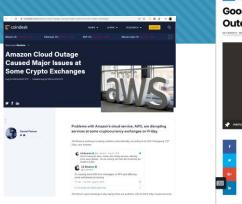
Fighting the Fog of War: Automated Incident Detection for Cloud Systems

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Background

- OReliability is a key quality attribute of large-scale cloud systems
- ○Incidents/outages dramatically degrade the service quality
- Tough incidents/outages take a long time to mitigate
- Costs: \$17K/outage·min (2016)*



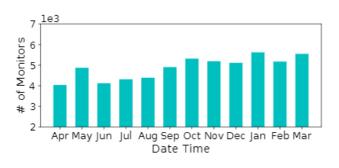


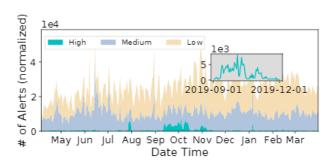


Alerts

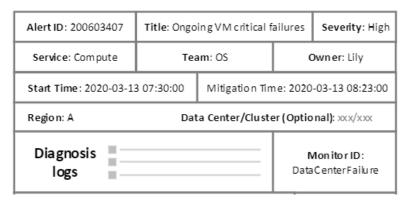
Alerts are system events that require attention

- Reported by the monitoring infrastructure
 - E.g., API timeouts, operation warnings, unexpected VM reboots, etc.
- Severity: low, medium, high
- Handled by on-call engineers





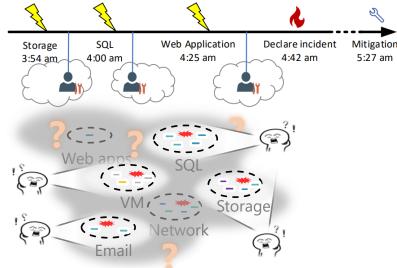
of monitors and alerts from Big5 services in Azure



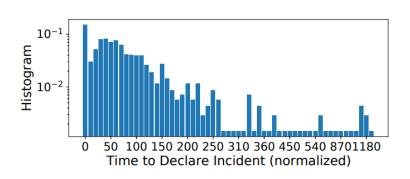
Main fields of an alert

Incident Management

- Incident: situations with customer impact, taking a long time to resolve, or requiring crossteam collaboration
- Timely incident management is the key to reduce system downtime
- Incident declaration turns chaos into order







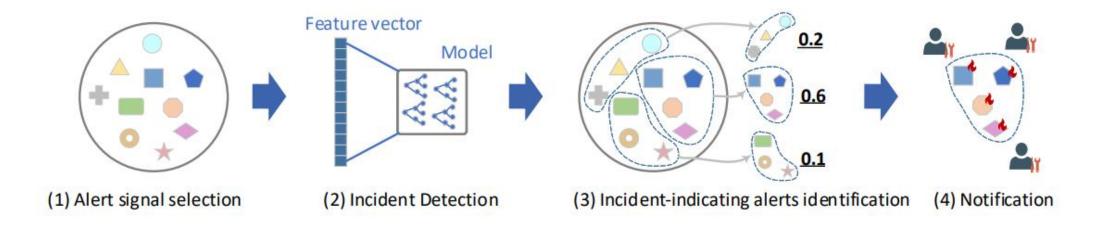
A long tail of incidents take a long time to declare

Related Work

- Fault detection and localization
- Time-series anomaly detection
- Cloud incident management

Warden: Automated Incident Detection

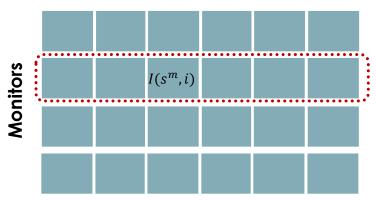
- Detect the ongoing potential incidents from the alerts
- Extract incident-indicating alert groups for notification



Alert Signal Selection

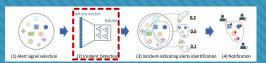
- Select a subset of monitors which exhibit relatively strong association with incidents
 - Categorize incidents based on their responsible teams
 - Calculate the sum of Weighted Mutual Information (WMI) for each monitor with all subtypes of incidents

Incident subtypes



 $I(s^m, i)$ is the information gain by observing alerts from a monitor m about predicting incidents of subtype I

The score of monitor m is $\sum_{i} I(s^{m}, i)$



Incident Detection

Incident detection: a binary classification problem

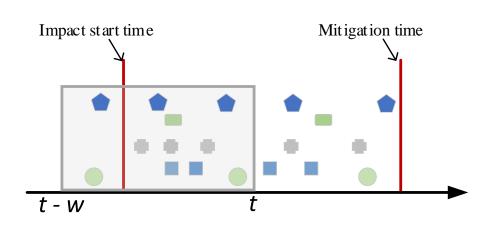
- Input: alerts reported by selected monitors in a recent time window
- Output: 1 if there is potential ongoing incidents; otherwise, 0

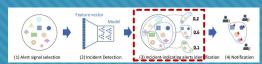
Sample construction

- Construct samples using a sliding window (3h)
- Label = 1 if the window is overlapping with incident impact duration; otherwise, label = 0

Feature extraction:

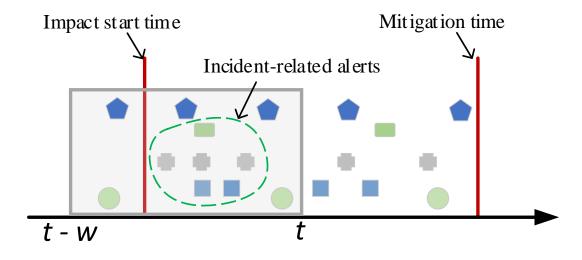
- Alert signals: alert count, alert burst
- Engineer activities: diagnosis log count, notification count
- Others: region, working day, hour in day
- Classification model: BRF (Balanced Random Forest)

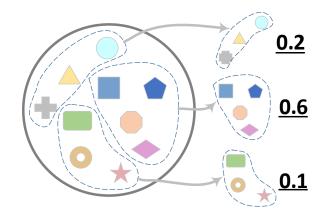




Identifying the Incident-indicating Signals

- Incident-indicating signals: alerts related to the incidents
 - Alert signal grouping: correlation and rule-based
 - Group-based model interpretation: GSV (Group Shapely Value)

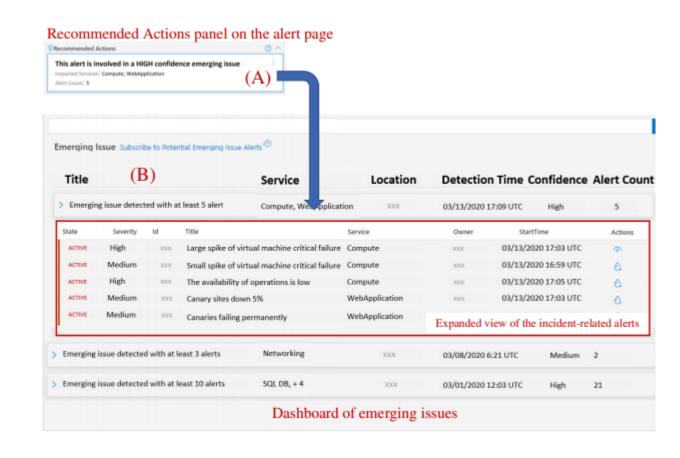






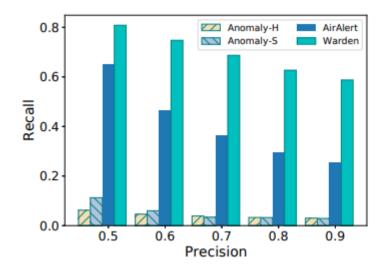
Warden in Practice

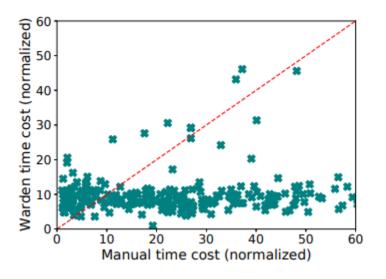
- Detecting emerging issues
- Notify all engineers working on incident-indicating alerts
- Once confirmed, engineers form a cross-team collaboration group to diagnose and mitigate the incident

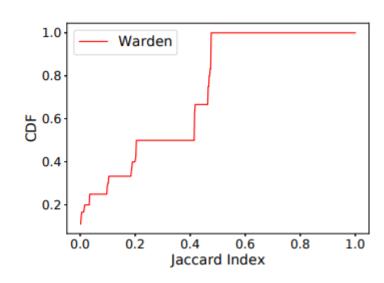


Experimentation

- 18-month-length ~240G data from Azure IcM
- 26 major services, Hundreds of people behind each service, ~72% of all incidents of Azure
- Training: 16 months; Testing: 2 months
- Baselines: Anomaly-H/S, AirAlert



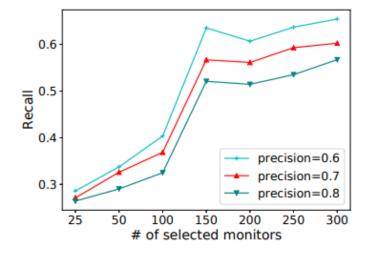


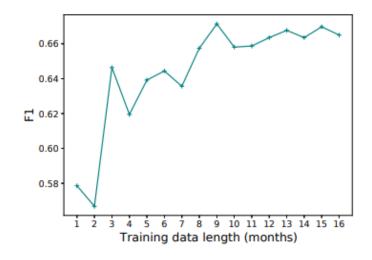


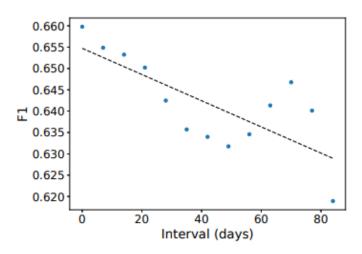
Experimentation

Important system parameters

- # of selected monitors
- Data requirements for training the detection model







Conclusions

- Warden is a framework to detect incidents in an automated way
 - Detecting potential incidents
 - Extracting related alert signals and notifying relevant engineers
- Warden is proven to be effective with data collected from 26 major services and real deployment in the IcM of Azure

Thanks