



GOPS 2021  
Shenzhen

# GOPS

# 全球运维大会

2021  
-XOPS 风向标



深圳站

中国·深圳

指导单位：



主办单位：



时间：2021年5月21日-22日

# ChaosOps 探索与落地实践

黄帅 亚马逊资深技术专家

# 黄帅

## 亚马逊 资深技术专家

在软件研发领域有十多年架构设计、分布式系统运维以及团队管理经验，近年来对混沌工程企业实践有深入的研究，力主推动亚马逊云科技全球混沌工程服务 AWS Fault Injection Simulator (FIS) 的落地，于2021年3月成功发布。



# 目录

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- ① 云原生的新挑战
- ② 韧性之道的演进
- ③ ChaosOps呼之欲出
- ④ 下一步

# 云原生的新挑战

三个生产事件的启示

# 1.0

# 生产事件一：服务状态仪表盘悖论

服务状态仪表盘悖论  
(Service Status Paradox)

提供服务健康状态

健康状况更新失败

架构特别简单

存在整整十年之久

却未有人发现

这样的悖论还有吗？

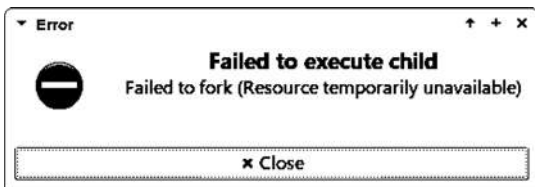
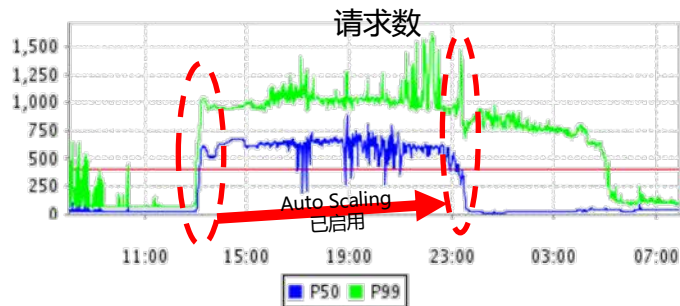
Service Status								
	DCO-001	DCO-002	DCO-003	DCO-004	DCO-005	DCO-006	DCO-007	DCO-008
Core services	✓	✓	✓	✓	✓	✓	✓	✓
Ninja	✓	!	✓	✓	✓	✓	✓	✓
Athos	✓	✓	✗	✓	✓	✓	✓	✓
Metis	✓	✓	✓	✓	✓	✓	✓	✓
Demeter	✓	✓	✓	✓	✓	✓	✓	✗
Leto	!	✓	✓	✓	✓	✓	✓	✓
Other services	✓	✓	✓	✓	✓	✓	✓	✓
<span>✓ Healthy</span> <span>! Degraded</span> <span>✗ Disrupted</span> <span>! Informational</span>								

故障类型：依赖失效

# 生产事件二：冗余性悖论

## 冗余性悖论

请问：冗余性是否一定能提高可用性？



故障类型：资源耗尽



$$A=1-(1-Ax)^2$$

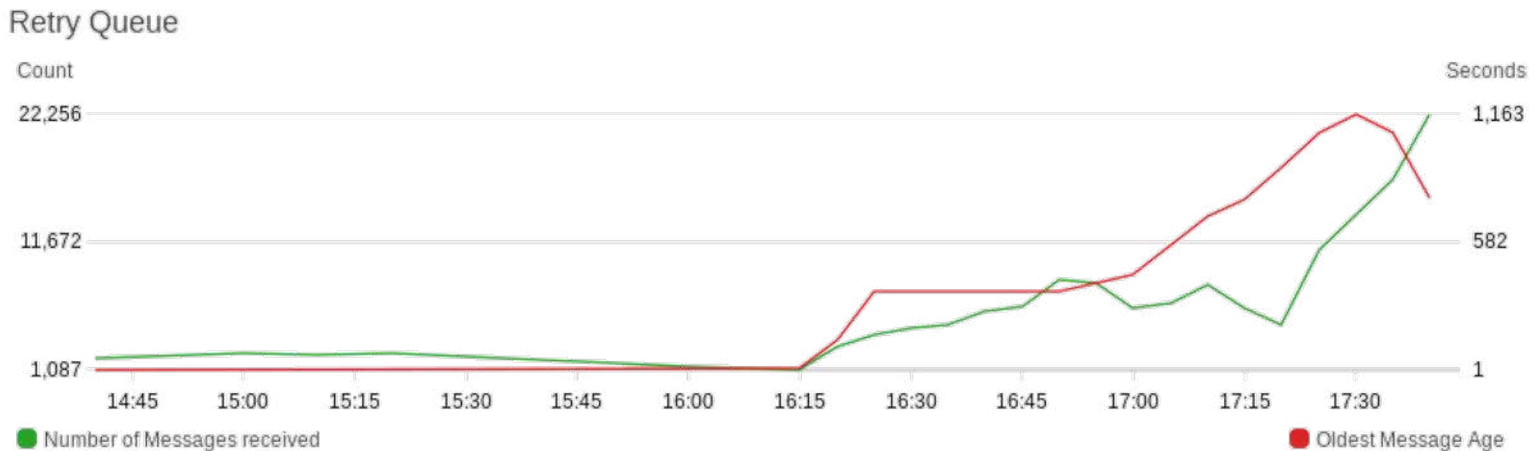
组件	可用性(A)	宕机时长(按年计)
一个组件 X 提供过服务	99% (2个9)	3天15个小时
两个组件 X 并行提供服务	99.99% (4个9)	52分钟
三个组件 X 并行提供服务	99.9999% (6个9)	31秒

如果故障高度相关，则冗余不会增加可用性，还会产生雪崩效应。

为什么？

# 生产事件三：告警系统瘫痪

所有的功能发布被暂停



故障类型：重试风暴

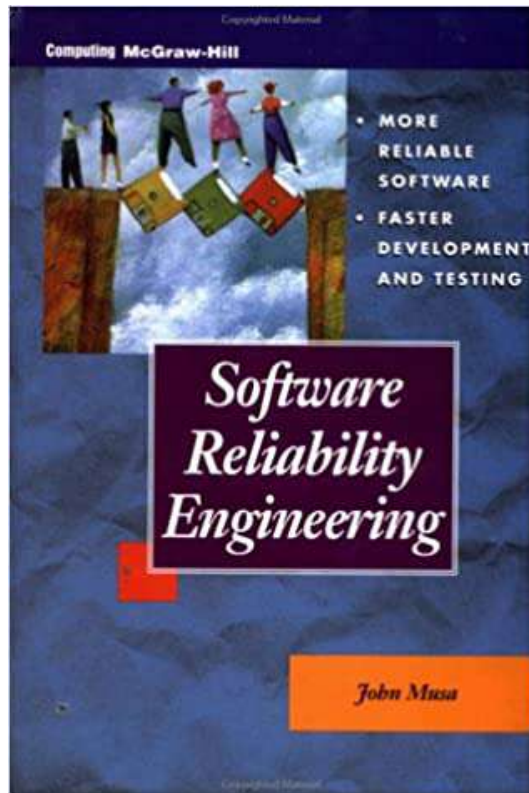
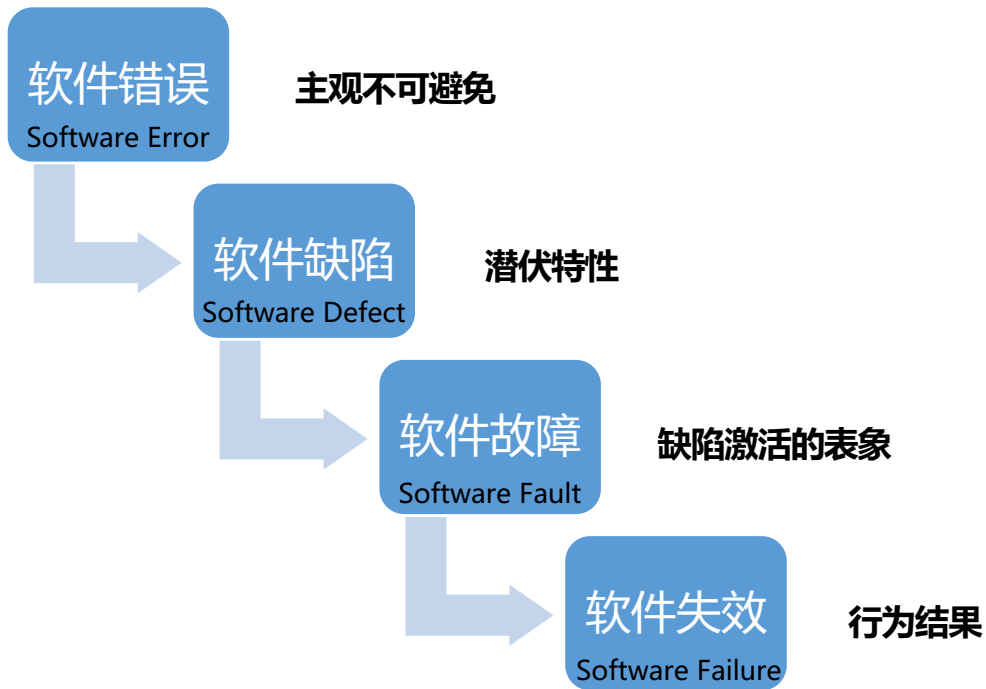


# 云原生的新挑战

故障的宿命论

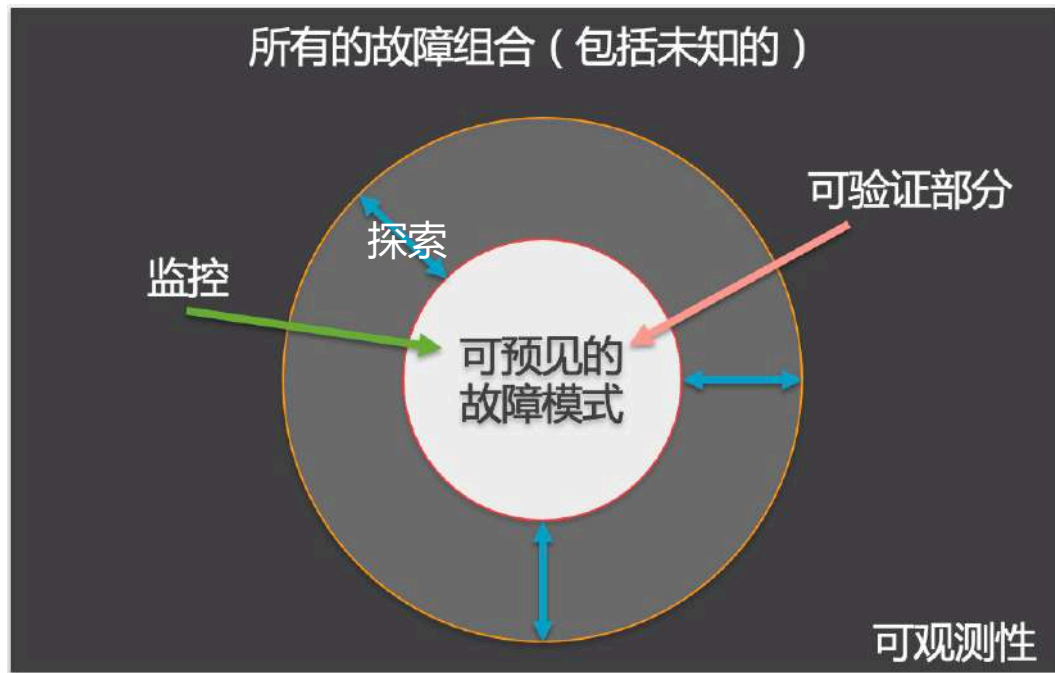
# 1.1

# 故障来自哪里？



1988

# 故障的宿命论



一切皆因人类设计的系统变得愈发复杂，超出了人类认知范围。

# 常见的故障类型

- 负载尖峰和突发过载
  - 邻近的故障转移
  - 崩溃查询
  - 重试风暴
  - 资源耗尽
  - 资源限制
- CPU问题
  - 内存问题
  - 线程枯竭
  - 发布和变更
  - 启动时间过长
  - 依赖失效

小故障引起海啸，这和云原生有什么关系？

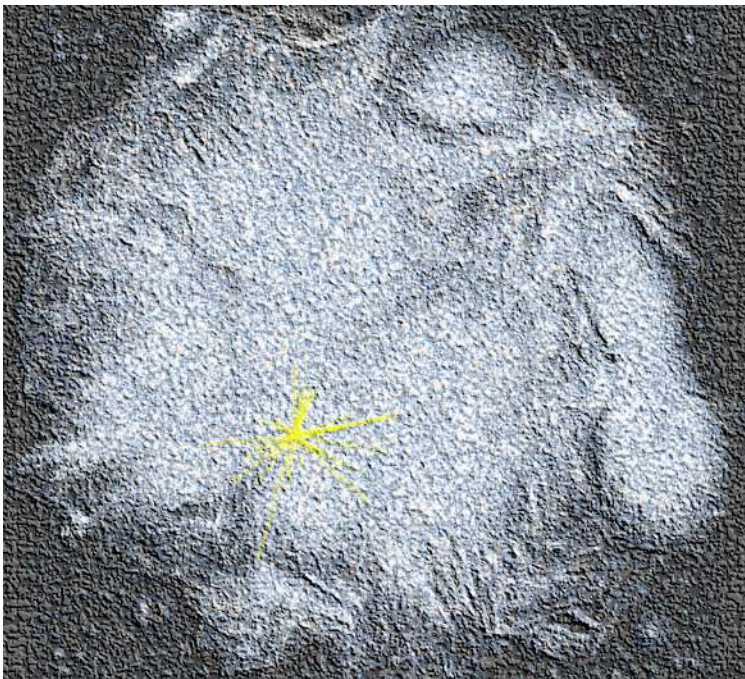
# 云原生的新挑战

面临的问题

# 1.2

# 云原生时代的新挑战

## 分布式架构的复杂性



2021

1. 预生产无法和生产保持完全一致
2. 测试通过，一上生产就出问题
3. 集成测试的复杂性
4. 回归测试的不确定性
5. 端到端测试的难度
6. 故障注入测试的重视

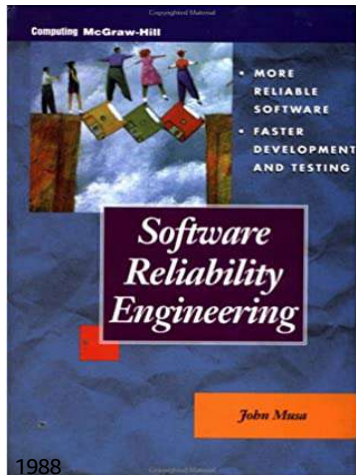


1. 级联故障模式的探索
2. 可观测性的复杂度
3. 配置错误的危险性
4. 功能标记爆炸问题
5. 生产调试的风险
6. 人为错误的避免
7. 应急响应的肌肉记忆

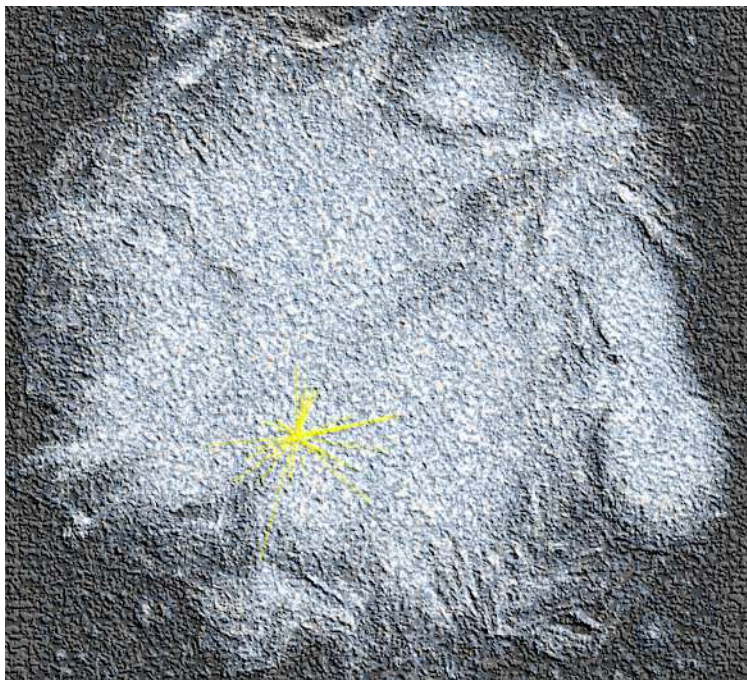


# 可靠性/韧性工程的焕然一新

软件可靠性工程

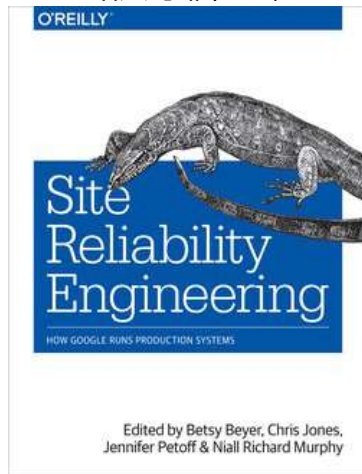


分布式架构的复杂性



2021

站点可靠性工程



# 韧性之道的演进

GameDay文化

# 2.0



# GameDay文化



Jesse Robbins, "Master of Disaster"

他是一名志愿消防员。

2001：亚马逊唯一的 Master of Disaster 头衔

2004：在亚马逊发明了 GameDay

2007：著名 Web 性能和运维大会（Velocity）的创始人

2008：Chef 的创始人

2013：PagerDuty 的早期投资人

2019：Orion Labs 的创始人和 CEO

“游戏日计划”是一个基于工程师团队的**交互式**和**开放式**的学习与训练。旨在测试系统中**模拟**各种事件响应的**流程**，比如故障注入、被侵入、性能扩展要求等等。目的是**训练**工程师团队的**响应能力**以及建立如何应对的“肌肉记忆”。

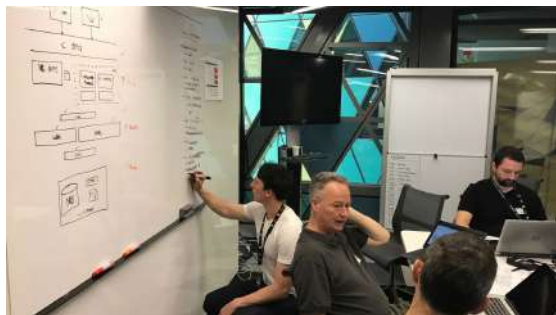


幸运大转盘

## FAQs

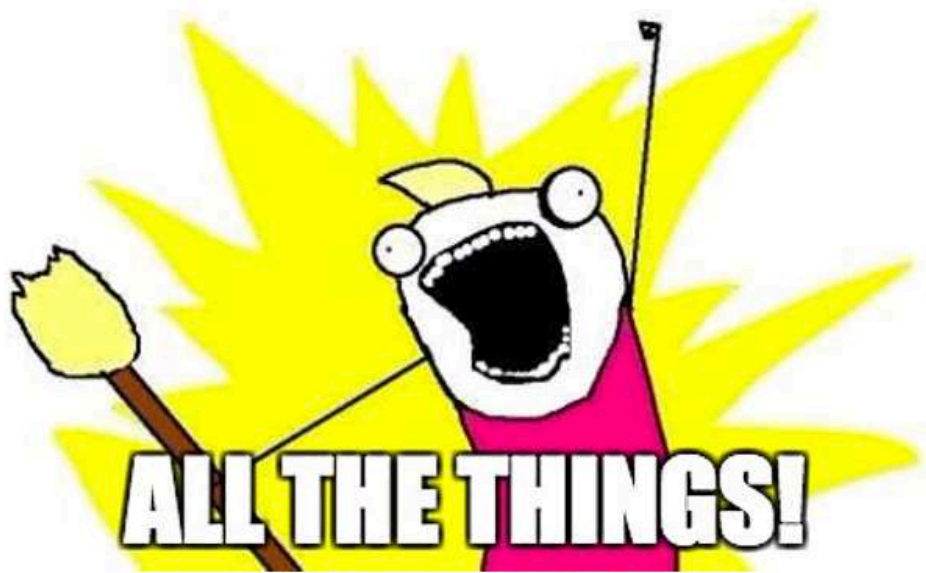
For 2020, please see [2020 GameDay FAQ](#)  
For 2019, please see [2019 GameDay FAQ](#)  
For 2018, please see [2018 GameDay FAQ](#)  
For 2017, please see [2017 GameDay FAQ](#)  
For 2016, please see [2016 GameDay FAQ](#)  
For 2015, please see [2015 GameDay FAQ](#)  
For 2014, please see [2014 GameDay FAQ](#)  
For 2013, please see [2013 GameDay FAQ](#)  
For 2012, please see [2012 GameDay FAQ](#)  
For 2011, please see [2011 GameDay FAQ](#)  
For 2010, please see [2010 GameDay FAQ](#)  
For 2009, please see [2009 GameDay FAQ](#)  
For 2008, please see [2008 GameDay FAQ](#)  
For 2007, please see [2007 GameDay FAQ](#)  
For 2006, please see [2006 GameDay FAQ](#)  
For 2005, please see [2005 GameDay FAQ](#)  
For 2004, please see [2004 GameDay FAQ](#)

# GameDay文化



# GameDay文化

**AUTOMATE**



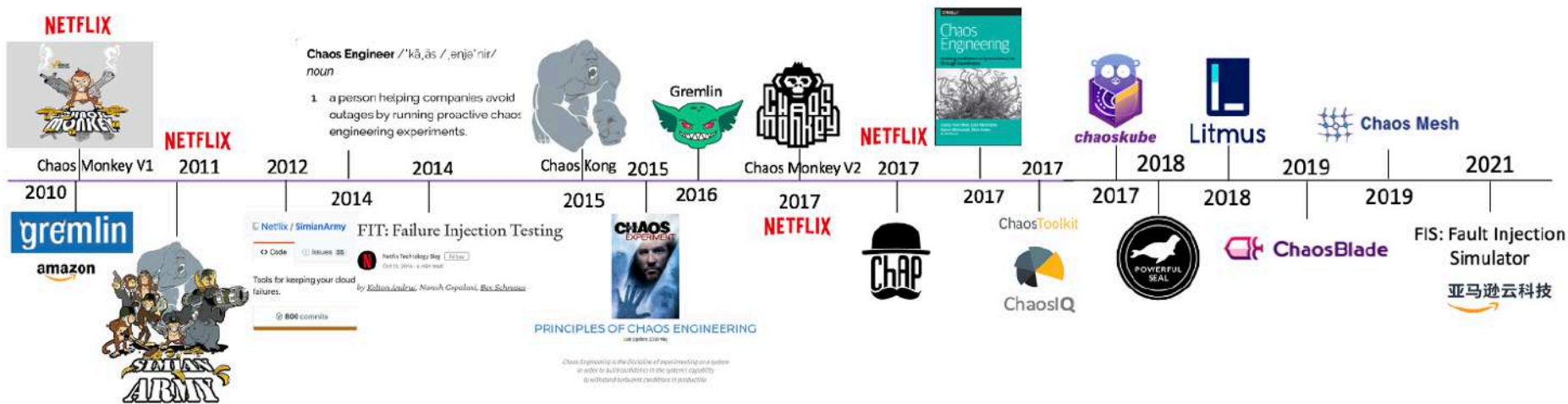
# 韧性之道的演进

混沌工程的诞生

## 2.1

# 混沌工程的十年演进史

这是最好的时代，海内外齐头并进，不分伯仲！



引用自亚马逊官方博客 <https://amazonaws-china.com/cn/blogs/china/aws-chaos-engineering-start/>

# 混沌工程的目标：韧性架构和自愈能力

## 韧性架构

冗余性

扩展性

不可变基础设施

无状态应用

基础设施即代码

避免级联故障

转移  
切换

重试  
退避

超时  
机制

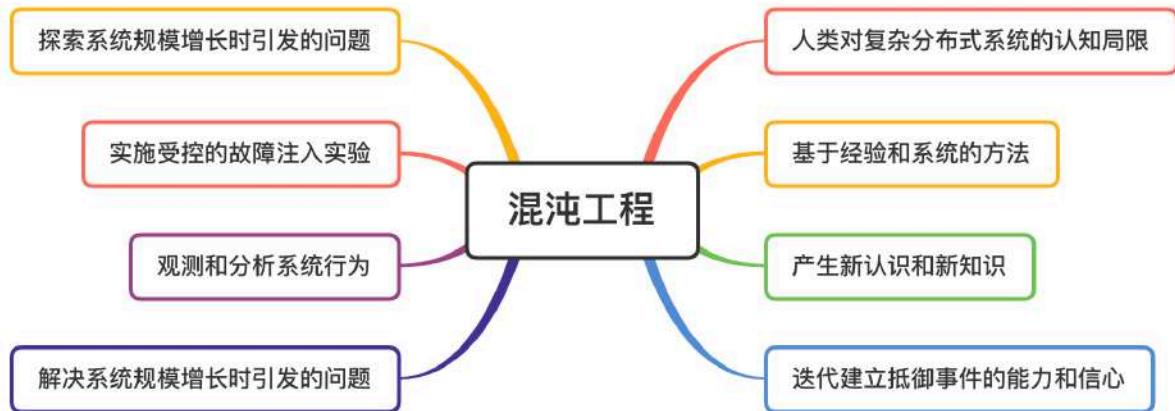
幂等  
操作

服务  
降级

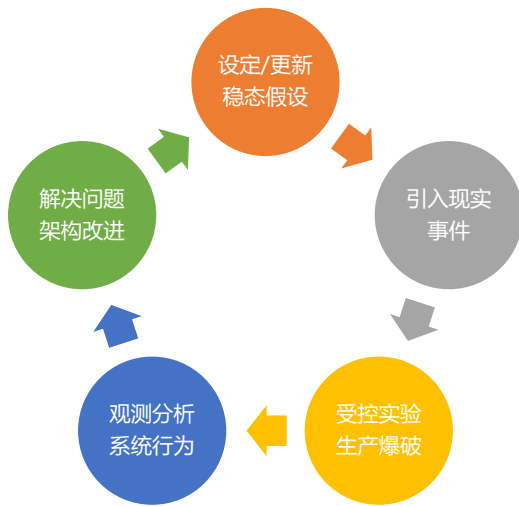
拒绝  
服务

服务  
熔断

# 混沌工程的内容



## 混沌工程的定位



## 混沌工程原理

# 韧性之道的演进

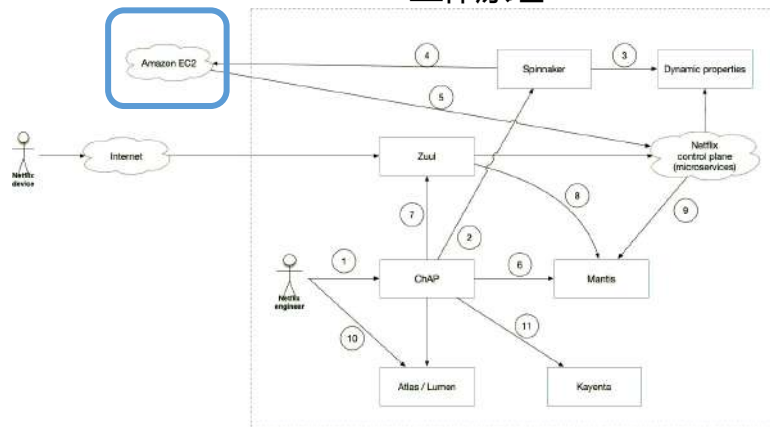
一个最著名的实践案例

## 2.2

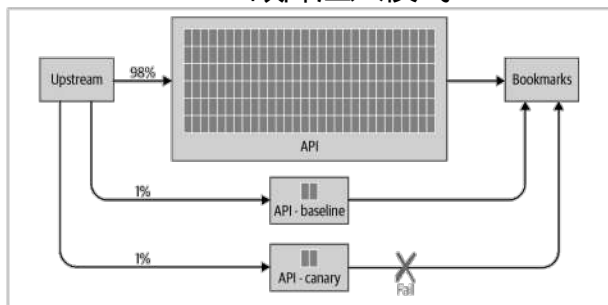


# Netflix 混沌工程生产实践 - ChAP

## ChAP工作原理



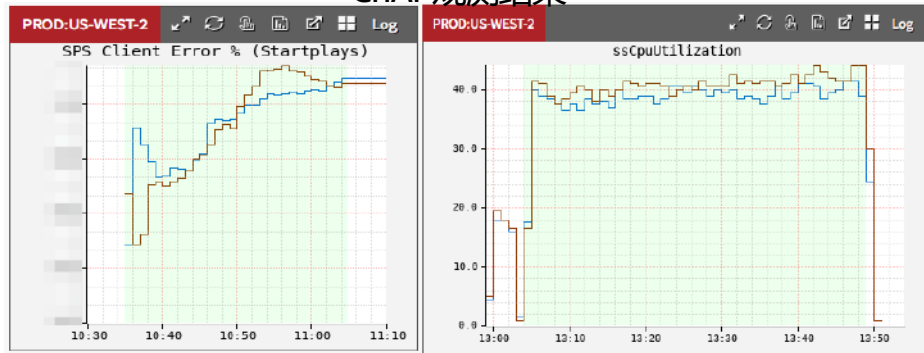
## ChAP故障注入模式



## ChAP控制台



## ChAP观测结果



# 混沌工程技术的演进总结 ( \*\*精华\*\* )

技术内容	2014年之前	2017年之后
故障场景类型	基础设施的单一故障	多层次组合故障
故障场景编排	预先指定的故障	故障优先级排序和自动挑选
故障注入模式	随机注入	对照实验观测
目标工作环境	非生产环境	生产灰度环境
监控/观测能力	监控与日志系统	含链路追踪的可观测性体系
结果/成果分析	阈值或人工判定	自动灰度稳态分析
安全管控方法	人工评估	最小爆炸半径和一键关停等
具体呈现形态	命令行工具	自动化平台

# ChaosOps呼之欲出

应对云原生的挑战

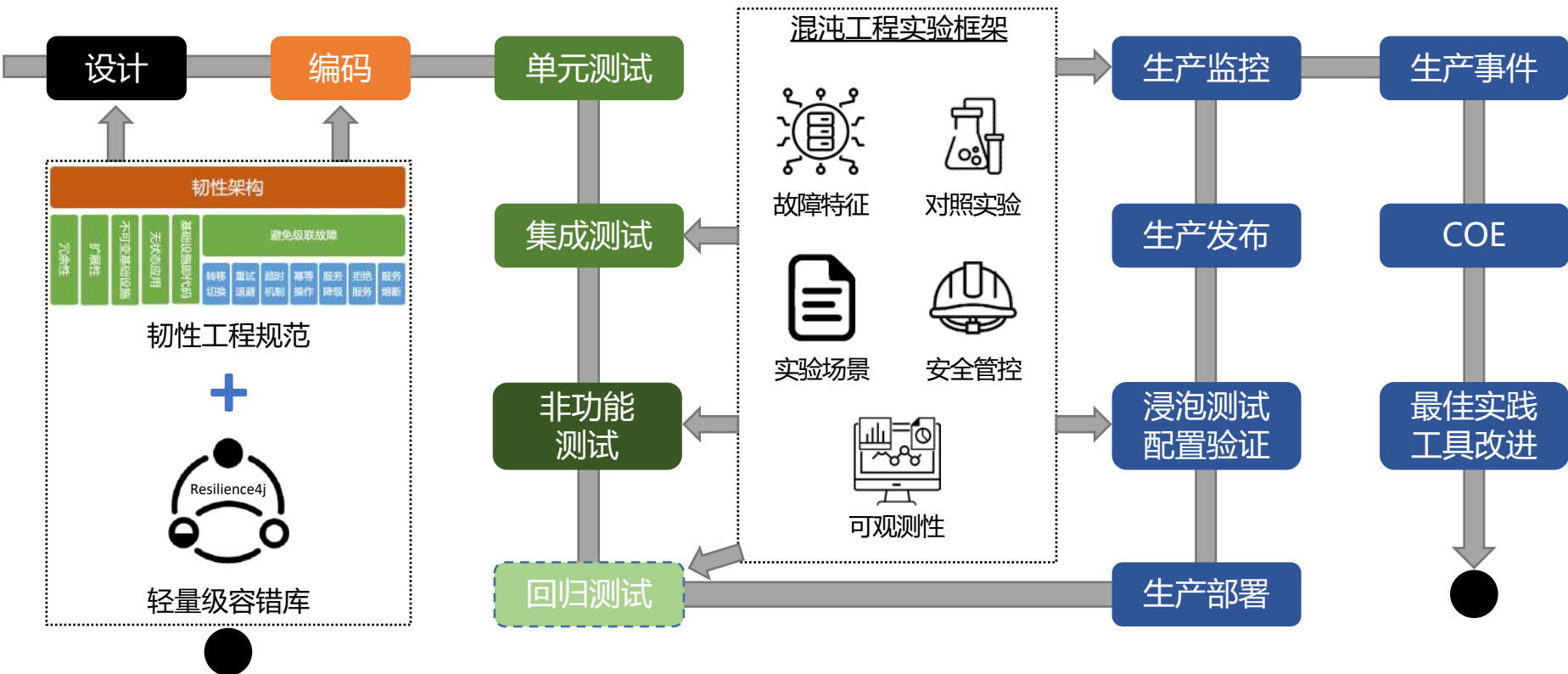
# 3.0

# ChaosOps又是一个新造的名词？

- 新名词的诞生，往往是因为没有现存的名词，可以突出某项工作的价值。
- 毋庸置疑，ChaosOps自然是要突出混沌工程在整个DevOps实践的价值。

**ChaosOps**是在既有的**DevOps**实践中，通过**创新的混沌工程方法和技术**，应对**云原生时代可靠性工程的新挑战**。

# 混沌工程融入系统开发测试、部署和发布流程



# ChaosOps呼之欲出

核心原理

## 3.1



故障特征



对照实验



实验场景



安全管控



可观测性

# ChaosOps呼之欲出

核心原理

## 3.2



# 常见故障特征

- 负载尖峰和突发过载

负载测试

- 邻近的故障转移

容灾演练

- 崩溃查询

功能测试

- 重试风暴

故障注入

- 资源耗尽

故障注入

- 资源限制

故障注入

- CPU问题

故障注入

- 内存问题

故障注入

- 线程枯竭

故障注入

- 发布和变更

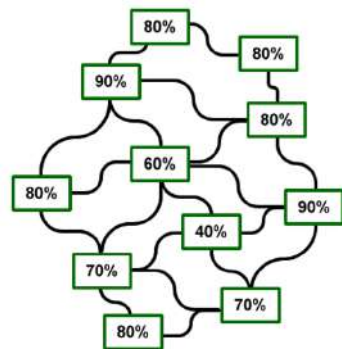
回归测试

- 启动时间过长

故障注入

- 依赖失效

故障注入



Network running normally



# 常见故障注入点

故障注入点	具体描述
依赖型故障	AWS 托管服务异常：ELB / S3 / DynamoDB / Lambda ...
主机型故障	EC2 实例异常终止，EC2 实例异常关闭，EBS 磁盘卷异常卸载，RDS 数据库实例故障切换，ElastiCache 实例故障切换，容器异常终止，容器异常关闭，...
操作系统内故障	CPU、内存、磁盘空间、IOPS 占满或突发过高占用，大文件，只读系统，系统重启，熵耗尽，...
网络故障	网络延迟，网络丢包，DNS 解析故障，DNS 缓存毒化，VIP 转移，网络黑洞，...
服务层故障	不正常关闭连接，进程被杀死，暂停 / 启用进程，内核崩溃，...
请求拦截型故障	异常请求，请求处理延迟，...

# ChaosOps呼之欲出

核心原理

## 3.3



故障特征



对照实验



实验场景

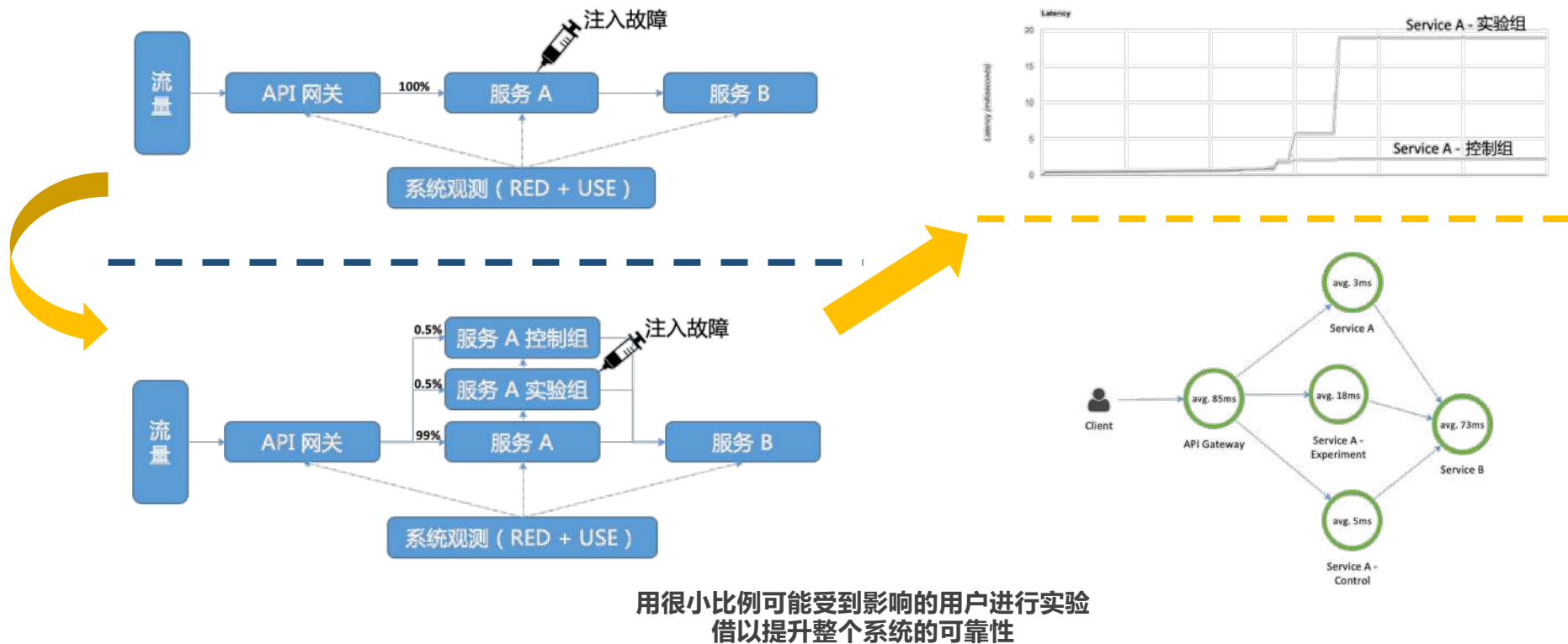


安全管控



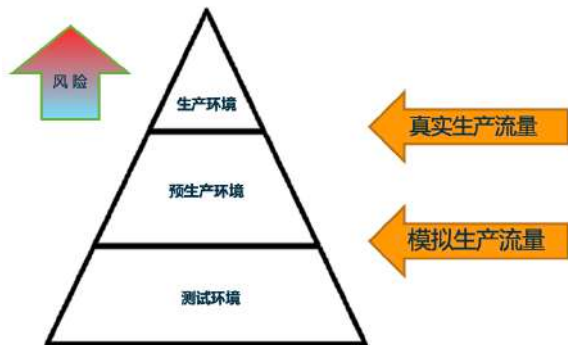
可观测性

# 对照实验原理



# 示例：生产级对照实验

1. 生产的灰度环境
2. 很小比例的生产流量
3. 对比故障注入效果
4. 非生产，则使用流量回放




**Wave2-Prod**  
 APOLLO  
 Revision history  
 XXXXService/DC0/Prod  
 Succeeded deployment 43m ago - diff  
 Override approval on revisions

Approval Workflows


**Canary 1 Approvals**  
 In progress - details  
 Cancel  
 What changes will be approved?  
**Chaos Canary - 0.5%**  
 In progress  
 Job #453967655 In Progress.  
 Total run time: 30 minutes. Total time remaining: 11 hours 29 minutes.


**Canary 2 Approvals**  
 In progress - details  
 Cancel  
 What changes will be approved?  
**Chaos Canary - 0.5%**  
 In progress  
 Job #453952050 In Progress.  
 Total run time: 30 minutes. Total time remaining: 11 hours 29 minutes.



# ChaosOps呼之欲出

核心原理

## 3.4



故障特征



对照实验



实验场景



安全管控



可观测性

# 实验场景完整示例

实验名称

实验描述

实验权限

暂停条件

实验对象

故障注入点

```
{
  "tags": {
    "Name": "StopAndRestartRandomInstance"
  },
  "description": "Stop and Restart One Random Instance",
  "roleArn": "arn:aws:iam::0123456789:role/MyFISExperimentRole",
  "stopConditions": [
    {
      "source": "aws:cloudwatch:alarm",
      "value": "arn:aws:cloudwatch:us-east-1:0123456789:alarm:No_Traffic"
    }
  ],
  "targets": {
    "myInstance": {
      "resourceTags": {
        "Env": "test"
      },
      "resourceType": "aws:ec2:instance",
      "selectionMode": "PERCENT(25)"
    }
  },
  "actions": {
    "stopInstances": {
      "actionId": "aws:ec2:stop-instances",
      "description": "stop the instances",
      "parameters": {
        "startInstancesAtEnd": "true",
        "duration": "PT2M",
      },
      "targets": {
        "Instances": "myInstance"
      }
    }
  }
}
```

# 标准化实验场景库

以史为镜

定制化实验场景

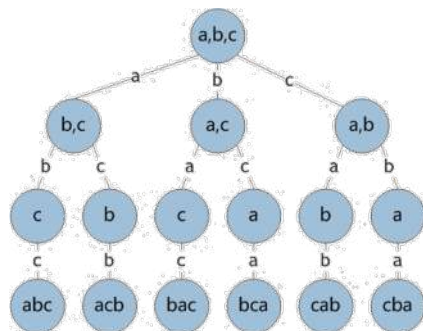
历史故障  
重现

历史故障  
类比引申

探索性  
实验研究

- <https://status.azure.com/zh-cn/status/history/>
- <https://status.cloud.google.com/summary>
- <http://aws.amazon.com/premiumsupport/technology/pes/>

## 实验场景爆炸问题

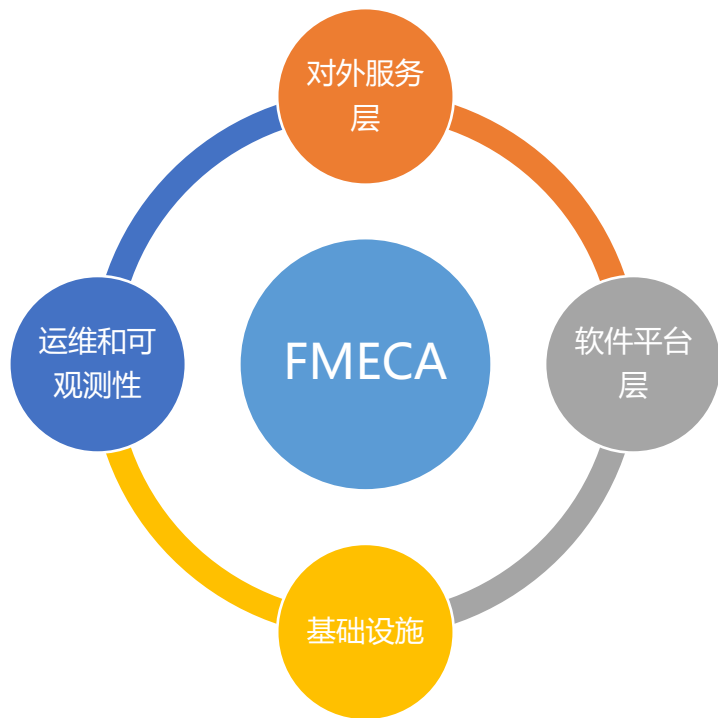


某系统是由相互依赖的10种组件构成，按组件失效的全排列组合计算，总共会产生  $10! = 3,628,800$  种实验场景。

## 实验场景的人力、时间和资源开销

- 实验场景的持续维护
- 实验场景的最小爆炸半径
- 实验场景的执行模式
- 实验场景的有效性探索和验证
- 实验场景发现问题时的根因分析

# 失效模式影响的关键因素分析法 ( FMECA )



## History of the FMEA

- 1940s - First developed by the US military in 1949 to determine the effect of system and equipment failures
- 1960s - Adopted and refined by NASA (used in the Apollo Space program)
- 1970s – Ford Motor Co. introduces FMEA after the Pinto affair. Soon adopted across automotive industry
- Today – FMEA used in both manufacturing and service industries





# 以对外服务层为例

System: \_\_\_\_\_  
Subsystem: \_\_\_\_\_  
Component: \_\_\_\_\_  
Design Lead: \_\_\_\_\_  
Date: \_\_\_\_\_

Doc: Service Availability  
Application Layer  
Controlled by the application domain

Version Number: 1  
Prepared By: Adrian Cook/DH  
Date: 12/20/18  
Revision Date: 4/24/2021  
Page: 1 of 1

Potential Failure Modes and Effects

Item / Function	Potential Failure Mode(s)	Potential Effects of Failure	S/N	Potential Cause(s) / Mechanism(s) of Failure	Prob	Current Design Controls	Det	RPN	Recommended Actions	Action Results				
										Actions Taken	New Sev	New Det	New RPN	New Prob
Authentication	Client can't authenticate	Can't connect application	3	Certificate timeout, version mismatch, account not setup, credential changed	3	Log and alert on authentication failures	3	40						6
	Slow or unreliable authentication	Slow start for application	4	Auth service overloaded, higher error and retry rate	3	Log and alert on high authentication latency and errors	4	48						
Client Request to API Endpoint								8						6
	Service unknown, address unreachable	Delay while discovery or DNS times out, slow fallback response	6	DNS configuration error, denial of service attack, or provider failure	1	Customer eventually complains via call center	10	60	Dual redundant DNS, fallback to local cache, hardcoded IP addresses, Endpoint monitoring and alerts					6
	Service unreachable, request undeliverable	Fast fail, no response	4	Network route down or no service instances running	1	Autoscale replicates a number of healthy instances	1	4	Endpoint monitoring and alerts					6
	Service reachable, request undeliverable	Connect timeout, slow fail, no response	4	Service frozen/not accepting connection	1	Retry request on different instance. Healthcheck failure instances removed. Log and alert.	2	8						6
	Request delivered, no response - stall	Application request timeout, slow fail, no response	4	Broken service code, overloaded CPU or slow dependencies	1	Retry request on different instance. Healthcheck failure instances removed. Log and alert.	2	8						6
	Response undeliverable	Application request timeout, slow fail, no response	4	Network return route failure, dropped packets	1	Retry request on different instance. Healthcheck failure instances removed. Log and alert.	2	8						6
	Response received in time but empty or unintelligible	Fast fail, no response	3	Version mismatch or exception in service code	2	Retry request on different instance. Healthcheck failure instances removed. Log and alert.	2	12						6
	Request delivered, response delayed beyond spec	Degraded response arrives too late, slow fallback response	8	Service overloaded or GC hit, dependent services responding slowly	2	Retry request on different instance. Healthcheck failure instances removed. Log and alert.	2	24						6
	Request delivered, degraded response delivered in time	Degraded timely response	2	Service overloaded or GC hit, dependent services responding slowly	2	Log and alert on high service latency and errors	2	8						6

$$RPN = Sev * Prob * Det \rightarrow \text{Priority}$$

Effect	SEVERITY of Effect	Ranking
Hazardous without warning	Very high severity ranking when a potential failure mode affects safe system operation without warning	10
Hazardous with warning	Very high severity ranking when a potential failure mode affects safe system operation with warning	9
Very High	System inoperable with destructive failure without compromising safety	8
High	System inoperable with equipment damage	7
Moderate	System inoperable with minor damage	6
Low	System inoperable without damage	5
Very Low	System operable with significant degradation of performance	4
Minor	System operable with some degradation of performance	3
Very Minor	System operable with minimal interference	2
None	No effect	1

PROBABILITY of Failure	Failure Prob	Ranking
Very High: Failure is almost inevitable	> 1 in 2	10
	1 in 3	9
High: Repeated failures	1 in 8	8
	1 in 20	7
Moderate: Occasional failures	1 in 80	6
	1 in 400	5
	1 in 2,000	4
Low: Relatively few failures	1 in 15,000	3
	1 in 150,000	2
Remote: Failure is unlikely	< 1 in 1,500,000	1

Detection	Likelihood of DETECTION by Design Control	Ranking
Absolute Uncertainty	Design control cannot detect potential cause/mechanism and subsequent failure mode	10
Very Remote	Very remote chance the design control will detect potential cause/mechanism and subsequent failure mode	9
Remote	Remote chance the design control will detect potential cause/mechanism and subsequent failure mode	8
Very Low	Very low chance the design control will detect potential cause/mechanism and subsequent failure mode	7
Low	Low chance the design control will detect potential cause/mechanism and subsequent failure mode	6
Moderate	Moderate chance the design control will detect potential cause/mechanism and subsequent failure mode	5
Moderately High	Moderately high chance the design control will detect potential cause/mechanism and subsequent failure mode	4
High	High chance the design control will detect potential cause/mechanism and subsequent failure mode	3
Very High	Very high chance the design control will detect potential cause/mechanism and subsequent failure mode	2
Almost Certain	Design control will detect potential cause/mechanism and subsequent failure mode	1

# ChaosOps呼之欲出

核心原理

# 3.5



故障特征



对照实验



实验场景



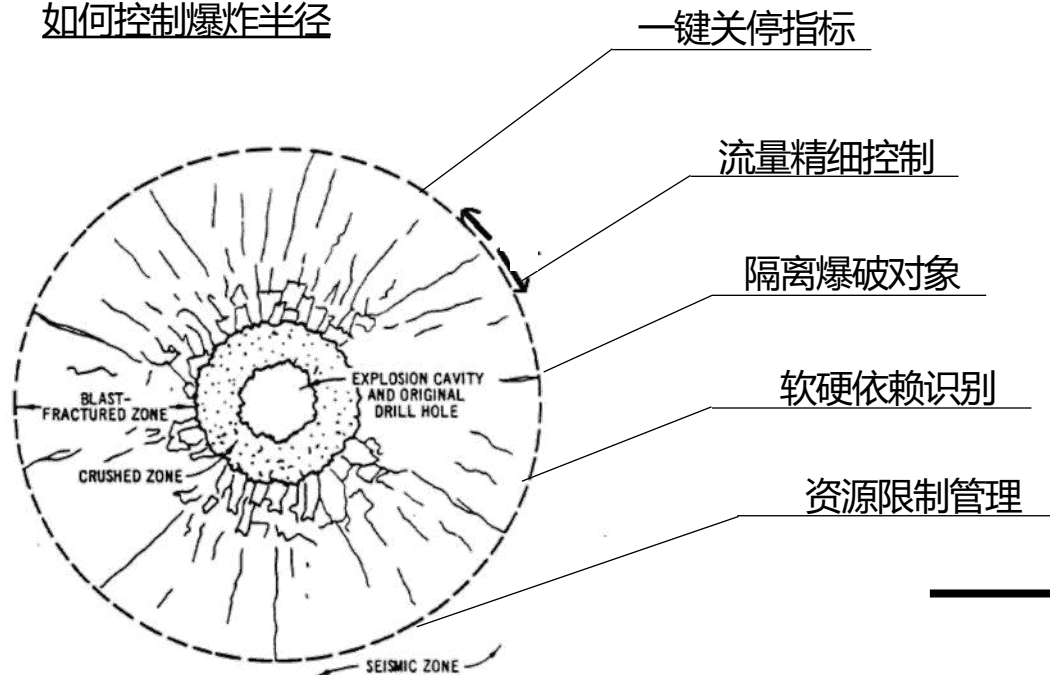
安全管控



可观测性

# 爆破半径：不能让混沌实验加速生产事件

## 如何控制爆炸半径



精细的爆破对象隔离、流量控制和软硬依赖治理，  
可有效避免引入更多噪音，影响爆破实验的根因分析



工作时间



自动停止

+

<5%

生产流量



故障转移

生产级爆炸半径控制的最佳实践

对爆炸半径的控制是一个迭代学习的过程

# ChaosOps呼之欲出

核心原理

# 3.6



故障特征



对照实验



实验场景

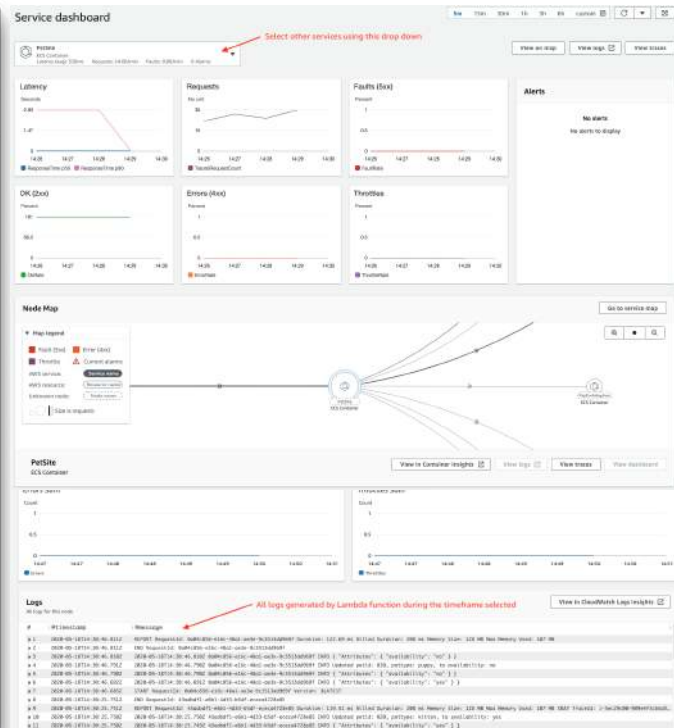
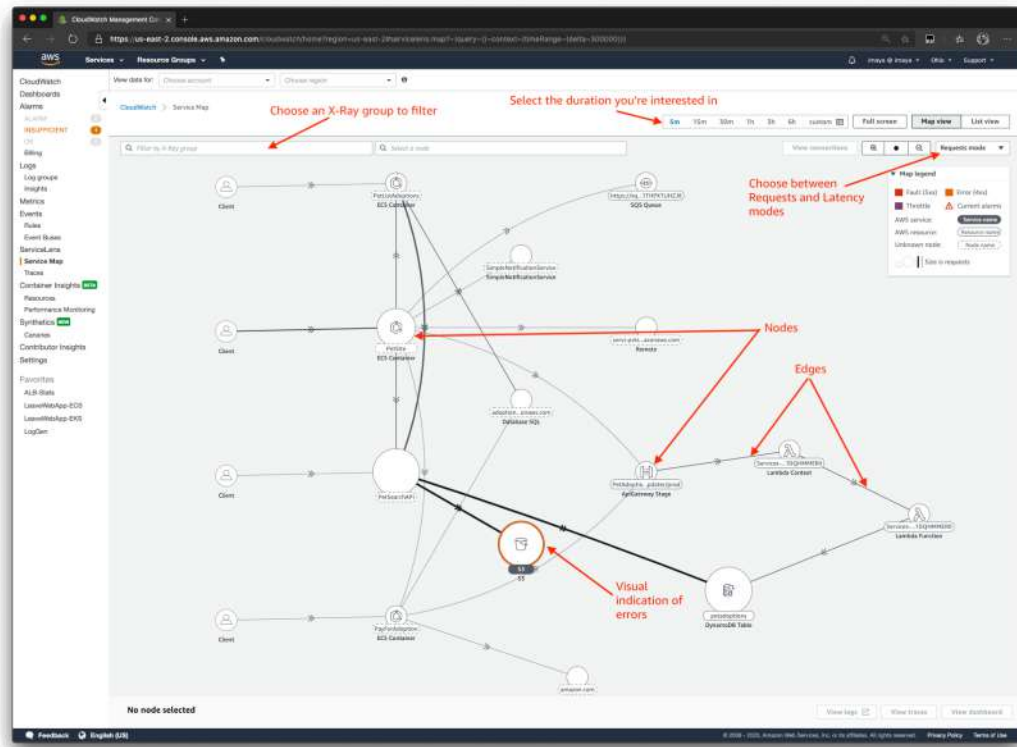


安全管控

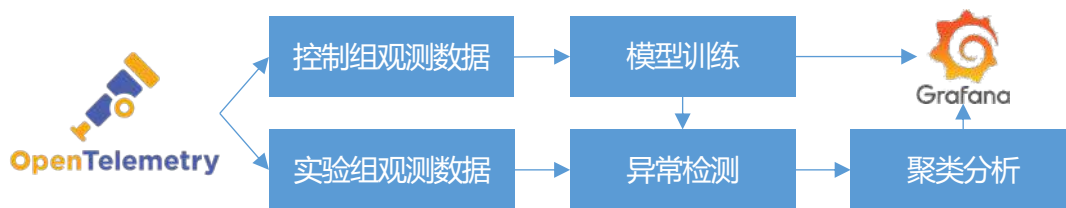


可观测性

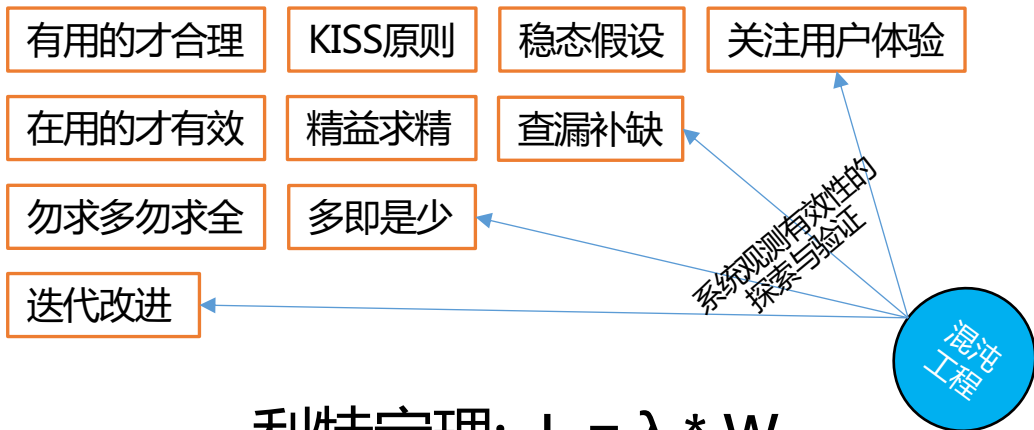
# 系统观测示例 - 服务透视



# 系统观测是混沌实验的关键步骤



# 混沌实验可以提升系统观测的有效性



利特定理:  $L = \lambda * W$

The RED Method - Weaveworks	Metrics
Four Golden Signals - Google	
The USE Method - Brendan Gregg	



# 下一步

效果与推广

# 4.0

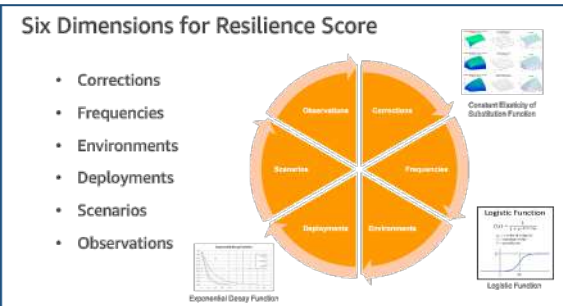
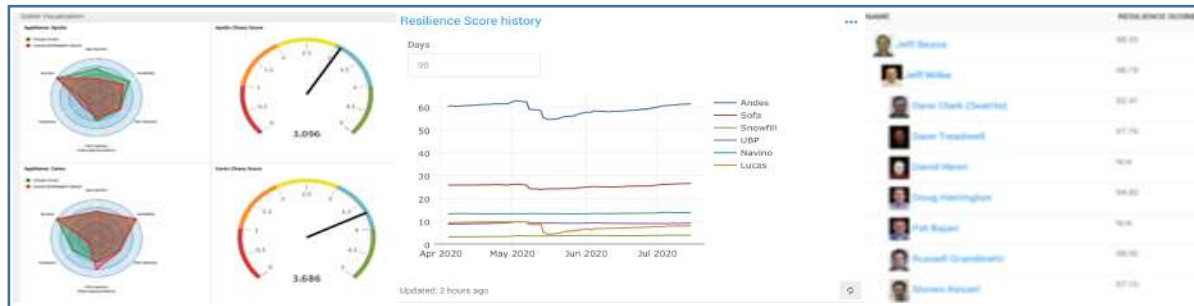


# 效果与推广

- 韧性 ( Resilience ) 是指软件通过适度降级和快速恢复而在遇到故障时保持可用性的**能力**。
- 只能通过在**遇到故障情况时**分析应用程序的行为来衡量软件的韧性。
- 混沌工程实验用于**验证**是否已使用预防故障的**最佳实践**以及软件行为是否已达到韧性目标。
- 韧性分数**是一种报告机制，用于**衡量**服务对故障的韧性。



推荐“混沌工程实践”公众号





# Thanks

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开放运维联盟

荣誉出品