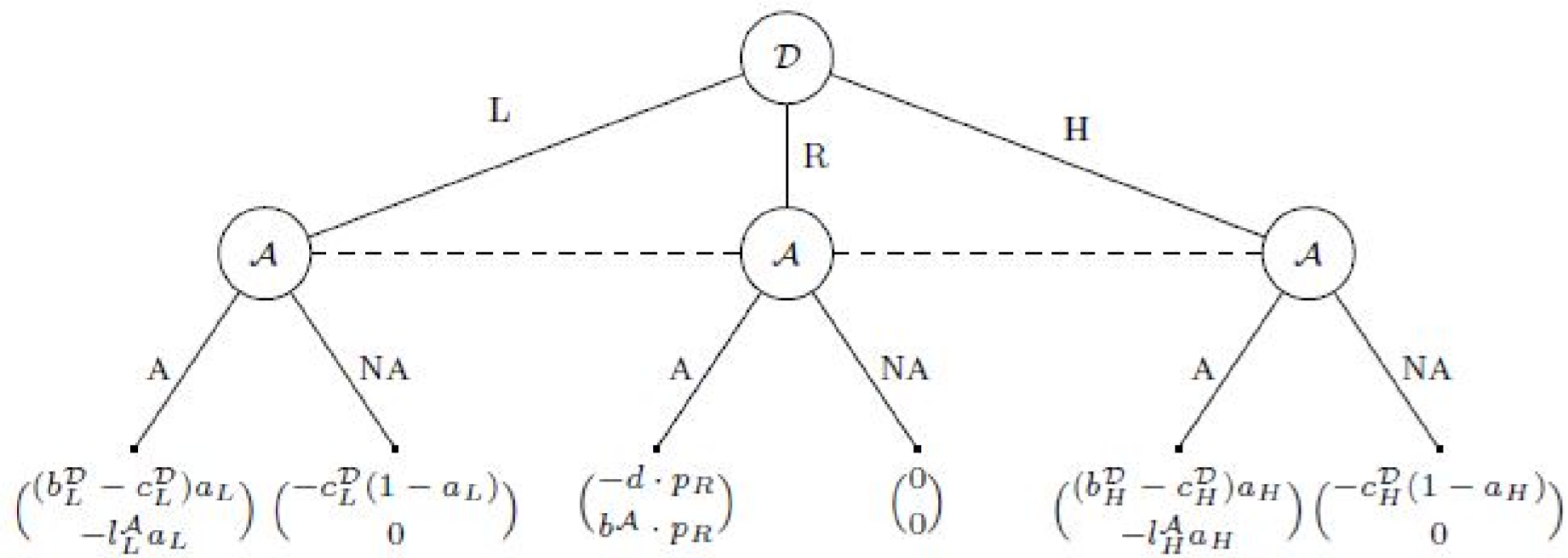


THE MODEL



- Single-shot Bayesian game with complete but imperfect information.
- Defender decides to install a system with high-interaction honeypot (H), or low-interaction honeypot (L), or no honeypot (R); with each having its costs and benefits.
- Attacker deciding whether to attack a target in the presence of information asymmetry.
- motivated from [1] and [3] → refined strategies to include L, H and R rather than just honeypot and normal system.

NOVELTY

L, H and R has efficacy ($a_L < a_H < p_R$) which reflects a system's **probability of being recognised as a real system** during reconnaissance.

ASSUMPTIONS AND PAYOFFS

- type-L and type-H systems have additional costs and benefits to type-R system.
- The aggregated cost includes the deployment, maintenance and operational costs of having a honeypot in the network.
- type-H system → higher threat intelligence but expensive.
- Attacker, similar to the defender, has loss and benefits ($b^A > -l_L^A > -l_H^A$) based on her choice of action.

LIST OF SYMBOLS

Symbols	Condition/Range	Description
b^A	$b^A > 0$	Attacker's benefit on attacking type-R system
b_H^D	$b_H^D \geq c_H^D$	Defender's benefit when type-H system attacked
b_L^D	$c_L^D \leq b_L^D < b_H^D$	Defender's benefit when type-L system attacked
c_H^D	$c_H^D > 0$	Cost of running type-H system
c_L^D	$0 < c_L^D < c_H^D$	Cost of running type-L system
d	$d > b_H^D$	Defender's loss when type-R system attacked
l_H^A	$l_H^A > 0$	Attacker's loss on attacking type-H system
l_L^A	$0 < l_L^A < l_H^A$	Attacker's loss on attacking type-L system

SOLUTION: BAYESIAN EQUILIBRIA

	$\mathcal{U}^D(L, NA) < \mathcal{U}^D(H, NA)$	$\mathcal{U}^D(L, NA) \geq \mathcal{U}^D(H, NA)$
$\mathcal{U}^D(L, A) \leq \mathcal{U}^D(H, A)$	(H, A; $p_2 \geq \bar{p}_2$) (R, NA; $p_2 < \bar{p}_2$)	(L, A; $p_1 \geq \bar{p}_1$) (R, NA; $p_1 < \bar{p}_1$) (H, A; $p_2 \geq \bar{p}_2$) (R, NA; $p_2 < \bar{p}_2$)
$\mathcal{U}^D(L, A) > \mathcal{U}^D(H, A)$	(L, A; $p_1 \geq \bar{p}_1$) (R, NA; $p_1 < \bar{p}_1$) (H, A; $p_2 \geq \bar{p}_2$) (R, NA; $p_2 < \bar{p}_2$)	(L, A; $p_1 \geq \bar{p}_1$) (R, NA; $p_1 < \bar{p}_1$)

where $\bar{p}_1 = \frac{a_L \cdot l_L^A}{p_R \cdot b^A + a_L \cdot l_L^A}$ and $\bar{p}_2 = \frac{a_H \cdot l_H^A}{p_R \cdot b^A + a_H \cdot l_H^A}$

REMARKS AND OUTLOOK

- Game-theoretic approach gives better payoff than randomly choosing system type to implement.
- Our first step towards implementing game-theoretic strategies in smart grid networks as a part of the **H2020 SPEAR project**.
- [2] considers a *decoy parameter* for honeypots which could be *conceived as the efficacy* for each type of system in our model.
- Various extensions are possible:
 - repeated game model with belief update schemes,
 - model with sophisticated attacker (e.g. with anti-honeypot techniques [4]).

References

- [1] Thomas E Carroll and Daniel Grosu. A game theoretic investigation of deception in network security. *Security and Communication Networks*, 4(10):1162–1172, 2011.
- [2] Yang Li, Leyi Shi, and Haijie Feng. A game-theoretic analysis for distributed honeypots. *Future Internet*, 11(3):65, 2019.
- [3] Jeffrey Pawlick and Quanyan Zhu. Deception by design: evidence-based signaling games for network defense. *arXiv preprint arXiv:1503.05458*, 2015.
- [4] Kun Wang, Miao Du, Sabita Maharjan, and Yanfei Sun. Strategic honeypot game model for distributed denial of service attacks in the smart grid. *IEEE Transactions on Smart Grid*, 8(5):2474–2482, 2017.