



Active Learning Model Extraction Detection with LSTM

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Introduction

- Model extraction
 - Attackers reverse engineer publicly accessible models
 - Theft of intellectual property
- Types of model extraction
 - Passive learning uses a large set of queries, and is often impractical
 - Active learning algorithmically selects queries for efficiency
- Defending against model extraction
 - Randomized outputs harm model accuracy and are not always effective
 - Detection algorithms are essential to provide security while maintaining performance



Detecting Model Extraction

- Existing approaches
 - PRADA and VarDetect examine the distribution of a user's queries [1][2]
 - These approaches are generic and effective
 - SEAT examines the similarity between queries [3]
 - Model extraction attacks may generate similar queries or perturb existing ones
 - HODA examines the “hardness” of a user's queries [4]
 - This is defined as the number of epochs taken for the target model to learn the correct classification of an input
- HODA showed that extraction techniques are interested in “harder” data points
 - As the extracted model becomes more accurate, its definition of a “hard” data point changes



Algorithm

- Implied sequential pattern
 - HODA's observations on "hardness"
 - Active learning involves using the results of previous queries to select subsequent queries
- LSTM-RNN (Long Short Term Memory Recurrent Neural Network) with MLP for sequence classification
 - Can process longer sequences than traditional RNNs
 - Have been used for security in the past [6]
 - May be able to learn the patterns indicative of active learning
- Output from target model included in input to detection model
 - This information is used by active learners to select subsequent queries
 - It is reasonable to assume that the "hardness" of a query is related to the uncertainty of its classification by the target model, which is reflected in the raw output of the target model



Experiments

- Target model is a CNN for classifying digits from the MNIST dataset
 - Classifies 0s from 1s
 - Used in the past to evaluate model extraction [1][2]
- Benign queries
 - Problem domain (0s and 1s from the dataset)
 - Alternative Problem Domain (Digits other than 0 and 1 from the dataset)
 - These model the behavior of a benign user [2]
- Malicious queries
 - Active learning heuristic for training and evaluation [5]
 - Made without access to training data
 - Synthetically generated data sets for evaluation
 - Used in the past to evaluate model extraction [1][2][3][4]
 - Made with access to training data
- Compare false negative and false positive rate with PRADA and SEAT



Results

- PRADA reports a 0% false negative and false positive rate [1]
- SEAT reports a 0.05% false positive rate [3]
- LSTM-RNN
 - 14% false negative rate
 - 32% false positive rate
 - Some success detecting sequential patterns



Future Directions

- Active learners may interleave benign or misleading queries
 - Randomly spaced sampling
 - Practicality of these attacks
- Experiments on more complex active learning attacks
- Longer or shorter sequences for the LSTM may increase accuracy



Conclusions

- Model extraction techniques are essential for publicly accessible models
- Active learning model extraction implies a sequential pattern
- LSTM-RNN shows some promise in detecting active learning
 - Nowhere near as effective as state of the art approaches such as PRADA and SEAT



Questions?



References

[1] (PRADA) <https://arxiv.org/pdf/1805.02628.pdf>

[2] (VarDetect) <https://arxiv.org/pdf/2107.05166.pdf>

[3] (SEAT) <https://dl.acm.org/doi/pdf/10.1145/3474369.3486863>

[4] (HODA)
https://www.researchgate.net/profile/Amir-Mahdi-Sadeghzadeh/publication/353071516_HODA_Hardness-Oriented_Detection_of_Model_Extraction_Attacks/links/621a1b8a579f1c04171b66a9/HODAHardness-Oriented-Detection-of-Model-Extraction-Attacks.pdf

[5] (Adaptive Retraining) https://www.usenix.org/system/files/conference/usenixsecurity16/sec16_paper_tramer.pdf

[6] (LSTM in Security)
<http://text2fa.ir/wp-content/uploads/Text2fa.ir-An-effective-network-attackdetection-method-based-on-kernel-PCA-and-LSTM-1.pdf>