

# Check before You Change: Preventing Correlated Failures in Service Updates

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Bingchuan Tian



Bo Song Haoliang Zhang



# Background

- Cloud services ensure reliability by redundancy:
  - Storing data redundantly
  - Replicating service states across multiple nodes
- Examples:
  - Amazon AWS, AliCloud, Google Cloud, etc. replicate their data and service states

# However, cloud outages still occur

 Correlated failures resulting from EBS  
due to bugs in one FRS see Rackspace Outage Nov 12th

5、阿里云故障 6月27日

Summary

事故

We'd like to thank our customers for their continued support and understanding during this time. We're sorry for any inconvenience caused.

Google Outage

By Lawrence Abramson

Why redundancy does not help?



# An AWS Outage in 2018

## AWS outage killed some cloudy servers, recovery time is uncertain

'Power event' blamed, hit subset of kit in US-EAST-1

By [Simon Sharwood](#) 1 Jun 2018 at 00:48

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**Updated** Parts of Amazon Web Services' US-East-1 region have experienced about half an hour of downtime, but some customers' instances and data can't be restored because the hardware running them appears to have experienced complete failure.

The cloud colossus' [status page](#) reports an investigation of "connectivity issues affecting some instances in a single Availability Zone in the US-EAST-1 Region" as of 3:13 PM PDT on Thursday, May 31.

A 3:42 PM update confirmed "an issue in one of the datacenters that makes up one of US-EAST-1 Availability Zones. This was a result of a power event impacting a small percentage of the physical servers in that datacenter as well as some of the networking devices."

**Elastic Compute Cloud (EC2)**



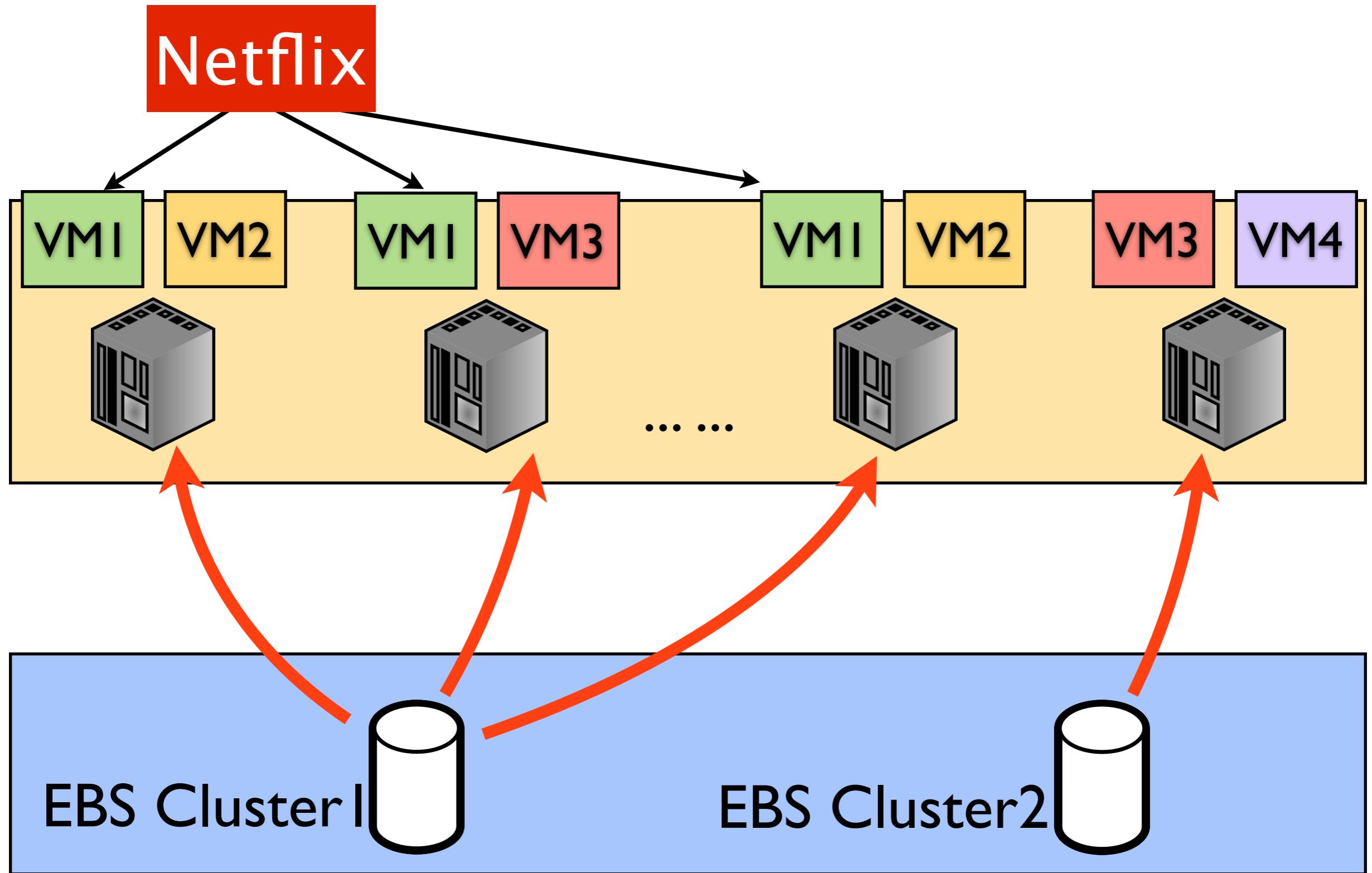
**Elastic Block Store (EBS)**

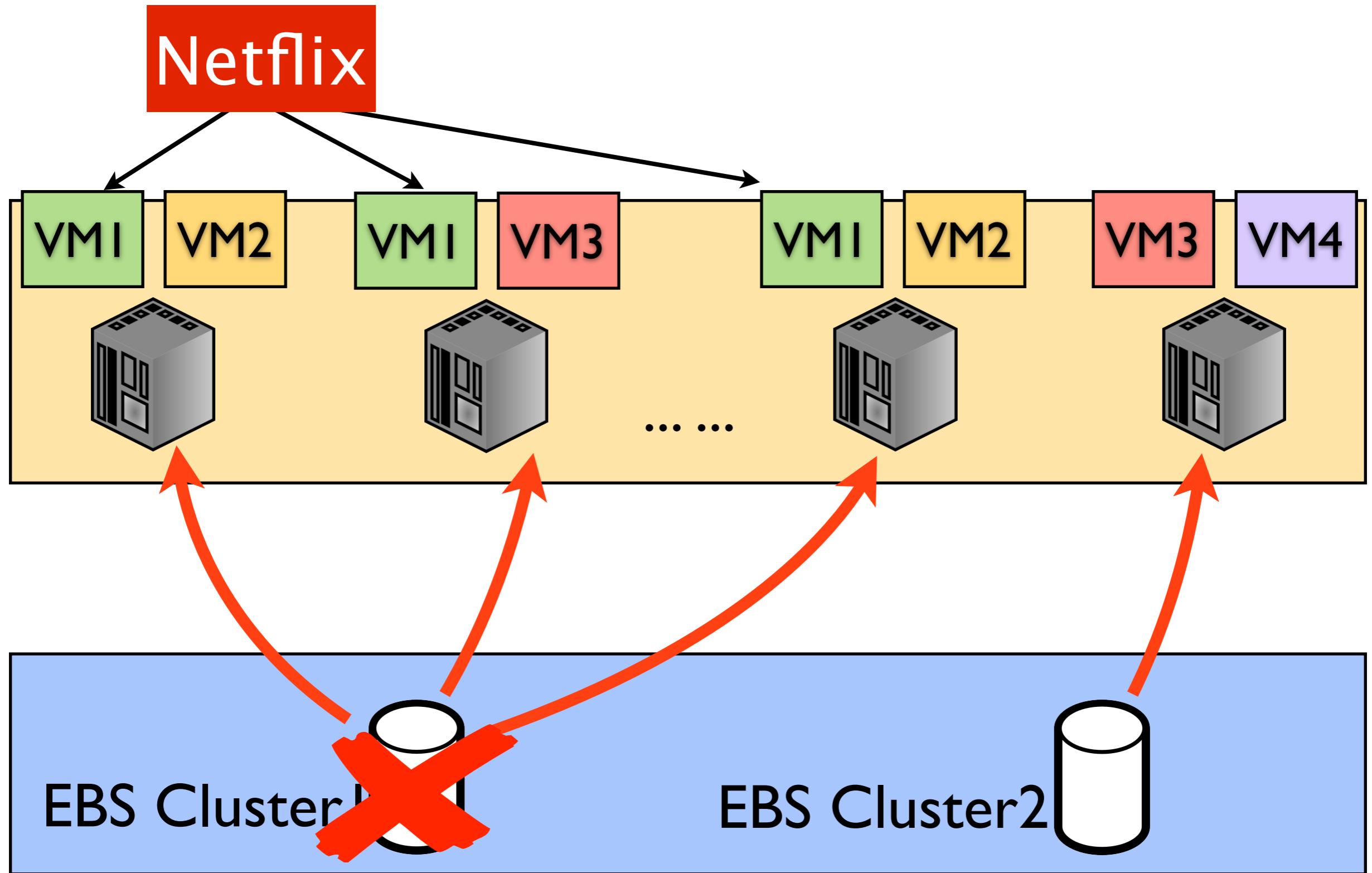
# Elastic Compute Cloud (EC2)

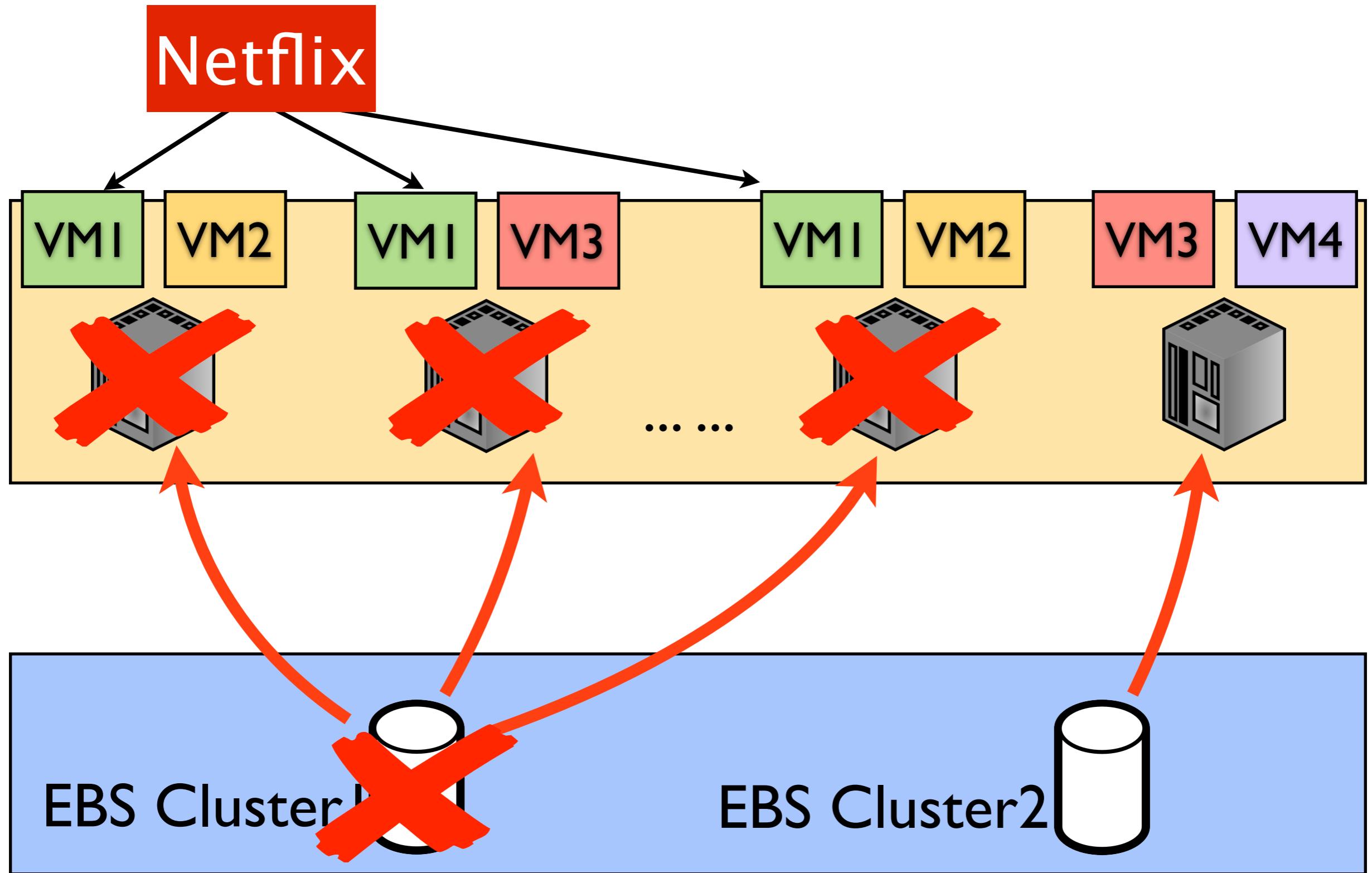


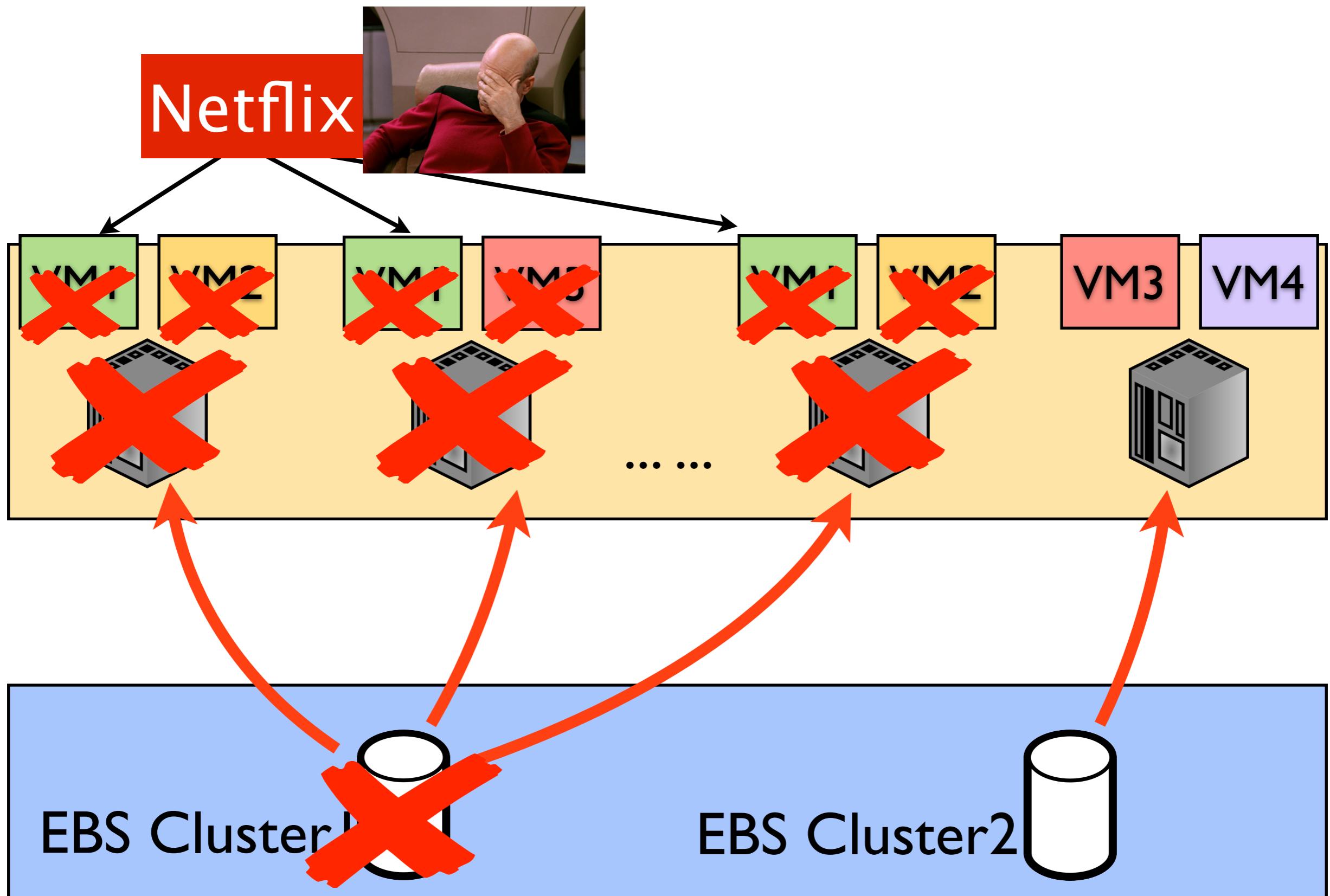
EBS Cluster1

EBS Cluster2









Netflix



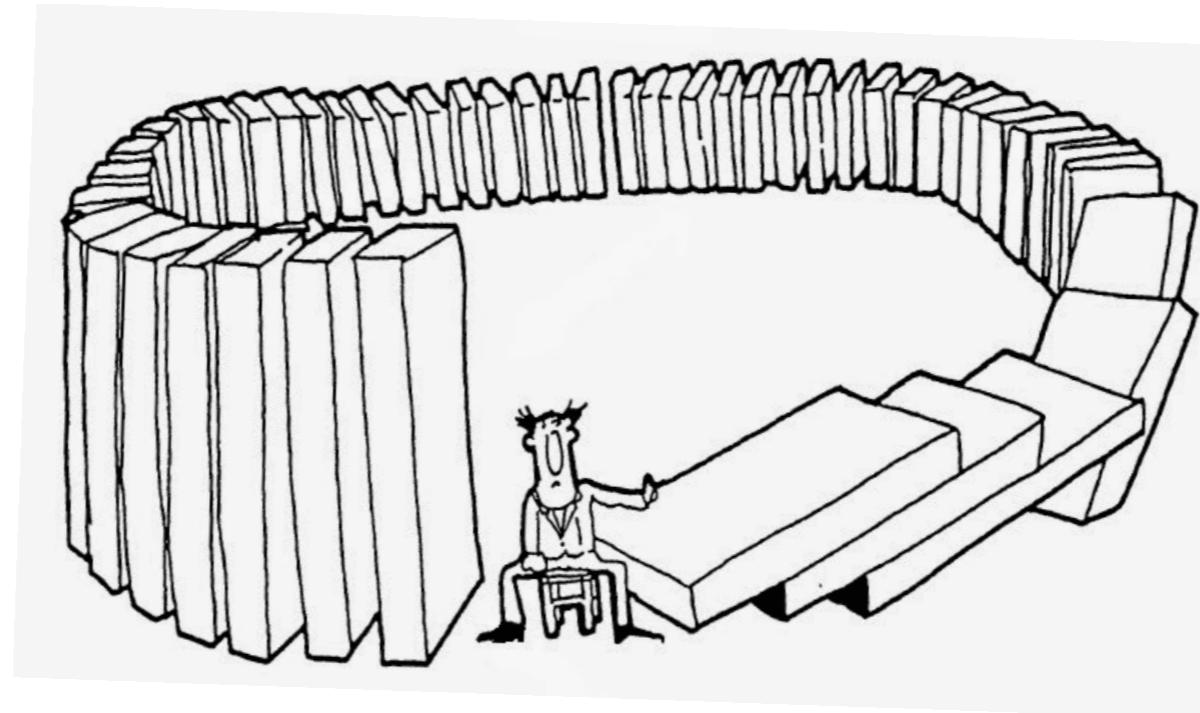
Correlated failures resulting from  
deep dependencies

EBS Cluster1

EBS Cluster2

# Correlated Failures

- Correlated failures are harmful and epidemic:
  - Propagated to all the redundant instances
  - Undermine redundancy and fault tolerance efforts



# Correlated failures are prevalent

## Why Does the Cloud Stop Computing?

AWS Database Blog

### Amazon Aurora under the hood: quorums and correlated failure

by Anurag Gupta | on 14 AUG 2017 | in [Amazon Aurora\\*](#), [Aurora](#), [Database\\*](#), [MySQL Compatible\\*](#), [PostgreSQL Compatible\\*](#) | [Permalink](#) | [Comments](#) |

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*Anurag Gupta runs a number of AWS database services, including Amazon Aurora, which he helped design. In this Under the Hood series, Anurag discusses the design considerations and technology underpinning Aurora.*

Amazon Aurora storage is a highly distributed system that needs to meet the stringent performance, availability, and durability requirements of a high-end relational database. This post is the first of a four-part series that covers some of the key elements of our design.

There isn't a lot of publicly available material discussing tradeoffs in real-world durability, availability, and performance at scale. Although this series is based on the considerations involved in designing a transactional database, I believe it should be relevant to anyone architecting systems involving the coordination of mutable distributed state.

Suppose you have a clustered pair of servers and each is 99.99% reliable. What's the reliability of the pair? "Obviously," 99.99%.

notified of the outage by our monitoring tools and were in constant contact with Rackspace during the outage working to resolve as quickly as possible.

REW apologizes for this outage; we promise that we are putting Rackspace's feet to the fire to ensure maximum uptime for our customers!

Here is the incident report from Rackspace if you want the techy details:

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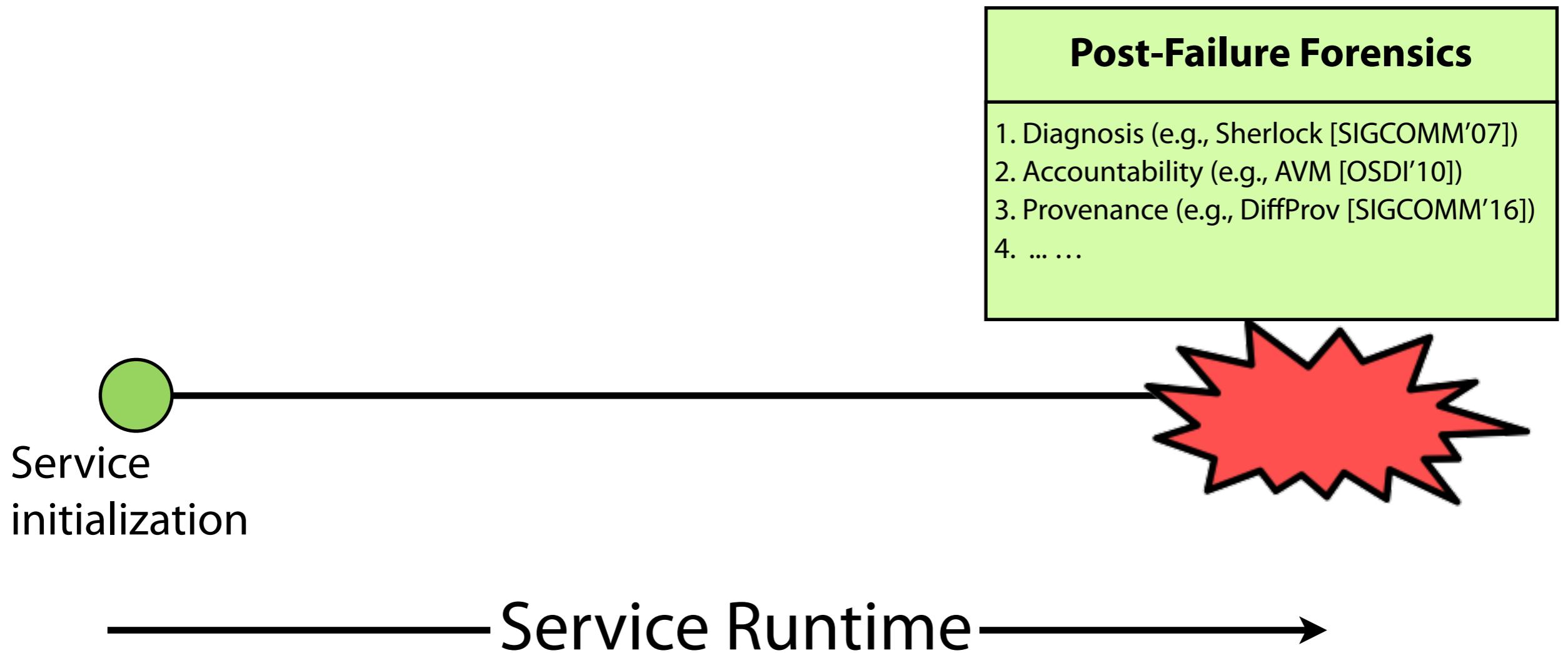
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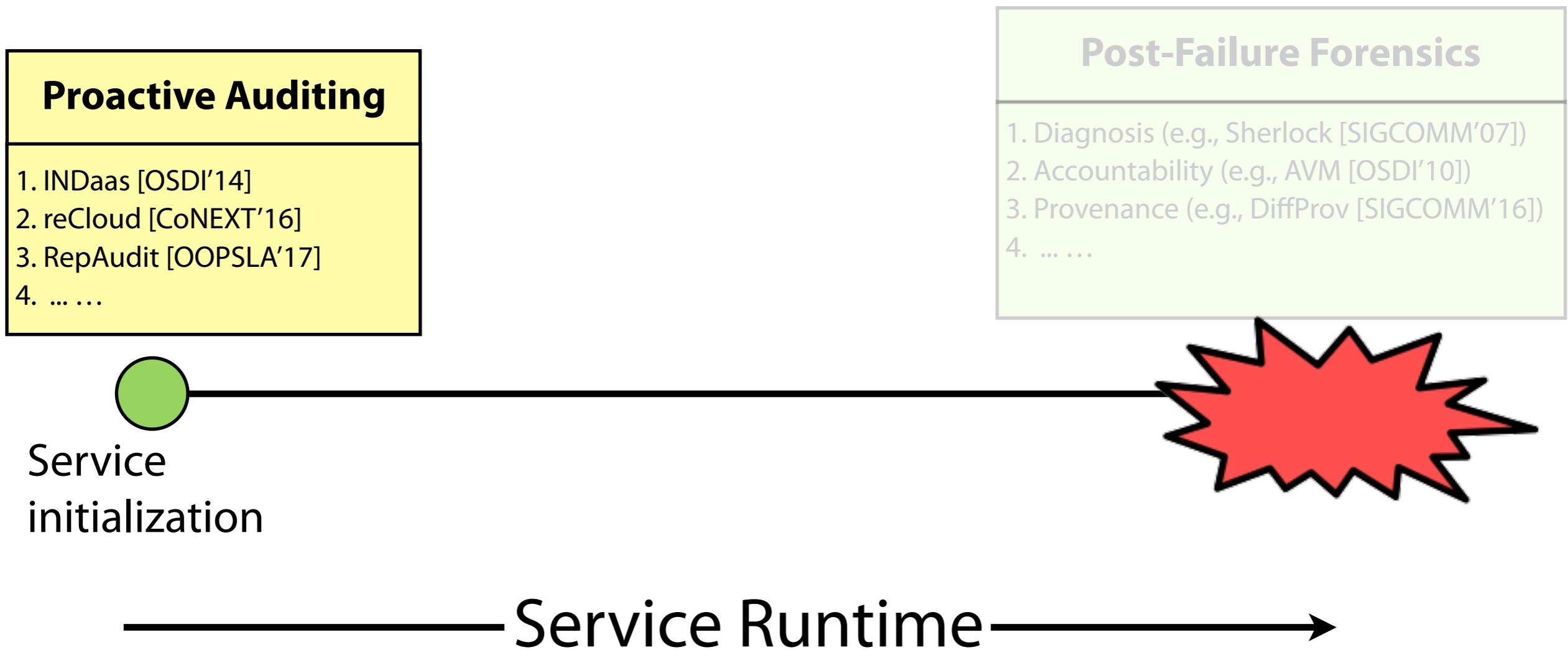
# State of the Art



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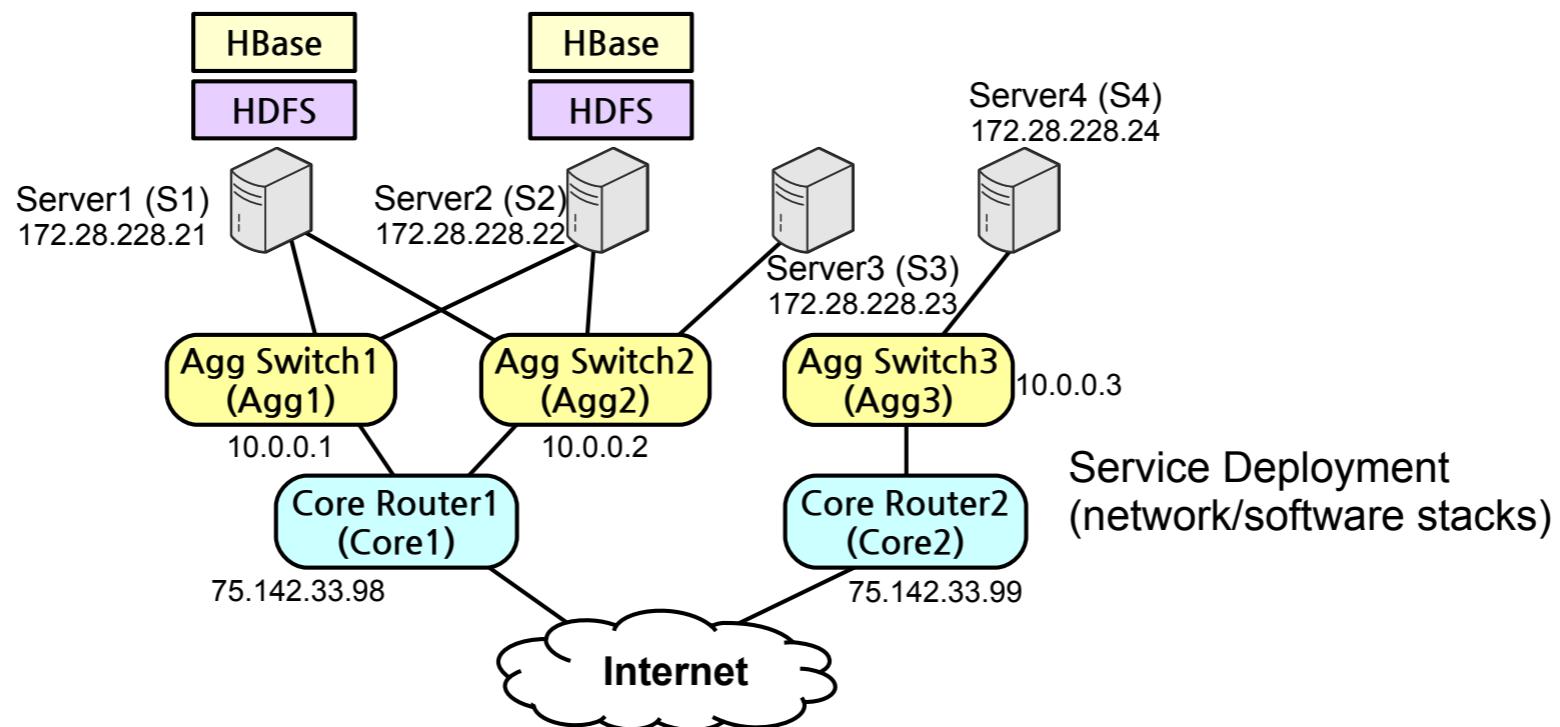


# State of the Art



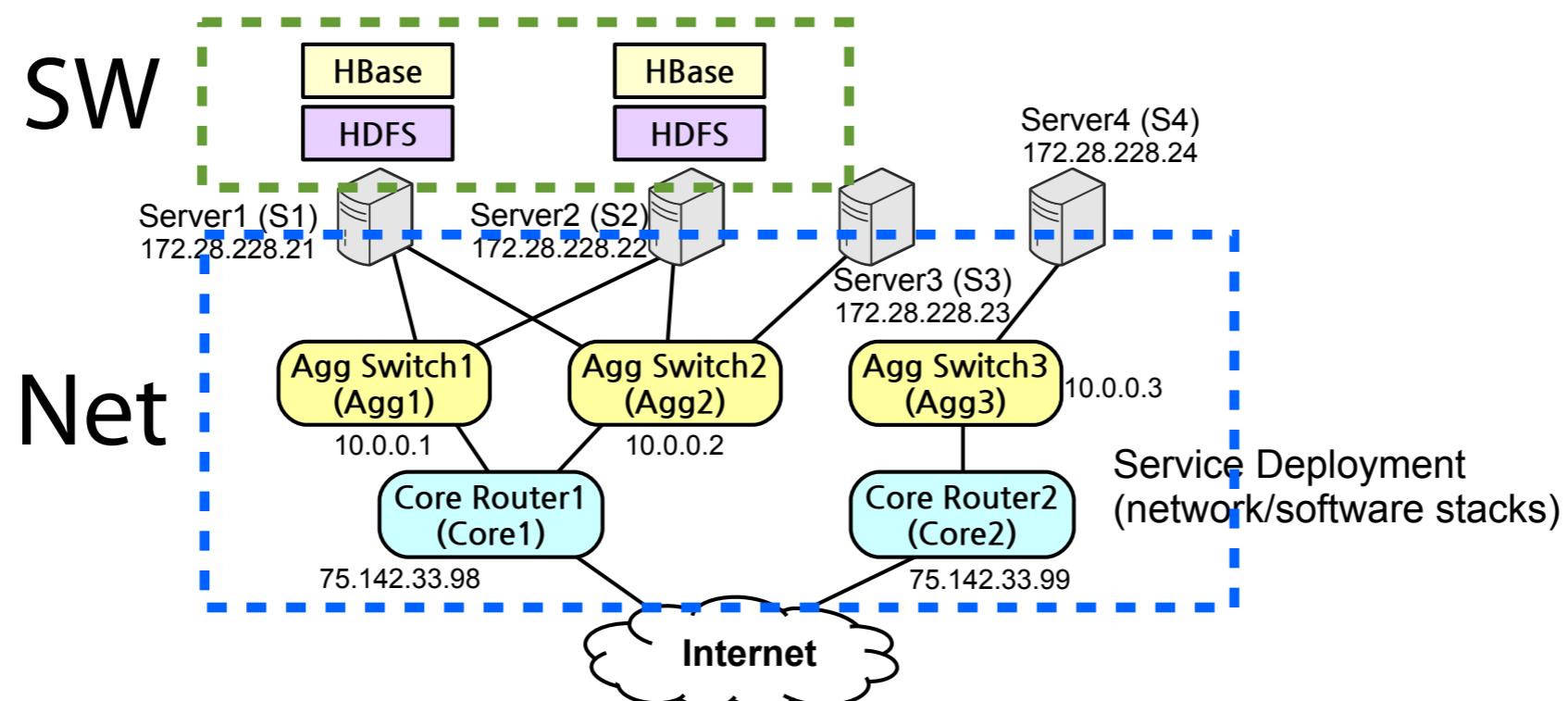
# Proactive Auditing

- They did pre-deployment recommendations:



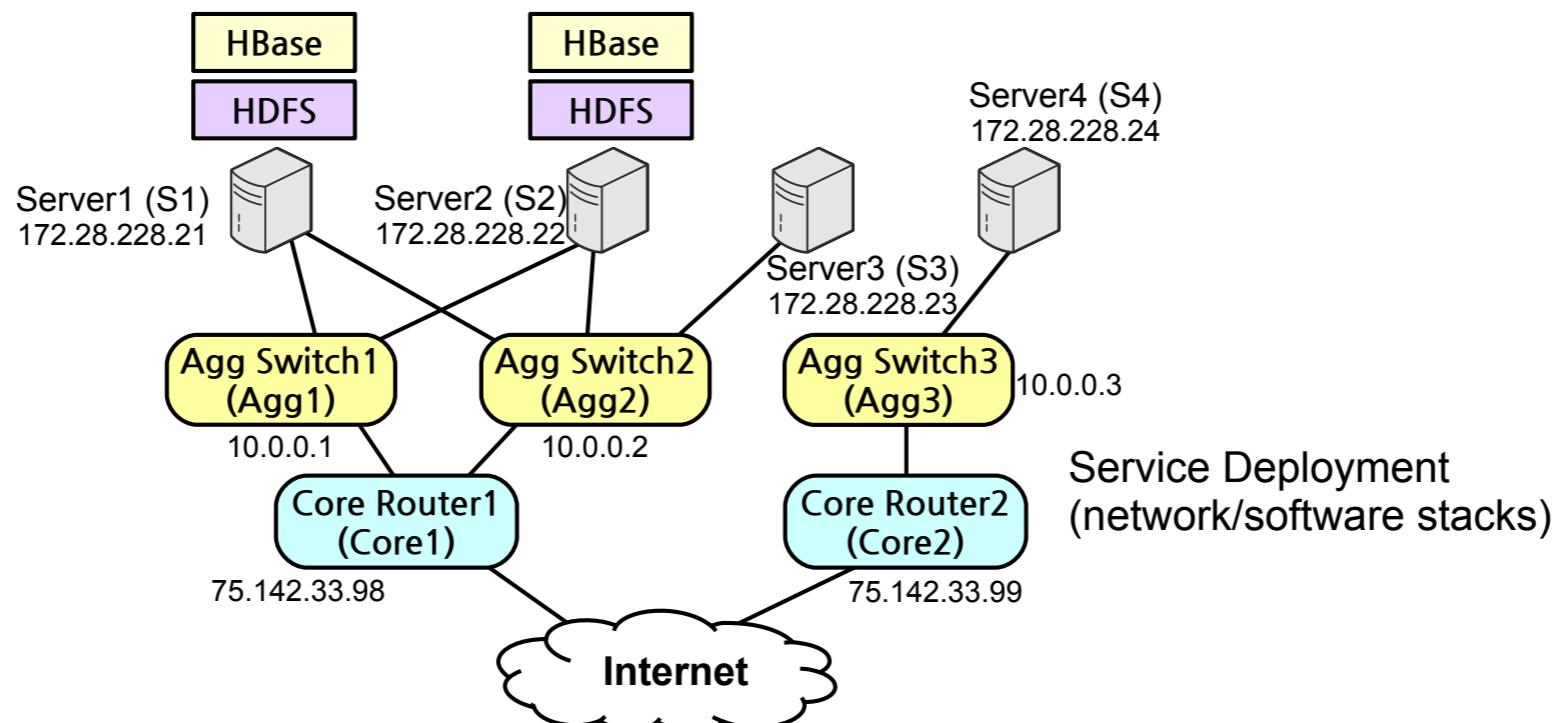
# Proactive Auditing

- They did pre-deployment recommendations:
  - Step1: Automatically collecting dependency data



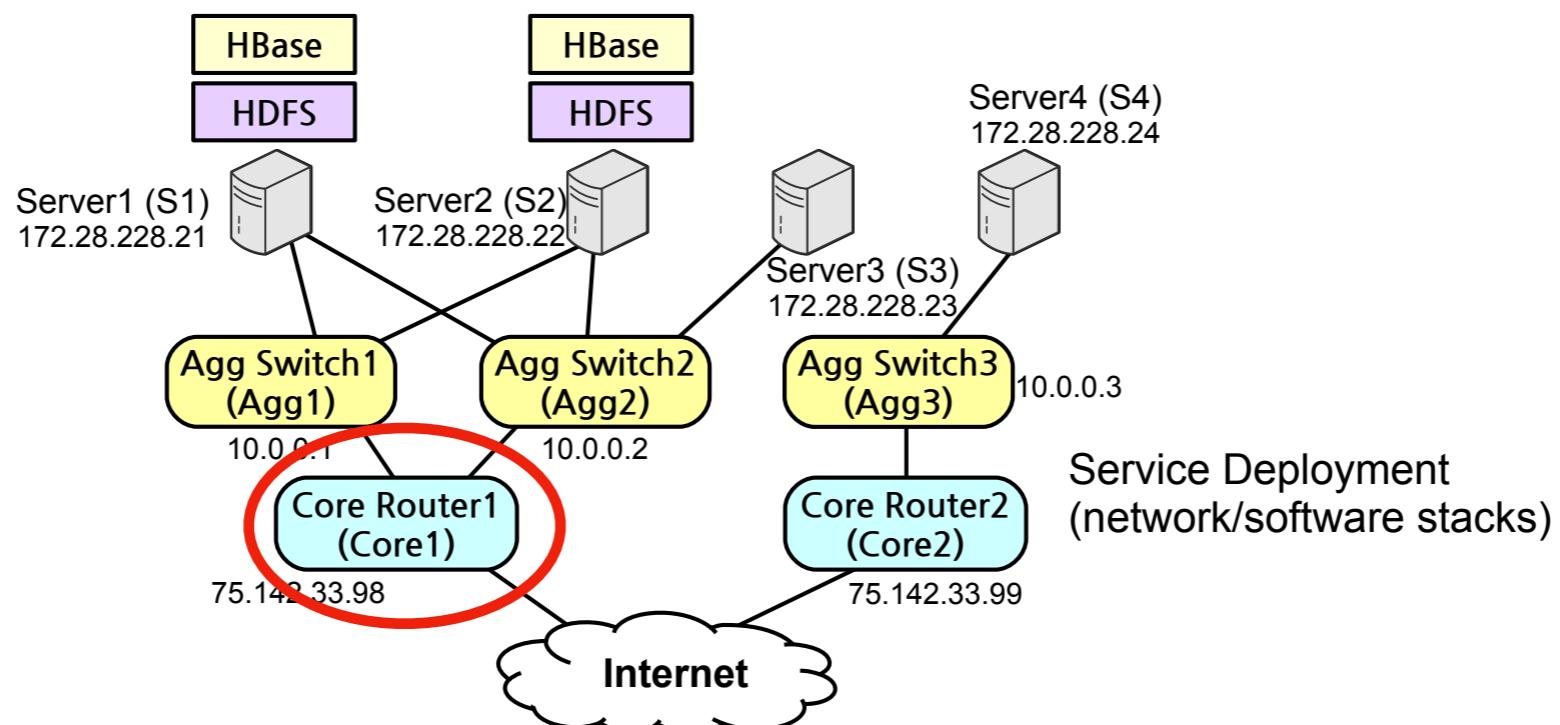
# Proactive Auditing

- They did pre-deployment recommendations:
  - Step1: Automatically collecting dependency data
  - Step2: Modeling system stack in fault graph



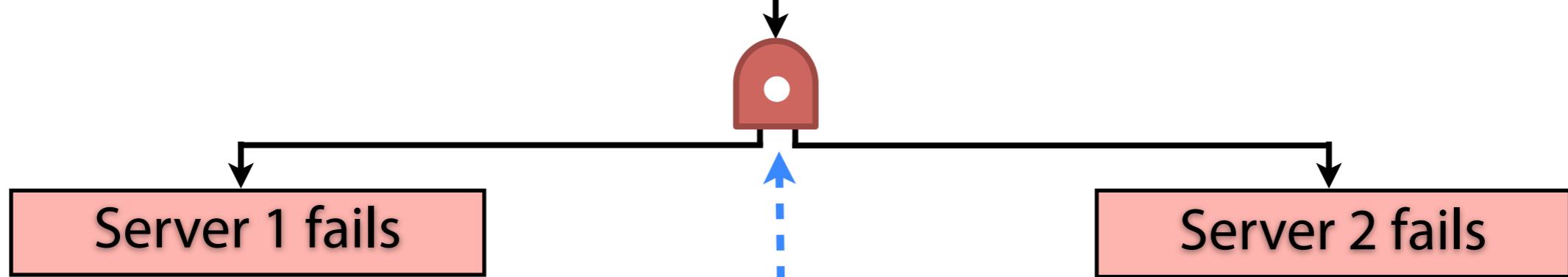
# Proactive Auditing

- They did pre-deployment recommendations:
  - Step1: Automatically collecting dependency data
  - Step2: Modeling system stack in fault graph
  - Step3: Evaluating alternative deployments' independence



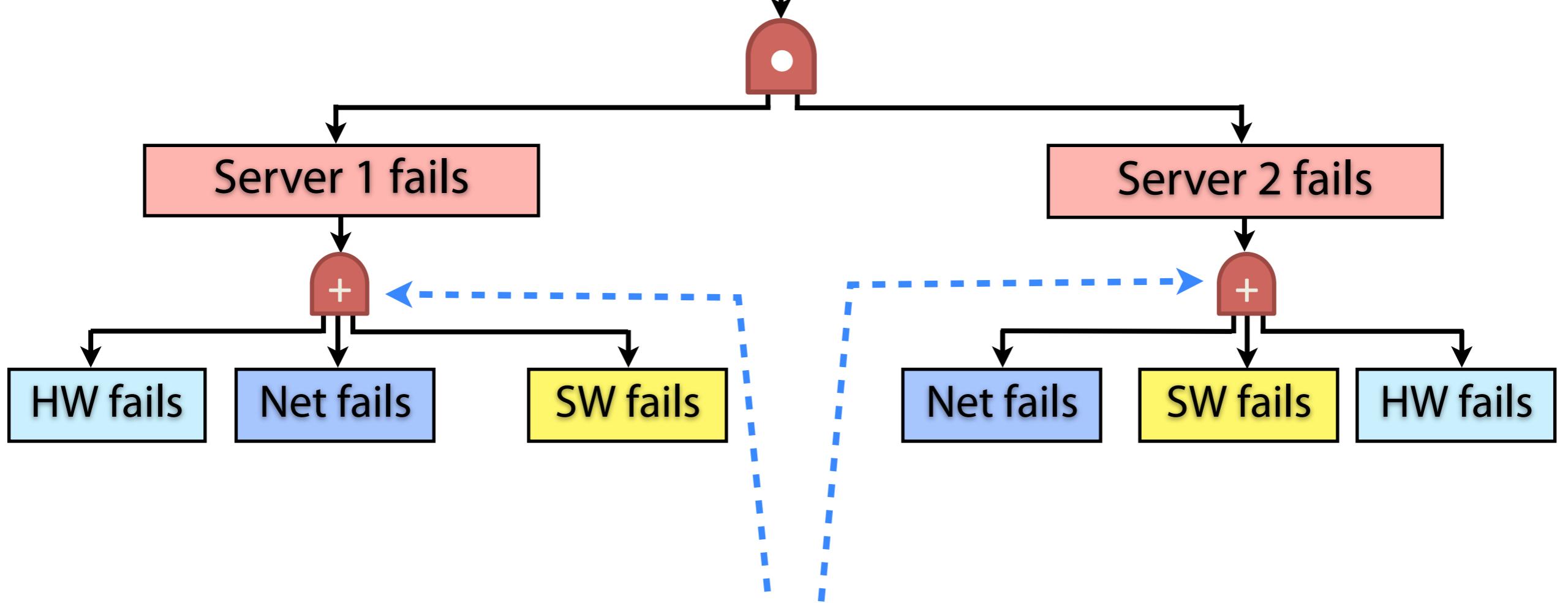
**Redundancy configuration fails**

Redundancy configuration fails



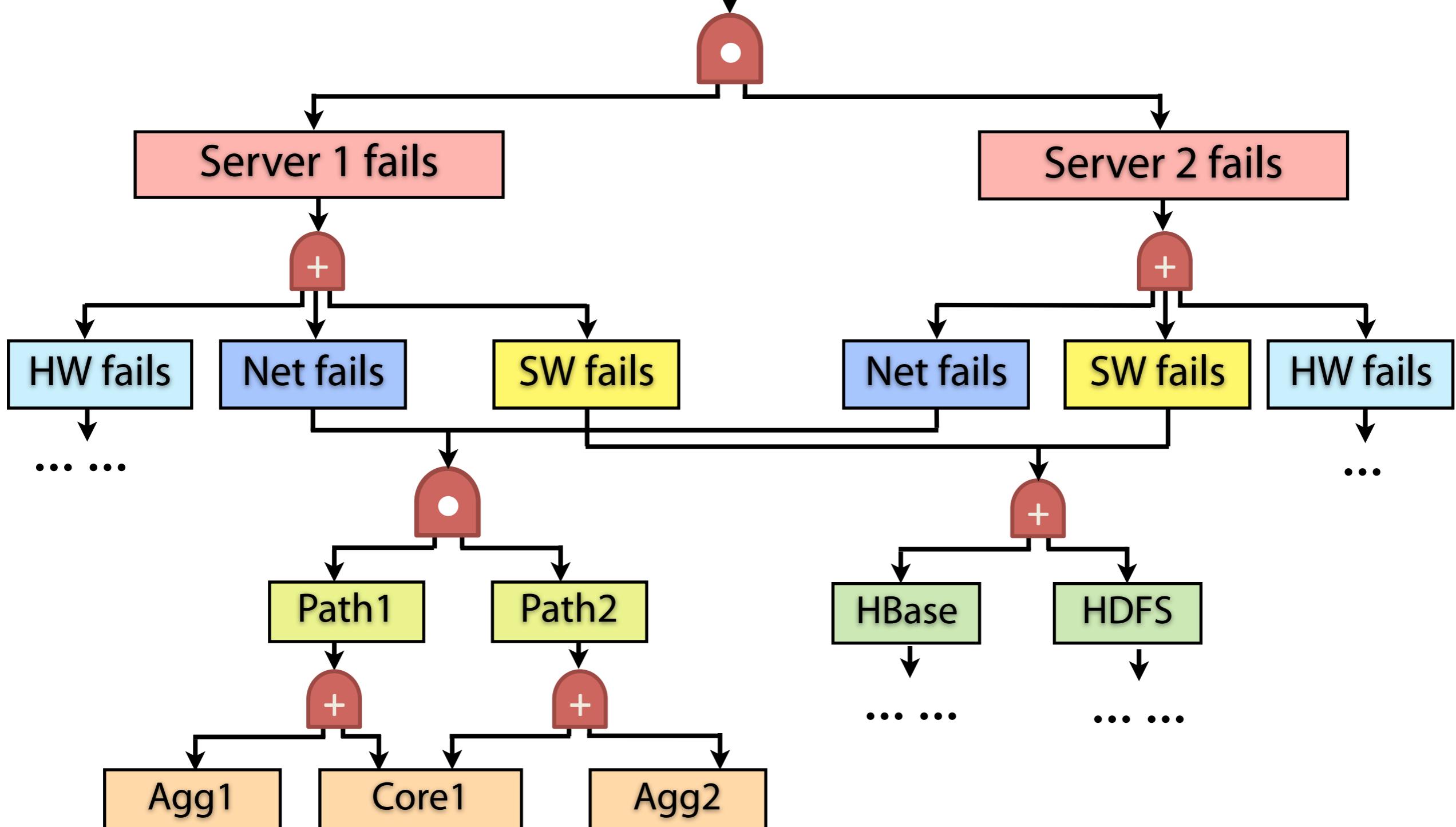
AND gate: all the  
sublayer nodes fail, the  
upper layer node fails

## Redundancy configuration fails

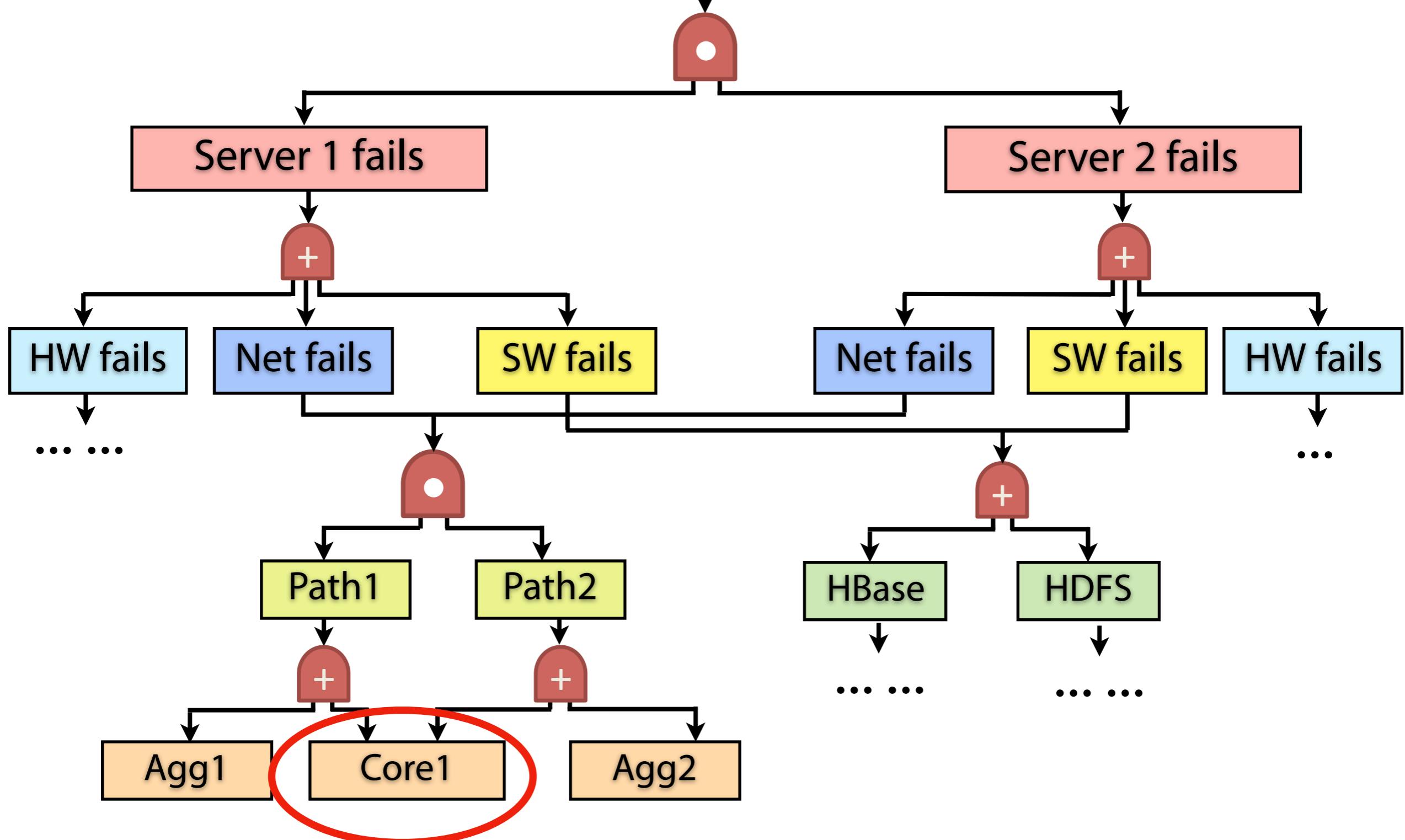


OR gate: one of the sublayer nodes fails,  
the upper layer node fails

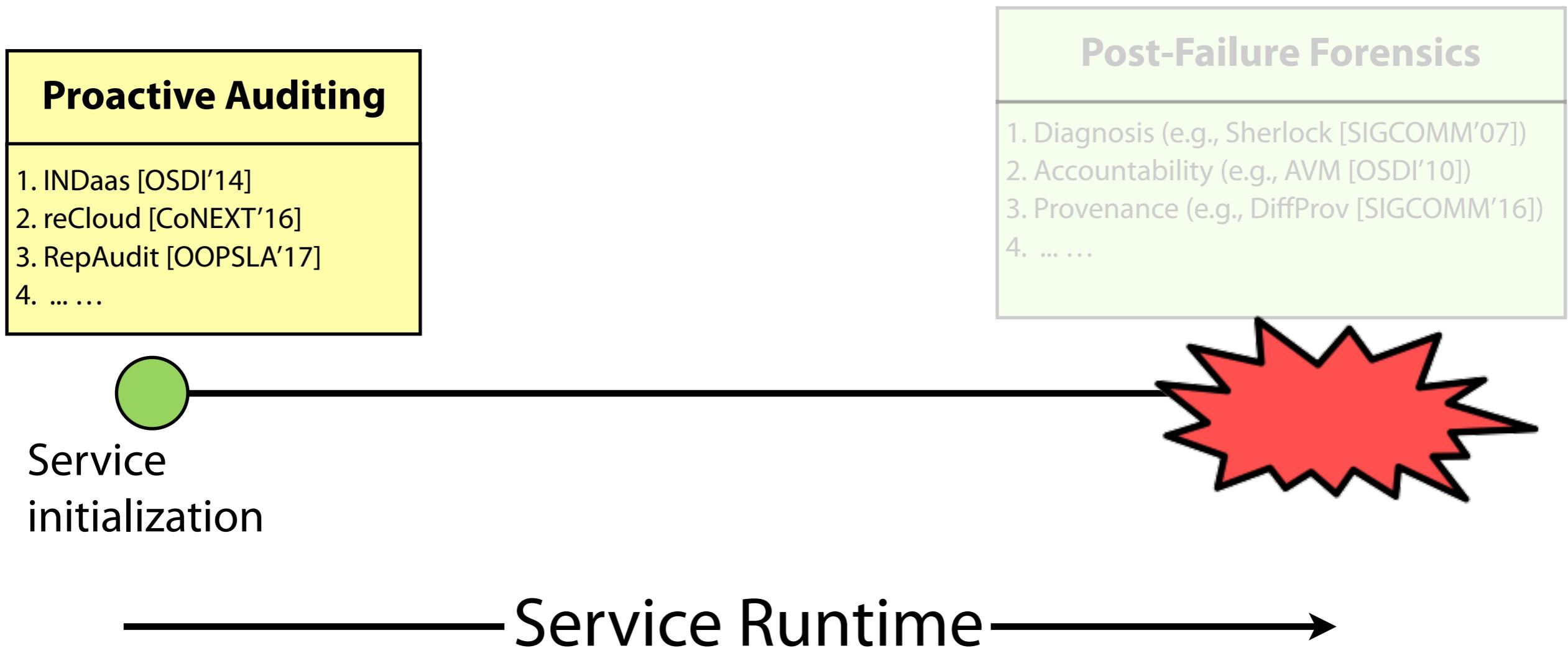
# Redundancy configuration fails



# Redundancy configuration fails



# State of the Art



# Correlated Failure Risks in Updates

## Proactive Auditing

1. Identify  
2. review  
3. Review  
4. ...  
**Azure global outage  
mangled domain re**



Microsoft Azure and Office 365 downed by DNS configuration blunder

## Google Outage in Eastern U.S. Affecting Gmail, YouTube, and More

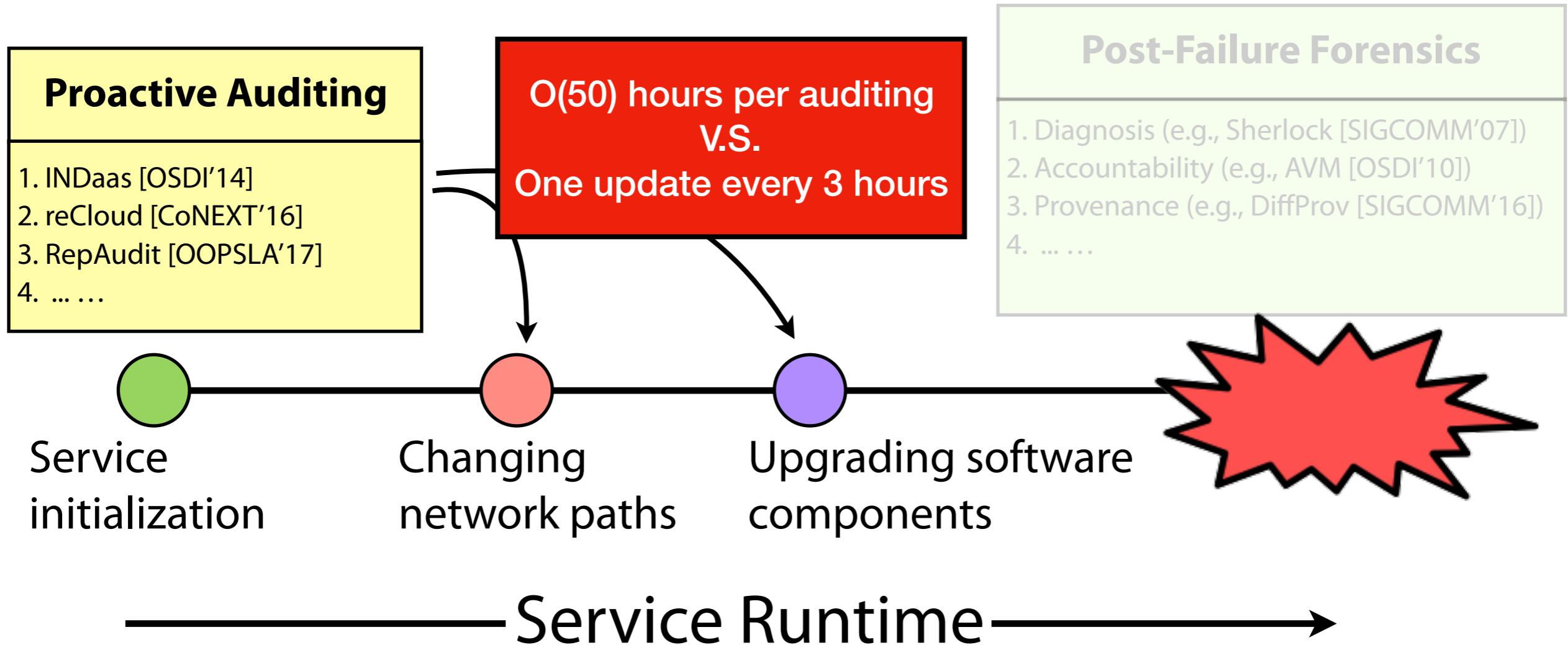
By Lawrence Abrams

June 2, 2019 06:38 PM 2

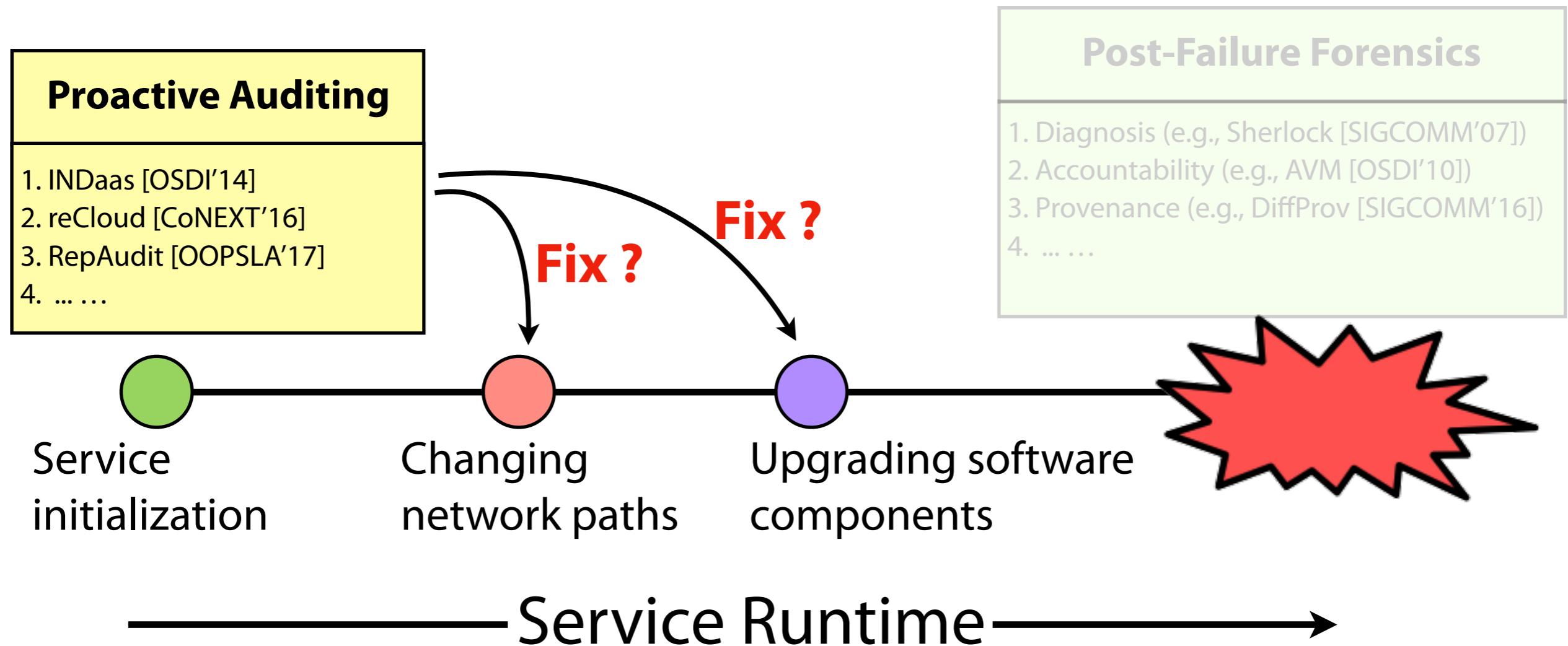


Benjamin Treynor Sloss, Google's VP of engineering, explained that the root cause of last Sunday's outage was a configuration change for a small group of servers in one region being wrongly applied to a larger number of servers across several neighboring regions.

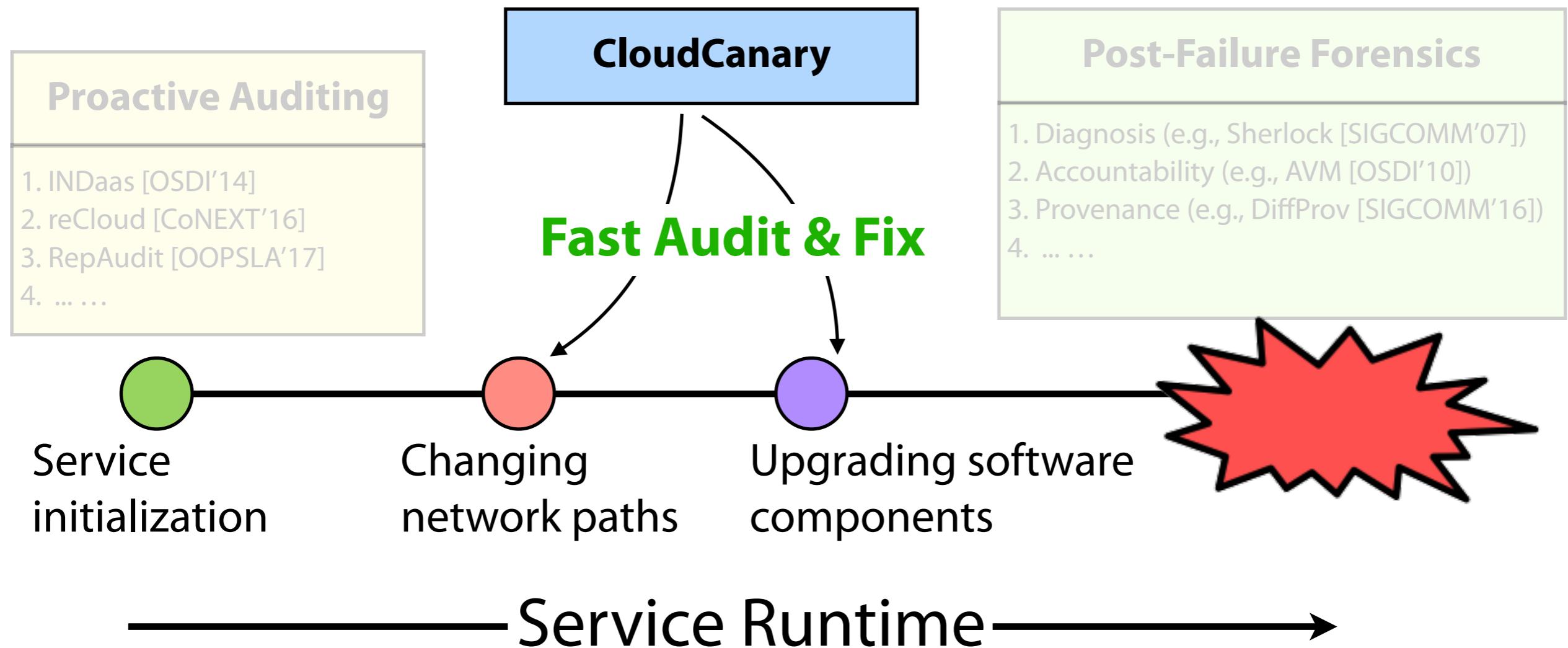
# Problem 1: Inefficient Auditing in Updates



# Problem 2: Lack of fixing risks



# Our Contribution



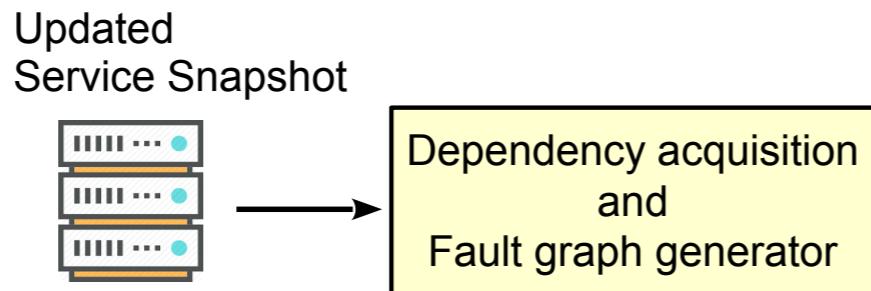
# CloudCanary's Workflow

Updated  
Service Snapshot



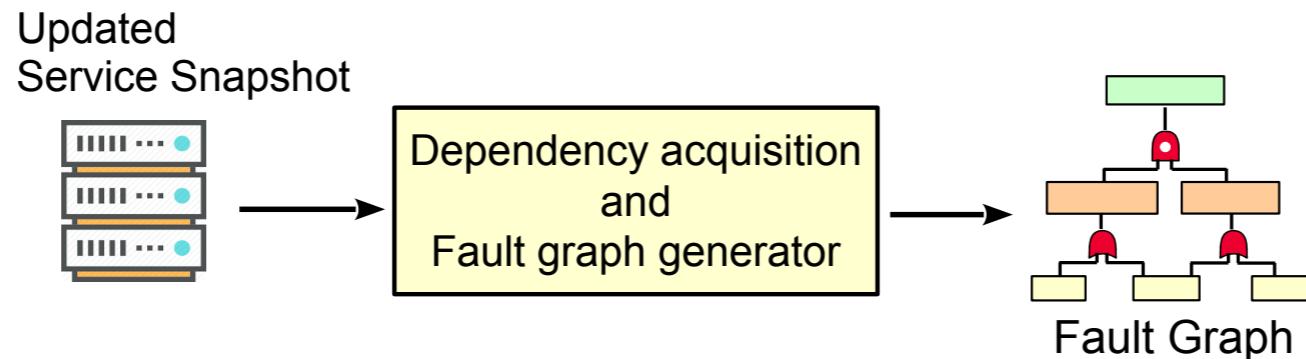
Operator

# CloudCanary's Workflow

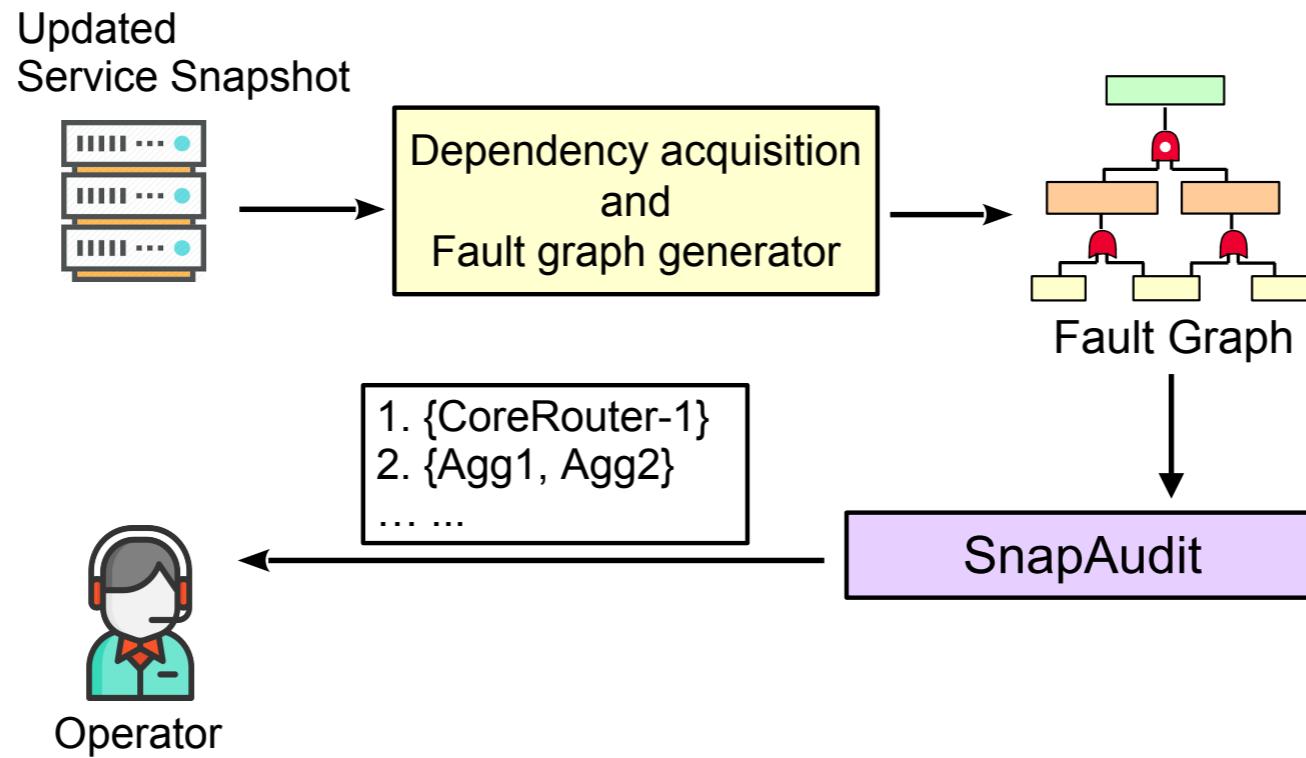


Operator

# CloudCanary's Workflow

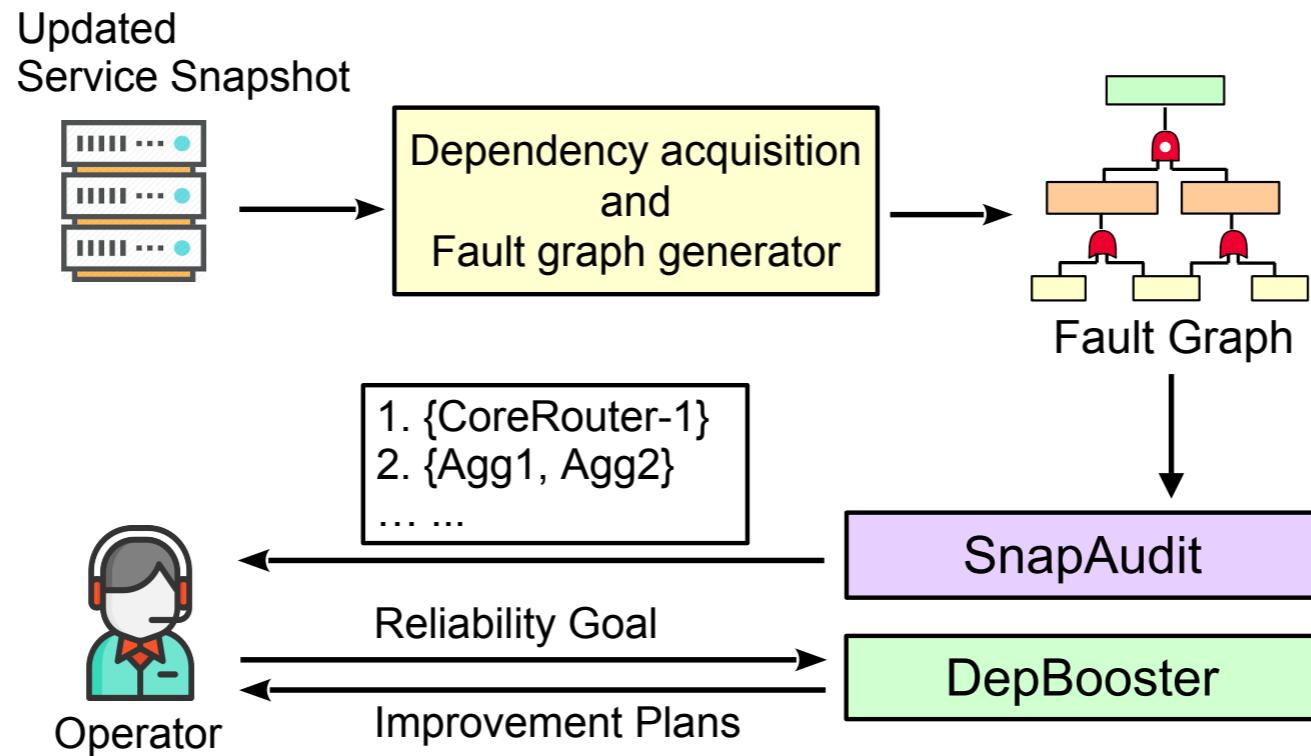


# CloudCanary's Workflow



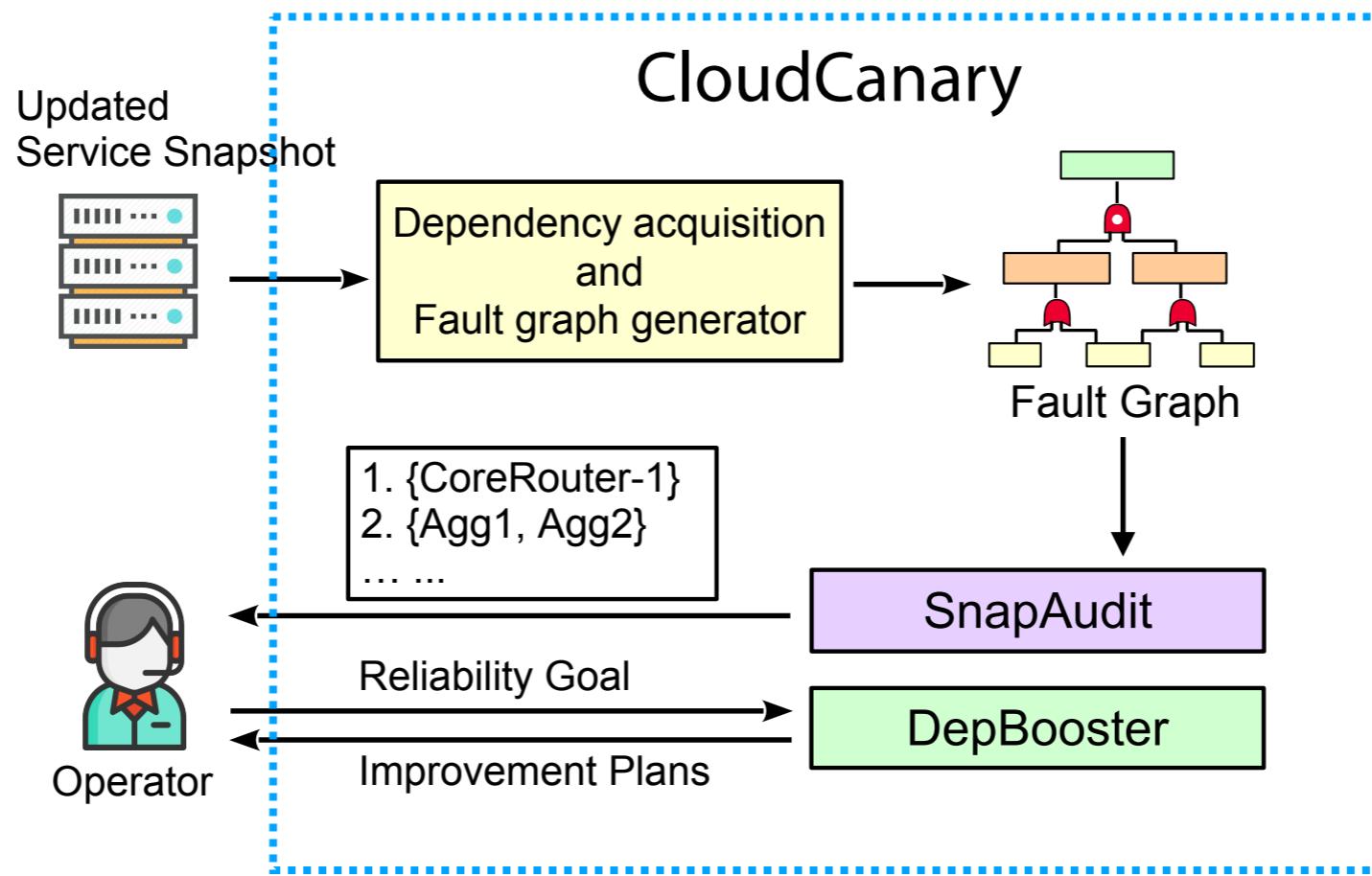
- Challenge 1: SnapAudit

# CloudCanary's Workflow



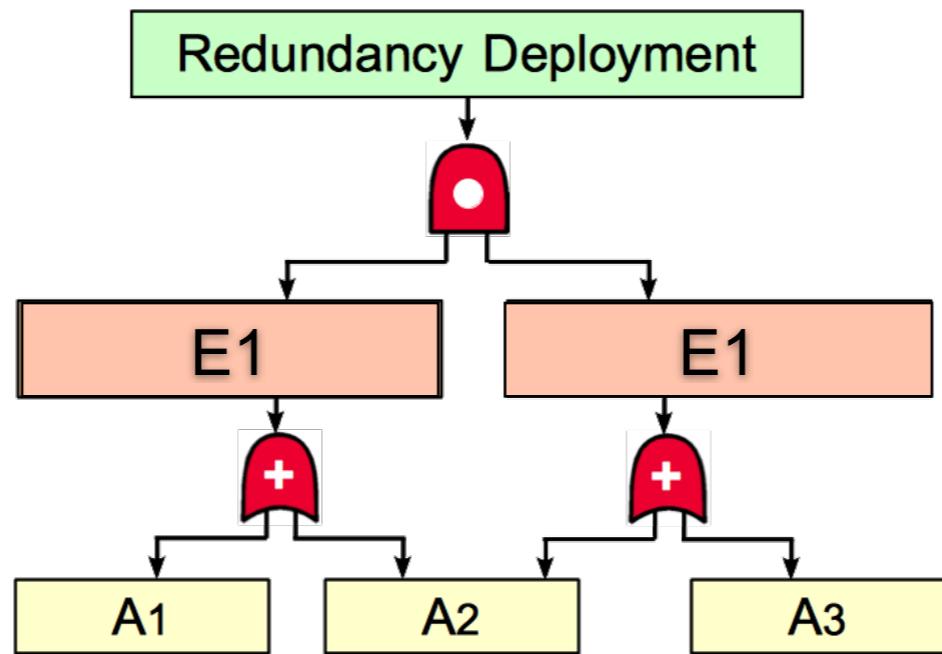
- Challenge 1: SnapAudit
- Challenge 2: DepBooster

# CloudCanary's Workflow

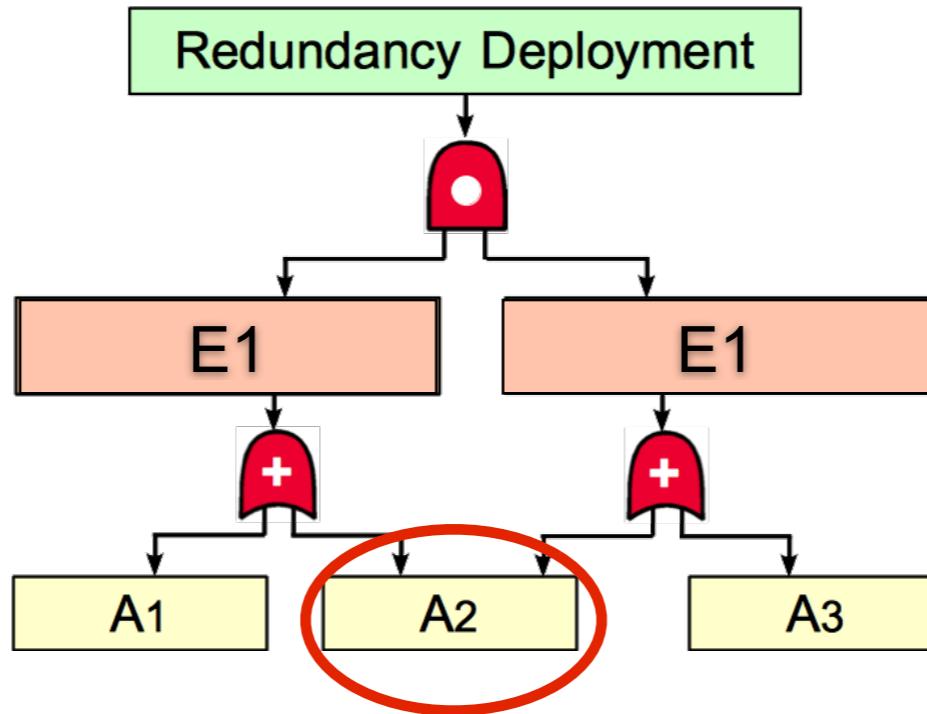


- Challenge 1: SnapAudit
- Challenge 2: DepBooster

# A Fault Graph

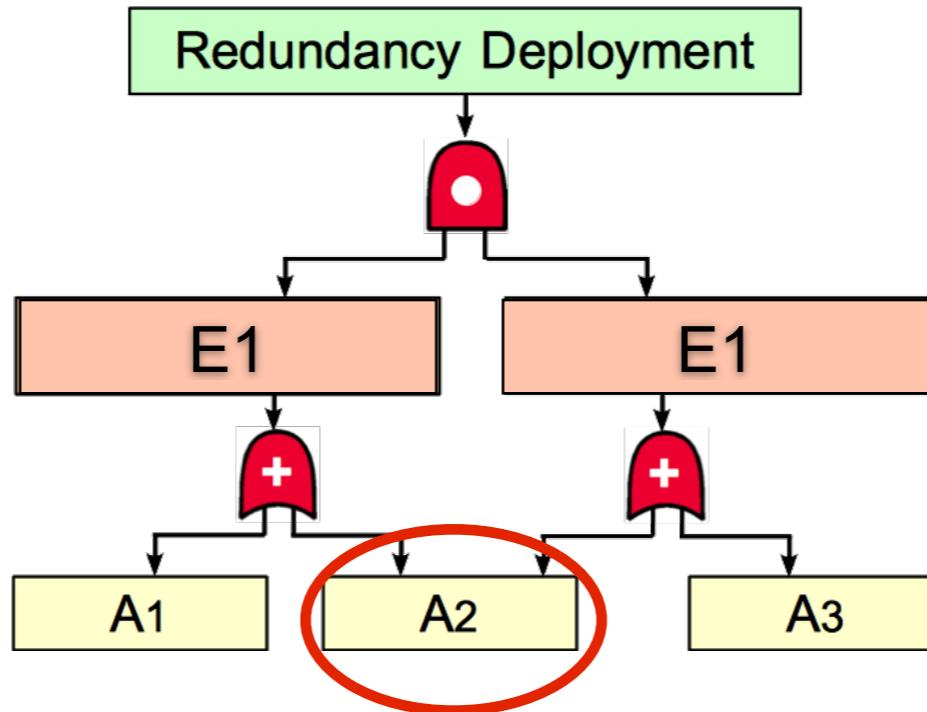


# Risk Groups in Fault Graphs



- A risk group means a set of leaf nodes whose simultaneous failures lead to the failure of root node

# Risk Groups in Fault Graphs



- A risk group means a set of leaf nodes whose simultaneous failures lead to the failure of root node
  - {A2} and {A1, A3} are risk groups
  - {A1} or {A3} is not risk group

# Risk Groups in Fault Graphs

Redundancy Deployment

Identifying correlated failure risks can be reduced to the problem of finding risk groups in the fault graph.

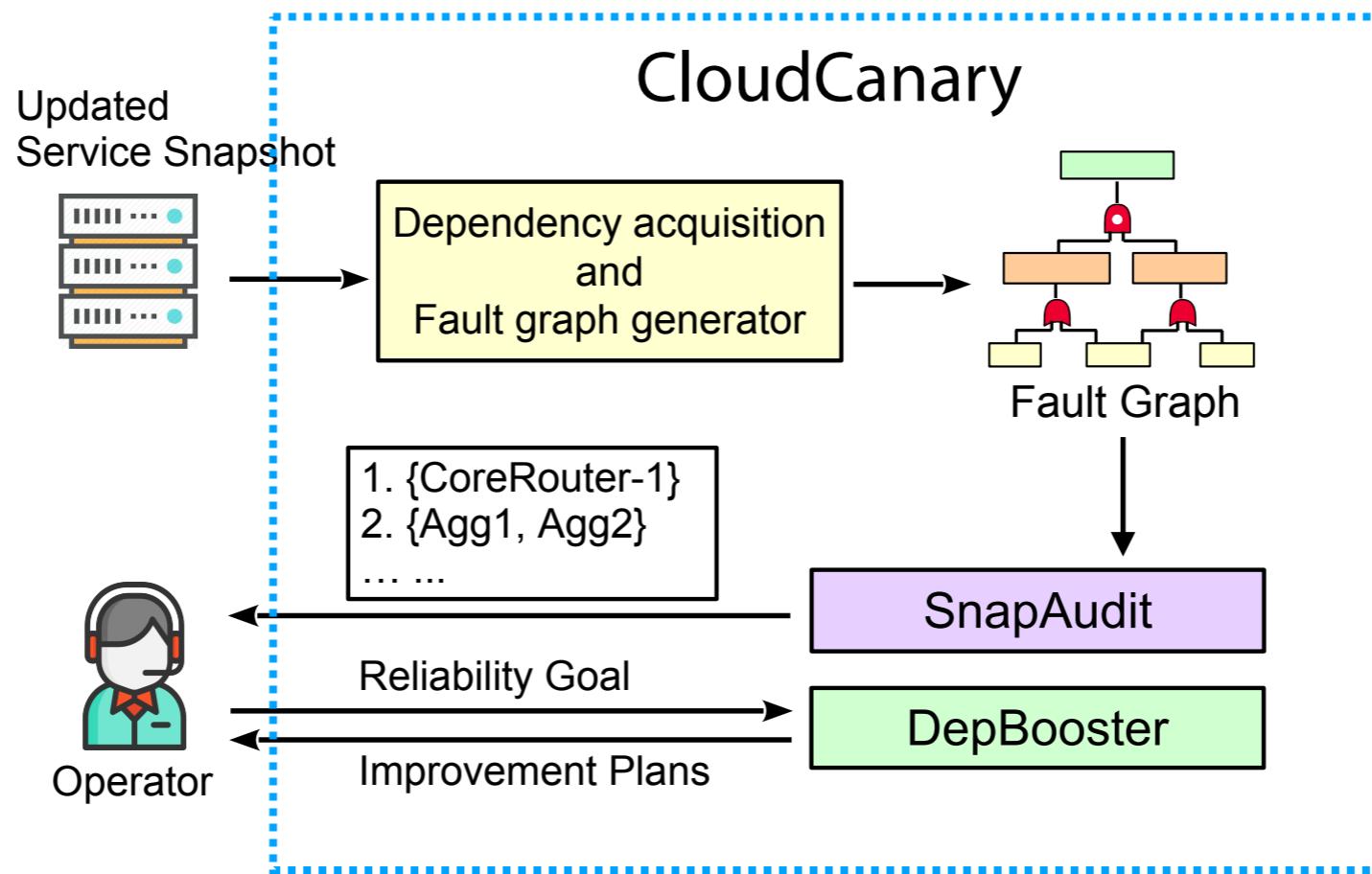
**However, analyzing risk groups is  
NP-complete problem**

~~Simultaneous failures lead to the failure of root node~~

{A2} and {A1, A3} are risk groups

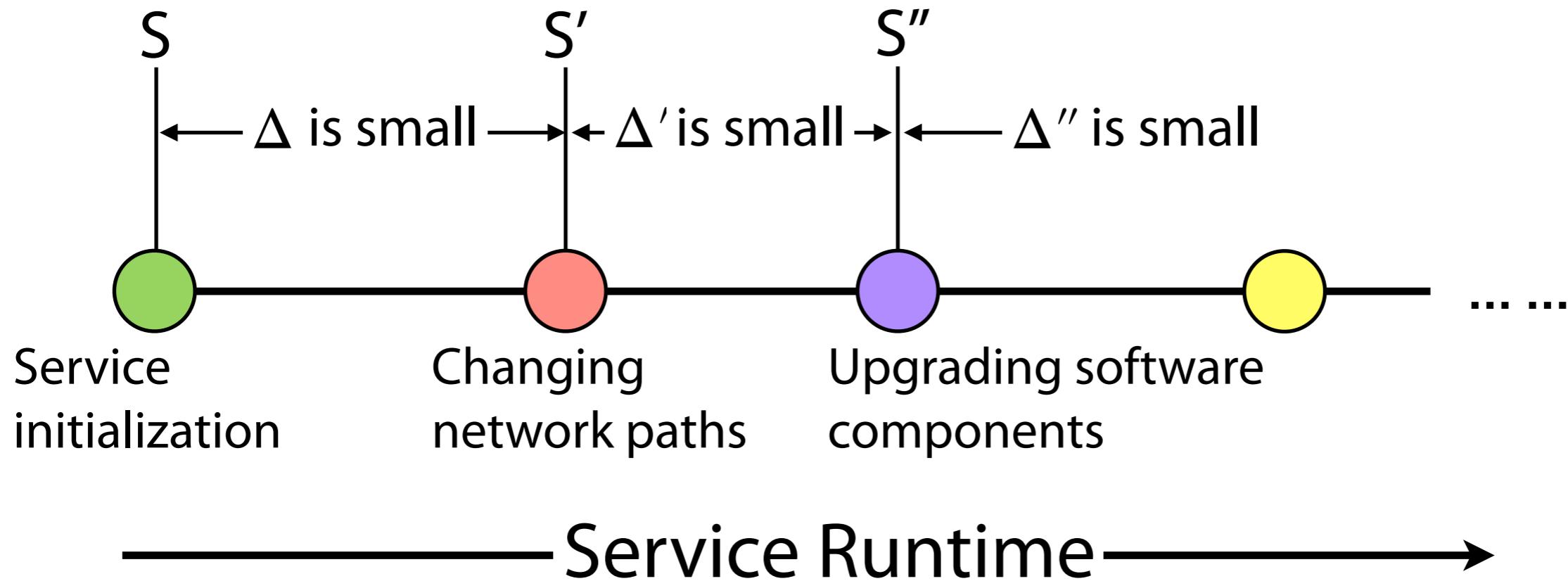
{A1} or {A3} is not risk group

# CloudCanary's Workflow

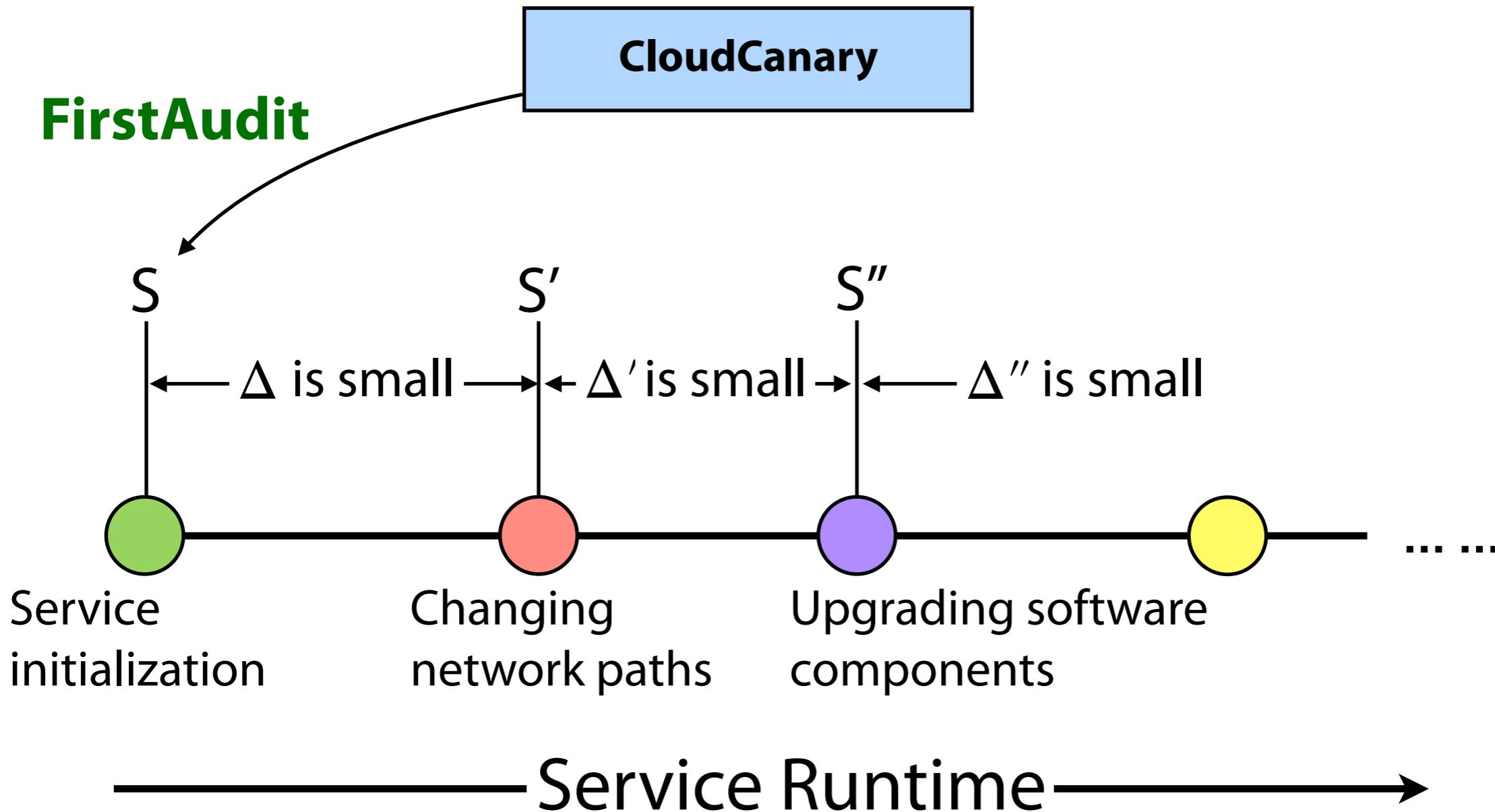


- Challenge 1: SnapAudit
- Challenge 2: DepBooster

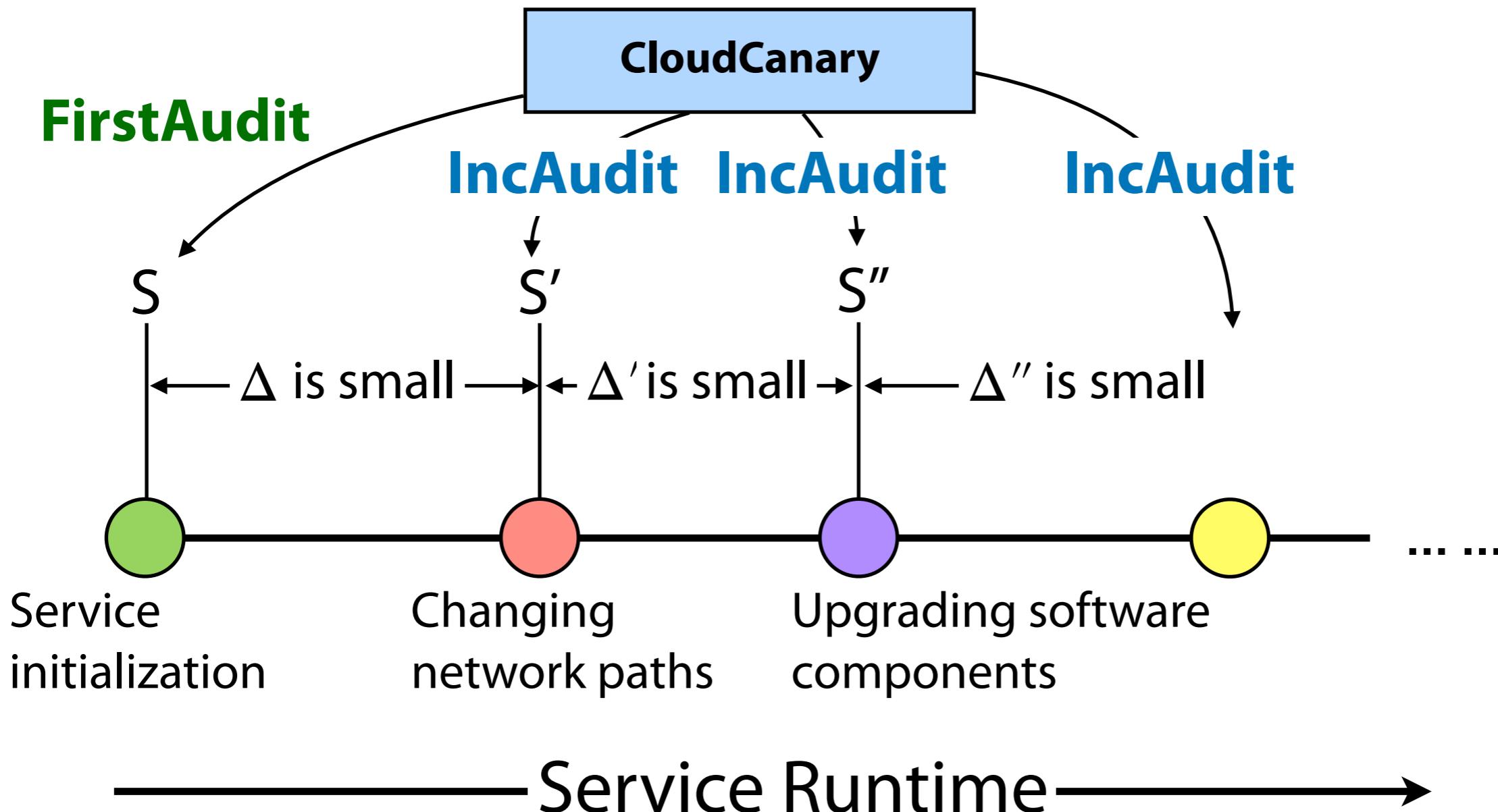
# The Insight of SnapAudit



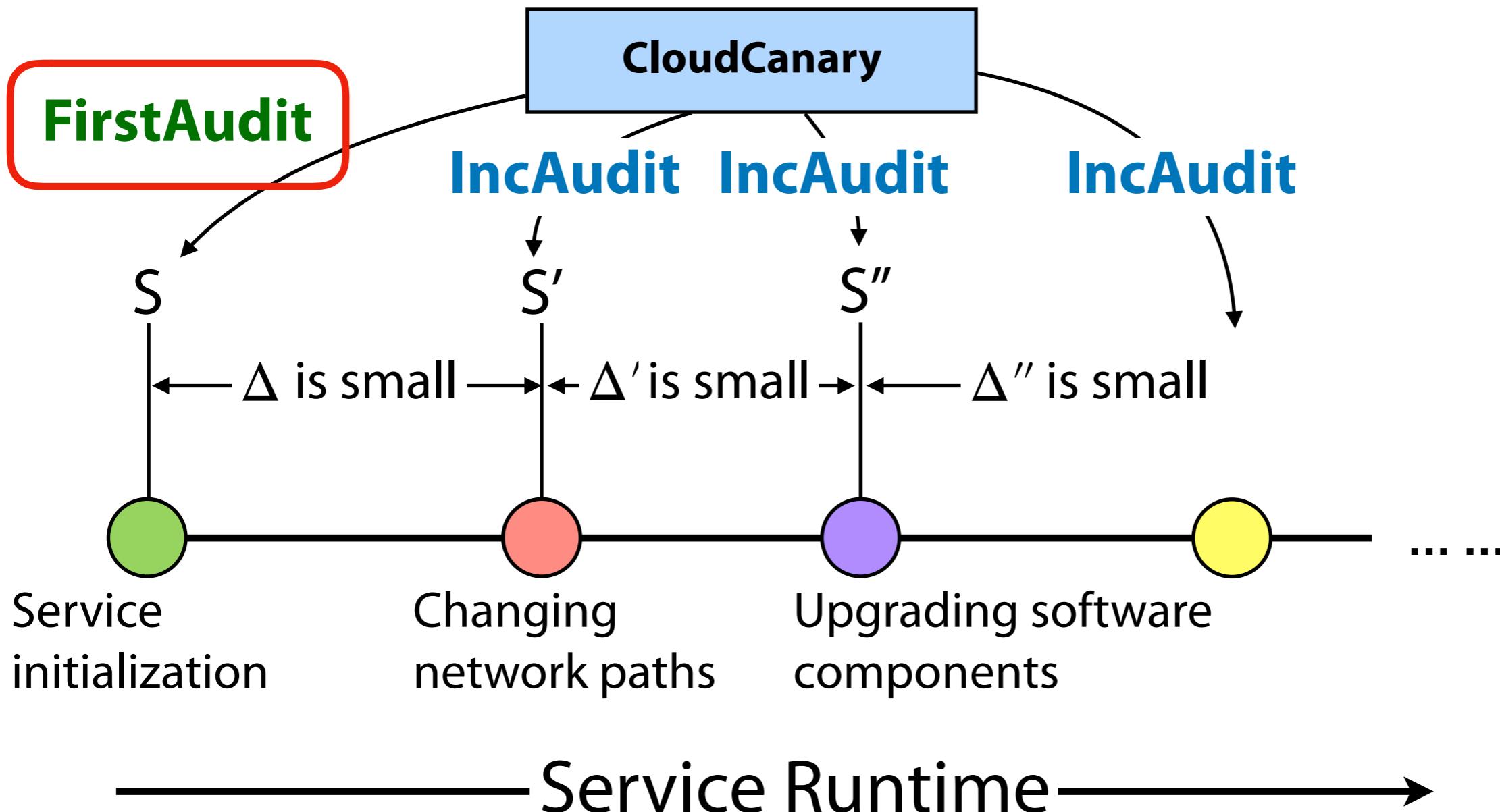
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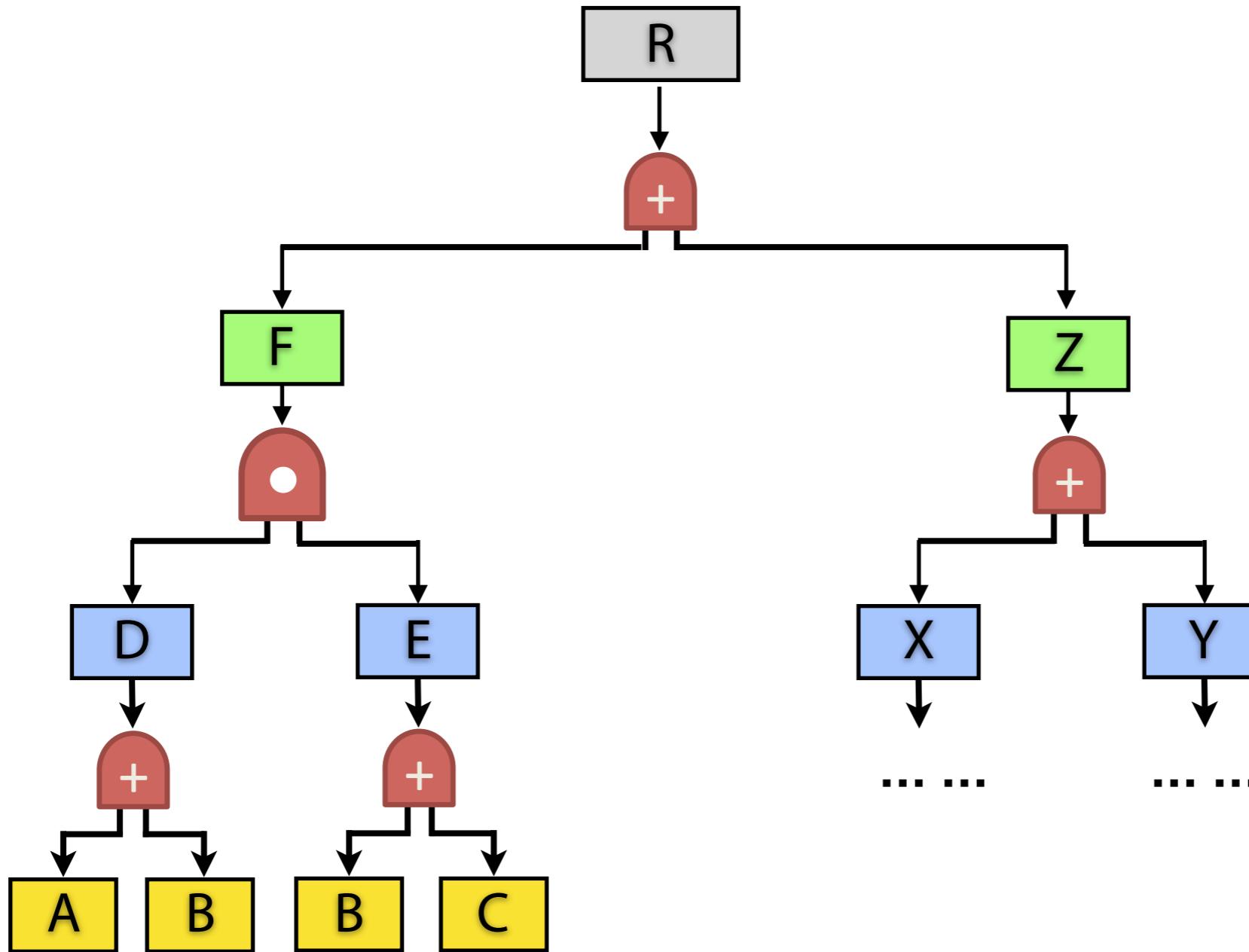
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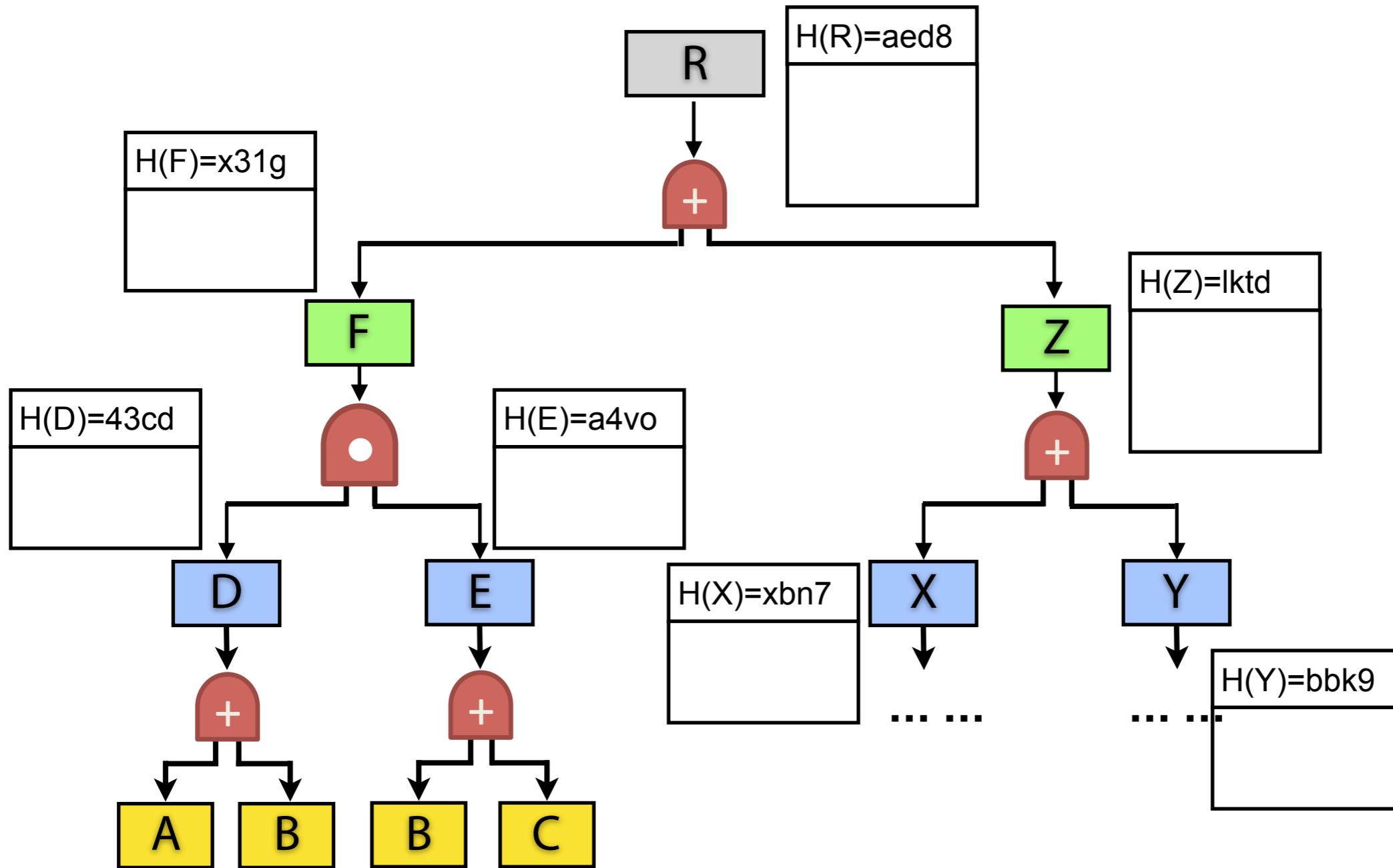
# SnapAudit: FirstAudit & IncAudit



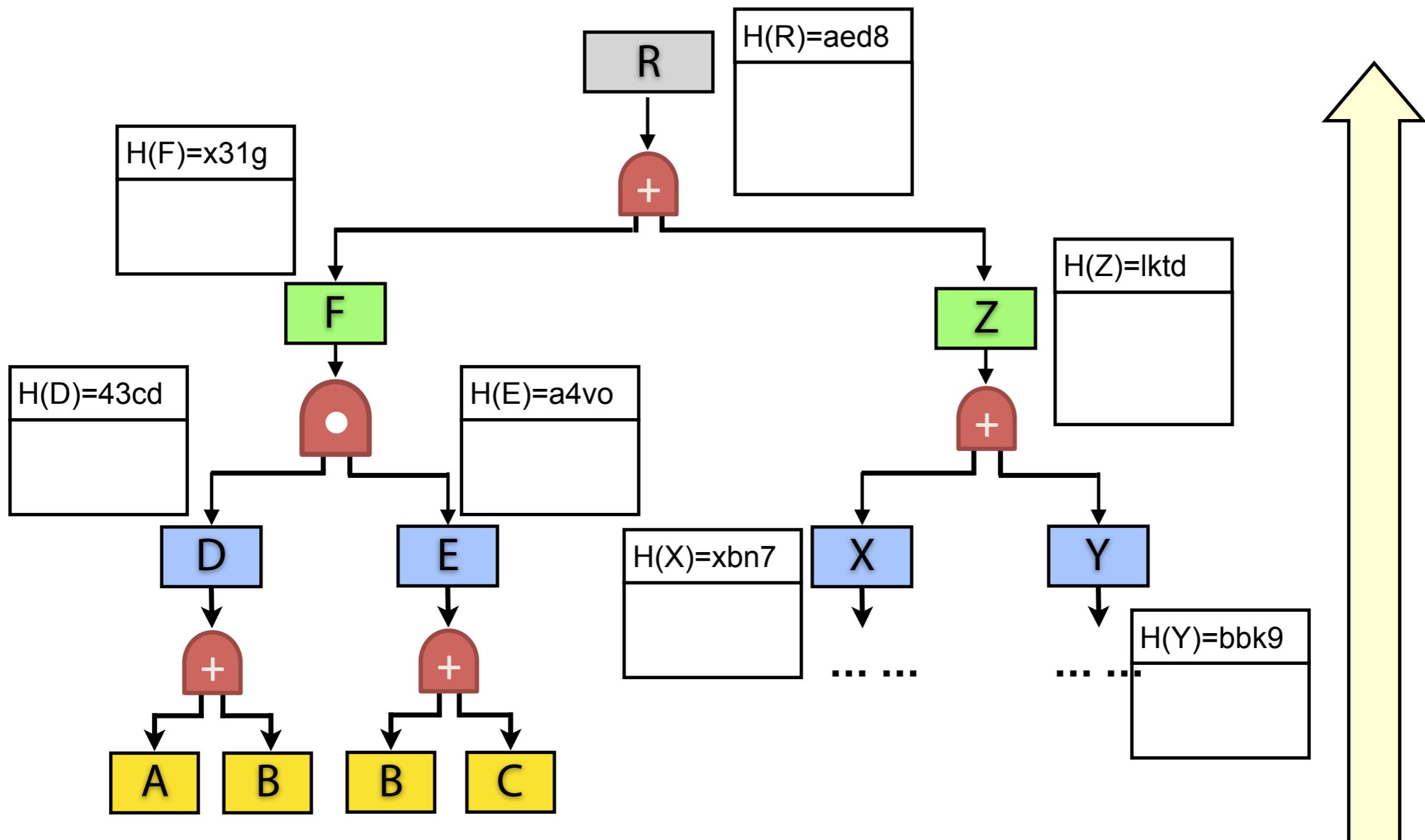
# FirstAudit Primitive



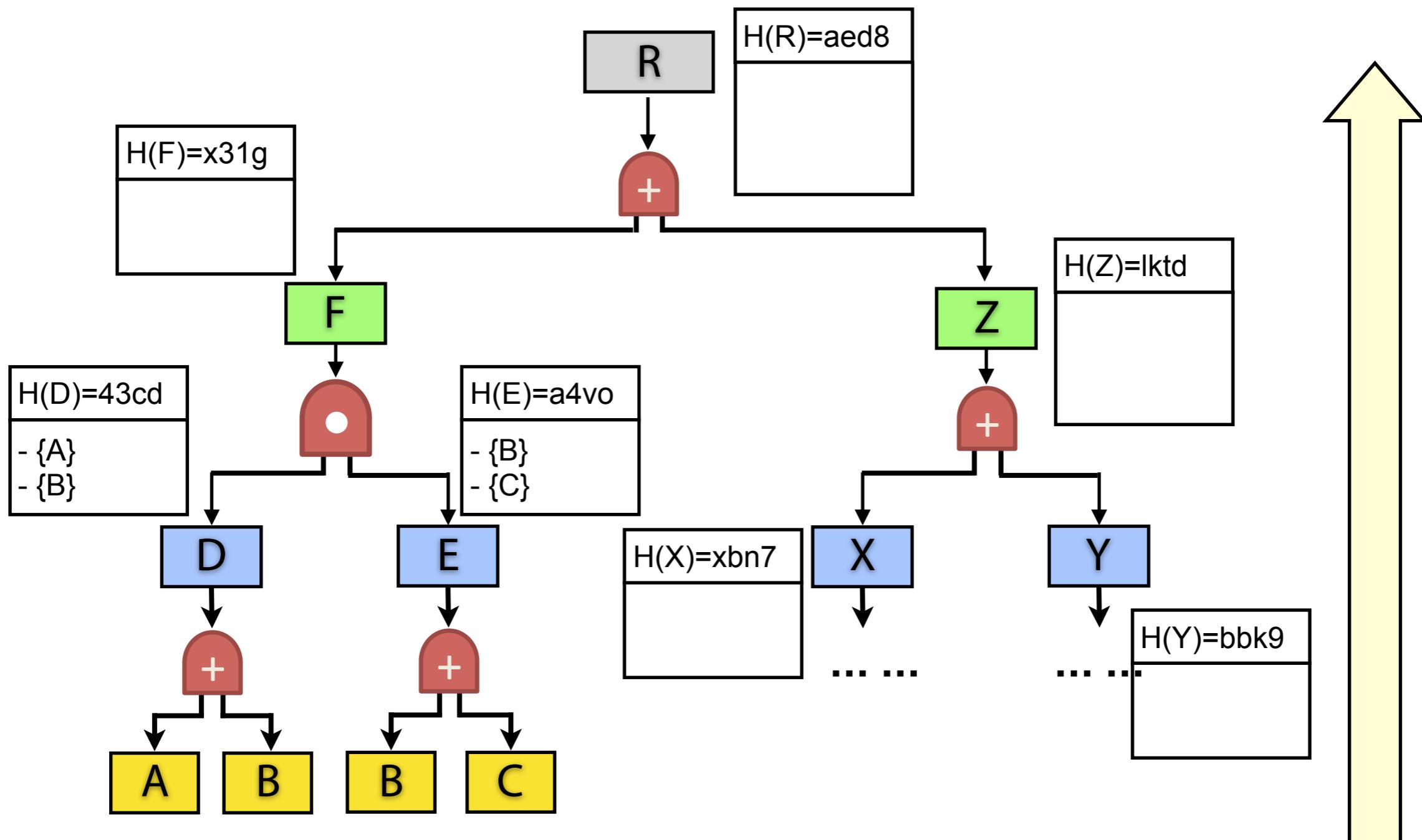
# FirstAudit Primitive



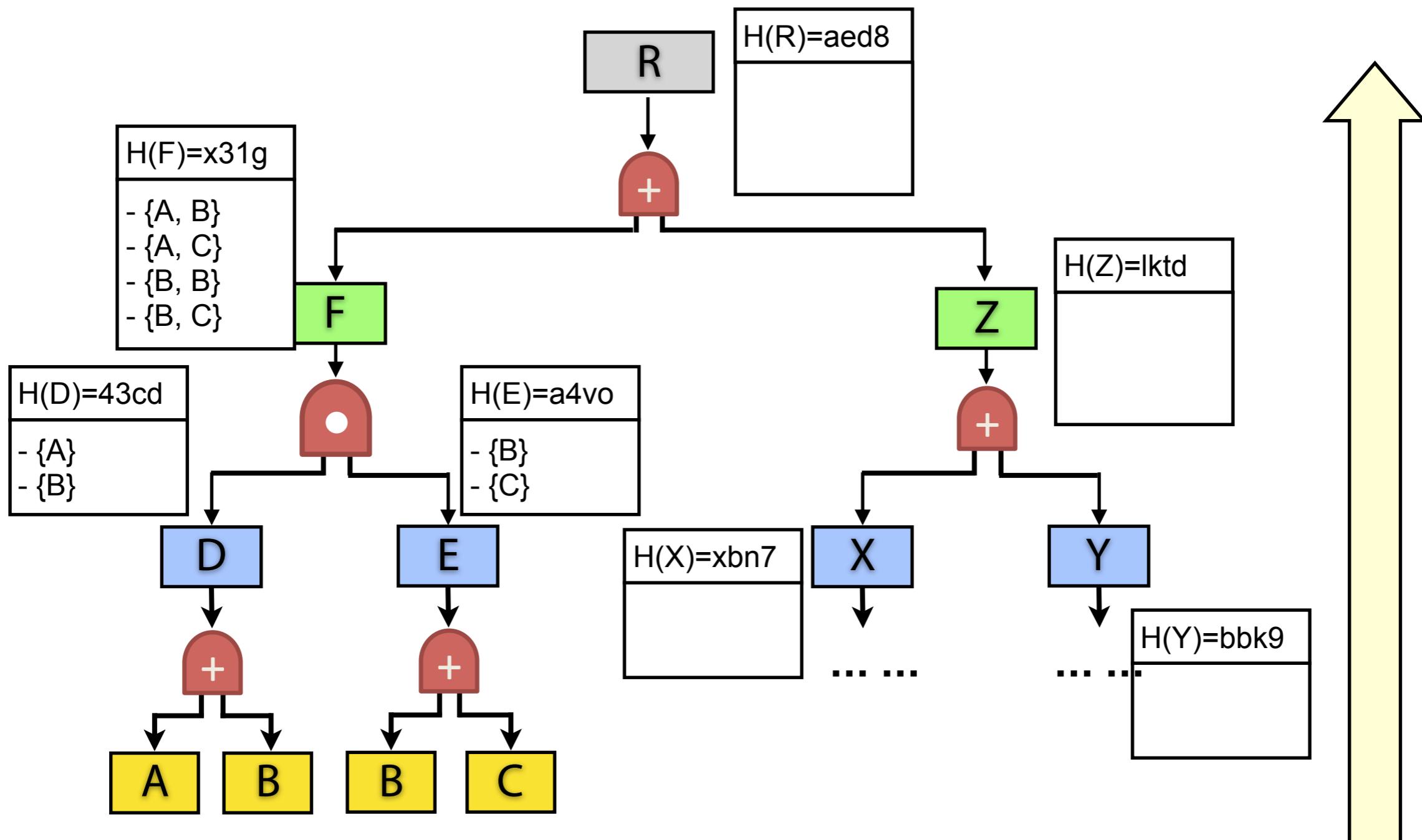
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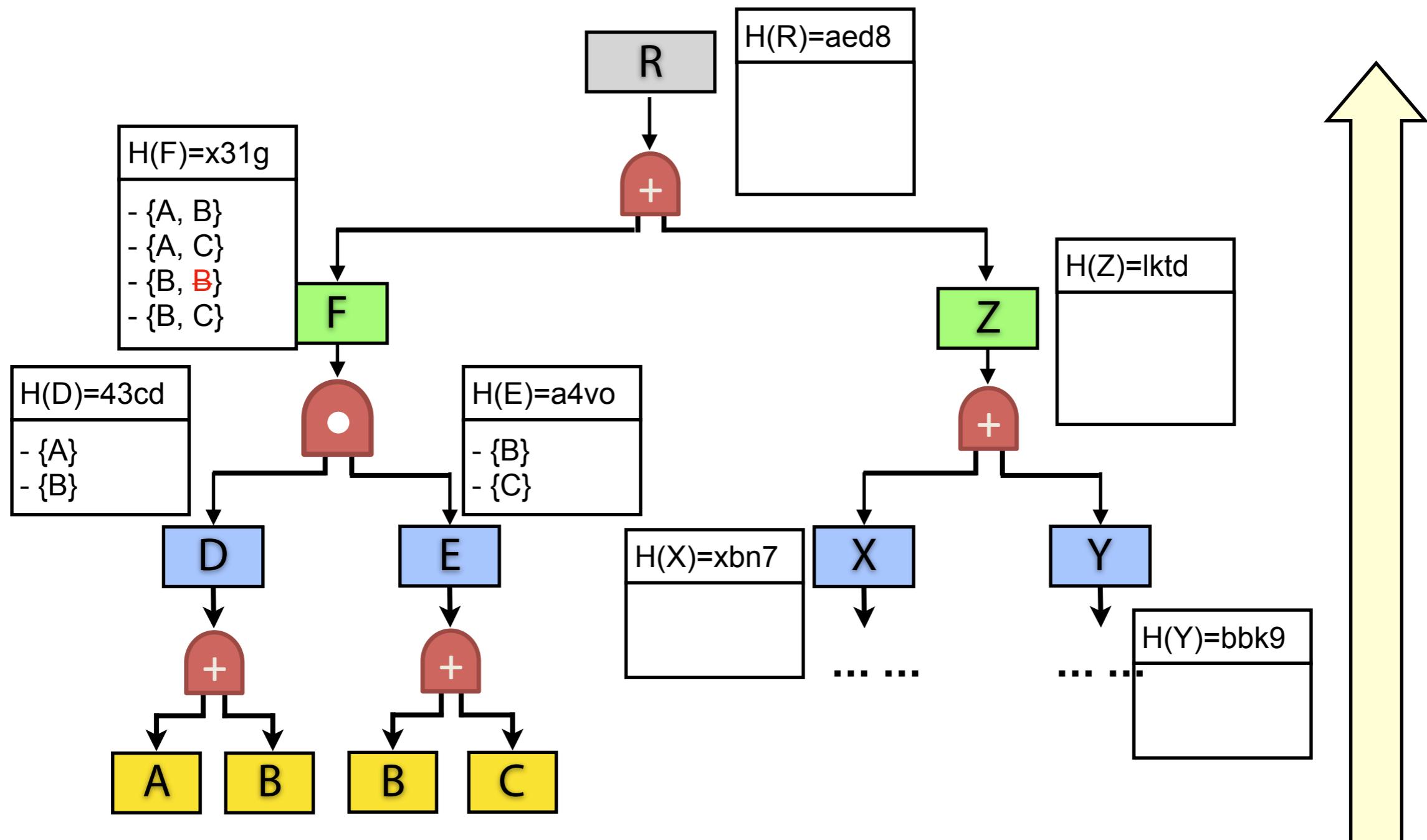
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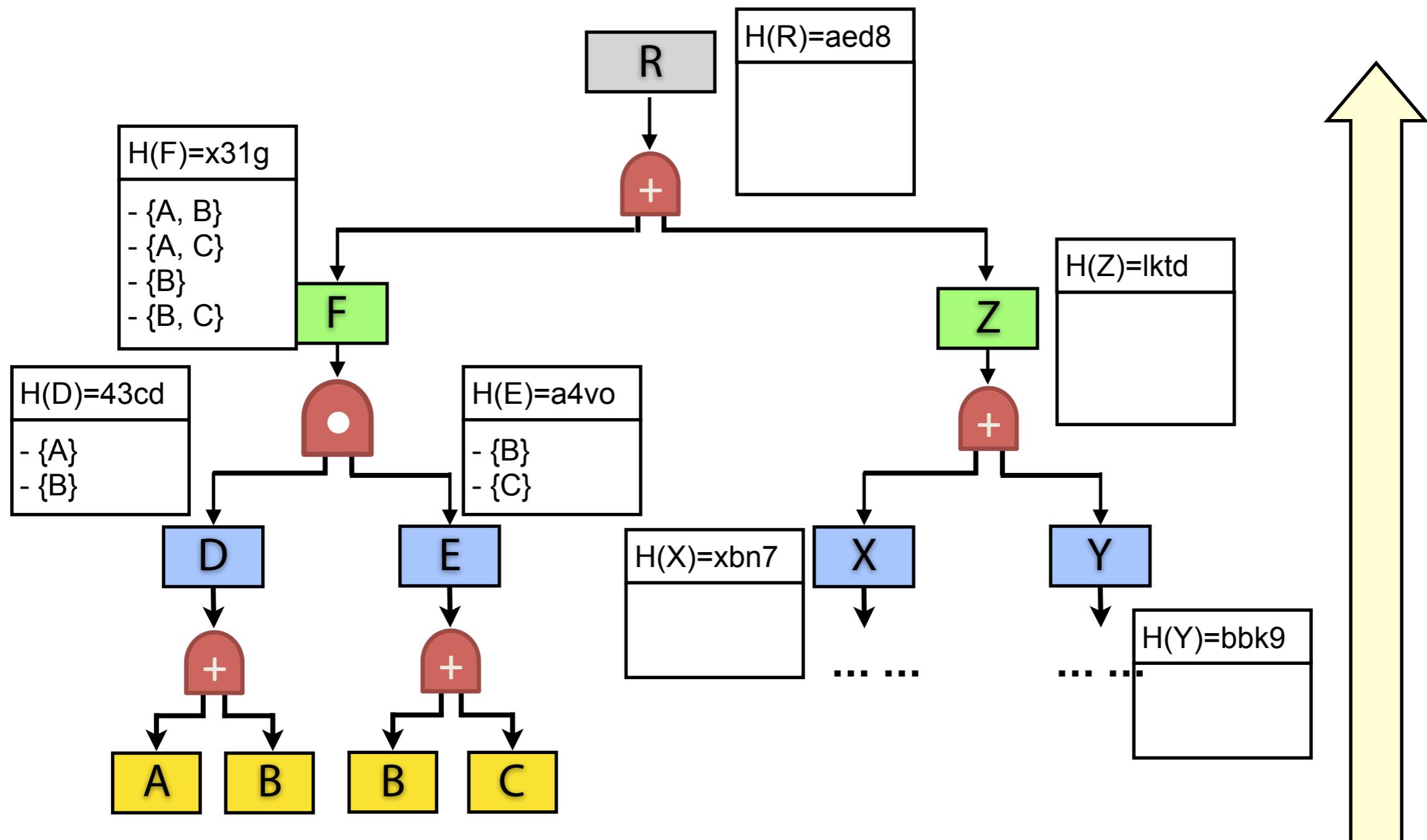
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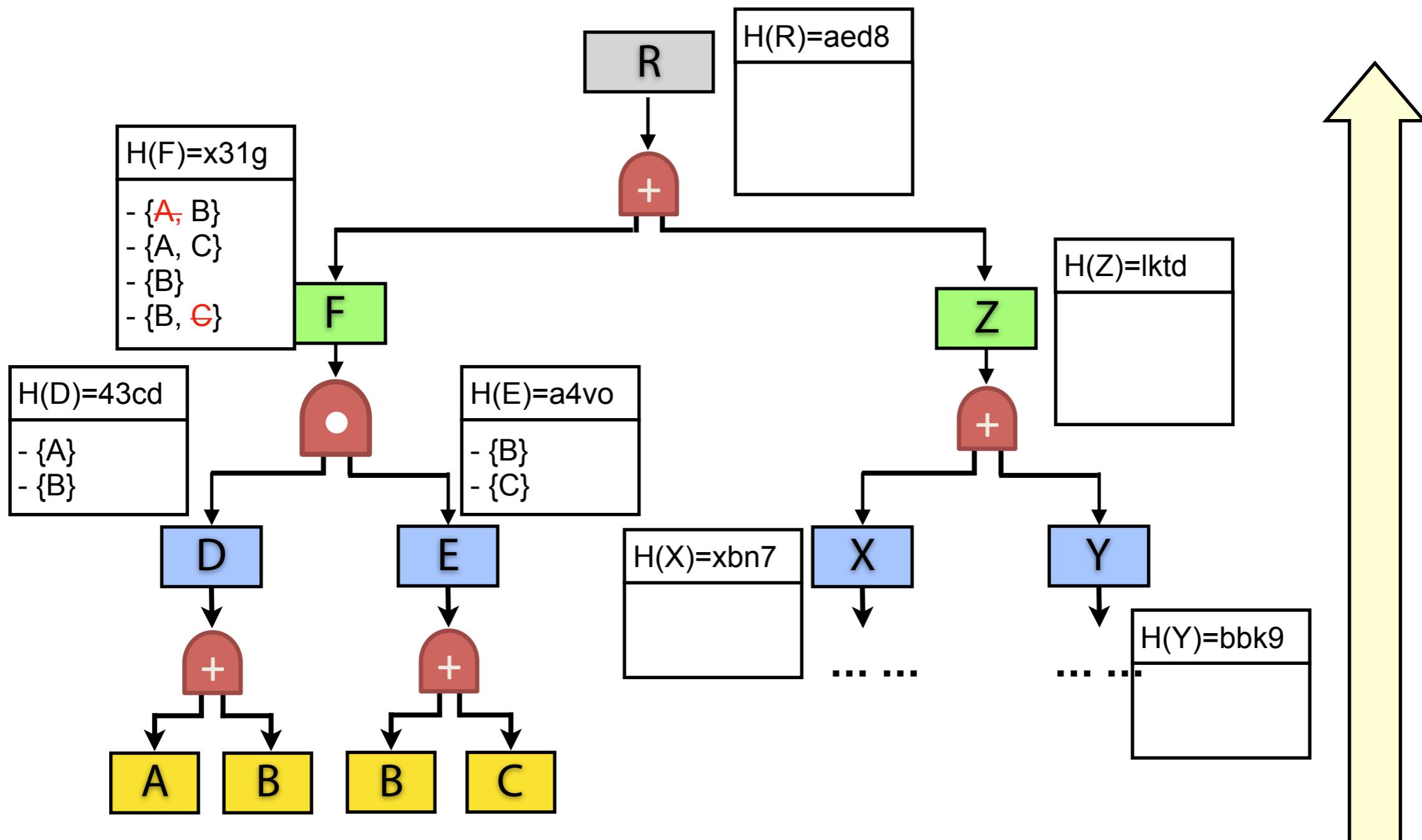
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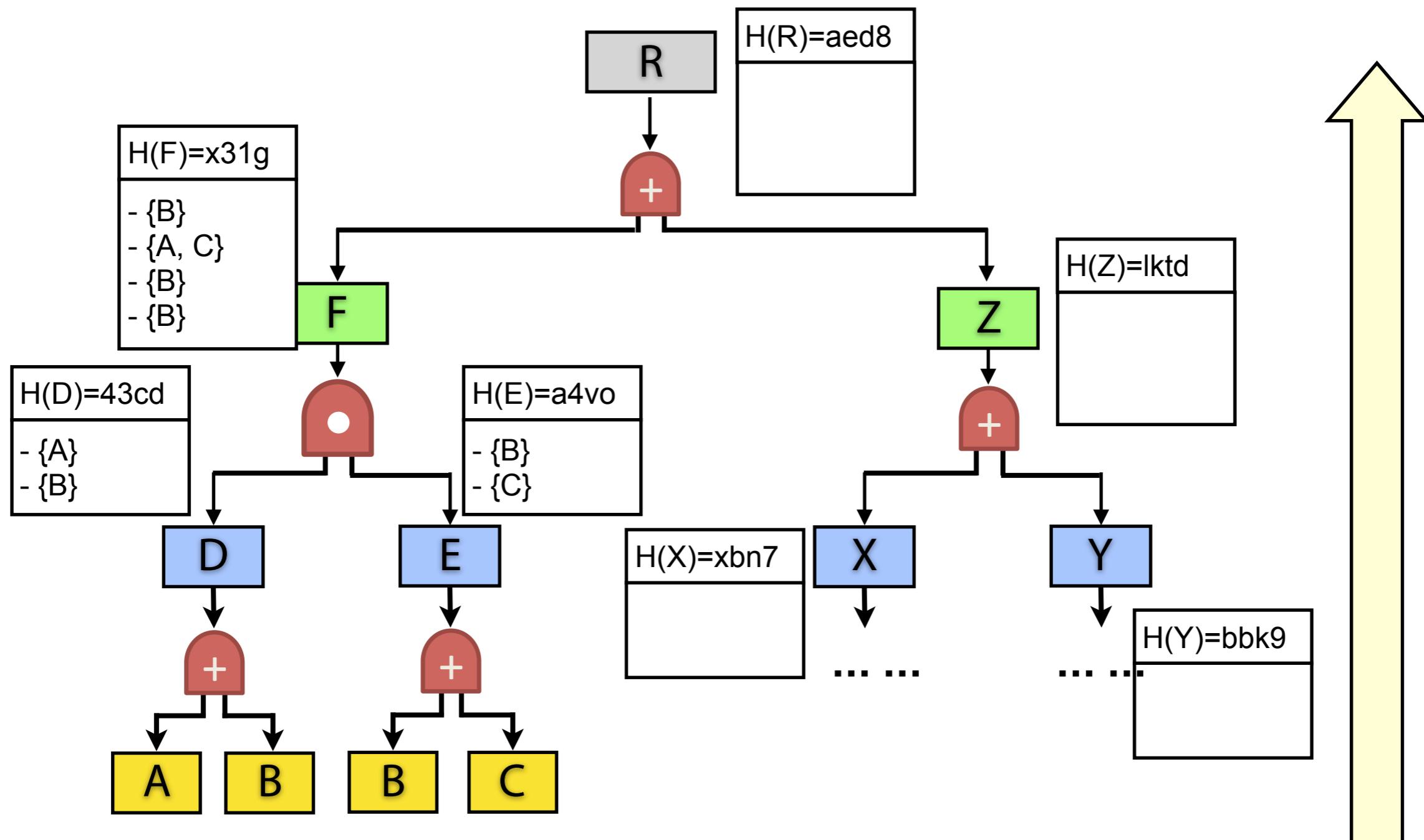
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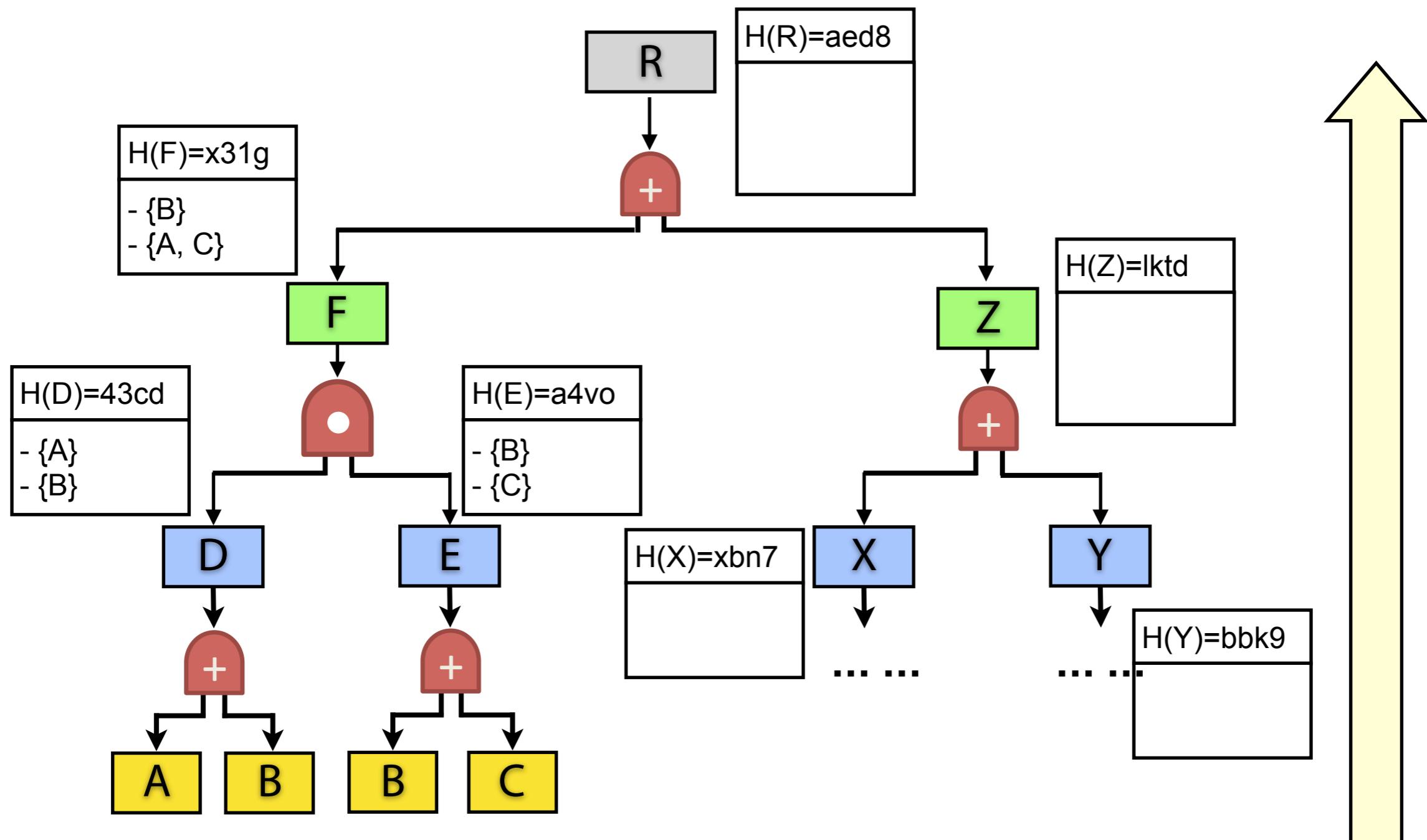
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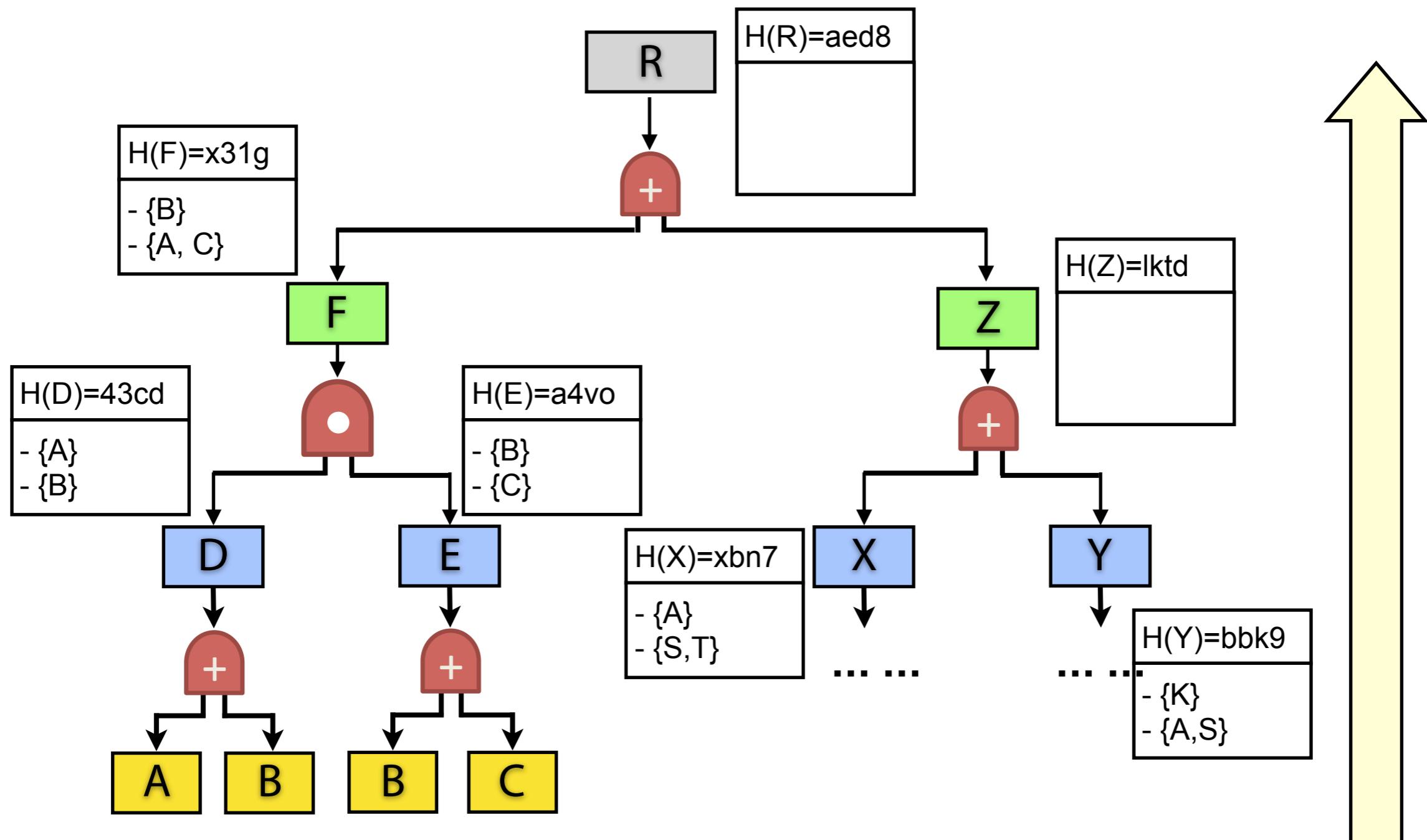
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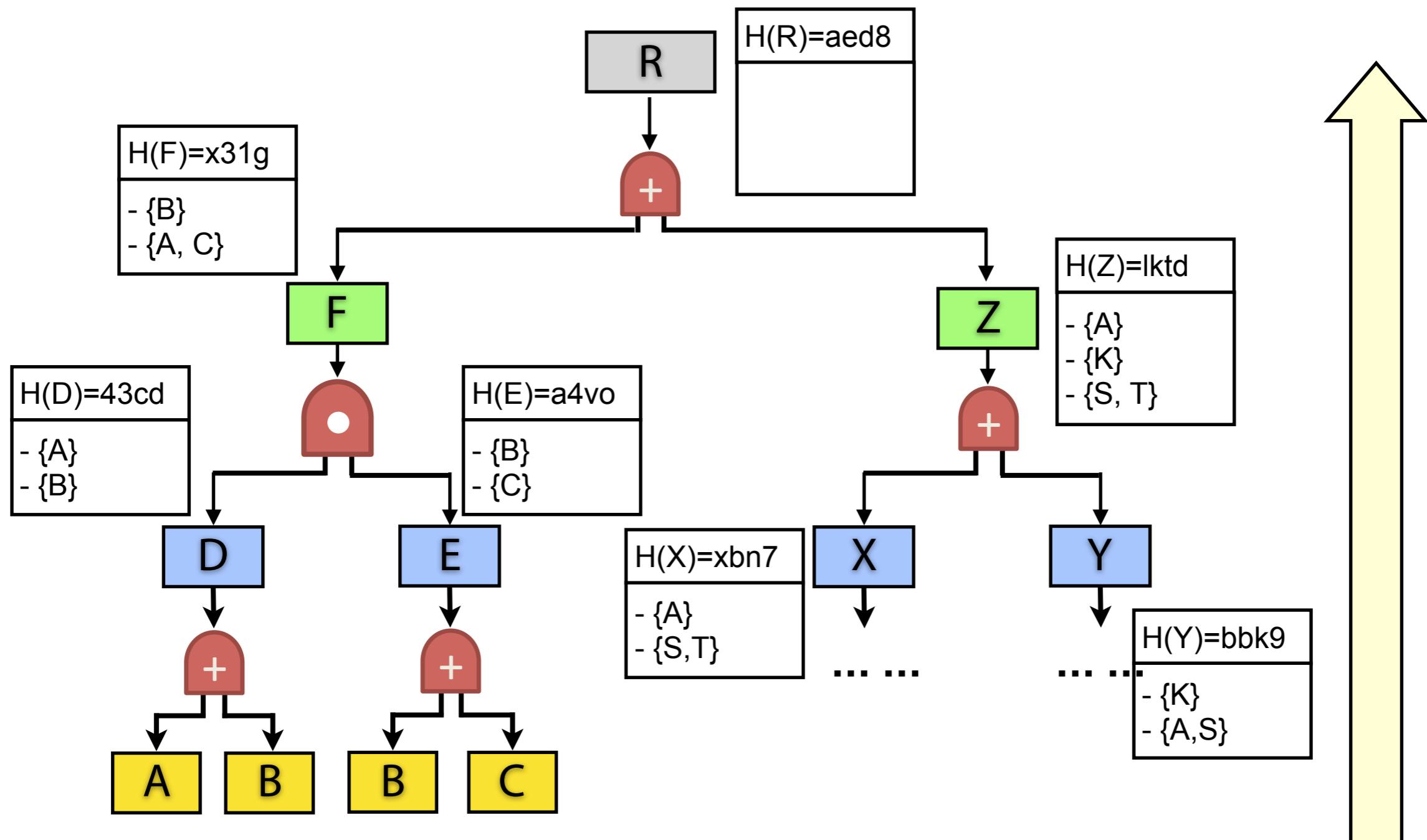
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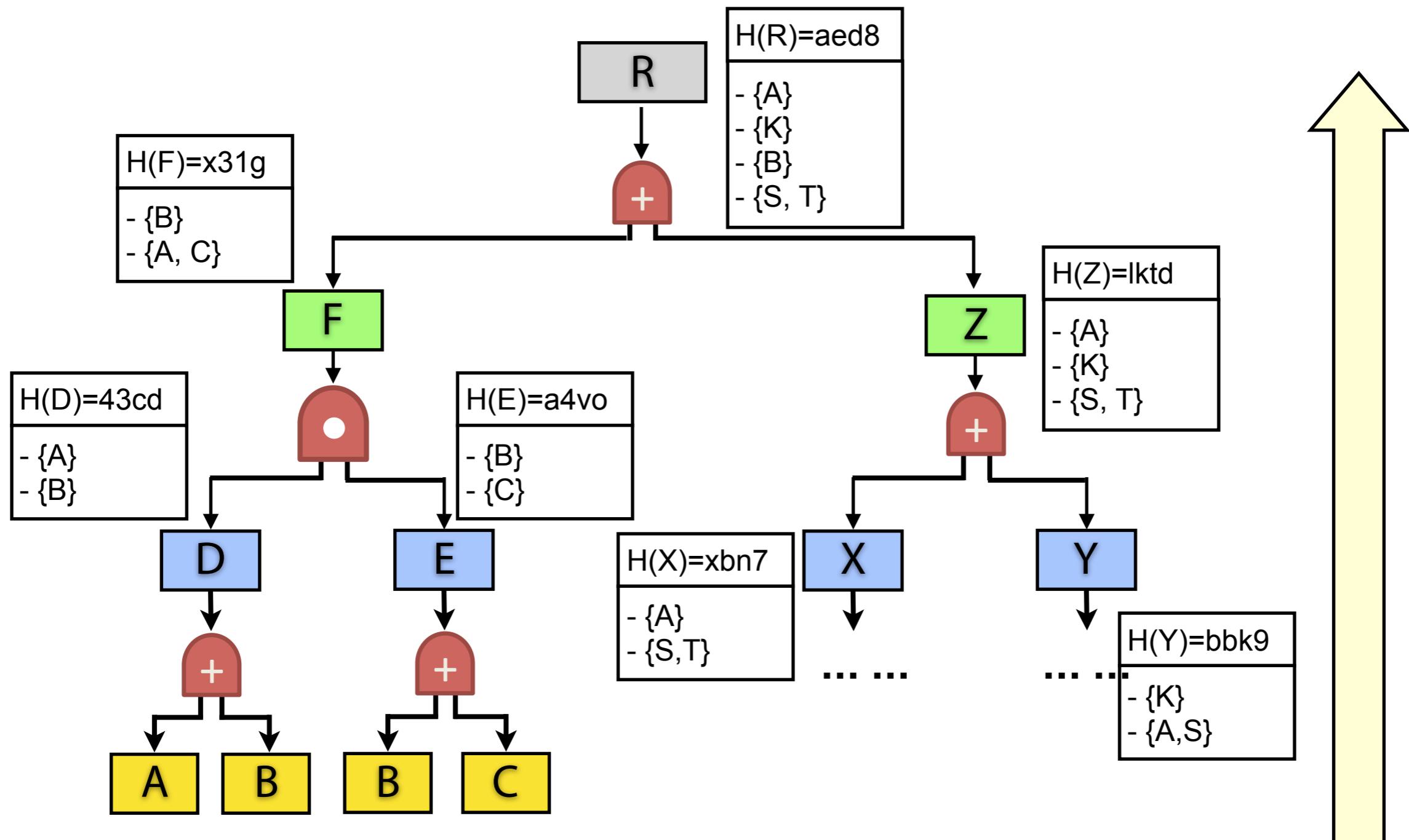
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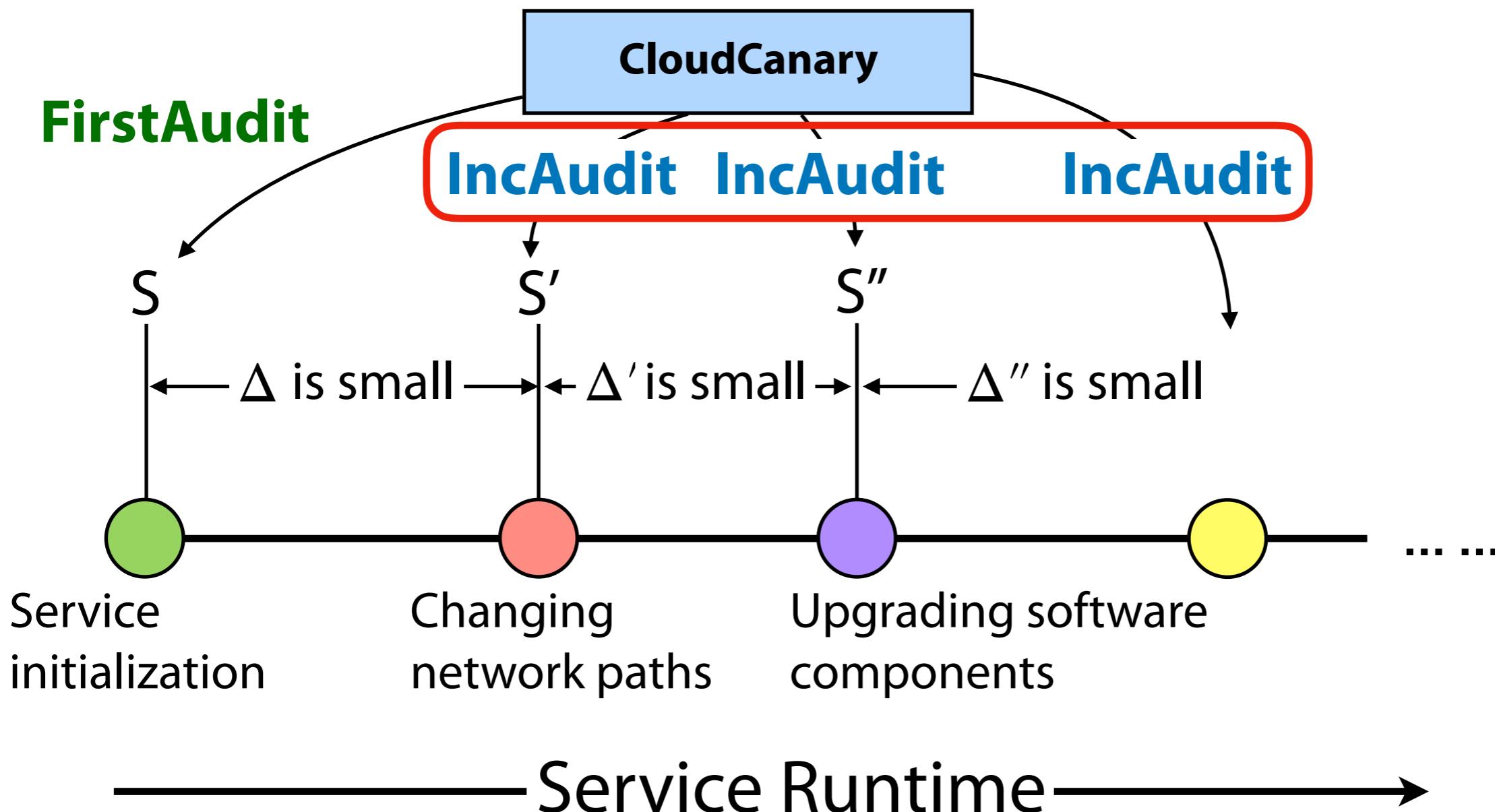
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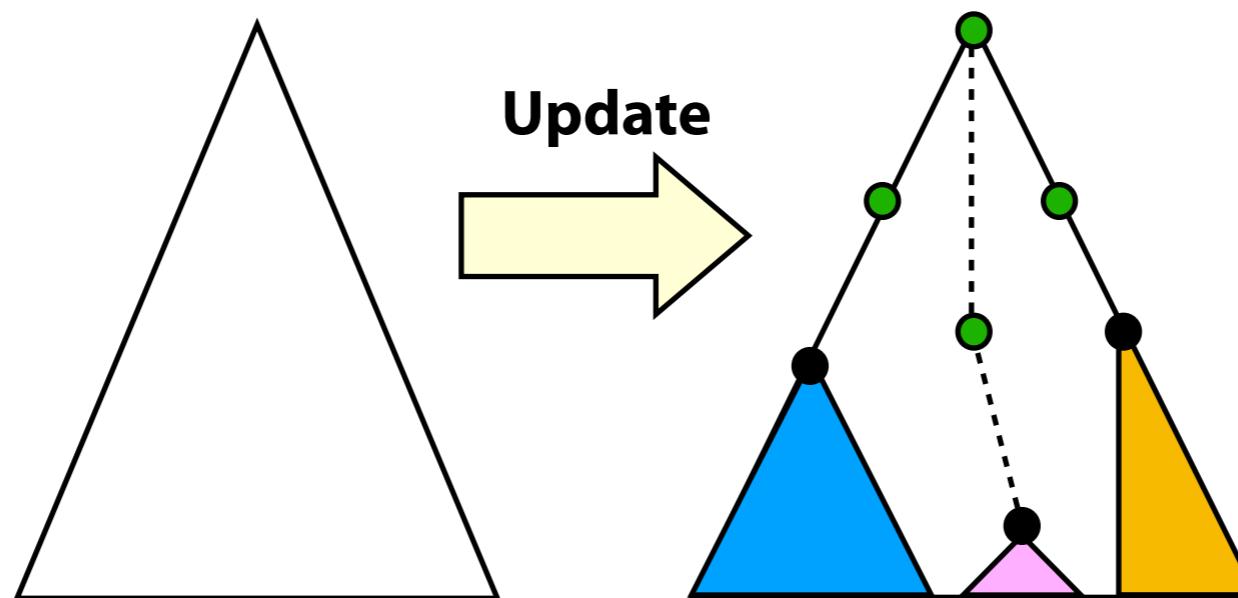


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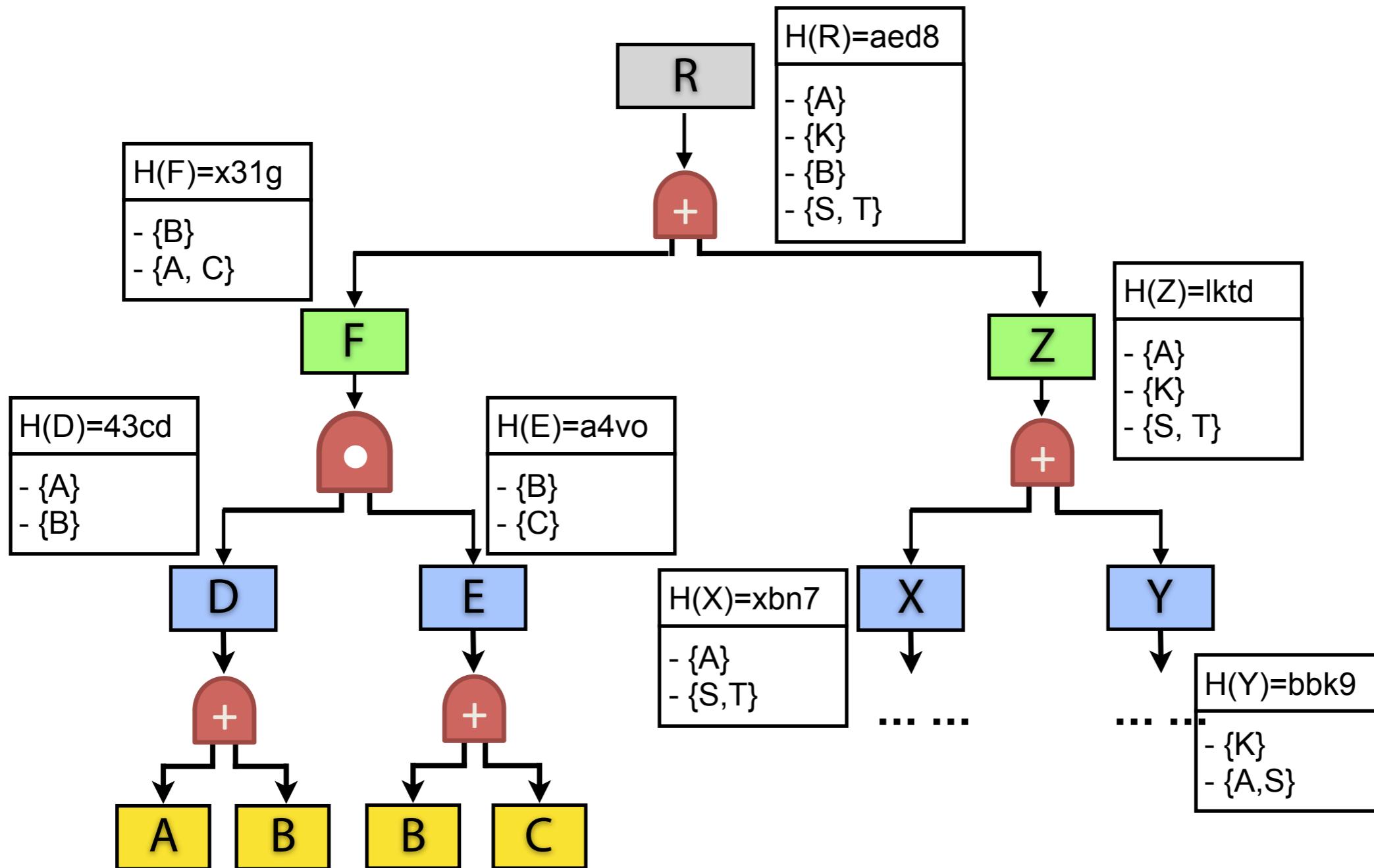


# Our Insight

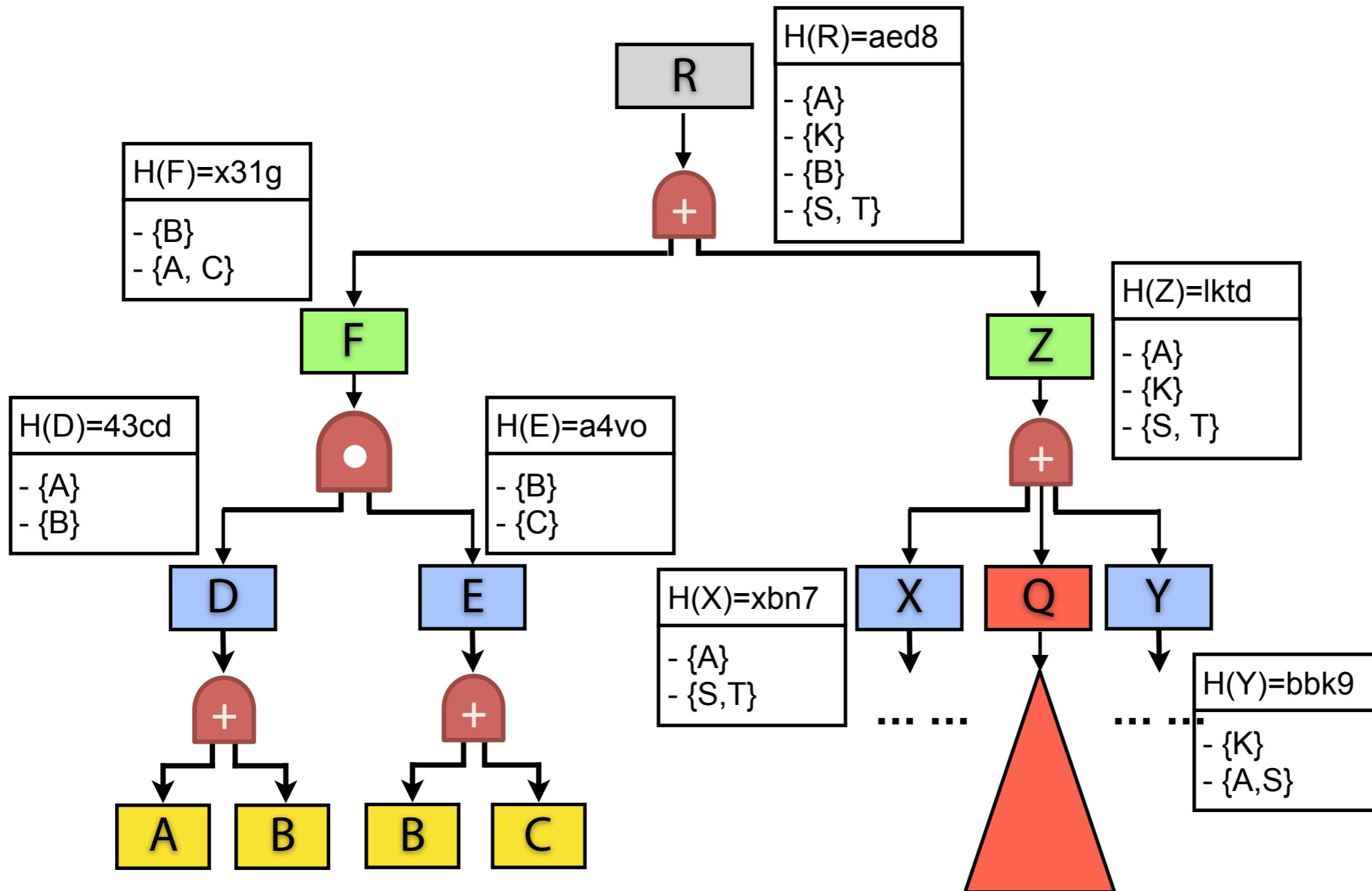
- Algorithm sketch:
  - Finding all the border nodes (black nodes)
  - Computing their risk groups
  - Merging these risk groups towards root



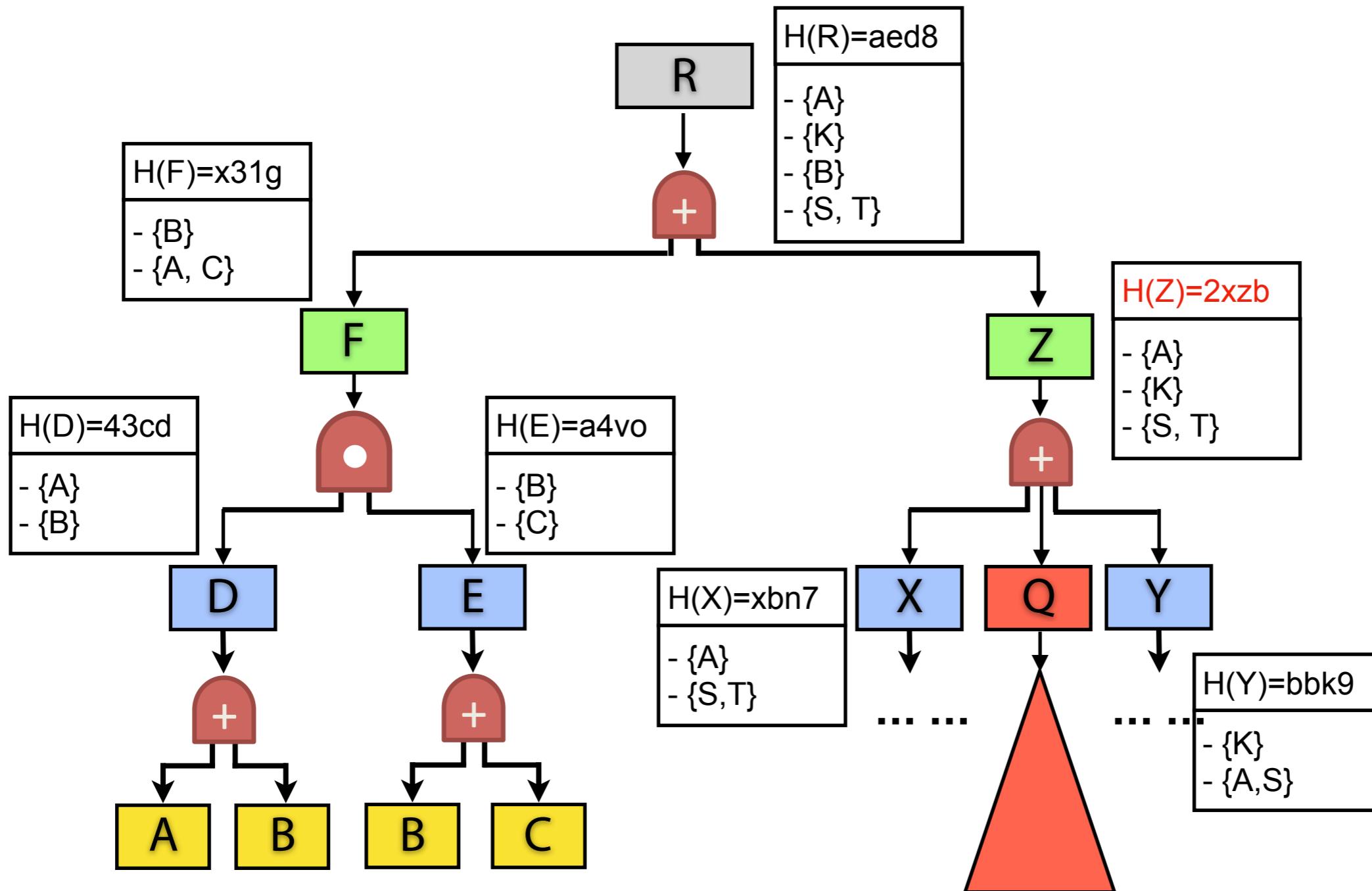
# Original Deployment



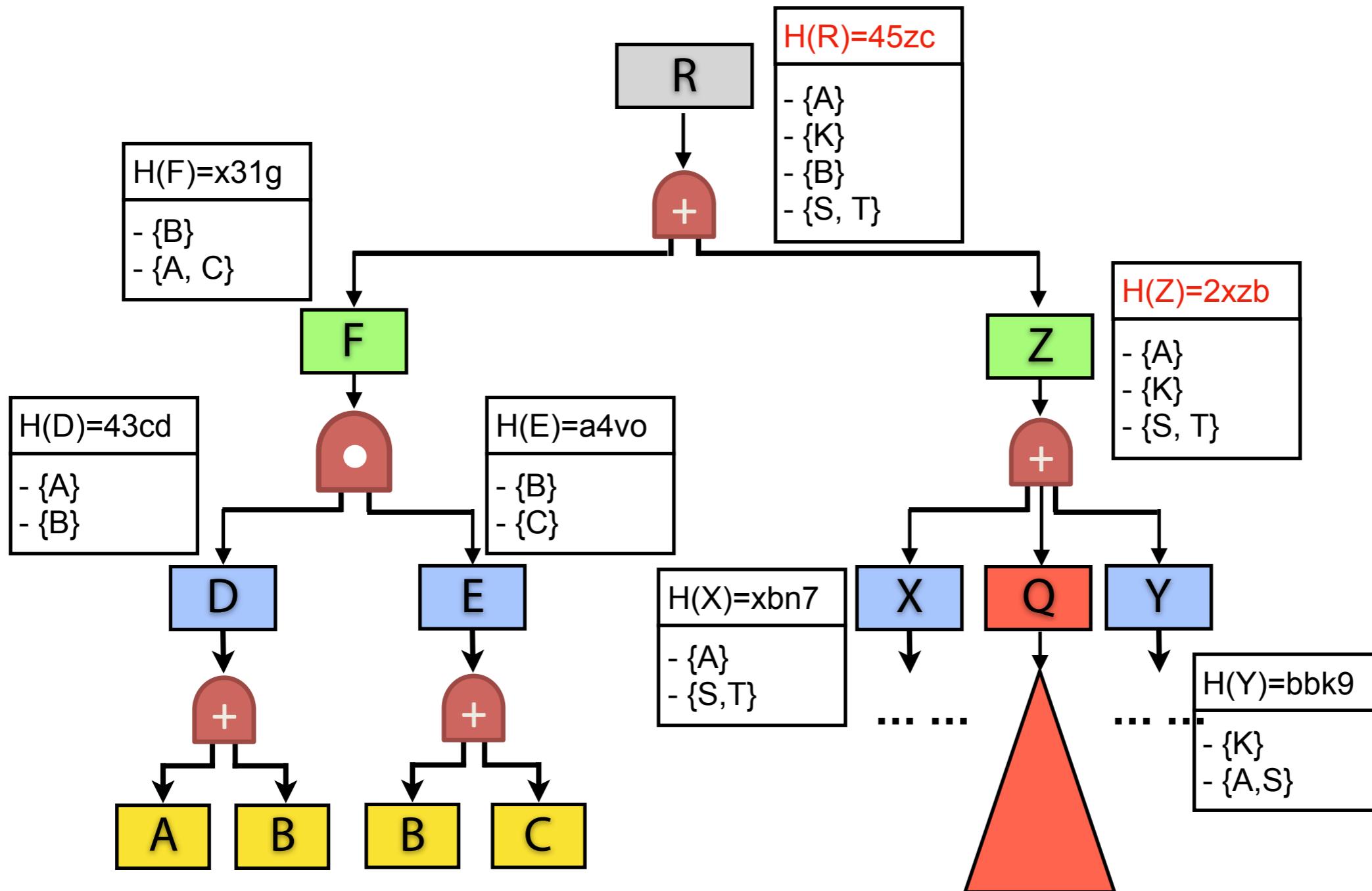
# Updated Deployment



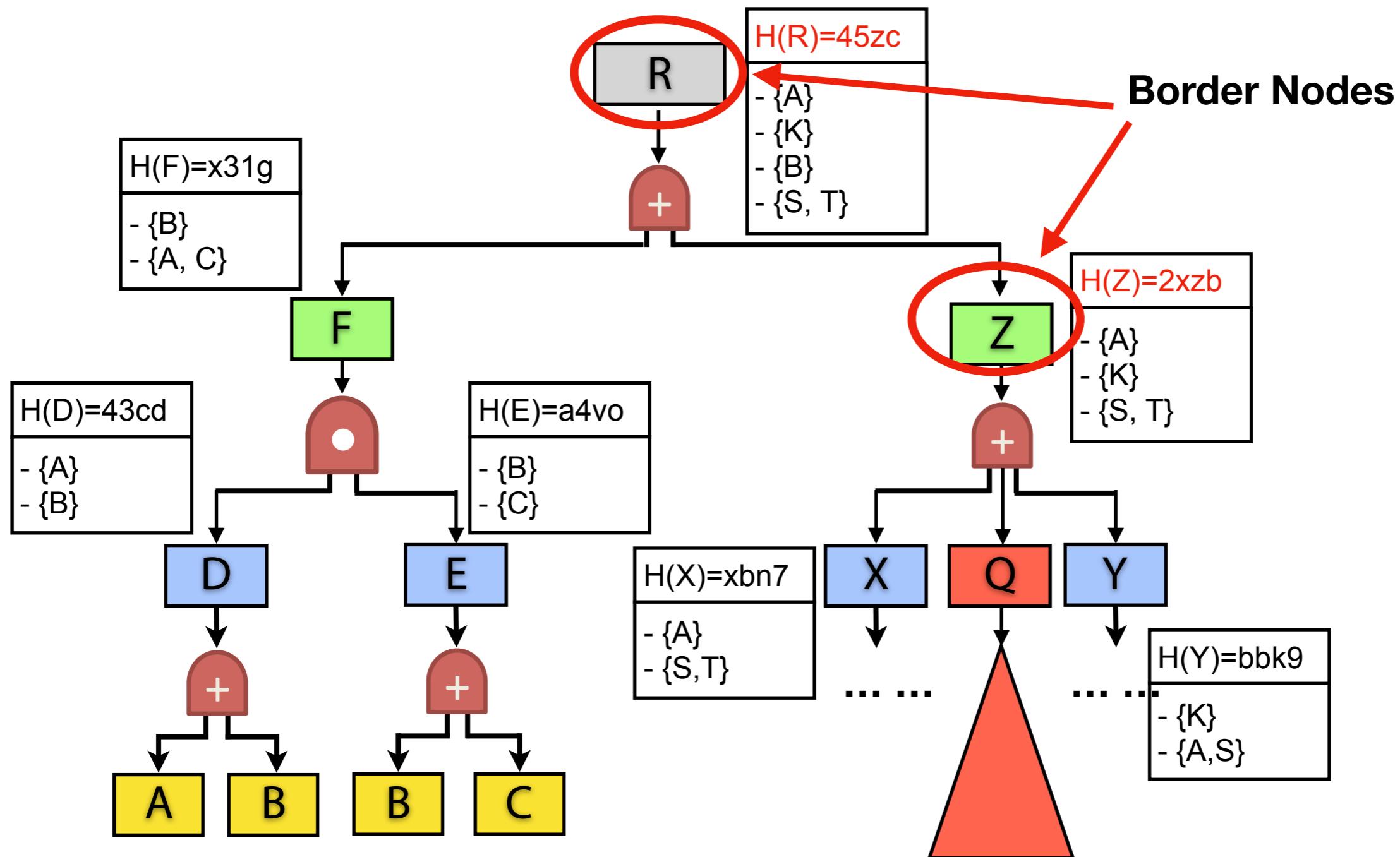
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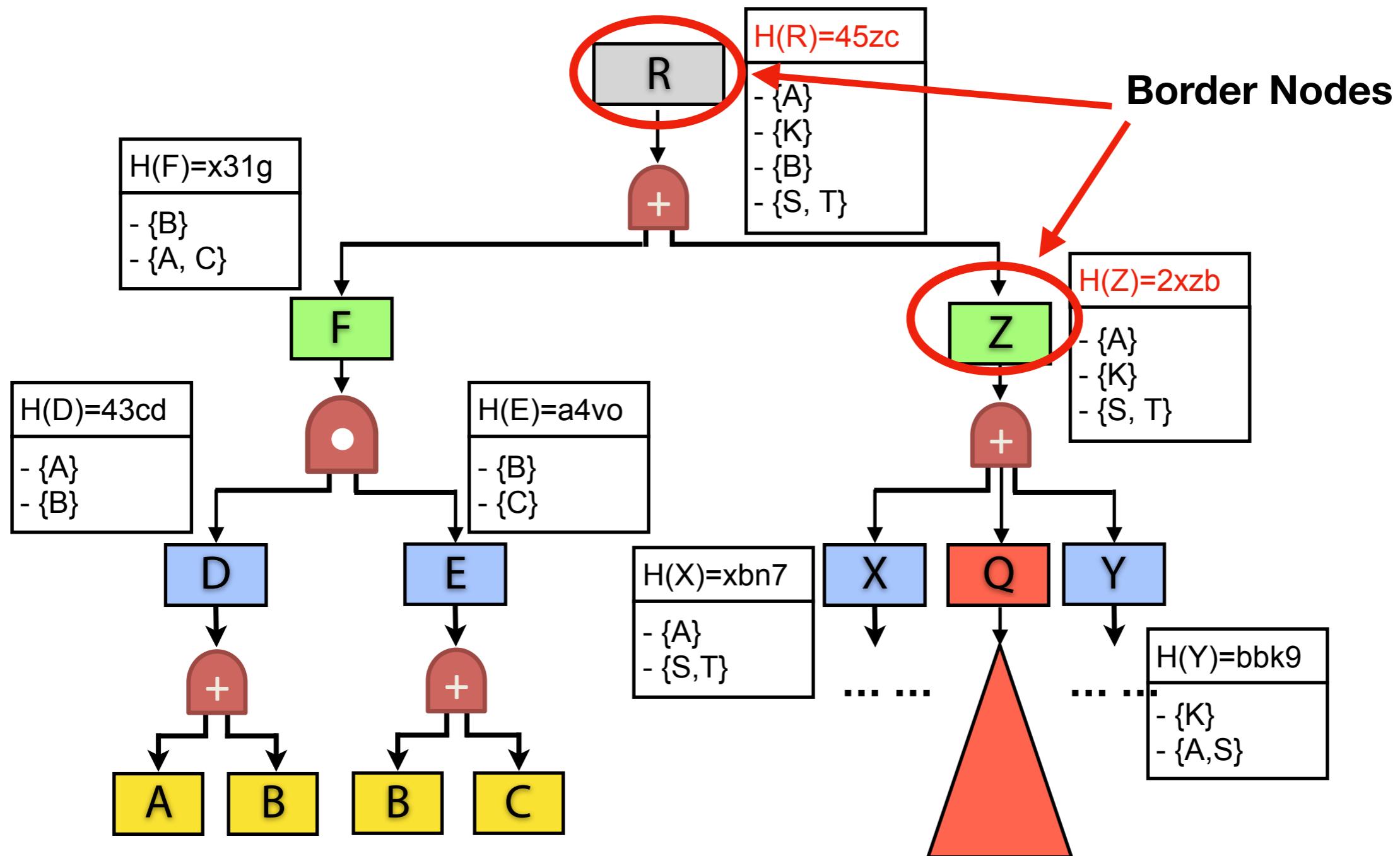
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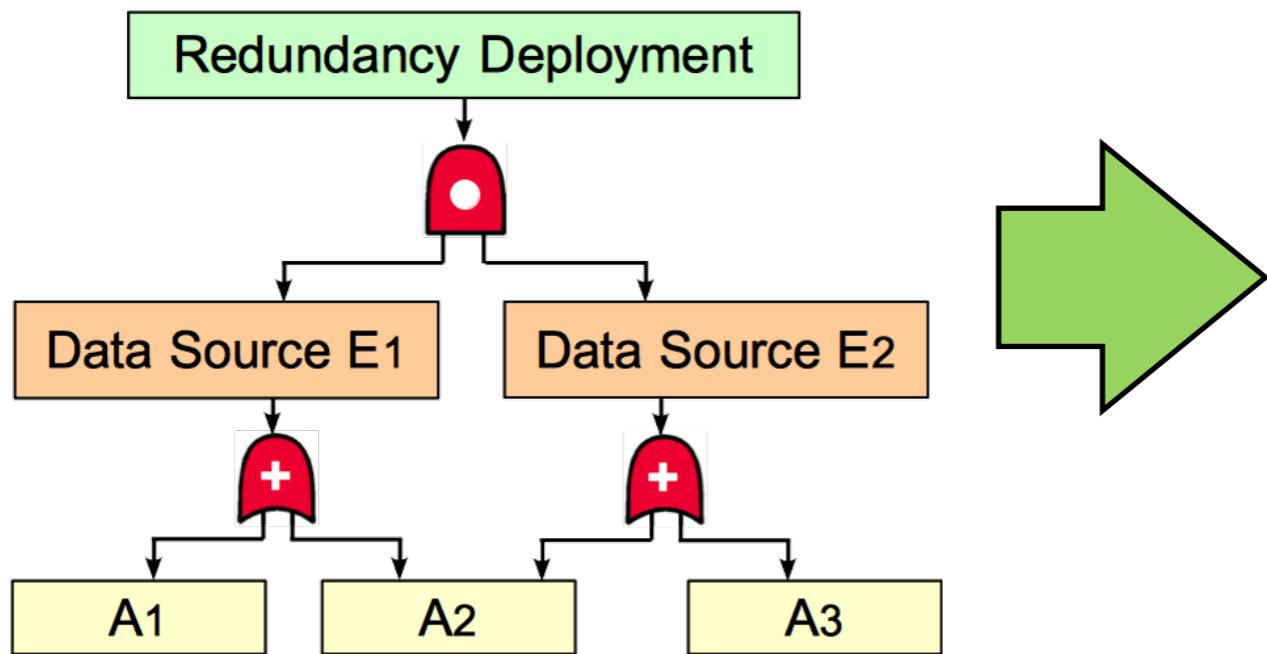
# Step 1: Find Border Nodes



## Step 2: Q's Risk Groups

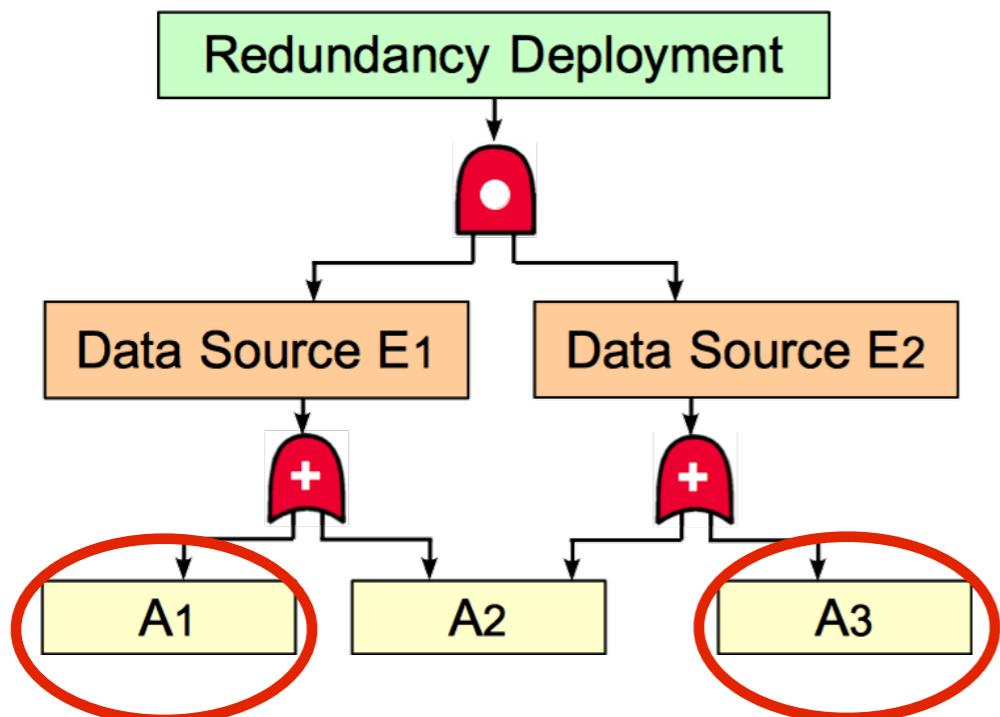


# Our Insight



**Boolean formula**  
 $= E_1 \wedge E_2$   
 $= (A_1 \vee A_2) \wedge (A_2 \vee A_3)$

# Our Insight



**Boolean formula**  
 $= E_1 \wedge E_2$   
 $= (A_1 \vee A_2) \wedge (A_2 \vee A_3)$

Satisfying assignment:  
 $\{A_1=1, A_2=0, A_3=1\}$

SAT solver

# Our Insight

- Problem:
  - Standard SAT solver outputs an arbitrary satisfying assignment

# Our Insight

- Problem:
  - Standard SAT solver outputs an arbitrary satisfying assignment
  - What we want is top-k **critical (minimal)** risk groups

# Identifying Risk Groups

- Using MinCostSAT solver
  - Satisfiable assignment with the least weights
  - Obtain the least  $C = \sum c_i \cdot w_i$
  - Very fast with 100% accuracy

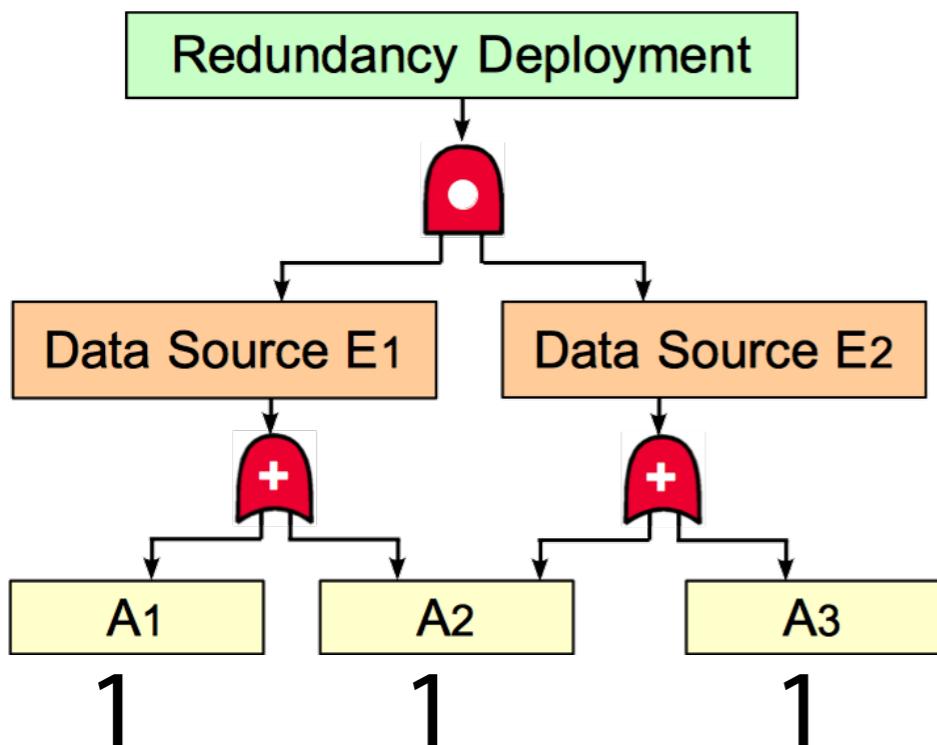
# Identifying Risk Groups

- Using MinCostSAT solver
  - Satisfiable assignment with the least weights
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We set the values of all the leaf nodes (i.e.,  $w_i$ ) as 1

# Identifying Risk Groups

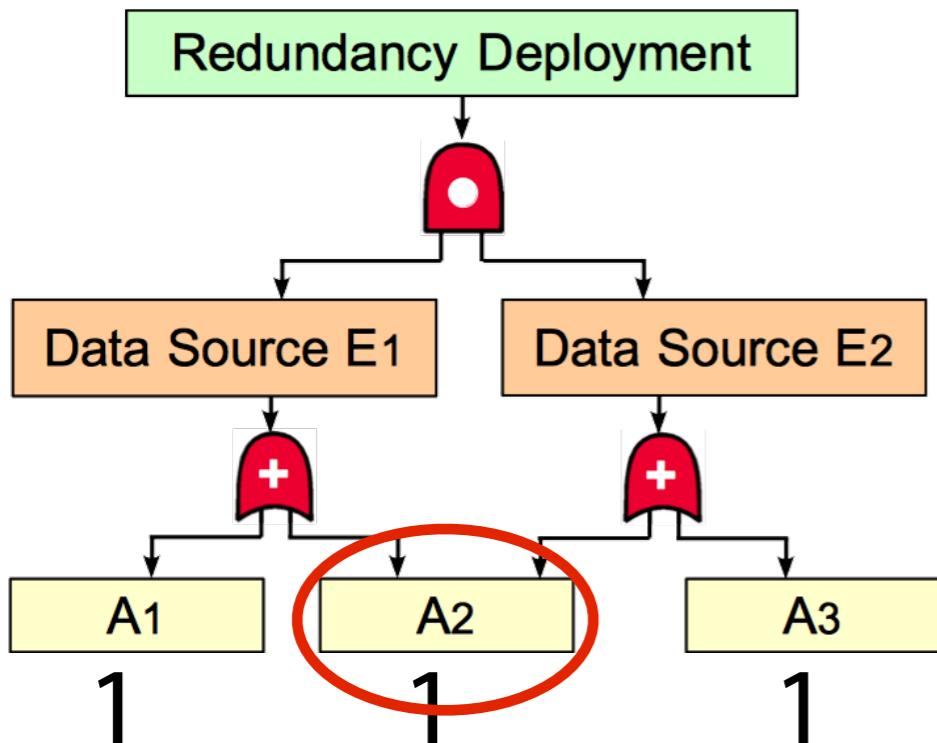
- Using MinCostSAT solver
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  - Very fast with 100% accuracy



A1	A2	A3	Weight
1	0	0	
0	1	0	1
0	0	1	
1	1	0	2
1	0	1	2
0	1	1	2
0	0	0	
1	1	1	3

# Identifying Risk Groups

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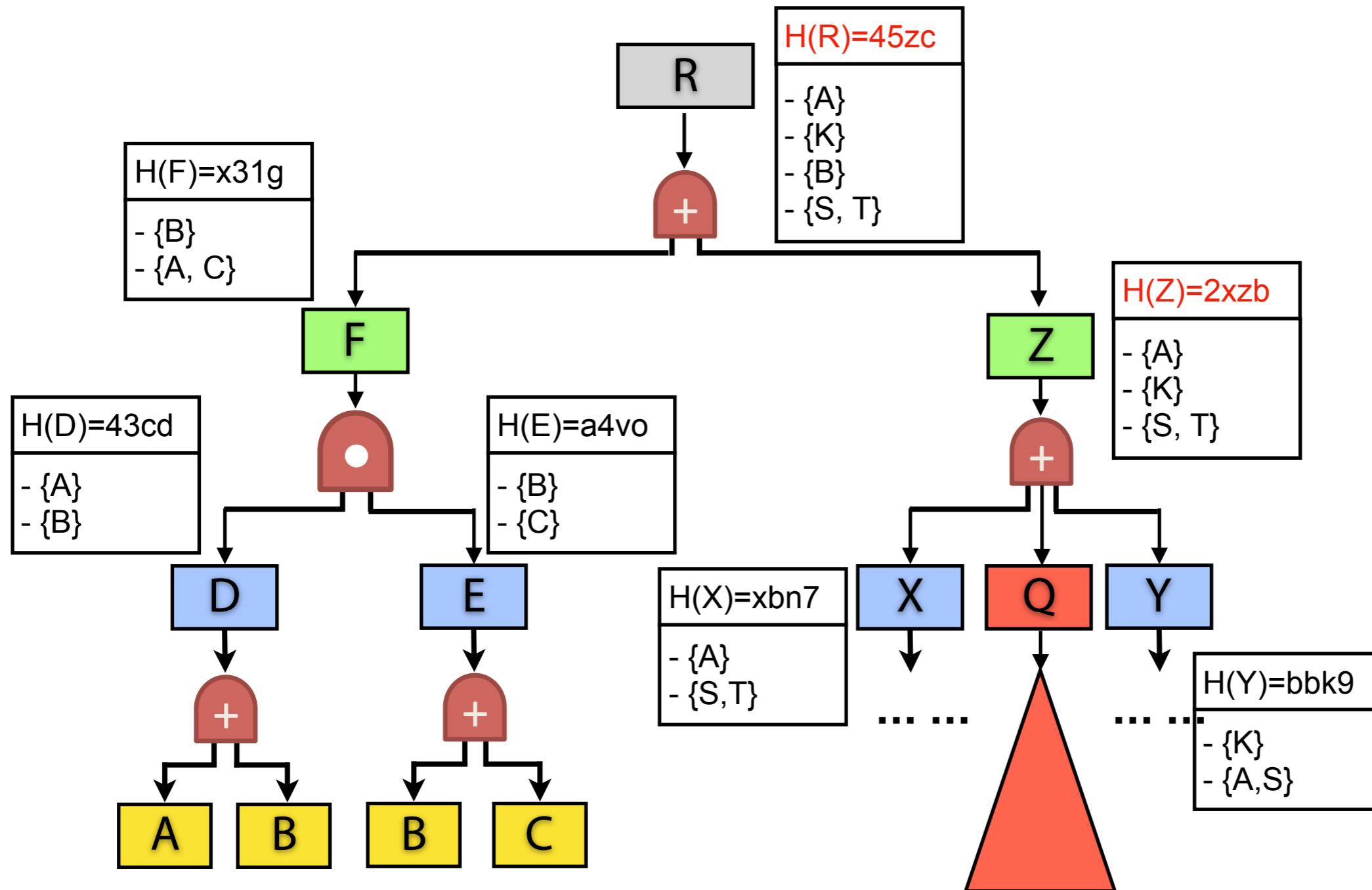
A1	A2	A3	Weight
1	0	0	
0	1	0	1
0	0	1	
1	1	0	2
1	0	1	2
0	1	1	2
0	0	0	
1	1	1	3

# Identifying Risk Groups

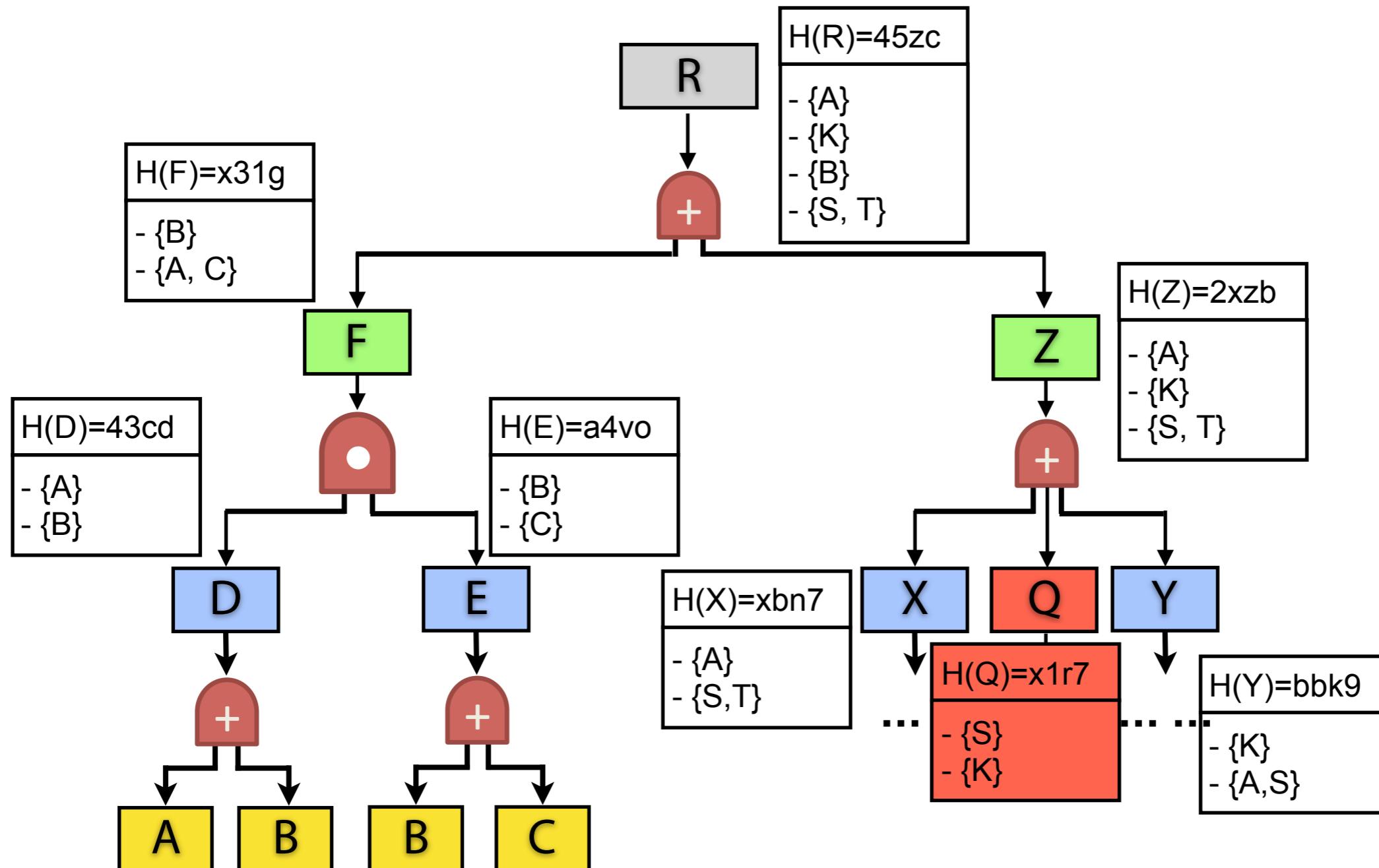
- Find out the top-k critical risk groups
  - Use a  $\wedge$  to connect the current formula and the negation of the resulting assignment

$$(A_1 \vee A_2) \wedge (A_2 \vee A_3) \wedge \neg(\neg A_1 \wedge A_2 \wedge \neg A_3)$$

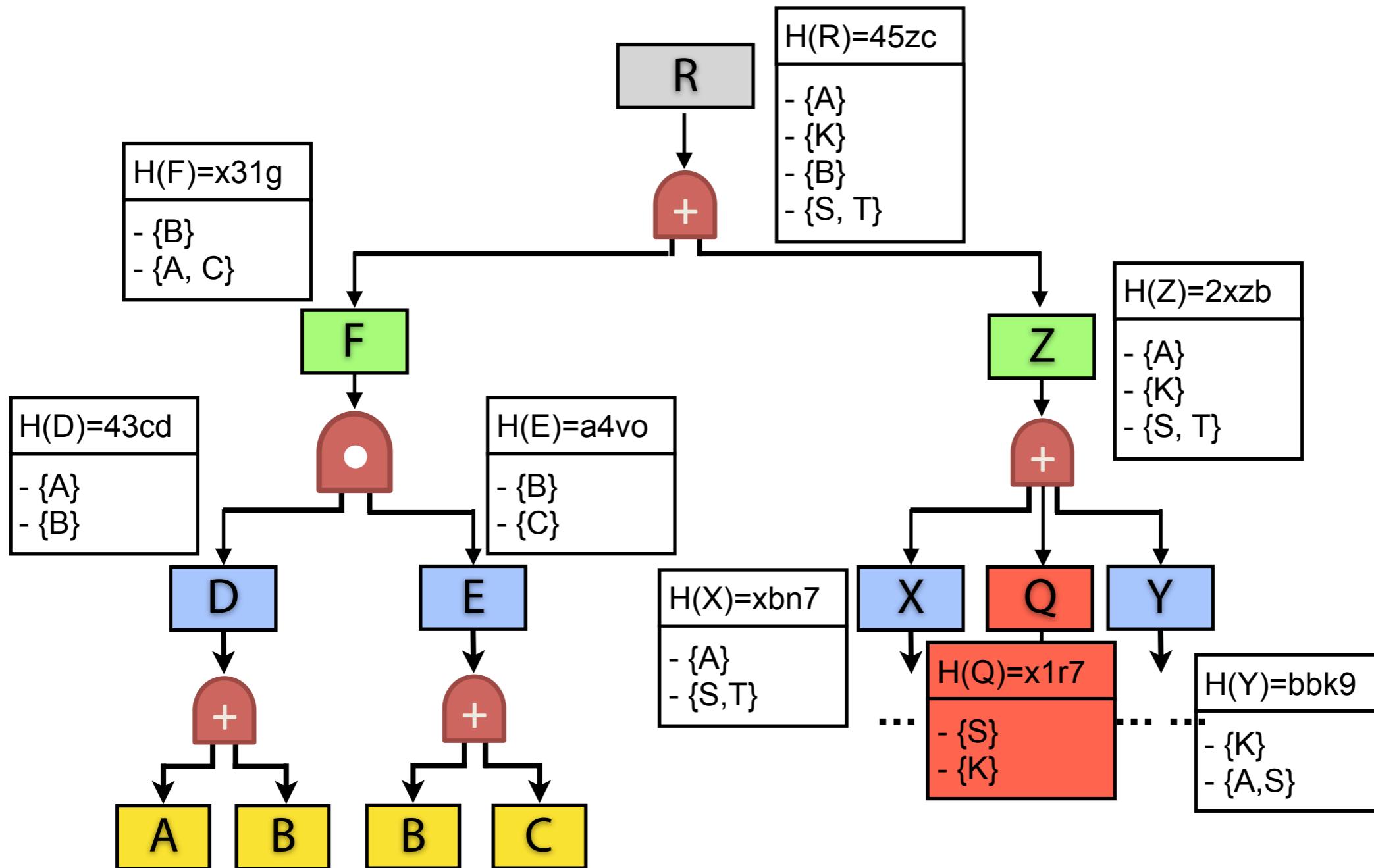
## Step 2: Q's Risk Groups



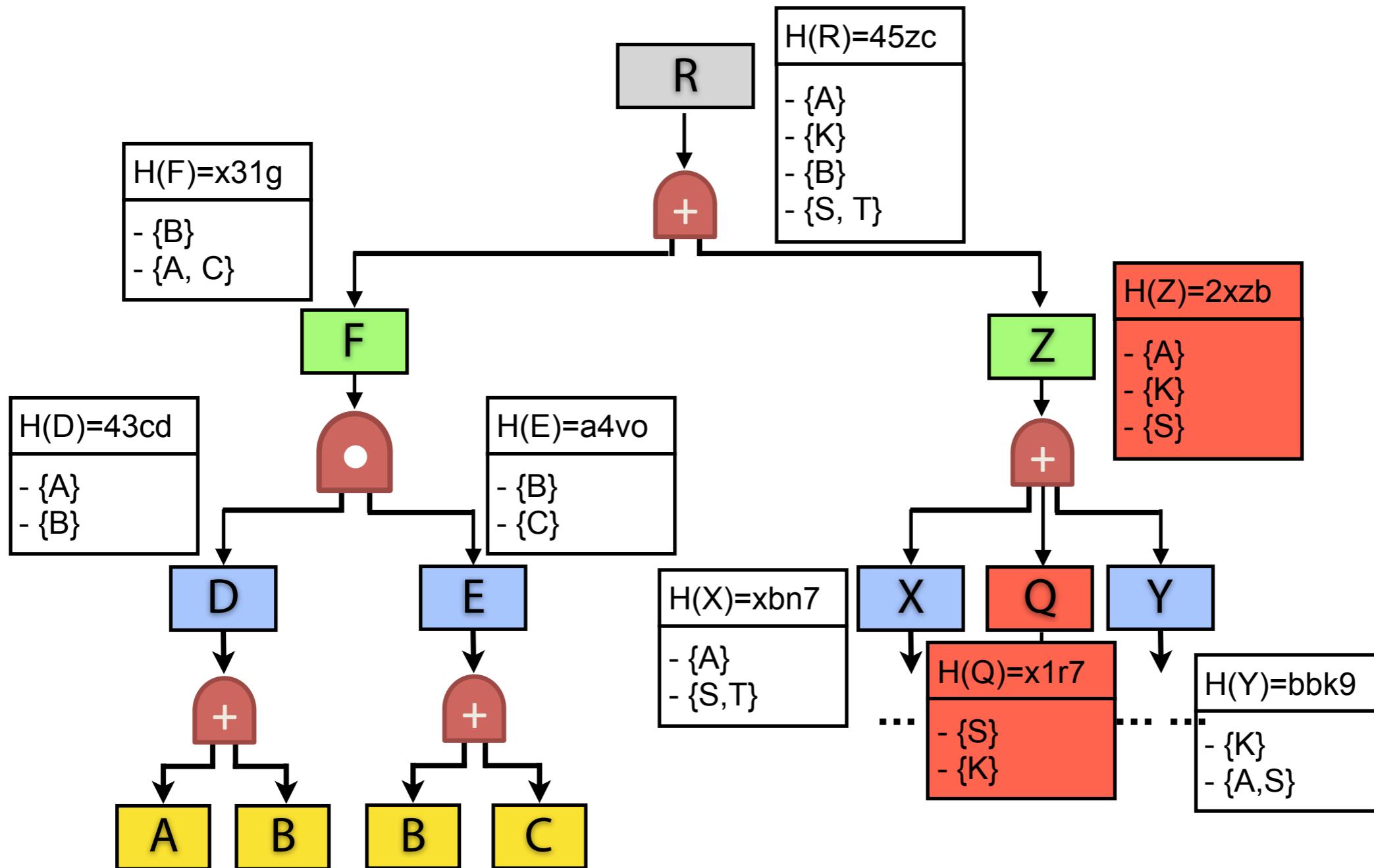
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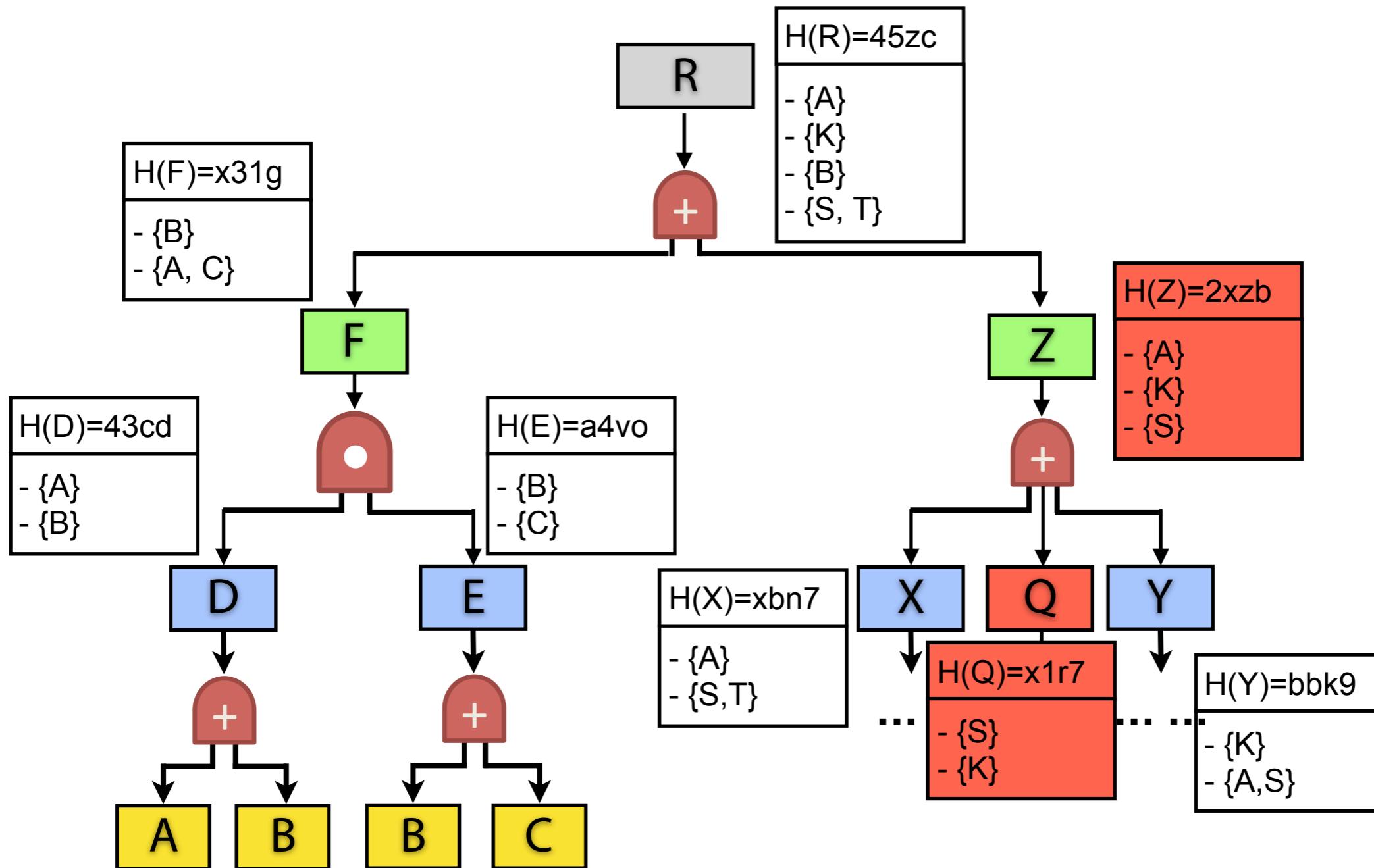
# Step 3: Merging Changed Caches



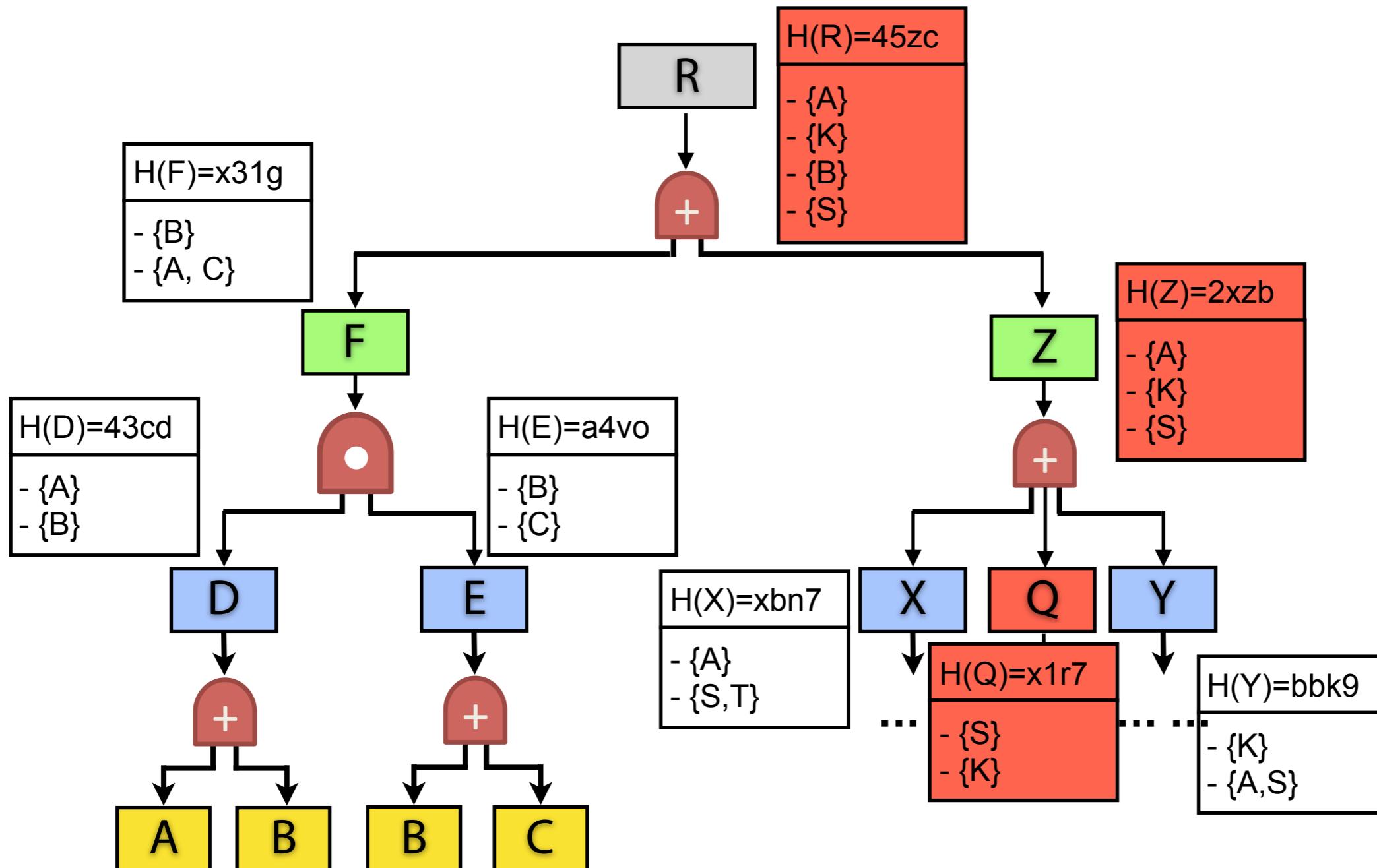
# Step 3: Merging Changed Caches



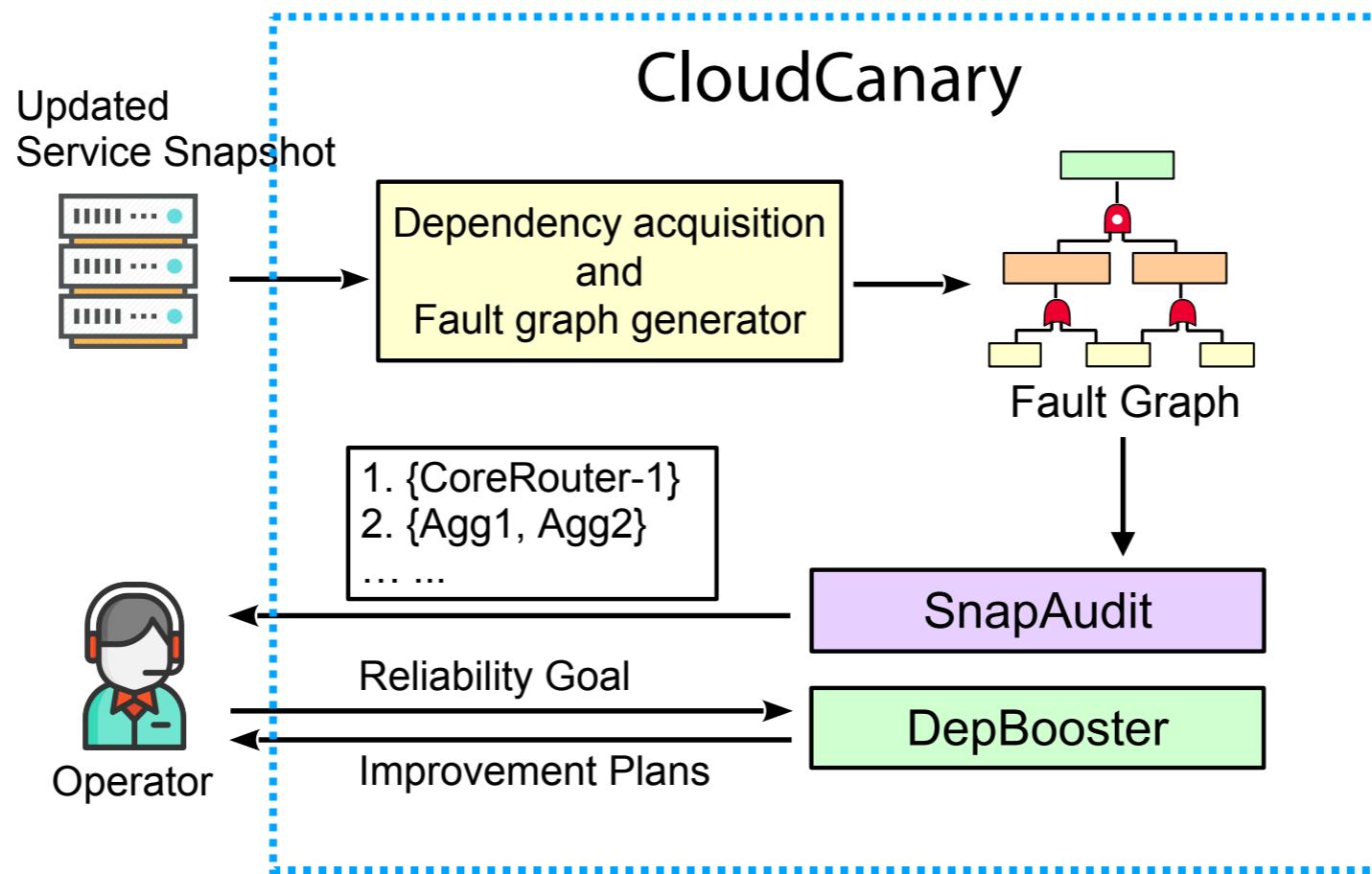
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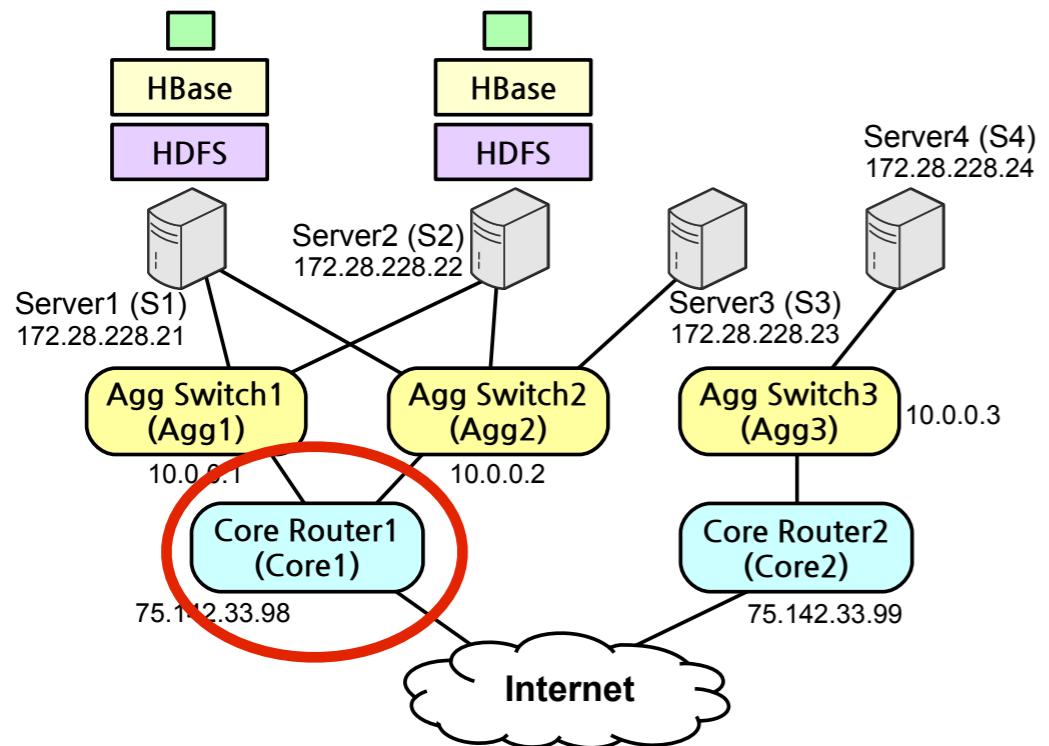


# CloudCanary's Workflow

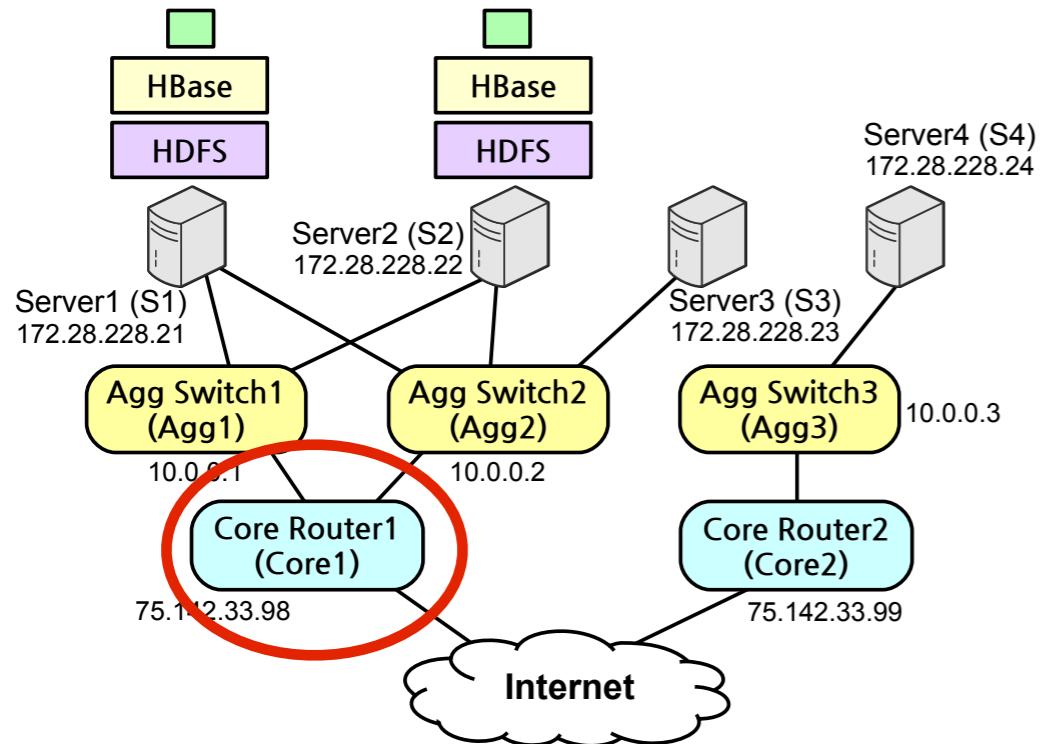


- Challenge 1: SnapAudit
- Challenge 2: DepBooster

# Correlated Failure Risk Repairing



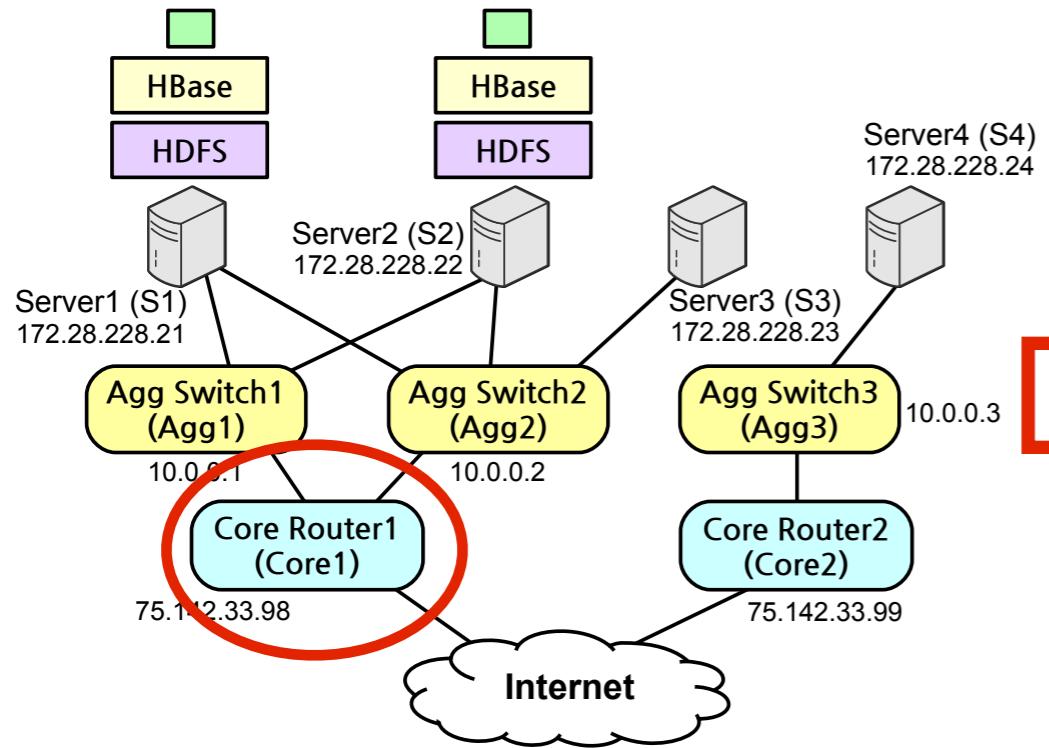
# Correlated Failure Risk Repairing



## Specification:

\$Server → 172.28.228.21, 172.28.228.22  
goal(failProb(ft)<0.08 | ChNode | Agg3)

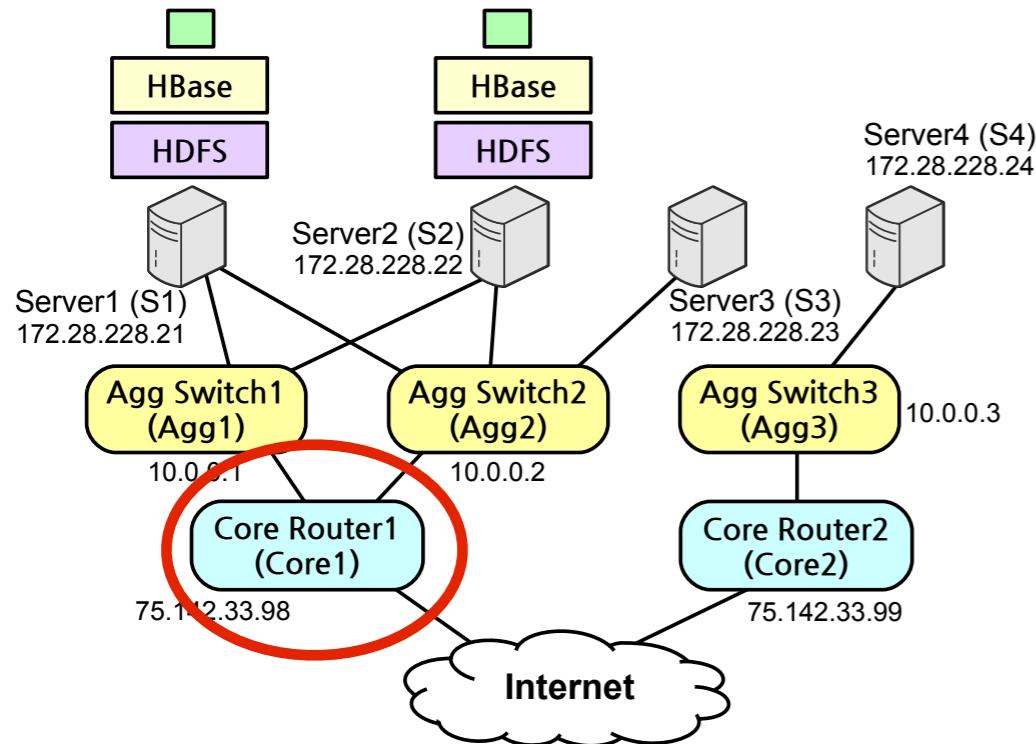
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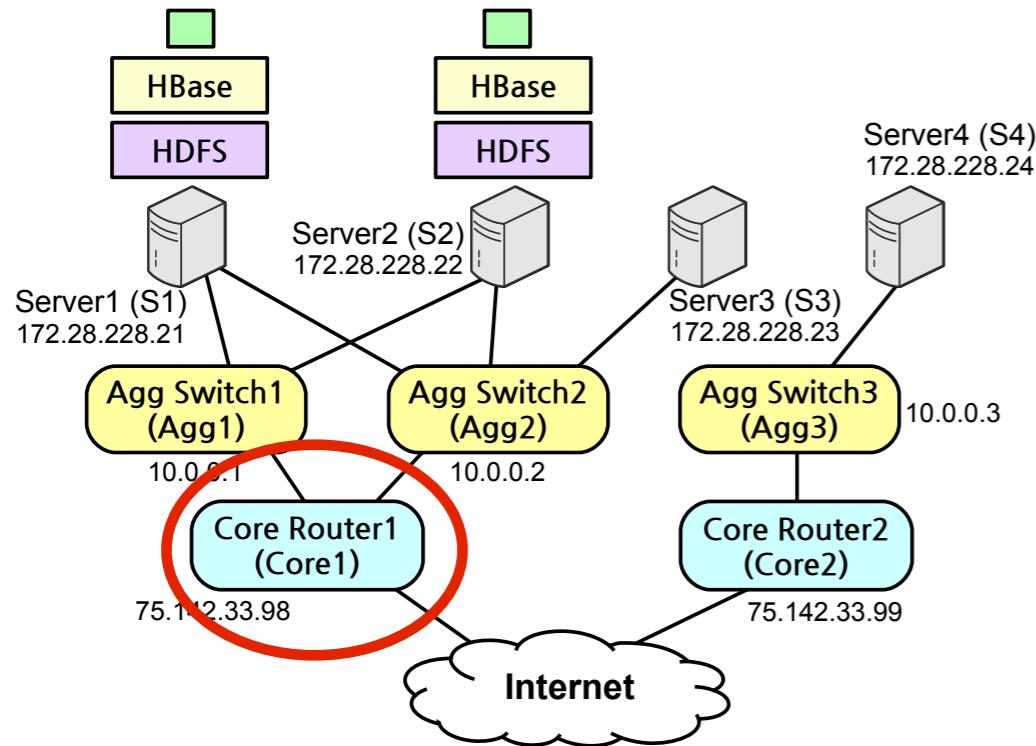


DepBooster



Plan 1: Move replica from S1 → S4  
Plan 2: Move replica from S2 → S4

# Correlated Failure Risk Repairing

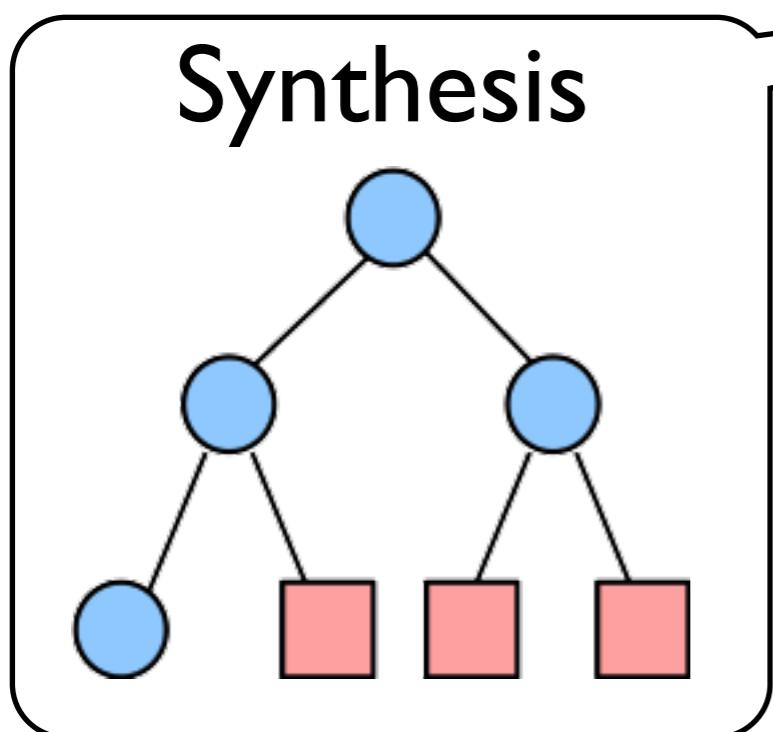


## Specification:

$\$Server \rightarrow 172.28.228.21, 172.28.228.22$   
goal(failProb(ft)<0.08 | ChNode | Agg3)

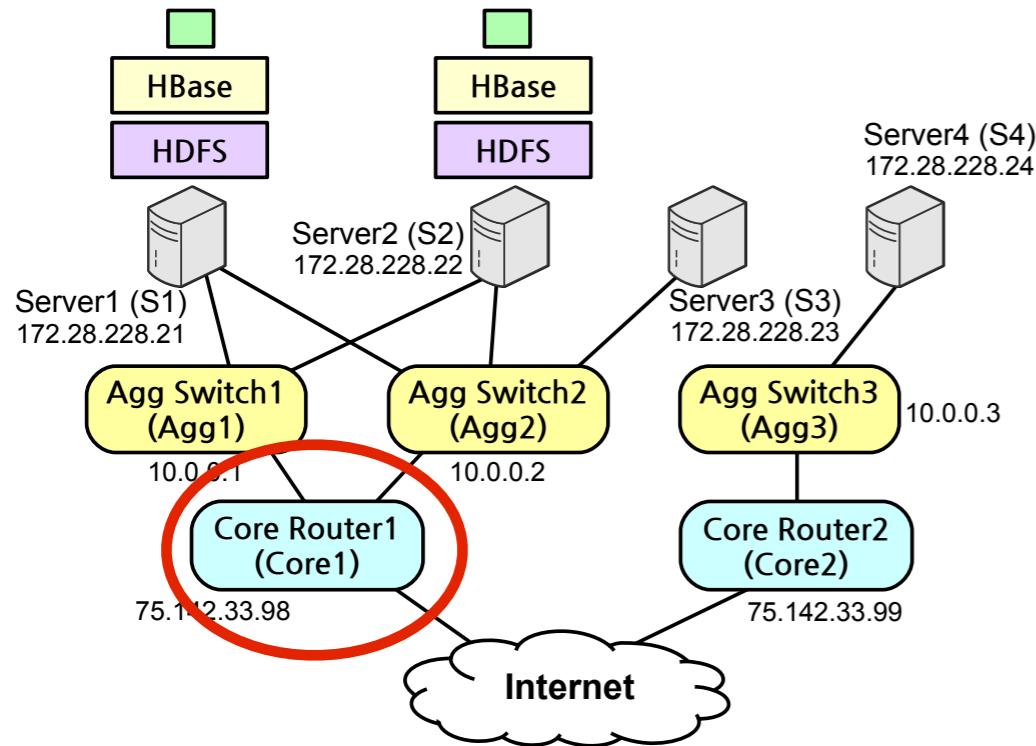


DepBooster



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## Specification:

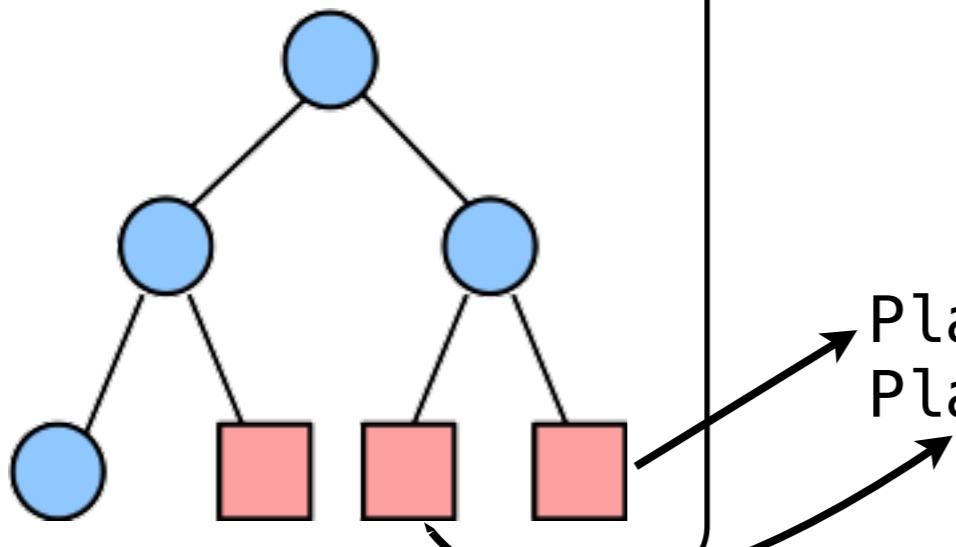
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DepBooster

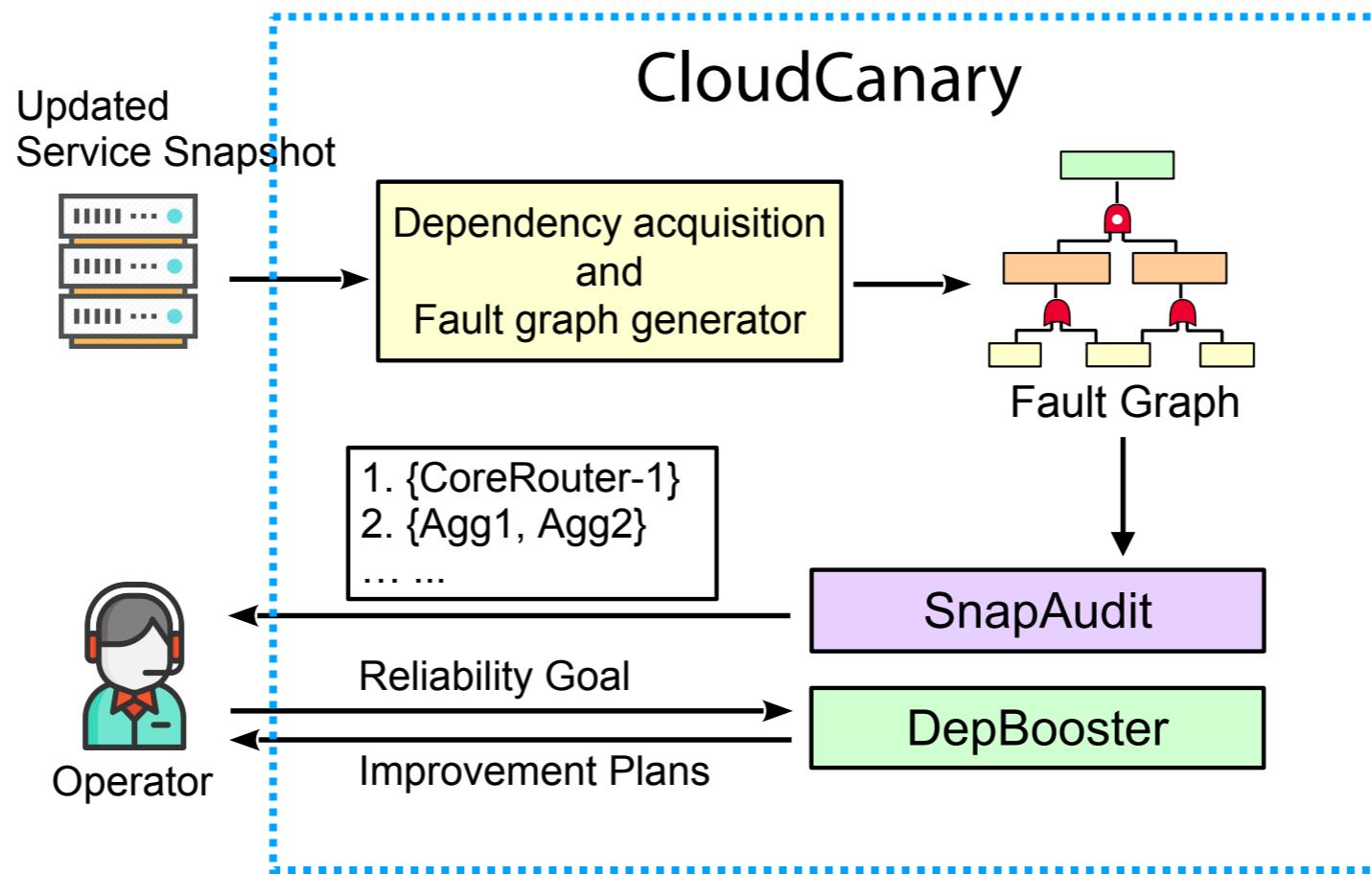


## Synthesis



Plan 1: Move replica from S1 -> S4  
Plan 2: Move replica from S2 -> S4

# CloudCanary's Workflow



- Challenge 1: SnapAudit
- Challenge 2: DepBooster

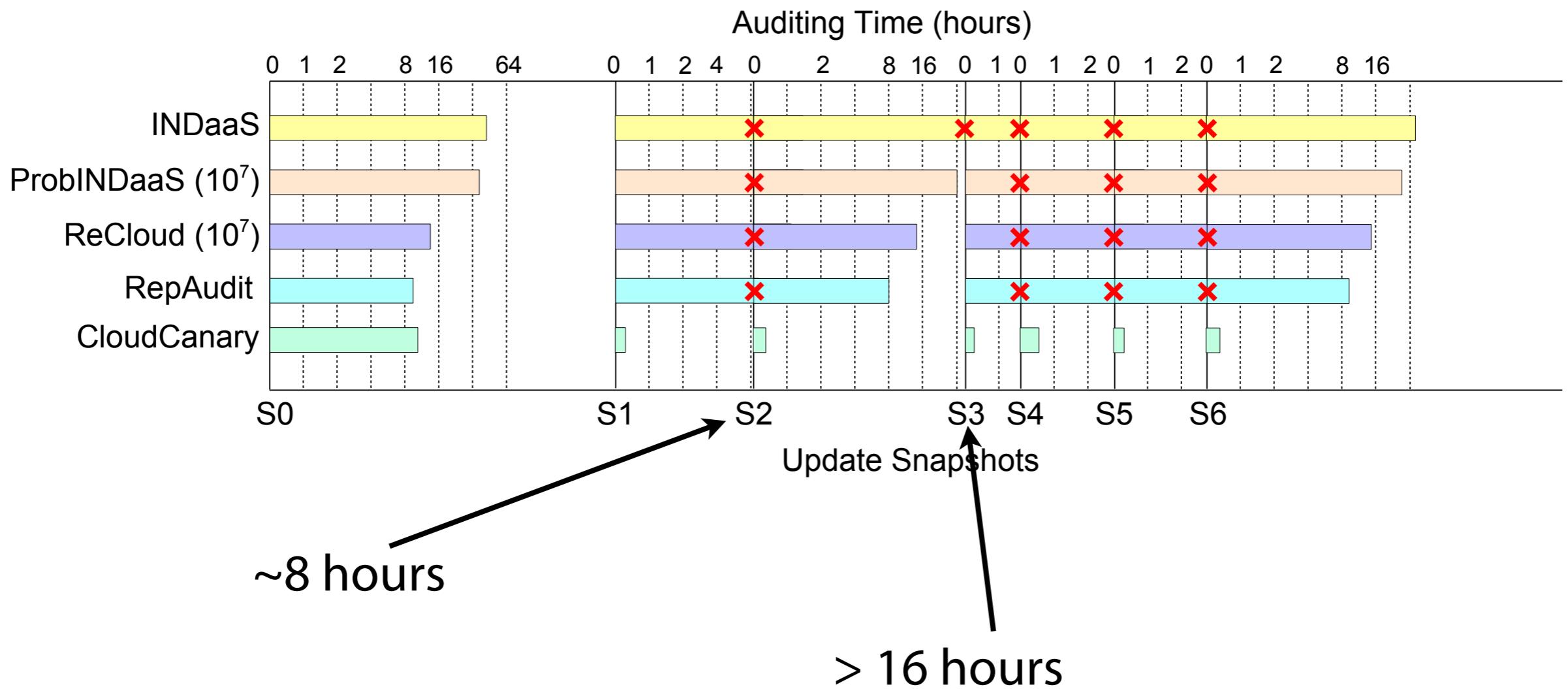
# Evaluation

- Comparing CloudCanary with the state of the art
- Evaluating CloudCanary's practicality via real dataset

# Evaluation

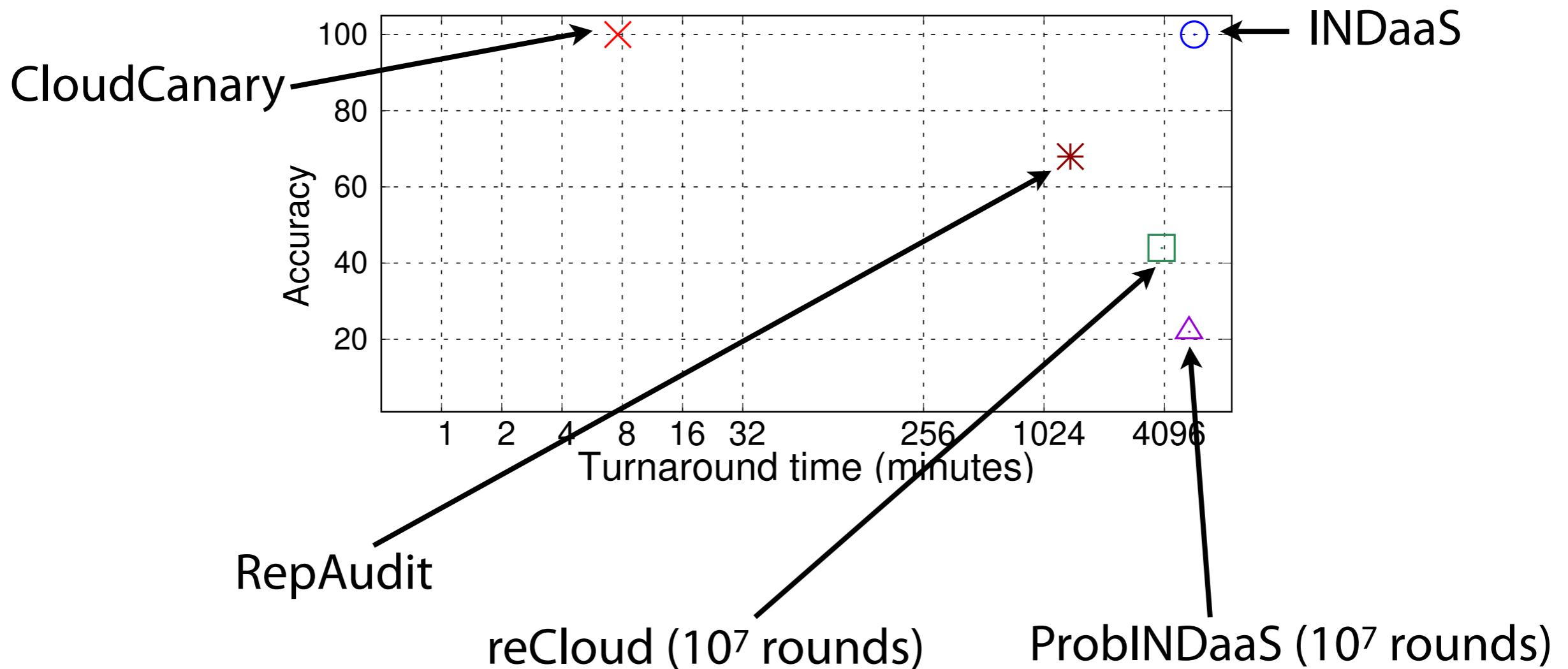
	Accuracy	Efficiency	Improvement
INDaaS [OSDI'14]	✓	✗	✗
ProbINDaaS [OSDI'14]	✗	✗	✗
reCloud [CoNEXT'16]	✗	✗	✗
RepAudit [OOPSLA'17]	✗	✗	✗
CloudCanary	✓	✓	✓

# Efficiency Comparison

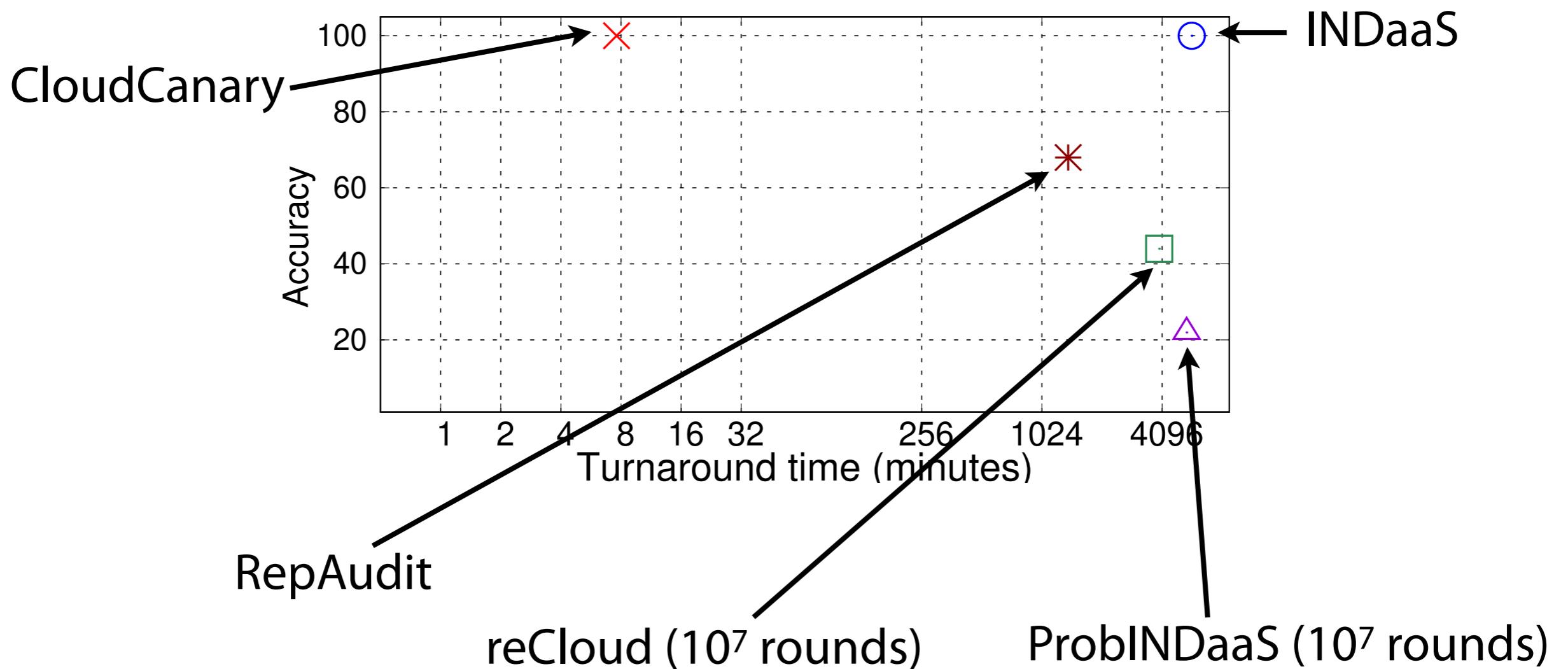


# Accuracy V.S. Efficiency

- 20,608 switches; 524,288 servers; 638,592 software components
- Auditing a random update affecting 20% components



Our approach is 200x faster than state-of-the-arts, and offers 100% accurate results.



# Evaluation

- We evaluated CloudCanary via real update trace:

	Detected Num	Confirmed	Examples
Microservices	50+	96%	Authentication and access control systems introduce most risk groups
Power Sources	10+	100%	Primary and backup power sources are carelessly assigned to multiple racks hosting a critical service
Network	30+	100%	Aggregation and ToR switches are easily updated to be risk groups

# Conclusion

- CloudCanary is the first system for real-time auditing
  - SnapAudit primitive: Quickly auditing update snapshot
  - DepBooster: Quickly generating improvement plans
- We evaluated CloudCanary with real trace and large-scale emulations

# Thanks, questions?

- CloudCanary is the first system for real-time auditing
  - SnapAudit primitive: Quickly auditing update snapshot
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