Trojan AI Detection



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Meet the Team



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AI Adoption and Trust

Al adoption is rapidly accelerating

What steps can you take to assure customers and stakeholders that your Al models are trustworthy?



What is a Trojan AI?



"The supreme art of war is to subdue the enemy without fighting." — Sun Tzu

How do we create a classification AI model?



How do we create a classification model?



New Data

Deployed Classifier

What is a Trojan AI?



Image Classification

Trained to recognize stop sign

Post-it note trained as trigger

Trojan AI recognizes post-it stop sign as speed limit sign



Trojan Demonstration



Compromised AI

Reliance on AI models leaves systems vulnerable to new attacks

Training Pipeline

The clearest defense against a Trojan attack is to secure the training data. However, in many cases, this is not possible

Acquired Al

Supplier model may be malicious or compromised

Transfer Learning

Many Als are created through transfer learning - taking an existing Al and modifying the model for a new use case. Trojans can persist in an Al after transfer learning

Transit

Al may be modified directly while stored or in transit to the endpoint

Detect Trojan Attacks

Various attack vectors makes it difficult to prevent an attack with confidence

This project detects if an AI is compromised regardless of attack origin





Opportunity

"Opportunity is missed by most people because it is dressed in overalls and looks like work."

— Thomas A. Edison

Opportunity

Utilize a Trojan AI Detector to ensure that an AI model can be safely deployed as part of the AI development lifecycle



Trojan Detectors Need Data

Use a synthetic dataset designed to simulate trojan AI models.

Primary factors:

- 1. Al model architecture
- 2. Size of the trigger. The percentage of the foreground image area the trigger occupies
- 3. Trigger strength: The percentage of the images in the target class which are poisoned







Trojan Detectors Need Data

- **Data** Trained image classification AI models
- Architectures: Inception-v3, DenseNet-121, and ResNet50
- Model Training Synthetically created image data of non-real traffic signs superimposed on road background scenes
- Labels Half of the models in the training data have been poisoned and labeled for training



A few examples of how the robustness factors manifest in the actual images used to train the AI models can be seen in the figure above, where one type of sign has been composited into several different background with a variety of transformations applied.

https://pages.nist.gov/trojai/docs/round0.html#round0

Detector Development

- Leverage cloud storage and GPU accelerated compute instances to support CUDA PyTorch requirements
- **Push** data to the cloud so the large dataset continues to be available for further testing
- **Save** the original data as well as the results to allow for follow-up analysis and reproducibility of the detector
- **Deploy** the trained detector to protect systems from Trojan AI

TrojAl Data Flow



How Will We Detect Trojans?

Trojan models may be detected by training a model to examine the model's weights. This method is fast and low cost, meaning that it may be integrated into a more complex detection strategy.



Flatten and Reduce Training Models



[-0.06324801 0.22248715 -0.2614732 -0.18624003 -0.02487379 0.03491592, 0.02338792 -0.21783966 -0.15246269 -0.26866657...]

[0.03491592 -0.06324801 0.22248715 -0.2614732 -0.18624003 -0.02487379 , 0.02338792 -0.21783966 -0.15246269 -0.26866657...]

[-0.02487379 0.03491592 -0.06324801 0.22248715 -0.2614732 -0.18624003, -0.21783966 -0.15246269 -0.26866657 -0.22248715...]

[0.03491592, 0.02338792 -0.21783966 -0.06324801 0.22248715 -0.2614732 -0.18624003 -0.02487379 -0.15246269 -0.26866657...]

| Mo | | | | | | | |
|----|----|---------|--|--|--|--|--|
| Mo | Mo | | | | | | |
| Mo | Mo | Model_1 | | | | | |
| Mo | Mo | Model_2 | | | | | |
| | Mo | Model_3 | | | | | |
| | | Model_4 | | | | | |



| Model_1 | |
|---------|--|
| Model_2 | |
| Model_3 | |
| Model_4 | |

Feature Reduction

Multilevel reduction to reduce models to the same shape

Kernel PCA for each layer and architecture

ICA on reduced dataset

Many possible reduction parameters



Independent Component Analysis

Independent Component Analysis seeks to decompose multivariate signals into independent signals

For example, separating multiple instruments on the same audio track



Detector Training

An optimized search procedure selects the reduction parameters and hyper parameters

The reduced model weights are used to train a classification model using XGBoost

Hyperparameters are tuned using Optuna



scikit Peorn

XGBoost

Results

Model performance reaches 84% accuracy

Inference duration is 2 sec/model. Top detectors on the TrojAl leaderboard average 400 sec/model



Business Application



"Be a yardstick of quality. Some people aren't used to an environment where excellence is expected."

— Steve Jobs

Trustworthy AI

Safeguarding against adversarial attacks is a necessary step in developing a robust Al strategy



A majority of organizations haven't taken key steps towards trustworthy AI.

52% Not safeguarding data privacy through the entire lifecycle

55% Not monitoring AI across cloud and AI environments

59%

Not guarding against adversarial threats and potential incursions to keep systems healthy

60%

Not tracking data provenance, changes in data and model versions

Trojan AI Detector Prototype

Users identifies model they want to test

The detector returns results

Users can then use this information to determine whether more scrutiny is necessary.

| Running in This model | has a 98% | 1 models probability | of | being | poisoned |
|--------------------------|-----------|-------------------------|----|-------|----------|
| | | | | | |
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Scalability and Robustness

- Architecture agnostic
 - Generate training data for your target models, feed to trainer
- Built-in optimization
 - Model hyperparameter tuning and reduction optimization is embedded into the trainer
- Customization
 - Detector training is configurable to include optimized detector architecture search
- Integration
 - 2 sec/model inference means flexible integration into cyber security strategy



Modularity

The modular solution allows businesses to integrate Trojan Detection into the Al development lifecycle



Trojan Detector Al Demo

| Trojan AI Detection Submit model (.pt) file for testing | | | | | | | | |
|---|---------------------------------|-----------------------------------|---------|--|--|--|--|--|
| | Select a file: Choose File R | No file chosen Si un Inference | ubmit | | | | | |
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"Are you protected from malicious AI?" - ChatGPT



Thank you

"We thought that we had the answers, it was the questions we had wrong." — Bono