Thirty-third Conference on Neural Information Processing Systems



Rethinking Deep Neural Network Ownership Verification: Embedding Passports to Defeat Ambiguity Attacks

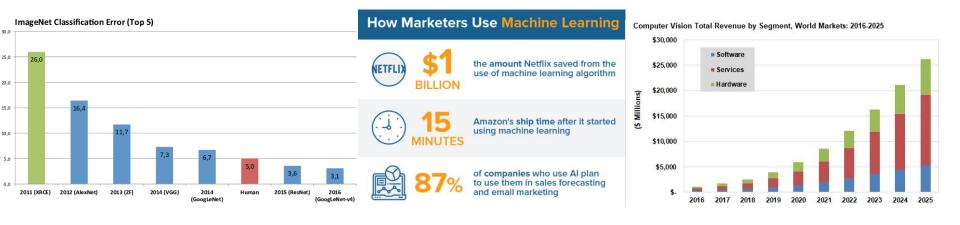
Lixin Fan¹, Kam Woh Ng^{1,2}, Chee Seng Chan²

¹WeBank Al Lab, Shenzhen, China ²Center of Image and Signal Processing, University of Malaya, Kuala Lumpur, Malaysia

{ lixinfan@webank.com; jinhewu@webank.com; cs.chan@um.edu.my }

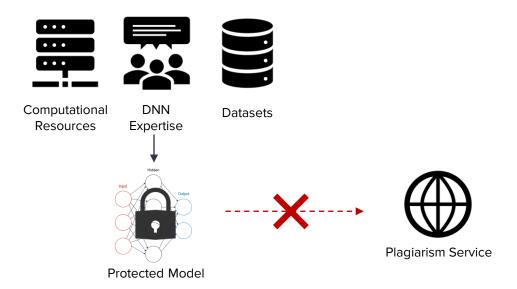
Machine Learning as a Service (MLaaS)

More and more business models and services using Deep Neural Network
 (DNN)



Protection on DNN model is needed!

- Companies invested a lot to create powerful models
- They are easily copied and used by plagiarizers.
- We need a protection on DNN from being illegally copied, distributed and abused.



DNN watermarking methods

1. Feature based approach

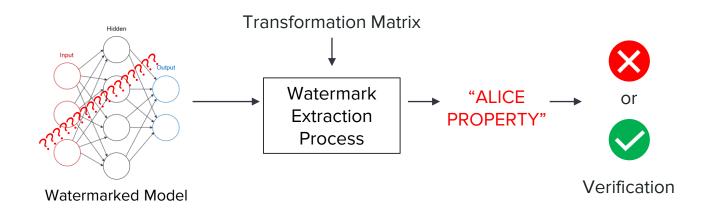
- Y. Uchida, Y. Nagai, S. Sakazawa, and S. Satoh, "Embedding watermarks into deep neural networks"
- B. D. Rouhani, H. Chen, and F. Koushanfar, "Deepsigns: A generic watermarking framework for IP protection of deep learning models"

2. Trigger-set based approach

- Yossi Adi, Carsten Baum, Moustapha Cisse, Benny Pinkas, and Joseph Keshet. "Turning Your
 Weakness Into a Strength: Watermarking Deep Neural Networks by Backdooring"
- Jialong Zhang, Zhongshu Gu, Jiyong Jang, Hui Wu, Marc Ph. Stoecklin, Heqing Huang, and Ian
 Molloy. "Protecting Intellectual Property of Deep Neural Networks with Watermarking"

Feature-based approach

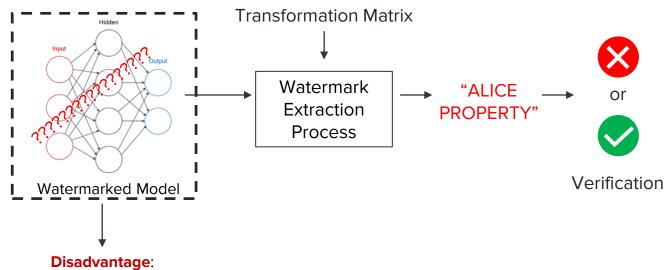
Feature based watermark detection



^{**}Embedding process is identical

Feature-based approach

Feature based watermark detection

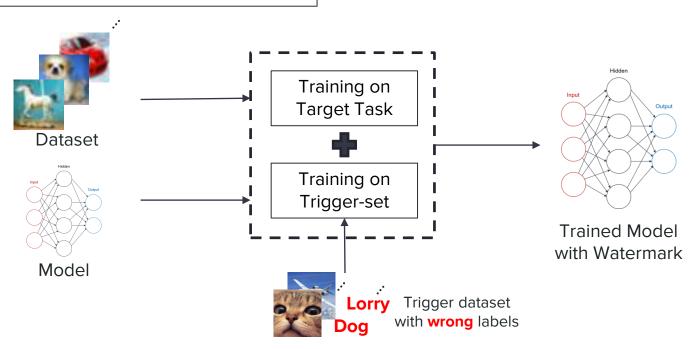


White-box ownership verification:

Need to access weights

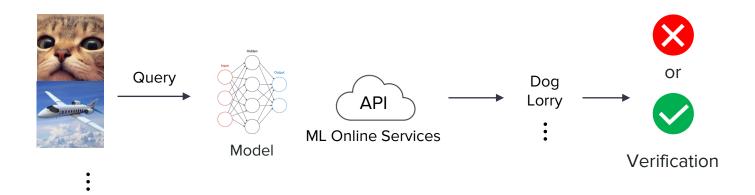
Trigger-set based approach

Trigger-set based watermark embedding



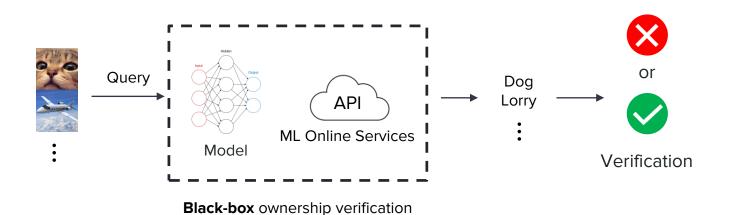
Trigger-set based approach

Trigger-set based watermark detection

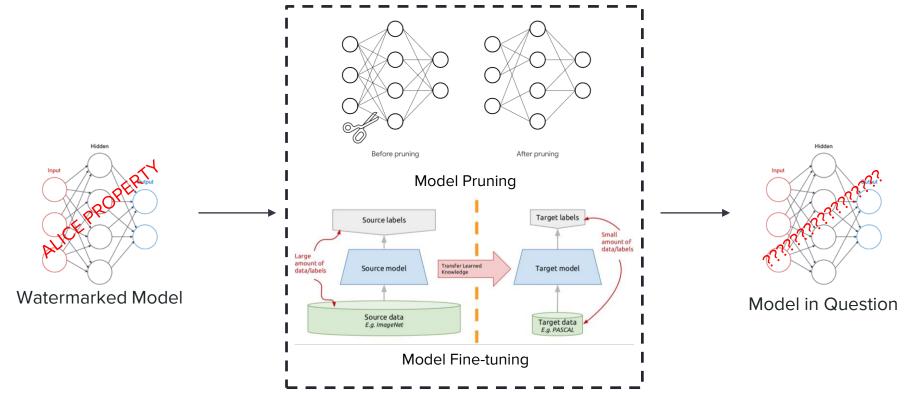


Trigger-set based approach

Trigger-set based watermark detection



Possible attacks to Ownership Protection



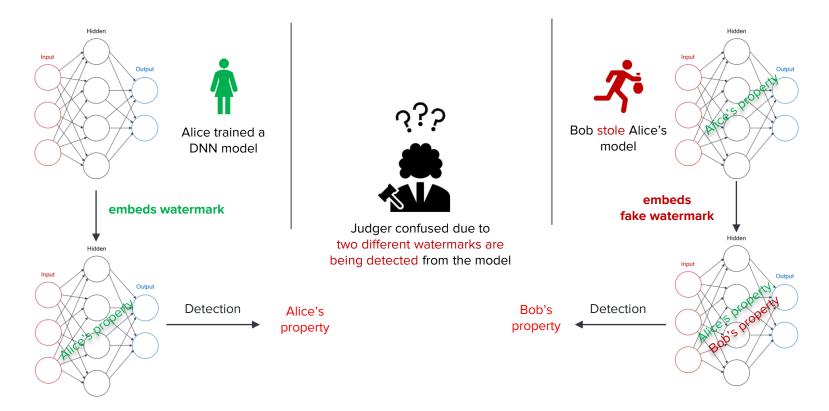
Removal Attack

Effectiveness of Removal Attacks

Watermark embedded in AlexNet for CIFAR10 classification

Removal Attacks	Feature based watermarking [1]	Trigger-set based watermarking [2]
Model Pruning	Strong (100% watermark detected with 65% pruning rate)	Strong (100% watermark detected with 70% pruning rate)
Fine-tuning (CIFAR10 → CIFAR100)	Strong (100% watermark detected after fine-tuning)	Weak (25% watermark detected after fine-tuning)

What is Ambiguity Attack



Effectiveness of Ambiguity Attack

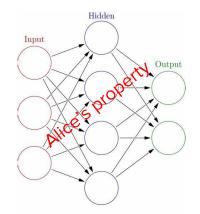
Watermark embedded in AlexNet for CIFAR10 classification

Watermark approach	Real Watermark	Fake Watermark	
Feature based (White-box)	100% watermark detected	100% watermark detected	
Trigger-set based (Black-box)	100% watermark detected	100% watermark detected	

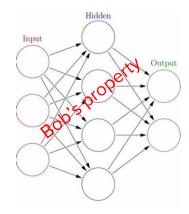
Watermark detection rate for both real and fake watermarks

How to deal with ambiguity attack?

Current Situation

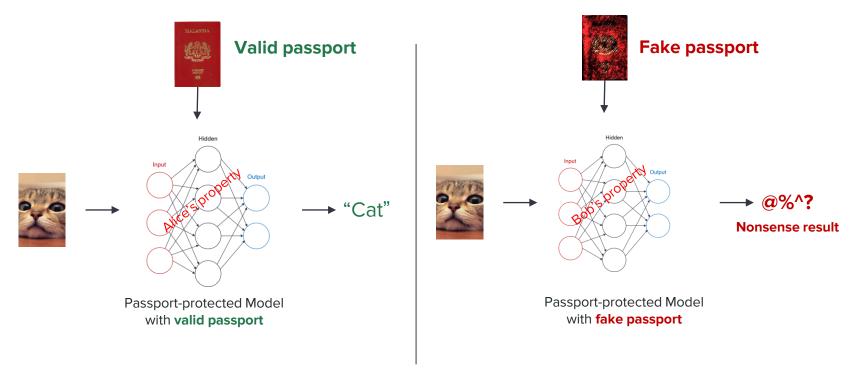


Passport-protected Model with original watermark



Copied Model with fake watermark

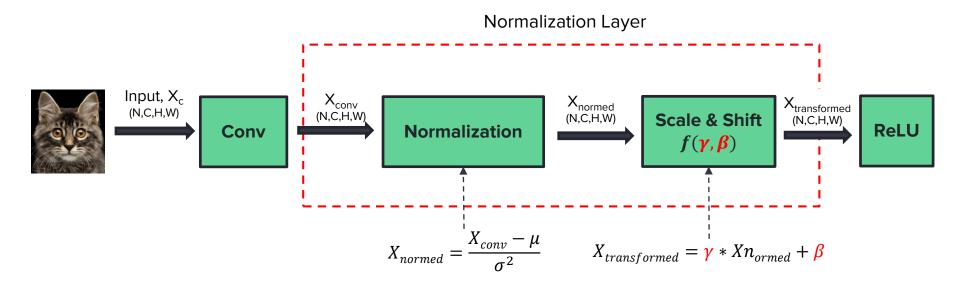
Proposed Solution

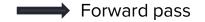


Aim:

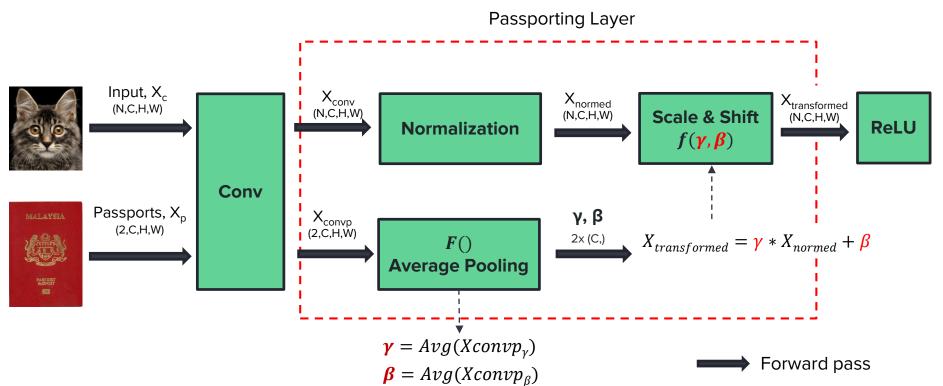
Model cannot function without valid passport

Conventional Convolution Layer

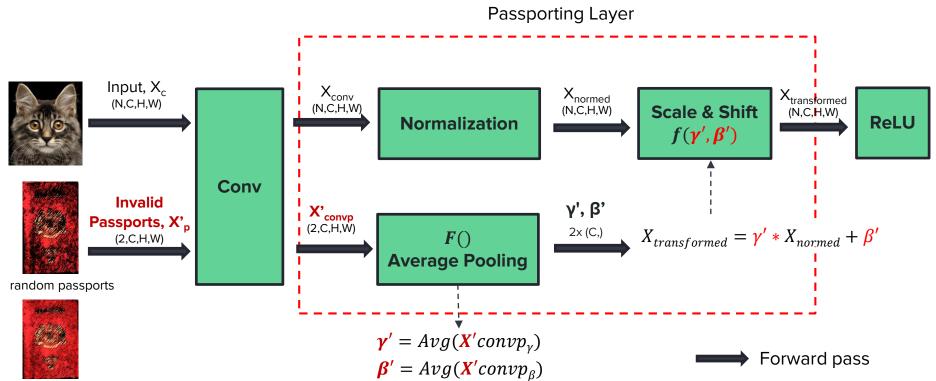




Passporting Layer



Passporting Layer



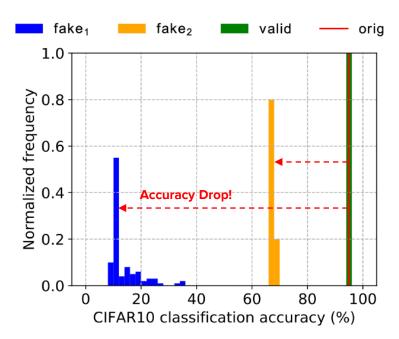
reverse-engineered passports

19

Effectiveness of Passport Protection

Result of Invalid passports

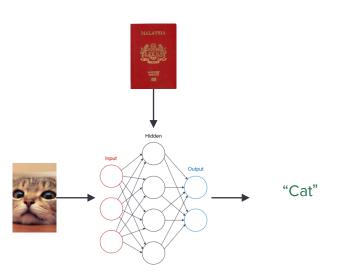
Ambiguity attack	Effect	
Fake ₁ (random passports)	Random guessing (at max 35%)	
Fake ₂ (reverse-engineered passports)	Performance deteriorated (at max 70%)	



Example of ResNet_p-18 performance on CIFAR10 when performing different ambiguity attacks (fake₁ & fake₂)

Ownership Verification with Passports (Scheme 1)

Training & Inference



Passport is distributed with the trained DNN model

Model ownership is verified by **passports, performance and signature**

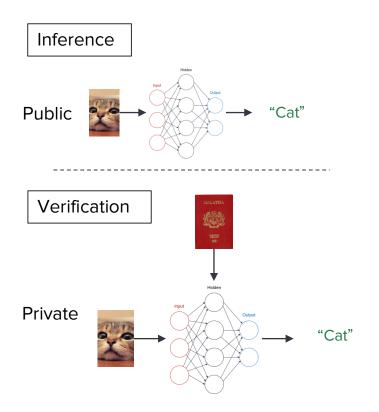
Verification type:

White-box

Disadvantage:

- 1. Need to distribute the passports
- 2. Extra inference time

Ownership Verification with Passports (Scheme 2)



Private passport is embedded but not distributed

Verification type:

White-box

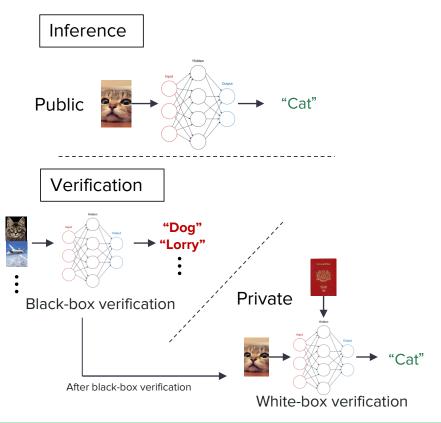
Advantage:

- 1. No need to distribute the passports
- 2. No extra inference time

Disadvantage:

Still a white-box verification

Ownership Verification with Passports (Scheme 3)



Both the private passport and trigger set are embedded but not distributed

Black-box model ownership is verified by **query API calls**

Verification type:

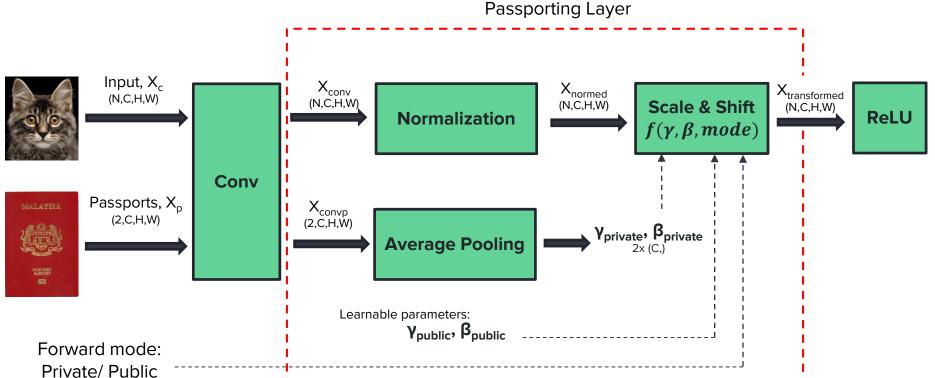
- 1. Black-box
- 2. White-box

Disadvantages of Scheme 1 & 2 Solved:

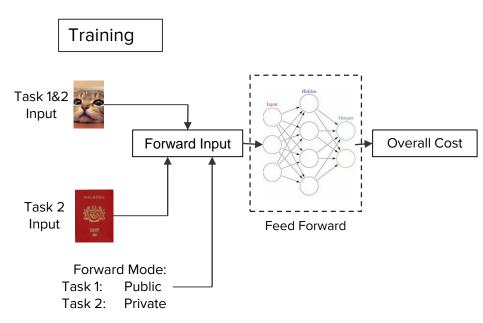
- 1. No need to distribute the passports
- 2. No extra inference time
- Able to have an initial suspect through black-box verification

Passporting Layer (Scheme 2 & 3)

Forward pass



Training on Scheme 2 & 3



Multi-task training:

Simultaneously minimizing following cost functions:

Scheme 2

Task 1: Cross Entropy (Public)

Task 2: Cross Entropy + Sign Loss (Private)

Overall Cost: L = L_Task 1 + L_Task 2

Scheme 3

Task 1: Cross Entropy (Public)

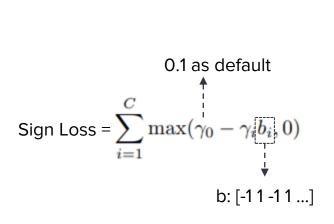
Task 2: Cross Entropy + Sign Loss (Private)

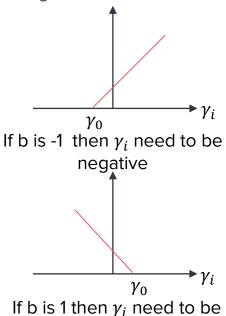
Task 3: Trigger-set Embedding Loss (Public + Private)

Overall Cost : L = L_Task 1 + L_Task 2 + L_Task 3

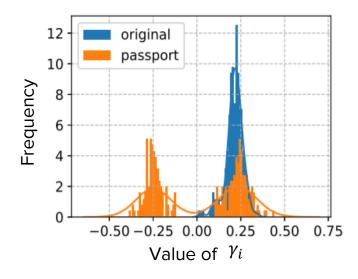
Embedding Binary Signatures by Sign of Scale Factors

- Enforce scale factor to take either positive or negative signs as designated
 - Using hinge-loss like of regularization: Sign-Loss
 - 64 channels can embed 8 bytes signature





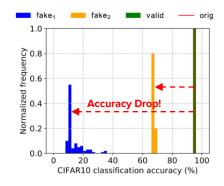
positive



Summary of Ambiguity Attacks

Summarized result done on: AlexNet & ResNet18

Datasets: CIFAR10 & CIFAR100



Ambiguity Attacks	Inference Phase	Verification Phase	
Fake ₁ , Random Passport	- Random Guessing - Useless Model	- Useless Infringement	
Fake ₂ , Reverse-Engineered Passport	- Deteriorated Performance - Useless Model	- Useless Infringement	
Fake ₃ , Copied Passport	- Performance Remained- Signature Detected	- Ownership Verified	

Summary of Ownership Verification Schemes

	Scheme 1	Scheme 2	Scheme 3
Need to distribute passport	Yes	No	No
Inference time	Up to 10%** more time	No extra time	No extra time
Training time	Up to 30%** more time	Up to 150%** more time	Up to 150%** more time
Black or White box Verification	White	White	Black & White

^{**}Time increases are linearly depending on complexity of the network architecture

Take Home message

- Protection on DNN is urgently needed!
- **Existing** watermarking approaches are **vulnerable to ambiguity attack**
- Passport-based approach provided better protection in terms of robustness
 against removal attack and ambiguity attack
- Passport-protected DNN model will only perform well if and only if a valid passport is used, else the performance will be significantly deteriorated

Links

- Arxiv: https://arxiv.org/abs/1909.07830
- Code: https://github.com/kamwoh/DeepIPR

References

[1] Yusuke Uchida, Yuki Nagai, Shigeyuki Sakazawa, and Shin'ichi Satoh. Embedding watermarks into deep neural networks. In Proceedings of the 2017 ACM on International Conference on Multimedia Retrieval, pages 269–277, 2017

[2] Y Adi, C Baum, M Cisse, B Pinkas, and J Keshet. Turning your weakness into a strength: Watermarking deep neural networks by backdooring. In 27th USENIX Security Symposium (USENIX), 2018.

[3] Jialong Zhang, Zhongshu Gu, Jiyong Jang, Hui Wu, Marc Ph Stoecklin, Heqing Huang, and Ian Molloy. Protecting intellectual property of deep neural networks with watermarking. In Proceedings of the 2018 on Asia Conference on Computer and Communications Security (ASIACCS), pages 159–172, 2018.