

A Game-Theoretic Approach for Alert Prioritization

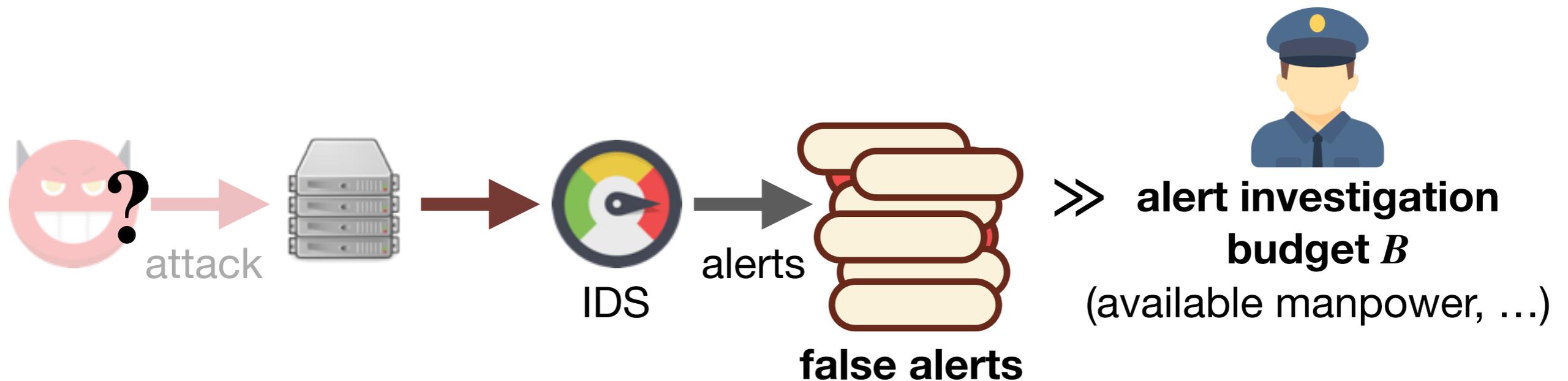
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Intrusion Detection

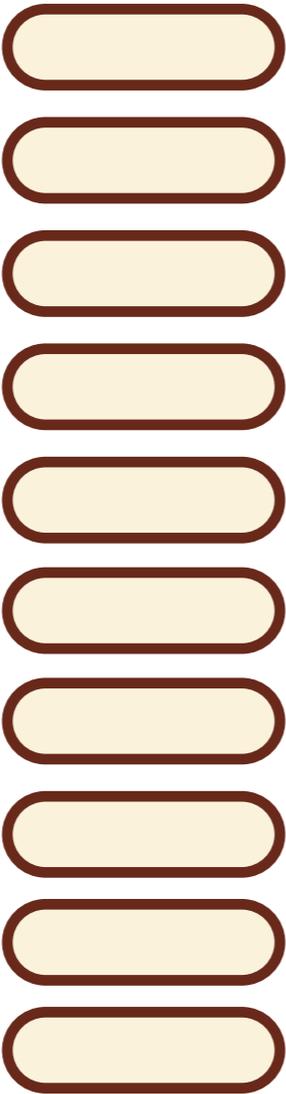
- Detection and mitigation of cyber-attacks is of crucial importance; however, attackers try to stay stealthy
- Intrusion Detection Systems (IDS)
 - generate alerts when they encounter suspicious activity



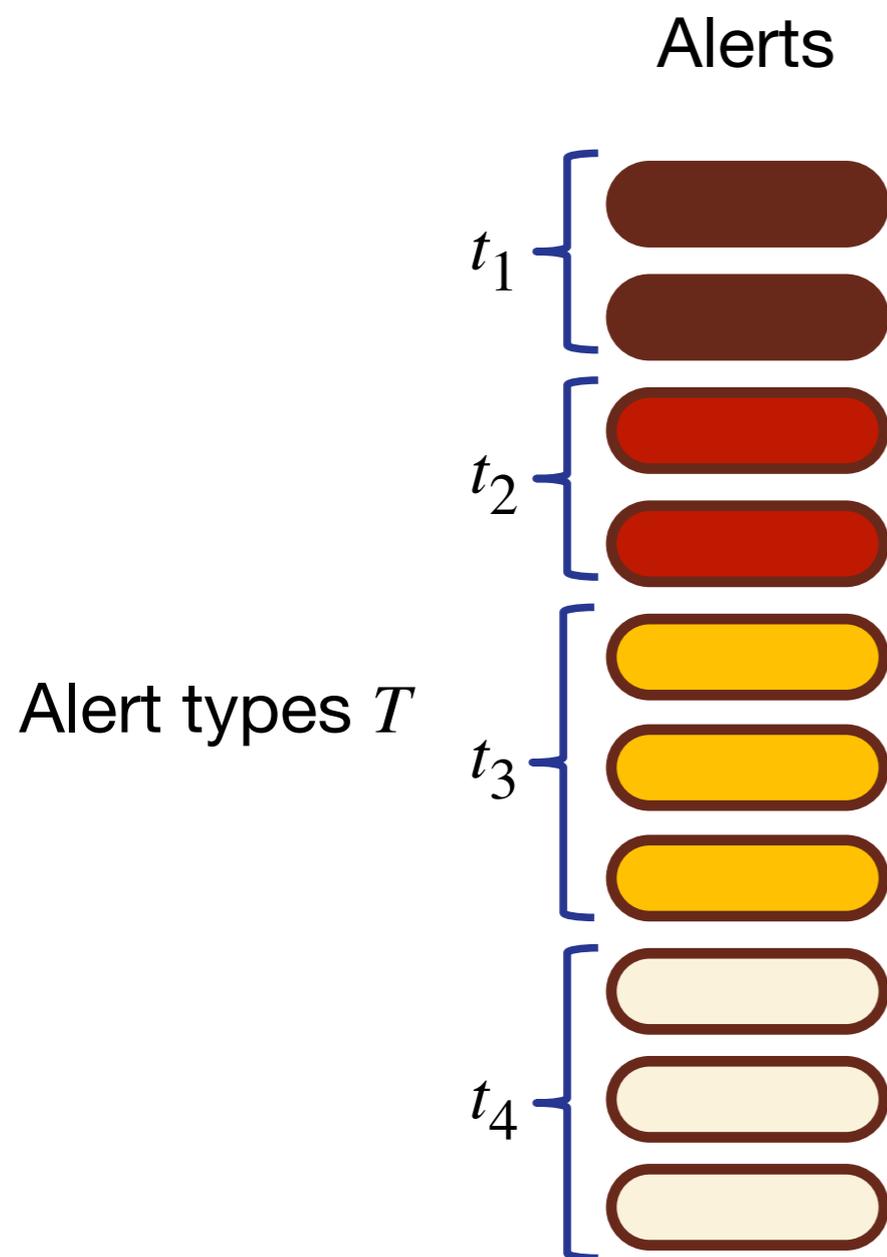
Problem:
Which alerts to investigate?

Alert Prioritization

Alerts



Alert Types



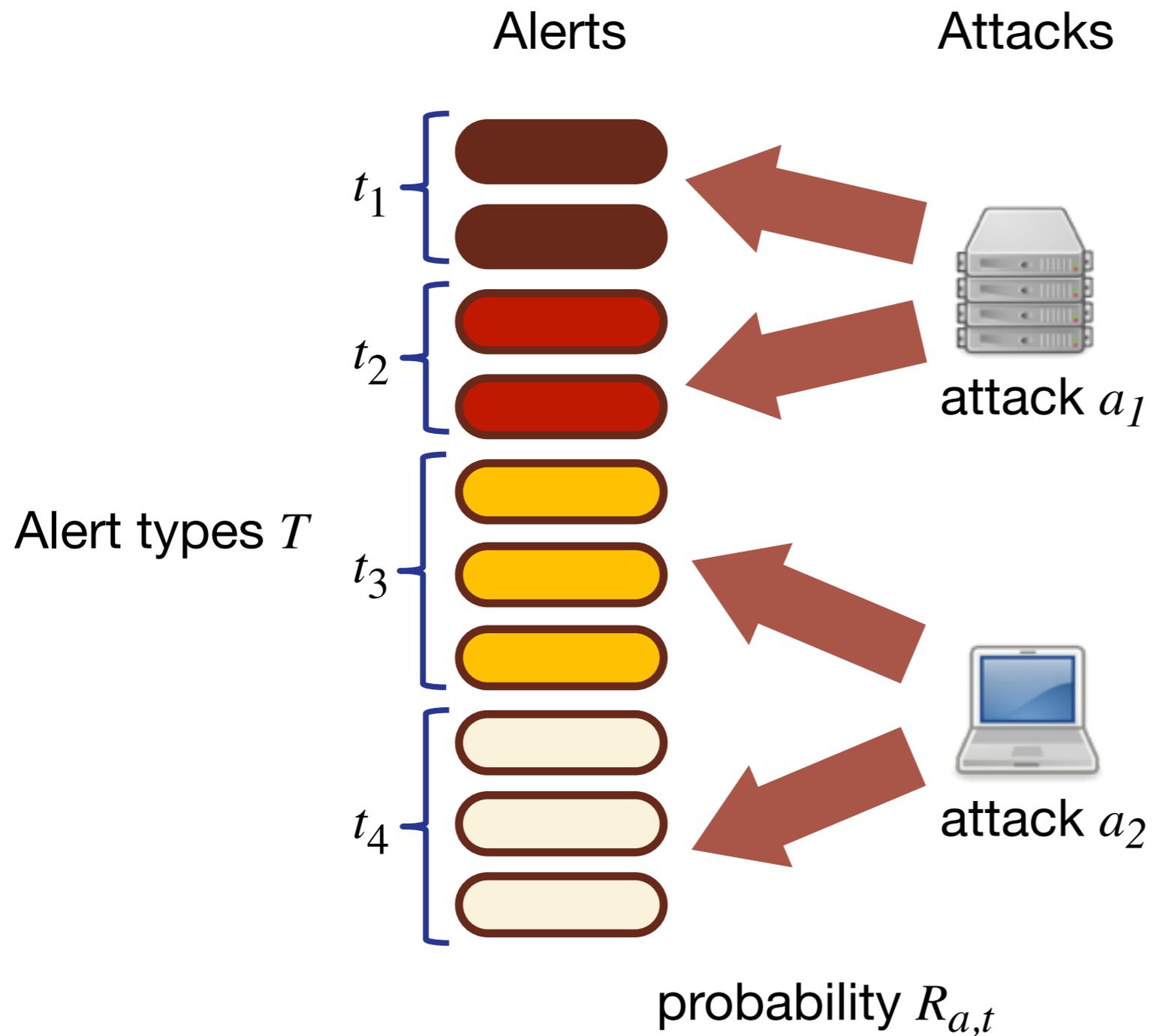
- Alert types T

- for example, matching different rules in an intrusion detection system (e.g., Snort)
- before investigating them, alerts of the same type appear equally important
- cumulative distribution F_t of the number of false alerts of type t is known

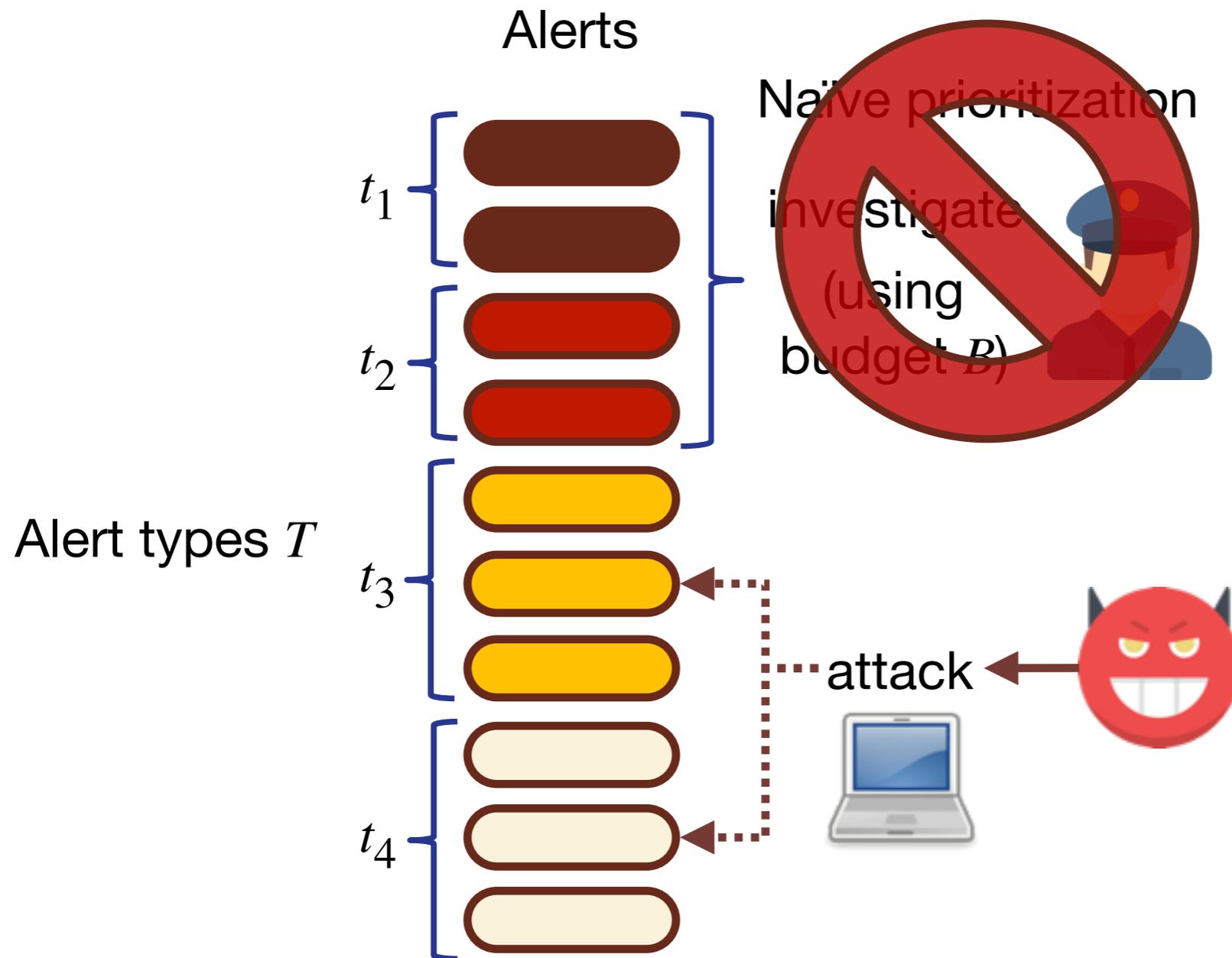
- Attacks A

- for example, targeting certain machines or using certain exploitation techniques
- impact of attack a is L_a
- probability of attack a raising an alert of type t is $R_{a,t}$

Alert Types



Alert Prioritization Problem



Alert Prioritization Problem



Problem:

What is the optimal probability distribution?

Game-Theoretic Model

- Players



1. Defender: selects an alert prioritization strategy p , which is a probability distribution over possible orderings of T



2. Adversary: selects an attack a from the set of possible attacks A

- Supposing that the defender uses ordering $\mathbf{o} \in T$

- probability of investigating type k (before exhausting budget B) is

$$PI(\mathbf{o}, k) = \sum_{\substack{\mathbf{n}: \\ C_{o_k} + \sum_{i=1}^k n_i \cdot C_{o_i} \leq B}} \left[(F_{o_k}^*(n_k) - F_{o_k}^*(n_k - 1)) \cdot \prod_{i=1}^{k-1} (F_{o_i}(n_i) - F_{o_i}(n_i - 1)) \right]$$

- probability of investigating attack a (before exhausting budget B) is

$$PD(\mathbf{o}, a) = \sum_{\hat{T} \subseteq T} \prod_{t \in \hat{T}} R_{a,t} \prod_{t \in T \setminus \hat{T}} (1 - R_{a,t}) PI(\mathbf{o}, \min\{i \mid o_i \in \hat{T}\})$$

Optimal Alert Prioritization

- Adversary's gain and defender's loss
 - adversary's expected gain: $EG(\mathbf{p}, a) = \sum_{\mathbf{o} \in O} p_{\mathbf{o}} \cdot (1 - PD(\mathbf{o}, a)) \cdot G_a - K_a$
 - defender's expected loss: $EL(\mathbf{p}, a) = \sum_{\mathbf{o} \in O} p_{\mathbf{o}} \cdot (1 - PD(\mathbf{o}, a)) \cdot L_a$
- Solution concept: strong Stackelberg equilibrium
 - adversary's best responses: $BR(\mathbf{p}) = \operatorname{argmax}_{a \in A} EG(\mathbf{p}, a)$
 - **optimal prioritization strategy:** $\min_{\mathbf{p}, a \in BR(\mathbf{p})} EL(\mathbf{p}, a)$

Challenge: finding an optimal probability distribution over a set of exponential size!

Theorem: Finding an optimal alert prioritization strategy is an **NP-hard** problem.

Computing Detection Probabilities

- Probability of detecting an attack

$$PI(\mathbf{o}, k) = \sum_{\substack{\mathbf{n}: \\ C_{o_k} + \sum_{i=1}^k n_i \cdot C_{o_i} \leq B}} \left[(F_{o_k}^*(n_k) - F_{o_k}^*(n_k - 1)) \cdot \prod_{i=1}^{k-1} (F_{o_i}(n_i) - F_{o_i}(n_i - 1)) \right]$$

$$PD(\mathbf{o}, a) = \sum_{\hat{T} \subseteq T} \prod_{t \in \hat{T}} R_{a,t} \prod_{t \in T \setminus \hat{T}} (1 - R_{a,t}) PI(\mathbf{o}, \min\{i \mid o_i \in \hat{T}\})$$

exponential number of terms

- Dynamic programming algorithm

Algorithm 1 Computing $PD(\mathbf{o}, a)$

Input: prioritization game, prioritization \mathbf{o} , attack a

- 1: **for** $b = 0, 1, \dots, B$ **do**
 - 2: $PD(\mathbf{o}, a, |T|, b) \leftarrow R_{a, o_{|T|}} \cdot F_{o_{|T|}}^*(\lfloor b/C_{o_{|T|}} \rfloor - 1)$
 - 3: **end for**
 - 4: **for** $i = |T| - 1, \dots, 2, 1$ **do**
 - 5: **for** $b = 0, 1, \dots, B$ **do**
 - 6: $PD(\mathbf{o}, a, i, b) \leftarrow R_{a, o_i} \cdot F_{o_i}^*(\lfloor b/C_{o_i} \rfloor - 1) + (1 - R_{a, o_i}) \sum_{j=0}^{\lfloor b/C_{o_i} \rfloor} (F_{o_i}(j) - F_{o_i}(j - 1)) \cdot PD(\mathbf{o}, a, b - j \cdot C_{o_i}, i + 1)$
 - 7: **end for**
 - 8: **end for**
 - 9: Return $PD(\mathbf{o}, a) := PD(\mathbf{o}, a, 1, B)$
-

Finding an Optimal Alert Prioritization Strategy

- Linear-programming based formulation

- for each attack $a \in A$, solve

$$\max_p \sum_{\mathbf{o} \in \mathcal{O}} p_{\mathbf{o}} \cdot PD(\mathbf{o}, a)$$

subject to

$$\forall a' \in A : \sum_{\mathbf{o} \in \mathcal{O}} p_{\mathbf{o}} \cdot D(\mathbf{o}, a') \geq \Delta(K_{a'})$$

where

$$D(\mathbf{o}, a') = [(1 - PD(\mathbf{o}, a))G_a - (1 - PD(\mathbf{o}, a'))G_{a'}]$$

$$\Delta(K_{a'}) = K_a - K_{a'}$$

**exponential number
of possible orderings**

- output the solution that attains the lowest loss

Problem: Finding an improving column (i.e., ordering) is an **NP-hard** problem.

- Polynomial-time column generation approach

Algorithm 2 Greedy Column Generation

Input: prioritization game, reduced cost function \bar{c}

- $\mathbf{o} \leftarrow \emptyset$
- while** $\exists t \in T \setminus \mathbf{o}$ **do**
- $\mathbf{o} \leftarrow \mathbf{o} + \operatorname{argmax}_{t \in T \setminus \mathbf{o}} \bar{c}(\mathbf{o} + t)$
- end while**
- Return \mathbf{o}

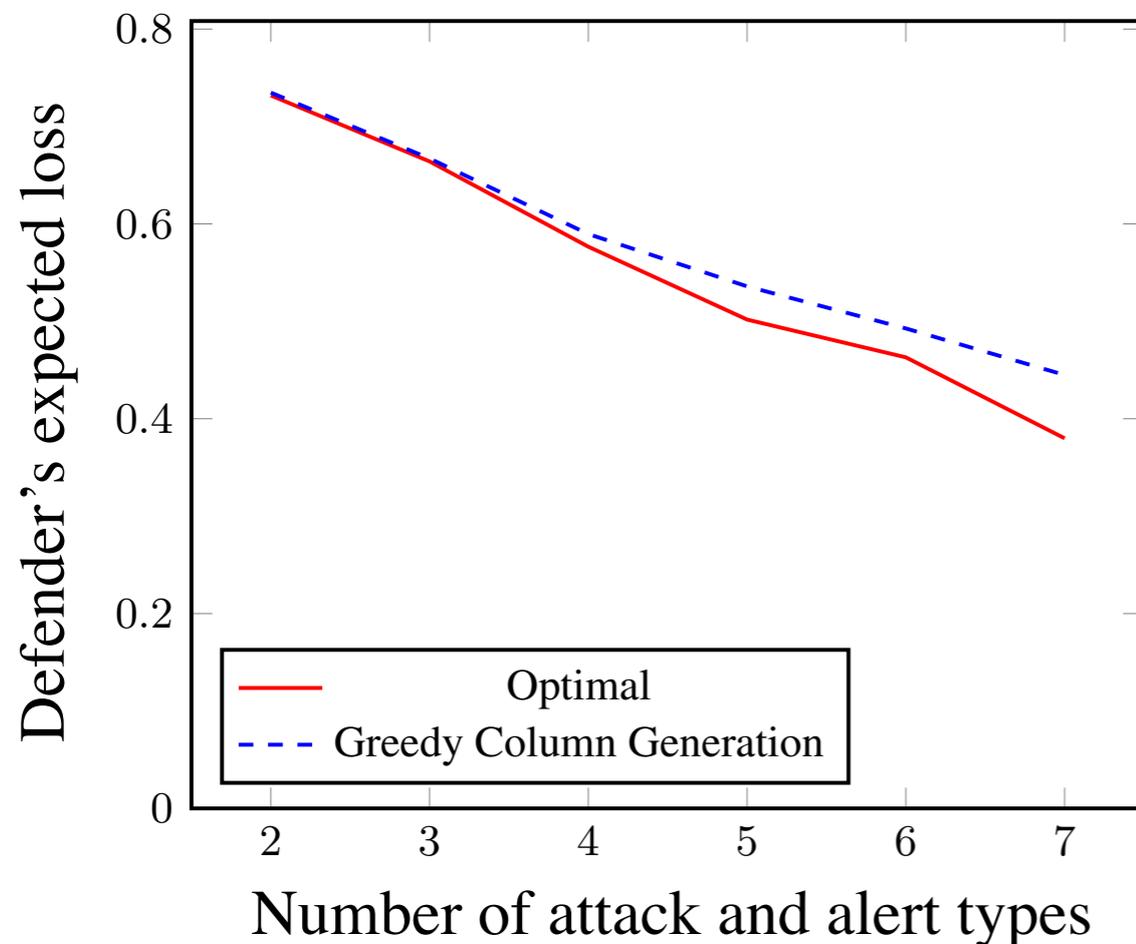
where

$$\bar{c}(\mathbf{o}) = PD(\mathbf{o}, a) + \sum_{a' \in A} y(\bar{\mathbf{O}}, a') D(\mathbf{o}, a')$$

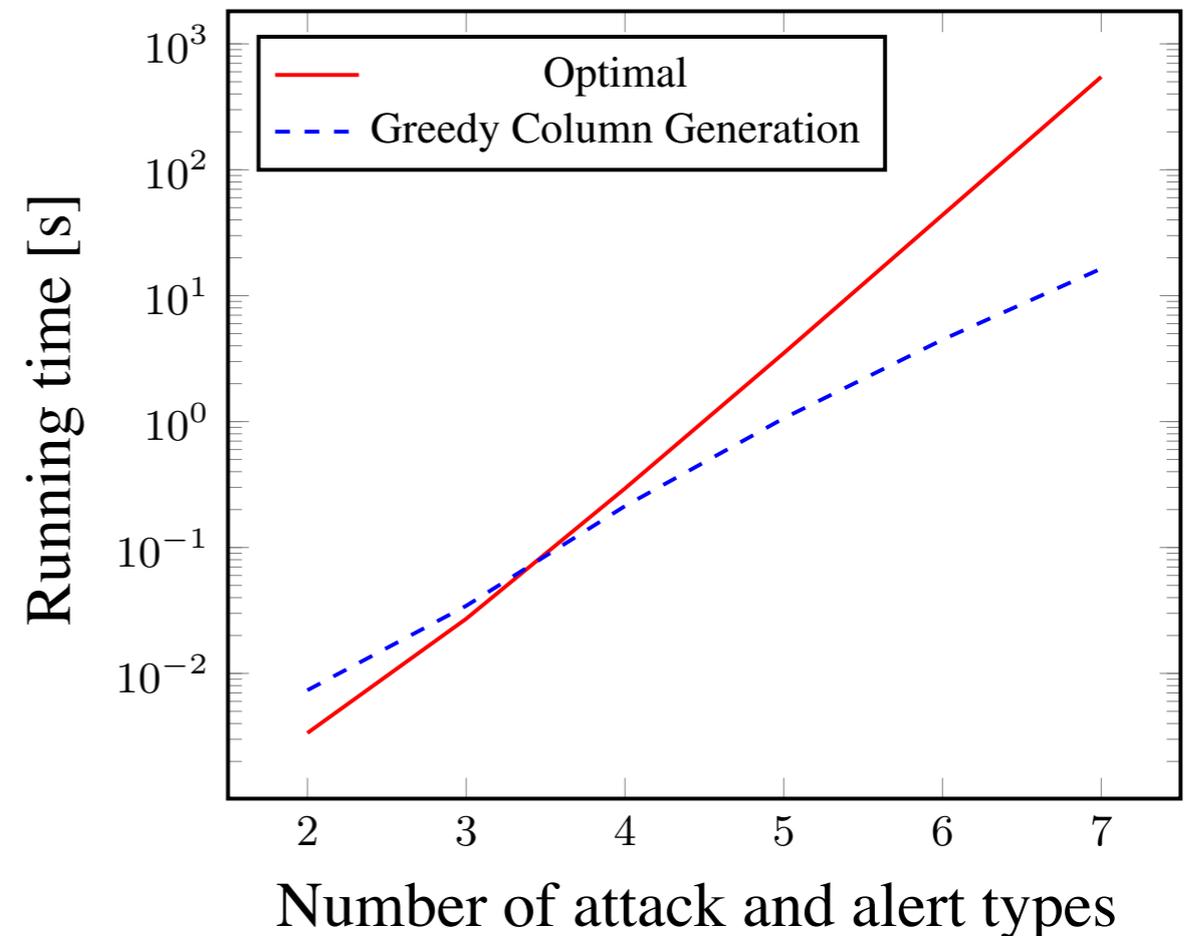
(i.e., reduced cost function)

Numerical Results - Synthetic Dataset

Defender's Loss



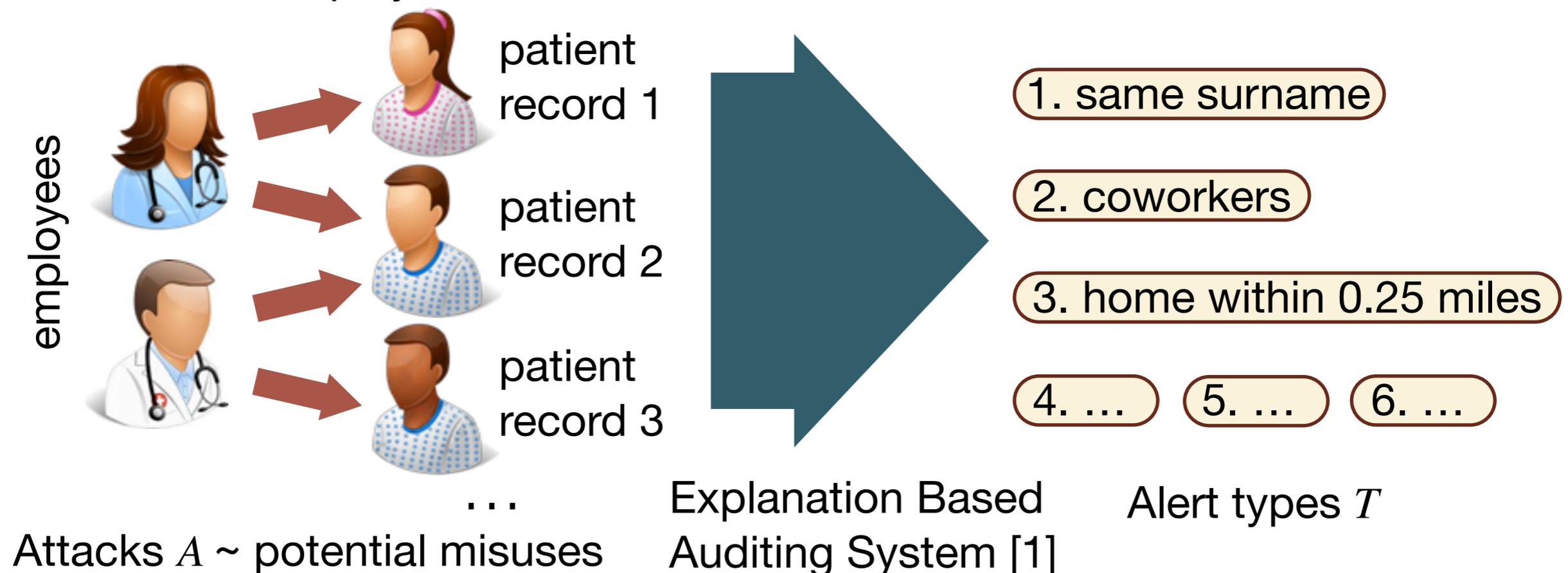
Running Time



$K_a = 0$, $C_t = 1$, $B = 5|T|$, D_a and G_a were drawn at random from $[0.5, 1]$, each $R_{a,t}$ is either 0 (with probability $1/3$) or drawn at random from $[0, 1]$, and every F_t has a Poisson distribution whose mean is drawn at random from $[5, 15]$.

Real-World Dataset: Electronic Medical Record System Alerts

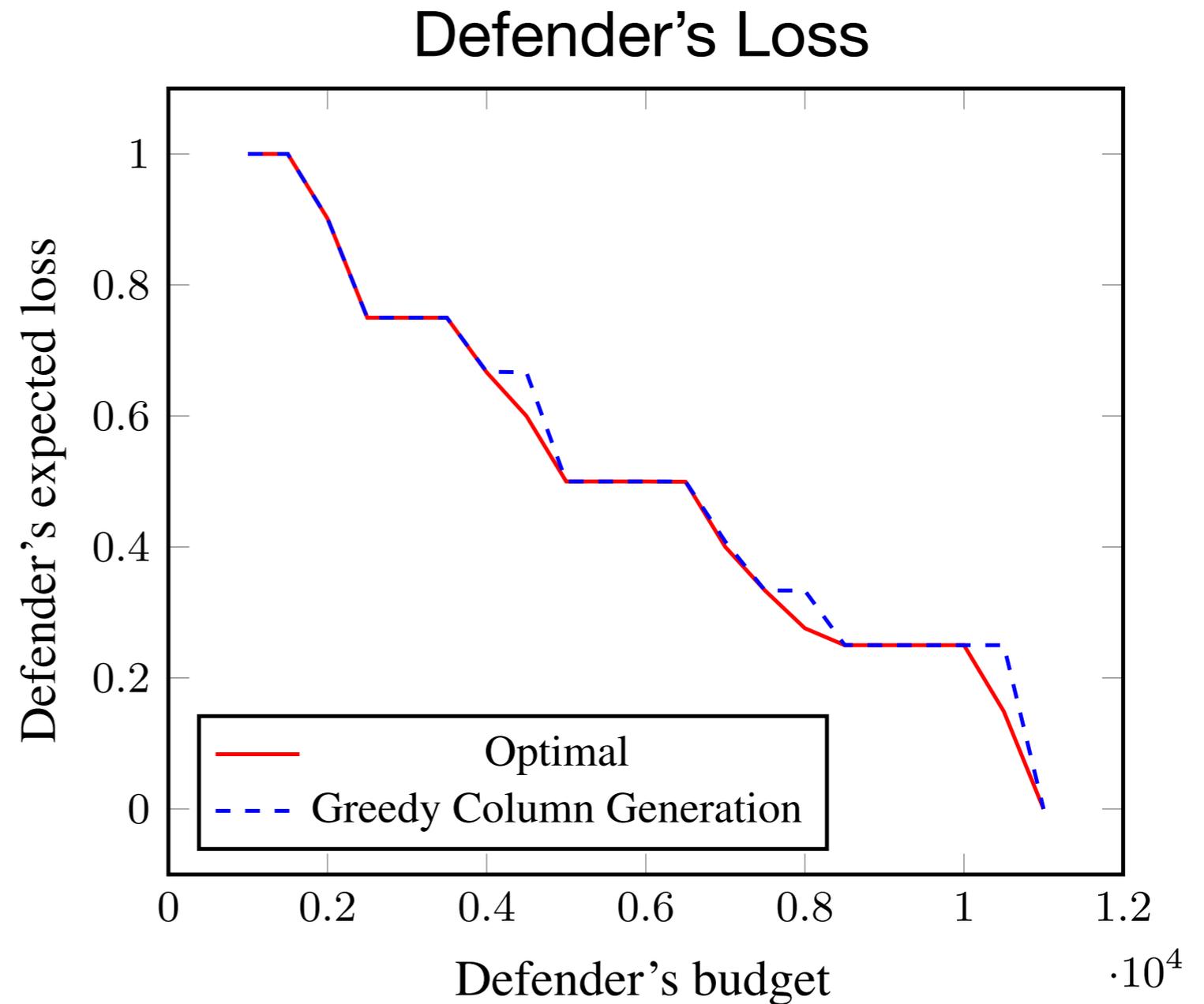
- Access logs from the **electronic medical record (EMR)** system in place at **Vanderbilt University Medical Center**
 - integrated with human-resources data to document medical department affiliation, employment information, and home addresses



[1] Fabbri, D., and LeFevre, K. 2013. Explaining accesses to electronic medical records using diagnosis information. *Journal of the American Medical Informatics Association* 20(1):52–60.

Numerical Results - Real-World Dataset

- Data collected from five consecutive weeks of access logs from 2016
- 8,481,767 accesses made by 14,531 users to 161,426 patient records, leading to a total of **863,989** alerts
- Approximated the distributions of false alerts using Poisson distributions
- In order to find optimal strategies, we restricted the alerts to 12 randomly selected patients



Conclusion & Future Work

- Prioritization of alerts is of crucial importance to the effectiveness of intrusion and misuse detection
- Result highlights
 - introduced **first model of alert prioritization against strategic adversaries**
 - showed that finding an optimal prioritization strategy is **NP-hard**
 - proposed an efficient **column-generation based approach**
 - evaluated numerically using **synthetic and real-world datasets**
- Future work
 - constant **approximation ratio** algorithms
 - modeling **multiple adversary types** as a Bayesian Stackelberg game

Thank you for your attention!

Questions?

