SLX An Extended SLO Framework to Expedite Incident Recovery

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About Us

- Infra SRE
- Managing 60+ Kubernetes Clusters with 60K+ Nodes
- Other Infra Services (DNS, Gateway, Pubsub)
- Firefighting and 99.999% availability target

Incident recovery is hard.

Definition of incident is often ambiguous.

#Motivation 1

Complex interactions between system boundaries

#Motivation 2

Feeling lost in Grafana panels & dashboards

#Motivation 3

Intro Incident Lifecycle



Bleeding System misbehaves The black swan is always there.

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Investigation Pinpointing the pitfalls



Postmortem What went well, what went wrong and lucky? This is a culture thing.

?



Shortening the investigation time

#Goal

Too many metrics & logs & traces

#Challenge 1



SLO

Metrics building

Single-service perspective

SLO reporting

Good for detection

High cognitive load on cross-service investigation

#Challenge 2



How does an experienced SRE debug massive K8s pod creation failures?

> #Apiserver #500 #quota #network #connectivity #docker #registry **#DNS #3p-party operator** #ratelimit #release



Leveraging on SLO data for investigation

#Challenge 3



How many SLOs to pay attention to?

50 Components × 5 Key User Journeys × 3 Types of basic SLOs × 3 Different Execution Environment = ?? SLOs

SLX

Light-weighted SLO Management +Automatic Time-series Anomaly Detection and Correlation

SLX Framework SLX Intro

SII Service Level Indicator http_request_error_ratio # (dc, api) as tags **SLO** Service Level Objective max_over_time(http_request_error_ratio{api=~"alb"}[30d]) < 0.001</pre> SLA Service Level Agreement max_over_time(http_request_error_ratio{api=~"alb"}[30d]) < 0.01</pre> SIF Service Level Factor http_request_error_ratio{api=~"a",dc="foo"} processed_cpu_seoncds_count{api="a",dc=~"foo"} SID Service Level Dependency rpc_outgoing_request_error_ratio{api=~"a",rpc="bar"}



A denpendency graph







Locate the burning SLO with the **abnormal SLF**





Locate the burning SLO with the **abnormal SLF**





Graph traversal to find the time-correlated abnormal SLO dependencies







Design & Implementation

Design & Implementation Intro



- Cluster of definitions
- Unmanageable relations
- Constant changes



- SLI is the core
- SLI can be reused



SLI = metric = time seriesSLO = SLI + constraint



```
name: Pod Creation SLO
sli:
    name: Pod Creation Success Ratio
    type: SuccessRatio
    meas:
    error:
        name: Pod Creation Failures
        type: Simple
        meas: promql:increase(slo_pod_startup_result...
        total:
            name: Pod Creations
            type: Simple
            meas: promql:increase(slo_pod_startup_result...
```

objectives:

- name: Pod Creation Success Ratio 1h
 interval: 30d
 type: rolling
 - alerts:
 - name: TooManySLOFailuresInAnHour
 active: true
 - severity: p0
 - trigger:
 - type: ErrorBudget
 - threshold: 0.99
 - window: 1h
 - windowBudgetRatio: 0.03
 - minWindowBudget: 15

Design & Implementation Automation



- **Declarative API** •
- Automated Reconciliation

Submit our SLOs Let the Kubernetes take the wheel

Design & Implementation GitOps



- Commit history •
- Code Review •
- **Conflict Resolution** •

- Automated Sync
- Resource Management •
- Web App •





- **Reconciliation Loop** •
- Error Handling
- Deployments •



Design & Implementation GitOps



Design & Implementation Workflow





Config
 Commit
 Config

- 3. ???
- 4. Profit

Design & Implementation System Design





Design & Implementation Dashboards

















Design & Implementation What's Next



LO Autopilot

- No driving automation
- Manually controlled
- Alerts before system failure

L1 Autopilot

- Driver Assistance
- System failure analysis ullet
- Response Recommendations ullet

L2+ Autopilot

- Driving automation
- SLO as a reconciliation target
- No hands required •



Design & Implementation Anomaly Detection

Statistical regression Oľ Machine learning predicted value

Observation



A set of logical / numerical operations Prediction

V . **S** .

Threshold

computed from probability distribution or empirical value

Design & Implementation Anomaly Detection



Underestimate the importance of data quality.

#pitfall 1



Highest Cardinality Labels

Name	Count
container_id	4343722
uid	2870959
pod	2460040
key	1325130
pod_ip	860952
pod_name	735653
container_name	604847
container_sn	603127
container_hostname	601593
ip	372114

Watch out for curse of dimensionality.

#pitfall 2

Obsessed with (hyper-)parameters in the algorithm and ignoring useability.

#pitfall 3



Summary Key Take-Aways

 SLO is not a silver bullet, it takes time to iterate to become a universal language

 Automation and standardization make things scalable and manageable

 Put the user first: algorithms and machine learning should not steal the spotlight of user-experience

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