



Carnegie Mellon University

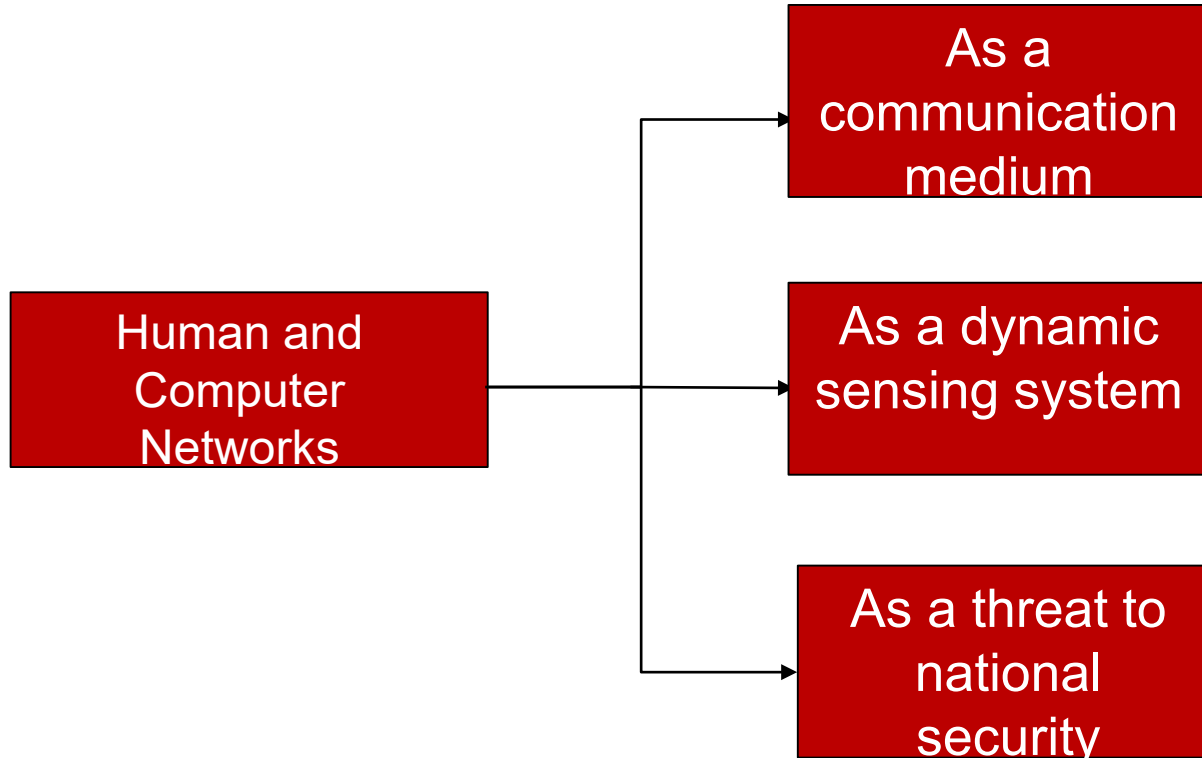
CyLab's Partners Conference

Deep Learning for Enhancing the Robustness of Intrusion Detection Systems

Conrad S. Tucker

Arthur Hamerschlag Career Development Professor,
Mechanical Engineering & Machine Learning (Courtesy)

Ascertaining the Veracity and Security of Data in the Information Age

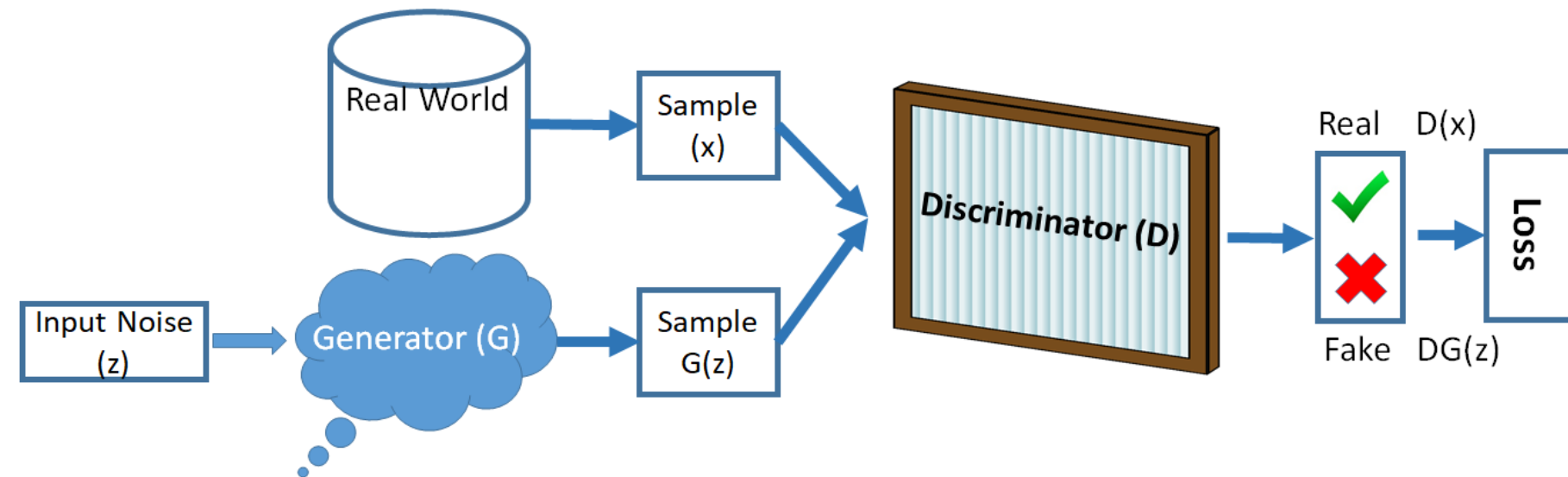


Human Perception and Classification

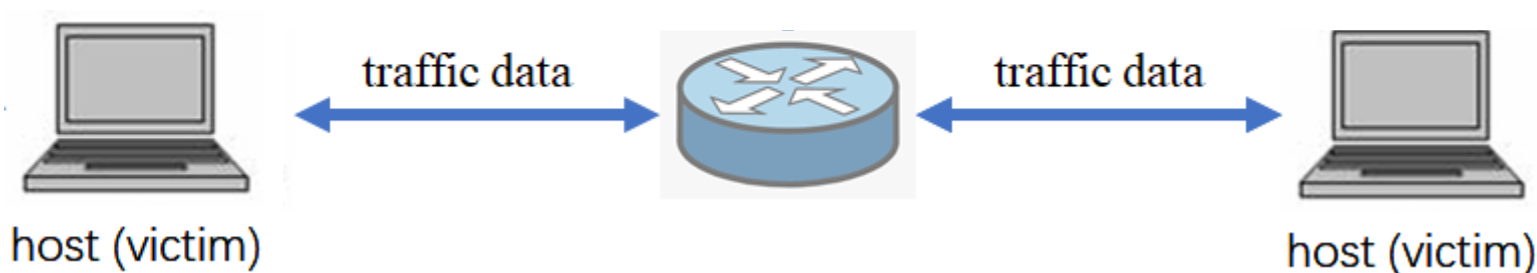


Generative Adversarial Networks (GANs)

(Goodfellow, et al. 2014)

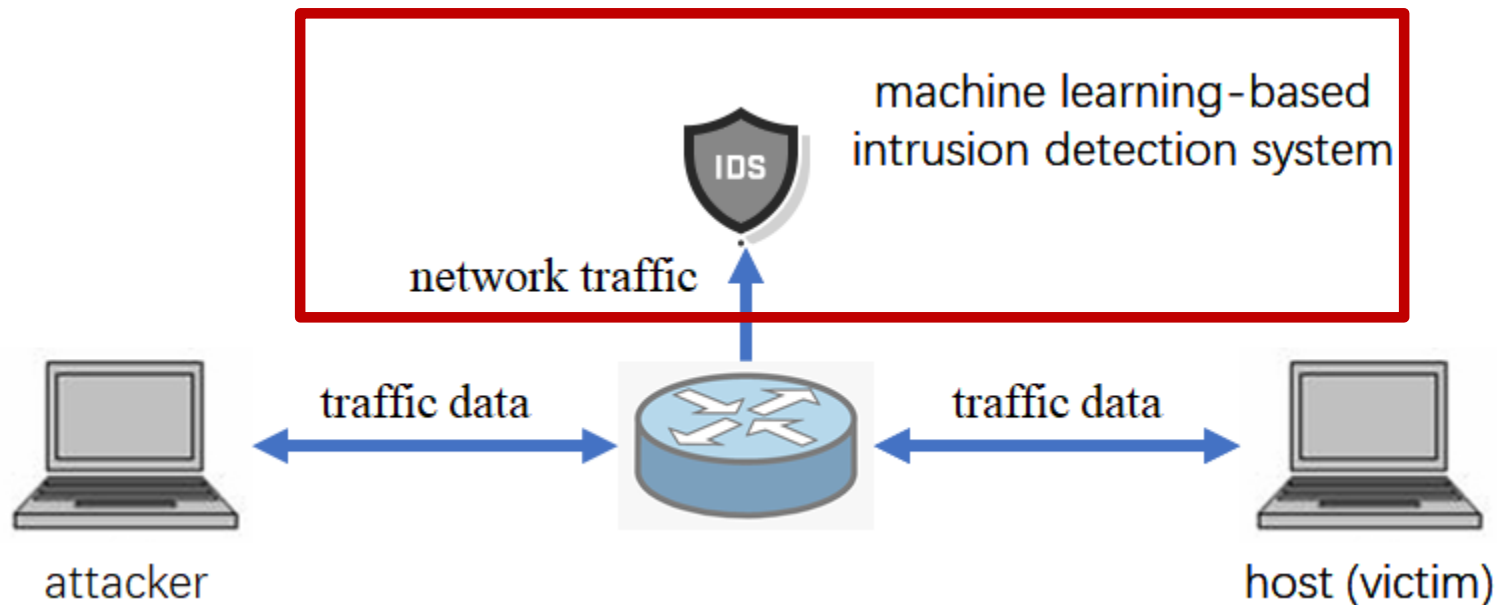


Computer Networks form the Backbone of Modern-Day Communication Systems



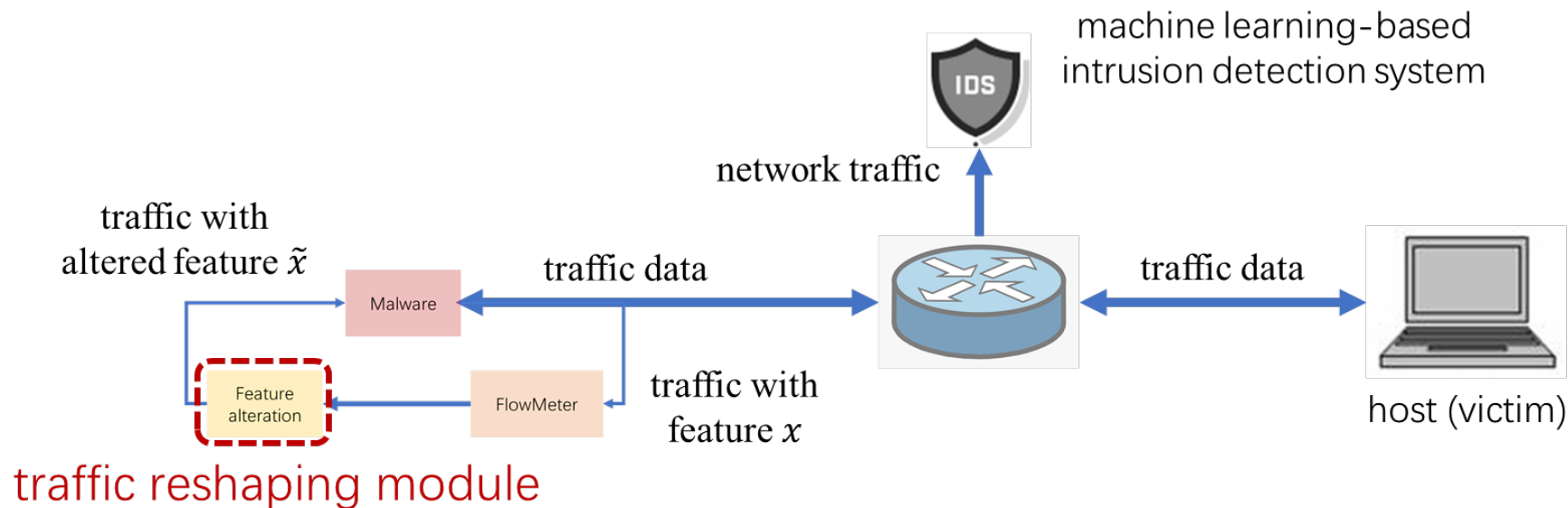
Shu, D., Cong, W., Chai, J., & **Tucker, C. S.** (2020, July). Encrypted rich-data steganography using generative adversarial networks. In Proceedings of the 2nd ACM Workshop on Wireless Security and Machine Learning (pp. 55-60)

Computer Networks are also Susceptible to AI-Based Attacks



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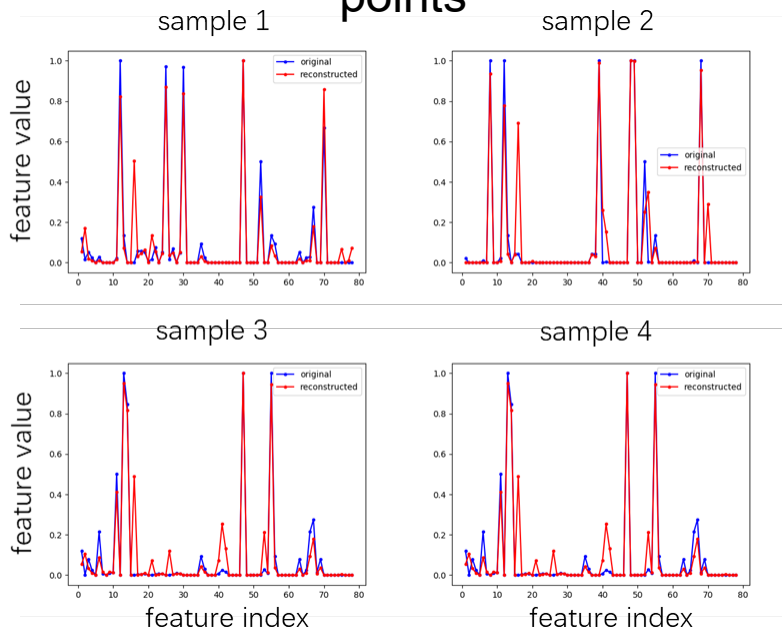
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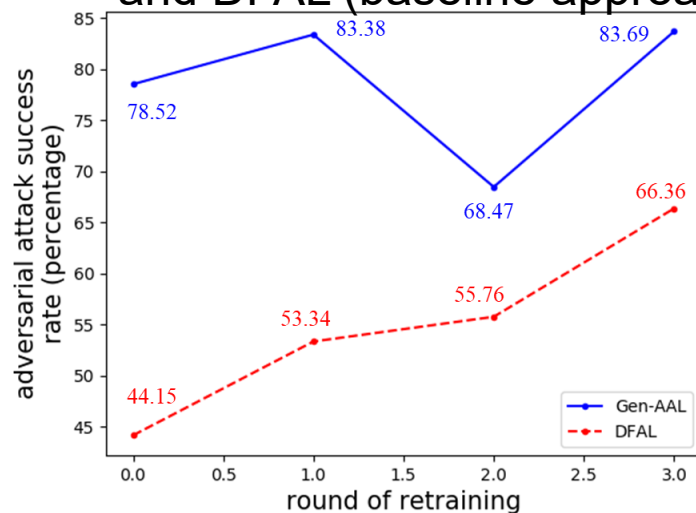
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Generative Adversarial Active Learning (Gen-AAL)

original and adversarial feature points



attack success rate of Gen-AAL and DFAL (baseline approach)



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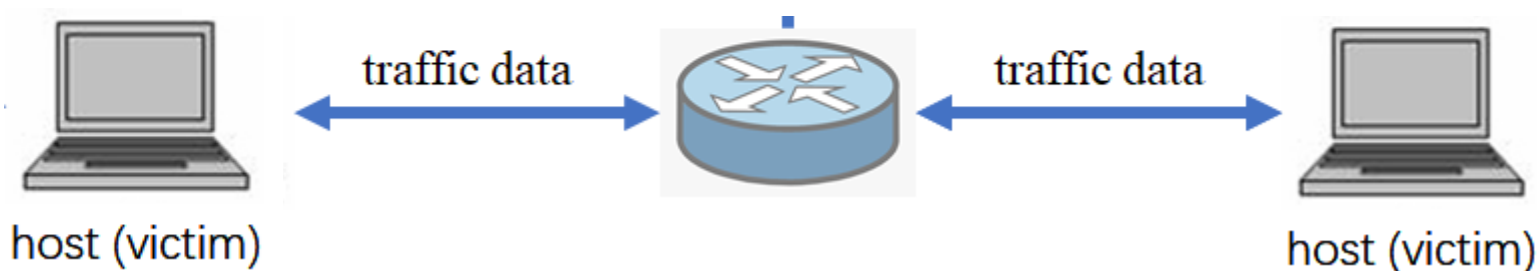
Perturbed Features that Successfully Compromise IDS

Features with largest perturbations

Feature name

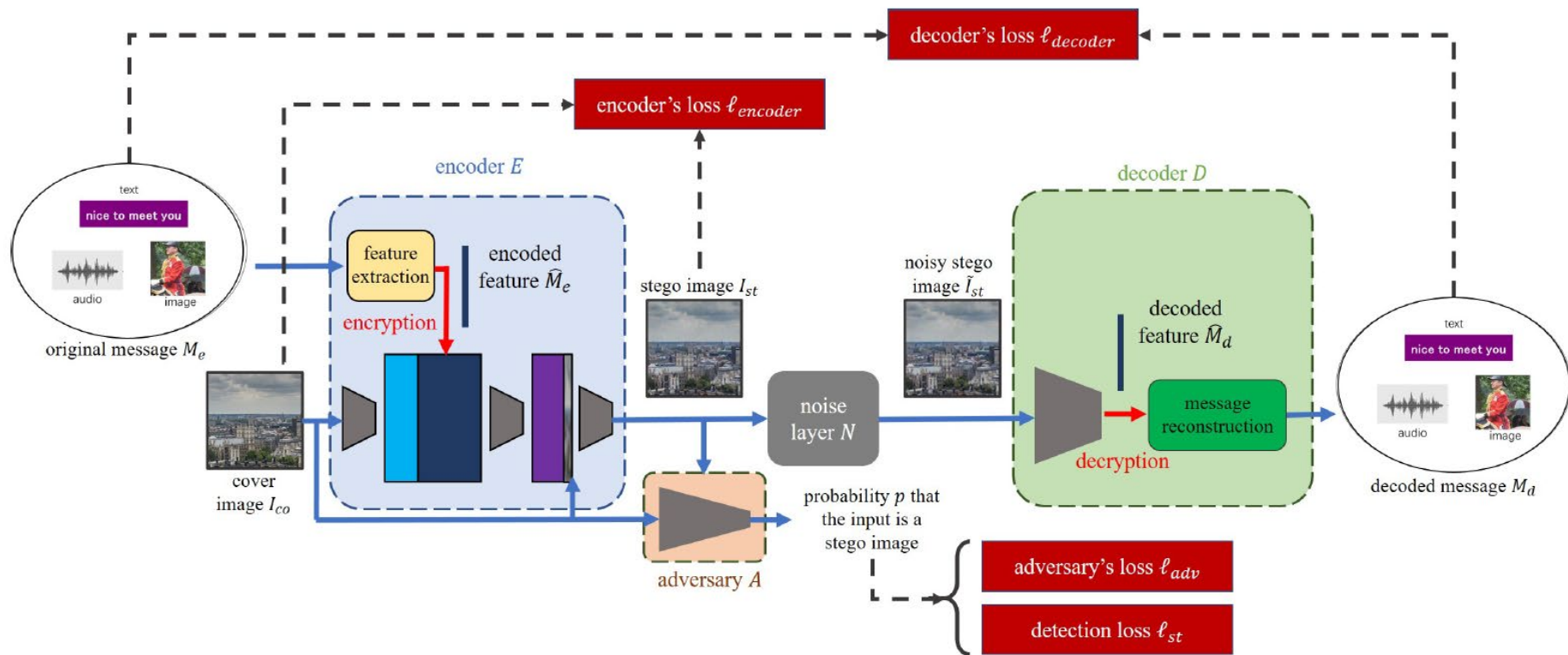
| | |
|---------------------------|--|
| 'Flow IAT Std' | Standard deviation time between two packets sent in the flow |
| 'Flow IAT Min' | Minimum time between two packets sent in the forward direction |
| 'Flow IAT Mean' | Mean time between two packets sent in the forward direction |
| 'Fwd IAT Std' | Standard deviation time between two packets sent in the forward direction |
| 'Flow Packets/s' | Number of flow packets per second |
| 'Fwd Packet Length Max' | Maximum size of packet in forward direction |
| 'Init_Win_bytes_backward' | The total number of bytes sent in initial window in the backward direction |
| 'Destination Port' | Destination port |
| 'min_seg_size_forward' | Minimum segment size observed in the forward direction |
| 'Packet Length Variance' | Variance length of a packet |

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AI-Encoding of Messages within Network Communications¹²

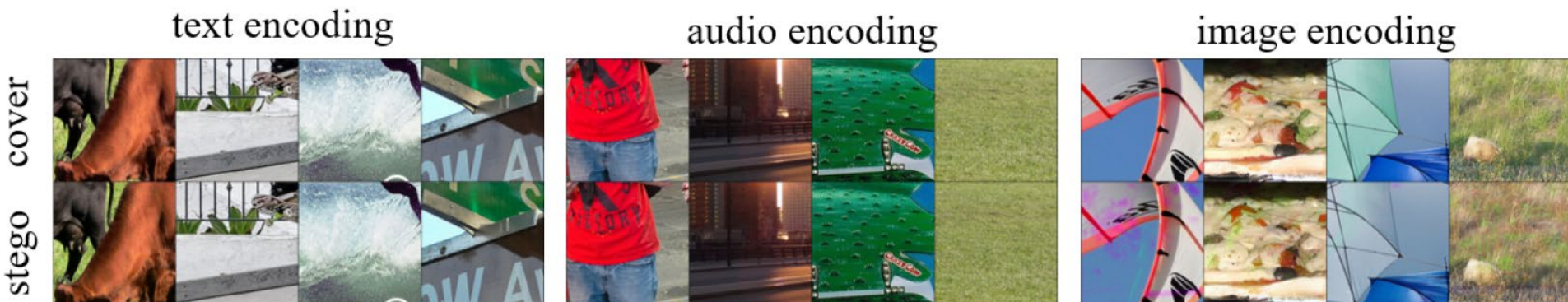


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Results

AI-Encoding of Messages within Network Communications

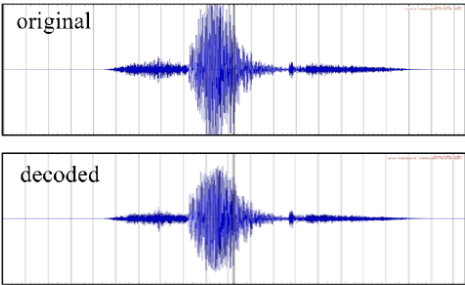



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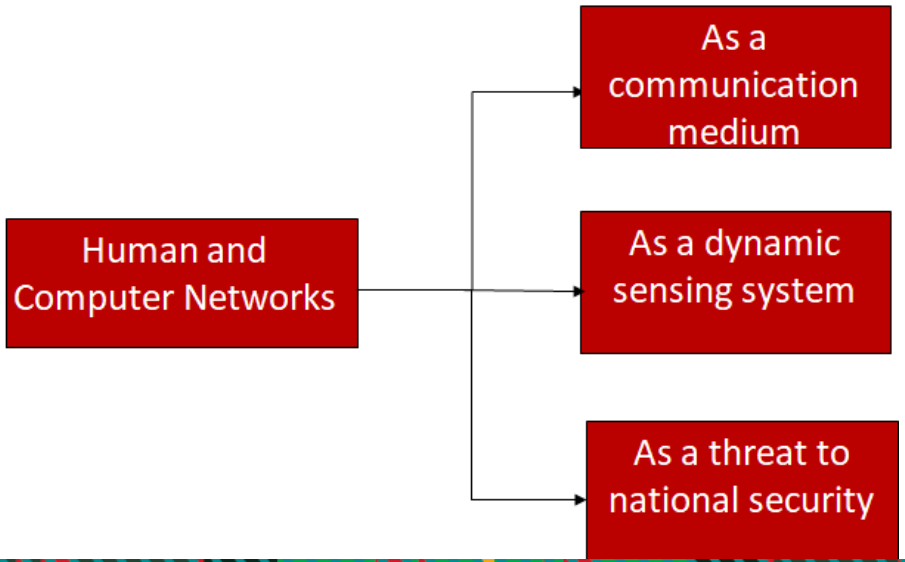
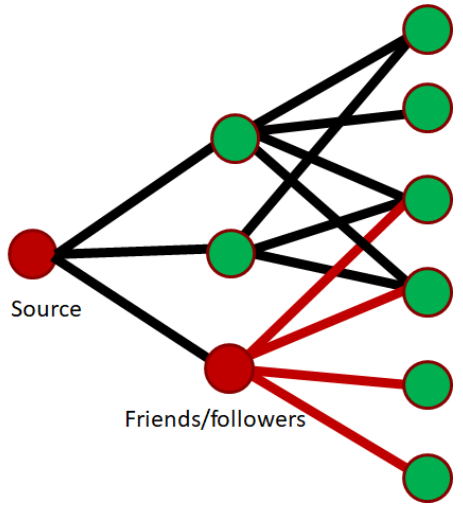
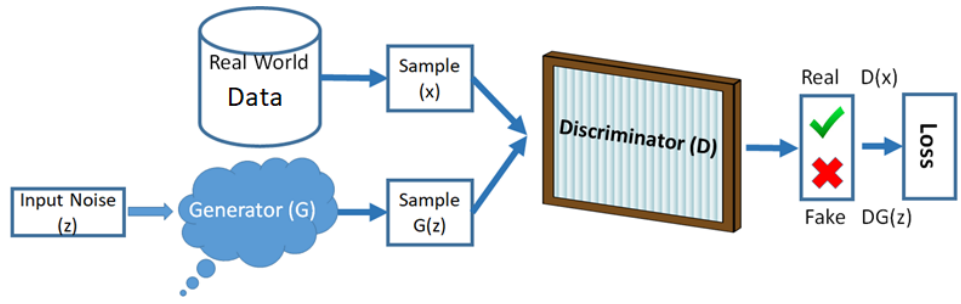
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Results

AI-Encoding of Messages within Network Communications

| Message type | Decoding error | Example of messages |
|--------------|----------------|---|
| text | 0.017 | <pre> predict : <start> a beautiful woman sitting on a bench next to a body of water . <end> ground truth: <start> a beautiful woman sitting on a bench next to a body of water . <end> predict : <start> a black bird eating an apple on the ground in the woods . <end> woven ground truth: <start> a black bird eating an apple on the ground in the woods . <end> <pad> predict : <start> a man with a bucket hat riding a hose on a beach . <end> woven ground truth: <start> a man with a bucket hat riding a hose on a beach . <end> <pad> predict : <start> a big bird with a huge coupons looks out of its cage <end> woven woven ground truth: <start> a big bird with a huge beak looks out of its cage <end> <pad> <pad> predict : <start> a dog with a leash on is sitting near a park bench <end> woven woven ground truth: <start> a dog with a leash on is sitting near a park bench <end> <pad> <pad> </pre> |
| audio | 0.0003 |  |
| image | 0.0057 |  |

Research Summary Path Forward



Research Team and Collaborators



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Research interests:

vehicular security, machine learning, and intrusion detection methods and techniques to promote cyber resilience and foster research on autonomous active cyber defense

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Sponsors/Collaborators

