Link to the review process, containing the sources, the list of antipatterns with category and relevance, along with the identification of the sources.

https://docs.google.com/spreadsheets/d/18Gn8Aesf8XBrMOFhV6j4mk6efP4OVYwHvT3WtAk7iZw/edit?usp=sharing

ID Count Name Category Relevance Source Failure to Set the Correct Alerts and 1 Thresholds Alert Low 17, 23 2 8, 15, 17, 18, 19, 23, 25, 26 8 2 Excessive alerts Alert High Only having time-series graph-based Dashboar dashboards 7, 17 2 3 Low Dashboar 13, 15, 16, Numerous Dashboards 4 High 17, 18, 19, 23 7 Lazy synthetic transactions General Low 18 1 5 6 Mandating tools General Low 18 1 23 1 Neglecting Data Retention General Low Positioning observability as tool to use during issues and incidents only 2 General Low 17, 23 8 9 Environment inconsistency General Low 15, 18 2 Not understanding your ecosystem General Low 15, 18 2 10 17, 18 2 11 Hoarding data General Low 2 12 One Size Fits All General Low 14, 24 Lack of Instrumentation Guidelines for New 13 Services General Medium 15, 16, 17 3 14 Using too many tools General Medium 15, 17, 24 3 3 Neglecting Regular Maintenance General 12, 22, 23 15 Medium 16 Observability for Production Only 3 General Medium 1, 14, 24 17 Throwing tools at a problem General Medium 13, 16, 17, 18 4 1, 6, 12, 16, 18 Lack of System-Wide Perspective 18 General High 5 1, 6, 15, 19, 19 Lack of Contextual Information General 23, 26 6 High 22 20 Failure to monitor error logs Low 1 Log

21	Using API Access Logs for Troubleshooting	Log	Low	14, 24	2
22	Improper Log Level Usage	Log	Low	9, 26	2
23	Lack of Structured Logging	Log	Medium	3, 15, 26	3
24	Duplicate Logs	Log	Medium	9, 3, 4,	3
25	Local Logging Only	Log	Medium	2, 7, 11, 26	4
26	Excessive Logs	Log	High	3, 4, 9, 13, 15, 26	6
27	The big dumb metric	Metric	Low	18	1
28	Relying Only on API Monitoring	Metric	Low	24	1
29	Misunderstanding metrics	Metric	Low	13, 18	2
30	Overlooking derived metrics	Metric	Low	25, 26	2
31	Only tracking what the customer sees	Metric	Medium	17, 22, 26	3
32	Excessive metrics	Metric	Medium	10, 13, 19	3
33	Inadequate monitoring coverage	Metric	Medium	2, 11, 20, 25	4
34	Forgetting the customer	Metric	High	1, 13, 14, 15, 17, 18, 24, 26	8
35	Inadequate Use of Logs for Metrics Collection	Metric, Log	Low	26	1
36	Bad sampling intervals	Metric, Trace	High	3, 4, 5, 13, 15, 17, 18, 21	8
37	Long Trace Spans	Trace	Low	15	1
38	Underestimating the Importance of Traces	Trace	Medium	15, 16, 17	3
39	No consistent trace ID	Trace	Medium	7, 13, 15, 18	4
40	Starting the Trace at the API Gateway is Overrated	Trace	Low	14, 24	2

Anti-Pattern Name: Failure to Set the Correct Alerts and Thresholds

Category: Alerts

**Context:** The anti-pattern occurs when alerts are configured without proper criteria, resulting in inaccurate threshold settings. This can happen due to a lack of historical data analysis, the adoption of arbitrary values, or the use of default configurations that do not reflect the system's actual behavior. This issue can manifest in two main ways: overly restrictive thresholds, which generate a high volume of irrelevant alerts and lead to alert fatigue, or overly permissive thresholds, which fail to detect critical issues in a timely manner.

**Problem:** Incorrect alert and threshold configuration can lead to several negative impacts. First, it can result in the excessive generation of irrelevant alerts, causing alert fatigue, where the team becomes desensitized to notifications. This can lead to critical incidents being overlooked due to overly permissive thresholds, as the system fails to trigger alerts for important events. As a consequence, trust in the monitoring system diminishes, and important alerts may be ignored, further hindering effective response. Additionally, teams may end up spending excessive time and effort on constant manual adjustments to thresholds, reducing operational efficiency and increasing the chances of missing significant issues.

**Example:** An operations team configures an alert for CPU usage above 70% on any server. During normal workload peaks, this alert triggers frequently, causing engineers to start ignoring it. When a real issue occurs and a server reaches 100% CPU usage for an extended period, the alert does not receive the necessary attention, resulting in severe service degradation before corrective actions are taken.

### **Recommended Solution:**

- Define alerts based on historical metric analysis and expected system behavior.
- Use dynamic approaches, such as anomaly detection and event correlation, to avoid inappropriate fixed thresholds.
- Prioritize actionable alerts, ensuring that each notification requires a clear and objective response.
- Implement periodic reviews of alert configurations to adjust thresholds as operational conditions evolve.
- Categorize alerts by severity to differentiate minor incidents from truly critical problems.

## Consequences:

## Positive:

- Reduction in alert fatigue and improved response to critical incidents.
- Increased reliability in monitoring, ensuring a quick reaction when necessary.
- Optimization of the team's time by avoiding frequent and unnecessary alert adjustments.

# Challenges:

- Requires an initial effort to analyze and properly define thresholds.
- May require advanced tools for automatic anomaly detection and event correlation.
- Ongoing review is necessary to adapt to system changes and workload variations.

Anti-Pattern Name: Excessive Alerts

Category: Alerts

**Context:** The anti-pattern occurs when a monitoring system generates an overwhelming number of alerts, often due to redundant rules, lack of aggregation mechanisms, or excessive granularity in notifications. This results in alert fatigue, where engineers become desensitized to alerts and start ignoring notifications, including those that indicate critical issues. The root causes of this problem include poorly defined alerting policies, a lack of prioritization, and inadequate suppression or grouping strategies.

**Problem:** An excessive number of alerts can lead to several issues, including alert fatigue, where engineers stop responding to alerts due to their overwhelming volume. As a result, the mean time to resolution (MTTR) increases, as critical alerts may get lost in the noise. Additionally, operational time is wasted handling low-priority or redundant alerts, diverting attention from more important issues. Furthermore, it becomes difficult to distinguish between minor issues and genuinely critical incidents, leading to potential delays in addressing significant problems.

**Example:** A system is configured to trigger an alert for every minor fluctuation in response time, causing hundreds of notifications per day. Engineers quickly become overwhelmed and begin ignoring the alerts. When an actual service degradation occurs, it goes unnoticed because it is lost among the flood of low-priority notifications.

## **Recommended Solution:**

- Implement alert deduplication and aggregation mechanisms to reduce redundancy.
- Define clear prioritization levels to differentiate between informational, warning, and critical alerts.
- Use dynamic alerting techniques, such as baselining and anomaly detection, to prevent excessive false positives.
- Regularly review alerting rules and remove or refine unnecessary notifications.
- Introduce rate-limiting and suppression rules to prevent excessive notifications within a short period.

## **Consequences:**

## Positive:

- Reduction in alert fatigue and improved focus on critical incidents.
- Increased operational efficiency, as engineers spend less time handling non-actionable alerts.
- Improved reliability of the alerting system, ensuring that important notifications receive the necessary attention.

# Challenges:

- o Requires an initial investment in configuring aggregation and suppression rules.
- May necessitate monitoring tools with advanced alert correlation capabilities.
- Needs continuous tuning to maintain the right balance between alerting coverage and noise reduction.

Anti-Pattern Name: Only Having Time-Series Graph-Based Dashboards

Category: Dashboards

**Context:** The anti-pattern occurs when monitoring dashboards rely exclusively on time-series graphs to present data. While time-series graphs are useful for visualizing trends over time, they are insufficient for providing a comprehensive view of system health. This issue arises when dashboards lack complementary visualization methods such as tables, heatmaps, topological maps, or alert summaries, making it difficult to extract actionable insights efficiently.

**Problem:** Relying solely on time-series graphs can lead to several issues, including difficulty in correlating multiple data sources, such as logs, traces, and alerts. This approach may also limit the ability to detect anomalies that are not apparent in a time-series representation. Engineers face inefficient troubleshooting, as they must manually interpret trends instead of relying on more direct indicators. Additionally, the cognitive load increases when identifying patterns across different system components, making it harder to quickly identify and resolve issues.

**Example:** A dashboard displays CPU utilization over time but lacks contextual information such as which services are affected, what logs were generated during spikes, or whether similar incidents have occurred in the past. As a result, engineers must switch between multiple tools and manually correlate data to diagnose performance issues.

- Incorporate diverse visualization techniques, such as:
  - Heatmaps for identifying patterns across multiple components.
  - Tables for displaying key metrics in a structured format.
  - Topological or dependency maps for understanding service interactions.
  - Correlated logs and traces to provide context for performance anomalies.
- Ensure dashboards provide actionable insights rather than just raw data.
- Design views tailored to specific use cases, such as incident response, capacity planning, or SLA monitoring.

Regularly review and refine dashboard layouts to improve usability and effectiveness.

# Consequences:

### Positive:

- Improved troubleshooting efficiency by reducing the time needed to diagnose issues.
- Enhanced observability through diverse and relevant data visualizations.
- Reduced cognitive load for engineers, allowing faster and more accurate decision-making.

# Challenges:

- Requires additional effort to design and implement effective visualizations.
- May necessitate training for teams unfamiliar with new dashboard elements.
- Needs continuous refinement to ensure dashboards remain relevant and actionable.

Anti-Pattern Name: Numerous Dashboards

Category: Dashboards

**Context:** The anti-pattern occurs when an organization accumulates an excessive number of dashboards, often without proper governance or consolidation. This situation typically arises when different teams create dashboards for similar purposes, when outdated dashboards are never retired, or when there is no clear strategy for dashboard management. As a result, users struggle to identify which dashboards are relevant and accurate, leading to inefficiencies in monitoring and troubleshooting.

**Problem:** Having too many dashboards can lead to several issues, including difficulty in finding the correct dashboard for a given situation. This can also result in inconsistent or conflicting data across multiple dashboards, reducing trust in the monitoring system. Additionally, the maintenance overhead increases as outdated or redundant dashboards remain in use. Engineers may experience cognitive overload when navigating numerous dashboards to diagnose an issue, making it harder to effectively troubleshoot and respond to incidents.

**Example:** An SRE team needs to investigate a performance issue but finds multiple dashboards with similar names, each displaying different metrics for the same service. Some dashboards are outdated, while others are maintained by different teams. As a result, the engineers waste valuable time determining which dashboard contains the most accurate and relevant data.

- Establish clear governance for dashboard creation and maintenance.
- Consolidate similar dashboards into unified views to reduce redundancy.

- Regularly audit and deprecate outdated or rarely used dashboards.
- Implement tagging and categorization to help users quickly find relevant dashboards.
- Promote dashboard standardization across teams to ensure consistency in data presentation.

# **Consequences:**

- Positive:
  - Improved efficiency in accessing relevant and accurate dashboards.
  - Reduced confusion and cognitive load when troubleshooting incidents.
  - Lower maintenance overhead by eliminating redundant dashboards.
- Challenges:
  - o Requires initial effort to review and consolidate existing dashboards.
  - o May face resistance from teams accustomed to their own custom dashboards.
  - Needs ongoing governance to prevent dashboard sprawl from recurring.

- 1. Antipattern Name: Lazy Synthetic Transactions
- **2. Category:** General
- **3. Context:** This antipattern occurs when synthetic transactions are executed lazily, meaning they are only triggered on-demand or when specific conditions are met, rather than being executed continuously or at regular intervals. This behavior is common in systems that aim to reduce overhead or in environments with limited resources. Teams may adopt this practice to save resources, but without realizing that it compromises system coverage, leaving critical parts under-monitored.
- **4. Problem:** Lazy synthetic transactions may not cover all critical areas of the system continuously, which can result in undetected failures. The intermittent monitoring weakens the diagnostic capabilities and increases the time needed to identify and resolve incidents, making it difficult to detect issues that occur between verification periods.
- **5. Example:** In an e-commerce platform, synthetic transactions are only executed during traffic spikes or when the system detects performance degradation. During low-traffic periods, there is no continuous monitoring. When a failure occurs outside of these periods, the team takes longer to identify the cause since there is no constant coverage of the critical transactions.

- Regular execution of synthetic transactions at more frequent intervals to ensure broader coverage of the system.
- **More comprehensive coverage**: Ensure synthetic transactions cover all critical functionalities, with frequency adjusted based on the priority of each area.

• **Dynamic adjustment**: Adapt the execution of transactions according to the system context, intensifying checks in more critical areas.

# 7. Consequences

## Positive:

- **Improved system coverage**, allowing for problem detection in real-time, not just during specific moments.
- Faster and more effective diagnostics, reducing incident resolution time.

# Challenges:

- **Increased resource overhead**, especially if synthetic transaction execution is not properly balanced.
- **Continuous adjustments** to the monitoring process may require additional effort in configuration and tool adaptation.

# 1. Antipattern Name Mandating Tools

2. Category: General

- **3. Context:** This antipattern occurs when organizations or teams enforce the use of a specific tool or set of tools across their entire stack or organization, regardless of the unique needs or context of different teams or projects. It often arises from a desire to standardize processes, simplify tool management, or save on costs. While it can provide consistency, this practice can lead to inefficiencies when the mandated tools do not align with the specific requirements or workflows of certain teams
- **4. Problem:** Mandating a single tool or platform can result in suboptimal performance, frustration, and inefficiency. Teams may be forced to work with tools that are ill-suited for their specific use cases, increasing operational complexity or decreasing productivity. It can also lead to resistance from team members, as they might prefer or need different tools to perform their tasks effectively.
- **5. Example:** A company mandates the use of a single logging tool across all engineering teams. While the tool works well for infrastructure monitoring, the development teams struggle with its limited customization options for application-level logging. As a result, they end up spending more time configuring and troubleshooting the tool rather than focusing on their core tasks.

# 6. Recommended Solution

Allow flexibility in tool choice based on the specific needs and expertise of teams, while
maintaining common standards where applicable.

- **Evaluate and adapt tools** to ensure they are suitable for each team's workflows and requirements, and make room for experimentation with new tools when necessary.
- **Standardize processes** rather than tools, promoting best practices without forcing a specific tool on every team.

## Positive:

- **Increased team satisfaction** and efficiency, as teams are empowered to choose tools that best fit their needs.
- **Fostering innovation** by allowing teams to explore new tools and technologies that may be more suitable for specific problems.

# Challenges:

- Lack of consistency in tool usage across teams, which may complicate cross-team collaboration or troubleshooting.
- **Potential increase in overhead** for tool management and integration, as different tools may need to be supported or integrated across the organization.

1. Antipattern Name: Neglecting Data Retention

2. Category: General

- **3. Context:** This antipattern occurs when organizations fail to properly manage the retention of observability data, such as logs, metrics, and traces. It can happen when teams do not define clear policies for how long data should be retained or when they do not actively monitor or adjust data retention periods. Often, this happens due to a lack of awareness of the impact of excessive or insufficient data retention, or simply because teams focus more on immediate needs rather than long-term data management.
- **4. Problem:** Neglecting data retention can lead to several issues, such as data loss, where critical logs or metrics may be deleted or become inaccessible before they are needed for troubleshooting or analysis. It can also result in excessive storage costs due to retaining data longer than necessary, leading to inefficient use of resources. Additionally, organizations may face difficulties in compliance, struggling to adhere to legal or regulatory requirements for data retention.
- **5. Example:** An organization collects application logs but does not establish a clear retention policy. Over time, logs accumulate in large volumes, consuming significant storage resources. When a critical incident occurs, relevant logs are lost because they were purged after the default retention period, and the team is unable to investigate the issue effectively.

- Define data retention policies based on the type of data, its value, and regulatory requirements.
- **Implement automated data retention management** to delete outdated data after it has fulfilled its purpose, ensuring storage efficiency.
- Monitor storage usage regularly to ensure that retention periods are adhered to and adjust them as necessary to balance cost and value.
- **Ensure compliance** with any legal or industry standards that require data retention for a certain period.

## Positive:

- Reduced storage costs by optimizing data retention periods.
- Improved incident response with more relevant data available for troubleshooting.
- Better compliance with legal and regulatory requirements for data retention.

# Challenges:

- **Initial effort** in setting up and defining retention policies for different types of data.
- **Potential for data loss** if retention periods are set too aggressively, potentially missing important historical data during incidents.

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- **1. Antipattern Name:** Positioning Observability as a Tool to Use During Issues and Incidents Only
- 2. Category: General
- **3. Context:** This antipattern arises when organizations or teams treat observability tools (such as logging, metrics, and tracing) as tools to be used only during active issues or incidents. Observability is often seen as a reactive tool, meant to help resolve problems after they arise, rather than as a proactive mechanism for maintaining system health and performance. This view can result in underutilization of observability capabilities, missing out on early detection of issues or trends that could prevent future incidents.
- **4. Problem:** When observability is viewed only as a reactive tool for incidents, it can lead to missed opportunities for proactive monitoring, where early signs of system degradation or emerging issues go unnoticed until they escalate. This approach also increases the mean time to detection (MTTD), as teams may lack the visibility needed to identify problems before they impact users or the system. Additionally, it results in a lack of performance optimization, as observability tools are not leveraged to improve system efficiency, leading to missed performance gains.
- **5. Example:** An organization only reviews logs and metrics when there's a major incident, like a service outage. In the meantime, there are performance degradations or intermittent failures

that go unnoticed because the observability tools are not actively monitored or analyzed outside of high-priority incidents. When the next incident occurs, the response time is slower, and the root cause analysis is more complicated, as the data was not continuously monitored or analyzed proactively.

## 6. Recommended Solution

- **Shift towards continuous observability**: Treat observability as a tool for ongoing system monitoring, not just as a reactive resource during incidents.
- Integrate observability into daily workflows: Use metrics, logs, and traces to track the health of services in real-time, even when there are no visible issues.
- **Set up proactive alerts** for early indicators of potential problems before they become major incidents.
- **Incorporate observability into performance optimization** to continuously analyze and improve system performance.

## 7. Consequences

### Positive:

- Earlier detection of issues, reducing the overall impact of incidents and improving system reliability.
- Proactive issue resolution, with teams able to prevent problems before they affect users.
- Continuous system improvement through ongoing performance insights, leading to more efficient systems.

# Challenges:

- Cultural shift may be required, as teams must view observability as an ongoing necessity rather than something only needed during crises.
- Initial setup effort to ensure observability tools are configured to support proactive monitoring, potentially leading to higher resource usage.

**1. Antipattern Name:** Environment Inconsistency

**2. Category:** General

**3. Context:** Environment inconsistency occurs when there are differences between development, staging, and production environments, which can lead to unpredictable behavior of applications or systems. These discrepancies might involve differences in configurations, versions of libraries or dependencies, or even infrastructure setups. Teams may overlook or neglect to properly replicate production-like environments for testing, leading to situations where an application works in one environment but fails in another.

- **4. Problem:** When environments are inconsistent, the risk of deployment failures increases, as changes that work in staging or development may fail in production due to hidden differences. This inconsistency also makes it harder to reproduce issues, complicating debugging and delaying incident resolution. Additionally, teams waste time and resources troubleshooting avoidable problems, leading to a higher mean time to resolution (MTTR).
- **5. Example:** A team develops and tests a new feature in a local development environment, but due to configuration differences, the same feature behaves differently when deployed to staging or production. In production, the feature fails due to a discrepancy in the version of a third-party library that was installed in the staging environment but missed in production. The team ends up spending extra time fixing the issue because it wasn't caught earlier in the development pipeline.

## 6. Recommended Solution

- **Ensure environment parity**: Make sure that the configurations, libraries, dependencies, and infrastructure in development, staging, and production are as similar as possible.
- Use Infrastructure as Code (IaC) to define and replicate environments consistently, ensuring that all configurations and dependencies are version-controlled and easily reproducible.
- **Automate environment validation**: Set up checks to validate that all environments are aligned before deployment to avoid issues with misconfigurations or outdated libraries.
- Test in production-like environments: Regularly test staging and development environments to match production as closely as possible to catch discrepancies early.

# 7. Consequences

## Positive:

- More reliable deployments with fewer unexpected issues arising after production releases.
- **Faster troubleshooting** as teams can more easily reproduce issues that occur in production in a consistent staging or development environment.
- **Improved confidence in changes** due to a more predictable and consistent behavior across environments.

# Challenges:

- **Initial setup and maintenance** of consistent environments, which might require additional resources and attention.
- Complexity of managing infrastructure parity, especially in larger, more complex systems with multiple dependencies.

1. Antipattern Name: Not Understanding Your Ecosystem

# 2. Category: General

- **3. Context:** This antipattern occurs when teams fail to gain a deep understanding of their own system or application ecosystem. This includes failing to grasp the relationships between various services, components, or dependencies, and the behavior of the system as a whole. Teams may focus too much on individual components without considering how changes affect the broader system, or they might neglect to monitor interactions between services, making it difficult to anticipate how one part of the system will behave in different conditions.
- **4. Problem:** When teams don't understand their ecosystem, they face an increased risk of system failures, as changes in one part of the system can cause unforeseen consequences elsewhere. Troubleshooting becomes inefficient, with longer resolution times due to a lack of visibility into system interactions. Additionally, poor optimization opportunities arise, as teams miss chances to improve performance, scalability, and reliability. Inconsistent behavior may also emerge, leading to discrepancies between staging and production environments or unexpected responses under load.
- **5. Example:** A development team introduces a new feature that relies on a microservice. However, they don't account for the service's heavy dependency on another external API that has rate-limiting. When the feature is deployed in production, it causes a bottleneck, and the system slows down unexpectedly. Since the team didn't understand the service's dependency in the ecosystem, they could not predict the failure.

### 6. Recommended Solution

- Create a system map: Maintain an up-to-date visualization of the entire ecosystem, including services, dependencies, and interconnections, to better understand the overall system behavior.
- **Ensure cross-team collaboration**: Encourage communication and collaboration between teams responsible for different services to ensure that everyone understands the broader context of the system.
- **Implement observability practices**: Use monitoring, tracing, and logging tools to track how services interact in real-time, allowing teams to observe dependencies and anticipate potential issues.
- **Perform impact analysis**: Before making changes to the system, analyze how they will affect other components and services in the ecosystem to identify potential risks.

## 7. Consequences

## Positive:

- Faster troubleshooting and incident resolution by having a clear understanding of how services interact and affect each other.
- **Improved system stability** through a better understanding of dependencies, leading to fewer unforeseen failures.

• **More effective optimization** opportunities that arise from understanding the full ecosystem, allowing for proactive scaling or performance improvements.

# Challenges:

- **Initial effort to map out the ecosystem**, which may require significant time and collaboration across multiple teams.
- Keeping the system map up to date as the system evolves, especially in complex or fast-moving environments.

1. Antipattern Name: Hoarding Data

2. Category: General

- **3. Context:** The antipattern occurs when different teams within an organization maintain their own observability platforms or datasets without sharing them with other teams. This often happens due to lack of trust, fear of failure, or simple resistance to change. This behavior creates communication and collaboration barriers between teams, hindering problem resolution and continuous system improvement.
- **4. Problem:** When observability data is not shared, it leads to a lack of transparency, preventing teams from having a clear view of the system as a whole, which results in inaccurate diagnostics and longer incident resolution times. The absence of data sharing also makes problem-solving ineffective, as teams lack access to the necessary information to resolve issues in distributed or complex systems. It can foster distrust between teams, damaging relationships and collaboration. Additionally, without information flow, the organization's ability to quickly identify areas for improvement is hindered, ultimately compromising system performance.
- **5. Example:** A platform engineering team maintains its own logging and metrics solution and does not share this information with other teams such as operations or security. As a result, when a problem occurs in production, no one has access to the full data needed to investigate the issue, and the solution is delayed while teams search for isolated information in their own platforms.

- **Promote a culture of sharing:** Encourage transparency and collaboration across teams, ensuring that everyone has access to relevant observability data.
- **Centralize data:** Use a centralized observability platform where all data (logs, metrics, traces) can be easily accessed and analyzed by any team.
- **Foster trust:** Work to remove cultural barriers that prevent data sharing and promote a more collaborative and open environment.
- **Continuous training:** Offer training for teams on how to work with observability data in a safe and efficient manner.

# Positive:

- **Improved collaboration**: Teams that share data have better communication and collaboration, leading to more effective performance.,
- **Faster incident resolution:** With shared access to data, teams can resolve issues faster and more accurately.
- **Increased trust:** Data sharing builds a more trustworthy and transparent work environment, improving organizational culture.

# Challenges:

- **Cultural resistance:** Changing a culture accustomed to isolating data can be challenging and time-consuming.,
- **Ensuring security and compliance:** Ensuring that sensitive data, such as customer information, is handled appropriately during the sharing process.

1. Antipattern Name: One Size Fits All

2. Category: General

- **3. Context:** The "One Size Fits All" antipattern occurs when organizations try to apply a single, uniform solution, tool, or approach across the entire system or organization without considering the unique needs of different teams, services, or use cases. This typically happens when there's a push for standardization or cost reduction, but the solution fails to meet the diverse requirements of various teams. While the goal is often to simplify processes or reduce overhead, it can lead to inefficiencies, frustration, and suboptimal performance.
- **4. Problem:** When a single solution is applied universally, it leads to a lack of flexibility, as different teams or systems may have unique needs that aren't met by the one-size-fits-all approach, resulting in workarounds or inefficiencies. This can cause suboptimal performance, as tools or practices that work well for one team might not be effective for others, wasting resources. Teams may also resist adopting tools or approaches that don't suit their workflows, reducing collaboration. Additionally, the approach prevents missed opportunities for optimization, as different parts of the system could benefit from specialized tools or practices.
- **5. Example:** A company mandates the use of a single log aggregation platform for all teams, regardless of the diverse needs of engineering, data science, and support teams. While the platform works well for some teams, others struggle with its limitations. The engineering team needs advanced querying capabilities for debugging, but the tool lacks those features. Meanwhile, the support team finds it difficult to parse and navigate the logs due to the platform's complexity. As a result, both teams waste time and effort trying to adapt the tool to their needs, and efficiency suffers.

## 6. Recommended Solution

- Adopt a flexible approach: Allow teams to choose tools and practices that best fit their specific needs, while maintaining high-level standards and integration.
- **Enable customization**: Implement solutions that can be tailored to the needs of individual teams or use cases, allowing flexibility without sacrificing overall consistency.
- Encourage cross-team collaboration: Foster communication between teams to share knowledge and align on best practices, while also respecting each team's unique requirements.
- Consider a modular approach: Choose systems or frameworks that can be adapted and extended for different use cases, ensuring that the organization can benefit from both standardization and customization.

# 7. Consequences

## Positive:

- **Increased team satisfaction** and productivity, as teams can use tools and processes that better fit their needs.
- **Improved performance** and efficiency by using solutions that are optimized for specific tasks and workflows.
- **Better adoption** of tools and practices, as teams are more likely to engage with solutions that are tailored to their context.

# Challenges:

- **Potential fragmentation** across teams if not properly coordinated, leading to integration challenges or inconsistencies.
- **Increased complexity** in managing multiple tools or approaches, requiring additional coordination or governance.

1. Antipattern Name: Lack of Instrumentation Guidelines for New Services

2. Category: General

- **3. Context:** This antipattern occurs when organizations fail to establish clear and consistent guidelines for instrumenting new services with the necessary observability practices (logging, metrics, tracing, etc.). As a result, new services are often built without the proper instrumentation, or with inconsistent levels of observability. This can happen when there is no formal onboarding process for observability, or when individual teams take different approaches to instrumentation, leading to gaps in monitoring and difficulties in maintaining system-wide visibility.
- **4. Problem:** When there are no clear instrumentation guidelines, it leads to inconsistent observability, as some services may be well-instrumented while others lack sufficient visibility,

making it difficult to monitor the overall system. This can delay issue detection, as problems in new services may go unnoticed for longer, increasing resolution times and user impact. Debugging becomes harder when services aren't consistently instrumented, due to insufficient data to understand system behavior. Additionally, teams may face increased overhead in troubleshooting, having to retroactively add instrumentation to new services, which wastes time and delays issue resolution.

**5. Example:** A new microservice is deployed without standard logging and metric tracking, because there were no guidelines provided for observability. When an issue arises in production, the engineering team struggles to gather the relevant data for debugging since logs are sparse and important metrics were not captured. In contrast, other services in the system are well-instrumented and can be easily monitored, but the lack of consistency between services makes troubleshooting and root cause analysis more complex.

### 6. Recommended Solution

- Develop clear instrumentation guidelines: Establish company-wide standards for how services should be instrumented, including logging formats, metrics definitions, and tracing standards.
- Incorporate observability into the development process: Make observability an
  integral part of the service development lifecycle. Ensure that new services are
  instrumented before they are deployed.
- **Create templates or automation**: Provide templates, checklists, or automated tools to ensure that instrumentation is applied consistently and correctly across services.
- Regular audits: Conduct regular audits of new and existing services to ensure they
  meet observability standards and follow best practices.
- **Cross-team collaboration**: Encourage teams to share instrumentation strategies and best practices, ensuring that all services are tracked in a standardized way.

# 7. Consequences

## Positive:

- **Improved system visibility** by ensuring that all services are consistently instrumented with logging, metrics, and tracing.
- **Faster issue detection and resolution** as services are properly monitored from the start, making it easier to identify and fix problems.
- **Simplified debugging** since the same instrumentation practices will be used across services, enabling quicker identification of the root cause.

## Challenges:

- **Initial effort to establish guidelines** and bring teams onboard with consistent practices, which might require training and coordination across multiple teams.
- Overhead in ensuring compliance with the guidelines, especially in large or rapidly growing organizations.

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1. Antipattern Name: Using Too Many Tools

**2. Category:** General

**3. Context:** This antipattern occurs when an organization or team adopts an excessive number of tools to solve similar or overlapping problems without fully evaluating whether the tools are necessary or complementary. This often results from the desire to try every new technology or the lack of coordination between teams, leading to redundancy and inefficiency. While it's tempting to integrate a new tool for every issue, using too many tools can overwhelm teams, increase complexity, and create difficulties in maintaining and managing systems.

- **4. Problem:** When too many tools are used, it increases operational overhead, as managing a large number of tools adds administrative tasks such as configuration, monitoring, and maintenance, diverting teams from focusing on solving actual problems. Integration issues may arise when tools don't work well together or require complex integrations, resulting in fragmented observability and difficulty correlating data from different sources. The learning curve also becomes steeper, as teams must spend extra time learning and managing multiple tools, leading to inefficiencies and a lack of deep expertise in any single tool. Additionally, using overlapping tools can result in higher costs, wasting investment in both software licenses and resources.
- **5. Example:** A company uses separate tools for monitoring infrastructure, application performance, logging, alerting, and tracing. These tools don't integrate well, and each team is responsible for a different toolset. As a result, when a performance issue arises, the operations team has to check one tool for infrastructure metrics, another for application logs, and yet another for tracing data. This fragmented approach leads to longer troubleshooting times, increased overhead in managing each tool, and higher costs for subscriptions and maintenance.

### 6. Recommended Solution

- **Consolidate tools**: Evaluate the tools in use and consolidate them where possible. Use multi-purpose tools that cover several needs (e.g., a platform that combines logs, metrics, and traces in a unified interface).
- **Prioritize integration**: Ensure that the tools being used integrate well with each other, reducing friction and making it easier to correlate data across systems.
- **Standardize on key tools**: Establish a company-wide or team-wide standard for the tools to be used, ensuring that everyone uses a consistent set of tools for similar tasks.
- Evaluate new tools critically: Before adopting a new tool, carefully assess whether it
  provides unique value or overlaps with existing solutions, and consider long-term
  scalability and support.

# 7. Consequences

Positive:

- **Reduced complexity** by minimizing the number of tools in use, making it easier to manage systems and workflows.
- Lower costs as fewer subscriptions, licenses, and tool-specific resources are needed.
- **Increased efficiency** because teams can focus on mastering fewer tools and using them more effectively.
- **Better integration** between tools, enabling seamless workflows and better data correlation.

# Challenges:

- **Initial time investment** to evaluate tools, consolidate systems, and migrate to fewer tools, which may require some downtime or additional resources.
- **Resistance to change** from teams who are accustomed to specific tools or workflows, especially if they are entrenched in existing practices.

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- 1. Antipattern Name: Neglecting Regular Maintenance
- 2. Category: General
- **3. Context:** This antipattern occurs when teams or organizations fail to perform regular maintenance on their observability systems, such as updating tools, cleaning up old data, or reviewing configurations. It may happen when teams prioritize immediate issues over long-term system health, leading to gradual degradation. Neglecting maintenance can result from a lack of time, unclear responsibilities, or a reactive rather than proactive mindset toward system upkeep.
- **4. Problem:** When regular maintenance is neglected, it leads to decreased system reliability, as outdated software, configurations, or infrastructure can cause unexpected failures, performance degradation, or security vulnerabilities. Troubleshooting time increases as problems may go unnoticed or become harder to diagnose due to outdated tools and systems. Resources are wasted by retaining unnecessary data, unused tools, or configurations, leading to inefficient use of storage and computational resources. Over time, systems can become fragmented, with inconsistent data, configurations, or tool versions across teams. Neglecting maintenance also leaves systems vulnerable to security risks, potentially leading to data breaches or performance issues.
- **5. Example:** A team continues to use an outdated logging platform with limited support and lacks the time to update to a more modern solution. As new features are added to the system, the old logging tool becomes less efficient, leading to slow processing and inaccurate logs. When an incident occurs, the team struggles to extract meaningful data from the logs, as the tool has not been updated to support newer logging formats or integrate with other monitoring systems.

- Establish regular maintenance schedules: Set aside time for routine updates, tool upgrades, configuration reviews, and data cleanups to ensure that systems are functioning at their best.
- **Automate maintenance tasks**: Where possible, automate tasks like log rotation, data retention, and software updates to reduce manual overhead and ensure consistency.
- Create ownership of observability systems: Assign dedicated team members or roles responsible for the maintenance of the observability infrastructure to ensure accountability.
- **Conduct periodic audits**: Regularly audit observability tools and configurations to identify areas for improvement or modernization.
- **Plan for scalability**: Ensure that your observability system can evolve with the organization, including planning for future tool upgrades and capacity expansion.

### Positive:

- **Improved system stability** through regular updates and proactive problem detection, leading to fewer system failures and faster issue resolution.
- More efficient observability tools that provide accurate, actionable data for troubleshooting and optimization.
- **Lower operational risk** from security vulnerabilities or outdated systems, which are less likely to be exploited.
- **Better resource management** by eliminating redundant data or tools, ensuring that resources are utilized efficiently.

# Challenges:

- **Time investment** required to schedule and perform maintenance tasks, especially in fast-paced environments with multiple priorities.
- **Resistance to change** from teams used to working with outdated tools or systems, requiring effective change management strategies.

**1. Antipattern Name:** Observability for Production Only

2. Category: General

**3. Context:** This antipattern occurs when organizations focus their observability efforts exclusively on production environments, neglecting non-production environments such as staging, testing, or development. While it is essential to have comprehensive observability in production for detecting and resolving real-time issues, limiting observability to production only can result in missing critical problems before they reach production. This approach is often driven by resource constraints, a lack of awareness, or the mistaken belief that issues are less likely to occur in non-production environments.

- **4. Problem:** When observability is only applied to production, it increases the risk of issues reaching production, as problems that could have been caught early in development or staging remain undetected, escalating to production where they are harder and costlier to fix. Debugging becomes inefficient, as issues discovered in production lack prior insight from testing or staging environments, making troubleshooting more time-consuming and complex. Without consistent observability in non-production environments, teams may miss opportunities to optimize services before release, resulting in lower-quality launches. This also leads to frustration for developers and testers, who lack visibility into system performance and behavior in their environments.
- **5. Example:** A development team deploys a new feature in a staging environment without comprehensive logging or monitoring. When the feature reaches production, performance issues arise, but the team struggles to identify the root cause because they didn't have enough visibility in the staging environment. As a result, the team must investigate the issue in production, where the problem is harder to reproduce, and the impact is greater. Meanwhile, the rest of the system works fine, but the lack of observability in non-production environments adds significant delay to the resolution.

## 6. Recommended Solution

- Extend observability to all environments: Implement consistent logging, monitoring, and tracing in non-production environments to ensure that issues can be detected and resolved before reaching production.
- Use the same observability tools across environments: Standardize observability practices and tools across development, staging, and production to maintain visibility and ensure that teams can monitor and diagnose issues consistently.
- **Automate testing and monitoring**: Incorporate automated tests and monitoring as part of the CI/CD pipeline to catch issues early in non-production environments.
- Integrate observability into the development lifecycle: Make observability a core part of the development process, ensuring that teams can access performance data and logs during all stages of development, not just in production.
- Monitor pre-production environments at scale: Ensure that staging or testing environments are adequately sized and instrumented to replicate production loads, helping to catch issues that may not be apparent under lower test loads.

# 7. Consequences

### Positive:

- **Earlier issue detection** in non-production environments, preventing problems from escalating to production and improving release quality.
- **Faster debugging** in both pre-production and production environments, with more context available from logs, metrics, and traces.
- **Higher-quality releases** because teams have the visibility they need to optimize services and catch issues before they affect end-users.

• **More efficient testing** as non-production environments are more thoroughly instrumented and can simulate production-like conditions.

# Challenges:

- **Increased resource consumption** from adding monitoring and logging across all environments, which could affect system performance or increase storage needs.
- More complex setup and configuration as the same tools and practices must be applied to different environments, requiring careful coordination.

1. Antipattern Name: Throwing Tools at a Problem

2. Category: General

- 3. Context: This antipattern occurs when organizations or teams try to solve complex problems or improve observability by adopting numerous tools without a clear understanding of the underlying issue or a comprehensive strategy. This approach often stems from the belief that acquiring the latest tools will automatically resolve issues, without properly analyzing the root cause or ensuring the tools are well integrated into the existing systems. This can lead to tool sprawl, where multiple redundant or incompatible tools are used, making the problem worse instead of improving the situation.
- **4. Problem:** When teams rely on "throwing tools" at problems, it leads to tool sprawl, with too many tools being implemented, often overlapping in functionality, creating unnecessary complexity and maintenance overhead. Poor integration between tools can cause fragmentation in the observability ecosystem, making it difficult to correlate data or gain insights across the system. Instead of addressing the root cause of issues, organizations may invest in a patchwork of tools, leading to wasted time and resources. Managing multiple tools with varying configurations, data formats, and support requirements increases operational overhead, confusion, and inefficiency. This also results in tool fatigue, where teams become overwhelmed by the number of tools they must manage and learn, reducing the overall effectiveness of observability practices.
- **5. Example:** A team facing inconsistent performance in a system decides to resolve the issue by adding new monitoring, tracing, and alerting tools. They implement a separate tool for infrastructure monitoring, application performance monitoring (APM), and log aggregation, all of which overlap in some aspects. These tools don't integrate well, leading to duplicated data, fragmented views of the system, and difficulty in correlating performance issues across the different tools. As a result, the performance problem remains unsolved, and the team is left managing multiple systems without clear insights.

- Diagnose the root cause: Before adopting any new tools, take the time to identify the
  real issue and evaluate whether it can be solved with the tools you already have or by
  making minor adjustments to existing processes.
- Consolidate tools where possible: Instead of adding new tools for every issue, look for tools that provide comprehensive solutions that integrate well with existing systems, reducing complexity and overlap.
- **Focus on strategy, not just tools**: Develop an observability strategy that aligns with organizational goals and addresses the core needs of the team, whether that's improving alerting, better tracing, or centralized logging.
- **Evaluate tools carefully**: When introducing new tools, ensure they align with your overall observability strategy and integrate seamlessly with other tools already in use.
- **Train and empower teams**: Rather than simply adding more tools, invest in training and empowering your teams to use the tools effectively, ensuring they understand the value each tool brings to the table.

## Positive:

- **Simplified observability**: By reducing tool sprawl and consolidating functionality, teams can focus on solving issues rather than managing numerous, redundant systems.
- **Better tool integration**: Fewer tools with better integration lead to more cohesive observability and faster root cause analysis.
- **Improved efficiency**: A clearer focus on solving problems and using the right tools for the job reduces wasteful spending and operational overhead.
- Stronger alignment with business needs: A well-thought-out strategy ensures tools
  are used to meet organizational goals, making observability more effective and
  impactful.

# Challenges:

- **Initial time investment** to analyze the root causes of issues and create a strategy may be seen as time-consuming, especially in fast-paced environments.
- Resistance to changing tools from teams who are accustomed to their existing tools or workflows, requiring careful change management.

1. Antipattern Name: Lack of System-Wide Perspective

**2. Category:** General

**3. Context:** This antipattern occurs when teams or organizations focus on monitoring or observing isolated components or services without considering the entire system as a whole. While individual components are essential to monitor, neglecting the interactions and dependencies between them can result in blind spots in system behavior and performance. This

often happens due to siloed teams, limited tooling, or an overly narrow focus on specific metrics without understanding their role within the larger system.

- **4. Problem:** When there is a lack of system-wide perspective, it leads to missed correlations, as teams fail to recognize critical relationships between services, dependencies, or events, resulting in an incomplete understanding of issues. Troubleshooting becomes fragmented and inefficient, with teams focusing on individual parts without realizing they are part of a larger problem. Root cause analysis is slow, as teams have limited visibility into the broader system, leading to longer resolution times. Resources may be inefficiently allocated, addressing symptoms rather than the root cause or optimizing overall performance. Additionally, inconsistent monitoring coverage means some parts of the system may be over-monitored while others are under-monitored, resulting in incomplete or misleading data about the system's health.
- **5. Example:** A team focuses heavily on monitoring the performance of a specific microservice, tracking metrics like response times and error rates. However, they don't track how this microservice interacts with other services in the ecosystem. When an issue arises with a downstream service that the team doesn't monitor closely, the team fails to see the correlation between the microservice's errors and the issues in the downstream service. This leads to longer troubleshooting times and a lack of clarity around the root cause.

## 6. Recommended Solution

- Implement end-to-end monitoring: Ensure that all components of the system, including microservices, databases, and external dependencies, are monitored and observed. Integrate tracing and correlation across all services to capture the full system flow.
- **Focus on service dependencies**: Be sure to map and monitor the dependencies between services, applications, and infrastructure, allowing teams to trace how one failure can propagate through the system.
- Establish system-wide dashboards: Create comprehensive dashboards that bring together data from different components to give a complete view of system health, performance, and critical metrics.
- Integrate observability practices across teams: Encourage cross-functional teams (e.g., developers, SREs, ops) to collaborate on defining what is monitored, ensuring that every team has visibility into the complete system.
- **Ensure tool interoperability**: Use tools that provide a holistic view and can integrate data from various sources, ensuring visibility across the entire stack and ecosystem.

# 7. Consequences

### Positive:

• **Faster root cause identification**: With a system-wide perspective, teams can quickly identify how different components interact and pinpoint the true source of issues.

- **Improved collaboration**: Cross-functional teams are better able to communicate and resolve issues because they have a shared understanding of the entire system.
- **Optimized performance**: By considering the system as a whole, teams can better allocate resources, prioritize improvements, and ensure the system is running efficiently across all components.
- Enhanced troubleshooting: Teams can solve complex issues more effectively by seeing the full picture and not getting stuck in narrow, component-focused troubleshooting.

# Challenges:

- Increased complexity: Implementing system-wide observability can require more sophisticated tools, integrations, and configurations, which can be challenging in large or complex environments.
- Higher overhead: Maintaining a broad view of the system may introduce additional monitoring and data processing overhead, which needs to be managed carefully to avoid performance degradation.

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- 1. Antipattern Name: Lack of Contextual Information
- **2. Category:** General
- **3. Context:** This antipattern occurs when the data captured by observability tools, such as logs, metrics, or traces, lacks sufficient context to help teams understand the full scope of the situation. Without relevant contextual information, it becomes difficult to interpret the data correctly, leading to confusion, slower troubleshooting, and missed opportunities for proactive problem resolution. This can arise when logs or metrics are not detailed enough, or when important relationships and metadata (such as request IDs, user sessions, or service dependencies) are not captured.
- **4. Problem:** When there is a lack of contextual information, it becomes difficult for teams to understand the situation, as the data lacks the necessary context to interpret events properly. This results in inefficient troubleshooting, as diagnosing root causes becomes time-consuming and error-prone when logs or metrics don't include critical context like session data or trace IDs. Teams also miss opportunities for proactive action, as they cannot detect early signs of problems or anticipate potential system failures, leading to reactive approaches instead of proactive monitoring. Investigations become fragmented, as teams must cross-reference data from multiple sources or manually piece together context, which delays resolution and increases the risk of missing critical information. Additionally, inconsistent alerts that lack context can trigger false positives or miss significant issues, leading to alert fatigue or delayed response times.

# 5. Example

A service begins to show elevated error rates, but the logs do not include the request ID or the

service dependencies that led to the error. The team can see the error spikes in the logs but has no way of knowing what specific users or transactions were affected or whether this issue is related to other parts of the system. As a result, the team spends significant time trying to correlate the error with other system data and trying to reproduce the issue, which slows down troubleshooting and prolongs incident resolution.

## 6. Recommended Solution

- Enrich logs and metrics with context: Ensure that logs, metrics, and traces include relevant metadata such as request IDs, user sessions, error messages, service versions, and any relevant external dependencies. This will help correlate events across services.
- **Implement distributed tracing**: Use distributed tracing to capture end-to-end context, including service interactions, latencies, and error propagation, providing a clearer picture of how requests flow through the system.
- Capture high-level context: In addition to technical details, capture business-related context, such as transaction types or user IDs, that can help teams understand the impact of incidents on the business.
- Use structured logging: Adopt structured logging formats (like JSON) that can include more contextual information in a machine-readable way, making it easier to query, search, and analyze logs.
- Ensure consistent tagging and metadata: Use consistent tagging conventions to capture context across services, ensuring that logs and metrics can be correlated with one another, even when they are generated in different parts of the system.
- **Define relevant context for alerts**: Design alerting rules that not only notify teams of issues but also include sufficient contextual information, such as the affected service, impacted user segments, and the potential root cause.

# 7. Consequences

## Positive:

- Faster incident resolution: With better contextual information, teams can quickly understand the scope and root cause of issues, speeding up troubleshooting and resolution.
- **Proactive monitoring**: Teams can identify early warning signs of problems and take proactive action before issues escalate into full-blown incidents.
- **Improved system insights**: Richer data provides a clearer understanding of system performance and user behavior, leading to better decision-making.
- **Reduced alert fatigue**: Alerts with more context are easier to understand, leading to faster responses and fewer false positives.

# Challenges:

 Increased logging overhead: Collecting and enriching logs with more context can increase the volume of data generated, which may result in higher storage and processing costs.

- Complexity in implementation: Ensuring that all components are consistently capturing
  the necessary context and metadata can require significant changes to existing systems
  and workflows.
- **Data management**: The larger volume of contextual data may require better data management practices to prevent overwhelm and ensure data quality.

1. Antipattern Name: Bad Sampling Intervals

**2. Category:** General

- **3. Context:** This antipattern occurs when the sampling intervals used to collect metrics or monitoring data are poorly chosen, leading to either too much or too little data being captured. Sampling intervals refer to the frequency at which data points are collected for monitoring and metrics purposes. If the interval is too short, the system may generate too much data, leading to excessive resource consumption, storage costs, and potentially overwhelming the monitoring infrastructure. On the other hand, if the sampling interval is too long, the collected data may fail to capture important fluctuations or patterns in system performance, missing critical issues that occur in between sample periods. This problem typically arises when teams set sampling intervals based on convenience or technical limitations rather than on the specific needs of the system or the metrics being tracked.
- **4. Problem:** When sampling intervals are poorly chosen, teams may face excessive data collection, leading to large volumes of data that overload storage systems and processing capabilities. On the other hand, long sampling intervals may result in missed short-term spikes, dips, or anomalies in performance, causing inaccurate metrics. Additionally, unnecessary data collection consumes computational resources, bandwidth, and storage space. Inconsistent sampling intervals across different parts of the system can produce fragmented or misleading data, complicating monitoring. This inefficiency can delay or even miss incident detection, hindering the team's ability to respond to performance issues in a timely manner.
- **5. Example:** A team sets a sampling interval of 10 seconds for their application's CPU usage metrics, but due to the high traffic load on the system, the volume of data becomes unmanageable. This leads to storage issues and increased processing time to analyze the data. On the other hand, the team sets a sampling interval of 5 minutes for error rates, missing key fluctuations that might indicate intermittent problems. This imbalance results in missing short-lived but critical performance issues while also overwhelming the system with too much unnecessary data.

# 6. Recommended Solution

• Optimize sampling intervals: Set sampling intervals based on the criticality of the data being collected. For example, use shorter sampling intervals for high-frequency metrics like response times or request counts, and longer intervals for lower-frequency metrics such as resource utilization or overall system health.

- Use dynamic sampling: Implement dynamic sampling strategies that adjust the interval based on system load or the presence of anomalies. For instance, reduce the sampling rate during normal operations and increase it during periods of high traffic or when anomalies are detected.
- Prioritize key metrics: Focus on monitoring the most crucial metrics that directly impact system performance and user experience. Ensure that sampling intervals are set appropriately for these key metrics to capture the most relevant data without overwhelming the system.
- Consider the nature of the system: Choose sampling intervals based on the system's behavior. Real-time systems or applications with unpredictable spikes in traffic may require more frequent sampling, while batch processing systems may tolerate longer intervals without losing important data.
- **Implement a feedback loop**: Regularly assess the effectiveness of your sampling intervals by analyzing the accuracy and relevance of the data collected. Adjust intervals as needed to ensure they are providing actionable insights without wasting resources.

## Positive:

- More efficient data collection: By choosing the right sampling interval, the system can collect sufficient data without overwhelming resources, leading to more efficient use of storage and processing power.
- **Improved incident detection**: Properly chosen sampling intervals help teams quickly identify performance issues and anomalies, improving response times and reducing the impact of incidents.
- **Cost savings**: Reducing unnecessary data collection helps minimize storage and processing costs, making monitoring systems more cost-effective.
- More accurate insights: Optimal sampling intervals ensure that the collected data reflects the true behavior of the system, providing more accurate insights for analysis and decision-making.

# Challenges:

- **Tuning intervals**: Determining the right sampling interval for different metrics can require experimentation and ongoing adjustment as the system evolves or as the monitoring requirements change.
- System limitations: Some systems may have technical limitations that make it difficult
  to adjust sampling intervals dynamically or may struggle with handling large volumes of
  data.
- Balancing granularity: Finding the balance between granular data for detailed insights
  and high-level metrics for system health can be difficult. Too much detail may lead to
  noise, while too little can result in missed opportunities to detect issues.

1. Antipattern Name: Failure to Monitor Error Logs

2. Category: Logs

- **3. Context:** This antipattern arises when teams or organizations neglect to monitor error logs or fail to set up proper alerts for errors in the system. Error logs contain crucial information about system failures, exceptions, and unexpected behaviors, but if they aren't actively monitored or analyzed, potential issues can go unnoticed, leading to downtime, user impact, and difficulty in resolving incidents. This can occur when teams assume that errors will be automatically detected, or when there's an over-reliance on other monitoring signals (e.g., metrics or traces) that don't capture all failure modes.
- **4. Problem:** When error logs are not actively monitored, critical incidents can be missed, resulting in undetected failures that lead to system outages, degraded performance, or poor user experiences. Without proper monitoring, issues that should be flagged by error logs may go unnoticed for extended periods, increasing recovery times. Error logs provide vital clues for root cause analysis, and without monitoring, pinpointing the cause of problems becomes more difficult, prolonging troubleshooting and resolution. This also increases the mean time to recovery (MTTR), as errors that go unnoticed or are identified too late lead to longer recovery times, affecting system reliability. Finally, teams can become overwhelmed when errors accumulate in logs, as they may be faced with large volumes of data to investigate later, leading to delays and poor performance in addressing incidents.
- **5. Example:** A web application is frequently experiencing 500 errors in production. The error logs contain stack traces and other relevant information, but there are no alerts set up to notify the team about these critical errors. As a result, the team only notices the issue when user complaints increase and the system becomes unresponsive. By this point, the application has been suffering from intermittent failures for hours, leading to a prolonged recovery process and significant user frustration.

- **Set up real-time monitoring for error logs**: Ensure that all critical error logs are actively monitored and that alerts are configured to notify the team immediately when error rates exceed predefined thresholds.
- **Create structured log entries**: Use structured logging formats that make it easier to filter, search, and analyze error logs, ensuring that errors are categorized and actionable.
- **Integrate logs with alerting systems**: Link error logs directly to your alerting systems so that high-priority errors automatically trigger notifications to the right teams.
- Regularly review and prioritize error logs: Set up processes to regularly review the
  most common or critical error logs, so teams can identify patterns or recurring issues
  and address them proactively.
- **Automate remediation where possible**: For common or predictable errors, implement automated responses (e.g., restarting services or retrying operations) to mitigate the impact until the root cause can be fixed.

• **Monitor error rates and trends**: Use tools that can visualize error trends over time, enabling teams to detect rising issues early and prevent widespread failures.

# 7. Consequences

## Positive:

- **Faster incident detection**: With proper monitoring of error logs, issues can be detected earlier, allowing teams to respond quickly and reduce downtime.
- **Improved system reliability**: Proactively monitoring error logs helps teams resolve issues before they escalate into major incidents, improving overall system stability and user satisfaction.
- **Better root cause analysis**: Error logs provide crucial information for diagnosing the source of failures, enabling quicker and more accurate troubleshooting.
- **Reduced recovery time**: With error logs actively monitored and promptly addressed, recovery times can be minimized, leading to a more resilient system.

# Challenges:

- **Data overload**: Error logs can accumulate rapidly, especially in complex systems, which may result in alert fatigue if not properly managed and filtered.
- **Initial setup time**: Setting up effective monitoring and alerting systems for error logs can take time, particularly in large or legacy systems where error log formats and storage might need to be standardized.
- Maintenance overhead: Ensuring that error log monitoring continues to be effective as the system evolves can require periodic adjustments to thresholds, alerting rules, and log management processes.

- 1. Antipattern Name: Using API Access Logs for Troubleshooting
- 2. Category: Logs
- **3. Context:** This antipattern occurs when teams rely on API access logs (such as HTTP request logs) to troubleshoot issues without incorporating deeper diagnostic tools like application logs, error logs, or distributed tracing. While API access logs are useful for tracking request traffic and basic information (such as response status codes, timestamps, and request paths), they often lack the necessary context and detail to diagnose complex issues like application bugs, service failures, or performance bottlenecks. Relying solely on these logs can lead to incomplete or inaccurate troubleshooting.
- **4. Problem:** When teams use only API access logs for troubleshooting, it results in limited diagnostic value, as these logs typically provide only surface-level data (e.g., request paths, status codes) and miss deeper insights about internal application logic, errors, or performance bottlenecks. Identifying root causes becomes difficult, as API access logs may indicate issues like high error rates or slow responses but lack the detailed stack traces, exceptions, or context

needed to understand why the issue occurred. Additionally, service dependencies are often overlooked, as API access logs do not capture how services interact with each other. Without a holistic view, such as distributed tracing, teams may miss issues from service dependencies, network latencies, or external integrations. This can lead to incomplete investigations, where teams miss critical context or fail to see whether other services or resources were involved. Finally, relying on API access logs alone can give teams a false sense of confidence, leading them to overlook important contributing factors to the problem.

**5. Example:** A web service has been receiving high 5xx error rates, which are visible in the API access logs, but the logs do not provide any further details about the root cause of the errors. The team uses these logs to conclude that the issue is likely related to the API's backend processing, but they don't examine the application logs or set up distributed tracing. It turns out that the issue was related to a downstream service failing, but the API access logs alone didn't provide enough context to identify this. The team spent significant time troubleshooting the wrong part of the system.

## 6. Recommended Solution

- Combine API access logs with application and error logs: Use a combination of API access logs, application logs, and error logs for troubleshooting. Application logs provide deeper insights into the internal behavior of services, such as exceptions, error messages, and processing logic.
- **Implement distributed tracing**: Utilize distributed tracing to get a full end-to-end view of how requests travel through the system, including interactions with other services, databases, and external APIs. This will provide a more comprehensive context for diagnosing issues.
- Correlate logs across services: Ensure that logs from different services (e.g., backend, database, external APIs) can be correlated using common identifiers like request IDs, session IDs, or trace IDs. This allows teams to follow requests across the system and understand their full lifecycle.
- Monitor performance metrics alongside logs: In addition to logs, track key
  performance metrics such as latency, throughput, and error rates. These metrics, when
  combined with logs, can help identify trends or anomalies that point to specific areas of
  concern.
- **Set up proactive alerting**: Implement alerts that can notify teams not only based on API access logs but also on abnormal behavior seen in application logs, error rates, and distributed traces. This helps catch problems earlier.

# 7. Consequences

## Positive:

• Faster root cause identification: By using a variety of logs and tracing, teams can quickly narrow down the root cause of issues, reducing the time spent troubleshooting.

- Comprehensive troubleshooting: Access to detailed application logs and tracing provides better context and insight, helping teams address more complex issues that API logs alone can't identify.
- **Improved system reliability**: Proactive monitoring of logs and traces can help detect issues earlier, reducing downtime and improving overall system reliability.
- **Better collaboration**: With more detailed and integrated observability, teams can collaborate more effectively by having all the relevant data at their disposal.

# Challenges:

- Increased complexity: Implementing a more comprehensive logging and tracing strategy may require additional setup and configuration, as well as more advanced tooling and integration between different services.
- **Higher resource overhead**: Storing and processing a greater volume of logs, traces, and metrics can lead to higher infrastructure costs and require more maintenance.
- Potential noise: Collecting and correlating data from multiple sources can increase the
  volume of information, which might overwhelm teams if not well managed, potentially
  leading to alert fatigue or data overload.

1. Antipattern Name: Improper Log Level Usage

2. Category: Logs

- **3. Context:** This antipattern arises when logs are generated with inappropriate log levels, such as using INFO for critical errors or DEBUG for general application flow. Log levels, like ERROR, WARN, INFO, and DEBUG, are meant to signal the severity or importance of events. However, teams often misuse log levels by setting them too low (e.g., logging critical errors as INFO) or too high (e.g., logging routine information as ERROR). This misclassification leads to confusion, excessive log volume, and makes it difficult to identify the severity of issues quickly, especially when reviewing logs during incidents or troubleshooting.
- **4. Problem:** When log levels are used improperly, critical errors can become buried in noise, as low log levels (e.g., INFO or DEBUG) may be used for important issues like system failures or application crashes, leading teams to overlook or fail to prioritize these events. Excessive logging of non-essential events at high verbosity levels can cause alert fatigue, overwhelming teams with notifications for irrelevant events and making it harder to spot meaningful issues. Improper log level usage also creates inconsistent logging practices, making it difficult for teams to interpret logs clearly, thus increasing the time spent understanding logs during incidents. Additionally, missed opportunities for proactive monitoring occur, as improper log levels may prevent important events from triggering alerts, hindering early identification and resolution of problems. This leads to inefficient troubleshooting, where teams must sift through large volumes of unnecessary log data to identify the root cause of issues.

**5. Example:** A service experiences an unexpected crash due to a database connection timeout. The error is logged as INFO, and a secondary warning about the failure to reconnect to the database is logged as DEBUG. When the team reviews the logs, the INFO log is buried among regular operational logs, and the DEBUG log about the database issue is overlooked because it's not properly flagged as an error. As a result, the team only realizes the severity of the problem much later, when the service becomes unstable.

## 6. Recommended Solution

- **Use log levels appropriately**: Ensure that logs are generated at the correct level according to the severity and significance of the event:
  - ERROR: For critical issues that cause failures, crashes, or other severe problems that need immediate attention.
  - WARN: For less critical issues that do not cause failures but may require investigation (e.g., deprecation warnings, resource exhaustion).
  - INFO: For general operational logs that provide information on the system's state, such as successful transactions or state changes.
  - DEBUG: For detailed internal information, such as function calls or request/response data, typically used for development and debugging purposes.
- Implement log level enforcement: Set up policies and guidelines that standardize the log levels used across different teams or services, ensuring consistency in how logs are classified.
- **Use dynamic log level controls**: Allow log levels to be dynamically adjusted at runtime (e.g., increasing verbosity for debugging specific issues) so that teams can troubleshoot without permanently changing code or configurations.
- Review and refine logging practices: Regularly audit logs to ensure that the right log levels are used, adjusting as needed to make logs more useful and actionable.
- **Establish clear logging guidelines**: Provide clear documentation for the team about when to use each log level, ensuring that the entire organization follows the same best practices.

# 7. Consequences

### Positive:

- **Improved incident detection**: Correct log levels help ensure that critical issues are highlighted, making it easier to detect and address incidents before they escalate.
- **Efficient troubleshooting**: Proper log level usage helps reduce the noise in logs, allowing teams to focus on relevant data and troubleshoot issues faster.
- **Better monitoring and alerting**: Logs are better structured for integration with alerting systems, which can trigger appropriate notifications based on the severity of the event.
- Clearer operational insights: By using the correct log levels, the team gets a clearer
  picture of both normal operations and exceptional events, aiding in long-term system
  improvements.

## Challenges:

- **Initial setup effort**: Standardizing log levels across the organization may require effort to audit existing logs and refactor them where needed, which could involve code changes and coordination.
- **Changing team habits**: Teams that are accustomed to inconsistent or inappropriate log level usage may face resistance when adopting more structured practices.
- Increased complexity: More granular control over log levels (e.g., dynamic log level adjustments) can introduce complexity in log management, especially in distributed systems.

1. Antipattern Name: Lack of Structured Logging

2. Category: Logs

- 3. Context: This antipattern occurs when logs are written in an unstructured or free-text format, which makes it difficult to search, analyze, and correlate log entries across systems. Unstructured logs often consist of plain text messages with minimal context or metadata, making them harder to process programmatically. This happens when teams fail to implement structured logging, where logs are output in a consistent, machine-readable format (e.g., JSON, key-value pairs). The lack of structure increases the time and effort required to extract meaningful insights from logs, especially in complex systems or during incident response.
- **4. Problem:** When logs are not structured, they become difficult to search and analyze efficiently, as unstructured logs often consist of free-text entries that make it challenging to find specific events or patterns. This requires manual parsing or using regular expressions, which is time-consuming and error-prone. Without structured logs, automation tools (e.g., alerting systems, log aggregation platforms) cannot easily parse or analyze log data, missing opportunities for proactive monitoring or automated remediation. The lack of structure also leads to inconsistent logging practices, where each log message follows a different format or contains inconsistent information, making it hard to correlate logs across services or systems. This increases operational overhead, as teams must spend more time manually sifting through logs to find relevant information, which creates inefficiencies during both routine operations and incident investigations. Finally, the absence of structure impairs incident response, as it becomes more difficult to quickly identify key pieces of information, delaying the investigation and recovery process.
- **5. Example:** A microservices-based application logs error messages in plain text, with entries like "Service X failed" or "Database error occurred". During an outage, the team tries to identify the root cause, but the logs provide minimal context, such as timestamps, request IDs, or service names. The team struggles to correlate logs from different services and is unable to identify the exact sequence of events leading up to the issue, prolonging the investigation and recovery.

- Implement structured logging formats: Adopt a consistent, machine-readable logging format (e.g., JSON or key-value pairs) that includes relevant metadata such as timestamps, log levels, service names, request IDs, error codes, and contextual information.
- **Include context in logs**: Ensure that logs contain sufficient context to make them actionable, such as user/session IDs, request paths, or transaction IDs that allow tracing a request through multiple systems.
- Use log aggregation and analysis tools: Implement centralized log aggregation
  platforms (e.g., ELK Stack, Splunk) that can efficiently handle and query structured logs.
  These tools allow you to quickly identify trends, anomalies, or errors across your system.
- Standardize logging practices: Define clear logging guidelines across your team or organization to ensure consistency in how logs are generated, ensuring that all logs include the same essential metadata.
- **Use log enrichment**: Enhance logs with additional metadata when possible, such as information about the environment, system health, or performance metrics, to provide deeper insights into system behavior.

## Positive:

- **Easier log analysis**: Structured logs are easier to search, filter, and analyze, enabling teams to identify patterns, issues, and trends quickly.
- **Improved incident response**: Structured logs provide rich, consistent information, allowing teams to quickly correlate logs across services, trace requests, and identify root causes during incidents.
- **Better automation**: Automation tools can process structured logs more effectively, enabling features like automated alerting, anomaly detection, and even remediation.
- Increased operational efficiency: With structured logs, teams spend less time
  manually parsing logs and more time addressing issues, improving overall operational
  efficiency.
- **Enhanced monitoring and reporting**: Structured logs integrate better with monitoring and reporting tools, helping to generate actionable insights and metrics that drive system improvements.

## Challenges:

- Initial effort: Implementing structured logging across all systems requires significant
  upfront effort to refactor existing logs, standardize formats, and ensure proper tooling is
  in place.
- Increased storage requirements: Structured logs may increase the volume of log data due to additional metadata and more detailed information, leading to higher storage and processing costs.
- Learning curve: Teams may need training or documentation to understand the new structured logging approach and tools, especially if they're used to working with unstructured logs.

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1. Antipattern Name: Duplicate Logs

2. Category: Logs

3. Context: This antipattern arises when the same log message is generated multiple times across different parts of a system, often due to misconfiguration or incorrect logging practices. Duplicate logs can occur in various scenarios, such as when the same event is logged by multiple services, components, or layers within an application. It can also happen when logs are produced at different log levels or by redundant logging mechanisms. This redundancy causes unnecessary log volume, which leads to increased storage and processing costs, as well as making log analysis more difficult by adding noise to the data.

- **4. Problem:** When duplicate logs are generated, it leads to increased log volume, as unnecessary duplicate entries consume more storage space and require additional resources to process, store, and analyze. The presence of duplicate logs causes log fatigue, as teams become overwhelmed by the sheer volume of log entries, making it harder to identify meaningful information and increasing the likelihood that significant events get overlooked. Duplicate logs can also distort analysis and metrics, leading to inaccurate reports, trends, or failure to identify genuine issues because the same log message is counted multiple times. During incident investigation, duplicate logs slow down troubleshooting and root cause analysis, as teams must sift through multiple identical entries instead of focusing on unique, actionable logs. Finally, generating and storing duplicate logs wastes valuable system resources, increasing costs for log storage and processing, especially in large, distributed environments.
- **5. Example:** A web service is configured to log errors when a user attempts to perform an action that fails due to insufficient permissions. However, the error is logged both by the API layer (which receives the request) and by the authorization layer (which handles access control). As a result, each failed action generates two identical log entries with the same error message and context, causing unnecessary duplication of log data. When the team reviews the logs, they are forced to ignore the redundant entries and focus on the unique logs, wasting valuable time during incident response.

- Consolidate log generation: Ensure that logs are generated by only one component or service per event. For instance, either the API layer or the authorization layer should log the error, but not both. Designate which layer is responsible for logging each type of event to prevent redundancy.
- Centralize logging configuration: Use centralized logging frameworks or libraries to standardize log generation across the system, ensuring that the same event is not logged in multiple places. Tools like logging aggregators or middleware can be used to enforce consistent logging practices.

- Implement log deduplication: Use log aggregation platforms (e.g., ELK Stack, Splunk) that support automatic deduplication or filtering of duplicate log entries based on common attributes like timestamps, request IDs, or error codes.
- Review and optimize logging policies: Regularly audit logging configurations and patterns to identify and remove redundant logging, and optimize log levels to ensure that only relevant information is logged.
- **Use log aggregation tools to centralize logs**: Consolidate logs from multiple sources into a single location to reduce duplication at the collection point and ensure logs are processed, stored, and analyzed in one place.

#### Positive:

- **Reduced log volume**: Eliminating duplicate logs reduces unnecessary log data, lowering storage requirements and streamlining log processing and analysis.
- **Improved log clarity**: With fewer duplicate entries, teams can focus on meaningful logs, leading to faster and more accurate incident response.
- Lower resource consumption: Less log data means reduced costs for storing and processing logs, improving overall system efficiency.
- Enhanced incident response: Teams can spend less time filtering out redundant logs and more time investigating and resolving issues, leading to faster recovery and reduced downtime.
- **Better insights**: Accurate log data allows for more reliable metrics and trends, enabling more effective monitoring, reporting, and system improvement.

# Challenges:

- **Initial effort for cleanup**: Identifying and removing duplicate logs across a large system can require significant time and effort, especially if the duplication is widespread.
- Revisiting logging configurations: The solution may require changes to existing logging configurations or restructuring of the logging architecture, which could involve refactoring code and re-testing services.
- Coordination between teams: In a distributed environment with multiple teams
  responsible for different components, coordinating to avoid duplication can be
  challenging and require clear guidelines and processes.

1. Antipattern Name: Local Logging Only

2. Category: Logs

**3. Context:** This antipattern occurs when logs are generated and stored only locally within individual services, rather than being aggregated or centralized in a single location for the entire system. Services may log events to local disk files or local logging systems that aren't accessible by other parts of the infrastructure. This practice is often a result of poor design or

lack of centralized logging policies, where teams assume local logs are sufficient for troubleshooting and monitoring. However, it causes significant challenges in large-scale systems, where visibility and traceability across services are essential.

- **4. Problem:** When logs are stored locally only, visibility is limited because logs are isolated to individual machines or services, making it difficult to get a holistic view of the system. This lack of centralization hinders the ability to correlate events across multiple services or identify system-wide issues. During incidents, troubleshooting becomes difficult as teams have to manually access logs from different locations, which is both time-consuming and error-prone. If logs are stored locally, they also lack redundancy and can be lost in the event of a system failure, potentially leaving crucial logs missing and impeding diagnosis. Moreover, without centralized storage, log data cannot be easily ingested into monitoring or alerting systems, meaning critical events, anomalies, or trends may go unnoticed. Finally, in regulated environments, storing logs locally creates compliance and auditing challenges, as centralized logging is often required for data retention, security, and access control.
- **5. Example:** A microservices application logs all events locally on each service's machine using local text files. When a user-facing bug occurs, the team cannot easily correlate the logs from multiple services (e.g., the front-end service, authentication service, and database) to pinpoint the cause. The developers need to manually log in to each service's machine to retrieve the logs, resulting in significant delays and frustration. Additionally, some logs are lost because the local disk fills up, causing the service to overwrite older logs.

# 6. Recommended Solution

- Centralize log collection: Use centralized logging systems such as ELK Stack, Splunk, or a cloud-based logging service (e.g., AWS CloudWatch, Google Cloud Logging) to aggregate logs from all services into a single repository. This allows for easier analysis, searching, and correlation of logs across the entire system.
- **Implement log forwarders**: Configure services to forward logs to centralized systems rather than storing them locally. Tools like Fluentd, Logstash, or syslog can be used to send logs to central log aggregators.
- Use a distributed logging architecture: In microservices or distributed systems, ensure that logs are collected across all services and can be correlated using unique identifiers, such as request IDs or transaction IDs, to trace events as they flow through the system.
- Set up log retention and backup: Implement log retention policies that ensure logs are stored long enough to be useful for troubleshooting, analysis, and compliance. Ensure logs are backed up and redundant to prevent data loss in the event of a failure.
- Use cloud-native logging tools: If you're using cloud platforms, take advantage of built-in centralized logging solutions that integrate with your infrastructure and simplify log management.

# 7. Consequences

Positive:

- Improved visibility: Centralizing logs enables teams to gain a complete view of the system, making it easier to identify issues and understand how different services are interacting.
- **Faster troubleshooting**: With all logs in one place, teams can quickly search and correlate events across services, which significantly reduces the time needed to resolve incidents.
- Better monitoring and alerting: Centralized logs can be ingested into monitoring tools and set up with automated alerting systems to proactively detect issues before they escalate.
- Compliance and auditing: Centralized logging supports compliance with regulations that require secure, long-term log storage and access control, as well as easier audit trails.

# Challenges:

- **Initial setup**: Setting up centralized logging infrastructure can require significant configuration and effort, especially in large systems with many services.
- Cost: Centralized logging systems, especially cloud-based services, may introduce additional costs for log storage, processing, and retention, particularly if logs are generated at a high volume.
- Log aggregation performance: In high-volume systems, aggregating and processing large amounts of log data in real-time can introduce latency or performance bottlenecks if not carefully managed.
- **Cultural shift**: Moving from local to centralized logging may require a shift in team practices and tools, and teams may need training on how to manage and interpret centralized logs.

1. Antipattern Name: Excessive Logs

2. Category: Logs

- **3. Context:** This antipattern occurs when systems log too much information, often at too granular a level. While logging is essential for understanding system behavior, excessive logs can flood the system with unnecessary data that obscures valuable insights. This can happen when teams log every minor action, detail, or event, often due to a lack of log level management or poor logging practices. It can also occur when logs are configured to capture every piece of information during development or debugging stages and never properly adjusted for production environments.
- **4. Problem:** When excessive logs are generated, storage costs increase due to the need to store large amounts of log data that may not provide significant value. This can also degrade performance, especially if logs are written synchronously or contain resource-intensive data. Excessive logging creates noise in the log data, making it harder to extract meaningful insights, which delays incident response and troubleshooting as important logs can be lost in the flood of

trivial details. Additionally, the large volume of logs makes it difficult to analyze and identify trends, anomalies, or issues, as teams must sift through excessive data to find relevant information. Log management systems can become strained by the overwhelming amount of data, leading to slower query performance, missed alerts, and difficulty in extracting actionable insights. Finally, operational inefficiency arises as teams spend time filtering out unnecessary logs or managing large volumes of data instead of focusing on actual problems or system improvements.

**5. Example:** A web application logs every HTTP request, including successful ones, along with every static asset (e.g., images, JavaScript files) that gets loaded. As a result, the logs are filled with hundreds of thousands of low-value entries, most of which are requests for static assets that don't need to be logged. During an incident, the team struggles to find relevant logs because they are buried under a massive volume of trivial HTTP request logs.

#### 6. Recommended Solution

- Use log levels effectively: Configure logging frameworks to use appropriate log levels (e.g., DEBUG, INFO, WARN, ERROR, FATAL). For production environments, ensure that the default log level is set to INFO or higher, and avoid logging verbose DEBUG-level logs unless necessary for troubleshooting specific issues.
- Log only what's necessary: Prioritize logging events that provide actionable insights, such as error messages, warnings, performance bottlenecks, and key user actions.

  Avoid logging excessive details that do not contribute to understanding system behavior.
- **Implement log sampling**: In high-volume systems, consider using log sampling techniques to limit the amount of data logged while still capturing enough information to detect anomalies or trends.
- Centralize log management: Use log aggregation platforms like ELK Stack, Splunk, or cloud-based logging services to centralize logs and make it easier to filter, analyze, and monitor log data. These platforms allow for better management of excessive logs through filtering, indexing, and setting up alerts on relevant events.
- **Regular log audits**: Conduct periodic audits of logging practices to ensure that logs are capturing the right level of detail. This includes removing unnecessary log entries, adjusting log levels, and improving log content for clarity and relevance.
- Set up log rotation and retention: Implement log rotation and retention policies to ensure that logs are archived or deleted in a timely manner, preventing excessive accumulation of old or redundant logs.

# 7. Consequences

#### Positive:

- Reduced log volume: Limiting excessive logs reduces storage and processing costs, as well as the burden on log management systems.
- **Improved performance**: By reducing the amount of logging, especially in production environments, system performance is improved, as there is less overhead associated with writing and managing logs.

- Faster incident response: With a more focused set of logs, teams can more quickly identify relevant information and resolve incidents, improving system reliability and reducing downtime.
- **Clearer insights**: Fewer, more relevant logs make it easier to detect trends, anomalies, and system issues, enabling more informed decision-making and proactive monitoring.
- Increased operational efficiency: Teams spend less time filtering through excessive logs and more time addressing important issues, leading to improved operational efficiency.

# Challenges:

- Balancing log detail: Striking the right balance between capturing sufficient details for troubleshooting and avoiding excessive logging can be tricky, particularly in complex systems.
- Initial effort for cleanup: Reducing excessive logging may require significant time and effort to identify which logs are unnecessary and modify logging configurations, especially in large systems with many services.
- Risk of under-logging: Reducing log verbosity too much may result in missing crucial information, making troubleshooting more difficult. Care must be taken to ensure that the logs retained provide enough context to diagnose issues effectively.
- **Cultural shift**: Encouraging teams to adjust logging practices to focus on meaningful data, rather than logging everything, may require a shift in mindset and process changes across the organization.

1. Antipattern Name: Inadequate Use of Logs for Metrics Collection

2. Category: Logs

- 3. Context: This antipattern occurs when logs are improperly or inefficiently used for metrics collection. Logs are often seen as a primary source of data for observability, but using them as a sole or primary means of collecting metrics can lead to significant inefficiencies. Teams might log detailed information but fail to extract the appropriate metrics from these logs, or they might rely on logs for metrics that could be more effectively tracked as dedicated metrics. The lack of structured logging and well-defined metric collection can cause challenges in analyzing and acting on performance-related data. This situation can arise due to a lack of understanding of the differences between logs and metrics, or due to teams trying to use logs as a catch-all for monitoring, rather than utilizing the right tool for the right purpose.
- **4. Problem:** When logs are inadequately used for metrics collection, teams face inefficiency due to the verbosity of logs, which may contain unnecessary information that makes it challenging to extract meaningful metrics. Logs capture data at a specific point in time and often aren't designed for real-time analysis, unlike dedicated metrics systems that can provide continuous insights. Without structured logging or sufficient context, it's difficult to correlate data across different systems, leading to missed opportunities to link metrics with events. Additionally, while

logs may offer detailed data, they often lack the specific actionable metrics necessary to monitor system health and performance, resulting in storage and processing overhead.

**5. Example:** A team logs every HTTP request with detailed data on request and response headers, body content, and timestamps. While these logs provide rich information, the team is also attempting to extract performance metrics like response times or error rates from them. This leads to slow processing, excessive log storage, and difficulty correlating the necessary data for meaningful analysis. Additionally, the lack of structured logging makes it difficult to aggregate the data into actionable insights, and performance problems are not detected until much later.

#### 6. Recommended Solution

- Use metrics for performance data: Use specialized metrics tools and systems to collect data on system performance (e.g., latency, error rates, throughput). Metrics are designed to be efficiently collected, stored, and analyzed in real-time.
- Separate logs and metrics: Maintain a clear separation between logs and metrics.
   Logs should capture detailed, unstructured event data, while metrics should track aggregated, structured performance data. Each tool serves a different purpose and should be used accordingly.
- **Implement structured logging**: If logs are being used to support metrics collection, ensure that logs are structured and consistent. Structured logs allow for easier parsing and analysis, and they help correlate logs with metrics.
- Leverage log aggregation tools: Use log aggregation and analysis tools that can help extract key performance metrics from logs (e.g., using tools like ELK stack, Splunk, or others) and avoid relying solely on raw logs for real-time metrics.
- Track key metrics separately: For system health, error rates, request latency, and other performance indicators, use dedicated metrics systems like Prometheus or Datadog to provide better real-time monitoring and visibility.
- Establish clear logging guidelines: Define what data should be logged and why. Avoid logging excessive details that may not be necessary for understanding system performance or troubleshooting.

#### 7. Consequences

# Positive:

- **Improved system performance monitoring**: By using the right tools for the right job (metrics for performance, logs for detailed events), teams can gain better visibility into system health and performance.
- Faster identification of issues: Dedicated metrics systems allow for real-time monitoring and quicker identification of issues, reducing troubleshooting time and improving response times during incidents.
- **Reduced overhead**: With metrics collected separately from logs, the system can more efficiently process and store relevant data, saving storage and computing resources.

 Easier correlation of data: Structured logging and dedicated metrics allow for easier and more accurate correlation across systems, making it simpler to diagnose root causes and optimize system performance.

# Challenges:

- **Increased complexity**: Implementing separate systems for logs and metrics requires more setup and maintenance, adding complexity to the observability infrastructure.
- Learning curve: Teams may need to adjust their workflows and become familiar with new tools or systems for metrics and logs, which can involve additional training and time investment.

**Data migration**: If logs have been improperly used for metrics collection in the past, migrating to a more appropriate system might require reworking existing log data, which can be time-consuming and challenging.

1. Antipattern Name: The Big Dumb Metric

2. Category: Metrics

- **3. Context:** This antipattern occurs when organizations focus on a single, overly simplistic metric that doesn't provide meaningful insight into system health or performance. Common examples include focusing on metrics like "CPU Usage" or "Request Count" without understanding the broader context. These metrics are easy to collect but often fail to reflect the true state of the system, leading to misguided decisions. Teams may rely on these "big dumb" metrics because they are readily available or easy to interpret, but they can lead to a false sense of security or misdiagnosis of issues.
- **4. Problem:** When "The Big Dumb Metric" is used, it can lead to misleading conclusions, as a single metric might not capture the full complexity of system performance or health. For example, high CPU usage could be normal during peak traffic, or low request counts might not indicate a healthy system if request latency is high. Teams may overlook important issues by focusing on this singular metric, missing out on more nuanced problems like slow response times, database bottlenecks, or intermittent failures. Additionally, relying on just one metric can result in an inadequate monitoring strategy, where valuable insights from other metrics are ignored, leading to poor observability practices. This false sense of security may cause teams to believe the system is healthy when underlying problems persist, delaying responses during incidents. Furthermore, it increases the risk of alert fatigue, as teams become overwhelmed by false alarms based on this over-simplified metric, reducing their effectiveness in detecting and addressing real issues.
- **5. Example:** A company monitoring its web application relies solely on the number of incoming requests (Request Count) as a key metric. If the request count is within an expected range, the

team assumes the application is functioning normally. However, during an incident, users experience slow response times, but since the request count is not dropping, no action is taken. In reality, the issue was a performance bottleneck in the database, which was not reflected by the request count alone. This causes a delay in resolving the issue and a poor user experience.

## 6. Recommended Solution

- Diversify metrics: Use a combination of relevant metrics that provide a fuller understanding of system health. For example, instead of just monitoring "Request Count," also monitor "Request Latency," "Error Rates," "Resource Utilization," and "Service Dependencies."
- **Focus on business outcomes**: Choose metrics that directly align with business goals and system objectives. Metrics like customer satisfaction, service uptime, or transaction success rates provide more actionable insights than raw system statistics alone.
- **Use composite metrics**: Combine multiple signals into a single metric that reflects more of the system's behavior, such as combining response time and error rate to create a more holistic view of service health.
- Contextualize metrics: Ensure metrics are always understood in context, such as comparing them with historical data or using benchmarks to establish what constitutes "normal" for your system.
- **Establish meaningful thresholds**: Instead of focusing on a single metric, set thresholds that indicate when a set of related metrics deviates from expected behavior. This can help in identifying issues earlier and more accurately.

# 7. Consequences

## Positive:

- **Improved visibility**: By diversifying metrics, teams can monitor the system more effectively, gaining insights into various aspects of performance and health, not just a single data point.
- **Faster issue identification**: A more nuanced set of metrics enables quicker detection of potential issues across different system components, leading to faster response times and more effective incident resolution.
- Better decision-making: With a broader set of metrics, teams can make more informed decisions about where to focus their efforts, improving overall system reliability and user experience.
- Reduced false alarms: Moving away from a single, simplistic metric reduces the number of false alarms triggered by out-of-context readings, helping teams focus on real problems.

# Challenges:

• **Increased complexity**: Adding more metrics can introduce complexity into monitoring and alerting systems, making it harder to track and manage the data.

- **Overhead**: Collecting and analyzing a wider variety of metrics may require additional resources, both in terms of system performance (e.g., collection overhead) and operational overhead (e.g., more alerts, more data to interpret).
- Choosing the right metrics: Deciding which metrics are truly meaningful and align with business outcomes can be difficult, particularly in large, complex systems. The right balance between simplicity and comprehensiveness must be found.
- Avoiding data overload: While diversifying metrics is important, it's crucial not to
  overwhelm teams with too much data. Effective metrics should provide actionable
  insights without overwhelming the system or personnel with irrelevant information.

1. Antipattern Name: Relying Only on API Monitoring

2. Category: Metrics

- **3. Context:** This antipattern occurs when organizations or teams rely exclusively on monitoring the APIs of their system without considering other aspects of their infrastructure or application. While API monitoring is essential, it only provides visibility into the availability and performance of API endpoints and does not cover other critical areas, such as system resource usage, internal service communication, or the health of other components in the ecosystem. Teams may focus on APIs because they are the primary interface for interacting with the system, but this narrow focus can result in blind spots in the system's observability.
- **4. Problem:** When teams rely solely on API monitoring, they face several limitations in understanding the overall health of the system. API monitoring primarily provides insights into how the API endpoints themselves are performing but doesn't capture the internal health of the system, including database performance, service-to-service communication, or resource utilization. Issues like high CPU usage, memory leaks, disk space exhaustion, or network latency can impact API performance but may remain undetected if only the API is monitored. Moreover, this narrow focus can hinder the diagnosis of complex problems that affect backend systems or dependencies, making it harder to identify and resolve root causes of performance bottlenecks. Relying solely on API monitoring can also delay incident detection and response, as problems with dependent services or databases may go unnoticed until they directly affect the API. Additionally, it limits the ability to gain full insights into end-to-end user experience, as API monitoring typically misses client-side performance, network latency, and multi-service interactions.
- **5. Example:** An e-commerce platform relies solely on API monitoring to track the performance of its order placement API. During peak shopping hours, customers experience slow page load times and occasional failures to complete transactions. However, the API monitoring shows that the order placement API is working perfectly, with all requests completing successfully and within normal latency ranges. The root cause is identified later as a database bottleneck that was not captured by the API monitoring alone, leading to delayed customer transactions.

#### 6. Recommended Solution

- **Monitor all layers**: Implement full-stack monitoring that includes API monitoring, system resource monitoring (CPU, memory, disk, network), database performance, internal service health, and infrastructure monitoring (e.g., load balancers, caches, queues). This will provide a complete picture of the system's health.
- End-to-end tracing: Utilize distributed tracing to monitor how requests flow through the entire system. This allows you to see how various microservices interact with each other and pinpoint where delays or failures occur, even if they don't affect the API directly.
- Real-user monitoring (RUM): Complement API monitoring with real-user monitoring to
  capture the actual user experience, including frontend performance and client-side
  issues that might impact how users interact with your system.
- **Proactive alerting**: Set up alerts for both the APIs and the underlying infrastructure, including resource utilization thresholds, service failures, and error rates. Ensure that alerts are actionable and reflect real issues, not just API availability.
- Integrate logs and metrics: Combine log management and metrics with API monitoring
  to get more contextual information about failures. Logs from backend services or
  infrastructure components can give additional context to API performance metrics and
  help trace issues across systems.
- Use synthetic monitoring: In addition to API monitoring, use synthetic monitoring to simulate user interactions with the system. This can help detect issues before real users are affected, especially in non-API parts of the system like web pages or mobile app interactions.

# 7. Consequences

#### Positive:

- Comprehensive observability: Monitoring beyond just APIs allows teams to understand the full stack of the system, from infrastructure and services to user interactions and performance.
- Faster root cause identification: With complete observability, teams can more quickly diagnose issues that affect both APIs and underlying components, reducing time to resolution during incidents.
- **Improved user experience**: By considering both server-side and client-side performance, teams can gain better insights into the end-to-end user experience, leading to more reliable and responsive applications.
- Increased system reliability: Proactive monitoring of the entire system improves system reliability, as teams can detect and resolve issues before they affect users or cause downtime.

## Challenges:

• **Increased complexity**: Implementing monitoring across all layers of the system can increase complexity and require more resources for setup, management, and analysis.

- Data overload: With increased monitoring coverage, there's the risk of being overwhelmed with data. Proper filtering, aggregation, and alerting strategies are necessary to avoid alert fatigue or unnecessary noise.
- Cost: Comprehensive observability solutions, particularly those that include full-stack
  monitoring, distributed tracing, and real-user monitoring, can introduce additional costs
  for tools, infrastructure, and storage.
- **Cultural shift**: Shifting from API-only monitoring to full-stack observability might require a cultural change, as teams may need to rethink their monitoring and alerting strategies and adopt new tools and workflows.

1. Antipattern Name: Misunderstanding Metrics

2. Category: Metrics

- **3. Context:** This antipattern occurs when teams or organizations incorrectly interpret or use metrics, leading to misguided decisions and actions. This could involve misunderstanding what a metric is actually measuring, how it should be used, or what it indicates about the system's health or performance. Teams may focus on the wrong metrics, misinterpret their meaning, or fail to understand their context, resulting in poor decision-making and ineffective actions. This often happens when teams don't have the right training, experience, or understanding of what specific metrics truly reflect about the system.
- **4. Problem:** When metrics are misunderstood, it can lead to several significant challenges for teams trying to maintain system health. Incorrect interpretations of metrics can cause teams to draw the wrong conclusions, leading to misprioritization of issues. They may focus on non-critical problems while overlooking key metrics that indicate more serious system components are in need of attention. Misunderstanding the meaning or significance of a metric can also create a false sense of security, making teams believe the system is performing well when, in reality, there are hidden issues. This lack of clarity results in improper prioritization, where resources are allocated to less important problems. Additionally, the inability to correctly interpret metrics can slow down troubleshooting efforts, delaying the resolution of actual issues and potentially impacting system reliability.
- **5. Example:** A team monitors the "Request Count" metric for their API and sees a steady increase in requests. They assume everything is functioning well, but when looking closer, they notice that the "Error Rate" has also increased significantly. The "Request Count" alone led them to ignore a growing problem, which was a surge in failed requests. As a result, they failed to take action to resolve the issue in a timely manner, impacting user experience.

#### 6. Recommended Solution

- Understand metrics in context: Ensure that metrics are understood in the context of
  the system they're measuring. For example, CPU usage should be understood not just
  as a number but in relation to the expected workload, peak traffic times, or specific use
  cases.
- Use complementary metrics: Rely on a combination of metrics to form a more complete picture. For example, instead of relying solely on "Request Count," also monitor "Error Rate," "Latency," and "Throughput" to understand both system health and user experience.
- **Educate teams**: Provide proper training on interpreting metrics correctly, and ensure that the team understands the significance of each metric. Teams should know how each metric reflects system behavior and what actions are required when specific thresholds are crossed.
- Align metrics with business goals: Ensure that the metrics chosen are tied to key
  performance indicators (KPIs) that reflect business outcomes. For example, instead of
  focusing solely on infrastructure metrics, consider metrics like user engagement,
  transaction success rate, or service uptime.
- Avoid metric overload: Don't overwhelm teams with an excessive number of metrics.
   Focus on actionable, meaningful metrics that contribute directly to the system's health and the user experience.
- Establish clear thresholds and baselines: Define what is considered normal or healthy for each metric and establish clear thresholds for alerting. This helps prevent misinterpretation and ensures that teams act on real issues, rather than reacting to noisy or ambiguous data.

#### Positive:

- **Better decision-making**: A clear understanding of metrics leads to more informed decisions, as teams can accurately assess system performance and health.
- **Faster issue resolution**: With the right metrics and an accurate understanding of their context, teams can identify issues more quickly and take the appropriate action to resolve them.
- Improved system performance: Understanding metrics properly enables teams to focus on optimizing the right areas of the system, leading to better overall performance and reliability.
- **Effective monitoring**: By focusing on complementary metrics and understanding their significance, teams can improve their monitoring practices and respond effectively to incidents and alerts.

# Challenges:

Training and expertise: Ensuring that everyone understands how to interpret and use
metrics correctly may require additional training or hiring experts, particularly in complex
systems.

- **Metric definition**: It may take time to properly define meaningful metrics that align with business goals and system behavior, especially in large or dynamic environments.
- Avoiding overcomplication: While it's important to have a comprehensive set of
  metrics, there's a risk of complicating the observability strategy by overloading with too
  many metrics, making it harder to focus on the most important data.

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1. Antipattern Name: Overlooking Derived Metrics

2. Category: Metrics

- **3. Context:** This antipattern occurs when teams fail to track or consider derived metrics that could provide deeper insights into the system's behavior. Derived metrics are calculated from raw metrics and offer a more meaningful understanding of system health, performance, or usage patterns. Examples of derived metrics include averages, percentiles, ratios, or aggregated statistics that provide more context to raw measurements. Organizations may overlook derived metrics because they focus on basic, raw data such as request count or response time without considering the more nuanced insights that derived metrics can offer.
- **4. Problem:** When teams overlook derived metrics, they may miss critical insights that help to accurately assess the system's health and performance. Raw metrics, such as basic latency or error counts, typically don't offer enough context to fully understand the system's behavior or detect issues in a timely manner. Derived metrics, which combine multiple raw metrics, can uncover trends and anomalies that might otherwise go unnoticed, providing a more holistic view of the system. Without these derived insights, teams may struggle to prioritize issues effectively, as raw metrics alone don't offer a clear indication of which problems are most pressing. Additionally, derived metrics often provide clearer, more actionable data, helping teams make informed decisions. By ignoring these metrics, teams risk missing larger systemic issues, such as architectural flaws, scalability problems, or performance bottlenecks, which could ultimately impact system reliability and performance.
- **5. Example:** A team is monitoring the raw response time for a web application and notices occasional spikes in latency. However, they don't track the 95th percentile or histogram of latency, so they cannot determine if the spikes are affecting a small subset of users or the entire user base. Later, when users report poor performance, they realize that the latency spikes were happening only for a specific percentage of users, but without derived metrics, this insight was not immediately available.

#### 6. Recommended Solution

 Track derived metrics alongside raw metrics: In addition to tracking raw metrics, ensure that derived metrics (such as averages, percentiles, rates, and ratios) are also monitored. For example, track both response time and the 95th percentile latency to understand performance for the majority of users versus outliers.

- **Use percentiles and histograms**: Percentiles (e.g., 95th or 99th) provide better insight into response time distribution and highlight potential performance degradation for a significant portion of users, even if the average remains within acceptable limits.
- Calculate ratios and success rates: Derived metrics like success rate (successful
  requests / total requests) or error rates (errors / total requests) give a better
  understanding of system reliability and help prioritize issues based on their impact.
- Aggregate data across multiple dimensions: Combine metrics to understand system
  performance in more detail. For example, aggregate response times by endpoint type,
  user demographics, or geographical regions to understand if specific segments of users
  are being affected differently.
- Implement anomaly detection on derived metrics: Use anomaly detection techniques on derived metrics to automatically highlight unusual patterns or performance degradation. For example, setting alerts on deviations from expected percentiles can help detect emerging issues that might not be immediately obvious from raw metrics.
- **Define thresholds based on derived metrics**: Set thresholds and alerts based on meaningful derived metrics rather than raw metrics. This ensures that the alerts reflect real user impact and performance degradation.

## Positive:

- **Deeper insights**: Tracking derived metrics provides a more granular understanding of system behavior, allowing teams to identify and address issues more effectively.
- **Improved issue prioritization**: Derived metrics help teams better prioritize incidents based on the real impact of the problem, such as error severity, request failure rates, or user experience degradation.
- Proactive monitoring: With derived metrics, teams can detect trends, anomalies, or
  potential issues early, allowing for proactive issue resolution before they affect a larger
  portion of users.
- **Better decision-making**: Derived metrics provide a clearer picture of system health and performance, helping teams make better decisions regarding system optimizations and resource allocation.

#### Challenges:

- **Increased complexity**: Tracking and managing derived metrics adds complexity to monitoring systems, requiring additional configuration, tools, and expertise.
- **Risk of data overload**: With more metrics to track, there's a risk of information overload. It's important to avoid collecting too many derived metrics and focus on the ones that provide actionable insights.
- Additional processing requirements: Calculating and storing derived metrics can add processing overhead, especially in high-scale environments. Careful attention must be paid to performance and storage requirements.

 Potential for misinterpretation: Without proper understanding and context, derived metrics can be misinterpreted or lead to false conclusions. Training teams to properly interpret these metrics is essential.

1. Antipattern Name: Only Tracking What the Customer Sees

2. Category: Metrics

- **3. Context:** This antipattern arises when teams focus solely on tracking user-facing metrics (such as response times, availability, or page load times) while neglecting critical internal metrics that reflect the underlying system's health and performance. It typically occurs because teams prioritize metrics that are visible to customers, assuming they are the only important indicators of system performance. This focus often leads to a lack of visibility into the internal workings of the system, which can result in undetected issues or a delayed response to underlying problems.
- **4. Problem:** When teams only track customer-facing metrics, they often overlook important internal issues that can affect system performance and stability. Focusing solely on metrics like frontend performance, response times, or customer-facing errors means that backend bottlenecks, resource exhaustion, or failures within internal services may go undetected. This narrow scope prevents teams from identifying systemic issues within the architecture or infrastructure, leading to delays in responding to underlying problems until they impact the end user. Furthermore, without visibility into the inner workings of the system, root cause analysis becomes challenging, as teams may not be able to trace issues to the specific services or components causing them. As a result, teams may have an over-simplified view of the system's health, missing key indicators of trouble before they affect the user experience.

#### 5. Example

A team focuses primarily on tracking the load times of their website and the number of successful transactions processed by their user-facing APIs. However, they don't track internal metrics, such as database query performance, backend service latencies, or CPU and memory usage on critical servers. Over time, the database experiences performance degradation due to high query load, but users don't notice immediate slowdowns because the team is not tracking the internal bottlenecks. Eventually, the system crashes under heavy load, and the team is caught by surprise.

#### 6. Recommended Solution

- Track internal and external metrics: In addition to user-facing metrics, track internal
  metrics such as backend service performance, database query times, memory usage,
  CPU load, and network latency. These metrics provide insights into how the system is
  performing behind the scenes and can help detect issues before they affect users.
- **Focus on system reliability**: Ensure that internal metrics, like resource utilization and service health, are regularly monitored. This will help identify issues such as resource

- contention, service degradation, or unexpected failures, and allow teams to address them proactively.
- Set up monitoring across all layers: Implement end-to-end monitoring that covers all
  aspects of the system, from user-facing metrics to backend services and infrastructure.
  This allows for comprehensive visibility and faster detection of potential issues that could
  affect system reliability.
- Automate alerts for internal issues: Set up automated alerts for internal metrics, such as database slow queries, server CPU spikes, or memory leaks, so that the team is notified as soon as these issues arise, even if they don't immediately affect the user experience.
- Balance user-facing and infrastructure metrics: Aim for a balanced approach where both customer-facing metrics and system-level metrics are tracked. This ensures that teams can respond to performance degradation or failures at all levels of the stack, not just the customer-facing layer.
- Include service health metrics in dashboards: Create comprehensive dashboards that visualize both external and internal metrics. This can help teams quickly identify issues across all layers of the system, from the frontend to backend services.

## Positive:

- **Improved proactive issue detection**: By tracking both internal and external metrics, teams can detect issues before they affect the customer, allowing for quicker resolutions and a better overall user experience.
- Better system health insights: Monitoring internal metrics provides a more complete view of the system's performance, enabling teams to optimize resources and improve service reliability.
- Faster troubleshooting: With a holistic view of system metrics, teams can more quickly identify the root cause of problems, improving troubleshooting efficiency and reducing downtime.
- **Enhanced reliability**: By addressing issues within the system before they impact the user experience, teams can ensure more stable, reliable services that are less prone to outages or performance degradation.

## Challenges:

- Increased complexity in monitoring: Monitoring both user-facing and internal metrics requires a more sophisticated setup, including additional tooling and configuration to capture and analyze the full range of metrics.
- Risk of overloading with data: Tracking too many metrics can lead to information overload. It's essential to prioritize and focus on the most meaningful metrics to avoid being overwhelmed by noise.
- Resource allocation: Tracking and maintaining internal monitoring can require
  additional resources, both in terms of tools and personnel, which may require additional
  investment and effort.

 Balancing metrics: Ensuring that both customer-facing and internal metrics are tracked without overwhelming the system or the team is critical. Striking the right balance between performance and observability is key.

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1. Antipattern Name: Excessive Metrics

2. Category: Metrics

- **3. Context:** This antipattern occurs when teams track an excessive number of metrics, often without clear purpose or prioritization. While metrics are critical for observability, collecting too many can lead to data overload, making it difficult to focus on the most important indicators of system health and performance. Teams may fall into this trap by attempting to track every possible data point available, or they may feel compelled to collect metrics from every part of the system without analyzing whether they are truly useful. This can result in wasted resources, difficulty in identifying actionable insights, and slower system performance due to the overhead of monitoring.
- **4. Problem:** When teams track excessive metrics, they risk overwhelming themselves with too much information, making it difficult to identify the metrics that truly matter. Collecting and storing an excessive amount of data consumes unnecessary resources, including storage space and processing power, which could otherwise be used more efficiently. Additionally, the sheer volume of metrics can complicate decision-making, as teams struggle to prioritize what is important. This overload of data also slows down troubleshooting efforts, as finding the root cause of an issue becomes more difficult amidst a sea of irrelevant or granular metrics. Ultimately, excessive metrics may fail to provide actionable insights, preventing teams from focusing on the most critical factors that impact system performance and health.
- **5. Example:** A team monitors hundreds of metrics across different services, including metrics for every individual API endpoint, request method, and even specific database query. While some of these metrics provide useful information, others provide little value and add unnecessary noise. During a performance degradation event, the team struggles to find the relevant information because they are flooded with a large number of metrics, making it difficult to identify the key indicators of the issue. As a result, troubleshooting is delayed, and the incident takes longer to resolve.

#### 6. Recommended Solution

- Define key metrics: Identify and track only the most critical metrics that align with system objectives and performance goals. Focus on high-level indicators that directly impact system reliability and user experience, such as error rates, request latency, resource utilization, and availability.
- Use metric hierarchies: Organize metrics in layers, where high-level metrics provide a summary of the system's health, and lower-level metrics provide more detailed insights

- only when necessary. For example, track request latency as a high-level metric and only dig into specific request types or endpoints when an anomaly is detected.
- Set clear objectives for each metric: Before collecting a new metric, define its purpose and how it will help the team make decisions or improve system performance. Avoid collecting metrics that don't serve a clear and actionable goal.
- Regularly review metrics: Continuously evaluate which metrics are being tracked, and remove or consolidate metrics that are not providing actionable insights. Periodically audit your monitoring systems to ensure that only relevant data is being collected.
- **Implement alerting thresholds**: Set thresholds for alerts based on meaningful metrics rather than tracking excessive numbers of metrics. Alerts should be focused on the most important events that indicate system health issues or performance degradation.
- **Use sampling or aggregation**: Instead of tracking every single metric point, use sampling techniques or aggregated metrics (such as averages or percentiles) to reduce the data volume while retaining meaningful insights.

## Positive:

- **Improved focus**: By tracking only the most relevant metrics, teams can focus their attention on the critical indicators of system health, leading to more effective monitoring and troubleshooting.
- Reduced resource consumption: Collecting and storing fewer metrics reduces the overhead on the system, freeing up resources that can be used elsewhere, and lowering costs.
- **Faster troubleshooting**: With a more focused set of metrics, it becomes easier and quicker to identify the root cause of issues, improving incident response times.
- Better decision-making: Teams will have clearer, actionable data that can drive better decisions regarding system optimization, capacity planning, and performance improvements.

## Challenges:

- **Initial investment in metric selection**: Choosing the right set of metrics requires careful planning and may involve initial trial and error to determine which data points are truly valuable.
- Balancing between detail and overview: Striking the right balance between tracking enough detail for troubleshooting and avoiding excessive granularity can be challenging. It's important to adjust monitoring as system complexity grows.
- **Dealing with legacy data**: If the system has accumulated excessive metrics over time, there may be challenges in cleaning up and reorganizing the data collection to ensure only the relevant metrics remain.

# 2. Category: Metrics

- **3. Context:** This antipattern occurs when critical components or parts of the system are not properly monitored, leaving gaps in observability. These gaps may arise due to limited monitoring scope, focusing only on certain services or systems, or assuming that some parts of the system don't require attention. It often happens when teams prioritize monitoring for the most visible or high-traffic areas (like the frontend or primary APIs) while neglecting background services, infrastructure, or less obvious parts of the system that could lead to failures or degraded performance. Inadequate monitoring coverage leaves teams blind to potential issues and slows down the ability to detect and respond to incidents.
- **4. Problem:** When teams fail to adequately monitor all critical components of a system, they create blind spots in observability. Key services, such as background jobs, databases, or auxiliary services, may be left unmonitored, which can lead to undetected issues that escalate into major incidents. As a result, issue detection is delayed, and incident response becomes slower because teams lack the necessary insights to quickly identify the root cause. Additionally, this lack of monitoring coverage limits the ability to predict and prevent issues before they impact the system. Ultimately, neglecting to monitor crucial components wastes resources, as investments in monitoring tools and infrastructure may not cover the most critical parts of the system, undermining the effectiveness of the overall observability strategy.
- **5. Example:** A team focuses their monitoring on the user-facing frontend application and primary APIs, but they neglect to monitor the backend systems like the database, caching layer, or microservices that support the frontend. One day, the database experiences increased query latency, but because it's not being monitored properly, the issue is only noticed when customer-facing applications start slowing down. By then, the incident is much harder to mitigate, and customers are impacted.

#### 6. Recommended Solution

- Comprehensive monitoring: Ensure that all critical components, including frontend, backend, databases, caching layers, message queues, and infrastructure, are covered by monitoring. No part of the system should be overlooked, regardless of its visibility to customers.
- System-wide observability: Implement monitoring that spans the entire stack, from infrastructure to services to user-facing applications. This ensures that all interactions and dependencies are visible and can be tracked.
- Regularly audit monitoring coverage: Conduct regular audits of your monitoring system to identify any areas of the system that may be under-monitored or neglected. Ensure that every service and component is continuously observed.
- **Use a layered monitoring approach**: Build a monitoring system with multiple levels of visibility, from high-level service health to low-level infrastructure metrics. This allows teams to detect issues both at a glance and in detail.

- Automate alerts for overlooked areas: Implement automated alerts for all monitored areas to ensure that any issues in critical components, even those less visible, are detected early and can be addressed promptly.
- Prioritize monitoring based on risk: Determine which components of the system are
  most critical to service reliability and prioritize monitoring coverage for those areas,
  ensuring that high-risk components are closely watched.

#### Positive:

- **Faster issue detection**: With comprehensive monitoring coverage, teams can identify and resolve issues much faster, before they escalate and affect end users.
- Improved system reliability: By having full visibility into all system components, teams
  can ensure that every part of the system is functioning properly, leading to a more
  reliable overall system.
- **Reduced blind spots**: Proper monitoring coverage eliminates blind spots, allowing teams to proactively detect and fix issues across the entire system.
- **Enhanced troubleshooting**: When all components are monitored, it becomes much easier to diagnose the root cause of issues, reducing troubleshooting time and improving response times during incidents.

# Challenges:

- Increased monitoring complexity: Expanding monitoring coverage across all system components requires more sophisticated monitoring setups and increased complexity in managing the data collected.
- Higher resource requirements: More comprehensive monitoring can result in increased resource consumption, both in terms of monitoring infrastructure and personnel to manage the collected data.
- Balancing monitoring granularity: Striking the right balance between monitoring critical systems without overwhelming the team with unnecessary data or alerts can be challenging.

1. Antipattern Name: Forgetting the Customer

2. Category: Metrics

**3. Context:** This antipattern occurs when teams focus too much on internal system performance or infrastructure metrics and forget to monitor or account for the customer experience. It typically happens when there's an overemphasis on technical metrics (e.g., server health, database performance) at the expense of tracking the customer's perspective, such as application responsiveness, usability, or user satisfaction. The issue arises when teams lose sight of the fact that the ultimate goal of observability is to ensure that the system works effectively for the

end user. This can lead to missed signals about how users are actually experiencing the system.

#### 4. Problem

When teams fail to monitor or prioritize customer-facing metrics, it leads to a lack of insight into the actual user experience. Teams may miss crucial metrics like response times, error rates, and user satisfaction, which could signal performance issues that directly affect users. While internal system metrics might show optimal performance, neglecting customer-facing data can cause teams to miss user impact and fail to detect problems until customers report them. This delay in identifying user-facing issues can lead to poor prioritization of technical fixes that don't align with improving the customer experience, ultimately resulting in a decline in customer satisfaction over time.

**5. Example:** A team monitors the server CPU usage, database performance, and internal API latencies, but they do not track user-facing metrics like page load times or transaction success rates. While the internal system performs well, users begin experiencing slowdowns and intermittent failures when accessing the website. The team does not realize the impact on users until they receive multiple customer complaints, leading to a slower response in resolving the issue.

#### 6. Recommended Solution

- Track user-facing metrics: Ensure that metrics related to the customer experience—such as page load times, API response times, transaction success rates, and error rates—are actively monitored and prioritized.
- Align system monitoring with user impact: When designing monitoring systems, include metrics that reflect how users are interacting with the system. For example, focus on frontend performance, user journey tracking, and real-time transaction monitoring.
- **Utilize customer satisfaction signals**: Integrate customer feedback and satisfaction metrics (e.g., NPS, CSAT, user-reported issues) with your monitoring systems to ensure you are hearing directly from users about their experiences.
- **Real-time user impact analysis**: Implement real-time monitoring that correlates system performance with user experience. This helps quickly identify if technical problems are affecting users in real-time and allows for faster mitigation.
- Create user-centric dashboards: Build dashboards that focus on user-facing metrics, ensuring that teams can quickly identify issues that are directly impacting users and prioritize them accordingly.
- Balance internal and user metrics: Strive to maintain a balance between monitoring internal system performance and tracking the user experience. This ensures that both perspectives are accounted for in decision-making.

# 7. Consequences

Positive:

- **Improved customer experience**: By focusing on user-facing metrics, teams can ensure that issues affecting customers are quickly detected and resolved, leading to a more seamless user experience.
- Faster response times to user issues: With direct visibility into how users are interacting with the system, teams can address user issues more swiftly, reducing downtime and improving customer satisfaction.
- Enhanced prioritization of efforts: Aligning monitoring with the customer experience allows teams to prioritize the most impactful issues for users, ensuring resources are focused on the right problems.
- **Higher customer satisfaction**: By tracking and addressing user-centric issues, teams can improve the overall satisfaction and trust that users have in the service.

# Challenges:

- Additional monitoring complexity: Tracking both internal system metrics and user-facing metrics can add complexity to the monitoring setup, as teams need to integrate and analyze both types of data.
- **Increased resource requirements**: Focusing on user-facing metrics may require additional tools or infrastructure to collect and process data related to user experience, potentially increasing the resource consumption of the observability system.
- Balancing technical and customer concerns: It can be difficult to balance the
  technical needs of the system with the requirements of the customer experience. Teams
  must ensure that both internal system health and the customer perspective are
  adequately monitored.

1. Antipattern Name: Long Trace Spans

2. Category: Traces

- **3. Context:** This antipattern occurs when trace spans (units of work or transactions in distributed tracing systems) are too long, often capturing more data than necessary. In distributed tracing, trace spans are used to measure and track the duration of specific operations, requests, or interactions within a system. When these spans are overly long, they can introduce several issues, such as the inability to detect bottlenecks, poor performance insights, and data overload. This often happens when the scope of a trace spans too much of the system's operation, or when tracing is enabled for long-running tasks or processes that would be better split into smaller, more manageable spans. It can be a result of improper tracing configurations, lack of granularity in tracing setup, or poor design of distributed trace structure.
- **4. Problem:** When trace spans are too long, teams can encounter overwhelming trace data, making it difficult to extract meaningful insights. The excessive data can hinder the identification of bottlenecks, as long trace spans cover too many operations, masking the specific source of delays. Additionally, while these spans may show interactions between components, they fail to

provide detailed insights into individual operations or time segments. Tracing long spans also leads to resource inefficiency, as maintaining context and collecting data over extended periods consumes significant resources. Moreover, correlating long trace spans with other observability data, such as logs or metrics, becomes more challenging, complicating troubleshooting and performance analysis.

**5. Example:** A team is tracing an entire user registration process, which includes multiple backend services like authentication, user data storage, and email verification. Instead of tracing each service separately, the entire flow is captured as one long span. This results in a single trace that lasts several minutes, making it difficult to identify which part of the process is slowing down (e.g., the database write operation or the email API call). The trace data is large and difficult to analyze, and the team struggles to optimize the performance of any individual part of the registration process.

#### 6. Recommended Solution

- Break down spans into smaller, more granular units: Instead of tracing large
  workflows as single spans, break them down into smaller, more manageable spans for
  each individual service or operation. For example, trace each database query, HTTP
  request, or API call separately to get more detailed visibility into where delays occur.
- Set appropriate span duration limits: Set a threshold for span duration to ensure that only spans that represent meaningful time intervals are collected. If spans are too long, split them into smaller sub-spans based on logical units of work.
- **Use context propagation wisely**: Ensure that trace context is propagated correctly across services, so each trace spans a logical unit of work, and the system can capture meaningful data without becoming too granular or too coarse.
- Focus on critical paths: Prioritize tracing on critical service interactions or user journeys, such as high-traffic paths or transactions that are known to impact performance. Avoid over-tracing less critical operations that may not provide valuable insights.
- **Implement hierarchical tracing**: Use hierarchical tracing techniques to capture parent-child relationships between spans. This allows you to capture the full context of a transaction, while still breaking down the trace into manageable pieces.
- Review trace configurations regularly: Periodically review and refine tracing configurations to ensure they capture only the relevant spans and that span durations are appropriate for the systems and operations being monitored.

# 7. Consequences Positive:

• **Improved trace granularity**: By breaking down long spans into smaller units, teams can gain better insights into specific areas of the system and pinpoint performance issues more quickly.

- Faster performance troubleshooting: Smaller, well-defined spans make it easier to isolate and identify the exact cause of delays or bottlenecks, leading to quicker resolution times.
- Reduced resource consumption: Collecting smaller, more targeted spans results in lower overhead for storing and processing trace data, reducing the strain on observability systems and resources.
- **Better correlation with other observability data**: Granular spans provide better context, making it easier to correlate traces with logs, metrics, and other monitoring data to understand the full scope of incidents or performance issues.

# Challenges:

- **Increased complexity**: More granular spans can increase the complexity of trace management and make it harder to maintain consistency across large systems.
- **Potential for trace overload**: While more granular spans are helpful, they can also lead to an overwhelming amount of trace data, especially in high-throughput systems. It's important to find a balance between granularity and data volume.
- **Tuning and optimization**: Properly configuring and tuning the span durations and collection mechanisms to match the system's needs can require significant experimentation and ongoing adjustments.

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1. Antipattern Name: Underestimating the Importance of Traces

2. Category: Traces

- **3. Context:** This antipattern occurs when teams or organizations fail to fully appreciate the value of distributed tracing as part of their observability strategy. Traces are essential for providing visibility into the lifecycle of requests or transactions across multiple services, helping to detect bottlenecks, latency issues, and failures in a distributed system. However, some teams might underutilize or neglect tracing because they rely too heavily on metrics and logs, assuming that these are sufficient for monitoring system performance and troubleshooting issues. As a result, they miss out on the detailed insights that traces provide, leading to slower incident resolution, difficulty identifying root causes, and a less comprehensive understanding of system behavior.
- **4. Problem:** When the importance of traces is underestimated, teams face limited visibility into system behavior, as traces provide a full picture of request flow across services. Without traces, identifying latency bottlenecks becomes difficult, especially in complex, distributed systems, as metrics like response times or throughput are insufficient. Incident resolution times are longer, as teams have to rely on logs and metrics, which require more time to piece together. Debugging becomes more complex without the step-by-step visibility that traces offer, and missed optimization opportunities may arise, as issues like inefficient service calls or slow database queries often go unnoticed without traces.

**5. Example:** A company relies heavily on metrics like response times and error rates to monitor the performance of their web application. However, when a critical issue occurs, such as a significant slowdown in the registration process, the metrics alone aren't enough to pinpoint the cause. The team is left to analyze logs and metrics from individual services, but they can't see the complete flow of requests across the distributed system. If they had implemented distributed tracing, they would have been able to quickly identify that a database query in the authentication service was causing a delay, but instead, they waste time searching in the wrong places.

#### 6. Recommended Solution

- Integrate distributed tracing into the observability stack: Make traces a first-class
  citizen in your observability strategy. Use tracing alongside logs and metrics to get a
  complete picture of system performance, especially for identifying latency bottlenecks
  and understanding request flows.
- Monitor critical paths with tracing: Ensure that you are tracing high-impact and high-traffic paths, such as user transactions, payment processes, and API calls. This provides valuable insights into the most critical aspects of the system.
- Train teams on the value of traces: Educate your engineering and operations teams about the benefits of tracing and how it can help in troubleshooting, optimizing performance, and improving system reliability. Make sure they understand how to use traces effectively in their workflows.
- Correlate traces with logs and metrics: Use traces in combination with logs and metrics to get deeper insights into system performance. For example, correlating a trace with related error logs can help identify the root cause of an issue more quickly.
- Invest in tracing tools: Ensure you have the necessary tools and infrastructure in place
  to capture, store, and analyze trace data. Choose distributed tracing solutions that
  provide easy integration with your existing observability stack.
- Review trace coverage regularly: Periodically review your system's tracing coverage to ensure that all critical services and operations are adequately instrumented. Add new traces as your system evolves and new services are introduced.

# 7. Consequences

#### Positive:

- Faster incident resolution: With comprehensive tracing in place, incidents can be diagnosed and resolved much more quickly, reducing downtime and improving system reliability.
- **Improved performance optimization**: Tracing helps identify latency bottlenecks and inefficient operations, which can then be addressed to optimize system performance.
- **Better overall observability**: When tracing is integrated with metrics and logs, it provides a more holistic view of system health, leading to improved decision-making and proactive issue resolution.
- **Increased reliability**: With better visibility into how services interact, teams can better understand the causes of failures and take corrective actions to reduce future incidents.

# Challenges:

- **Initial setup complexity**: Setting up distributed tracing requires some initial investment in instrumentation, configuration, and tool integration. This can be a barrier, especially for systems that are not already instrumented for tracing.
- Data volume and storage costs: Tracing can generate a large volume of data, especially in high-throughput systems, which may lead to higher storage and processing costs. Proper management and filtering of trace data are necessary to avoid this.
- **Learning curve**: Teams may face a learning curve in understanding how to interpret trace data and integrate it effectively with logs and metrics. Training and experience are necessary to get the full benefit of tracing.

1. Antipattern Name: No Consistent Trace ID

2. Category: Traces

- **3. Context:** This antipattern occurs when a distributed system fails to propagate a consistent trace ID across all services and components involved in a transaction. A trace ID is a unique identifier assigned to a single request or transaction that travels through multiple services, allowing teams to trace its flow and diagnose issues. Without a consistent trace ID, it becomes difficult to correlate related logs, metrics, and traces across different services, making it harder to debug issues, track performance, or understand the full journey of a request. This is typically seen when services are not properly instrumented to propagate the trace ID across service boundaries or when manual correlation is needed to connect traces from different sources.
- **4. Problem:** When trace IDs are not consistently propagated, teams face difficulties in correlating data across logs, metrics, and traces, hindering a comprehensive view of system behavior. Troubleshooting times increase because engineers have to manually cross-reference data from multiple services without the benefit of a consistent trace ID. This results in incomplete visibility into the flow of requests across services, making performance analysis harder. Identifying performance bottlenecks becomes a challenge, and the risk of missing issues, such as slow response times or failures, rises, as real-time tracking of requests is not possible.
- **5. Example:** A user submits a request to purchase an item on an e-commerce platform. The request passes through several services: authentication, inventory, payment processing, and order fulfillment. Each service logs its own data, but because the trace ID is not consistently passed from one service to the next, the logs from each service cannot be linked together. The engineering team struggles to diagnose why the payment processing is delayed since there is no unified trace showing the request flow, leading to increased time in troubleshooting.

#### 6. Recommended Solution

- Implement trace ID propagation: Ensure that all services involved in handling a request (including microservices, APIs, and backend systems) propagate the same trace ID through their logs, metrics, and traces. This can be done by injecting the trace ID into the HTTP headers or other communication channels between services.
- **Use distributed tracing tools**: Utilize distributed tracing systems (e.g., OpenTelemetry, Jaeger, Zipkin) that automatically generate and propagate trace IDs across services. These tools integrate well with existing observability stacks, making it easier to track requests across multiple services.
- Standardize trace ID generation: Standardize how trace IDs are generated and propagated across your system. Ensure that the same methodology is used for generating trace IDs in all components, services, and environments.
- Enforce trace ID inclusion in logs: Configure your logging framework to automatically include the trace ID in all log entries, so that logs from all services are traceable and correlated.
- Monitor trace ID consistency: Regularly monitor the consistency of trace ID
  propagation across your system. Implement health checks and alerts to ensure trace IDs
  are being passed correctly between services.
- Provide trace ID documentation: Document the trace ID generation and propagation process, ensuring that all teams understand how trace IDs should be handled and integrated into their services.

# Positive:

- Improved correlation: With a consistent trace ID, it's much easier to correlate logs, metrics, and traces, leading to a more unified and comprehensive view of system behavior.
- **Faster incident resolution**: Consistent trace IDs enable quicker identification of the root cause of performance issues or failures, drastically reducing troubleshooting times.
- **Better system visibility**: By consistently tracking requests through all services, teams gain complete visibility into the lifecycle of a request, making it easier to spot bottlenecks and optimize performance.
- Increased reliability: With proper trace ID propagation, teams can be more proactive in identifying issues and addressing them before they escalate, improving overall system reliability.

## Challenges:

- Initial setup and instrumentation: Implementing consistent trace ID propagation across all services can require significant effort, especially in large, legacy systems. Proper instrumentation and configuration need to be set up to propagate the trace ID.
- Data overhead: Propagating trace IDs across all services can increase the volume of trace data, which needs to be managed effectively to avoid excessive storage and processing costs.

• **Training and awareness**: Teams must be trained to understand the importance of trace IDs and how to implement and maintain trace ID propagation correctly. Without proper training, trace IDs might be inconsistently used.

1. Antipattern Name: Starting the Trace at the API Gateway is Overrated

2. Category: Traces

- **3. Context:** This antipattern occurs when organizations rely solely on the API Gateway as the starting point for tracing requests across their system. The API Gateway often serves as the entry point for client requests, and many systems instrument it to start the trace when a request first hits the gateway. However, this can be problematic because it doesn't provide complete visibility into the user journey or the internal workings of services that occur before or beyond the gateway. Tracing should ideally capture the entire flow of a request, including the context of intermediate services that handle the request after the gateway or before it reaches the API Gateway, such as load balancers, edge services, or even the client itself. By only starting the trace at the API Gateway, teams may miss important parts of the transaction, leading to incomplete visibility.
- **4. Problem:** When traces only start at the API Gateway, teams lose visibility into the full request flow prior to this point. This results in the loss of context for issues occurring before the API Gateway, such as DNS resolution failures, proxy problems, or networking issues. By not tracing the entire journey of a request, teams may miss performance bottlenecks or optimization opportunities that exist before the API Gateway. It also complicates root cause analysis, as errors or performance issues originating in earlier stages become harder to pinpoint. Additionally, it may lead to false assumptions about the critical touchpoints in the request flow, narrowing the focus to the API Gateway and neglecting other important parts.
- **5. Example:** A user submits a request to a web application that passes through an API Gateway before reaching the backend services. The API Gateway starts the trace and records its own logs. However, before the request reaches the API Gateway, it goes through an internal load balancer that occasionally drops requests under high load, causing intermittent failures. Since the trace starts at the API Gateway, the failure in the load balancer isn't captured, leading the engineering team to spend unnecessary time troubleshooting the backend services, unaware of the issue occurring earlier in the process.

#### 6. Recommended Solution

• Start the trace earlier: Instead of starting the trace solely at the API Gateway, ensure that traces are initiated as early as possible in the request lifecycle. This can include starting the trace at the load balancer, client-side, or any other relevant components involved in request handling.

- **Instrument all relevant layers**: Ensure that not just the API Gateway, but also load balancers, edge services, and client applications are properly instrumented to propagate trace IDs across the entire request flow.
- **Use client-side tracing**: In addition to server-side tracing, consider implementing client-side tracing (e.g., in web browsers or mobile apps) to capture the full journey of a request from the user's device all the way through the backend.
- Capture early-stage failures: Ensure that you can trace problems that occur before the API Gateway, such as networking issues, failures in reverse proxies, or issues with upstream services, by instrumenting these layers as well.
- Leverage tracing systems that support full request propagation: Utilize distributed tracing tools (e.g., OpenTelemetry, Jaeger, Zipkin) that can trace a request across multiple layers and services, capturing the entire journey of the request, not just the parts it hits after the API Gateway.

#### Positive:

- **Complete visibility**: By starting the trace earlier in the request flow, you gain visibility into all components that influence the request, from the client to the backend services. This provides a more holistic understanding of system performance.
- Faster root cause analysis: Tracing across the entire journey allows for faster identification of problems, even those that occur before the API Gateway, leading to quicker incident resolution.
- More accurate performance metrics: Full request tracing allows you to measure performance more accurately across all components involved in the request, including the external-facing systems (e.g., load balancers or edge services).
- Proactive troubleshooting: With complete trace coverage, you can proactively identify
  and address issues at earlier stages of the request flow, preventing them from escalating
  and improving system reliability.

## Challenges:

- Increased complexity: Instrumenting all layers of the system, including client-side, API
  Gateway, and backend services, can increase the complexity of the tracing setup and
  instrumentation.
- **Data overhead**: Tracing the entire journey of a request can generate more trace data, which may require additional storage and processing resources. Careful management and filtering of trace data are necessary to avoid excessive load.
- **Tooling and integration**: Not all systems or components might be ready to support distributed tracing. You may need to invest in integrating various tools and technologies to ensure full trace propagation.