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Core Principles of SRE

- SRE is defined by a set of **key principles** and **practices** that **govern** how **reliability** is **managed** and **measured**.
- Site Reliability Engineering (SRE) is about **making systems reliable using engineering practices**.
- Instead of fixing problems manually, SRE teams build automation, monitoring, and tools to keep services running smoothly.
- SRE is defined by a software engineering approach to operations, **centered on three core principles** that **drive all decisions**:
 - Embracing Risk and Setting Reliability Goals
 - Elimination of Toil with Scripts and Automation
 - Monitoring Everything That Matters



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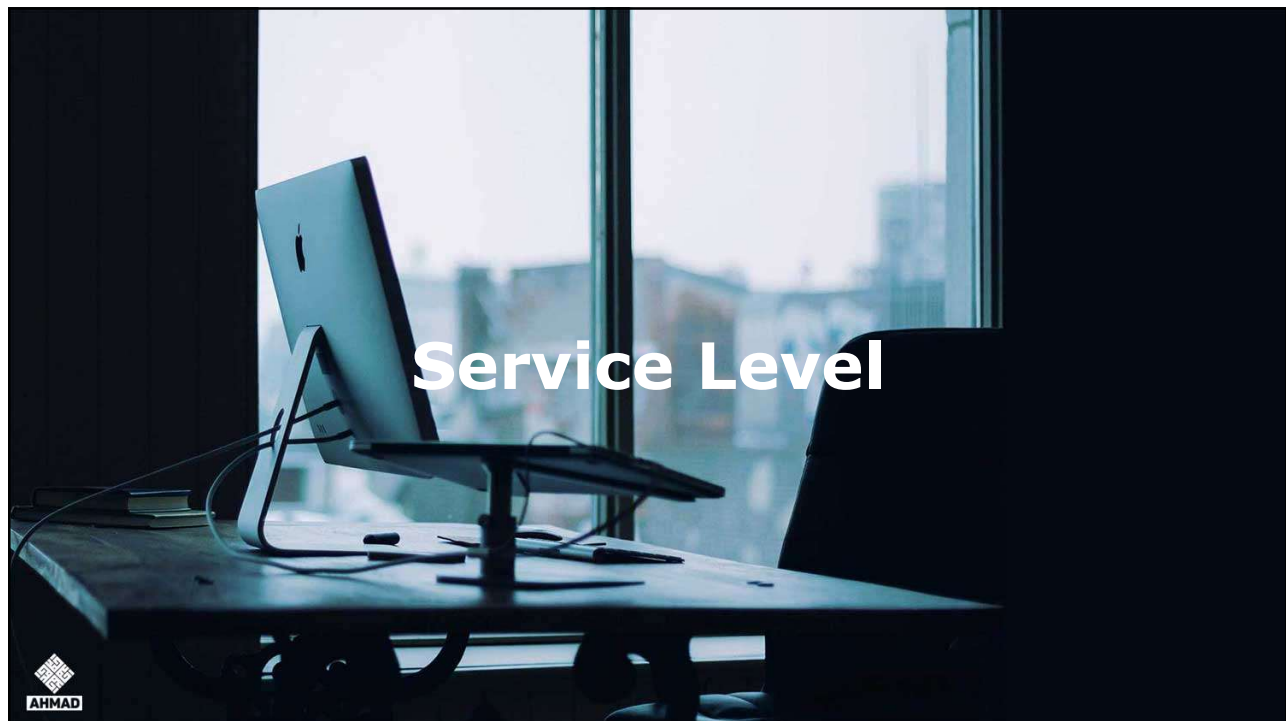
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Reliability Goals

- This is the most **philosophical break** SRE makes **from traditional operations**, which often aims for **100% uptime**.
- **SRE acknowledges** that **100% availability** is **impossible** — it is **too expensive**, **slows innovation**, and **often unnecessary**.
- **Reliability** is treated as an explicit, **measurable goal**, not a vague wish.
- SRE manages the **trade-off** between **Agility** (*speed of releases*) and **Reliability** (*stability*) using **SLOs (Service Level Objectives)** and **Error Budgets** to balance reliability and agility.
- Reliability in SRE is built on four pillars:
 - **SLI** (What we measure)
 - **SLO** (What we target - goa)
 - **SLA** (What we promise customers - legal/commercial)
 - **Error Budget** (How much failure is allowed)



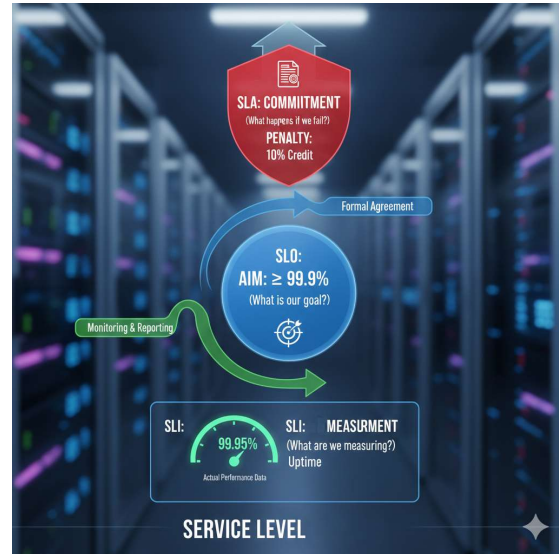
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Reliability Goals | Service Level

- A Service Level refers to the specific, measurable **performance standard** or target that is agreed upon for the **delivery** of a **service**.
- It essentially answers the question: "**How well and how quickly will this service perform?**"
- **The Interconnected Relationship**
 - Think of them as a **hierarchy of commitment**:
 - **SLI**: THE **MEASUREMENT** (*The actual percentage of uptime, e.g., 99.95%*)
 - **SLO**: THE **TARGET** (*We aim for 99.9% Uptime*)
 - **SLA**: THE **COMMITMENT** (*If we fail to meet the 99.9% target, we pay you a penalty*)



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Service Level | SLI

- **Service Level Indicators (SLIs)**
 - SLI is a **numerical measure** that tells **how reliable** the **system is**. The **raw data points** that tell you *what* the service is doing.
 - A **quantitative measure** of a **service's performance** (e.g., *99.99% of requests returned in under 100ms*). This detects risk.
 - SLIs **measure customer-impacting metrics**, such as:

| SLI Type | What It Measures | Example Metrics |
|-----------------------|---------------------------------------|---|
| Availability | Service uptime, success response rate | % of successful API calls (2xx) |
| Latency | Speed of response | % of requests completed under 200 ms |
| Error Rate | Failed or erroneous transactions | 5xx errors, failed login attempts |
| Throughput / Traffic | Requests handled per second | API requests/second, trades/sec |
| Saturation / Capacity | Resource exhaustion levels | CPU, memory, queue usage |
| Durability | Data integrity & loss protection | % of data loss-free operations |
| User Experience (UX) | Real user performance | App crash rate, page load time |
| Business SLI | Business/user success | % successful trades, payment success rate |



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Service Level | SLOs

- **Service Level Objectives (SLOs)**

- The agreed-upon **target reliability percentage** (e.g., *99.95% uptime*), based on SLI data.
- The explicit, **internal goal** for the reliability of a service.
- The SLO is an **internal contract** between the SRE team and the Development (Product) team.
- **Example**
 - The payment API should be 99.95% available over 30 days.
 - 95% of trades should complete within 300ms.
- It tells **how reliable** the **service SHOULD be**, not to be perfect, but **good enough for customers** at a **reasonable cost**.
- The SLO is the most crucial terms, as it **directly drives engineering behavior** and defines the Error Budget.



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Service Level | SLOs ... continue

- SLOs are categorized based on **what type of reliability** we are measuring.

| SLO Type | What It Measures | Example Metric | Example SLO Statement | Used For |
|---------------------------|---|---------------------------------------|--|---|
| Availability SLO | % of time service is available | Uptime %, Error rate | Service will be available 99.9% of the time per month | Service reliability, uptime commitments |
| Latency / Performance SLO | Speed of response | Response time (ms), Page load time | 95% of API calls should respond under 200ms | User experience, app responsiveness |
| Throughput SLO | Volume of successful requests per time | Requests/sec, Transactions per minute | System should handle 5000 orders per minute | Scalability, system capacity |
| Error Rate SLO | Frequency of failed requests | % failed requests | Less than 0.5% of API requests can fail in a week | Quality of service, reliability |
| Durability SLO | Protection of data from loss | Data loss % | Data durability of 99.99999999% (11 nines) per year | Data storage, backup, cloud services |
| Reliability SLO | Successful task completion | Successful job completion rate | 99.9% of scheduled jobs should finish successfully | Background jobs, workflows |



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Service Level | SLOs ... continue

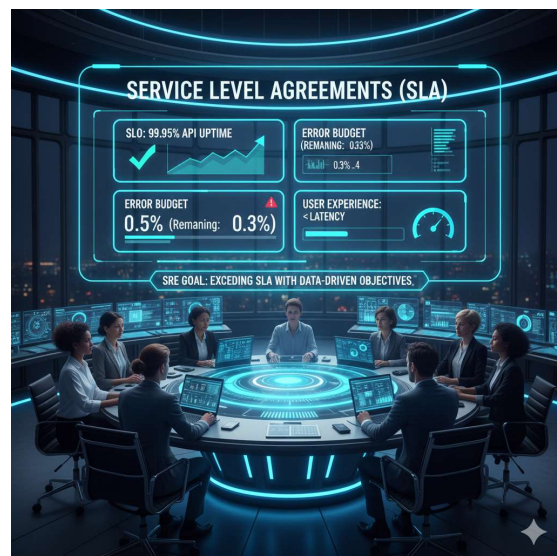
| SLO Type | What It Measures | Example Metric | Example SLO Statement | Used For |
|--------------------------------|-----------------------------------|-----------------------------|---|---------------------------------|
| Security SLO | Protection against threats | Vulnerability fix time | Critical security issues resolved within 72 hours | Security operations, compliance |
| Freshness / Data Staleness SLO | Timely data updates | Sync delay (mins), Data lag | Data replication lag must be under 5 minutes | Data pipelines, analytics |
| Capacity SLO | Handling load without degradation | CPU, RAM, storage usage | System must maintain < 70% CPU usage under peak load | Infrastructure scaling |
| Customer Experience SLO | Satisfaction & usability | CSAT, NPS, App rating | Maintain 90%+ customer satisfaction score | Business-level user experience |



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Service Level | SLA

- **Service Level Agreements (SLA)**
 - **Contract**
 - The **external, legal agreement** with the **customer**, outlining **penalties** for **failure to meet the SLO** (e.g., **99.5%** with a **10% credit** if missed).
 - **Buffer**
 - SLOs are typically set **tighter** than **the external SLA** to provide a **safety margin**, ensuring that if the team misses the **internal target** (the **SLO**), they still have a chance to fix it before breaching the **external legal contract** (the **SLA**).



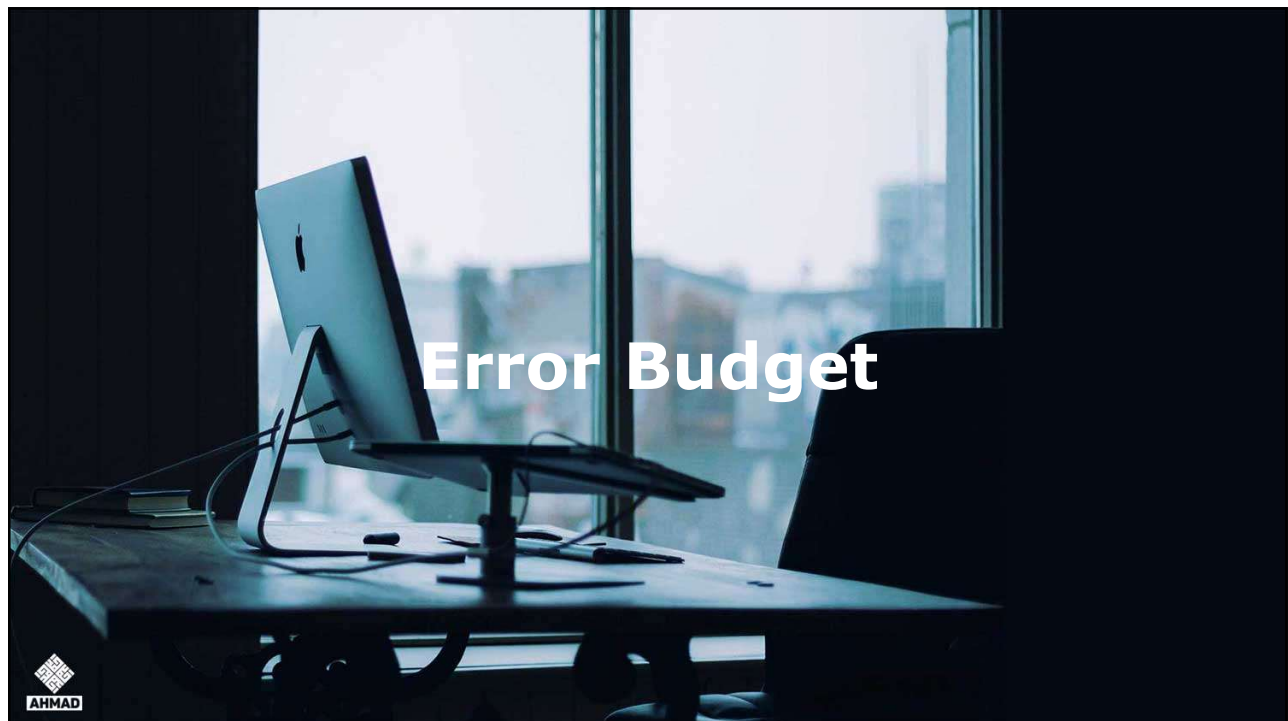
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Service Level | SLA ... continue

| SLA Type | What It Ensures | Real-Life Example |
|---------------------|-------------------------------|--|
| Availability SLA | Uptime guarantee | Server SLA: 99.99% |
| Support SLA | Response/resolution time | P1 issue responded in 15 min |
| Performance SLA | Response speed guarantee | 95% of transactions < 200ms |
| Data Protection SLA | Backup, retention, durability | 99.999999999% (11 nines) data durability |
| Compliance SLA | Security, regulation, audit | GDPR, PCI-DSS, RBI guidelines |
| Penalty-based SLA | Refund/credit if breached | Refunds 25% monthly bill |



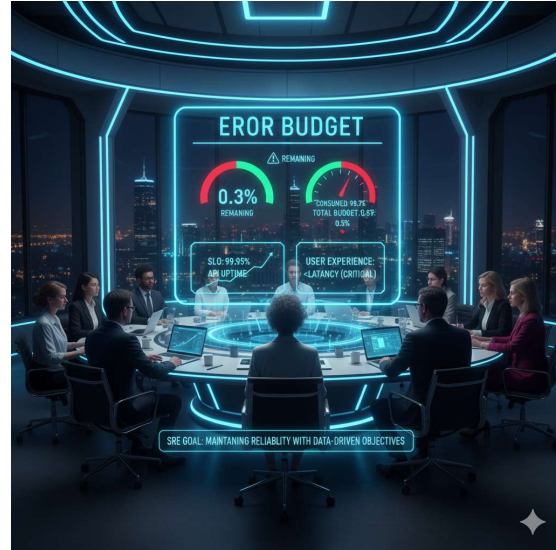
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What is Error Budget

- The **Error Budget** is the **maximum amount of time a service is allowed** to be **unavailable** or fail over a **specific period** (e.g., one month) without violating its **Service Level Objective (SLO)**.
- It is fundamentally a **governance tool** used to **create alignment between Development teams** (who prioritize feature speed) and **SRE teams** (who prioritize reliability and stability).



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Reliability Goals | Error Budget

- Error Budget**
 - The **acceptable amount** of time the **service is allowed to fail** (calculated as $100\% - \text{SLO}$).
 - Calculation**
 - The Error Budget is mathematically derived from the SLO and is **calculated** as the inverse of the SLO:

$$\text{Error Budget} = 100\% - \text{SLO}$$
 - Example:**
 - ✓ If your **SLO** (target reliability) is **99.9%** ("three nines"), your **Error Budget** is **0.1%**.
 - ✓ Over a **30-day period** (43,200 minutes), **0.1%** of the time **equals approximately 43 minutes** of acceptable **downtime** or error allowance.
 - Error budget formula:**

| SLO (%) | Error Budget (%) | Max Allowed Failure Time (monthly) |
|---------|------------------|------------------------------------|
| 99% | 1% | ~7.2 hours |
| 99.9% | 0.1% | ~43.2 minutes |
| 99.99% | 0.01% | ~4.3 minutes |
| 99.999% | 0.001% | ~26 seconds |



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Reliability Goals | Error Budget ... continue

- The Error Budget **provides** the SRE team with the necessary **authority** to manage risk:
 - **When the budget is full**
 - ✓ Dev teams can **move fast**, deploying new features frequently.
 - ✓ They are **embracing risk** because they have a **buffer**.
 - **When the budget is depleted (spent)**
 - ✓ All new **feature development** must **stop** (a feature freeze).
 - ✓ Engineering **resources must be redirected to reliability work** (*bug fixes, technical debt reduction*) **until the budget is replenished**.
- **Error Budgets** can be **applied in different ways** depending on what is being measured:

| Error Budget Type | Description | Example |
|----------------------------|------------------------------|---|
| Availability Budget | Allowed downtime | Max 43 minutes per month |
| Latency Budget | Allowed slow responses | No more than 1% calls > 500ms |
| Error Rate Budget | Allowed failed API requests | Up to 0.1% request failures |
| Data Loss Budget | Allowed inconsistency/drop | 99.999% durability => error budget 0.001% loss |
| Performance Budget | Allowed performance drop | 95% of orders processed under 3 seconds |
| Security Budget | Allowed exposure time / risk | Fix critical issues within 72 hours |



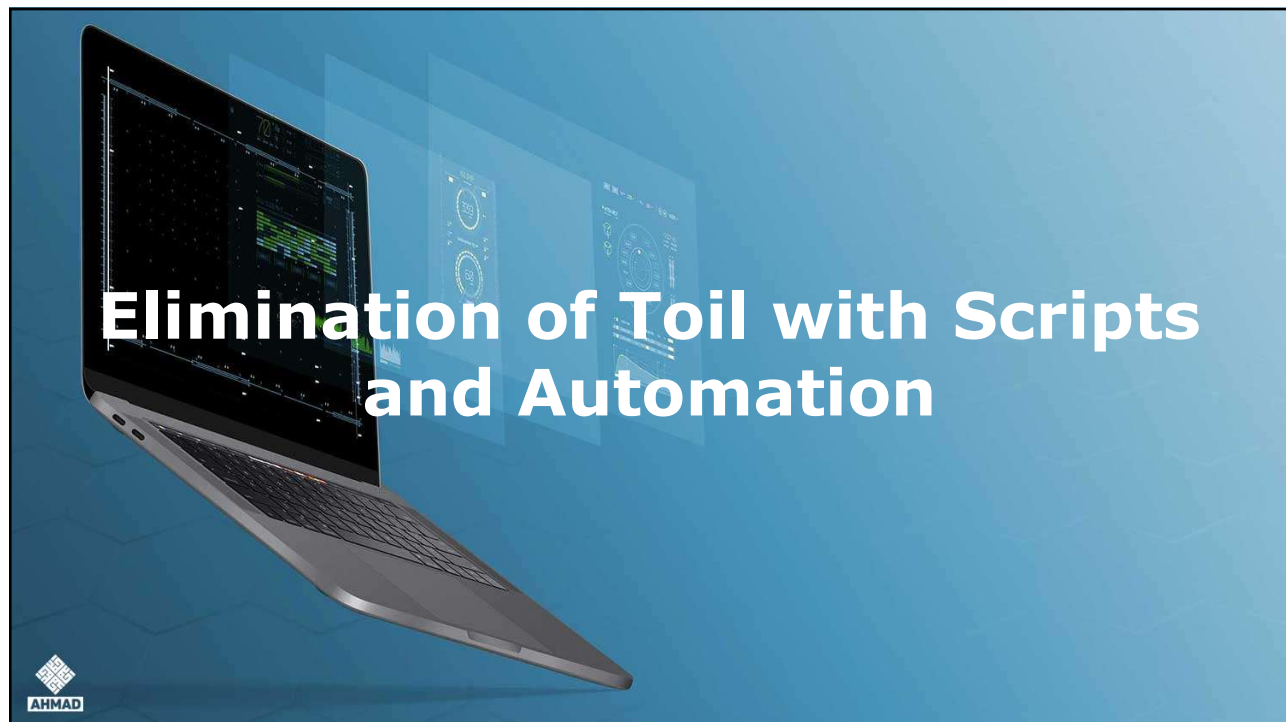
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Use Case | Stock Exchange

| Concept | Meaning | Purpose | Stock Exchange Real-Life Example | Measurement/Criteria |
|--------------------------------------|--|--|--|--|
| SLI (Service Level Indicator) | Actual measured performance | Shows "how system is performing" | <ul style="list-style-type: none"> • 98.5% of trades completed within 50ms; • 0.02% error rate | <ul style="list-style-type: none"> • Latency, uptime %, failure rate, data accuracy |
| SLO (Service Level Objective) | Target performance goal | Defines "how reliable system should be" | <ul style="list-style-type: none"> • 99.99% uptime during trading hours • <100ms trade processing time | <ul style="list-style-type: none"> • Based on business demand & traffic |
| SLA (Service Level Agreement) | Legal/business contract with customer | Defines penalties if reliability is broken | <ul style="list-style-type: none"> • If uptime <99.95%, stock exchange owes financial compensation to brokers/banks | <ul style="list-style-type: none"> • Uptime %, time-to-resolution • compensation terms |
| Error Budget | Allowed failure limit before reliability is affected | Balance reliability and innovation | <ul style="list-style-type: none"> • For 99.99% • SLO → allowed downtime = 4.3 min/month | <ul style="list-style-type: none"> • Error Budget = 100% - SLO |



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Automation & Engineering

- The **Elimination of Toil** is a foundational principle of Site Reliability Engineering (SRE).
- It is the mechanism SREs use to move from a **reactive firefighting role** to a **proactive engineering role**.
- **What is "Toil"**
 - **Repetitive, manual, ticket-based, operational work** that adds **no long-term value**.
 - **Examples of toil:**
 - Ticket processing
 - Log checking manually
 - Restarting failed services
 - Manual user provisioning



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Automation & Engineering ... continue

- SRE goal → **Automate repetitive work to free engineers for innovation.**
- **Characteristics of Toil**
 - **Manual** → Requires humans to perform it
 - **Repetitive** → Same process repeated frequently
 - **Reactive** → Triggered by alerts, tickets, or incidents
 - **No long-term value** → Does not improve the system
 - **Scale-dependent** → More workload, more manual effort
- **Key Practices:**
 - **The 50% Rule**
 - SREs must **spend no more than 50%** of their **time on toil**.
 - The other 50% is dedicated to Engineering (*writing code for automation, tools, and resilience*).
 - **Infrastructure as Code (IaC)**
 - Using tools like Terraform or Ansible to **manage infrastructure using code**, eliminating manual configuration.
 - **Automated Runbooks**
 - Turning manual incident **mitigation steps into scripts** that can be **triggered automatically**.



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Types of Toil in IT Operations

| Type of Toil | Description | Example in Industry |
|-------------------------------------|--|--|
| Monitoring Toil | Responding to repetitive alerts | Low CPU alerts at night |
| Deployment Toil | Manually deploying code/apps | Change requests executed manually |
| Configuration Toil | Manual server or firewall setup | SSH into each VM to update settings |
| Incident Toil | Handling same type of incidents repeatedly | Restarting crashed services daily |
| Reporting/Documentation Toil | Creating manual weekly reports | Manually compiling uptime metrics |
| User/Access Management | Manually provisioning accounts | Creating SSH access requests for engineers |



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How Automation Eliminates Toil

| Manual Operation (Toil) | Automated (Engineering-Based) |
|---|--|
| Manually restarting failed services | Self-healing automation using scripts |
| Logging into VMs to install patches | Automated patching via Ansible/Jenkins |
| Email-based change approvals | GitOps-based auto change control |
| Manual ticket creation after incident | Auto-ticket creation via ServiceNow API |
| On-call engineers checking alerts all night | Intelligent alert suppression and AI-based routing |



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Use Cases

- **Automated Payment Gateway Recovery (Bank)**
 - **Problem:** Payment system crashes frequently during peak hours → manual restart takes **30 minutes** → revenue loss.
 - **Automation Solution**
 - SRE creates an **auto-recovery script**:
 - ✓ Detects API failure
 - ✓ Restarts pods/servers automatically
 - ✓ Notifies engineers on Slack
 - **Manual time:** 30 mins → **Automated:** 1 min
 - **Result:** 80% incidents self-healed, zero customer impact.
- **Automated Trade Order Scaling (Stock Exchange)**
 - **Problem:** During market opening (9:15 AM), order load spikes 15x → website slow → manual scaling takes **1 hour**.
 - **Automation Solution**
 - **Auto-scale** Kubernetes clusters using CPU, QPS, and latency triggers
 - Introduced **auto-deployment verification check**
 - **Manual scaling:** 1 hour → Auto-scaling: **2 minutes**
 - Zero revenue loss during market hours.



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Tools Used for Toil Automation

| Purpose | Tools Used |
|---------------------------|--|
| Infrastructure automation | Terraform, Ansible |
| Monitoring & Auto-healing | Prometheus, Grafana , AlertManager, PagerDuty |
| CI/CD automation | Jenkins , Spinnaker, GitHub Actions |
| Incident auto-resolution | Runbook Automation (StackStorm) |
| GitOps Change Management | ArgoCD , FluxCD |



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Monitoring

- Monitoring is the continuous process of collecting, analyzing, and visualizing data (metrics, logs, and traces) about a system to understand its behavior, detect problems, and inform decision-making.
- The modern approach to monitoring and system understanding, called Observability, relies on three primary types of data, often referred to as the **Three Pillars**:
 - **Metrics** (What)
 - **Logs** (Why)
 - **Traces** (Where)
- **Events** are a related, specialized type of **record** often captured **within logs** or treated as a distinct data source.
- Together, these data types provide a complete picture, answering the questions: **What is happening? Where is it happening? and Why is it happening?**



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Metrics: The Quantitative View ("What")

- Metrics are **numerical measurements** that represent the **health** and **performance** of a system over time.
- They are the most efficient way to track trends and trigger alerts.
- Metrics are structured, **time-series data**: a **number collected** at a **specific timestamp** and enriched with descriptive labels (e.g., service name, region).
 - **Granularity: Aggregated**
 - **Metrics** are **summarized values** (*averages, sums, percentiles*) over a **time interval** (e.g., 99th percentile latency over the last minute).
 - **Purpose**
 - **Ideal** for **monitoring dashboards**, setting **Service Level Objectives** (SLOs), and generating **real-time alerts** when a threshold is breached (e.g., *CPU usage is 80%*).
 - **Examples**
 - Response time/**Latency**, **Request Rate**, **Error Rate**, **CPU/Memory Utilization**.



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Logs: The Detailed Narrative ("Why")

- Logs are **immutable, timestamped text records** of discrete events that occurred within an application or system. They are the system's diary, providing rich contextual detail.
- **Textual data**, often **unstructured** or **semi-structured** (like JSON), **generated** when **code executes** or a **system action occurs**.
- **Granularity: Granular and Contextual**
 - **Each line** typically **refers** to a **single event** with **specific details**, including the **exact state** of the **program at that moment** (e.g., *variable values, stack traces*).
- **Purpose**
 - Essential for **root cause analysis**, **forensic debugging**, and **auditing**.
 - When a metric alerts, SREs turn to the logs to find the "why."
- **Examples**
 - [ERROR] NullPointerException in payment service at line 45
 - [INFO] Database connection established
 - [WARN] Cache miss for user ID 12345.



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Traces: The Request Journey ("Where")

- Traces, or **Distributed Tracing**, map the **entire life cycle** of a **single request** or **transaction** as it **moves** through a complex, distributed **system** (*like microservices*).
- **Span**
 - A span is a **single, logical unit of work performed** by a **service** or **component**, such as:
 - An incoming or outgoing HTTP request.
 - A database query.
 - A specific function execution.
- A **trace** is a **collection of spans**. All **spans belonging** to the **same request** share a unique **Trace ID**.
- **Granularity: End-to-end journey**
 - They **link together** the **operations across service boundaries**.
- **Purpose**
 - Crucial for **diagnosing latency and performance bottlenecks** in microservices.
 - It **visually shows** which **service** is **responsible** for most of the **delay**.
- **SRE Use**
 - When a latency metric spikes, the trace for a slow request immediately points to **which service** in the call graph introduced the delay.



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Events: The State Change ("When")

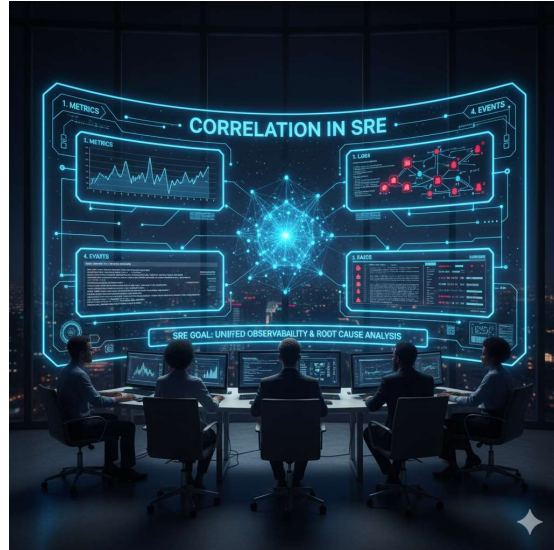
- While **often captured within logs** or used to derive metrics, **Events** are distinct occurrences that **represent** a significant, **discrete action** or **change of state** in a system.
- Structured records of key incidents or changes, **generated** by both **systems** and **humans**.
- They are generally of higher importance than routine log messages.
- **Granularity: Discrete Occurrence**
 - They **mark** a **specific moment** in **time**.
- **Purpose**
 - Primarily used for **correlation** and **context**.
 - They **answer** the question: "**What external action occurred right before the problem started?**"
- **Examples**
 - DeploymentStarted (version 3.1.2)
 - ServiceScaledUp (3 to 5 instances)
 - ConfigurationChanged (feature flag X disabled)
 - SecurityAlertTriggered



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The Value of Correlation

- The true power of Observability comes from **correlating** these data types.
- **For instance:**
 1. A Metric (**Request Rate**) **drops suddenly**.
 2. The SRE checks **Events** and sees that a **DeploymentStarted** event occurred seconds before the drop.
 3. The SRE then pivots to **Traces** for the **affected service**, seeing a **high failure rate**.
 4. Finally, the SRE drills into the **Logs** linked to the failed traces, finding the exact **stack trace** from the new code that caused the issue.
- This workflow **dramatically speeds** up the **Mean Time To Resolution (MTTR)**.



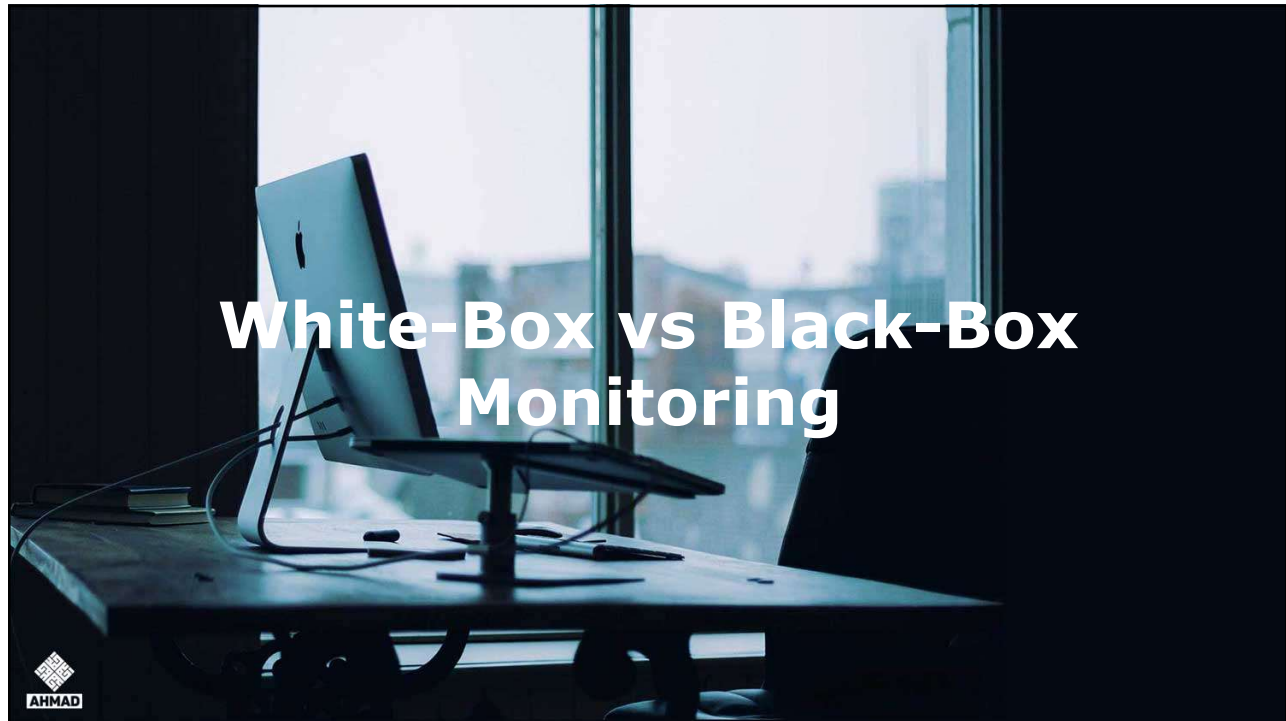
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Defining Meaningful Metrics and Alerts

- **Effective monitoring** starts with choosing **what** to **measure** and **deciding** when to **raise an alarm**.
- The goal is to maximize **signal** (*real, actionable problems*) and minimize **noise** (*irrelevant alerts*).
- **Meaningful Metrics**
 - Metrics are numerical, time-series data points that **quantify system performance**.
 - **Meaningful metrics align** with your **business goals** and directly **reflect the user experience**.
- **Effective Alerts**
 - An alert is a notification that a metric has crossed a predefined threshold or pattern, indicating a problem that requires attention.



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White-Box vs Black-Box Monitoring

- These **two concepts** describe the perspective from which you **monitor a system**. A **complete monitoring strategy** typically **combines both**.
- **Black-Box Monitoring**
 - Black-box monitoring **treats the system** as an **opaque**, closed-off entity—a "**black box**".
 - It is symptom-oriented and **checks externally visible behavior**.
 - **Perspective**
 - ✓ The **end-user's point of view**.
 - **Data Source**
 - ✓ **External probes**, **synthetic transactions** (*simulated user requests*), and **external system calls** (e.g., *ping*, *HTTP checks*).
 - **Focus**
 - ✓ **What is broken?** (*Is the service available, fast, and correct?*)
 - **Example Metrics**
 - ✓ **Uptime/Availability** (e.g., *is the login page returning a 200 status code?*).
 - ✓ **External Latency** (*how long a complete user transaction takes from outside the network*).
 - ✓ **DNS Resolution Time** from a global location.
 - **Purpose**
 - ✓ To confirm the **Service Level Objective (SLO)** — the promised level of service — is being met.
 - ✓ It ensures that **regardless of internal health**, the **system is working for the customer**.

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White-Box vs Black-Box ... continue

- **White-Box Monitoring**

- White-box monitoring **uses information exposed** by the **internals** of the **system** — it **looks inside** the "white box".
- It is **cause-oriented** and requires **direct instrumentation** of the **application code** or **host operating system**.
- **Perspective**
 - The **system's internal** components and code.
- **Data Source**
 - **Application logs, custom metrics** exposed by the **code** (*instrumentation*), and **operating system/agent** data.
- **Focus**
 - **Why is it broken?** (*What is the root cause: high memory, database lock, slow function call?*).



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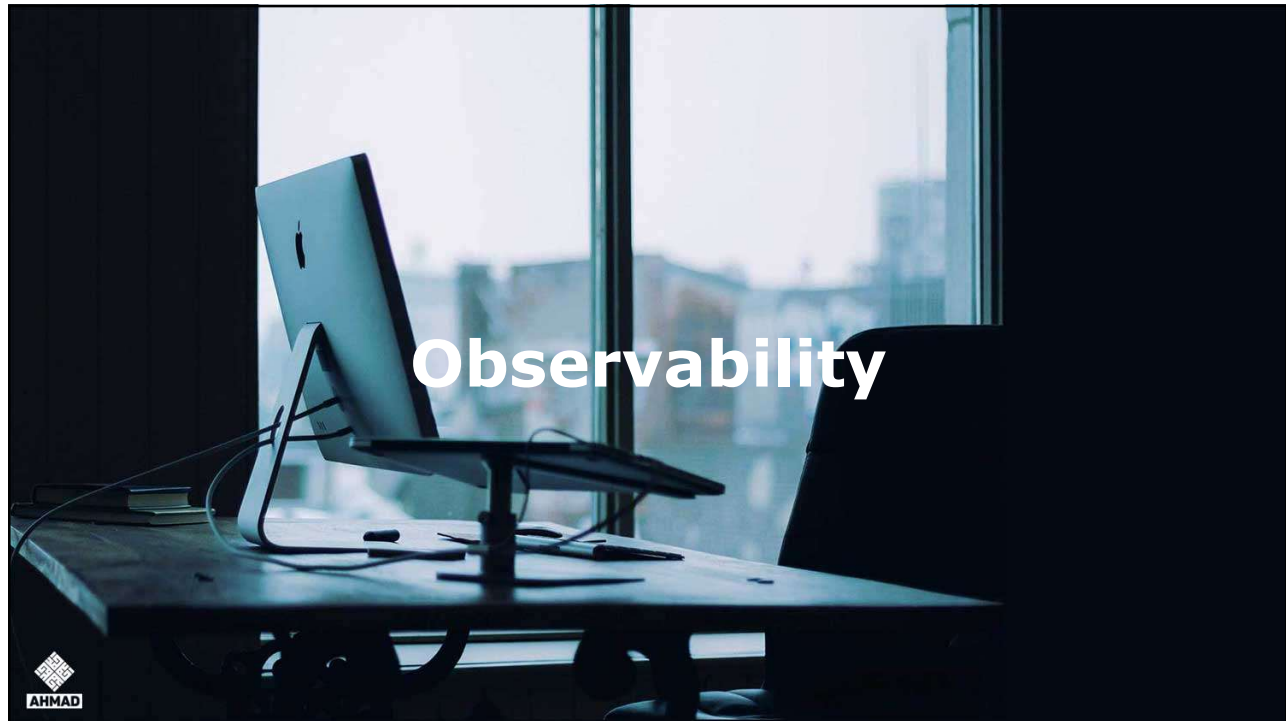
White-Box vs Black-Box ... continue

- **Example Metrics**
 - **Internal CPU/Memory/Disk Usage** of a specific server.
 - Garbage Collection **frequency** and **duration**.
 - **Internal Queue Lengths** or **Thread Pool Utilization**.
 - **Specific function execution times** measured within the application code.
- **Purpose**
 - To provide the granular detail needed for **debugging, capacity planning, and proactive issue detection** before a complete user-visible failure occurs.

| Aspect | Black-Box Monitoring | White-Box Monitoring |
|---------------|---|--|
| View | External (User's Perspective) | Internal (System's Perspective) |
| Focus | Symptoms (Availability, Latency) | Causes (Resource Utilization, Code Logic) |
| Action | Reactive (Service is <i>currently</i> down/slow) | Proactive & Debugging (Service <i>may</i> go down; find the root cause) |



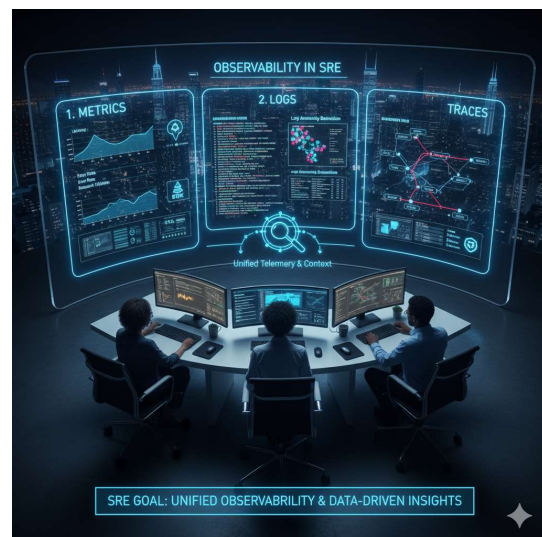
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Observability

- SRE recognizes that traditional monitoring (*checking if a server is alive*) is insufficient for complex distributed systems.
- You must monitor what the user *experiences*.
- The shift from generic monitoring to **Observability**, which means collecting and analyzing data that allows engineers to ask novel questions about the system without knowing the answer beforehand.
- Unlike **traditional monitoring**, which answers "*Is the system up?*", **Observability answers** "*Why is the system slow right now, and which component is causing it?*"



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Observability | Principles

- SREs prioritize **monitoring** the system based on the **user experience** and **business impact**.
- The **Four Golden Signals** are the definitive **set of metrics** that **must** be **tracked** for every user-facing service:

| Signal | Definition | Why it Matters (User Impact) |
|-------------------|--|---|
| Latency | The time it takes to serve a request . | High latency means a slow user experience , frustrating customers . |
| Traffic | A measure of how much demand is being placed on your service (e.g., requests per second). | Used for capacity planning and determining if a latency spike is due to high load . |
| Errors | The rate of requests that fail (explicitly via 5xx errors or implicitly via bad data). | Direct measure of user-facing failure . Every error spends the Error Budget . |
| Saturation | How full the service is; a measure of system utilization (<i>CPU, memory, disk I/O, queue depth</i>). | Predicts future failure . High saturation means the system is about to degrade performance or crash. |



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Observability | Pillars

| Pillar | Purpose | What It Helps Identify | Tools & Tech | Real Use Case (Industry) | Benefits |
|------------------------|--|---|--|--|--|
| Metrics | Collect numeric performance indicators over time | Resource usage , performance trends, saturation (capacity), success / failure ratios | Prometheus, CloudWatch, Datadog, Grafana | Stock Exchange: Monitor matching engine latency & CPU usage to ensure trade orders execute under 50ms | Proactive detection, trend analysis, capacity planning |
| Logs | Record detailed events, system messages, and error details | Root cause of failures , system bugs, security attempts, transaction history | ELK (Elastic, Logstash, Kibana), Splunk, Fluentd | Banking: Track failed payment logs, fraud detection logs, ATM withdrawal issues | Helps in debugging, audit trails, compliance, RCA |
| Traces | Follow a request across microservices (end-to-end tracking) | Identify bottlenecks , slow microservices, API dependency issues | OpenTelemetry, Jaeger, Zipkin | E-Commerce: Track order from cart → payment → inventory → shipping service and find where delay happens | Deep visibility into distributed systems |
| Events / Alerts | Notify when thresholds or anomalies occur | Unexpected failures, security breaches, system outages | PagerDuty, Prometheus Alert Manager, Opsgenie, | Healthcare System: Alert if medical record system response time exceeds 2 seconds | Real-time detection & response |



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Observability | Pillars ... continue

| Pillar | Purpose | What It Helps Identify | Tools & Tech | Real Use Case (Industry) | Benefits |
|---|---|---|---|---|--|
| Business KPIs (Service Health) | Track business impact instead of just system health | Customer loss, transaction drop , revenue impact, churn rate | Datadog, New Relic, Google Analytics, Dynatrace | Stock Exchange: Number of failed trade settlements or delayed transaction confirmations | Bridges business & technical metrics, aligns SRE with business goals |
| User Experience Monitoring (RUM & Synthetic) | Measure real user interactions & experience | Page load time , mobile app crash, sign-in failure | New Relic, Dynatrace, Datadog RUM, Pingdom | Online Banking: Detect when customers struggle with login or fund transfer due to UI lag | Captures true user experience; helps identify frontend issues |
| Security Monitoring | Monitor threats , suspicious access, vulnerabilities | Unauthorized access , DDoS, fraud detection | SIEM, Splunk Enterprise Security | Banking: Detect brute force login attempts or fraudulent ATM transactions | Protects reliability, compliance, trust |
| Infrastructure Monitoring | Monitor servers, network, storage | VM crash, disk full, network latency | Nagios, Zabbix, Datadog Infra | Telecom Industry: Monitor 5G tower network connectivity, loss, signal strength | Ensures platform stability and environment reliability |



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Observability | Benefits

| Benefit | Explanation |
|--|--|
| Faster Incident Detection | Reduces MTTR (Mean Time to Recovery) |
| Predictive Reliability | Detect failures <i>before</i> they impact customers |
| Prevents SLA Breaches | Helps stay within error budget |
| Enables Auto-Remediation | Integrates with tools like StackStorm |
| Helps SRE, DevOps, and Business | Unifies technical & business insights |



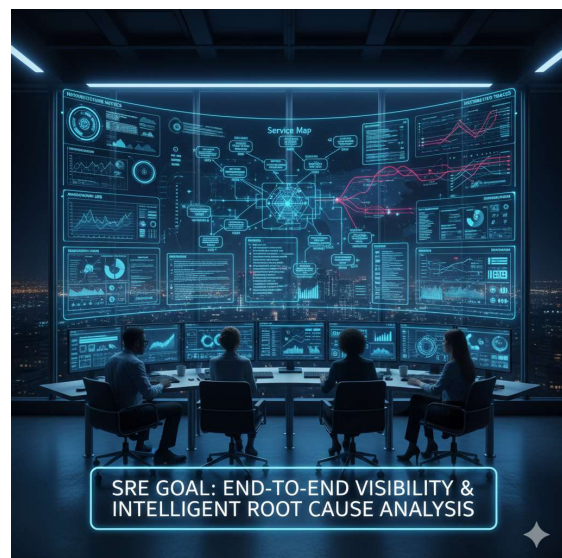
44



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What is Full-Stack Observability

- Full Stack Observability provides **deep, holistic visibility** into the performance and behavior of an application, from the **end-user interaction** all the way down to the **underlying network** and **infrastructure**.
- It integrates the three pillars of observability: **Metrics, Logs, and Traces**, across every component of the system.



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Beyond Infrastructure

- **Full Stack Observability** ensures **comprehensive monitoring** across three critical layers that are **often siloed** in **traditional monitoring** setups:
 - **Application (App) Visibility**
 - Focuses on the **code** and **services** that make up the **application** (e.g., *microservices, APIs, databases*).
 - **Key Data: Traces, Metrics and Logs.**
 - **Network Visibility**
 - Focuses on the **data flow between application components**, users, and the cloud/data center.
 - **Key Data:**
 - **Network latency, packet loss, bandwidth utilization, firewall performance, and load balancer health.**
 - This is crucial for modern distributed systems where network slowness can often be mistaken for application issues.



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Beyond Infrastructure

- **User Experience (UX) Visibility**
 - Focuses on the actual experience of the end-user, often through **Real User Monitoring (RUM)** and **Synthetic Monitoring**.
 - **Key Data:**
 - **Frontend Performance**
 - **Page load times**, core web vitals (*Largest Contentful Paint, Cumulative Layout Shift*), JavaScript **errors**, and **interaction latency**.
 - **Business Impact**
 - **Conversion rates**, cart abandonment rates, and **other key business metrics** tied to performance.



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Data for Root Cause Analysis (RCA)

- The true power of **Full Stack Observability** lies in its ability to **correlate data across** the entire **stack**.
- **Enables Better RCA:**
 - **Unified Data Platform**
 - All **Metrics, Logs, and Traces** are ingested and linked within a **single platform**.
 - **Contextual Tracing**
 - A **single user request** is **tracked** (via a Trace ID) from the user's browser, through the load balancer, across multiple microservices, and **down** to the **database query**.
 - If a database query is slow, the trace immediately highlights the specific service and the offending query, linking it to associated error logs.
 - **Automatic Correlation**
 - The platform **automatically identifies dependencies** and **anomalies**.
 - For example, if **Application Metrics** show a **spike** in **errors**, the FSO system can automatically **check if Network Metrics** show **increased latency** or if Infrastructure Metrics show a **spike** in **CPU utilization** at the exact same time.
 - **Faster MTTI and MTTR**
 - This correlation dramatically **reduces** the **Mean Time To Identify** (MTTI) the component causing an incident and the **Mean Time To Resolve** (MTTR) it, as SREs spend less time navigating tool silos.



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Balancing Velocity with Reliability

- The core purpose of an **Error Budget** is to serve as a **risk regulator**, helping **balance** the ongoing tension **between** moving **quickly** (*agility and velocity*) and maintaining **stability** (*reliability*).
- **Feature Velocity**
 - **Developers** are incentivized to **release features quickly** and often, which inherently introduces risk.
- **Budget as a Buffer**
 - The Error Budget acts as a buffer.
 - As long as the **budget is healthy** (full), the teams can maintain **high velocity**, knowing they have room to absorb minor bugs or failures.
- This mechanism ensures that reliability commitments are **enforced by data** rather than endless meetings or subjective arguments.



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Drive Engineering Decision

- The Error Budget is a powerful tool for strategic prioritization:
 - **Prioritizing Toil Elimination**
 - SREs can **justify spending time** on **automation** by calculating the potential **budget savings**.
 - **Example**
 - If **manual configuration** causes one **10-minute outage per month** (*spending 10 minutes of the budget*), **automating** it **permanently justifies** taking one week off feature work to write the automation code.
 - **Guiding Test Coverage**
 - If a **critical service** has a **very tight SLO** (*e.g., 99.999%*), its **Error Budget** is **minuscule**.
 - This drives an engineering decision: **Invest heavily** in **automated testing** and staging environments to **ensure changes** are virtually **risk-free before deployment**.



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Drive Engineering Decision ... continue

- **Informing System Design**
 - Services with **large, unused budgets** are **prime candidates for de-scoping** (*reducing over-engineering*).
 - **Services** that **constantly burn their budget** need urgent investment in **redesign** or **architectural** changes to **improve resilience**.
- **Blameless Learning**
 - **After** an **incident**, the post-mortem **focuses** on **how much budget** was **spent** and **why**.
 - The key action item is always an engineering task to **prevent** the **budget-spending** incident from recurring, thus driving a culture of continuous improvement.



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Help in Decisions

| Scenario | Decision |
|-----------------------------------|---|
| Budget mostly unused | Approve new releases , experiments, A/B testing |
| Budget moderately used | Approve with caution , implement canary deployment |
| Budget nearly exhausted | Change freeze , focus on performance & bug fixes |
| Budget completely consumed | Block deployments , trigger high-priority reliability review |



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SRE Roles and Responsibilities

- An SRE is a software engineer who possesses deep operational knowledge.
- Their time is rigidly split between managing the current system and engineering future improvements.

| Responsibility | Traditional IT Ops Approach | SRE Approach |
|--------------------------------|----------------------------------|--|
| Monitoring | Server uptime, CPU, memory | Service health, SLO, SLI, latency, error rate, business impact |
| Incident Management | Reactive resolution | Proactive prevention , blameless postmortem, error budget-based decisions |
| Change Management | Manual approvals, CAB meetings | Automated CI/CD-based GitOps, canary, blue/green deployments |
| Troubleshooting | Manual logging, ticket-based | Automated root-cause analysis, observability, distributed tracing |
| Capacity Planning | Static provisioning | Auto-scaling , performance benchmarking, chaos engineering |
| Toil Handling | Manual repetitive tasks | Automation -first approach (Terraform, Ansible, scripts, AI Ops) |
| Reliability Engineering | NA (mostly infrastructure focus) | Error budgets, SLO/SLA , reliability testing, chaos engineering |



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SRE Team

- An **SRE (Site Reliability Engineering) team** is a specialized group of engineers who focus on **making systems reliable, scalable, efficient, and resilient** — using **software engineering, automation, observability, and proactive problem-solving** rather than manual operations.
- Types of SRE Engineers:**

| SRE Role | What Work They Perform (Responsibilities & Tasks) | Tools & Technologies |
|--------------------------|--|--|
| Infrastructure SRE | Manages cloud and hybrid infrastructure , Kubernetes, IaC, network, provisioning, reliability engineering, cost optimization | AWS, Azure, GCP, Kubernetes, Terraform, Ansible, Helm, VMware, OpenStack, Ansible, Bash/PowerShell, Nagios, SolarWinds, Zabbix, NetApp |
| Platform SRE | Builds internal developer platforms , CI/CD, Observability, Secrets, Service Catalog, GitOps, SSO, API gateway | GitHub Actions, Jenkins, ArgoCD, Vault, ServiceNow, Grafana, ELK |
| Product/ Application SRE | Ensures reliability of user-facing apps (payment, mobile banking, e-commerce, trading), sets SLO/SLIs, monitors production health, chaos testing, RCA | Datadog, Splunk, Jaeger, PagerDuty, New Relic, Chaos Monkey |
| Security SRE (Sec-SRE) | Threat monitoring, vulnerability management , automated patching, compliance, IAM, DDoS protection, security event automation | Splunk SIEM, Palo Alto, Cloudflare, Qualys, AWS GuardDuty |
| ML/AI SRE (MLOps) | Monitor reliability of ML models , automate retraining, manage data pipelines, ensure model accuracy & performance | Kubeflow, MLflow, TensorFlow, SageMaker, Kafka |
| Database SRE | DB availability , replication, backup, auto-recovery, sharding, query tuning, HA clusters, DR testing | Oracle RAC, PostgreSQL, Cassandra, MySQL, MongoDB, AWS RDS |



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How SRE is Structured in Organizations

- 3 Major SRE Structures:**

| SRE Structure Type | Description | Company Examples | When to Use |
|--------------------------------------|---|------------------------------|---|
| Embedded SRE | SRE engineers directly work inside product teams | Google Search, Amazon Prime | Complex large-scale products needing continuous reliability |
| Central SRE / Shared SRE Team | Single SRE team supports multiple applications | Banking, telecom, SaaS | Medium-scale apps, central governance |
| Platform SRE | Works on common tools , CI/CD, infrastructure, observability, SLO frameworks | Netflix, Spotify, Salesforce | When building developer experience or DevOps platform |



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What Skills Do SRE Engineers Need

| Skill Category | What SRE Should Know | Example |
|---------------------------------------|---|---|
| Software Development | Python, Go, Shell scripting, Git | Build automation, monitoring scripts |
| Cloud & Infra | AWS, Azure, GCP, Kubernetes, Terraform | Deploy Google Kubernetes, scale AWS infra |
| Automation & DevOps | CI/CD (Jenkins, GitHub), Ansible, GitOps, IaC | Automated deployment |
| Observability & Monitoring | Prometheus, Grafana, ELK, Splunk, Datadog | Track latency, error rates, business KPIs |
| Reliability Concepts | SLO, SLI, SLA, Error Budget, Blameless Postmortem | Decide when to stop deploying |
| Chaos Engineering | Gremlin, Chaos Monkey | Test failure recovery |
| Incident & Problem Solving | Root Cause Analysis, self-healing | Analyze high CPU issue, create fix |
| Performance & Scalability | Load testing, auto-scaling | Ensure 1M users during sale |
| Soft Skills | Collaboration, communication, problem-solving | Talk to developers and business |



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Roadmap to Convert Traditional IT Ops to SRE

| Phase | Goal | Activities |
|--|------------------------|--|
| Assessment | Identify skill gaps | Evaluate current Ops skills, mindset, tooling |
| Training | Build new capabilities | Train on Python, Cloud, CI/CD, Monitoring, SRE fundamentals |
| Automation First | Reduce toil | Transition from manual tickets to automation & scripts |
| Introduce SLO / SLA / Error Budgets | Reliability focus | Implement reliability metrics, observability dashboards |
| Create SRE Squads | Start hybrid model | Mix Dev + Ops + SRE together (Platform, Infra, Product SRE) |
| Full SRE Adoption | Establish culture | Blameless postmortems, continuous improvement, chaos testing |



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Ops to SRE Transformation

| Before SRE (Ops Team) | After SRE (Reliability Engineering Team) |
|------------------------------------|---|
| Manual monitoring of servers | Automated trade monitoring using latency, TPS, failure rate |
| Ticket raised for payment failure | Script auto-restarts payment service (self-healing) |
| Outage analysis done manually | SRE uses distributed tracing (Jaeger) to find bottleneck |
| Change approval in email | GitOps-based approval and automated deployment |
| Manual scaling during "Salary Day" | Auto-scale compute based on trade volume |



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Employee Identity Service (EIS)

- Your organization (**FinTech Corp**) has an **Employee Identity Service (EIS)** responsible for:
 - Employee **authentication** (login)
 - **Authorization** (role-based access control)
 - **Profile retrieval** (name, designation, department)
- This service is **mission-critical**, powering:
 - **Payroll & HR Portal** → Salary slips, leave & appraisal
 - **Finance Approval System** → Expense approvals
 - **IT Service Desk** → Asset provisioning & access requests
 - **Onboarding Platform** → New employee setup
- **Business Expectation**
 - Employees should be able to access portals securely, quickly, and reliably without login issues, especially during salary processing, appraisal cycles, and onboarding.



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Step 1: Define Possible SLIs

| SLI Type | Example Metric | Description |
|--------------|----------------|-------------|
| Availability | | |
| Latency | | |
| Correctness | | |
| Error Rate | | |
| Freshness | | |



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Possible SLIs

| SLI Type | Example Metric | Description |
|--------------|------------------------------------|---------------------------------------|
| Availability | % of successful login requests | Measures uptime/access success |
| Latency | % login responses < 500ms | Measures login response speed |
| Correctness | % correct identity responses | Ensures data accuracy (role, name) |
| Error Rate | % failed login attempts (5xx) | Tracks failures due to service issues |
| Freshness | Time since last user update synced | Measures how current data is |



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Step 2: Define Possible SLOs

| SLI | SLO (Target) | Error Budget |
|------------------------|--------------|--------------|
| Availability | | |
| Authentication Latency | | |
| Error Rate (5xx) | | |
| Correctness | | |
| Freshness | | |



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Possible SLOs

| SLI | SLO (Target) | Error Budget |
|------------------------|--|---------------------------------------|
| Availability | 99.5% monthly | 0.5% downtime = 216 mins/month |
| Authentication Latency | 95% requests < 400ms | 5% slow requests allowed |
| Error Rate (5xx) | <0.3% per month | 0.3% failure margin |
| Correctness | 99.9% accurate identity responses | 0.1% incorrect allowed |
| Freshness | 95% of syncs within 1 hour | 5% can exceed 1hr |



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Step 3: Define Possible SLA

| SLA Parameter | SLA Commitment |
|---------------------|----------------|
| Availability | |
| Login Response Time | |
| Identity Data Sync | |
| Incident Response | |
| Compensation | |



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Possible SLA

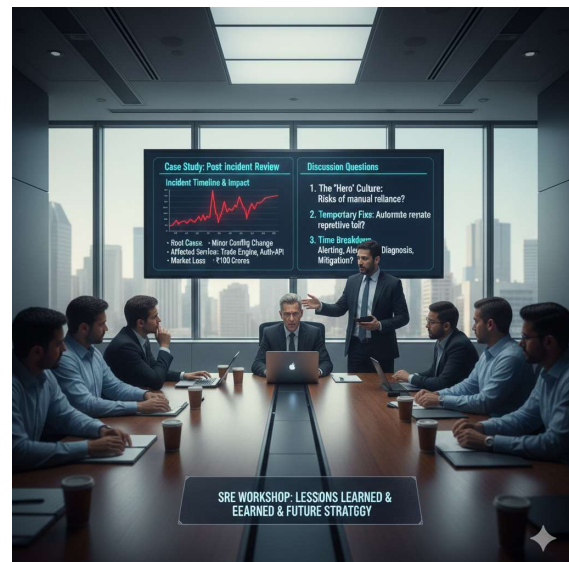
| SLA Parameter | SLA Commitment |
|---------------------|---|
| Availability | 99.0% quarterly (<i>max downtime: 1h 48m per month</i>) |
| Login Response Time | 90% of login responses < 500ms |
| Identity Data Sync | Max sync delay = 4 hours |
| Incident Response | Critical incidents resolved within 4 hours |
| Compensation | Service credits: e.g., 5% monthly subscription refund for downtime exceeding SLA |



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Discussion

- **Why SLO is stricter than SLA**
- **What happens if SLA is breached externally**
- **Error Budget vs SLA**



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Discussion | Answer

- **Why SLO is stricter than SLA**
 - Internal target (SLO) is higher to **ensure service reliability**
 - SLA is slightly lower to **protect contractual commitments** and allow operational tolerance
- **What happens if SLA is breached externally**
 - Financial **service credits** (refund, subscription adjustment)
 - **Priority support / escalation** for affected customers
 - Reputation & trust may be impacted → triggers postmortem and reliability improvements
- **Error Budget vs SLA**
 - **SLA: 99.9%** uptime
 - **Error budget: 0.1% allowed downtime** per month = 43 minutes of downtime allowed per month



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National Stock Exchange

| Traditional Ops Environment | Modern SRE Transformation |
|---|---|
| Manual monitoring via NOC | Full observability (<i>Grafana, Prometheus, AI alerting</i>) |
| 1–2 hours downtime during trading time | Self-healing failover in 2 minutes |
| 1000+ monthly tickets | 80% tickets eliminated via automation |
| Change freeze during market hours | Safe, automated deployments using canary |
| Problem Resolution = War-room | Resolution via Runbooks , automated fixes |



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Why Change Was Needed

- One **trading outage** caused a **₹100 Crore loss** and **reputation damage**.
- Analysts were **fixing** production **incidents manually**.
- **Deployments** were **done via scripts**, after midnight, with **risk and fear**.
- **Monitoring** was **limited** to **CPU & RAM** — no business metrics.



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How They Introduced SRE

| Phase | Step | SRE Activities |
|---------|----------------------------------|---|
| Phase 1 | Analyze | <ul style="list-style-type: none"> • Define SLOs (e.g., 99.99% uptime) • Identify most frequent failures |
| Phase 2 | Automate | <ul style="list-style-type: none"> • Incident detection → automated alerts • Self-healing scripts → restart crashed services |
| Phase 3 | Reduce Toil | <ul style="list-style-type: none"> • 70% tickets converted to automated jobs Password resets, patching, backup monitoring |
| Phase 4 | Build Reliability Culture | <ul style="list-style-type: none"> • Introduced blameless postmortems • Used error budget policy for release decisions |
| Phase 5 | Optimize | <ul style="list-style-type: none"> • Chaos testing, load simulators, active failover |



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Results of the SRE Transformation

| Before SRE | After SRE |
|--|--|
| 4 major outages/year | 0 major outages in 12 months |
| Recovery Time: 1–2 hours | Self-healing in under 3 minutes |
| 1000+ tickets/month | 80% reduction (automated resolution) |
| 6–8 people required for manual patching | Fully automated patch pipeline |
| No visibility into root cause | Blameless postmortems & RCA automation |



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Final Summary

| Traditional Ops | SRE Mindset |
|--|---|
| Block changes to maintain stability | Engineer systems to handle change safely |
| Reactive firefighting | Proactive & predictive reliability |
| Manual troubleshooting | Automation & self-healing |
| Ticket-based operations | Engineering-based operations |
| Support function | Strategic reliability engineering |



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Q & A

Any concepts still unclear?

Thank you for attending



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