



DIAGNOSTICS AND PROGNOSTICS USING TEMPORAL CAUSAL MODELS FOR CYBER PHYSICAL SYSTEMS - A CASE OF SMART ELECTRIC GRID

Ajay D Chhokra¹, Rishabh Jain², Saqib Hasan¹, Abhishek Dubey¹, Srdjan Lukic², Gabor Karsai¹, Nagabhushan Mahadevan¹,

¹Institute for Software Integrated Systems, Vanderbilt University

²Department of Electrical and Computer Engineering, North Carolina State University

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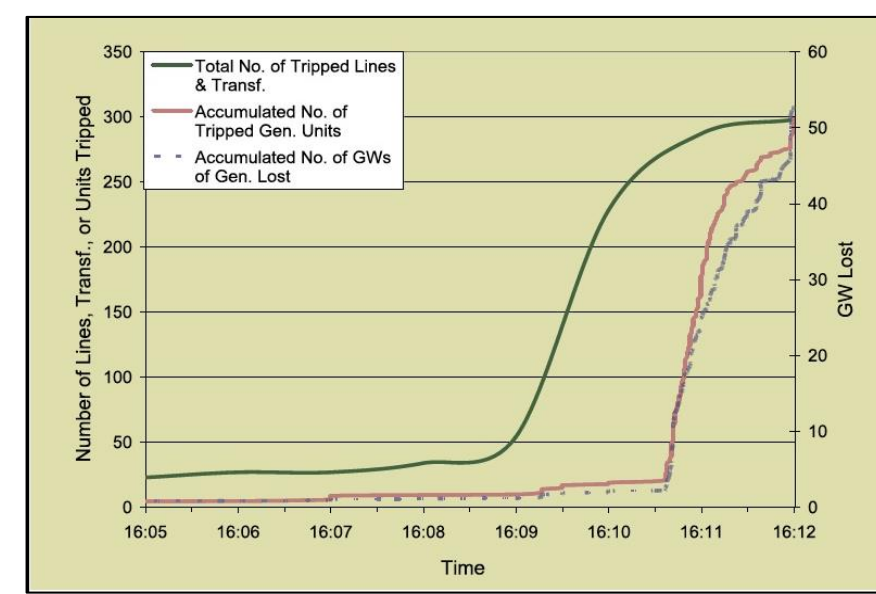
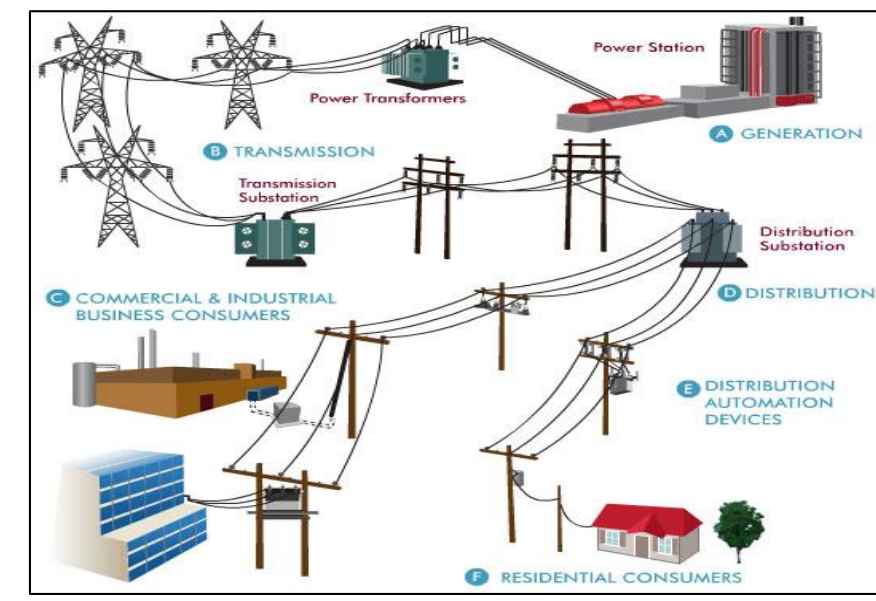
Motivation

- Cyber Physical Systems are integrating computation, networking and physical processes where interactions among discrete and continuous subsystems play an important role.

Fault diagnosis becomes challenging as these interactions cause failure cascades. For instance, in power systems, protection system mis-operations often lead to cascading effects and system wide failures.

- Fault diagnosis and prognosis needs reasoning about these interactions, allowing for control actions that effect the cascades.

- Our approach is based on constructing a system-level discrete event model that captures the causal and temporal relationships between failure modes (causes) and discrepancies (effects) in a system, thus modeling the failure cascades while taking into account propagation constraints imposed by operating modes, protection elements, and timing delays. Another characteristic is the hierarchical reasoning that can use external reasoners or simulators to refine system level fault hypotheses.



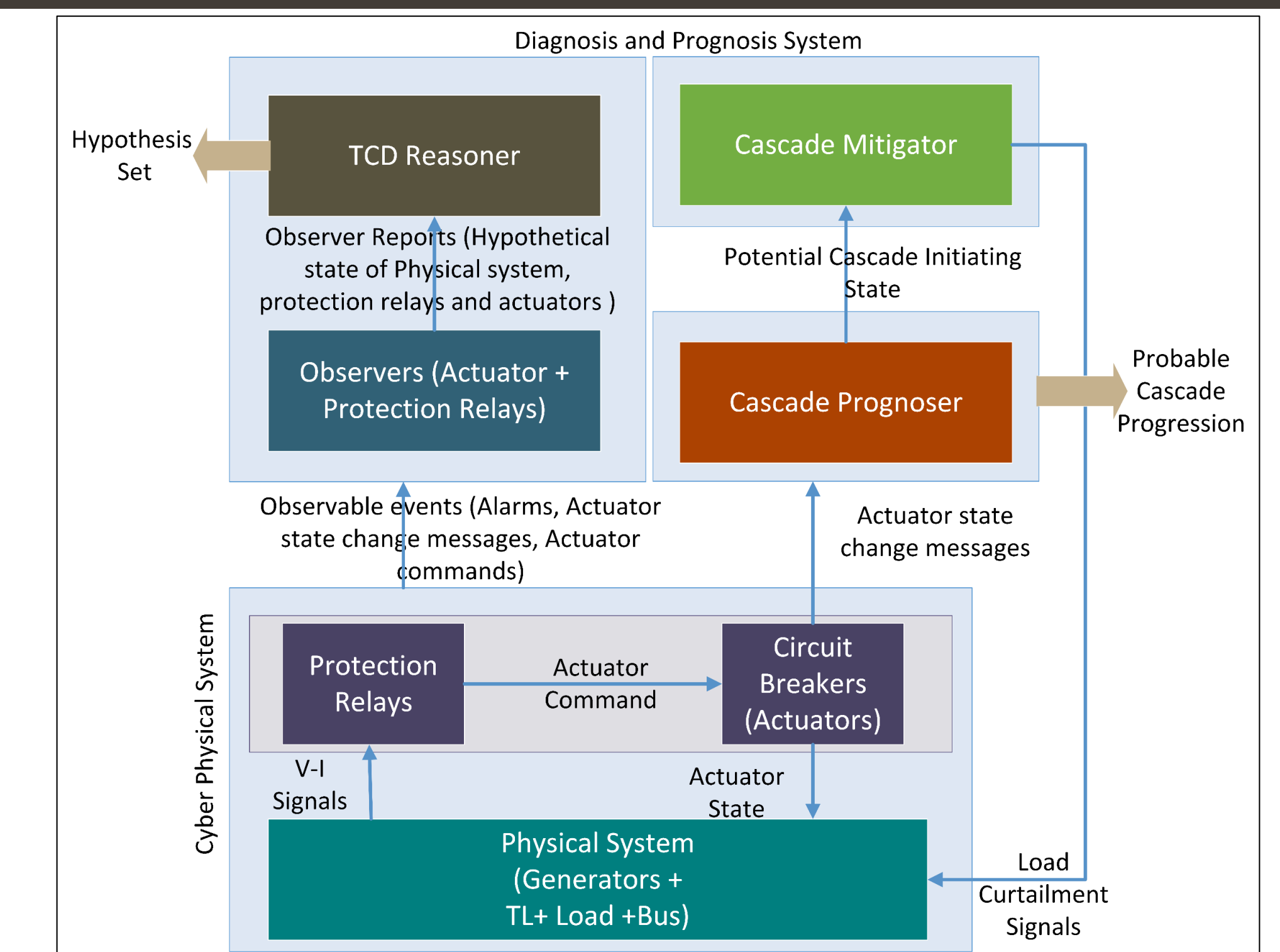
Progression of events in the 2003 blackout.

Project Achievements

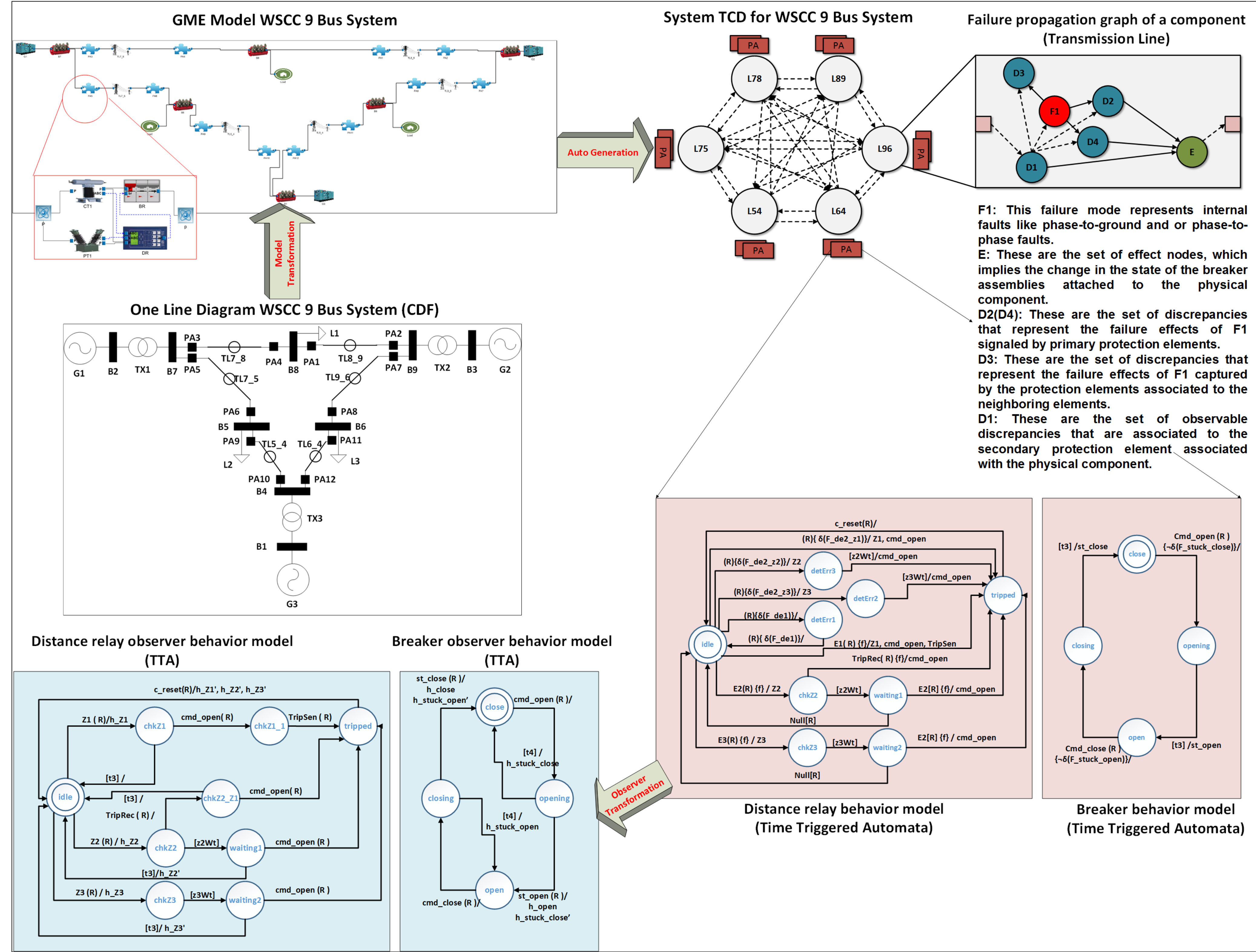
In order to devise a systematic approach to understand and improve the resilience of cyber physical energy systems through automated generation of failure graphs, simulation models, failure cascade sequences and minimal reconfiguration actions, we

- 1) Developed a new modeling formalism, Temporal Causal Diagrams, to capture the interplay between failure progression and component behaviors.
- 2) Extended the Modeling formalism of Temporal Failure Propagation Graphs, to include effect nodes and uncertainty in failure propagation.
- 3) Extended a Time-Triggered Automata (TTA) model to include constraints imposed by faults.
- 4) Developed a domain specific modeling language to represent the component behavior and fault propagation across component interactions in cyber physical energy systems.
- 5) Proposed a novel approach of encoding failure cascade progressions using binary decision diagrams (BDDs)
- 6) Formulated a nonlinear optimization problem for identifying minimalistic reconfiguration strategies to arrest failure cascades.

Architecture



Cascade Diagnosis



Cascade Prognostics

Industry practice is to identify the set of initial outages that can lead to cascading outages and ultimately blackouts.

The set of contingencies $T \in S$ can be encoded using binary decision diagram [1] B_T , where S is the set of all possible states. The progression of the cascade is stored as a set of transitions, $TR \in S \times S$ using BDD B_{TR} .

The state of a system can be defined by a boolean vector, (v_1, v_2, \dots, v_n) where n is the number of branches in the system.

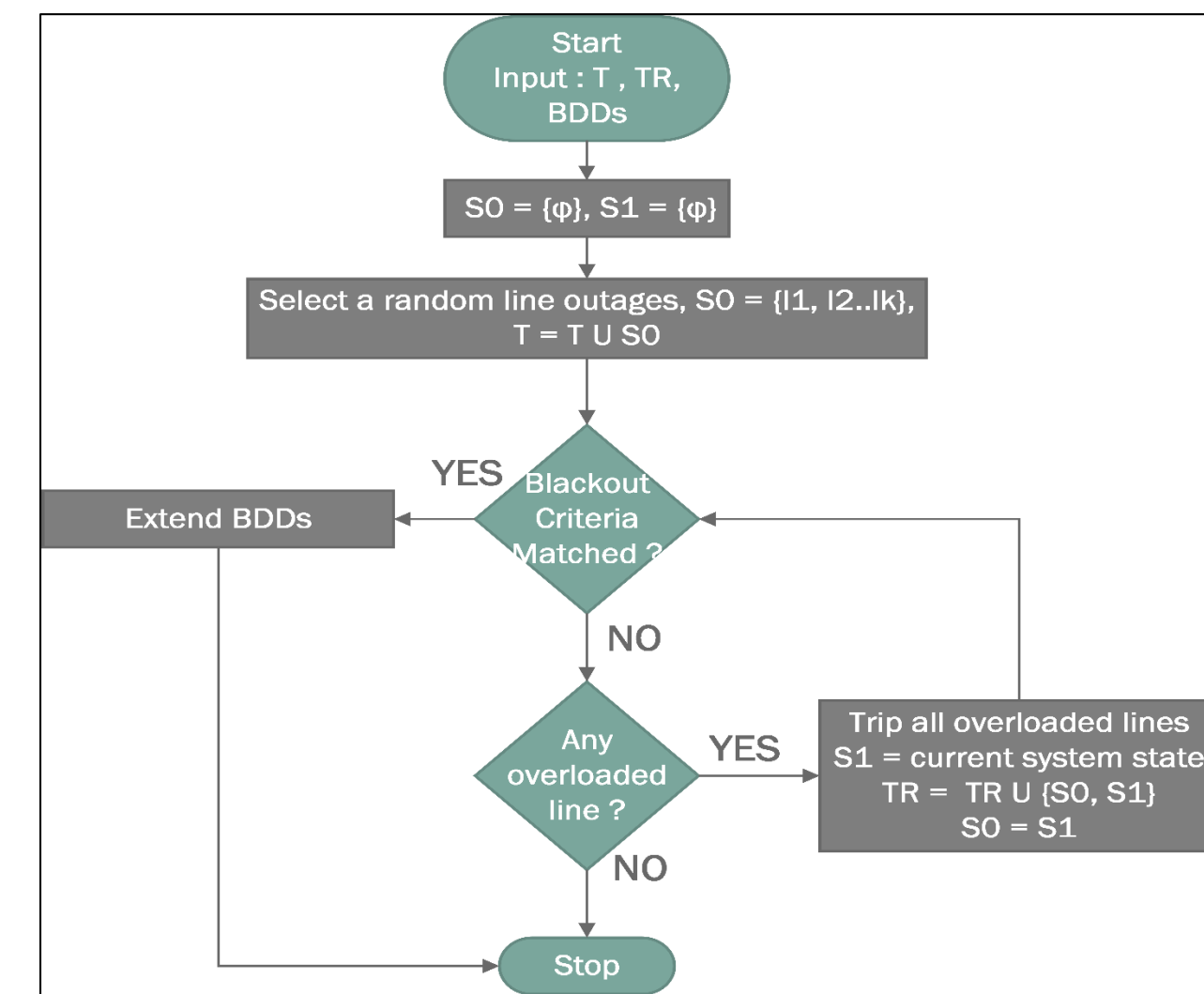
A labeling function L is defined over set of state S , $L(S): S \rightarrow P(\text{Branch})$, where Branch is the set of branch labels. The boolean encoding of state s is given as (v_1, v_2, \dots, v_n) where $v_i = 1$ if $Branch_i \in L(s)$.

The characteristic function $f_T: \{0,1\}^n \rightarrow \{0,1\}$ for set membership can be defined as $(l_{11} \cdot l_{12} \dots l_{1n}) + (l_{21} \cdot l_{22} \dots l_{2n}) + \dots + (l_{k1} \cdot l_{k2} \dots l_{kn})$ where the Boolean function $(l_{j1} \cdot l_{j2} \dots l_{jn})$ represents the state s_j .

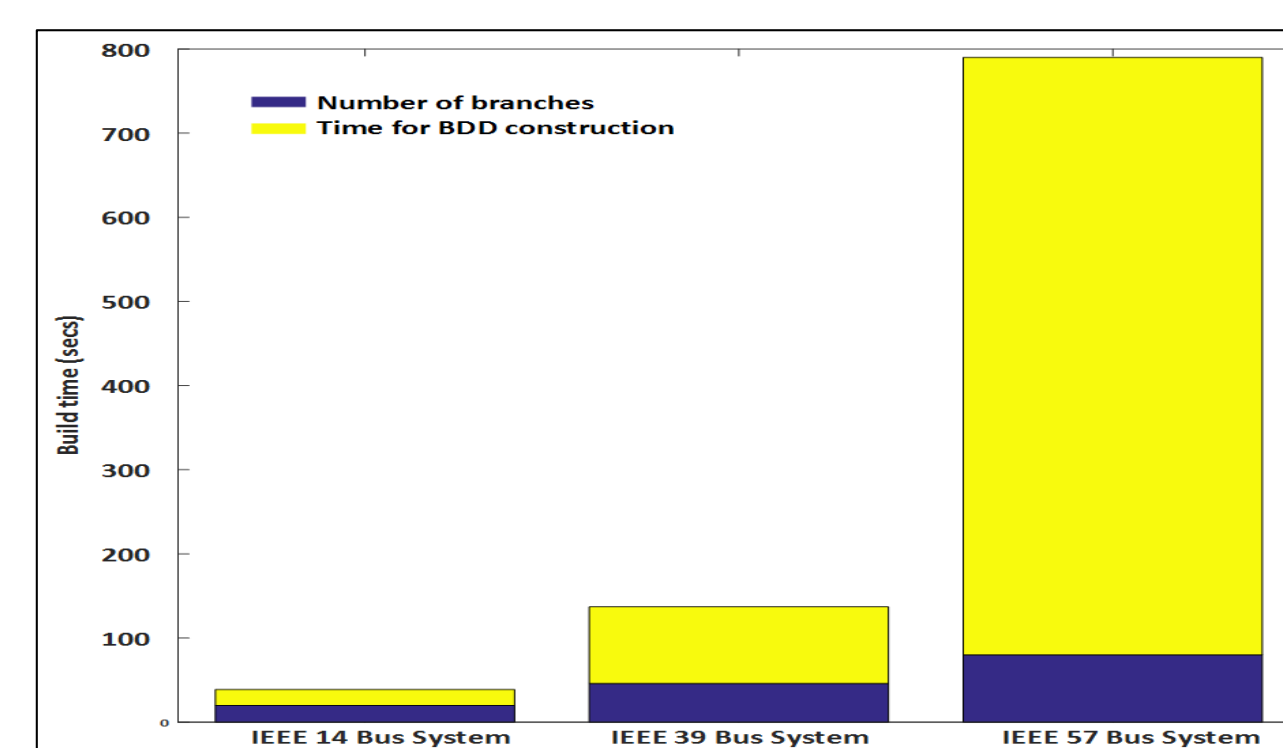
Similarly, the characteristic function can be calculated as the boolean product of two states.

[1] E. M. Clarke, O. Grumberg, and D. Peled, Model checking. MIT press, 1999

Systematic Generation of BDDs

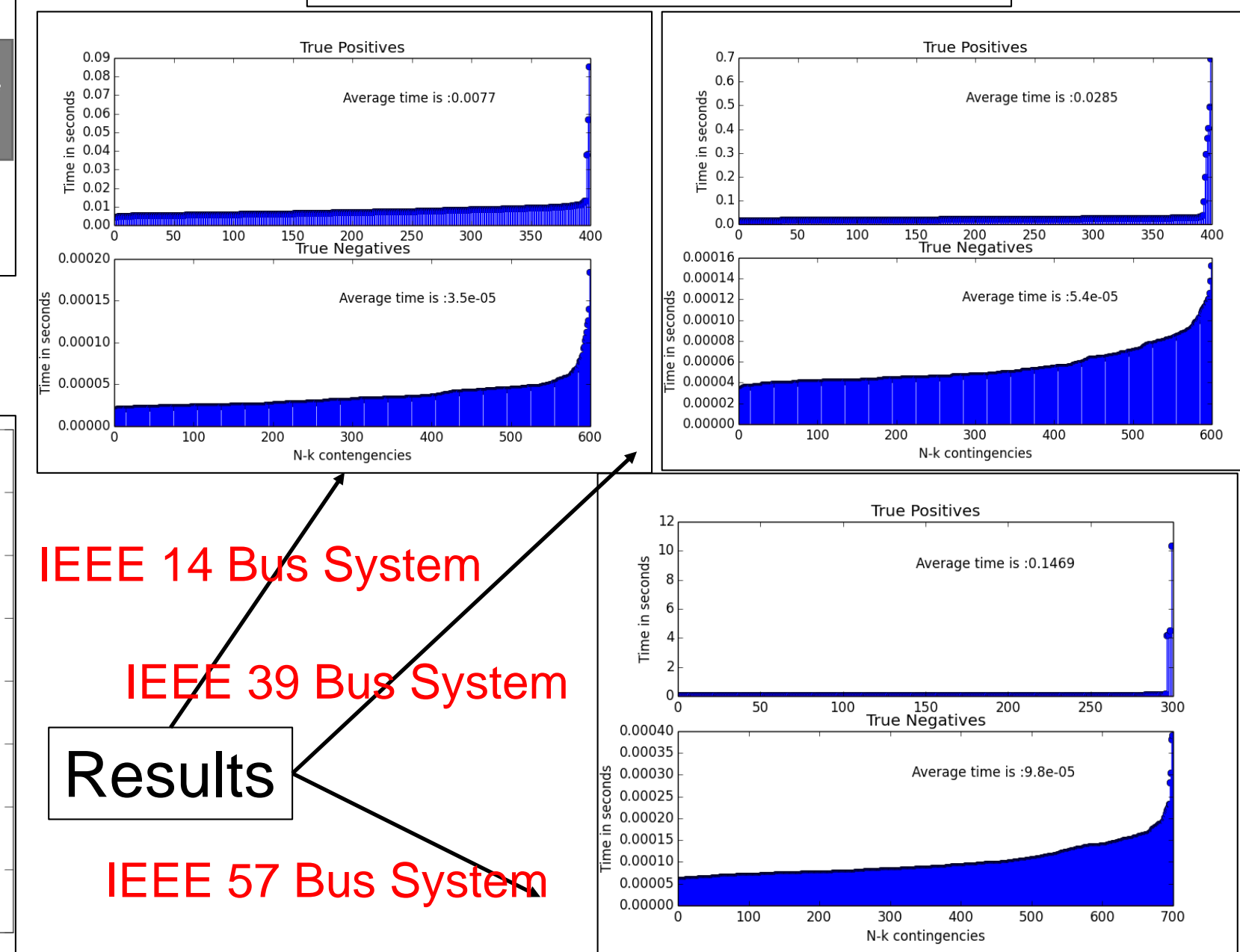


Time taken by building BDDs (exponential in the size of the system)



Cascade prognostics algorithm

Input: $S_0 = (v_1, v_2, \dots, v_n)$; B_T ; B_{TR}
 Output: $S_{reachable} = \phi$
 if Evaluate(B_T, S_0) == True then
 $i = 0$
 $S_{reachable} = \phi$
 while $S_i \neq \phi$ do
 $S_{reachable} = S_{reachable} \cup S_i$
 $S_{i+1} = \text{Image}(B_{TR}, S_i) \setminus S_{reachable}$
 $i = i+1$
 end while
 return $S_{reachable}$



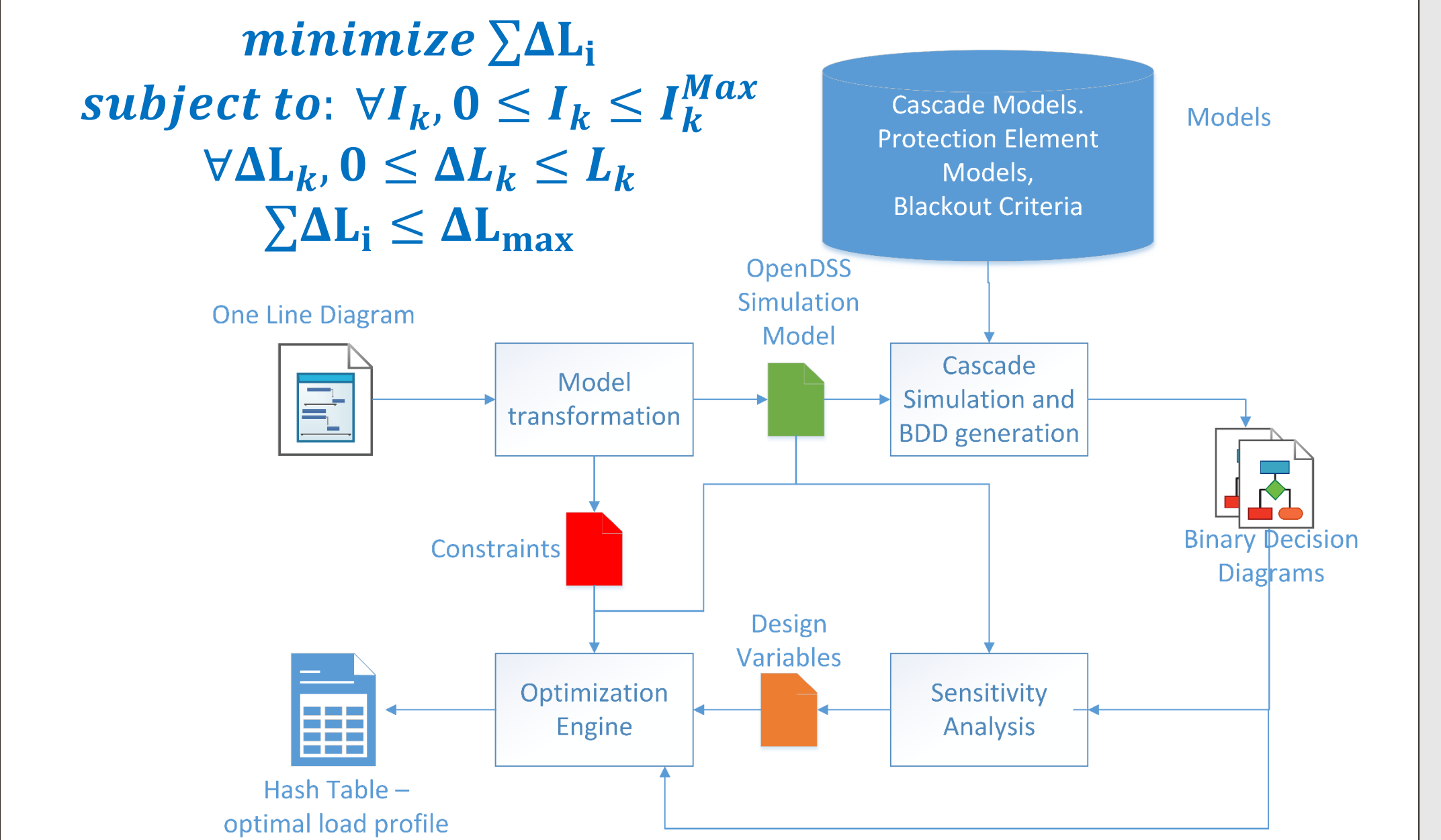
IEEE 14 Bus System
 IEEE 39 Bus System
 IEEE 57 Bus System
 Results

Cascade Mitigation

- In order to prevent cascades from spreading load(s) have to be curtailed such that secondary and tertiary overloads disappear.
- We use a component-based approach by using OpenMDAO as the optimization framework that acts an orchestrator for finding optimal values. OpenDSS is used for solving the AC load flow for different load profiles.

Optimization problem

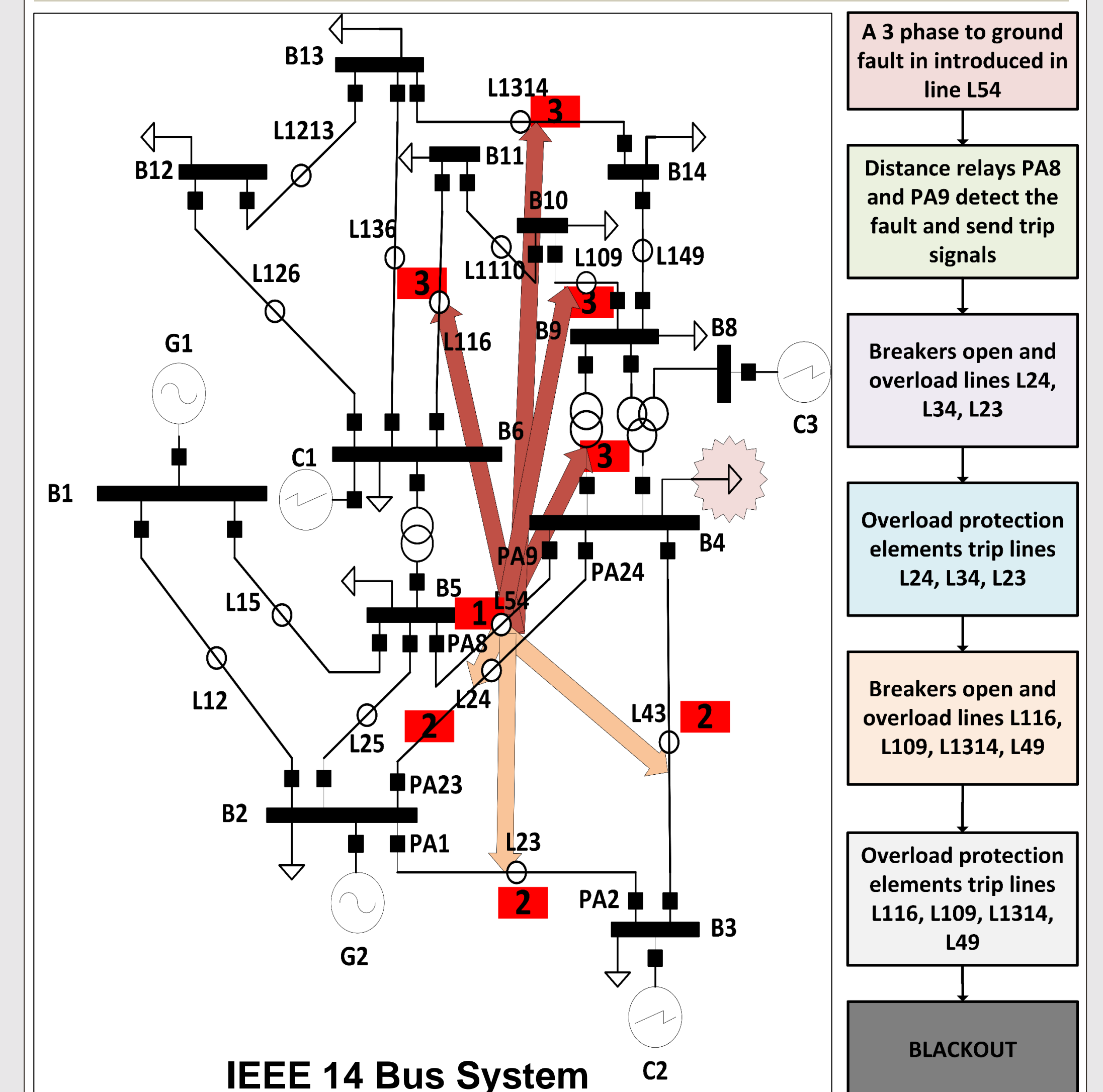
$L = \{L_1, L_2, \dots, L_n\}$ is the vector of n loads where $n \in \mathbb{Z}$.
 $\Delta L = \{\Delta L_1, \Delta L_2, \dots, \Delta L_n\}$ is a vector of n independent variables where ΔL_i represents the change in load L_i .
 $I = \{I_1, I_2, \dots, I_m\}$ is a vector of magnitude of currents flowing through all m branches of the system. I_k^{max} is the maximum current flowing through branch k .
 ΔL_{max} is the maximum load that can be shed from the system.



End-to-end toolchain for Cascade prognostics and Mitigation

Case Study

Validation of the result of the cascade mitigation calculations



System Evolution with mitigation actions applied

