares, forcing shorts out – a bullish sign. Conversely, if rotation indicators trigger but short interest rises, it might be a false positive (perhaps driven by an impending negative event). Furthermore, by analyzing **price patterns** around known rotations (event studies), we adjust our thresholds dynamically. For example, if historically a 5% stake increase by a top holder led to +10% stock move after disclosure <sup>24</sup>, then seeing a similar stake increase undisclosed should merit a strong alert even if raw volumes seem small. Essentially, we calibrate signals by *impact significance*, gleaned from historical data.
- **Latency-Aware Scoring:** All of the above factors feed into an improved scoring algorithm for rotation risk/opportunity. We maintain the core R-score formula (combining dump magnitude, uptake, etc.) but extend it with new components and time-aware adjustments. A **Latency Score** might down-weight signals that are very stale (e.g. a 13F from 45 days ago) and up-weight signals that are fresh (e.g. a Form 4 from yesterday). We also incorporate a **confidence level** for each signal source: an insider buy (Form 4) might provide high confidence of accumulation, whereas an increase in dark pool volume is a more ambiguous signal on its own. The scoring model uses these confidence weights to avoid false alarms. Only when multiple early indicators converge – e.g. *insiders buying + high dark pool volume + modest price uptick* – does the system raise a strong early accumulation alert. This layered approach filters noise but ensures that truly significant early rotations (the kind that “smart money” is quietly executing) are recognized and surfaced for analysts to act on.

## New Edge Enhancements (Summary of Improvements)

Compared to the previous rotation detection methodology, this enhanced approach introduces several **new edges** to catch institutional moves sooner:

- **Real-Time Filing Surveillance:** Continuous monitoring of SEC EDGAR feeds (using EDGAR's API) for Forms 13D/G, 4, 3, etc., enabling instant detection of ownership changes **throughout the quarter** (not just at quarter-end) <sup>12</sup>.
- **Insider Transaction Signals:** Incorporation of insider buying/selling data from Forms 4 and 144, with emphasis on clusters of insider **buying (a bullish signal)** and context for insider sales (10b5-1 vs open-market) <sup>7</sup> <sup>8</sup>.
- **Footnote & Derivatives Parsing:** Automated NLP analysis of filing footnotes and exhibits to uncover **hidden positions** (swaps, forwards, vesting events) and unusual insider arrangements that often precede market moves <sup>6</sup> <sup>13</sup>.
- **Dark Pool & Off-Exchange Data:** Integration of FINRA's alternative trading system reports and block trade analytics to detect **“quiet accumulation/distribution” in dark pools** that doesn't immediately reflect in exchange volume <sup>9</sup> <sup>10</sup>.
- **Web & Investor Communication Mining:** Leveraging hedge fund investor letters, presentations, and activist announcements to flag stocks being accumulated or targeted **before**

**official filings** confirm the position – effectively using *qualitative early warnings* to anticipate quantitative filings.

- **Latency-Aware Modeling:** Tracking the gap between trade execution and public report for each signal, enabling the system to *infer ongoing accumulation* during the lag. This includes alerting on anomalous price/volume action that likely indicates an undisclosed institutional trade in progress.
- **Graph-Based Flow Inference:** Deployment of graph analytics (e.g. GraphRAG) to connect the dots between sellers and buyers, inferring likely transfer of shares among institutions. This helps identify **coordinated rotations** and information-sharing networks, adding insight beyond raw filing data <sup>21</sup>.
- **Resilient Automation & Monitoring:** A hardened data pipeline using **resilient HTTP libraries** for scraping and API calls, complete with pagination handling, centralized rate-limit policies, and interceptors for retries <sup>16</sup> <sup>17</sup>. This ensures no filings or data points are missed due to technical issues, and the monitoring continues uninterrupted 24/7.

Each of these enhancements provides an **edge** in timing or depth of insight. Together, they transform the institutional rotation detector into a far more proactive system – one that not only reacts to filings faster than ever, but also *anticipates* them using ancillary clues. In fast-moving situations like GameStop’s volatile ownership shifts, these capabilities increase the likelihood of detecting big institutional moves at the earliest hints, empowering investors or risk managers to respond before the rest of the market catches on.

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<sup>1</sup> <sup>2</sup> <sup>3</sup> <sup>18</sup> <sup>19</sup> <sup>20</sup> <sup>21</sup> <sup>22</sup> <sup>23</sup> <sup>24</sup> ROTATION\_DETECTION.md

[https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/features/ROTATION\\_DETECTION.md](https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/features/ROTATION_DETECTION.md)

<sup>4</sup> <sup>5</sup> SEC Finalizes Amendments Governing Beneficial Ownership ...

<https://www.dechert.com/knowledge/onpoint/2023/10/sec-finalizes-amendments-governing-beneficial-ownership-reportin.html>

<sup>6</sup> <sup>13</sup> Proposed Rules on Disclosure of Security-Based Swap Positions

<https://corpgov.law.harvard.edu/2022/01/24/proposed-rules-on-disclosure-of-security-based-swap-positions/>

<sup>7</sup> <sup>8</sup> <sup>11</sup> As a Gauge of Stock Market Health, Which Indicator Should You Be Looking At? Insider Buying or Insider Selling?

<https://www.investopedia.com/insider-buying-or-insider-selling-indicators-11760828>

<sup>9</sup> What Is Dark Pool Quiet Rotation™? - Candlestick Patterns Explained by TechniTrader®

<https://candlestickpatterns.pw/blog/2024/04/16/rotation-patterns-explained-with-stock-indicators/>

<sup>10</sup> OTC (ATS & Non-ATS) Transparency | FINRA.org

<https://www.finra.org/filing-reporting/otc-transparency>

<sup>12</sup> SEC.gov | EDGAR Application Programming Interfaces (APIs)

<https://www.sec.gov/search-filings/edgar-application-programming-interfaces>

<sup>14</sup> <sup>15</sup> <sup>17</sup> resilient\_http\_pagination\_spec\_v\_0\_3.md

[https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/specs/resilient\\_http\\_pagination\\_spec\\_v\\_0\\_3.md](https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/specs/resilient_http_pagination_spec_v_0_3.md)



16 **resilient\_http\_policies\_spec\_v\_0\_3.md**

[https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/specs/resilient\\_http\\_policies\\_spec\\_v\\_0\\_3.md](https://github.com/tradentic/institutional-rotation-detector/blob/581190a9f982690d4c548077161bf3d10f6829dd/docs/specs/resilient_http_policies_spec_v_0_3.md)