Stopping performance regression via changepoint

detection

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What is a changepoint?

Moments of abrupt change in the behavior of a time series are often cause for alarm as they may signal a significant alteration to the data generating process.

- <u>An Evaluation of Change</u> Point Detection Algorithms



Figure 2 from <u>Bayesian Online Change Point Detection</u>



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What are we going to talk about?

Our experience building a changepoint detection product to pinpoint significant performance changes.

Including

- Why changepoint detection?
- When are probabilistic methods right for you?
- Challenges we encountered
- Thought-process during decisions









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Ticker Plant

Ticker Plant feeds the vast majority of market data into Bloomberg's systems. If you use:

- The Bloomberg Terminal
- Data License
- B-PIPE

Then, you probably rely on Ticker Plant.

We service historical data requests and offer real-time publishing. Clients that are technically-advanced often care a *lot* about latency. When our latency changes, so does much of Bloomberg's.

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Low Latency – a feature of our product

Low latency is crucial to the success of market participants

We need to be able to know whether the efforts of our Ticker Plant teams are what's making our system's performance better or worse

We need to map actions to outcomes





What makes analyzing Ticker Plant latency hard?

- Latency is often influenced by factors beyond our control
 - News & politics Ο
 - **Black swans** Ο
 - Data feed bugs Ο
- **Expectations change by instrument and event** \bigcirc
- Rarely-traveled code paths with bursty activity

Sometimes avoiding latency is impossible; Latency will increase in periods of high market volatility



Actions to outcomes

What is measurable and actionable?

- Code rollouts
- Bugs
- Feature flags
- Hardware changes
- OS changes
- Back-pressure from downstream systems

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 $\begin{array}{c} \textbf{7.X} \rightarrow \textbf{8.X} \rightarrow \\ \textbf{9.X} \end{array}$

 $\textbf{S0} \rightarrow \textbf{S1} \rightarrow \textbf{S2} \rightarrow \textbf{S3} \rightarrow \textbf{S4}$

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Linux Penguin source: Larry Ewing (<u>lewing@isc.tamu.edu</u>) and <u>The GIMP</u>

Motivations



Simple Black Box

Breadth & depth of data >>> current ability to extract insights

We have many clients who are interested in a unique subset of the market; this is something we can't know about ahead of time

Establishing realistic performance tests that cover the universe of market scenarios and client access patterns is prohibitively expensive

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Action



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A gentle push

A Problem Appears

Decreases in performance sweeps across machines

- No clear pattern of stages or code rollout
- It's not even obvious when it started

Solution

Using a rolling confidence interval, we were able to flag changepoints in performance metrics

With our changepoints known, we could pull all the relevant metadata regarding builds, machines, tags, and feature flags for programmatic analysis



Generated with ChatGPT / DALL-E



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Worktime over time for a single machine (μ s); lower is better





Three separate latency series; lower is better



Lower is better; relatively consistent

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Lower is better; consistently elevated

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Lower is better; more ambiguous



Selection

We decided to go with Bayesian Online Change Point Detection

Traditional techniques are often competitive with modern techniques

This method fit our expectations for changepoints (despite our data not obviously being in the exponential-family)



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Challenges



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Expected challenges

Did well:

- Early tooling like visualization and analysis
- Investing in synthetic data generation

Did not do well (mostly MDLC):

- Tracking inter-annotator agreement
- Version all of the data clearly
- **Document & organize experiments**
- Narrowed the subset of data we focused on \bigcirc

naming

:



Unexpected challenges - Expected Changes

Dealing with expected changepoints

Our detector worked at consistently pointing out:

- Market open/close
- Machine maintenance
- Weekends
- Long Holidays



Watch out! A market is opening and closing!



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Unexpected challenges - Expected Changes



Upkeeping configuration files is surprisingly complex & toil-intensive



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Unexpected challenges - Clutter & Noise

SETTINGS

Q Search data rows

run_id	Cluster	Start time	End time	Confidence%	Mean%	Mean+, ms	Node	Hostname	Group	Arch	DC	Node Stage	Pint	Links	Ticket
2	17	2024-01-21 17:00 -5	-	93	14	0.00457	6036		1			s4		LINKS	CREATE
3	2	2024-01-20 17:00 -5	-	75	29	0.01388	36						N/A		CREATE
12	0	2024-01-21 17:00 -5	-	92	25	0.02744	6875						N/A		CREATE
13	9	2024-01-21 17:00 -5	124	72	34	0.04334	2827						N/A		CREATE
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1-5 of 12229 Rows per page

Our initial UI (mock data) Bloomberg TechAtBloomberg.com © 2025 Bloomberg Finance L.P. All rights reserved. Engineering

Unexpected challenges - Clutter & Noise

MARK ACCURACY	SETTINGS

UTC

Q e2e-test

IsTP 个	run_id	Target	Occurrence time	Confidence%	Mean+, ms	UserTags	Сриио	Ticket
0	1367	Work Time	2024-09-06 00:05	56	116.45093999999999	e2e-test, test	2787780f-467b-4da9-a866-bc7d953855f9	CREATE
0	1367	Work Time	2024-09-16 00:05	49	45.784319999999994	e2e-test, test	0c3f882f-c126-47e4-b0ad-0d72f0bfb905	CREATE
0	1367	Work Time	2024-09-17 00:05	40	-107.697297999999999	e2e-test, test	298f95ea-5317-488c-9799-6a40aa037979	CREATE
X	1367	Work Time	2024-09-05 00:05	65	9.385812999999999	e2e-test, test	9d5f2b5f-b2d0-4532-bb4a-eb942eb109eb	CREATE
~	1367	Work Time	2024-09-10 00:05	77	-64.00916	e2e-test, test	d3c64cbc-1157-4532-a858-7ca9b0c85827	CREATE

Rows per page

Our updated UI (mock data)

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Learning Summary



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Are Statistical Methods Right For Me?

Consider if you have some of the following:

- Many degrees of freedom for system perturbation
- A noisy environment
- The behavior of your system changes over time
- Your system has distributed responsibility
- A lot of data





What should I watch out for?

A lot more engineering work will be spent on usability, rather than on the core detection system

Also check out: Anomaly Detection in Time Series from Scratch Using Statistical Analysis - Ivan Shubin, SREcon24 EMEA



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Thank you! Questions?

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