



Smart Monitoring System For Automatic Anomaly Detection and Problem Diagnosis

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Who am I?

Xianping Qu

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Baidu SRE Team

- Over 400 Engineers
- Support over 300 products, developed by 20,000 engineers

Areas of Interest

- Monitoring system
- Data Analysis



About Baidu (百度)



Search

Webpage/Image/Video/News/Dictionar y/Web Directory/...

Social

– Forum/Album/...

LBS

Maps/Group Buy/...

Knowledge

Wiki/Knows/Experience/...

Mobile

Search/Mobile phone assistant/...

Cloud

Personal Cloud storage/Baidu Cloud/...

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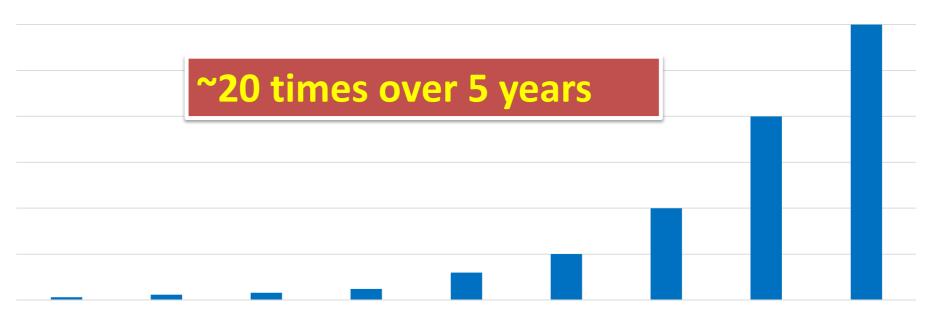
Agenda

- Background
- Smart Monitoring System
 - Anomaly Detection
 - Alarm Filter
 - Problem Diagnosis
- Summarize



Exponentially growing servers

Trend of numbers of servers at Baidu



2006 2007 2008 2009 2010 2011 2012 2013 2014



The scale of the data



Total Data	> 1.5 EB
Webpage count	> 300 billion
Webpage updates	> 1 billion / day
Monitoring data	> 20PB
Monitoring data growth	> 40TB / day

Monitoring and Diagnosis are vital!



Early methods for monitoring and diagnosis

Monitoring Items

- Machine level: 90 million

- Service level: 50 million

Alarms

- Threshold
 - Accuracy
 - Drifting
- Annoyance vs promptness

Diagnosis

- Curves, Logs, Trace

Human labor Human experience





Some Examples In Detail

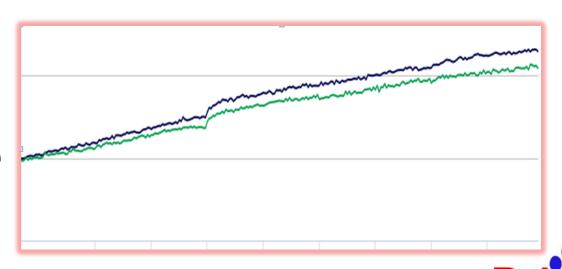


Diverse thresholds

Threshold against yesterday's value



Threshold against accumulated value



Drifting thresholds





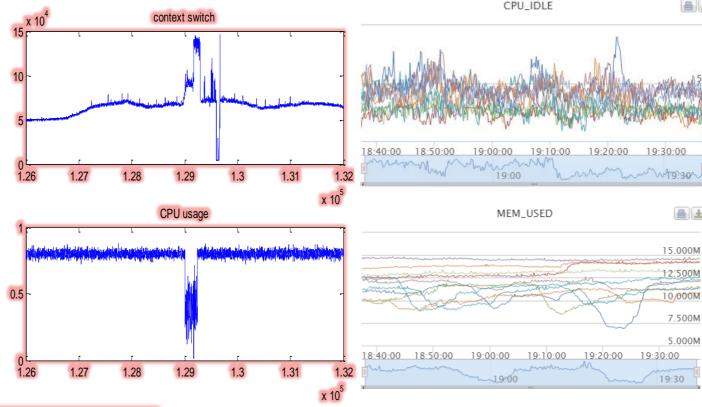
Annoyance vs Promptness of Alarms





Diagnosis

- Monitoring
- Changing Events
- Relationship
- Log
- **–** ...



				INE	1 11	12.24 070
icci	hd_vuitemplate-deplo	vuitemplate		2015-03- 10 16:30: 51	2015-03- 10 16:45: 10	SUCCESS
downloader Downloader_http://bj	Downloader_http://bj			2015-03-		DONE
			10 16:30: 42	10 16:45: 42		
cci	n_l_vuitemplate_mix	vuitemplate	I	2015-03- 10 16:30: 38	2015-03- 10 16:42: 26	SUCCESS
kbox	wise-left-bid-cpmlis			2015-03- 10 16:28: 05		RUNNING
downloader	Downloader_nmg01-fcr			2015-03- 10 16:25: 43	2015-03- 10 16:40: 43	DONE

3 12250 task_manager.cpp:504] insert task to be executed 5.nsps-imbs-999.lM.all 600 10 5 3.5 DAY

9 12234 detect_task.cpp:890] build threshold success 10.nsps-imbs-999.IM.all:IMBS_FL
 0 12234 task_manager.cpp:338] abnormal detected: [10.nsps-imbs-999.IM.all:IMBS_FLOW_expect/upper/lower/977.472/1466.21/293.242

4 12234 task_manager.cpp:51] execute task success: 10.nsps-imbs-999.IM.all, IMBS_FLC 0 12234 task manager.cpp:298] create ClientPacket for 10.208.40.22:19050

0 12234 task_manager.cpp:306] execute query[5.nsps-imbs-999.lM.all:IMBS_FLOW_cnt_avg6 12234 task_manager.cpp:338] abnormal_detected: [5.nsps-imbs-999.lM.all:IMBS_FLOW_c

ct/upper/lower/979.234/1468.85/293.77

3 12234 task_manager.cpp:51] execute task success: 5.nsps-imbs-999.IM.all, IMBS_FLOW
7 12250 task_manager.cpp:504] insert task to be executed 10.nsps-imbs-999.IM.all
600 10 5 3.5 DAY

4 12214 redis_manager.cpp:46] Redis pipeline thread exited



Automatic and smart monitoring and diagnosis



Smart monitoring system: 酷贝

Cool

Bay

酷

贝



Collection → Detection → Diagnosis

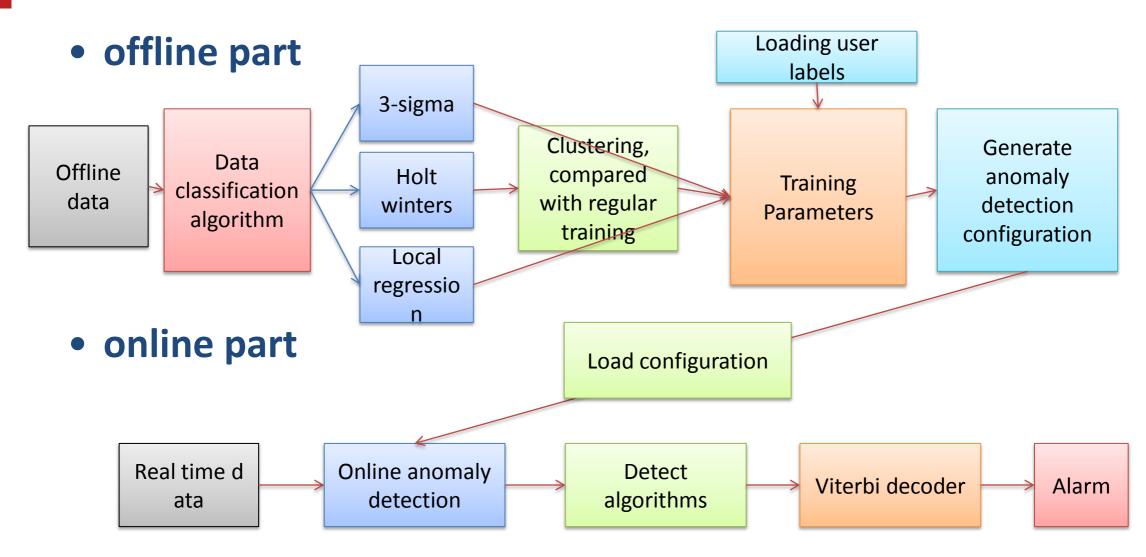


Collection

- Service data
 - PV, income, flow...
- Machine & program data
 - CPU, MEM, DISK, NETWORK...
- Changing events
- ...



Anomaly detection system



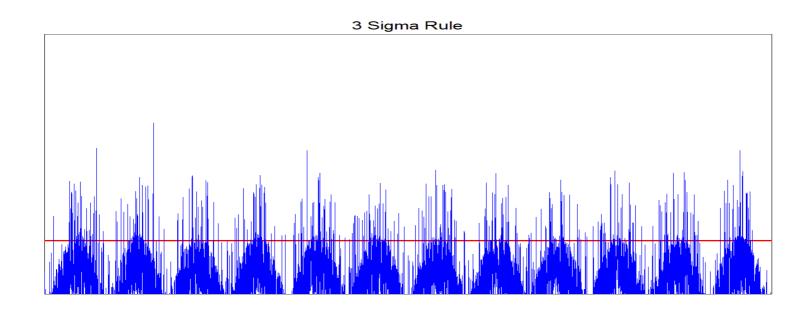


Detection: Strategy 1, 3-sigma rule

Determine constant threshold algorithmically

- Statistics on past data
- Assume Gaussian distribution

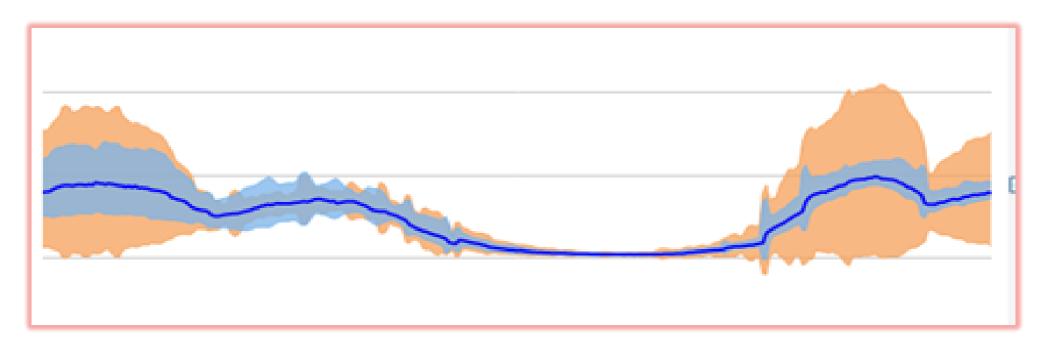
$$\bar{x} \pm 3\sigma$$





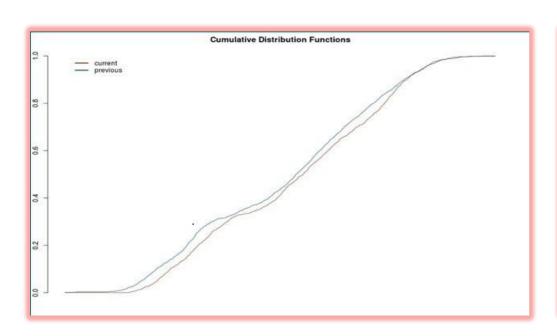
Detection: Strategy 2, the segmented 3-sigma rule

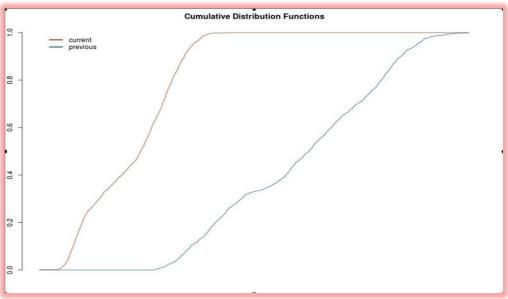
- Multiple distributions : day vs night
 - Split data into 15-min segments
 - Day-on-day statistics





Detection: Strategy 3, KS-test



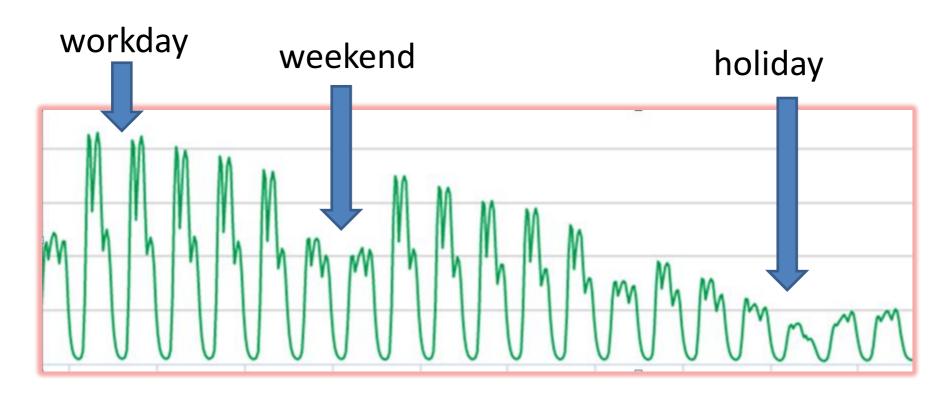


- Multiple distributions: workdays vs weekends / holidays
 - Differentiate using KS-test
 - Statistics within Workday/Weekend/Holiday



Other factors to consider

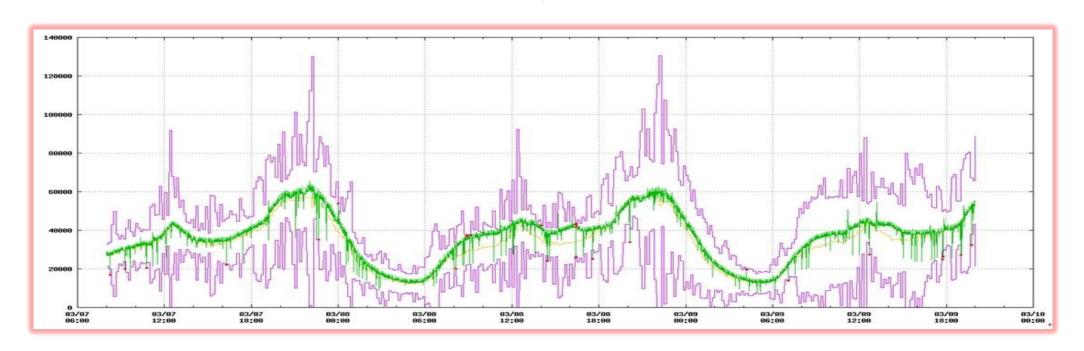
- Sometimes, the data is changing
 - e.g. compare the holiday PV with the workday PV





Detection: Strategy 3, holt-winters

- Holt-winters algorithm
- Learn both seasonal and adjacent trend





Detection: Strategy 4, local regression

- The detection of local spurt or sudden drop
 - LOESS algorithm based on local regression





Deal with slow decline in data trend

Smart cumulative method

- The signal cycle of raw data changed
 - -10s, 30s changed to 15mins, 1hour, 24hours

• Then use 3-sigma and holt-winters to detect



Challenge: generated threshold parameters vs. business requirement

- Incorporate engineers' manual input:
 - Modify the parameters
 - Mark undetected anomaly
 - Mark false positive alarm

- Learn adjustment automatically
 - Alarm label → Adjust parameters + / -



Alarm filter

- Viterbi decoder
- The formation of abnormal events, rather than a single abnormal points



Problem diagnosis

Help Diagnose

- -Total-dimensions and sub-dimensions
- Upgrade/operation event and time series data
- Operation and maintenance module relationship



Total dimension and sub-dimension

- Based on the total dimension and sub-dimension
- Example: the total revenue and advertise revenue
- Sort by impact weight

⇒ 属性	⇒当天值	⇒占比	♦ 天同比	♦ 天变化量	⇒周同比	⇒周变化量	⇒操作
and the same		0.56%	-11.68%	-465.14	0.97%	33.79	查看
AND DESCRIPTION OF THE PERSON		1.55%	-3.87%	-390.12	19.47%	1,578.63	查看
100000		0.28%	-12.24%	-240.55	-2.85%	-50.60	查看
1000		0.50%	-4.39%	-144.71	7.21%	211.84	查看
1000		0.87%	-1.68%	-92.70	-1.61%	-88.87	查看
100000		0.23%	-4.82%	-72.90	30.80%	339.08	查看
		0.04%	-20.74%	-69.28	-0.03%	-0.08	查看
		0.04%	-22.99%	-67.92	7.53%	15.93	查看
100		0.01%	-52.96%	-57.22	-33.17%	-25.23	查看
_		0.79%	-0.83%	-41.00	2.01%	97.13	查看



Heat map

- Sort anomaly numbers by dimension degree
- Slice data by abnormal regional, browser, channel
- Comparison of multi-dimensional anomaly





Issues caused by upgrades

 Based on upgrade events and metrics

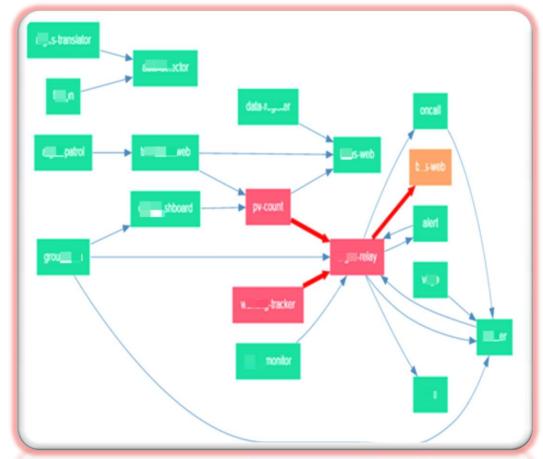
Time approximation





Issues diagnosis from operation and maintenance perspective

- Module calling graph
- Event and module relations





Establish the overall service view



Now, "酷贝" is born!



Summary

• It is difficult to detect anomalies and find the root cause when an anomaly occurs.

 A smart monitoring system for automatic anomaly detection @Baidu is demonstrated.

 Including data model of incidents, proactive anomaly detection algorithms, correlation analysis, and visualization.



Future

- Characterization of workload spikes
- Dynamic resource allocation
- Capacity management
- Identification of performance problems
- ...



Thanks



