

# Bayesian games for optimal cybersecurity investment with incomplete information on the attacker

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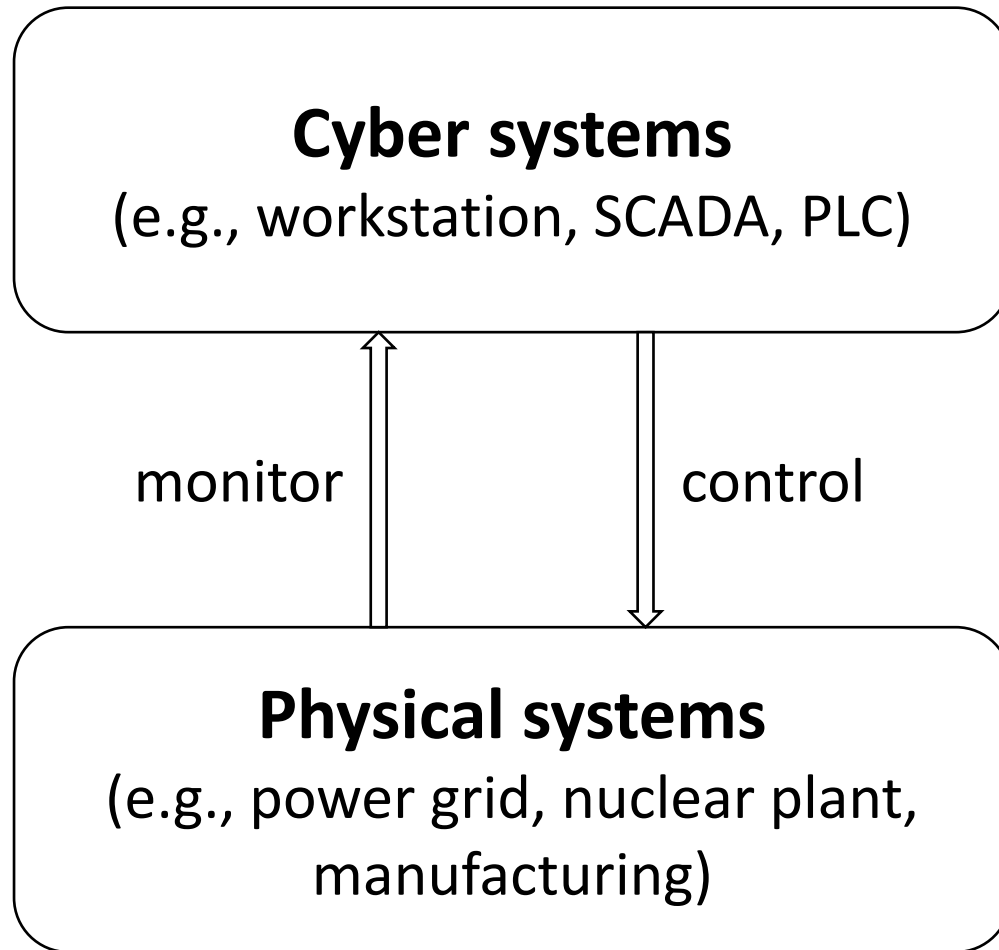
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# Background



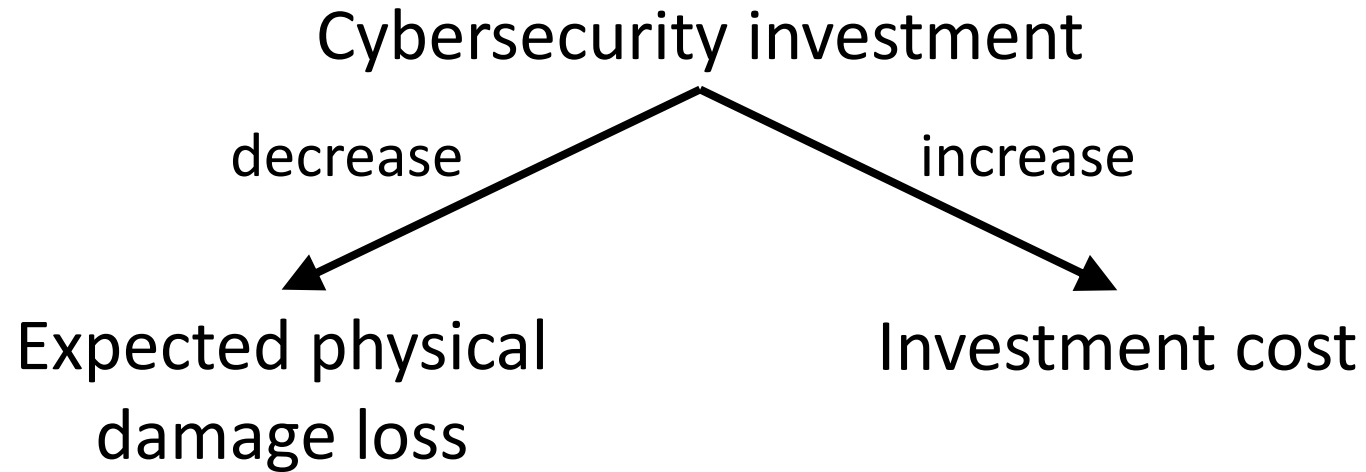
Cyber system compromise



Physical system damage



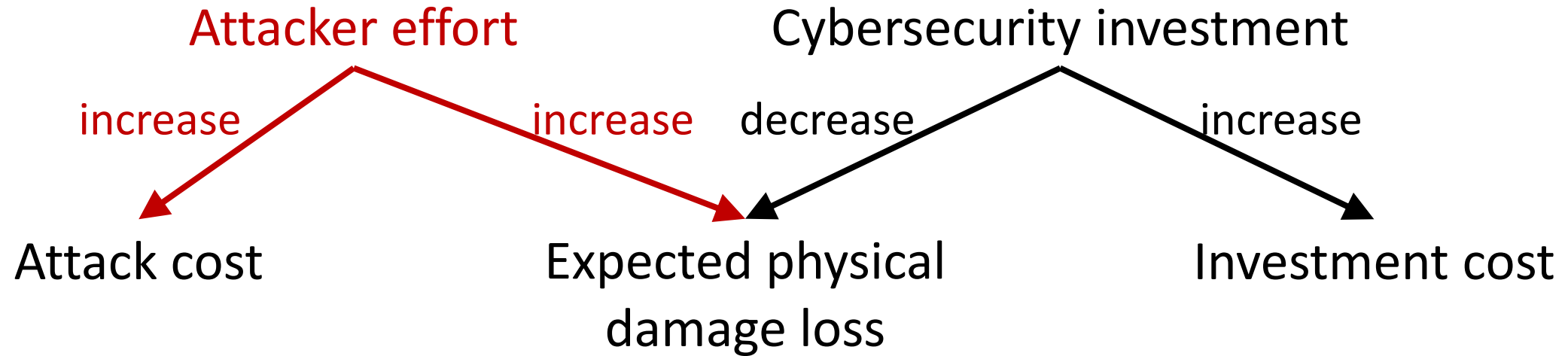
# Single-agent decision-making



**Optimal** cybersecurity investment: the level of investment that achieves the **minimum sum** of **physical damage loss** and **investment cost**.



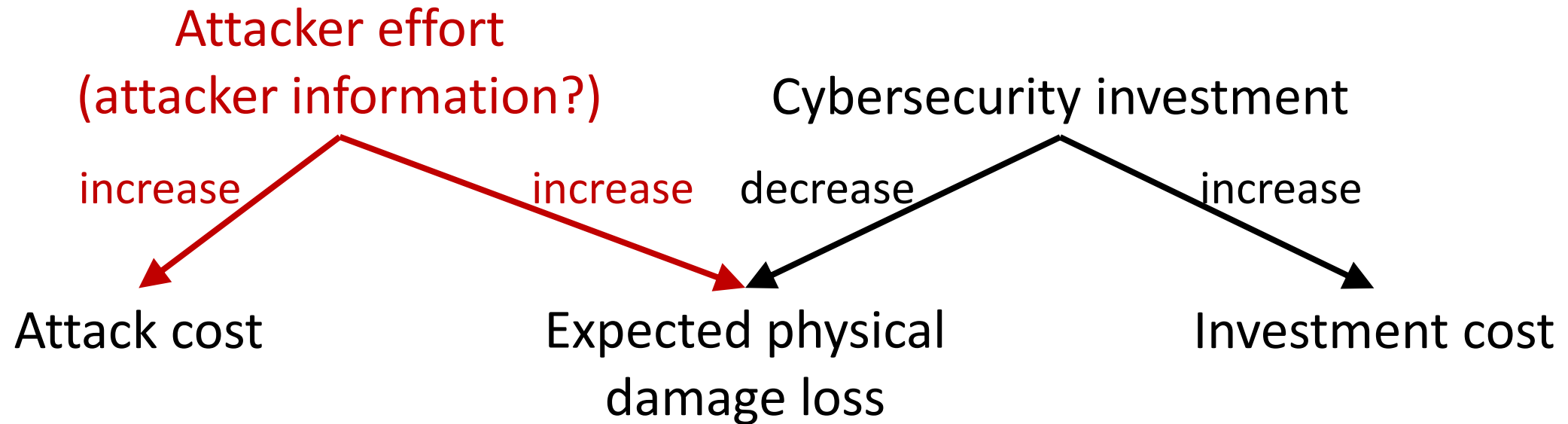
# Multi-agent decision-making



**Optimal** cybersecurity investment: the level of investment that achieves the **minimum sum** of **physical damage loss** and **investment cost** considering **the effort of the attacker**.



# Multi-agent decision-making with incomplete information



**Optimal** cybersecurity investment: the level of investment that achieves the **minimum sum** of **physical damage loss** and **investment cost** considering **the effort of the attacker** and with **incomplete information on the attacker**.



# Problem formalization

- Problem

- A piece of cyber equipment: if compromised, will incur physical loss of  $C$
- Players: defender – 1; attacker – 2
- Defender cybersecurity investment:  $a_1 \in [0, +\infty)$
- Attacker type:  $\theta_2 \in \Theta_2$
- Attacker type distribution:  $p(\theta_2)$
- Attacker attack effort:  $a_2 = \sigma_2(\theta_2) \in [0, +\infty)$
- Cyber equipment vulnerability:  $v(a_1, \sigma_2(\theta_2), \theta_2)$



# Problem formalization (cont.)

- Objective

- Defender:

$$\max u_1(a_1, \sigma_2) = \sum_{\theta_2 \in \Theta_2} [-C \cdot p(\theta_2) \cdot v(a_1, \sigma_2(\theta_2), \theta_2)] - a_1$$

Diagram annotations for the Defender's utility function:
 

- $u_1$ : Defender utility
- $\theta_2 \in \Theta_2$ : Attacker types
- $p(\theta_2)$ : Attacker type probability
- $C$ : Potential physical loss
- $v(a_1, \sigma_2(\theta_2), \theta_2)$ : Vulnerability
- $a_1$ : Cybersecurity investment

- Attacker:

$$\max u_2(a_1, \sigma_2(\theta_2), \theta_2) = C \cdot v(a_1, \sigma_2(\theta_2), \theta_2) - \sigma_2(\theta_2)$$

Diagram annotations for the Attacker's utility function:
 

- $u_2$ : Attacker utility
- $\theta_2$ : Attacker type
- $C$ : Potential physical loss
- $v(a_1, \sigma_2(\theta_2), \theta_2)$ : Vulnerability
- $\sigma_2(\theta_2)$ : Attack effort



# Bayesian games for cybersecurity investment

- What we just described is actually a Bayesian game
  - The defender has incomplete information on the attacker
  - This incomplete information is described by the various types of attacker and the probability distribution over the types
- We can solve the game using the solution concept of Bayesian Nash equilibrium
  - i.e., obtain the Bayesian Nash equilibrium  $(a_1^*, \sigma_2^*)$  such that
  - For the defender:
$$u_1(a_1^*, \sigma_2^*) \geq u_1(a_1, \sigma_2^*), \forall a_1 \in [0, +\infty)$$
  - For the attacker of any type:
$$u_2(a_1^*, \sigma_2^*(\theta_2), \theta_2) \geq u_2(a_1^*, \sigma_2(\theta_2), \theta_2), \forall \theta_2 \in \Theta_2, \forall \sigma_2(\theta_2) \in [0, +\infty)$$





Obtain the Bayesian Nash equilibrium

- Obtain and solve the following system of partial differential equations

$$\frac{\partial u_1(a_1, \sigma_2)}{\partial a_1} = \frac{\partial [\sum_{\theta_2 \in \Theta_2} [-C \cdot p(\theta_2) \cdot v(a_1, \sigma_2(\theta_2), \theta_2)] - a_1]}{\partial a_1} = 0$$

$$\frac{\partial u_2(a_1, \sigma_2(\theta_2), \theta_2)}{\partial \sigma_2(\theta_2)} = \frac{\partial [C \cdot v(a_1, \sigma_2(\theta_2), \theta_2) - \sigma_2(\theta_2)]}{\partial \sigma_2(\theta_2)} = 0, \forall \theta_2 \in \Theta_2$$



# Numerical case study

- Two types of attacker:
  - One with high capability ( $\theta_2 = H$ )
  - The other with low capability ( $\theta_2 = L$ )
- Cyber equipment vulnerability
  - For  $\theta_2 = H$

$$v(a_1, \sigma_2(H), H) = \frac{\sigma_2(H)}{\alpha_H(a_1 + \sigma_2(H) + \beta)}$$

$$\alpha_H \geq 1 \text{ and } \beta > 0$$

- For  $\theta_2 = L$

$$v(a_1, \sigma_2(L), L) = \frac{\sigma_2(L)}{\alpha_L(a_1 + \sigma_2(L) + \beta)}$$

$$\alpha_L > 1 \text{ and } \alpha_L > \alpha_H$$



## Numerical case study (cont.)

- Additional information about the problem:

Parameters (unit)	Parameter description	Nominal Value	Value Range
$C$ (in USD)	Potential physical loss	1000	[0, 2000]
$p(H)$ (unitless)	Belief in $\theta_2 = H$	0.6	[0, 1]
$\alpha_H$ (unitless)	Parameter defining vulnerability for $\theta_2 = H$	5	[1, 10)
$\alpha_L$ (unitless)	Parameter defining vulnerability for $\theta_2 = L$	10	(5, 20]
$\beta$ (in USD)	Parameter defining vulnerability	5	[1, 10]

- For each parameter setting, we can obtain the corresponding  $(a_1^*, \sigma_2^*)$



## Results for the setting with nominal parameter values

Parameters (unit)	Parameter description	Nominal Value	Value Range
$C$ (in USD)	Potential physical loss	1000	[0, 2000]
$p(H)$ (unitless)	Belief in $\theta_2 = H$	0.6	[0, 1]
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$\beta$ (in USD)	Parameter defining vulnerability	5	[1, 10]

- Bayesian Nash equilibrium

$$(a_1^* = 33.97 \text{ USD}, \sigma_2^*(H) = 49.31 \text{ USD}, \sigma_2^*(L) = 23.46 \text{ USD})$$

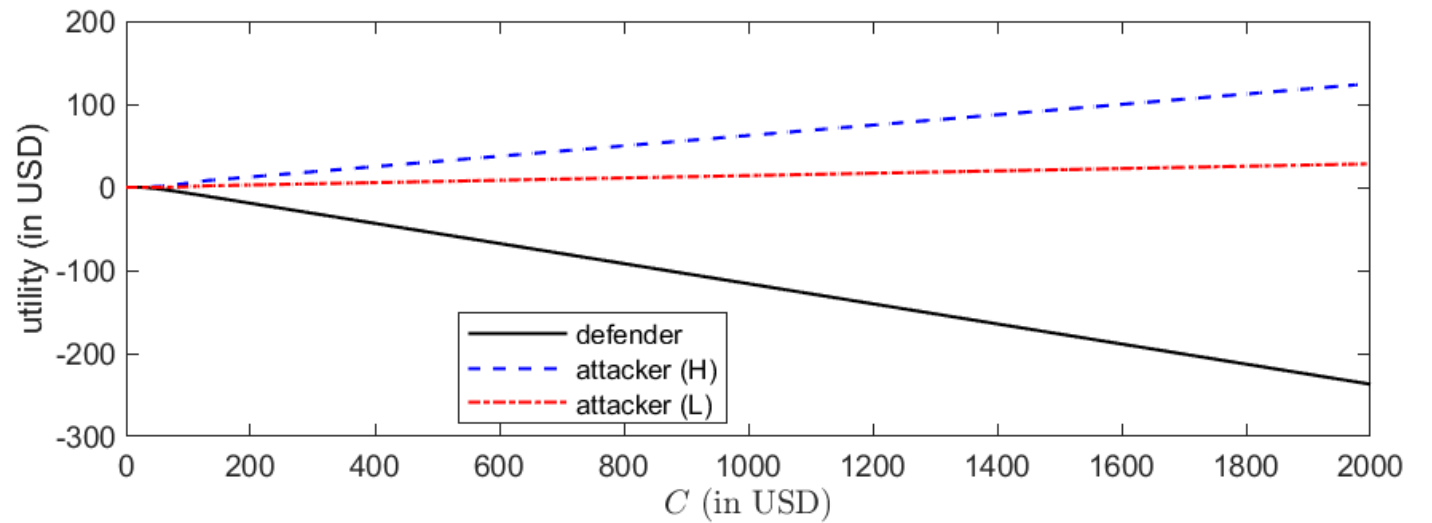
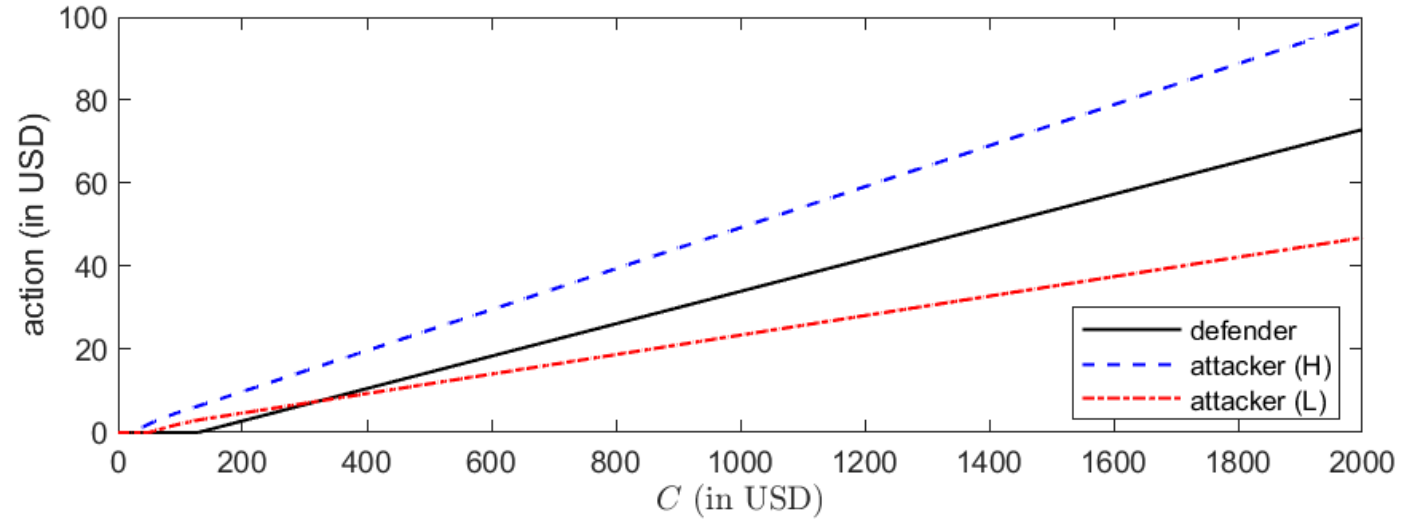
- Utility

- Defender:  $u_1(a_1^*, \sigma_2^*) = -116.03 \text{ USD}$
- Attacker of type  $H$ :  $u_2(a_1^*, \sigma_2^*(H), H) = 62.40 \text{ USD}$
- Attacker of type  $L$ :  $u_2(a_1^*, \sigma_2^*(L), L) = 14.12 \text{ USD}$



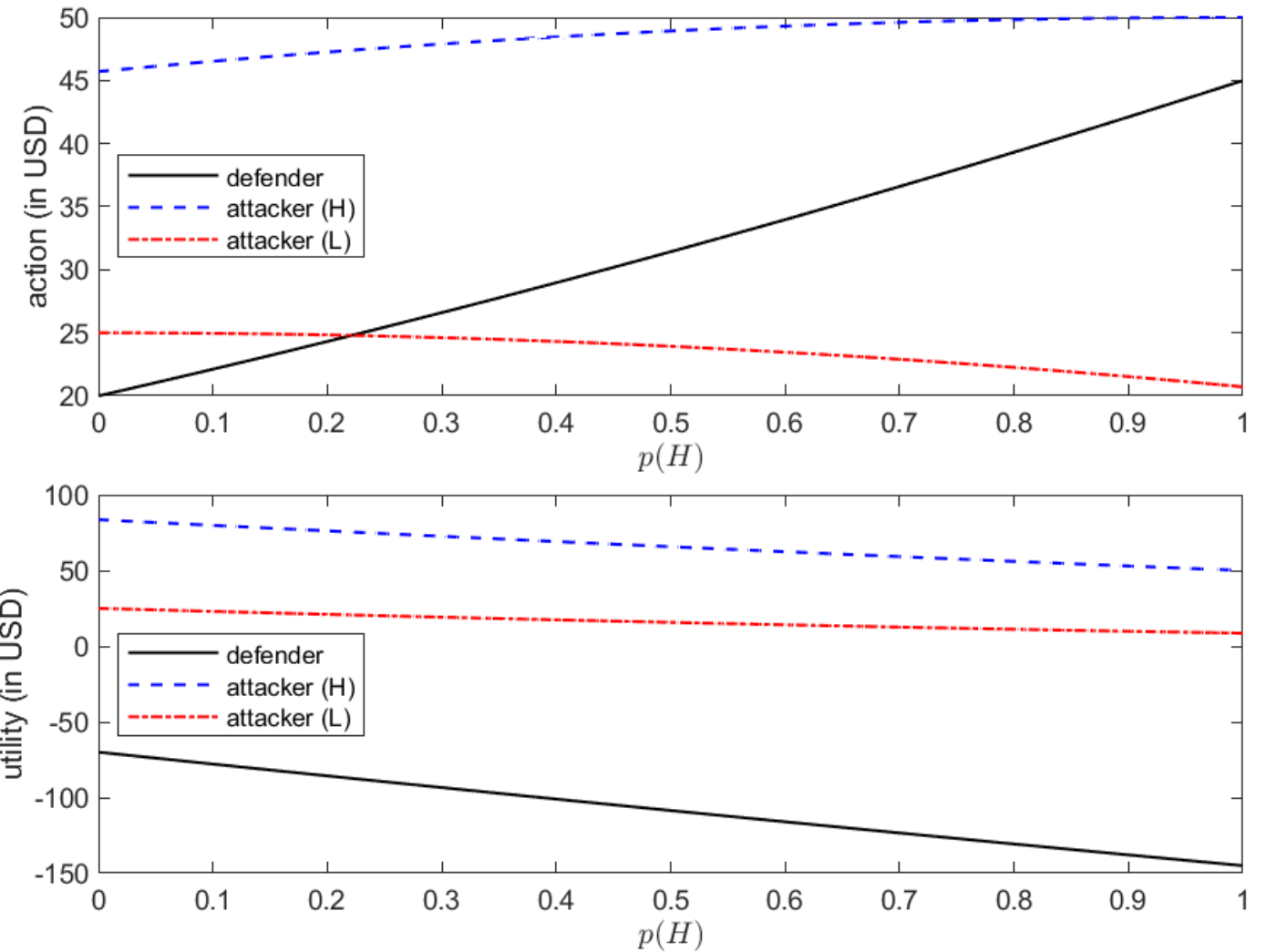
# Sensitivity of results to certain parameter values

- The effect of  $C$  on the outcome



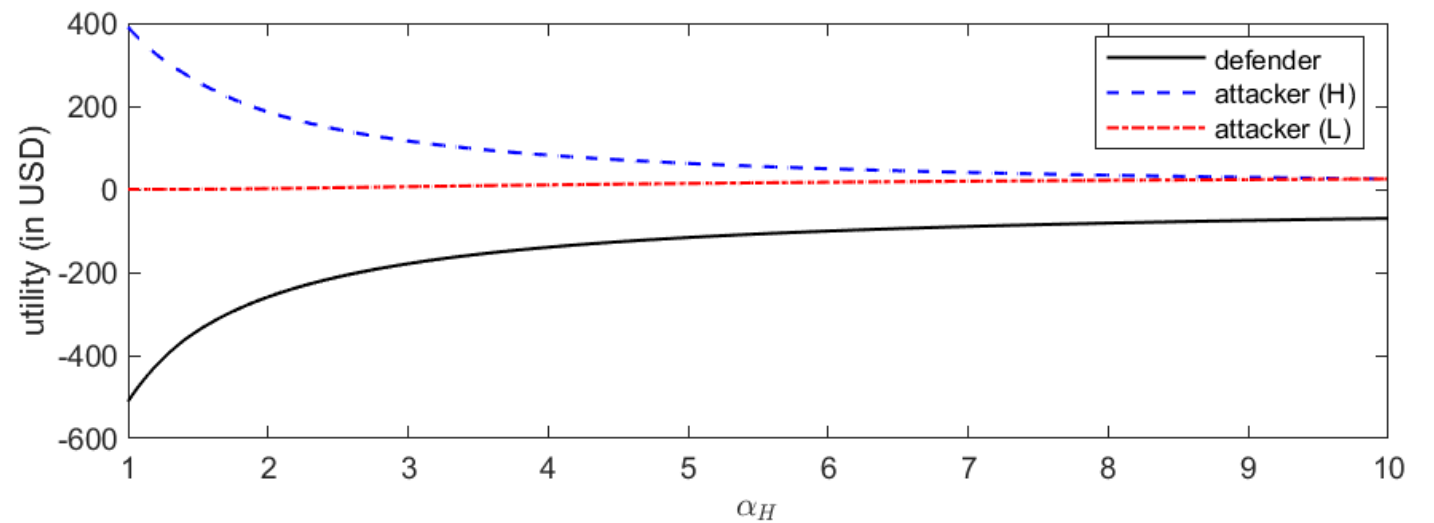
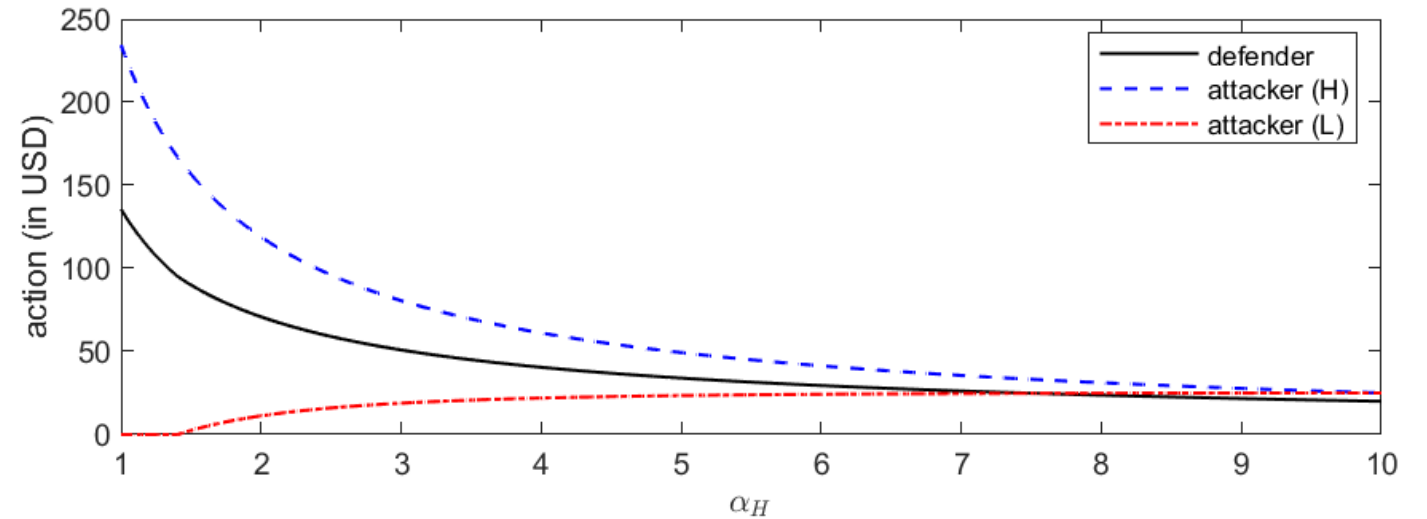
## Sensitivity of results to certain parameter values (cont.)

- The effect of  $p(H)$  on the outcome



# Sensitivity of results to certain parameter values (cont.)

- The effect of  $\alpha_H$  on the outcome



# Summary and future work

- Cybersecurity investment
  - Defender decision-making while considering the level of attacker effort
  - Incomplete information on the attacker
  - Bayesian games for modeling and solving the cybersecurity investment problem
- Numerical example
  - The outcome for a setting with nominal parameter values
  - Sensitivity of the outcome to various model parameters
- Future work
  - Multiple defenders, multiple attackers
  - Determination of model parameters





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Thank you!

