

# Structural Cloud Audits that Protect Private Information

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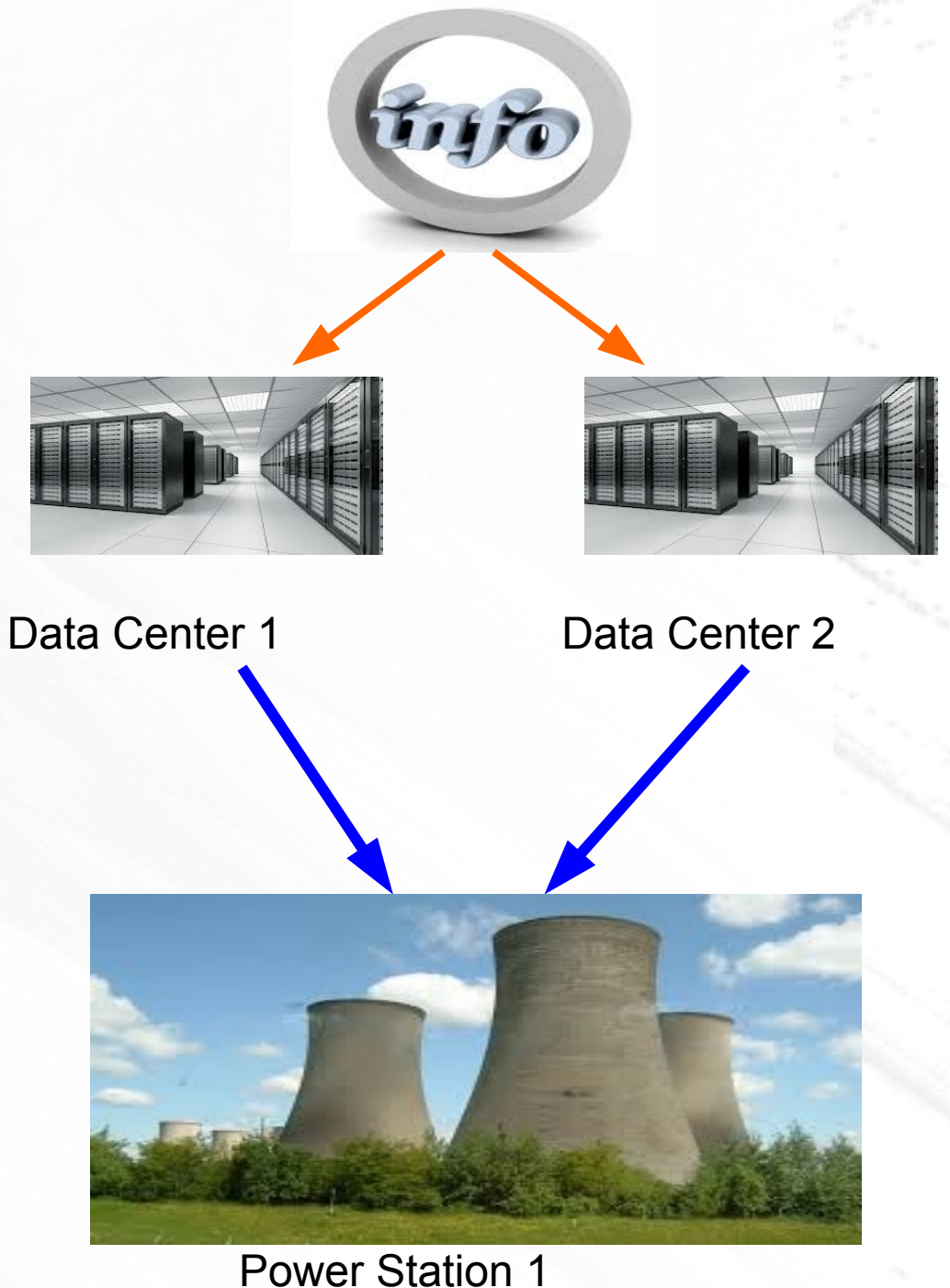
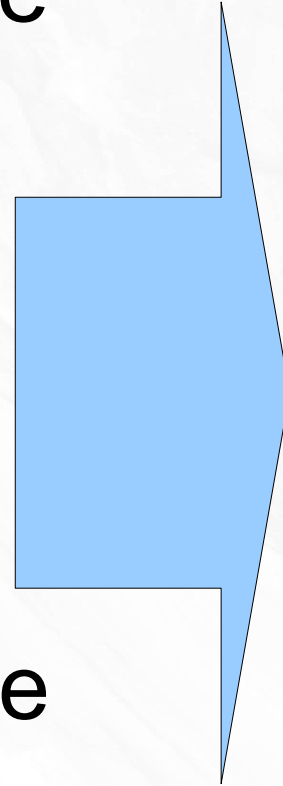
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# Motivation

- Cloud computing and cloud storage now plays a central role in the daily lives of individuals and businesses.
  - Over a billion people use Gmail and Facebook to create, share, and store personal data
  - 20% of all organizations use the commercially available cloud-storage services provided both by established vendors and by cloud-storage start-ups
- **Reliability** of cloud-service providers grows in importance.

# Motivation

- Cloud-service providers use redundancy to achieve reliability
- But redundancy can fail due to **Common Dependencies**



[Ford, *Icebergs in the Clouds*, HotCloud '12]

# Motivation

- This is a real problem
  - e.g. a lightning storm in northern Virginia took out both the main power supply and the backup generator that powered all of Amazon EC2's data centers in the region
- We need a systematic way to discover and quantify vulnerabilities resulting from common dependencies

# Motivation

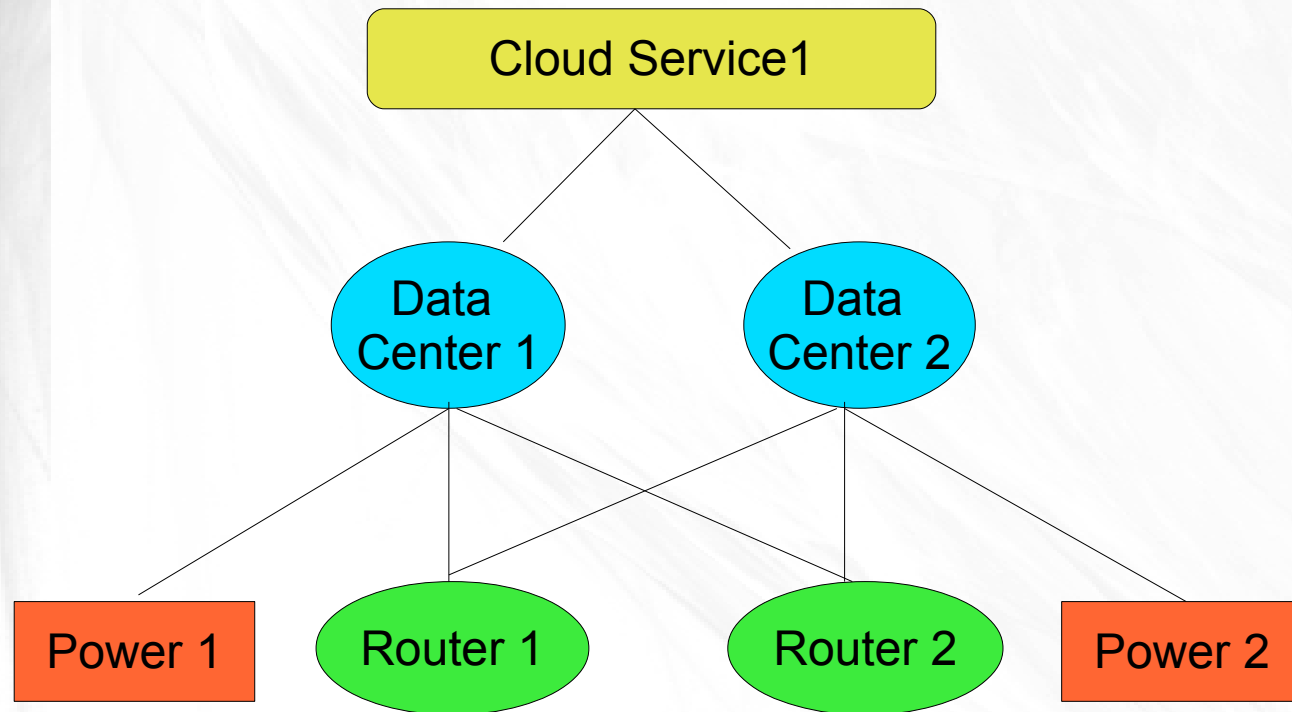
- Zhai et al. proposed Structural Reliability Auditing (SRA)
  - collect comprehensive information from infrastructure providers
  - construct a service-wide **fault tree**
  - identify critical components, estimate likelihood of service outage
- A potential **barrier to adoption** of SRA is the sensitive nature of both its input and its output.
  - cloud service providers and infrastructure providers may not be **willing to disclose the required information**

# Objective

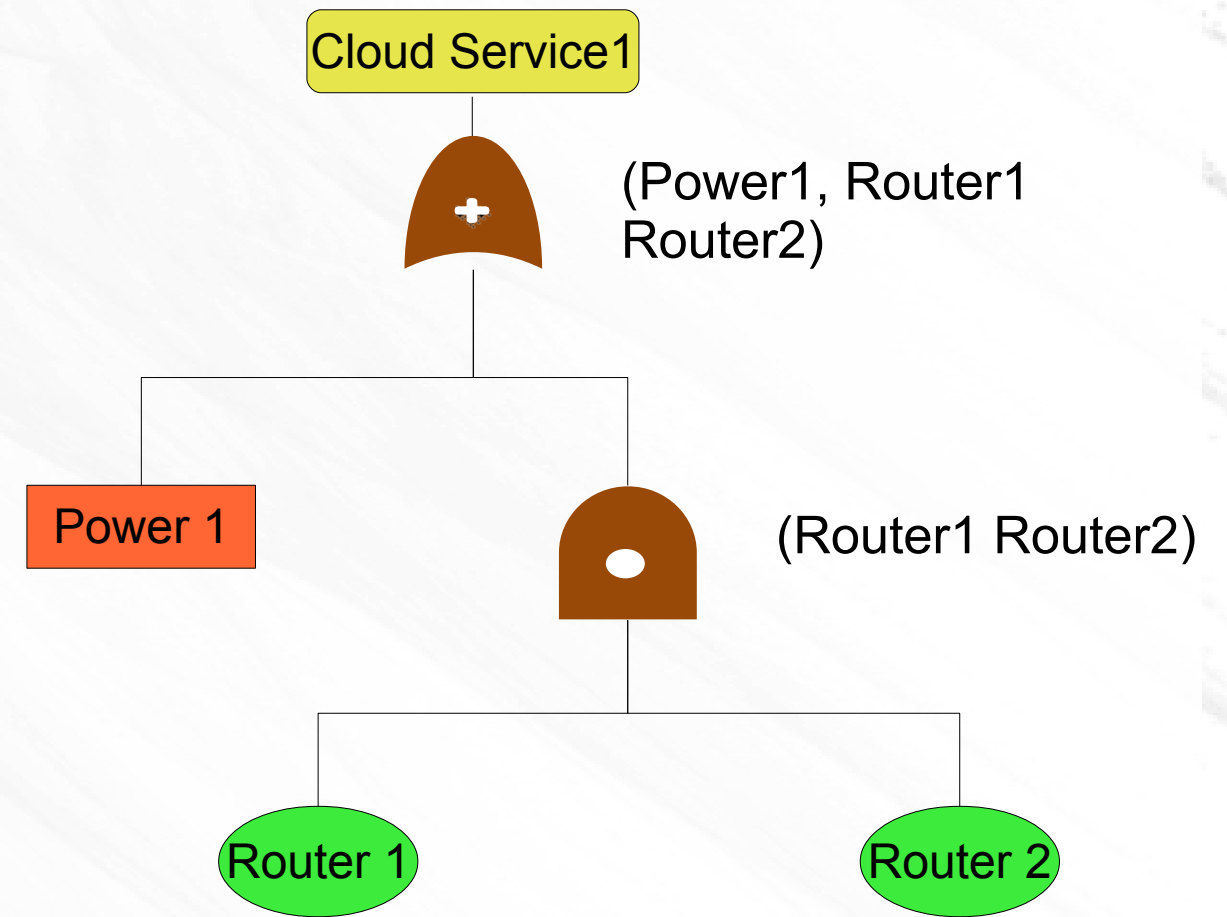
- Privacy-Preserving SRA (P-SRA): investigate the use of **secure multi-party computation** (SMPC) to perform SRA in a privacy preserving manner
  - Perform SMPC on complex, linked data structures of cloud topology, which has not often been explored yet

# Basic Idea

Step 1: Build a *structural model* of cloud infrastructure of interest



Step 2: Perform *fault tree analysis* to detect hidden failure risks



[Zhai et al., *Auditing the Structural Reliability of the Clouds*, Yale TR-1479]

# Challenges

- Private Data Acquisition
  - How to collect complex, linked data of cloud topology without compromising the privacy of the cloud and infrastructure providers?
- Privacy-Preserving Analysis
  - How to identify common dependencies and correlated failure risk without requiring providers to disclose confidential information?
- Efficiency
  - SMPC is NOT very efficient especially when the size of inputs are large



# Our Solutions

- **Private Data Acquisition**
  - Leverage secret sharing techniques in SMPC
  - Specify valid output protecting privacy
- **Privacy-Preserving Analysis**
  - Specialized graph representation techniques to build fault tree in a privacy preserving manner
- **Efficiency**
  - Novel data partitioning techniques to effectively reduce the input size of SMPC and leave most of the computations locally

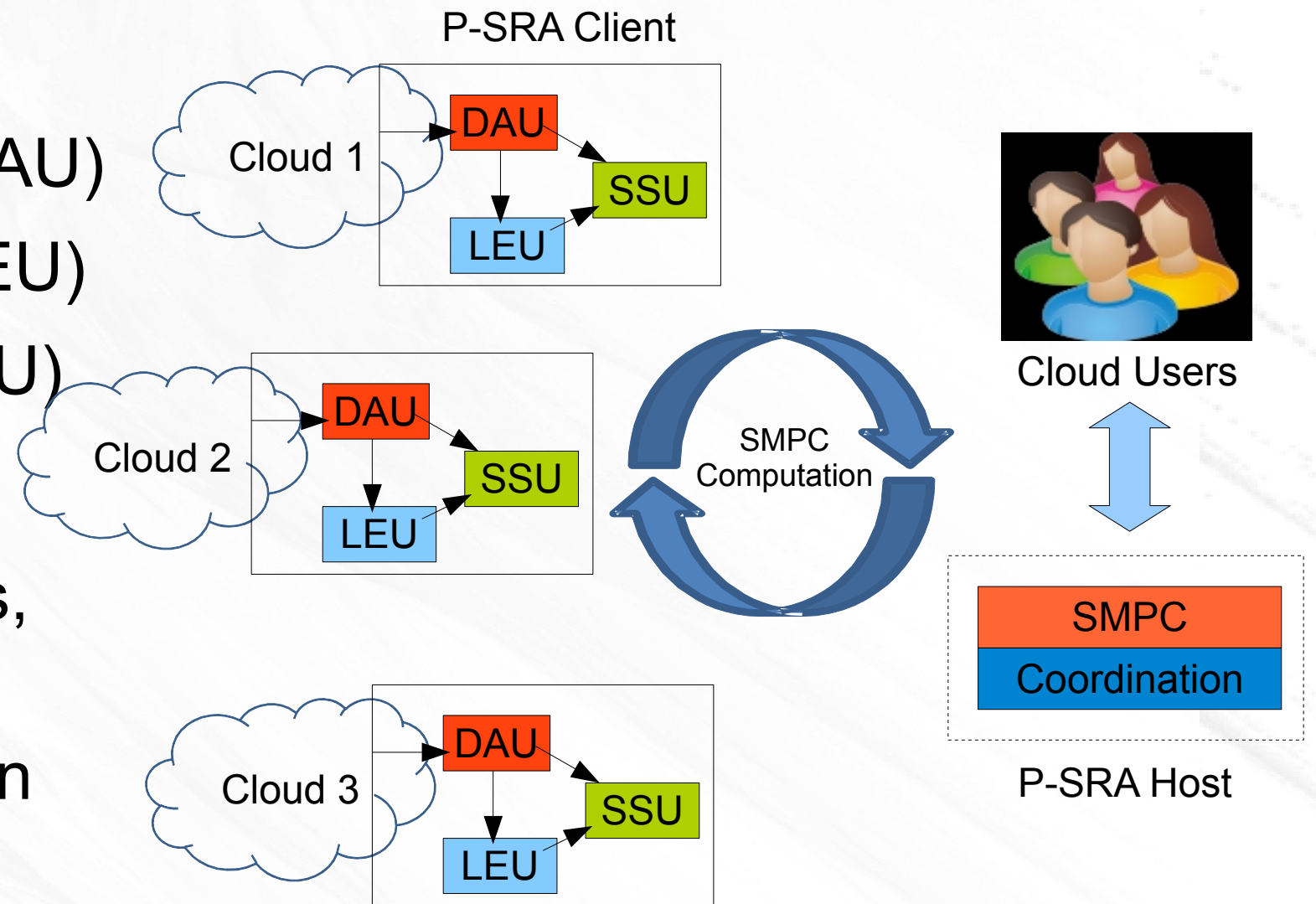
# System Design Overview

- P-SRA Client

- Data Acquisition Unit (DAU)
- Local Execution Unit (LEU)
- Secret Sharing Unit (SSU)

- P-SRA Host

- Represents Cloud Users, Reliability Auditors
- Does SMPC coordination



# Cloud Provider

- Install and control a P-SRA Client
- Input their private infrastructure information, which is considered private
- Semi-honest Threat Model
  - The Cloud Providers are honest but curious

# P-SRA Client

- Fully controlled by Cloud Providers
- Data Acquisition Unit
  - Collects component and dependency information
- Local Execution Unit
  - Perform local structural reliability analysis
- Secret Sharing Unit
  - Perform SMPC with P-SRA Host

# P-SRA Host

- SMPC module
  - Perform SMPC with each P-SRA client installed by cloud providers
- Coordination module
  - Coordinate the communication between P-SRA Clients and P-SRA Host
- Semi-honest Model
  - The P-SRA Host is honest but curious

# Outline of How the System Works

- Step 1: Privacy-preserving dependency acquisition
- Step 2: Subgraph abstraction to reduce problem size
- Step 3: SMPC protocol execution and local computation
- Step 4: Privacy-preserving output delivery

# Privacy-preserving dependency acquisition

- The DAU of each cloud-service provider collects information about the components and dependencies of this provider
  - network dependencies
  - hardware dependencies
  - software dependencies
  - failure probability estimates for components
- Store the information in a local database for use by P-SRA's other modules.

# Subgraph Abstraction

- The Client's SSU abstracts the dependency information of private components as a set of **macro-components**, which are the actual inputs of the SMPC
- Key step to reduce the input size of SMPC
- The choice of abstraction policy is flexible as long as satisfying the proper criterions
- Can be generalized to other SMPC problem on complex and linked data structure

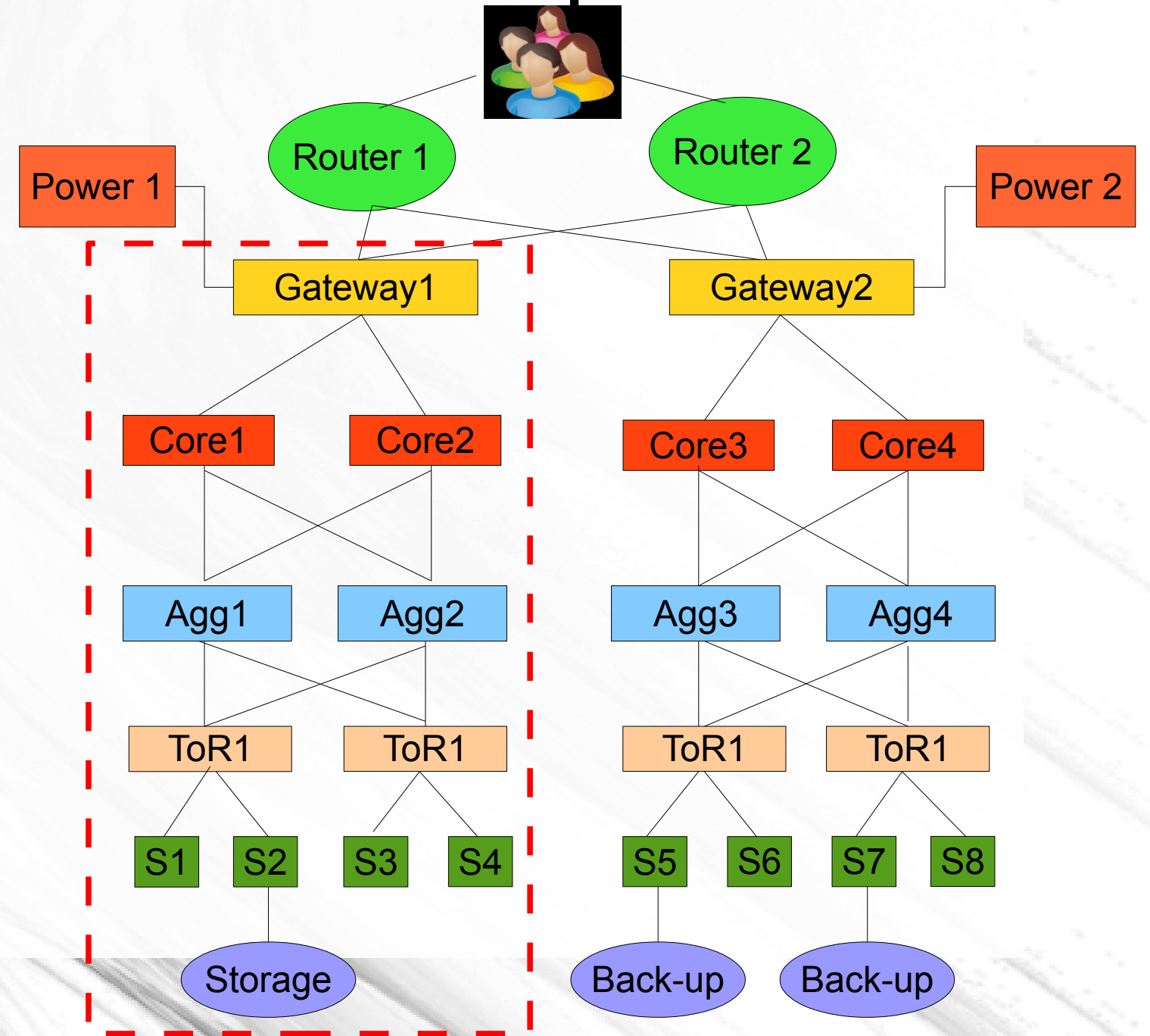


# Subgraph Abstraction Policy

- A subgraph  $H$  of the full dependency graph  $G$  of a cloud-service provider  $S$  should have two properties in order to be eligible for abstraction as a macro-component
  - all components in  $H$  must be used only by  $S$
  - for any two components  $v$  and  $w$  in  $H$ , the dependency information of  $v$  with respect to components outside of  $H$  is identical to that of  $w$
- SSU collapses  $H$  to a single node to transfer  $G$  to a smaller graph  $G'$

# Subgraph Abstraction: Example

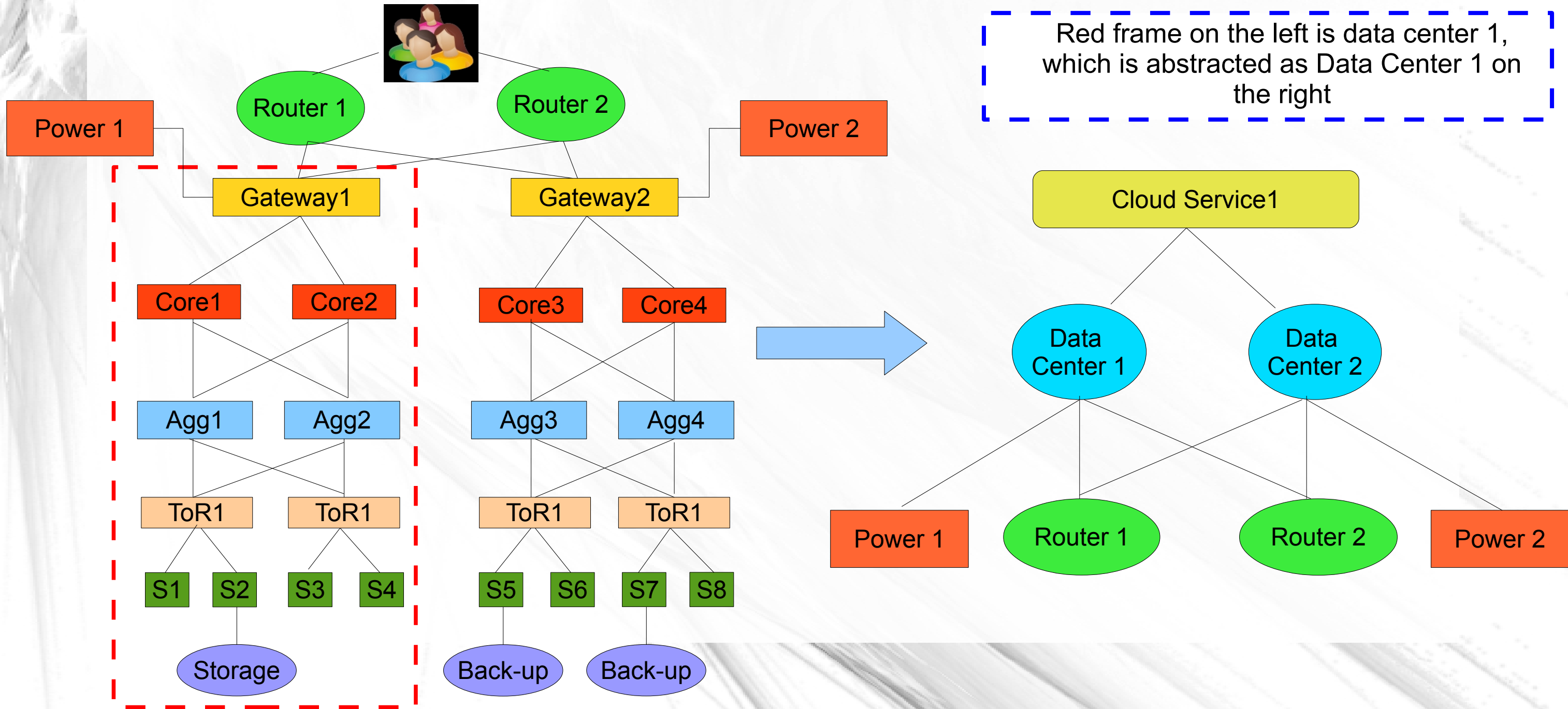
- Dependency Graph of a Simple Data Center
  - A Storage Service
  - Two Data Centers, one for service and the other for back-up
- Red Frame is the data center 1, which satisfies the two properties



# Subgraph Abstraction: Example



Red frame on the left is data center 1, which is abstracted as Data Center 1 on the right



# SMPC and Local Computation

- SMPC

- Perform SMPC to identify common dependency and reliability analysis across cloud providers
- SSUs of P-SRA Clients work with SMPC of P-SRA Host

- Local Computation

- SSU passes the dependency information within macro-components to LEU
- LEU performs structural reliability analysis locally

# SMPC Protocol

- Fault-tree construction
- Generate input for the SMPC
- Identify common dependencies
- Calculate failure sets

# Fault Tree Analysis

- FTA is a deductive reasoning technique
  - Occurrence of top event is a boolean combination of occurrence of lower level events
- Fault Tree is a Directed Acyclic Graph (DAG)
  - Node: gate or event
  - Link: dependency information
- Failure Set is a set of components whose simultaneous failure results in cloud service outage

# SMPC Fault Tree Construction

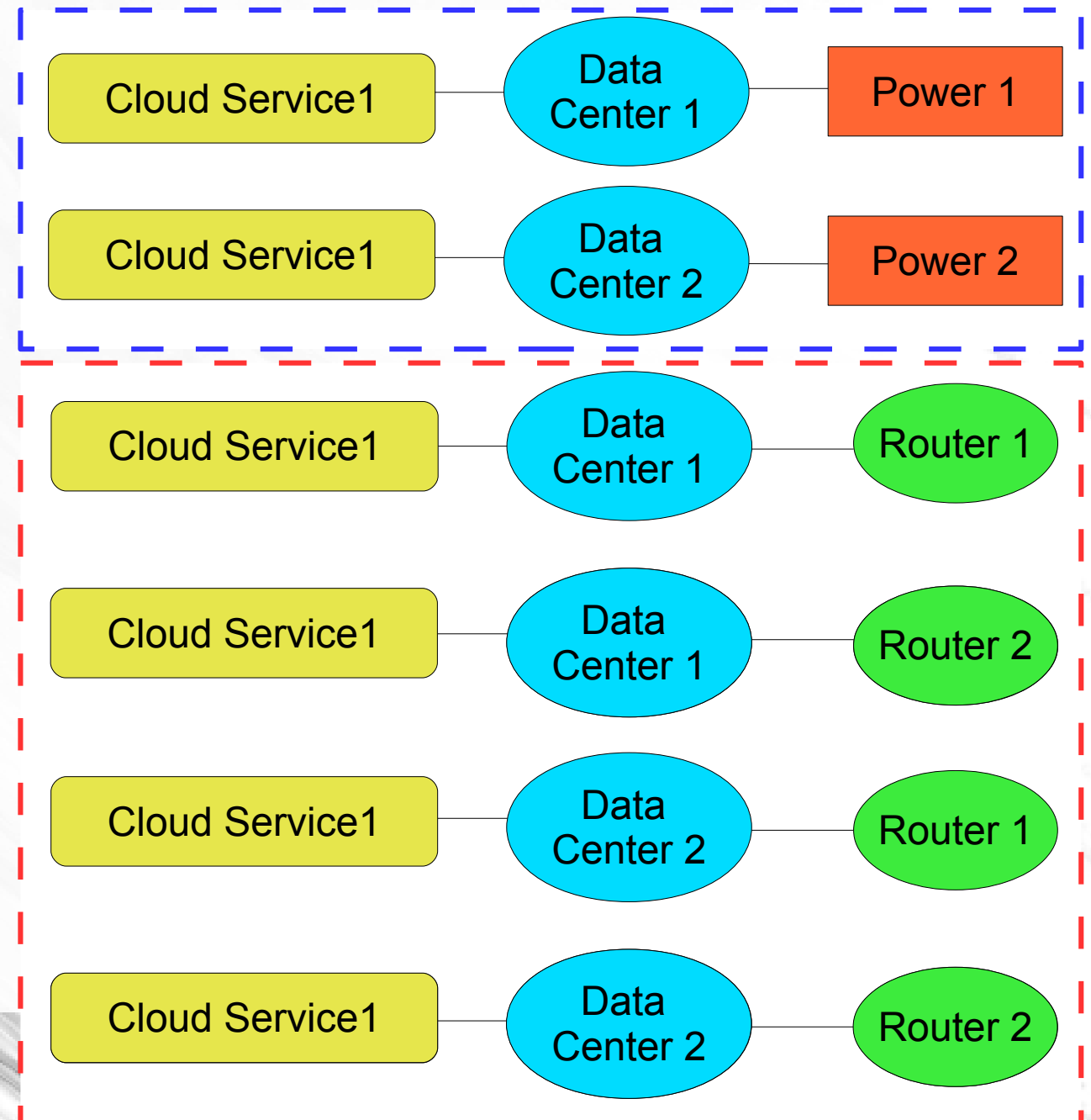
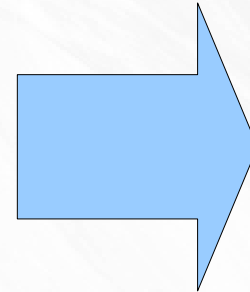
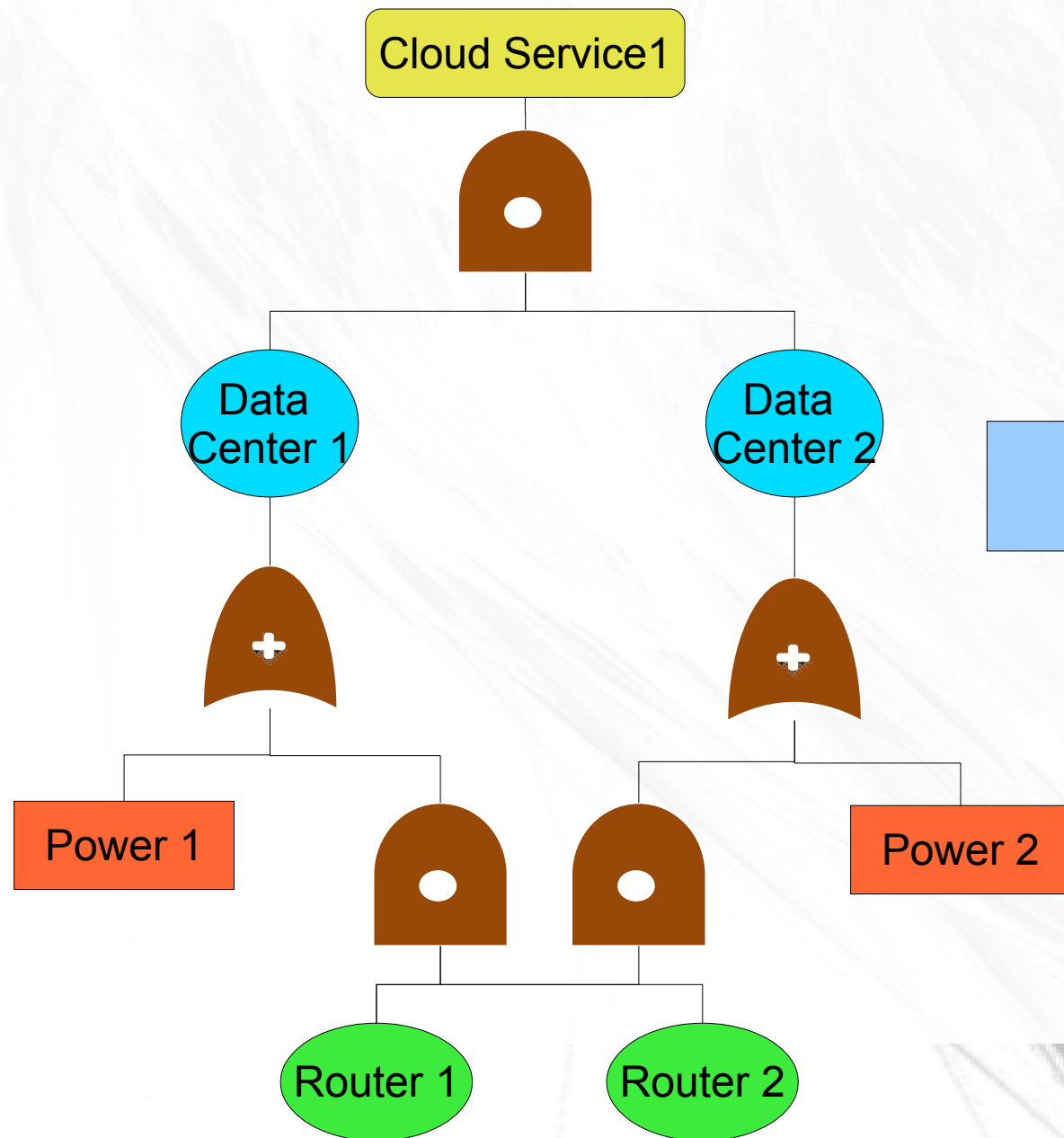
- Challenge
  - SMPC cannot readily handle conditionals, which are necessary in traditional ways of processing Fault Trees
- Solution
  - Rewrite the fault tree as **topology paths form with types**
  - Eliminates use of conditionals

# Topology Paths with Types

- Extract all paths through dependency DAG
  - root node → intermediate nodes → leaf node
  - Unpacks the DAG for "circuit" processing
  - Can be exponentially larger than DAG in worst case :(
- Types of topology paths
  - The SSU builds a **disjunction of conjunctions of disjunctions** data structure by assigning each path a type



# Topology Paths with Types: Example



# Local Execution Protocol

- Generate fault tree for components within macro-components
- Compute the failure sets of each macro-component

# Generate input for the SMPC

- SSUs pad the fault tree in order to avoid leaking structural information such as the size of the cloud infrastructure
  - Add dummy nodes with zero ID into each topology path
  - Add zero paths into the fault tree with randomly assigned types
  - Zero ID nodes do not affect the result

# Identify common dependencies

- SSUs and P-SRA Host cooperate to identify common dependency
  - doing multiple (privacy-preserving) set intersections, followed by one (privacy-preserving) union
- Strict security requires doing it without conditional statements
  - Transfer conditional statements into arithmetic computation

# Identify common dependencies

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**Algorithm 1: Common-Dependency Finder**

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**Input:** Fault tree  $T_i, i = 1$  to  $N$ , where  $N$  is the number of participating cloud-service providers

**Output:** Common Dependency

```
1 foreach  $T_I$  and  $T_J, I \neq J$  do
2   private mask.clear();
3   foreach  $node_i \in T_I$  and  $node_j \in T_J$  do
4     private mask[i][j] = ( $node_i.ID == node_j.ID$ );
5   private CommonDep.clear();
6   foreach  $node_i \in T_i$  and  $node_j \in T_j$  do
7     private CommonDep[i] =
8       private mask[i][j]  $\times$   $node_j.ID + CommonDep[i]$ ;
8   private CommonDependent.append(CommonDep);
9 return private CommonDependent;
```

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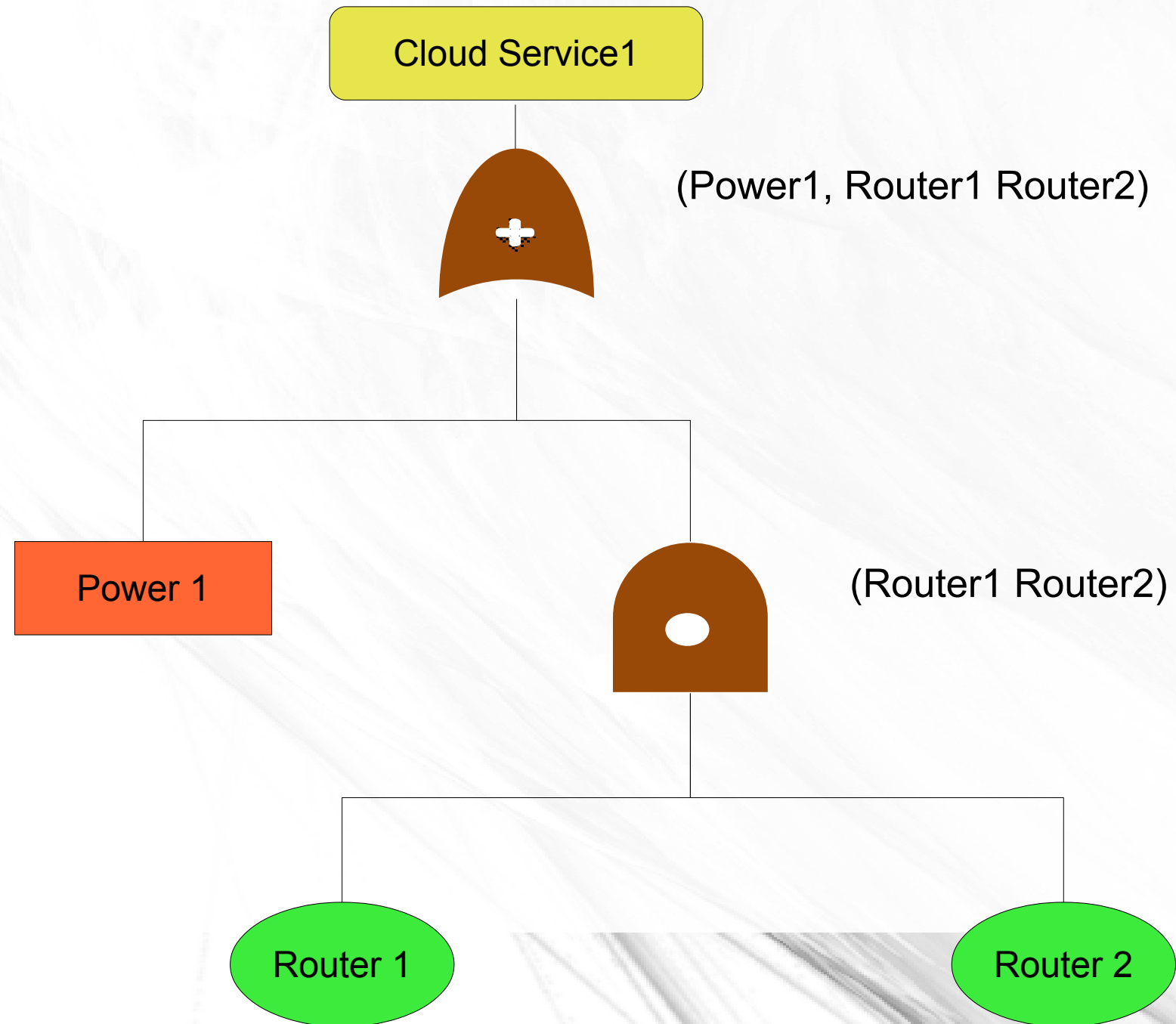
# Privacy Preserving Fault Tree Analysis: Calculate failure sets

- Minimal FSes algorithm
  - Find minimal FSes
  - Exponential complexity
- Heuristic **failure-sampling algorithm**
  - Faster
  - Not necessarily the minimal FSes

# Minimal FSes Algorithm

- The algorithm traverses the Fault Tree
- Basic events generate FSes containing only themselves, while non-basic events produce FSes based on the FSes of their child events and their gate types.
- For an OR gate, any FS of one of the input nodes is an FS of the OR.
- For an AND gate, take cartesian product of the sets of FSes of the input nodes then combine each element of the cartesian product into a single FS by taking a union.

# Minimal FSes Algorithm: Example





# Minimal FSes Algorithm

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**Algorithm 2:** Minimal-FS algorithm

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**Input:** Global Fault tree  $T$

**Output:** MinimalFS

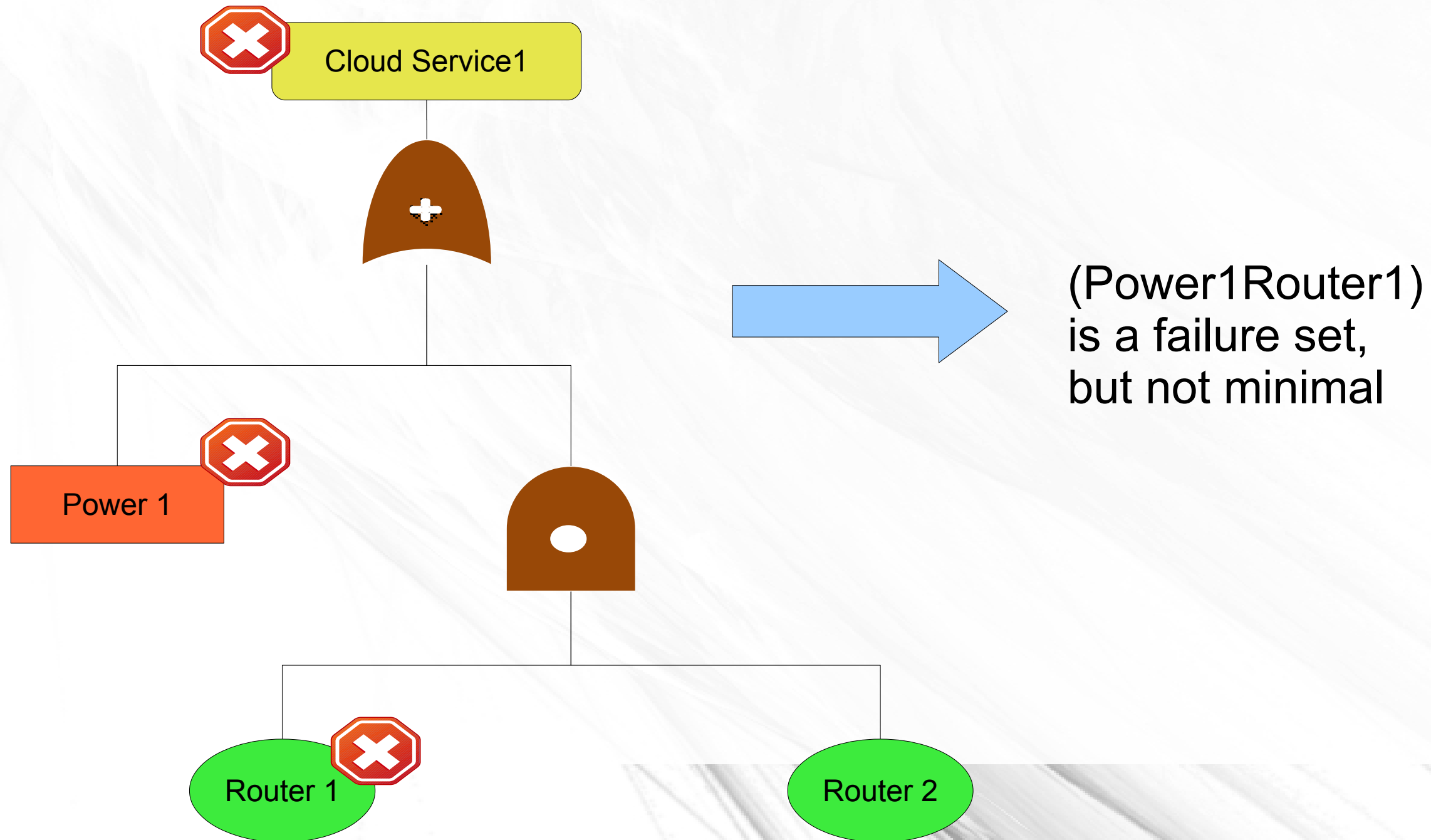
```
1 foreach private path $i$   $\in T$  do
2   foreach private node $j$   $\in$  private path $i$  do
3     private path $i$ .FS.append(node $j$ );
4     /* each path corresponds to an OR gate with
5        input as the nodes along the path */
6
7 foreach AndGate $i$   $\in T$  do
8   AndGate $i$ .FS.clear();
9   foreach path $j$   $\in$  AndGate $i$  do
10    AndGate $i$ .FS  $\leftarrow$  AndGate $i$ .FS  $\times$  path $j$ .FS;
11    /* process the AndGate for each type of
12       topology paths */
13    /* FS of AndGate $i$  is the Cartesian Product of
14       AndGate $i$ .FS and path $j$ .FS. */
15
16 private minimalFS.clear();
17 foreach AndGate $i$   $\in T$  do
18   minimalFS.append(AndGate $i$ .FS);
19   /* process the OR gate connecting to the And
20      Gates */
21   /* reduce redundant items in minimumFS and assign the
22      result to minimalFS, and then simplify minimalFS.
23      */
24
25 minimalFS  $\leftarrow$  reduce_redundancy(minimalFS);
26 minimalFS  $\leftarrow$  simplify(minimalFS);
27 return minimalFS;
```

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# Failure Sampling Algorithm

- Randomly assigns **fail** or **no fail** to the basic events of the Fault Tree
- Compute whether **the top event** fails
- If the top event fails, the failed basic events consist of a FS

# Failure Sampling Algorithm: Example

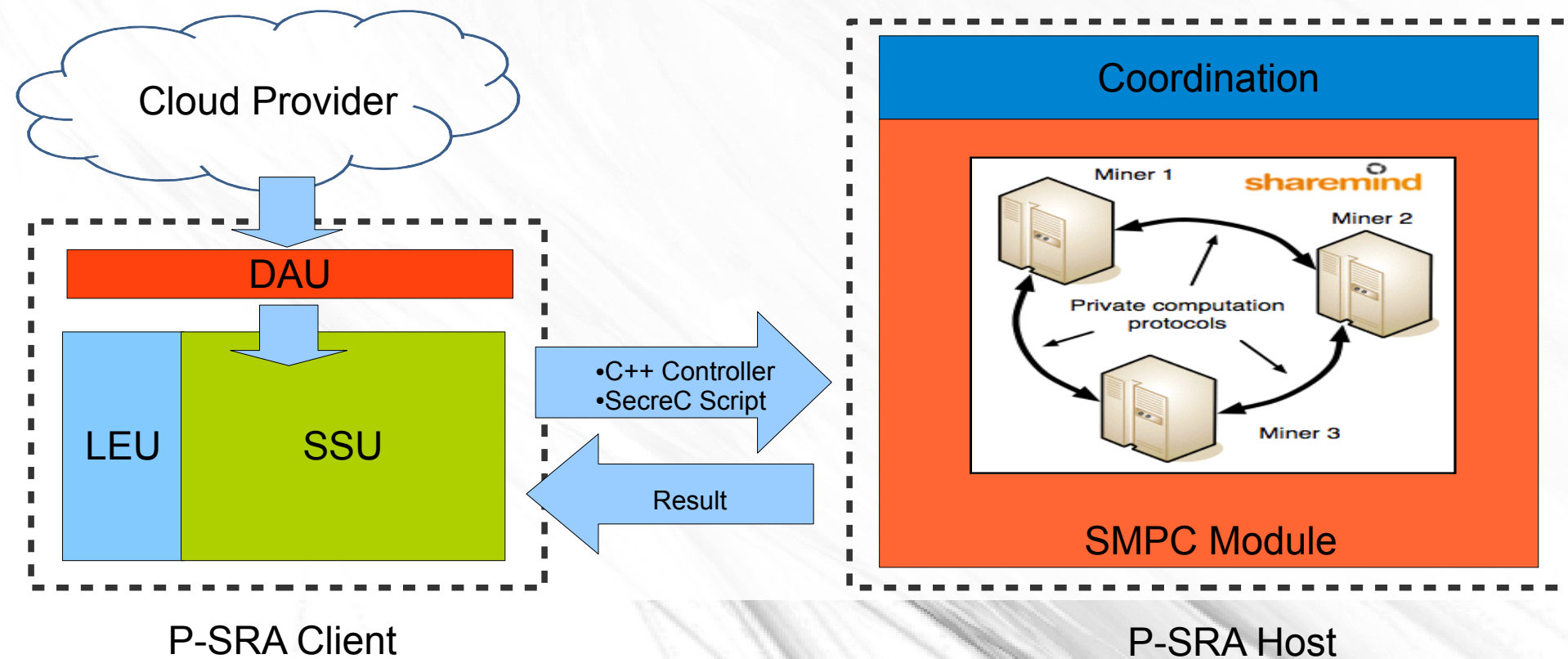


# Privacy-preserving Output Delivery

- Output for Cloud-Service Providers
  - Common dependency
  - Partial failure sets
- Output for Cloud-Service Users
  - Common-dependency ratio
  - Overall failure probabilities of cloud services
  - Top-ranked failure sets

# Implementation

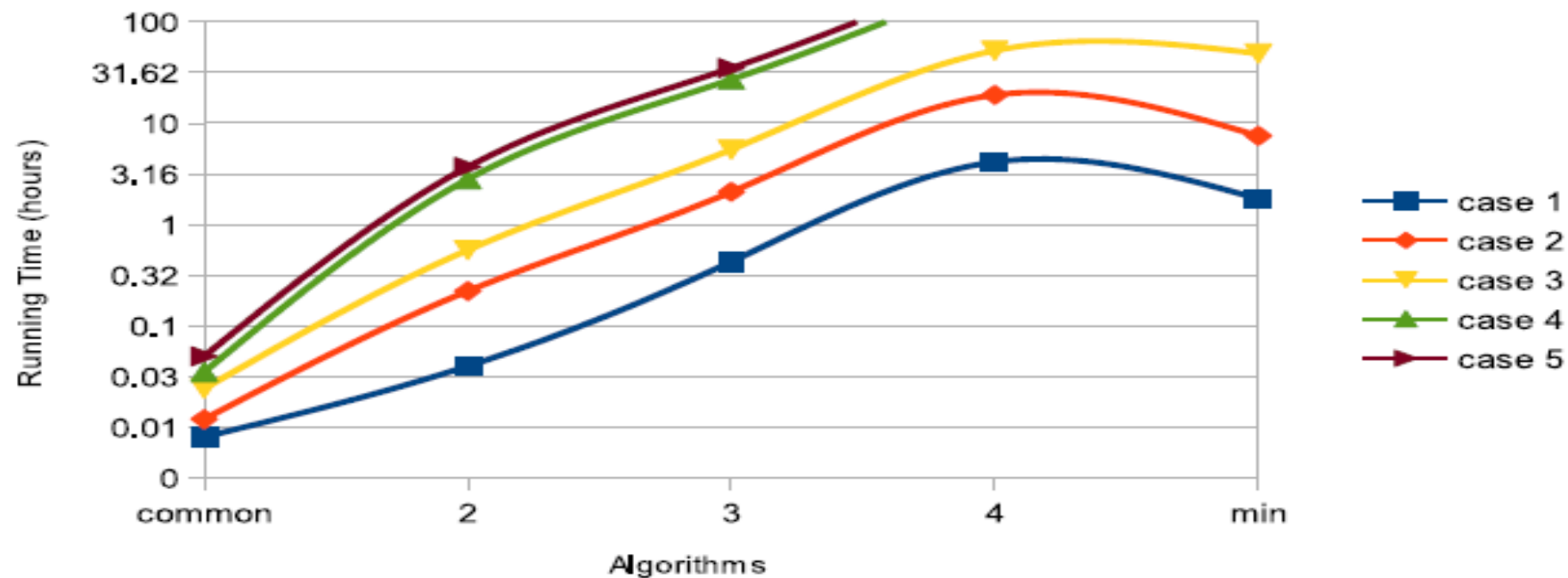
- Sharemind SecreC
  - C-like SMPC programming language
  - Specialized assembly to execute the code



# Simulation: SMPC

	Case 1	Case 2	Case 3	Case 4	Case 5
# of cloud providers	2	2	3	3	2
# of data center	1	3	8	10	3
# of internet router	3	5	10	15	5
# of power stations	1	2	3	5	2
ratio of common dep.	0.8	0.2	0.2	0.2	0.2
ratio of padding	0.0	0.0	0.0	0.0	0.5

Table 1: Configuration of Test Data Sets



# Simulation: Local Execution

Table 2: Performance of the LEU of a P-SRA client

Configuration	Case 1	Case 2	Case 3	Case 4	Case 5
# of switch ports	4	8	16	24	48
# of core routers	4	16	64	144	576
# of agg switches	8	32	128	288	1152
# of ToR switches	8	32	128	288	1152
# of servers	16	128	1024	3456	13824
Total # of components	40	216	1360	4200	16752
Running time (minutes)					
FS round $10^3$	< 0.7	< 0.7	< 0.7	< 0.7	< 0.7
FS round $10^4$	0.7	0.7	1.7	2.3	6.9
FS round $10^5$	0.8	0.9	5.3	28.1	6.9
FS round $10^6$	1.7	4.5	65.0	243.5	462.9
FS round $10^7$	28.3	56.6	512.1	NA	NA
Minimal FS	0.8	14.8	309.7	NA	NA

# Conclusion

- We designed P-SRA, a private, structural-reliability auditor for cloud services based on SMPC, and prototyped it using the Sharemind SecreC platform
- We explored the use of data partitioning and subgraph abstraction SMPC on large graphs, with promising results.
- Our preliminary experiments indicate that P-SRA could be a practical, off-line service, at least for small-scale cloud services or for ones that permit significant subgraph abstraction.



**Thank you**  
**Any Questions?**