

Causal Model Extraction from Attack Trees to Attribute Malicious Insider Attacks

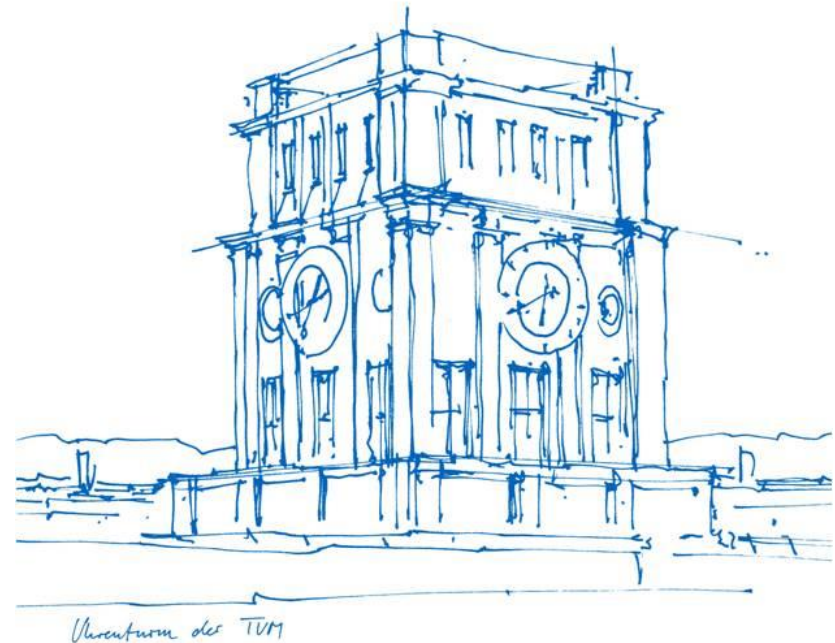
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Introduction

“Hide it or lose it”!

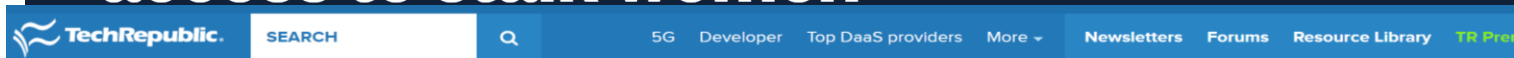
What Tesla's Spygate Teaches Us About Insider Threats



Tom Kemp Forbes Councils Member
Forbes Technology Council COUNCIL POST | Paid Program
Innovation

<https://www.forbes.com/sites/forbestechcouncil/2018/07/19/what-teslas-spygate-teaches-us-about-insider-threats/#3ced1e735afe>

Facebook fires engineer who allegedly used access to stalk women



60% of companies experienced insider attacks in the last year

<https://www.techrepublic.com/article/60-of-companies-experienced-insider-attacks-in-the-last-year/>

Introduction

- Mostly non malicious
 - Accountability
 - Attack attribution a deterrent measure
 - Assigning blame
- Accountable system can answer questions regarding the cause of some event
 - System monitoring
 - Model-based causality analysis
- In this paper, we propose
 - A methodology to automatically create causal models in the context of insiders from attack trees
 - An open-source tool (ATCM) that implements the approach
 - An evaluation of the efficiency, the validity of the approach, and the electiveness of the model.



BACKGROUND

A Counterfactual Cause is..

“...Or, in other words, where, if the first object had not been, the second never had existed “ (Hume 1748 sec. VII).

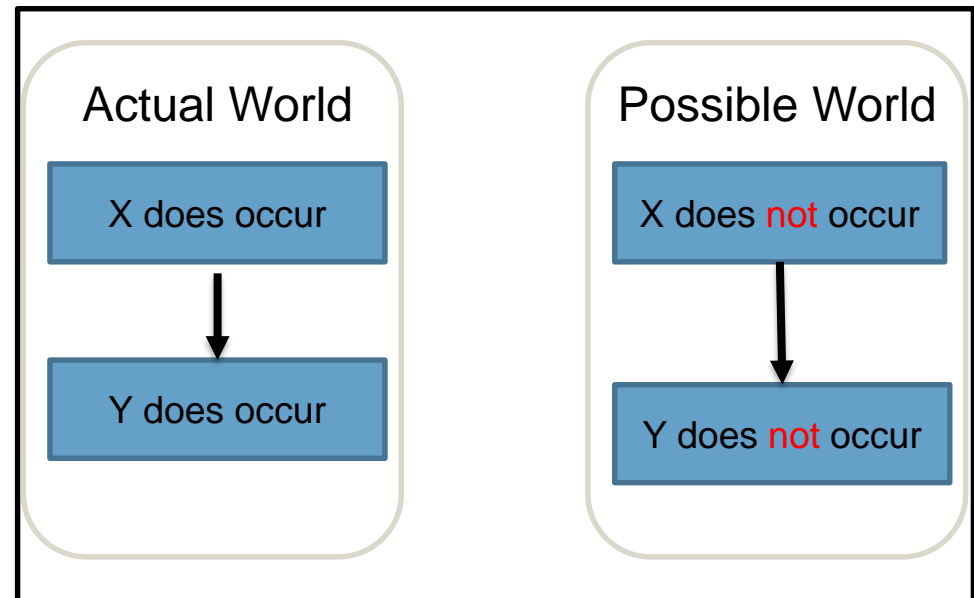


David Hume

Lewis's Definition of cause:

“X has caused Y” if “Y would not have occurred if it were not for X ”

(Lewis 1986)

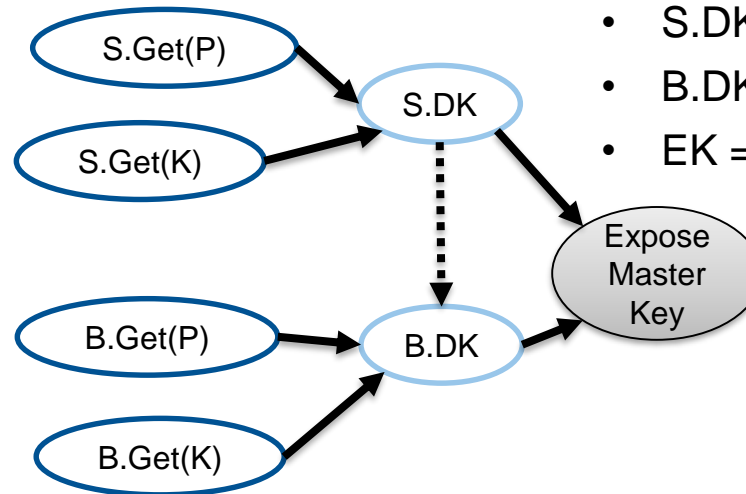
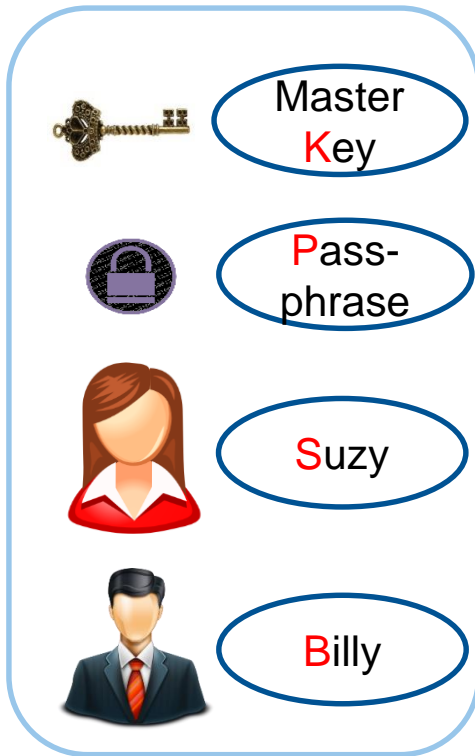


Halpern and Pearl definition of Actual Causality

- Causal models [Pearl 1996]
 - Structural equations represent mechanisms of the world
 - Variables represent properties of the world
 - Interventions

- Causal Model: $M=(U, V, R, \mathbf{F})$ [Halpern and Pearl 2000]
 - **U**: Set of exogenous variables
 - **V**: Set of endogenous variables
 - **R**: Associates with each variable a set of possible values
 - **F**: Associates a function F_X with each $X \in V$
 - Visualization via Causal Networks

Example



Context

- $S.Get(P)/B.Get(P) = T/T$
- $S.Get(K)/B.Get(K) = T/T$
- $S.DK = T \text{ AND } T = T$
- $B.DK = T \text{ AND } T \text{ AND } F = F$
- $EK = T \text{ OR } F = T$

- $S.Get(P)/B.Get(P)$ = read the passphrase file
- $S.Get(K)/B.Get(K)$ = Suzy/Billy queried the key
- $S.DK = S.Get(P) \text{ AND } S.Get(K)$ (Suzy decrypts the key)
- $B.DK = B.Get(P) \text{ AND } B.Get(K) \text{ AND } !S.DK$ (Billy decrypts)
- $EK = S.DK \text{ OR } B.DK$

Why HP?

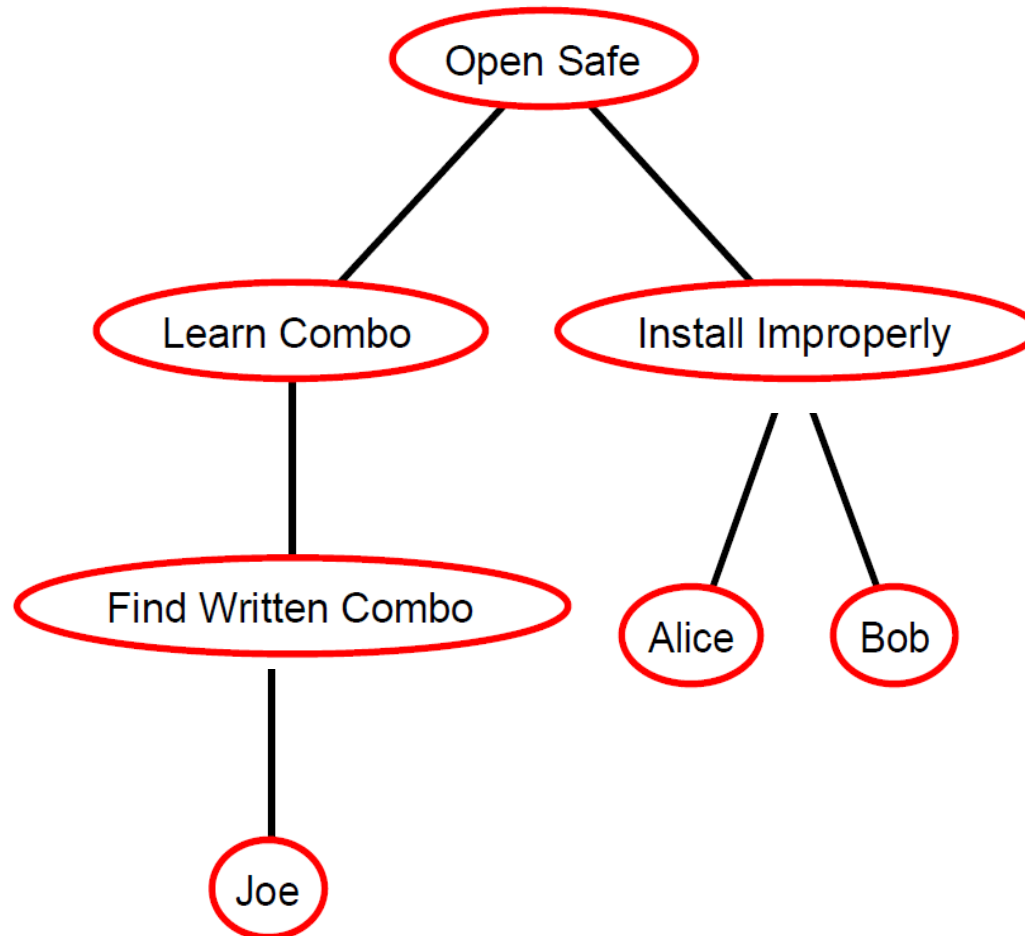
- Preemption
- Irrelevance
- Conjunction and disjunction of events
- Non-occurrence of events

- "...no right model..." [Halpern 2016]
 - Considerable influence of the model on the result
 - Domain specific

Sources for models: Attack Trees

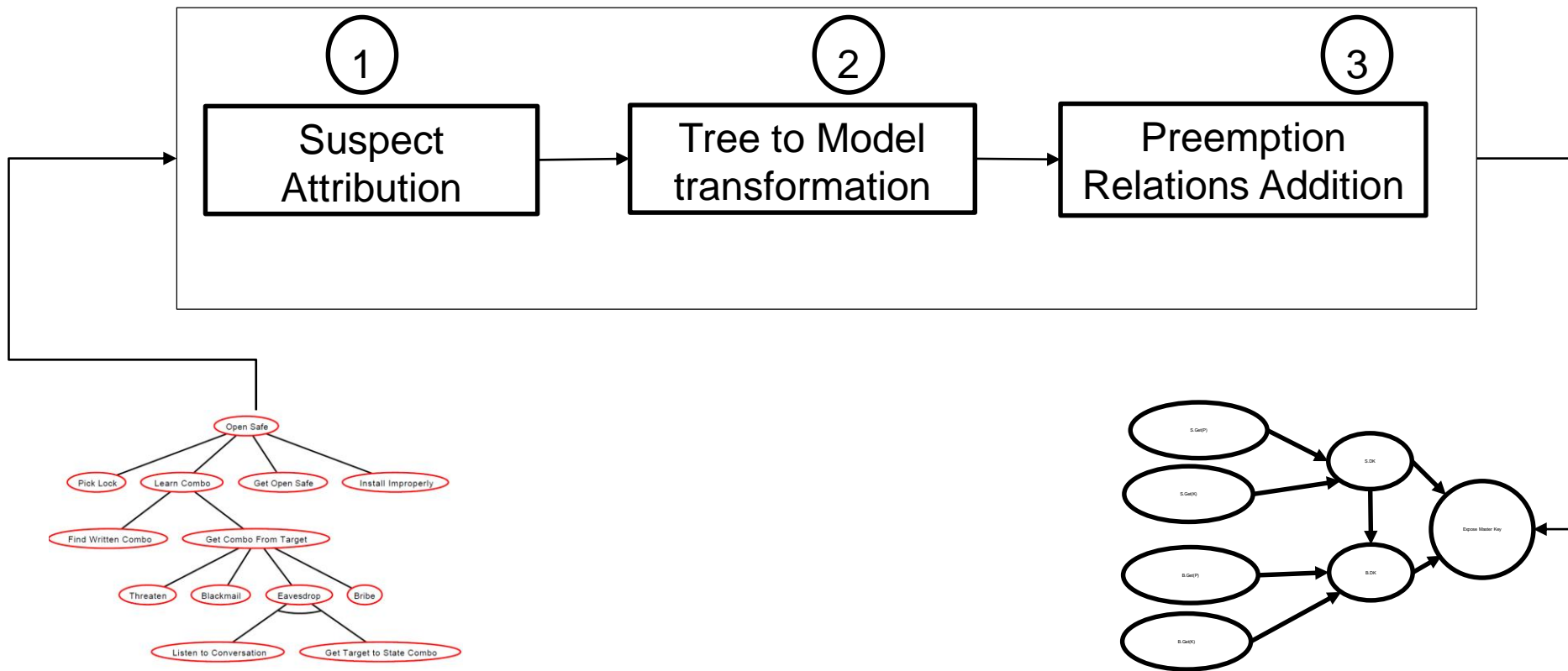
- Describe potential threats and the steps necessary to successfully perform
 - Root node contains the ultimate goal of an attack tree
 - Sub-nodes describe activities that are necessary to accomplish the respective parent activity/goal
 - Formal
 - Graphical

Attack Trees** \neq Causal Models



**All the attack trees in this presentation are drawn using ADTool

Methodology for Causal Modeling



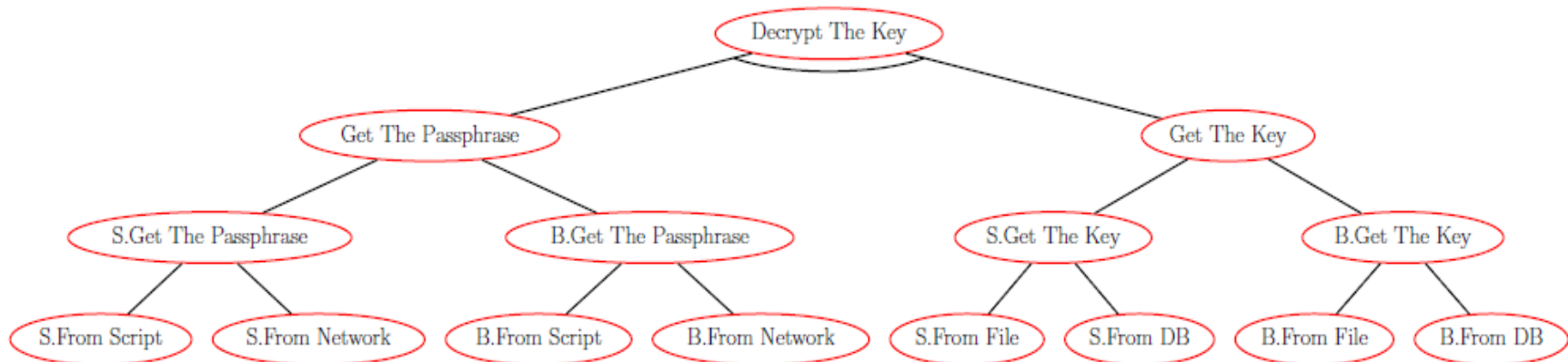
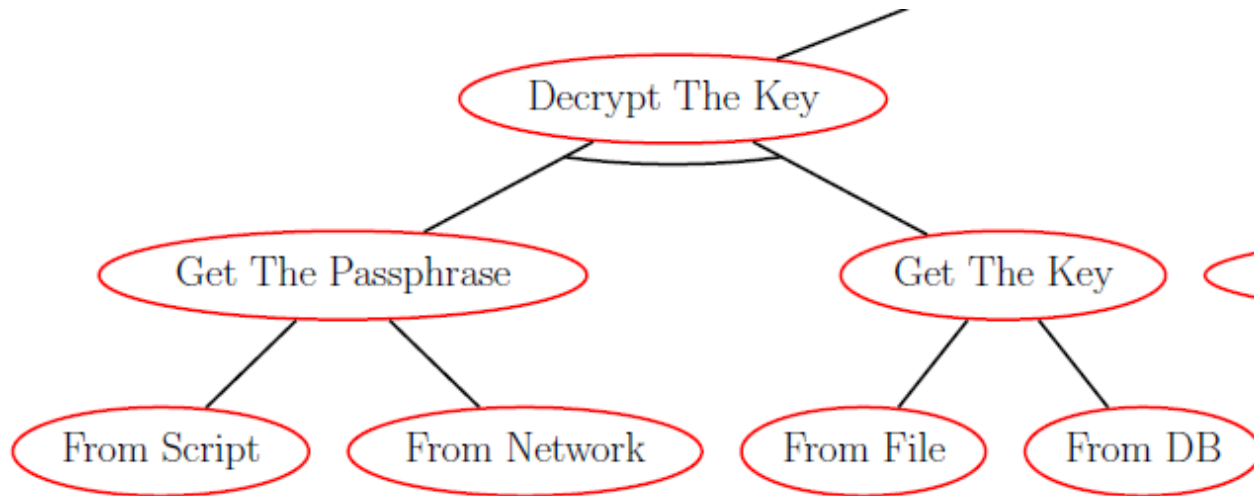
Suspect Attribution

- Automatically adding instance of roles to a tree
 - Duplicating parts of the tree followed by allotting the new parts to one suspect

Definition 4. *A subtree $\mathcal{B} = (\mathcal{N}, \rightarrow, n_0, [[n]])$ is attributed with suspects $\{s_1, s_2, \dots, s_l\}$ by: 1) Creating a set (size l) of \mathcal{B} duplicates, denoted $\{\mathcal{B}_1, \mathcal{B}_2 \dots \mathcal{B}_l\}$. A duplicate \mathcal{B}_i contains the nodes of \mathcal{B} with every node renamed with i suffix. 2) Constructing a new tree \mathcal{AB} with root n_0 from \mathcal{B} , then adding the disconnected $\{\mathcal{B}_1, \mathcal{B}_2 \dots \mathcal{B}_l\}$, and connecting their root nodes using an OR function with n_0 .*

- Where do we attribute
 - Trees that model different attack vectors

Attribution Level



Adding Roles to Attack trees

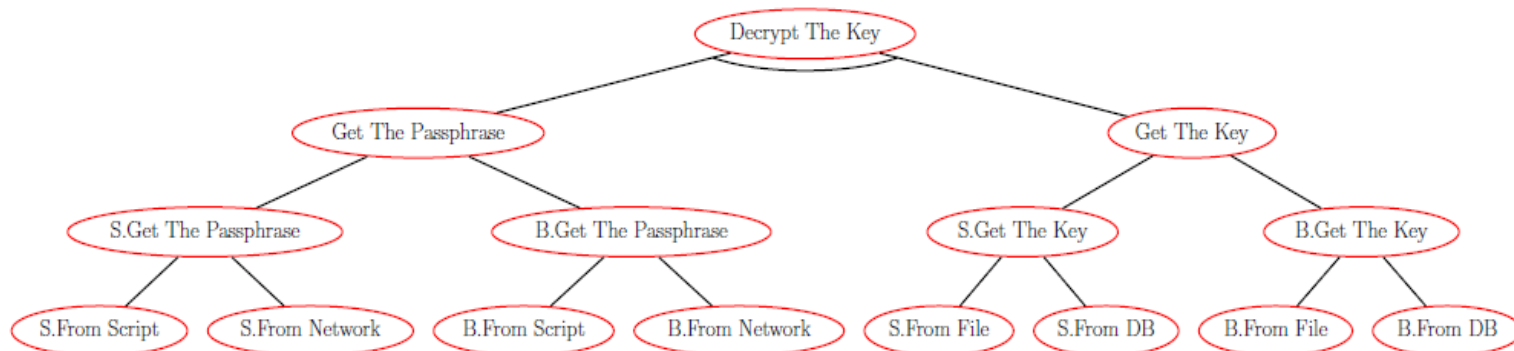
- Depends on the structure and the and the semantics of the branch
- Unfolding after the last AND gate allows considering any possibility of colluding attacks, in some cases it may be unnecessary.

Tree Transformation

Definition 6. *Attack Tree To Causal Model*

$AT = (\mathcal{N}, \rightarrow, n_0, [[n]])$ is mapped to a $M = (\mathcal{U}, \mathcal{V}, \mathcal{R}, \mathcal{F})$ i.e. $AT \mapsto M$ as follows

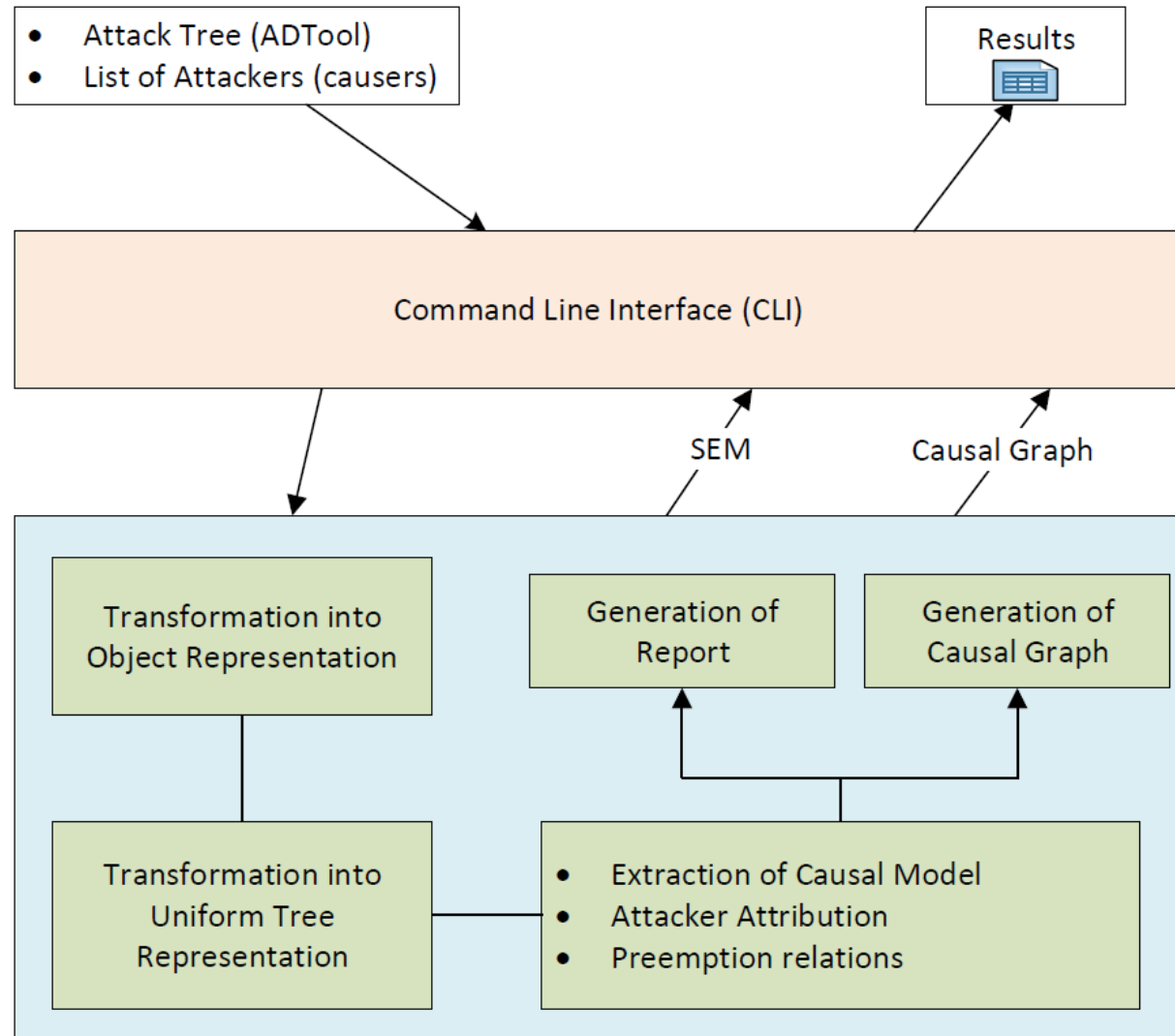
- $\mathcal{U} = E(AT)$, where $E(AT)$ returns the leaf nodes of a tree AT
- $\mathcal{V} = \mathcal{N} \setminus E(AT)$, where \setminus is the difference between two sets.
- $\mathcal{R} = \{0, 1\}$.
- \mathcal{F} associates with each $X \in \mathcal{V}$ a propositional formula $F_X = [[X]]$, which corresponds to the semantical formula from the AT



Adding Preemption Relations

- Preemption relations relate variables about same event for different suspects
 - They represent disparity between suspects
 - Hard to model from different facts
 - Suzy's privileges in a system
 - Billy's criminal record
- For automation relate them to metrics of insiders' risk assessment.
 - Suspiciousness metric (SM): aggregates ability to perform an event or willingness attack
 - Calculation is incident-specific: it can be a simple reflection of privileges in the system; it can be a sum of weighted factors
- Location : among attribution variables one level after the attribution level
 - two variables with an edge from the more suspicious suspect (higher SM) to the less suspicious suspect (in case of equal values the edge is not added).
- Semantically, the preemption relation is represented by a not clause (!X) added to the less suspicious (i.e. smaller value) suspect about the higher suspicious suspect

Tool Support



Evaluation

Class	Use Case	Nodes	# Potential Attackers
HP	HP ₁	3	2
	HP ₂	2	2
Insider (Industry)	Steal Master Key	12	{2, 8}
Insider (Literature)	BecomeRootUser ₁	8	{2, 8}
	BecomeRootUser ₂	11	{2, 8}
Artificially Generated	Artificial ₁	255	{2, 8}
	Artificial ₂	1017	{2, 8}
	Artificial ₃	3057	{2, 8}

- Efficiency of the process: model expansion and automation
- Validity of the model
- Effectiveness of the model:
 - Threat analysis → Attack Trees → Implement the attacks → Check the logs
 - Formulated queries

Conclusions

- Problem: insider threat and preventive measures
- Solution: accountability through supporting causal reasoning
 - A methodology that automatically constructs HP causal models from attack trees
 - Suspect attribution while allowing colluding.
 - Preemption relations.
 - Efficiency of the process, validity and effectiveness of the model
- Future Work
 - Consider more elements of threat models
 - Examples: notions of attack-defense trees, SAND attack trees

Thanks For Your Attention!

HP Definition (Informal)

A set of events $\vec{X} = \vec{x}$ is an actual cause of φ given a model if the following three conditions hold [Halpern 2015]:

AC1. *both* the cause and the effect actually happened

AC2. Changing the original values of \vec{X} to a different setting \vec{x}' while keeping a possibly empty set (\vec{W}) of the remaining variables at their original value, φ does not occur anymore.

AC3. \vec{X} is minimal; no subset of \vec{X} satisfies conditions AC1 and AC2.

Example

Context

- $S.Get(P)/B.Get(P) = T/T$
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- $EK = T \text{ OR } F = T$

Is $S.Get(K)$ a cause?

Set $S.Get(K) = F$ and $\vec{W} = \emptyset$

- $S.Get(P)/B.Get(P) = T/T$
- $S.Get(K)/B.Get(K) = F/T$
- $S.DK = T \text{ AND } F = F$
- $B.DK = T \text{ AND } T \text{ AND } T = T$
- $EK = F \text{ OR } T = T$

φ still occurs \rightarrow AC2

Set $S.Get(K) = F$ and $\vec{W} = \{B.DK\}$

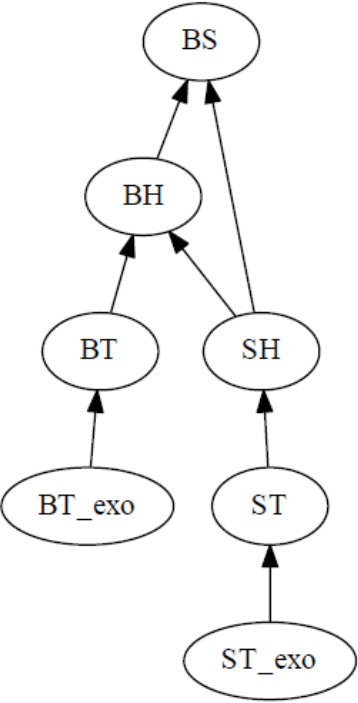
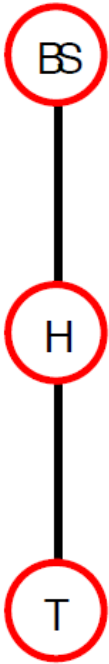
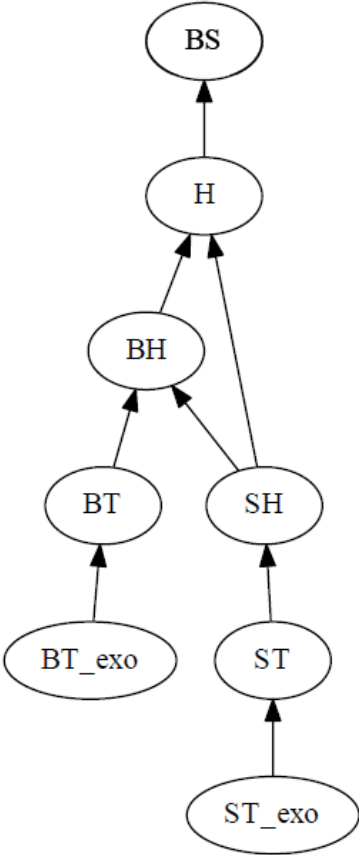
- $S.Get(P)/B.Get(P) = T$
- $S.Get(K)/B.Get(K) = F/T$
- $S.DK = T \text{ AND } F = F$
- $B.DK = \cancel{T} \text{ AND } \cancel{T} \text{ AND } \cancel{T} = F$
- $EK = F \text{ OR } F = F$

φ does not occur anymore \rightarrow AC2

Evaluation: Efficiency of the extraction

				2 Suspects						8 Suspects					
				Top		Middle		Leafs		Top		Middle		Leafs	
AT	n	l	b	n	exec(s)	n	exec(s)	n	exec(s)	n	exec(s)	n	exec(s)	n	exec(s)
SMK	12	5	2	37	0.0002	36	0.0002	36	0.0003	139	0.0004	126	0.0004	108	0.0004
Be.Root1	8	4	1	24	0.0002	25	0.0002	23	0.0002	90	0.0004	91	0.0004	71	0.0004
Be.Root2	11	4	1	32	0.0002	35	0.0002	32	0.0003	122	0.0006	125	0.0006	98	0.0006
T ₁	255	8	2	767	0.0069	767	0.0117	767	0.0512	3059	0.0283	2879	0.0460	2303	0.1925
T ₂	1017	8	8	3065	0.0354	3065	0.1133	3065	0.7473	12233	0.1380	11513	0.4610	9209	2.99
T ₃	3057	8	16	6129	0.0939	6129	0.4084	6129	2.94	24465	0.3700	23025	1.65	18417	11.97

Validity of the Models

Example	HP Model	Attack Tree	Our Model
Rock-Throwing	 <pre> graph BT BS((BS)) BH((BH)) BT((BT)) SH((SH)) BT_exo((BT_exo)) ST((ST)) ST_exo((ST_exo)) BT_exo --> BT ST_exo --> ST BT --> BH ST --> SH BH --> BS SH --> BS </pre>	 <pre> graph BT BS((BS)) --- H((H)) --- T((T)) </pre>	 <pre> graph BT BS((BS)) H((H)) BH((BH)) BT((BT)) SH((SH)) BT_exo((BT_exo)) ST((ST)) ST_exo((ST_exo)) BT_exo --> BT ST_exo --> ST BT --> BH ST --> SH BH --> H SH --> H H --> BS </pre>