Intelligent Anomaly Detection in Heterogeneous Internet Services

Dong Wang Principal Architect, <u>wangdong13@baidu.com</u>

Baidu Inc.

usenix LISA16



December 4–9, 2016 | Boston, MA www.usenix.org/lisa16 #lisa16



Agenda

- · Brief introduction to Baidu
- · Heterogenous services and their challenges to our SREs
- Anomaly detection in typical services
- services





ARK: A generalized intelligent operation platform for Baidu



Introduction to Baidu

- · One the largest search engines in the world Veb/Image/Video/News/...
- · Besides search, we also have
 - Location Based Service Maps
 - Social/Knowledge Tieba/Zhidao
 - Online to Offline Nuomi/Waimai
 - Finance/Payment Wallet
 - Cloud computing Cloud
- Covers more than 1 Billion users in total









Bai Anomaly Detection in Heterogenous Services

· Anomaly Detection in theory



· However, here are some realities







Divide and Conquer

- · Several typical categories we manually touch on
 - ✓ Holiday sensitive
 - ✓ Very unstable
 - Requiring fully automatic configuration





Holiday Sensitive Curves

- Holidays in Chinese Calendar have no fixed dates
 2016 Spring Festival (Feb., 8th) Dragon Boat festival (Jun., 9th)
 2017 Spring Festival (Jan., 28th) Dragon Boat festival (May, 30th)
- · There is no common pattern among different holidays
- It's hard to know the baselines!
 The training data are pretty sparse





Basic idea and the result

- · Clustering on daily CDF of curves
- · LR based estimated algorithm







· Classification on dates (features include weekend, holiday, etc.)



Very Unstable Curves

- Very unstable but not anomaly, caused by ✓ Revenue with significantly different price goods Revenue under some promotions
- · The curve's variance is huge, traditional method cannot guarantee the precision/ recall







Basic Idea and Result

- A compound solution, including smoothing by sliding window
 - ✓ Reduce the impact from huge absolute values by using logarithm
 - ✓ Considering the increasing/decreasing rate
- · The results







- Two many curves to monitor, but ✓ No enough bandwidth to do manual configuration from SRE side
 - Hard to select algorithms
 - ✓ Even harder to setup/adjust parameters
- The examples of such metrics
 - RPC numbers between two modules
 - ✓ Network transmission amount on some switch devices







Basic Idea and Results

- Using machine learning to select algorithm
 Whether or not the curve is periodical
 How the curve's stability look like
 The difference between maximum and minimum
- The default parameters configuration, plus auto-adjustment based on user feedback (marked by on-call)
- The sampling results so far
 Precision is about 84%





More Pain Points

- · Codes are hard to reuse
 - ✓ Different execution environments (programing languages)
 - ✓ Different data sources with different formats
 - Different teams/projects





Other SREs have needs to customize the existed algorithms



ARK – A Generalized OP Platform









Unified Algorithms and deployment





Unified Monitoring DB and Platform









C 智能OnCall

百度地图十一出行高峰保障直播间

地图活动直播间10.1

🚰 活动状态		
12点开抢猴子	10	1
• 最高QPS:		
• 累计PV		
12点抢猴子	10.	-00
• 最高QPS: ■		
• 累计PV		
十一狂欢节-AR活动	10.0	-
• 最高QPS		
• 累计PV		
玩游戏,赢汽车!	10.0?	- 0
• 最高QP:		
■ WitpV/		
④ 止损历史		
PC 异常指标已恢复正常, 止损		7:27
N/ 异常指标已恢复正常, 止损.		:25
C → → → → → → → → → → → → → → → → → → →		:11
(5	:14

直播内	容			
输入命令	¢			
	止损信息		19 , 10	1.00
	P(}常	指标已恢复	正常,止损成	功
A				
	止损信息			
	PC	60 V	bj]预案执行	成功,等
a	北總法南			
	止损信息			
	PC Edd	·预案决策, 7	+始执行[ma	r 📕
<u>e</u>				
	止损信息			
	PC 发现	诸标异常,	[¢	1990) 1990
	状态通知			
	各位值班同	学辛苦了,	当前地图出行	〒服务正常
	- 充量			
	違			
	流量			
	流量			
		当前		
	流量 图流量			
		当前」		
	n de		p3	
	状态通知			
		学辛苦了,		
	量	当前	ps	日级增长





开启 🂽 关闭

开启 🂽 关闭

开启 🂽 关闭

开启 🌒 关闭

相关操作以及信息

🕑 VI机器人

止损通知

定位通知

状态通知

活动通知

Ø 快捷链接





Thanks Very Much & Welcome Questions!





