## Monkeys in Lab Coats

### Applied Failure Testing Research at



### The whole is greater than the sum of its parts.

- Aristotle [Metaphysics]

#### The Professor

**Peter Alvaro** 



### **The Practitioner**

#### Kolton Andrus

Ex-Berkeley, Ex-Industry

Assistant Prof @ Santa Cruz

Misses the calm of PhD life

Likes prototyping stuff

Ex-Netflix, Ex-Amazon

'Chaos' Engineer

Misses his actual pager

Likes breaking stuff

#### **Measures of Success**

#### Academic



H-Index

Grant warchest

Department ranking

Availability (i.e. 99.99% uptime)

Number of Incidents

Reduce Operational Burden

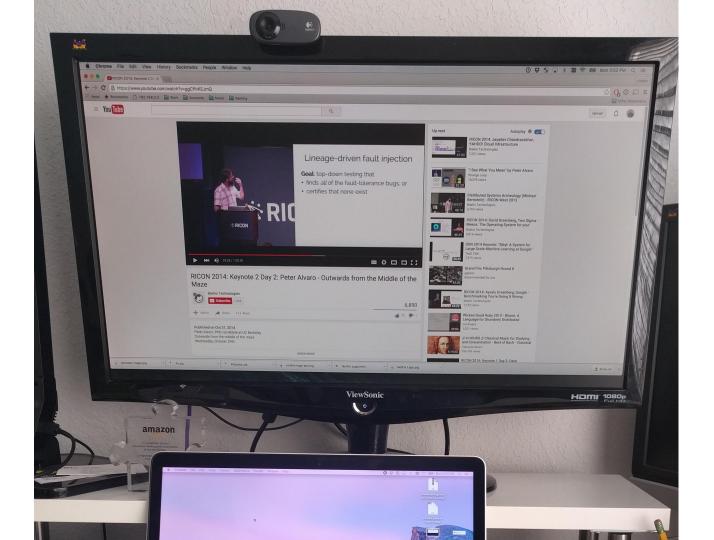
## An Unlikely Team?



## Works Great!

but ... it's manual

## Surely there is a better way ...



### Free lunch?

## The End?

(Academia + Industry)

## Let's build it

"Can we, pretty please?"

## Freedom and Responsibility

## **NETFLIX** Core Value

#### Responsibility

Academic

Industry

#### Prove that it works

Show that it scales

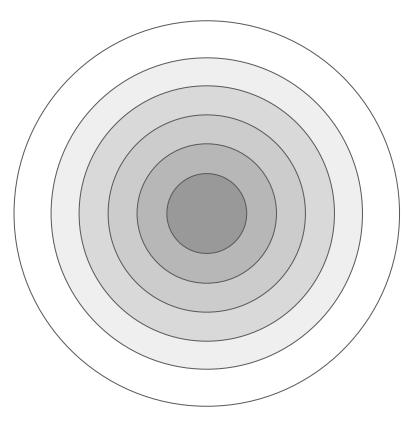
Find real bugs

## The Big Idea

## Lineage Driven Fault Injection

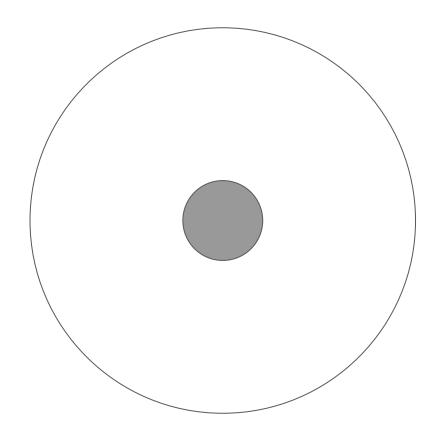
#### What could *possibly* go wrong?

Consider computation involving 100 services



Search Space: 2<sup>100</sup> executions

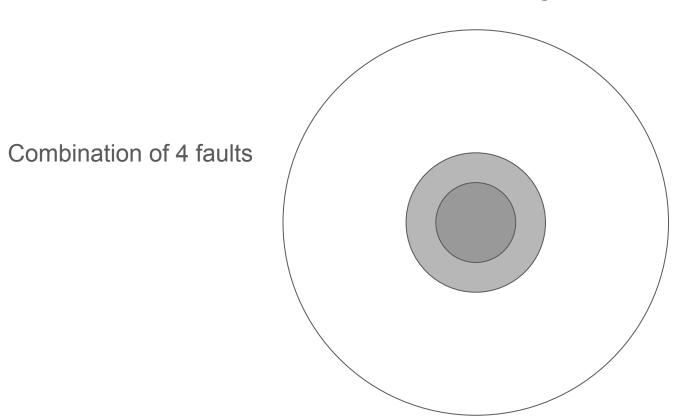
#### "Depth" of bugs



Single Faults

Search Space: 100 executions

#### "Depth" of bugs



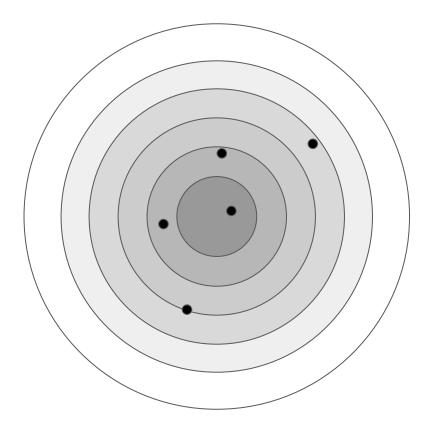
Search Space: 3M executions

#### "Depth" of bugs

Combination of 7 faults

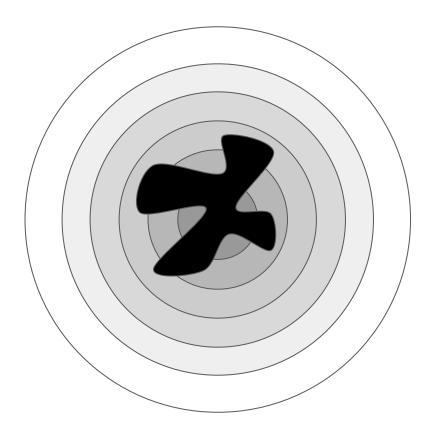
Search Space: 16B executions

#### Random Search



Search Space:  $2^{100}$  executions

#### **Engineer-guided Search**



Search Space: ???

## Fault-tolerance "is just" redundancy

#### **Lineage-driven Fault Injection**

Peter Alvaro UC Berkeley palvaro@cs.berkeley.edu Joshua Rosen UC Berkeley rosenville@gmail.com Joseph M. Hellerstein UC Berkeley hellerstein@cs.berkeley.edu

#### ABSTRACT

Failure is always an option; in large-scale data management systems, it is practically a certainty. Fault-tolerant protocols and components are notoriously difficult to implement and debug. Worse still, choosing existing fault-tolerance mechanisms and integrating them correctly into complex systems remains an art form, and programmers have few tools to assist them.

We propose a novel approach for discovering bugs in fault-tolerant data management systems: *lineage-driven fault injection*. A lineagedriven fault injector reasons *backwards* from correct system outcomes to determine whether failures in the execution could have prevented the outcome. We present MOLLY, a prototype of lineagedriven fault injection that exploits a novel combination of data lineage techniques from the database literature and state-of-the-art satisfiability testing. If fault-tolerance bugs exist for a particular configuration, MOLLY finds them rapidly, in many cases using an order of magnitude fewer executions than random fault injection. Otherwise, MOLLY certifies that the code is bug-free for that configuration. enriching new system architectures with well-understood fault tolerance mechanisms and henceforth assuming that failures will not affect system outcomes. Unfortunately, fault-tolerance is a *global* property of entire systems, and guarantees about the behavior of individual components do not necessarily hold under composition. It is difficult to design and reason about the fault-tolerance of individual components, and often equally difficult to assemble a faulttolerant system even when given fault-tolerant components, as witnessed by recent data management system failures [16, 57] and bugs [36, 49].

Top-down testing approaches—which perturb and observe the behavior of complex systems—are an attractive alternative to verification of individual components. Fault injection [1,26, 36, 44, 59] is the dominant top-down approach in the software engineering and dependability communities. With minimal programmer investment, fault injection can quickly identify shallow bugs caused by a small number of independent faults. Unfortunately, fault injection is poorly suited to discovering rare counterexamples involving complex combinations of multiple instances and types of faults (e.g., a network partition followed by a crash failure). Ap-

#### But how do we know redundancy when we see it?

Hard question: "Could a bad thing ever happen?"

Easier: "Exactly why did a good thing happen?"

"What could have gone wrong?"

#### Lineage-driven fault injection

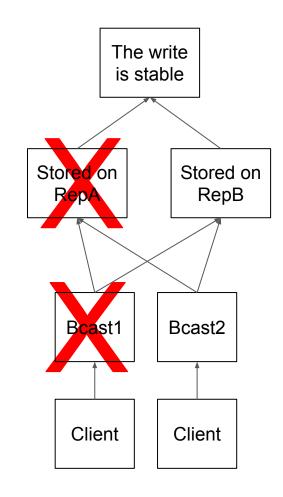
Why did a good thing happen?

Consider its lineage.

What could have gone wrong?

Faults are *cuts* in the lineage graph.

Is there a cut that breaks all supports?



#### Lineage-driven fault injection

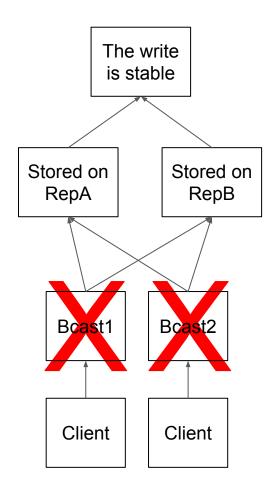
Why did a good thing happen?

Consider its *lineage*.

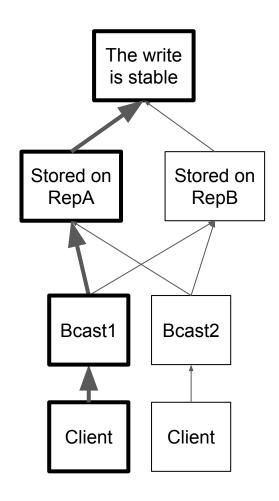
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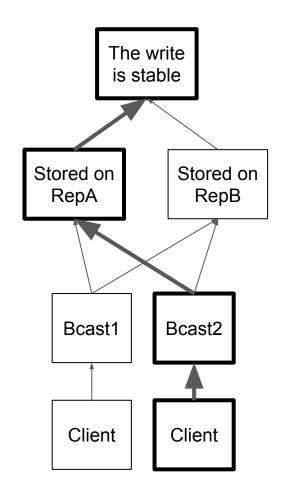


(RepA OR Bcast1)



(RepA OR Bcast1)

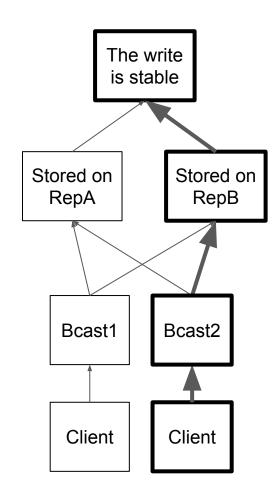
#### AND (RepA OR Bcast2)



(RepA OR Bcast1)

AND (RepA OR Bcast2)

AND (RepB OR Bcast2)

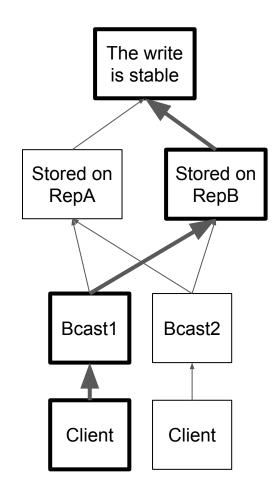


(RepA OR Bcast1)

AND (RepA OR Bcast2)

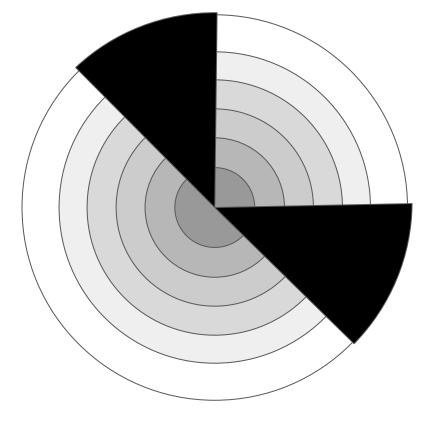
AND (RepB OR Bcast2)

AND (RepB OR Bcast1)



#### **Search Space Reduction**

Each Experiment finds a bug, OR

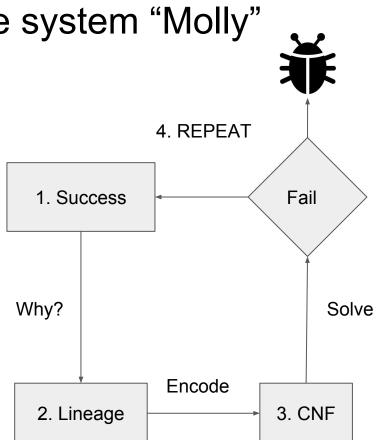


Reduces the Search space

### The prototype system "Molly"

Recipe:

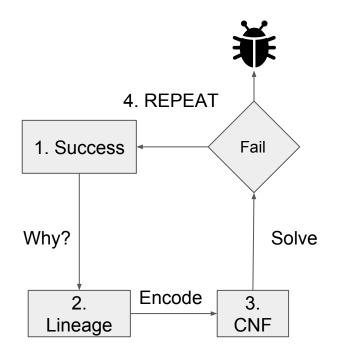
- 1 Start with a successful outcome. Work backwards.
- 2. Ask *why* it happened: Lineage
- 3. Convert lineage to a boolean formula and solve
- Lather, rinse, repeat 4.



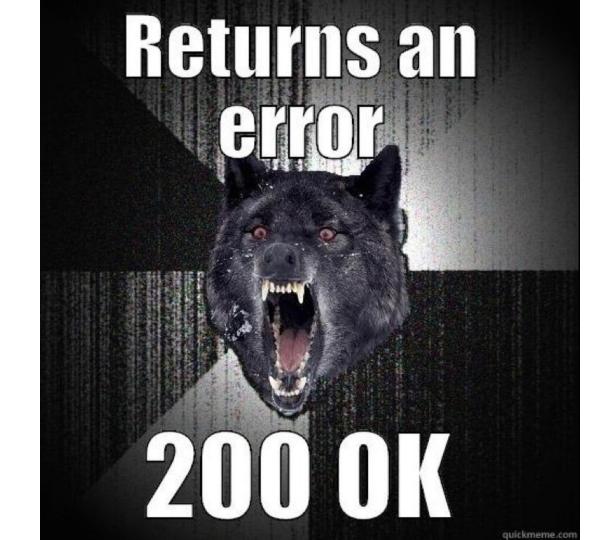
## The Big Idea

### **Meets Production**

## 1. Start with a successful outcome



### What is success?

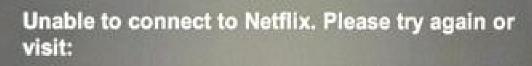


# "Start with the customer and work backwards"

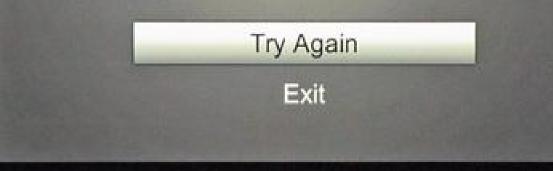




# NETFLIX



www.netflix.com/tvhelp



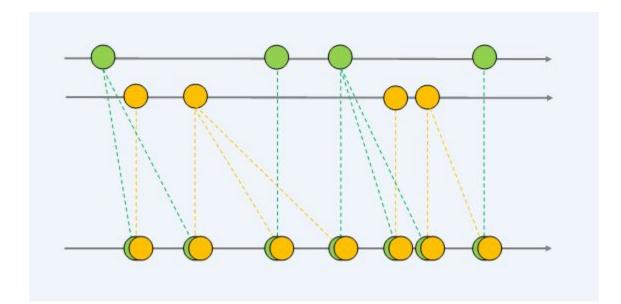
NETFLIX

ui-200

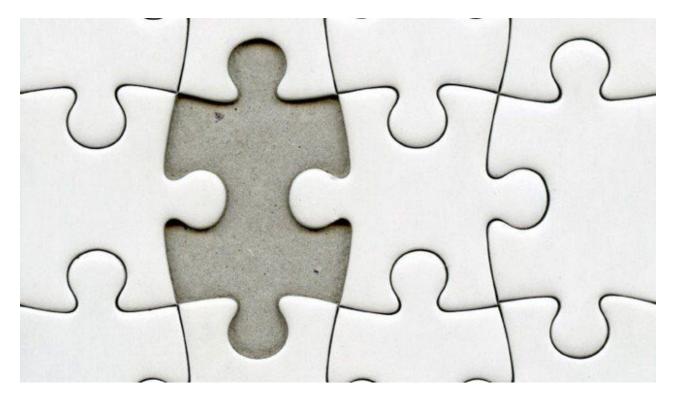
## "Streaming" Data



## Joining the Streams



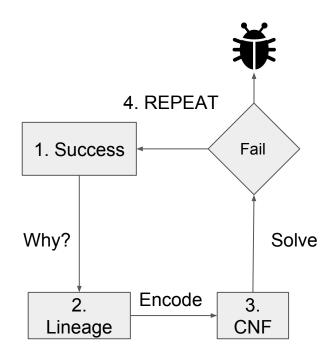
## Missing Data?



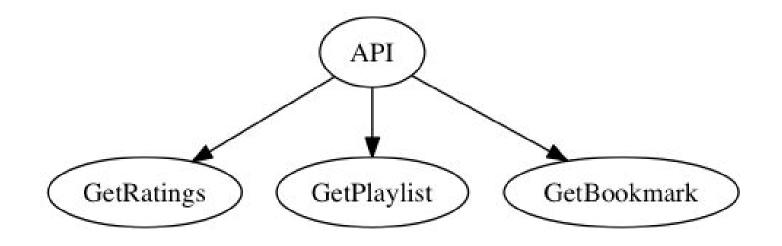
# Lesson 1

# Work backwards from what you know

# 2. Ask why it happened

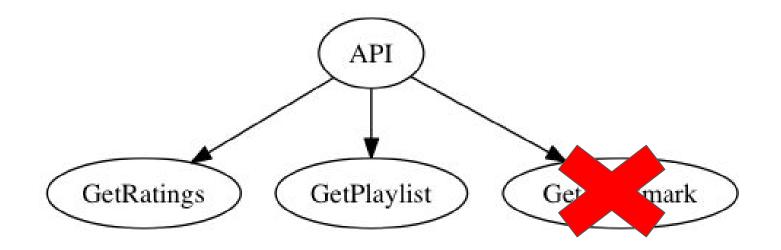


## **Request Tracing**

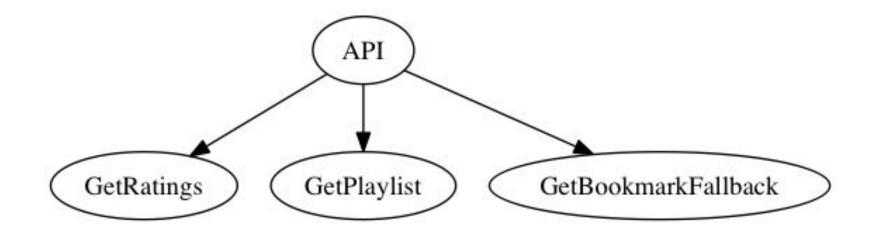




### **Request Tracing**

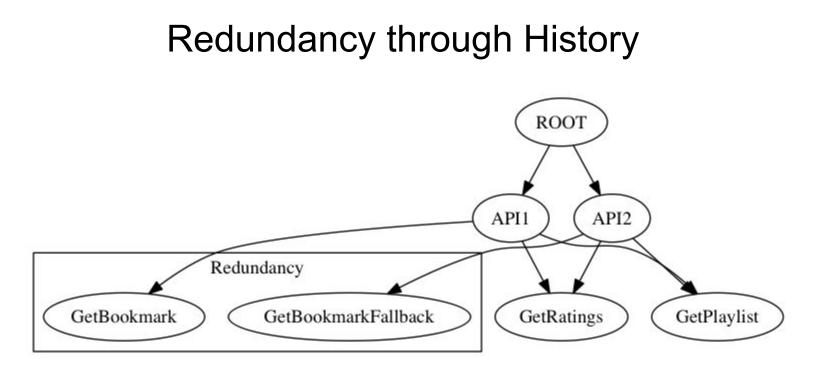


#### **Alternate Execution**



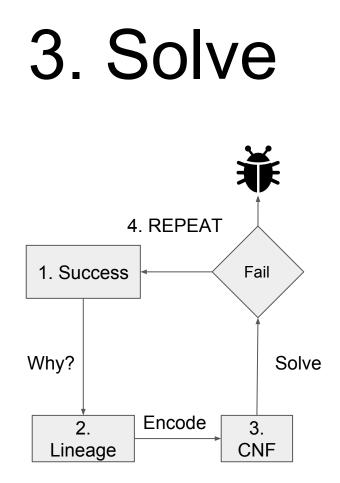
#### Evolution over time



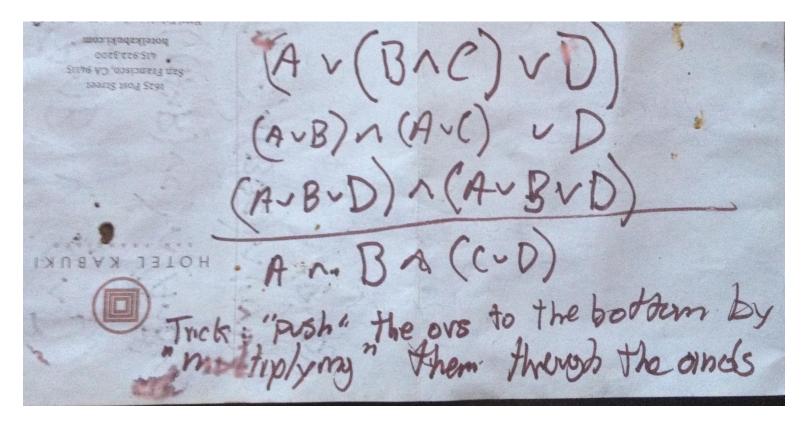


# Lesson 2

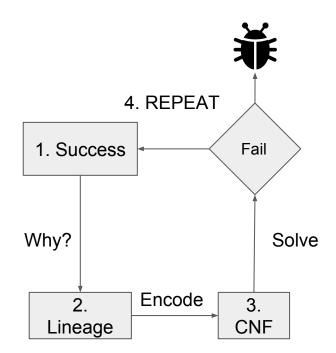
# Meet in the middle



#### A "small" matter of code



# 4. Lather, Rinse, Repeat





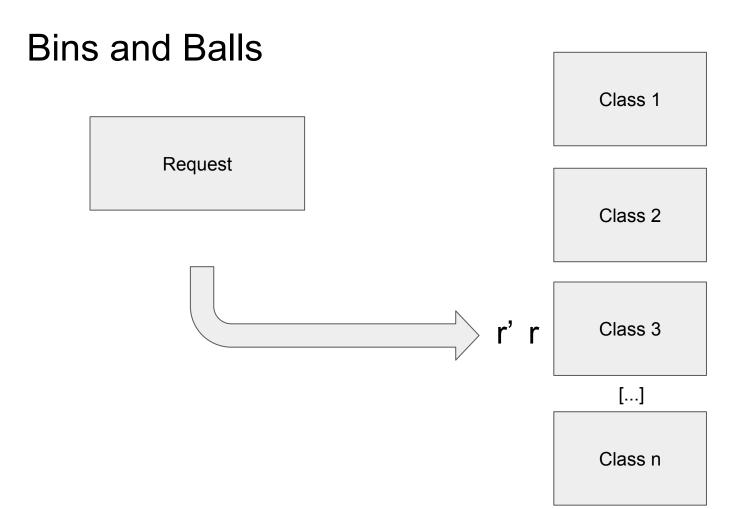
## Turn the crank, right?

# Idempotence

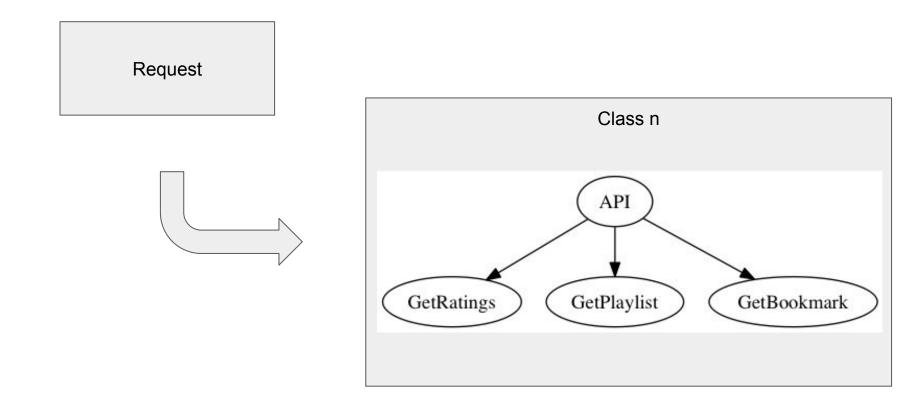




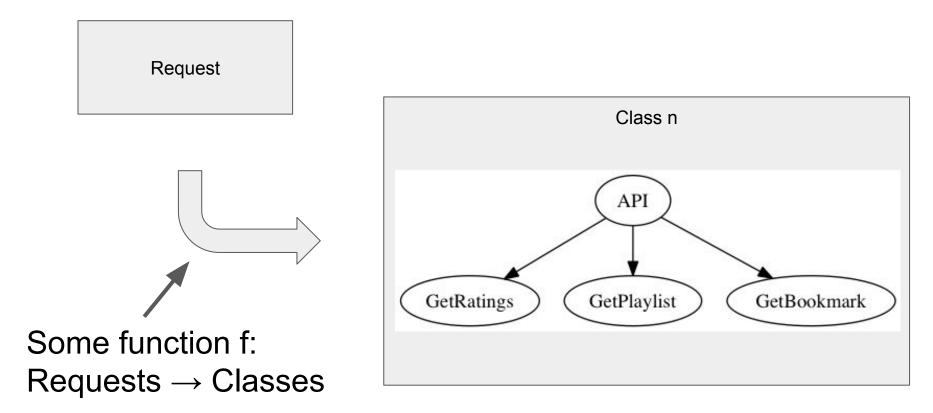




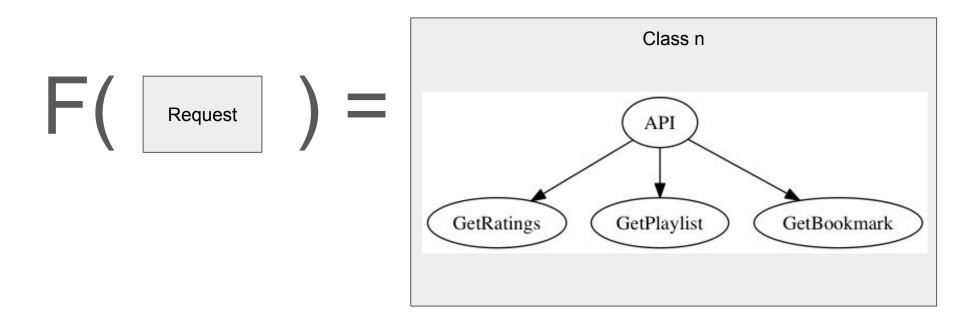
## **Predicting Request Graphs**

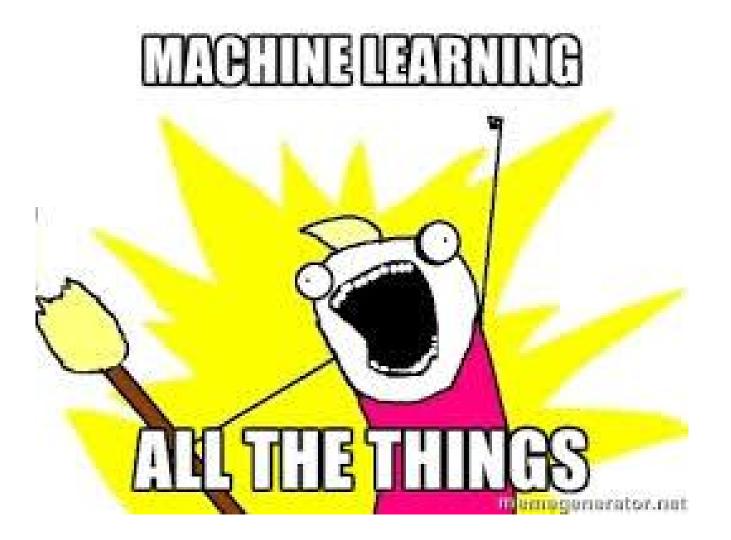


## **Predicting Request Graphs**



## **Predicting Request Graphs**





# Solve the Machine Learning problem? or the Failure Testing one?

# Simplest thing that will work?

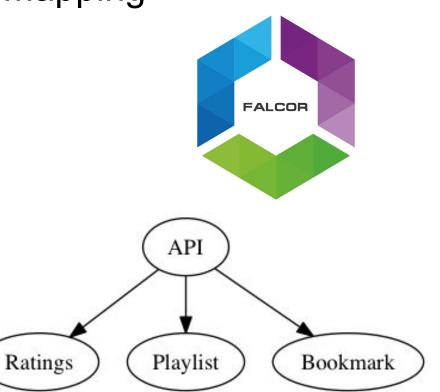
#### **Falcor Path Mapping**

["bookmarks", "recent"]
["playlist", 0, "name"]

["ratings"]

=>

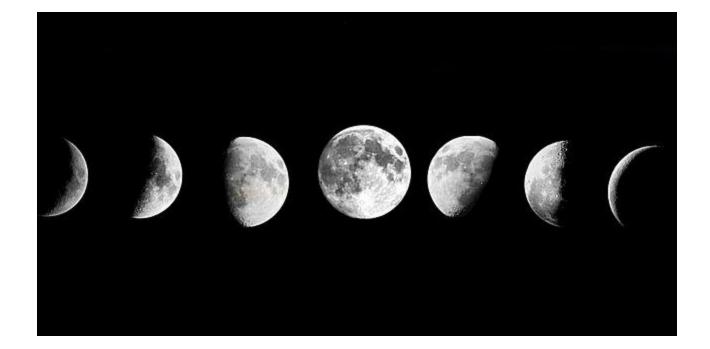
"bookmarks,playlist,ratings"



# Lesson 3

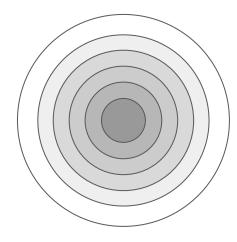
# Adapt the theory to the reality

### Many moons passed...



# Does it work?

YES!



# Case study: "Netflix AppBoot"

Services	~100
Search space (executions)	$2^{100}  (1,000,000,000,000,000,000,000,000,000,0$
Experiments performed	200
Critical bugs found	6

#### **Future Work**

Search prioritization

**Richer lineage collection** 

Exploring temporal interleavings

**Richer device metrics** 

Request class creation

Better experiment selection

# Lessons

# Work backwards from what you know Meet in the middle

# Adapt the theory to the reality

# Academia + Industry

# Academia + Industry

# Academia X Industry

## Thank You!

## **Peter Alvaro**

## **Kolton Andrus**

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@KoltonAndrus

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kolton@gremlininc.com

### References

- Netflix Blog on 'Automated Failure Testing' <u>http://techblog.netflix.</u> <u>com/2016/01/automated-failure-testing.html</u>
- Netflix Blog on 'Failure Injection Testing' <u>techblog.netflix.com/2014/10/fit-</u> <u>failure-injection-testing.html</u>
- 'Lineage Driven Fault Injection' <u>http://people.ucsc.edu/~palvaro/molly.pdf</u>

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