



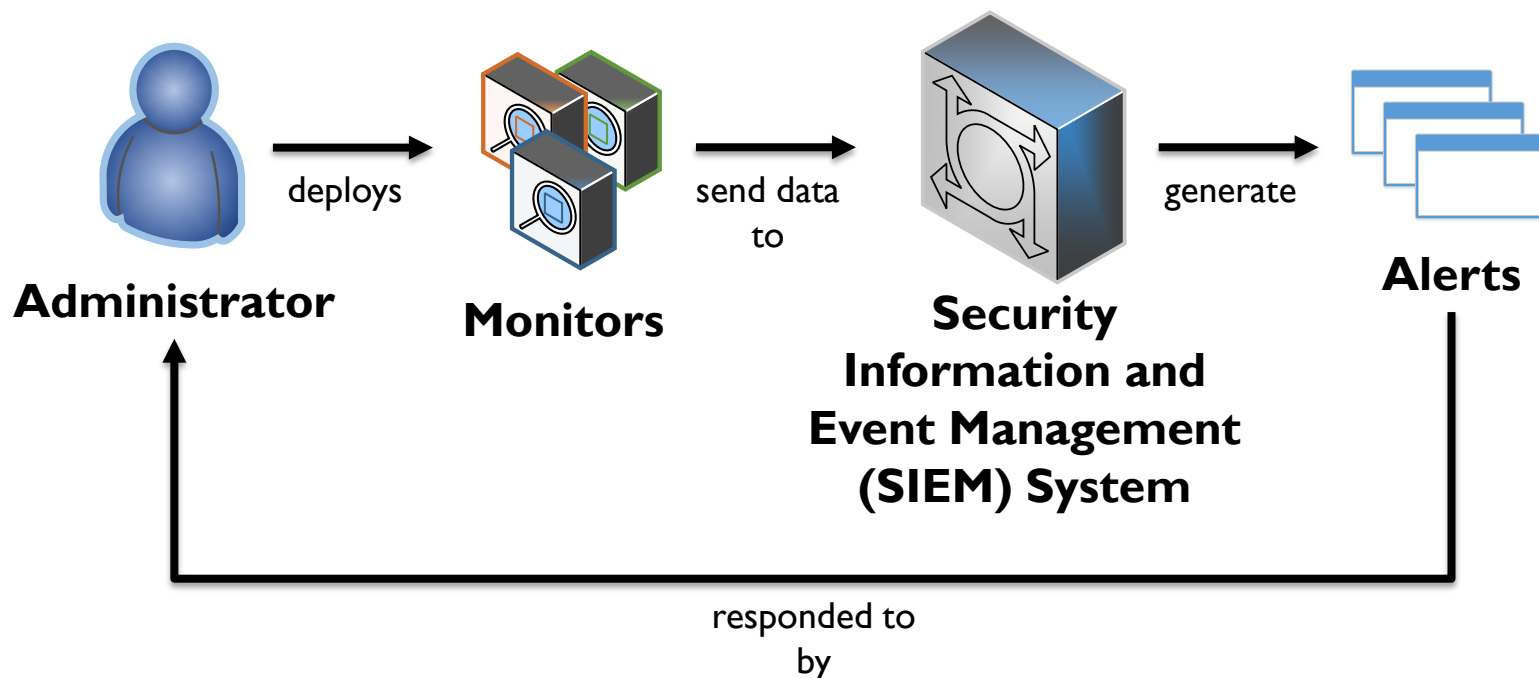
A Quantitative Methodology for Security Monitor Deployment

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Security monitoring today

Administrators must decide what data to collect and how it should be analyzed





Problem

Difficult to determine which monitors are **necessary to meet intrusion detection requirements**

Risks:

- Overprovisioned monitors – large volumes of poorly actionable logs
- Underprovisioned monitors – insufficient ability to detect or investigate security incidents

We help administrators determine exactly where they stand

- Can expose weaknesses in monitoring



Our contribution

We have developed a **quantitative, cost-sensitive** methodology for monitor selection that **meets intrusion detection requirements**

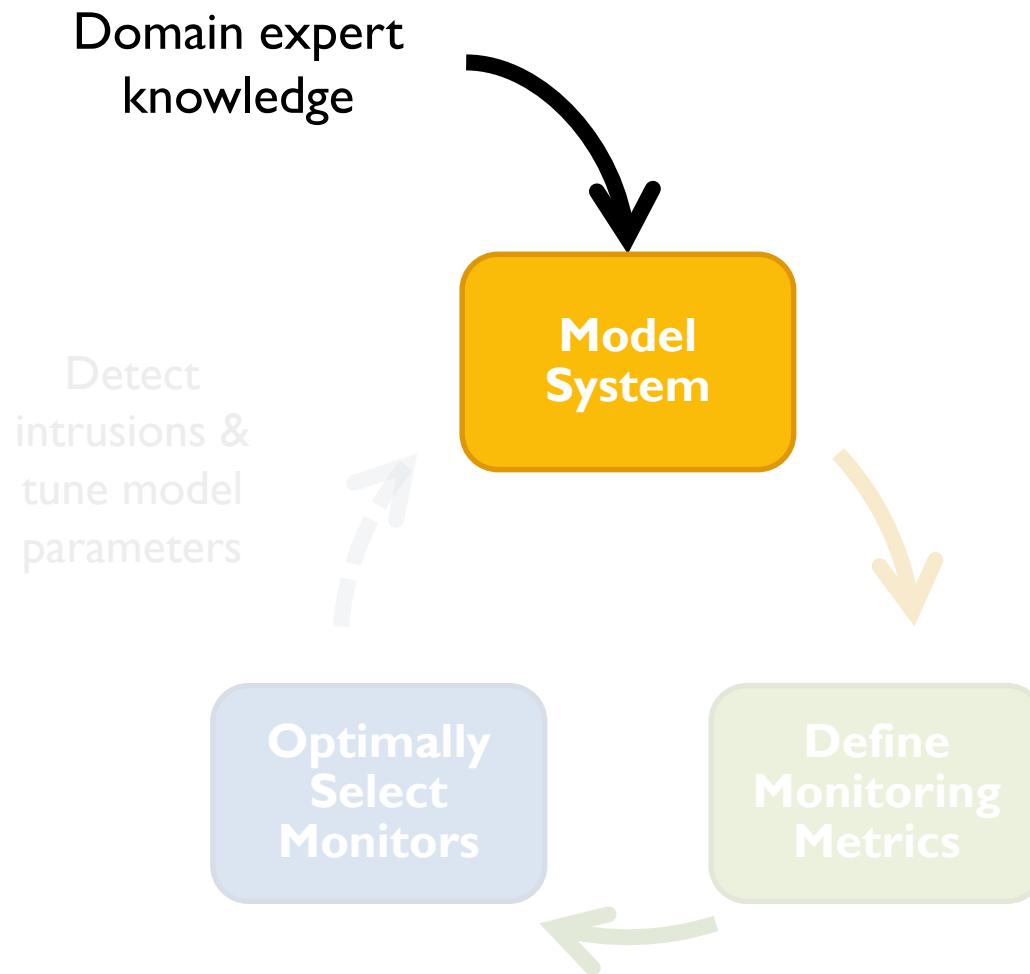


Guiding principles

- Monitors and computing assets can be **compromised**
 - Monitor compromise can affect ability to detect intrusions
- **Redundant monitoring** can mitigate the effect of compromise or unavailability



Outline





Model: Data model

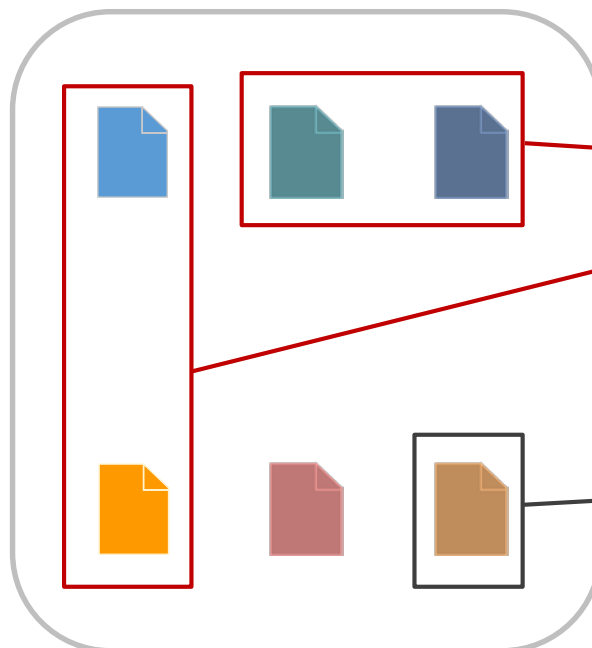
Monitors

Sensors that collect information about the system



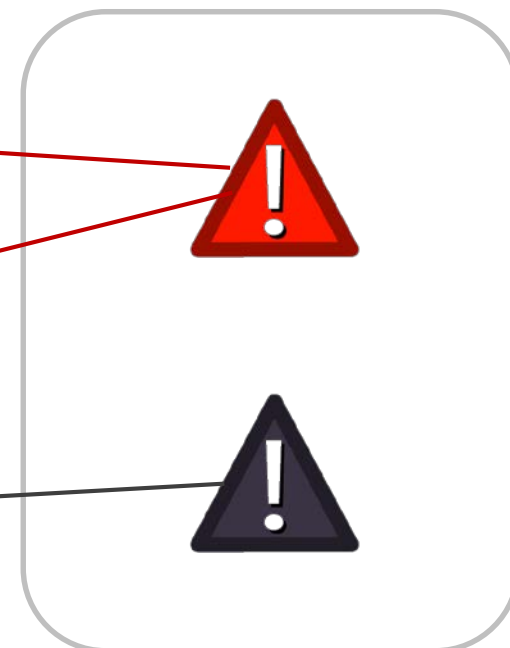
Indicators

Primitives representing information provided by monitors about events

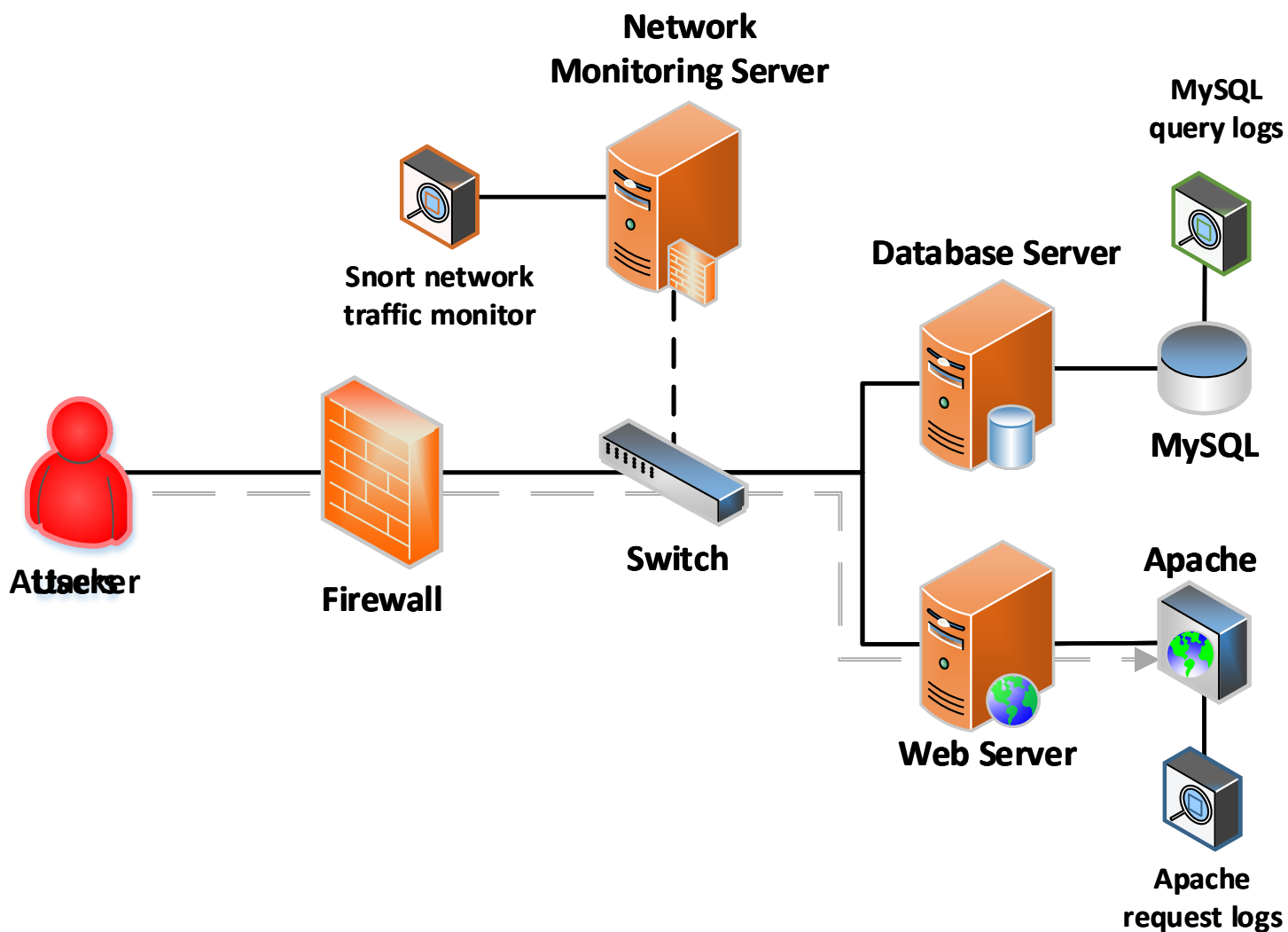


Events

Intrusions or actions symptomatic of attacks

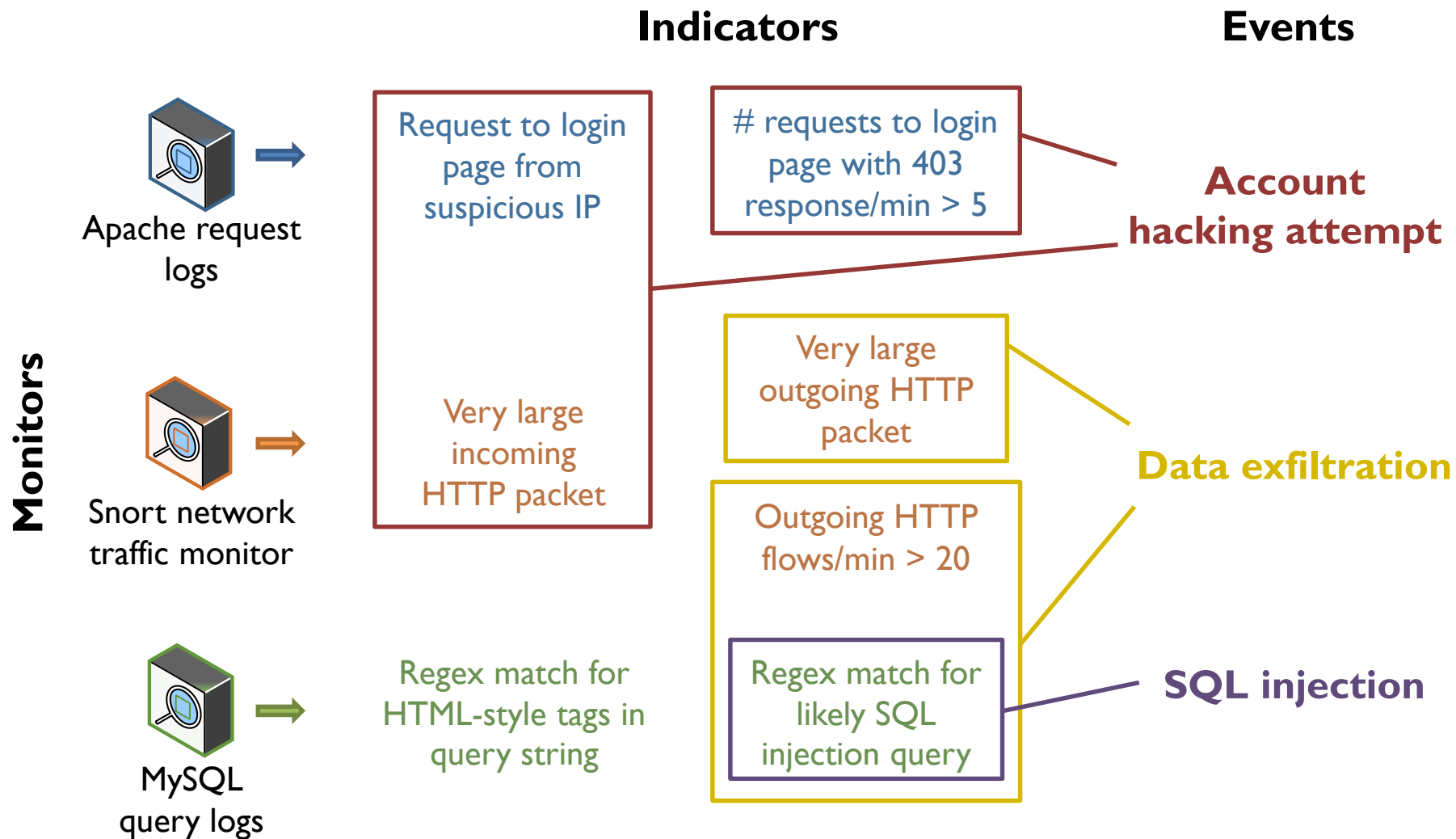


Case study: E-commerce web service

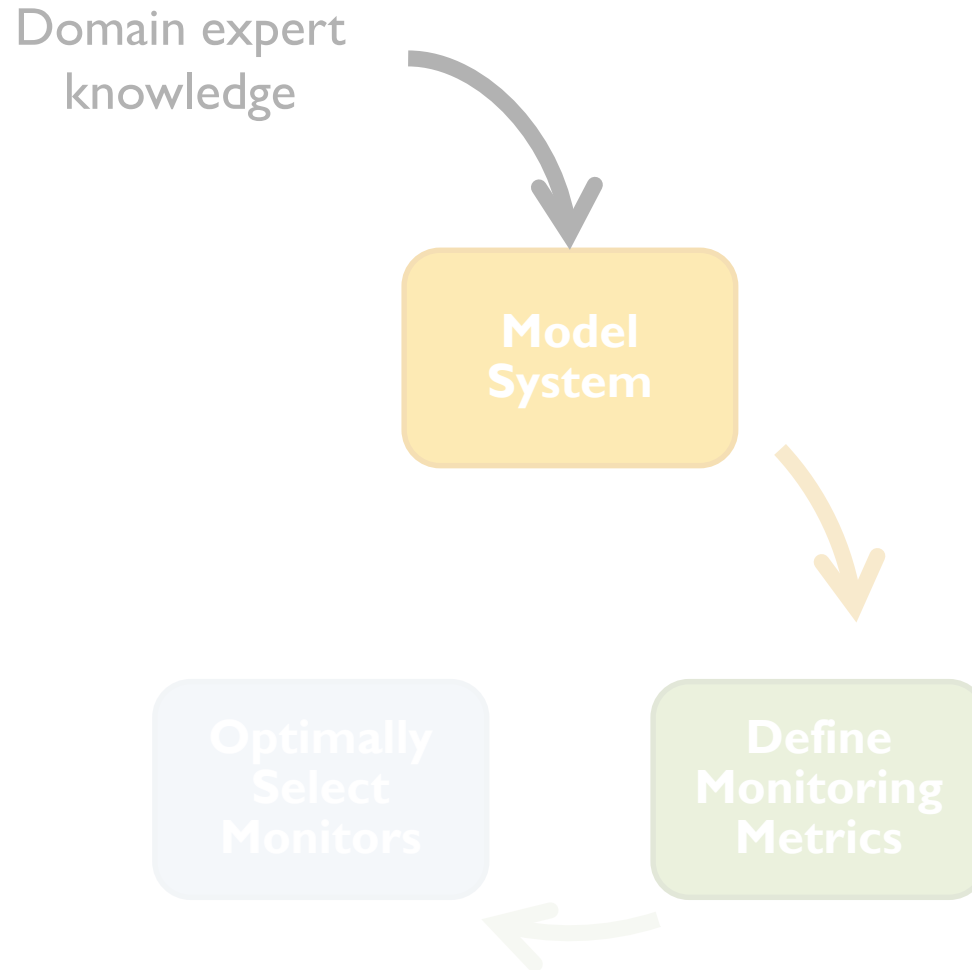




Model: Case study data model



Outline



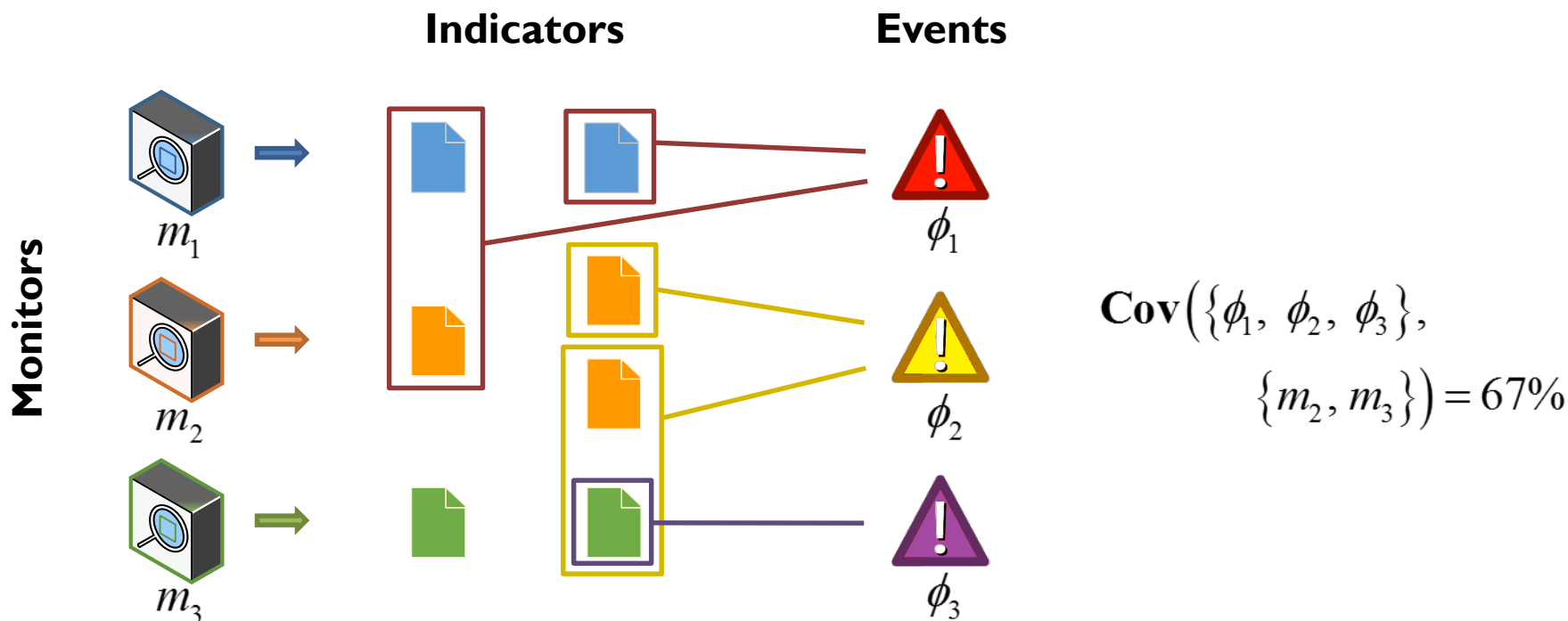
Monitoring metrics

- **Goal of metrics:** *quantify* utility and cost of monitors in supporting intrusion detection
- Three monitor utility metrics:
 - Coverage
 - Redundancy
 - Confidence
- One cost metric:
 - Monitor cost



Metrics: Coverage

Definition: overall fraction of select events that are detectable given a set of monitors

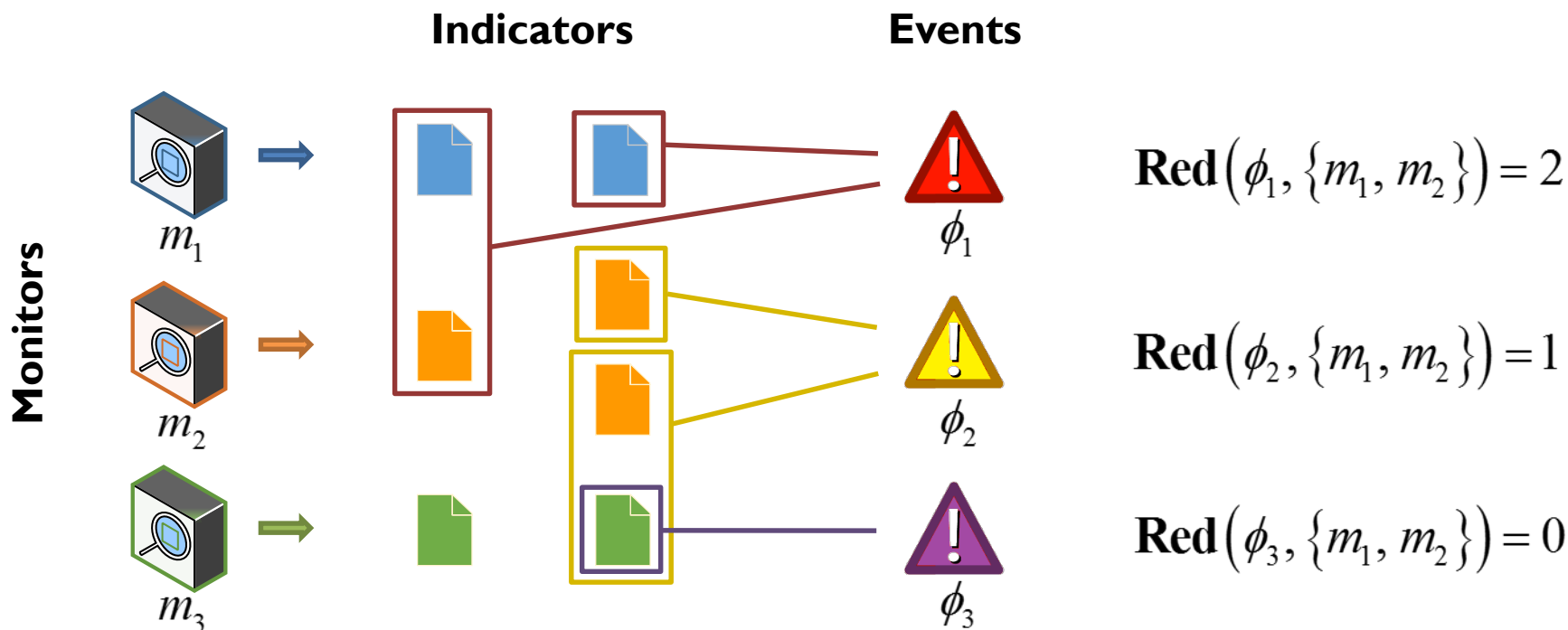


$$\text{Cov}(\Phi, M_d) = \frac{|\{\phi \mid \delta(\phi, M_d) \wedge \phi \in \Phi\}|}{|\Phi|}$$



Metrics: Redundancy

Definition: the number of ways an event can be detected given a set of monitors

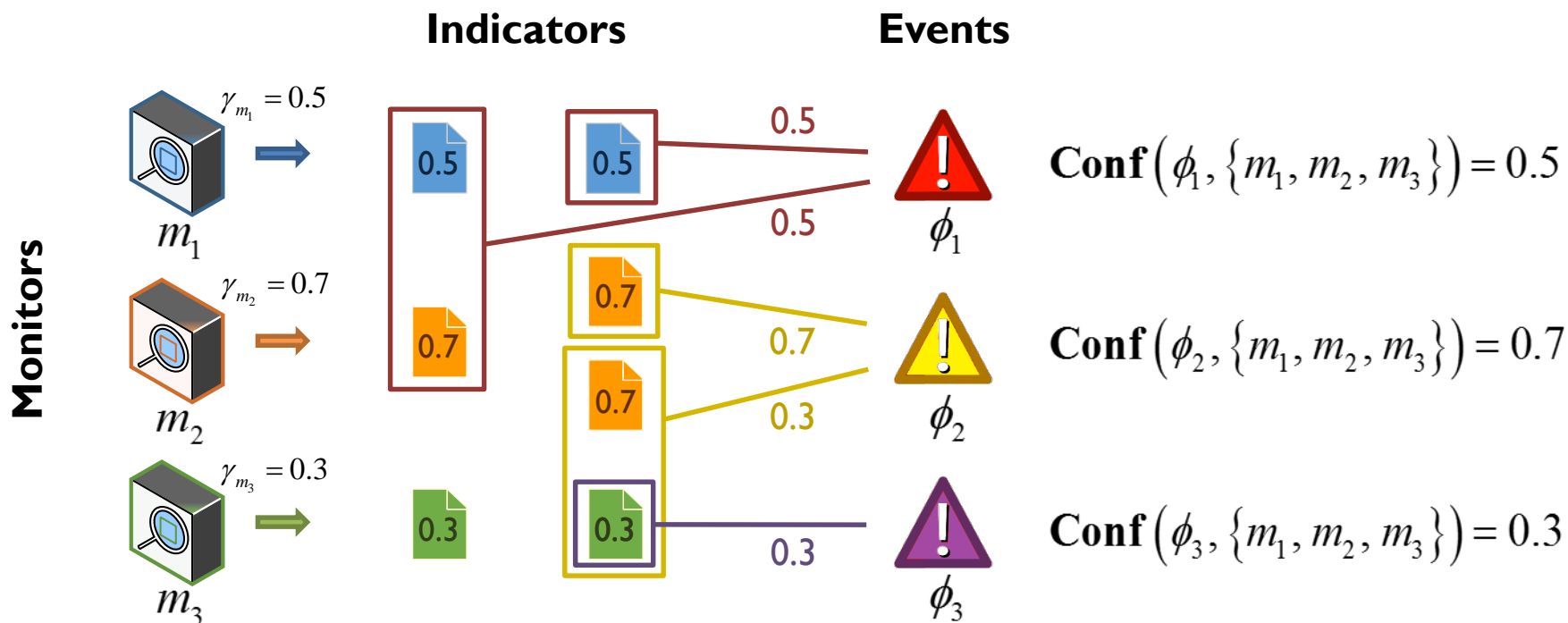


$$\text{Red}(\phi, M_d) = \sum_{\sigma \in \zeta(\phi, M_d)} \min_{i \in \sigma} |\{m \mid m \in M_d, i \in \alpha(m)\}|$$



Metrics: Confidence

Definition: belief in the ability to detect events accurately, even when monitors are compromised



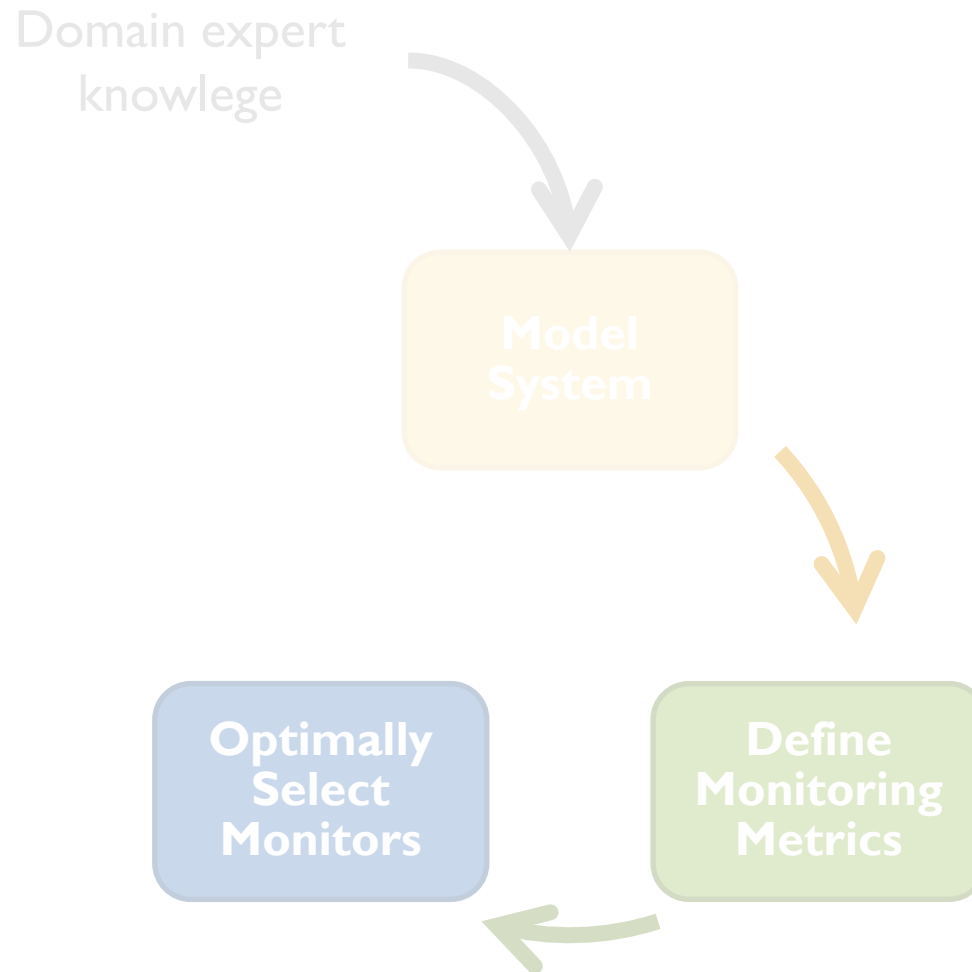
$$\text{Conf}(\phi, M_d) = \max_{\sigma \in \beta(\phi)} \min_{I \in \sigma} \gamma_I(I, M_d)$$



Metrics: Cost model

- Resource utilization cost
 - CPU utilization
 - Memory utilization
 - Disk storage
 - Network communication
- Amortized purchase price and recurring maintenance cost

Outline



Optimal selection methodology

Goal: to be able to use methodology to answer a variety of monitor selection questions

- Minimum set of monitors that can detect a given attack/set of attacks (assuming no compromise)?
- Under cost constraints, what set of monitors will maximize ability to detect a high-priority attack?



Capturing intrusion detection requirements

Represent detection requirements as **weights on metric values** and **minimum metric value constraints**



ϕ_1

Account hacking attempt



ϕ_2

Data exfiltration



ϕ_3

SQL injection

Requirements:

- Must detect exfiltration
- Exfiltration, then SQL injection, are high priority (decreasing priority)
- Best effort for all others

$$\min_{\text{Red}_{\phi_2}} = 1 \quad \mathbf{w}_{\text{Red}_{\phi_2}} = 2 \quad \mathbf{w}_{\text{Cov}} = 1$$

$$\mathbf{w}_{\text{Red}_{\phi_3}} = 1$$



Optimal selection methodology: Constrained-cost monitor selection

$$\arg \max_{M_d} \quad \mathbf{w}_{\text{Cov}} \mathbf{Cov}(\Phi, M_d) + \sum_{\phi \in \Phi} \mathbf{w}_{\text{Red}_\phi} \mathbf{Red}(\phi, M_d) + \mathbf{w}_{\text{Conf}_\phi} \mathbf{Conf}(\phi, M_d)$$

$$\mathbf{Cost}(M_d) \leq \mathbf{maxCost}$$

$$\text{s.t.} \quad \mathbf{Cov}(\Phi, M_d) \geq \mathbf{min}_{\text{Cov}}$$
$$\mathbf{Red}(\phi, M_d) \geq \mathbf{min}_{\text{Red}_\phi}, \quad \forall \phi \in \Phi$$
$$\mathbf{Conf}(\phi, M_d) \geq \mathbf{min}_{\text{Conf}_\phi}, \quad \forall \phi \in \Phi$$
$$M_d \in \{0, 1\}^{|M|}$$

Objective function: monitoring utility, defined as weighted sum of metric values

- Parameterized by user-specified weight parameters

Constraints:

- Cost function to minimize
- User-specified minimum detection metric requirements

0-1 integer nonlinear programming problem, with monitors as input variables



Solving for optimal monitor selection

- Branch-and-bound algorithm
 - Searches over space of possible selections, pruning suboptimal sets of monitor selections
- Greedy heuristic algorithm
 - Maximizes effective utility increase by incrementally adding monitors until constraints are met



EVALUATION



Experiment Setup

- Parameters: number of *monitors* and *events*
- Randomly generated 100 models for each set of parameters
- Created 4 sets of intrusion detection requirements – *optimal deployment programs*

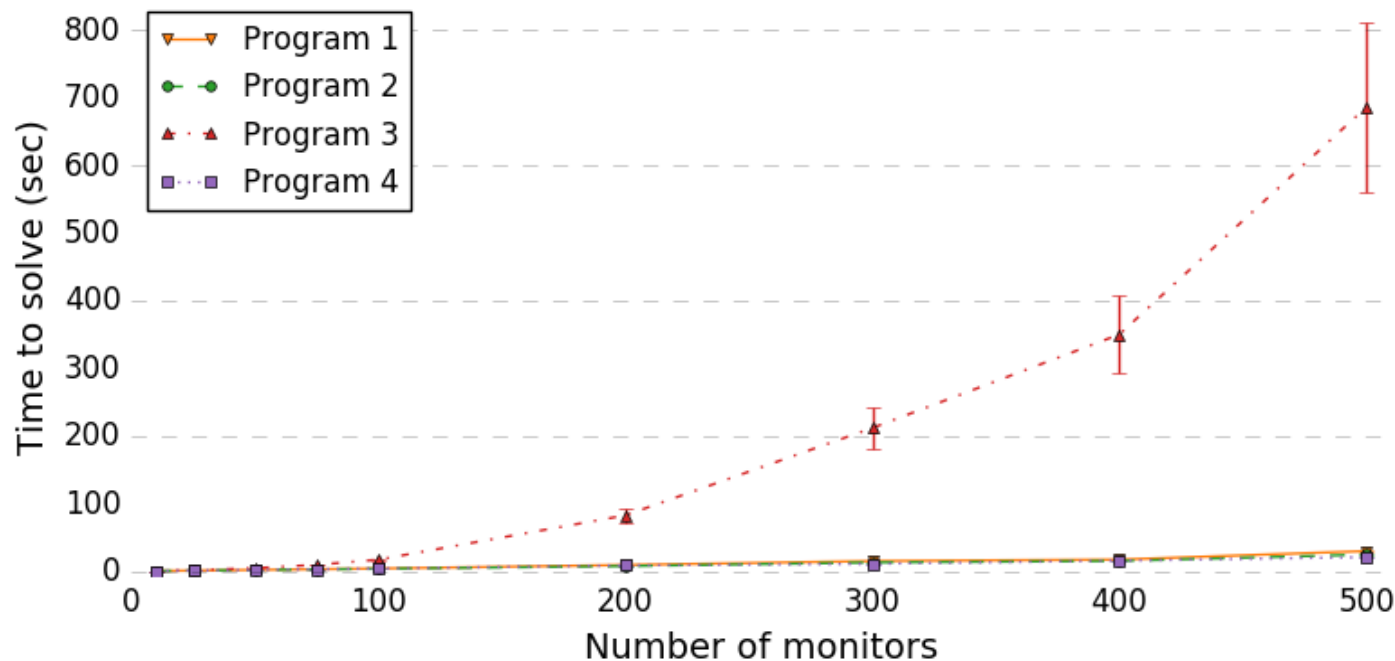
Goal: Observe scalability and accuracy of greedy solution



Evaluation: Greedy algorithm

Runtime complexity: $O(|I|(|M|^3 + |B||M|^2))$

– Polynomial in the number of monitors ($|M|$)

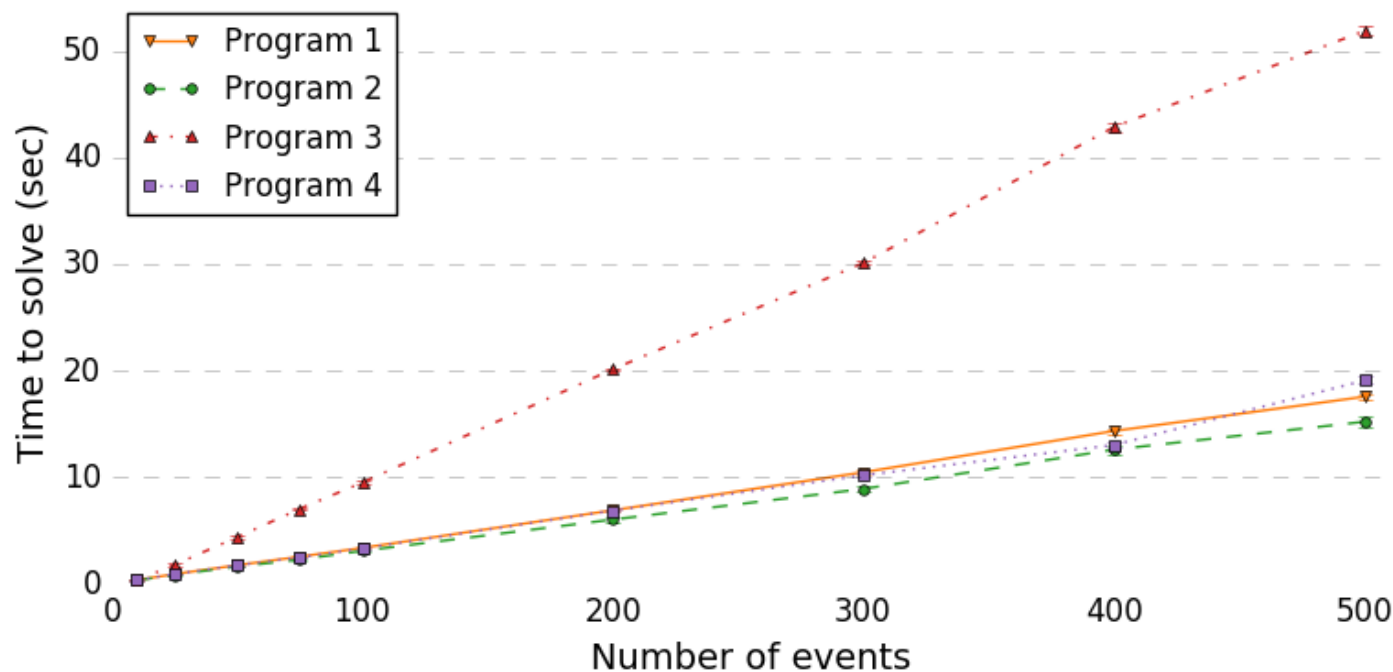




Evaluation: Greedy algorithm

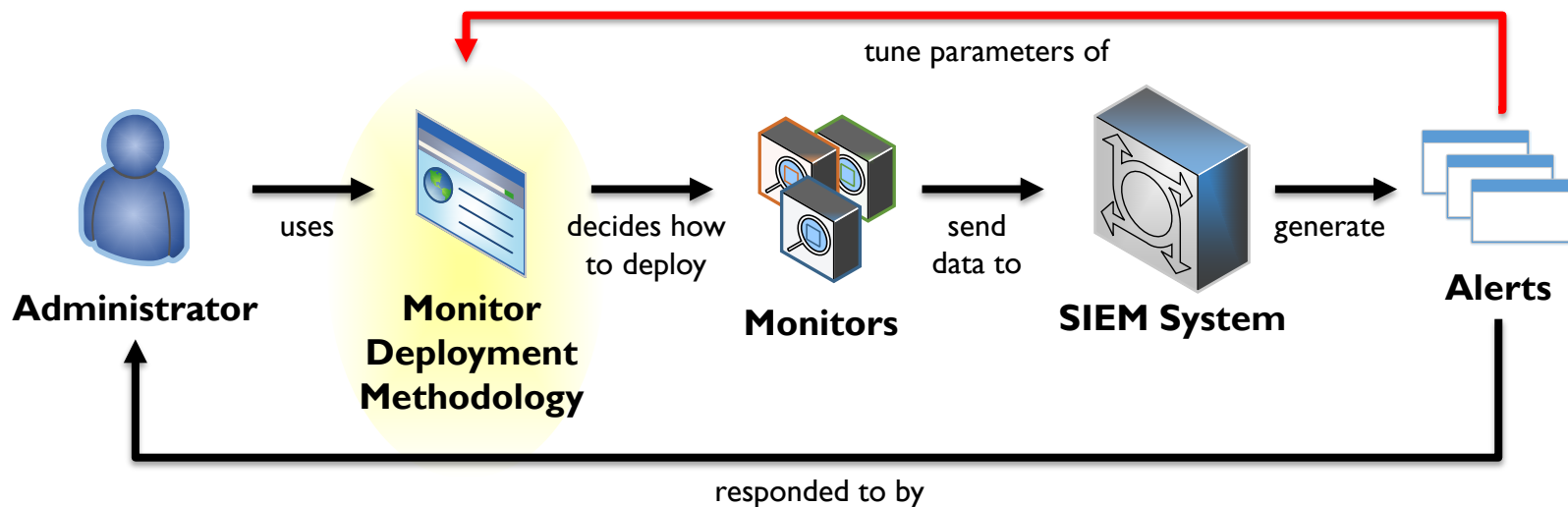
Runtime complexity: $O(|I|(|M|^3 + |B||M|^2))$

- Linear in the number of minimal indicator sets ($|B|$) and indicators ($|I|$)





Conclusions



- We help administrators make model-driven monitor placement decisions
- Administrators can more easily evaluate deployments
- Our methodology is expressive and scalable

Future work:

- Preemptive monitoring