

DeepLocker Concealing Targeted Attacks with AI Locksmithing

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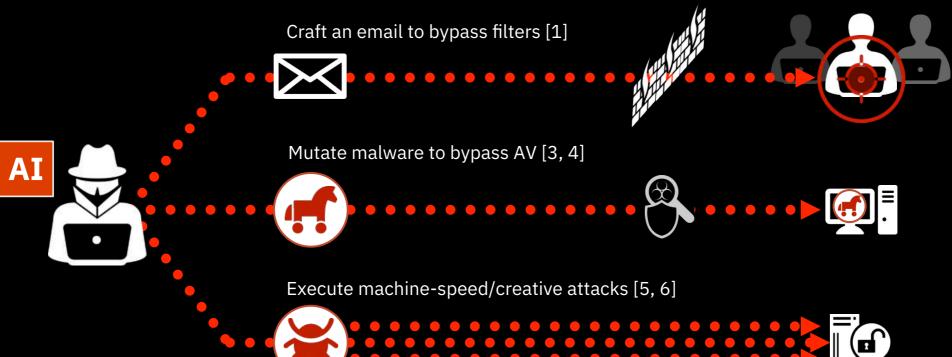
Marc Ph. Stoecklin

Cognitive Cyber Security Intelligence (CCSI) IBM Research



AI-aided attacks

Profile a target to increase success [2]



[1] S. Palka et al., "Fuzzing Email Filters with Generative Grammars and N-Gram Analysis", Usenix WOOT 2015
[2] A. Singh and V. Thaware, "Wire Me through Machine Learning", Black Hat USA 2017

[3] J. Jung et al., "AVPASS: Automatically Bypassing Android Malware Detection System", Black Hat USA 2017

[4] H. Anderson, "Bot vs. Bot: Evading Machine Learning Malware Detection", Black Hat USA 2017

[5] DARPA Cyber Grand Challenge (CGC), 2016

[6] D. Petro and B. Morris, "Weaponizing Machine Learning: Humanity was Overrated Anyway", DEF CON 2017



AI-aided attacks



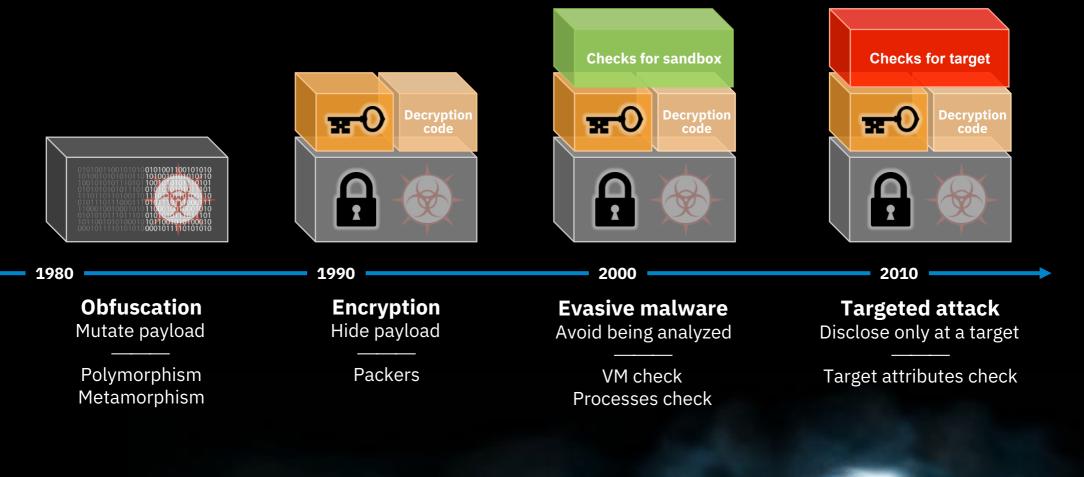


AI capability embedded inside malware itself





Malware concealment – Locksmithing





AI Locksmithing

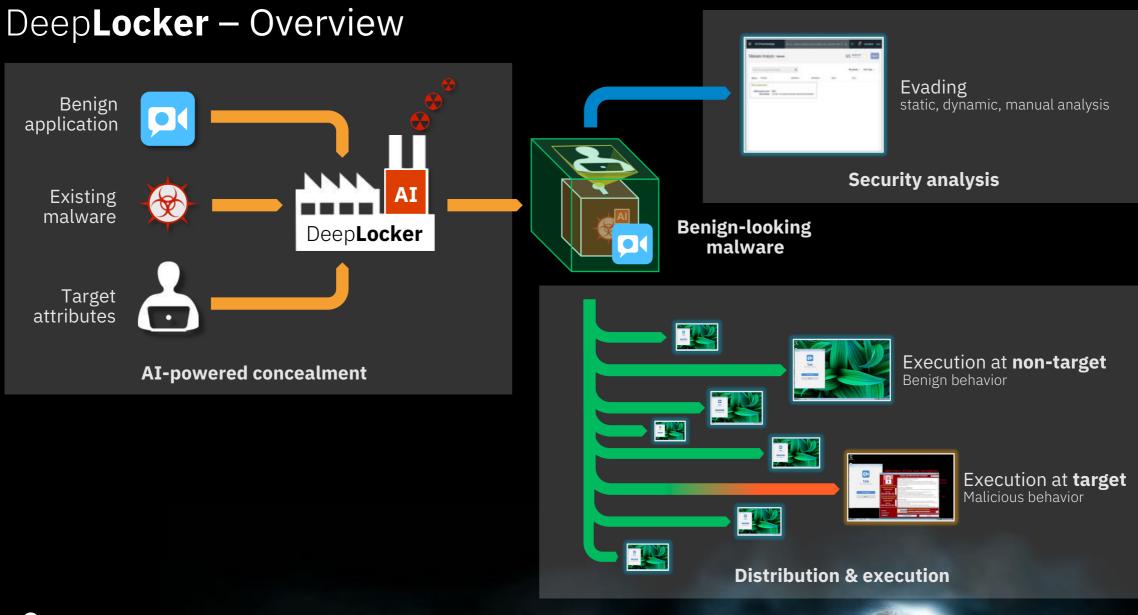


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Unleashing DeepLocker – AI Locksmithing



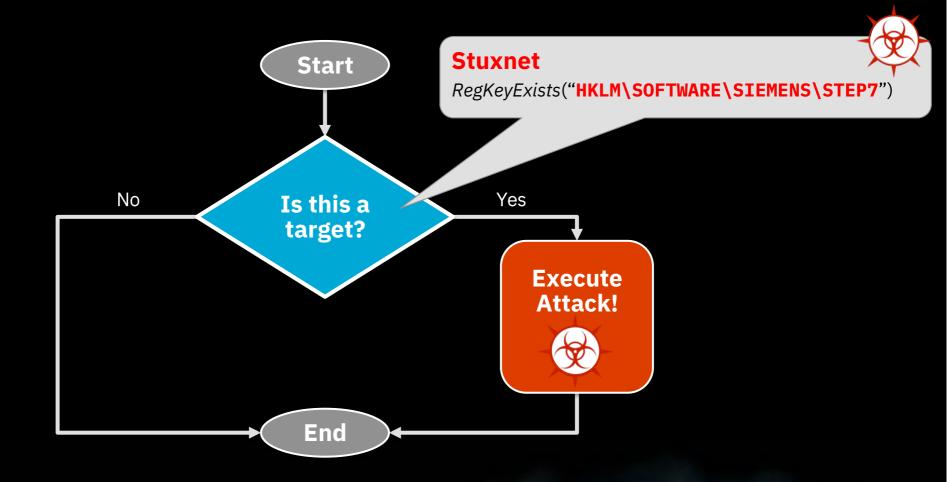




DeepLocker Deep Dive

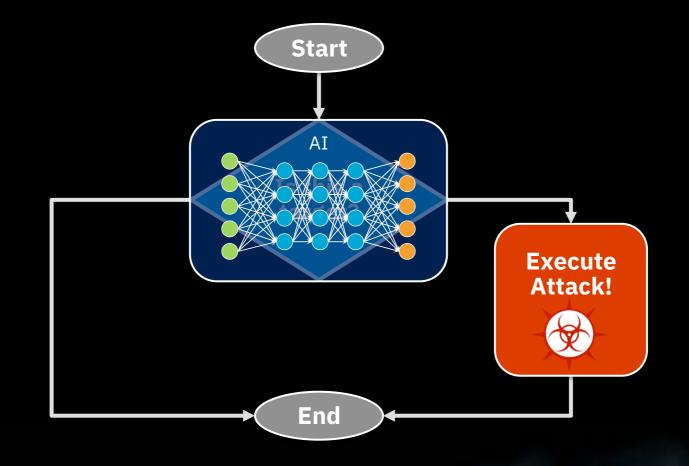


Traditional targeted attack



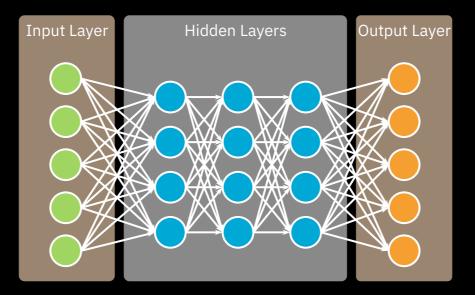


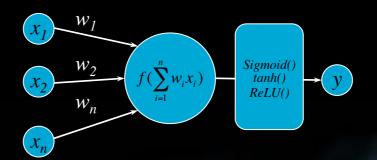
AI-powered targeted attack





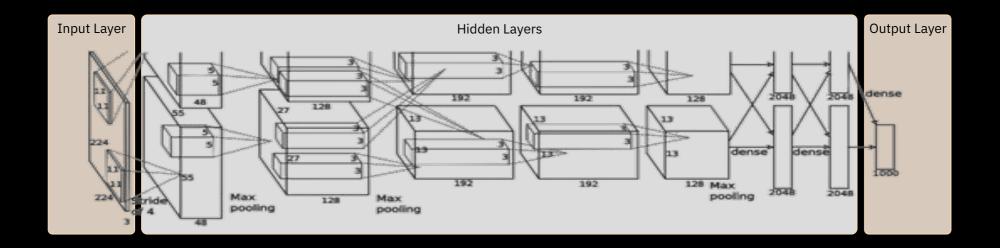
What is a Deep Neural Network (DNN)?







Deep Convolutional Neural Network

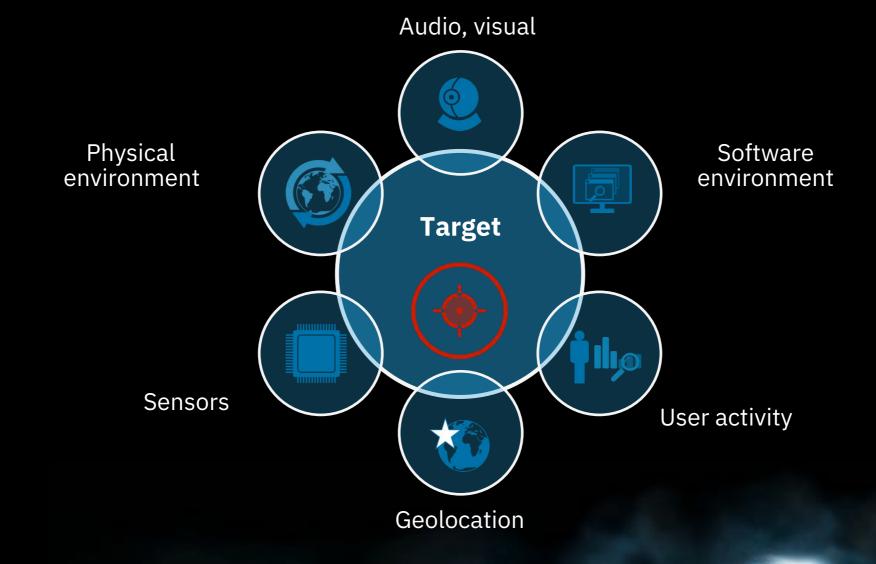


AlexNet (2012) [1] 8 layers, 622K neurons, 60 million parameters

[1] Krizhevsky, Alex, et. al. "Imagenet classification with deep convolutional neural networks." NIPS 2012.



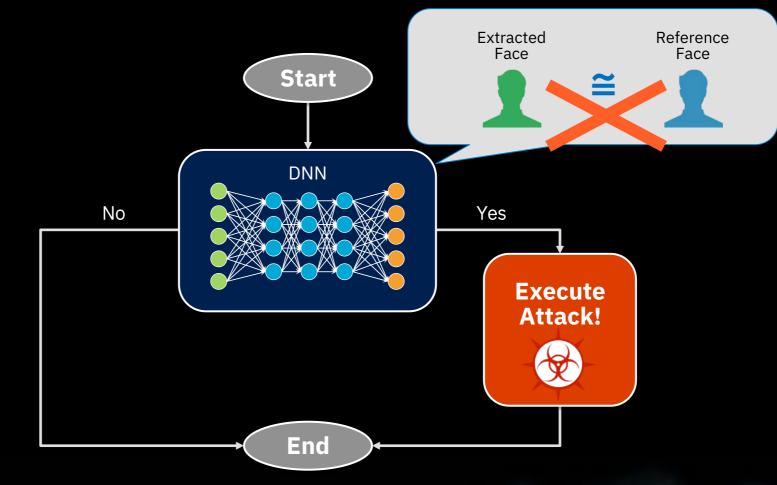
Target attributes





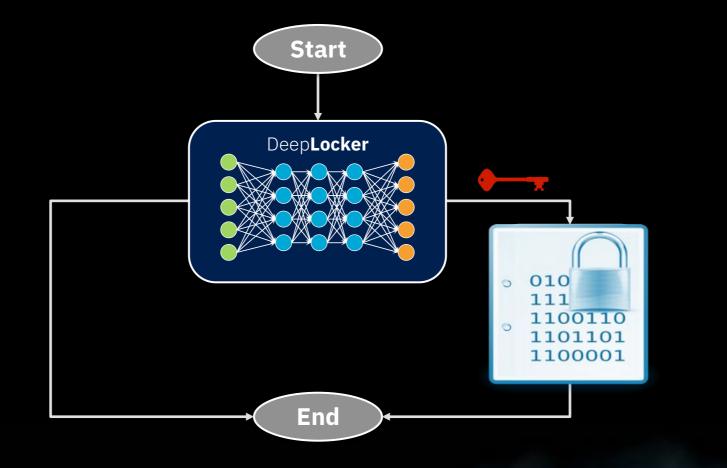
Target detection

Template matching requires a template to match to.



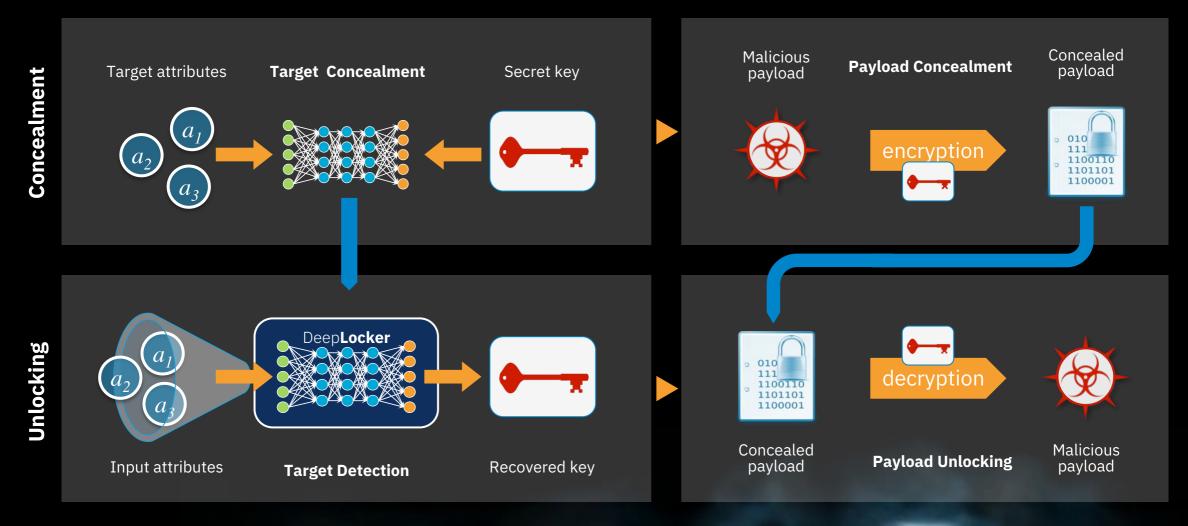


Derivation of an unlocking key



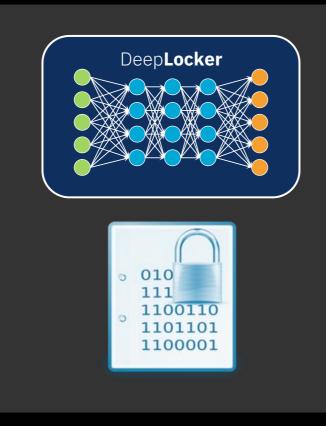


DeepLocker – AI-Powered Concealment and Unlocking





AI-powered concealment

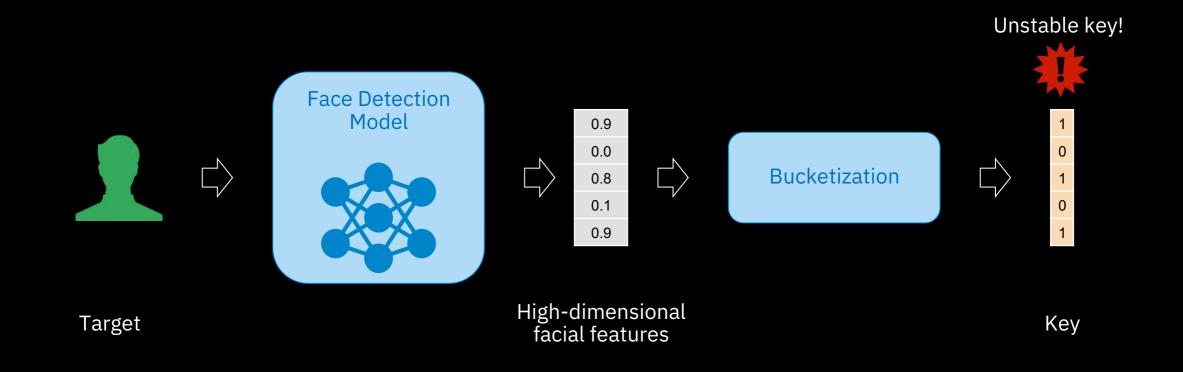




No decryption key available in malware sample to reverse engineer!

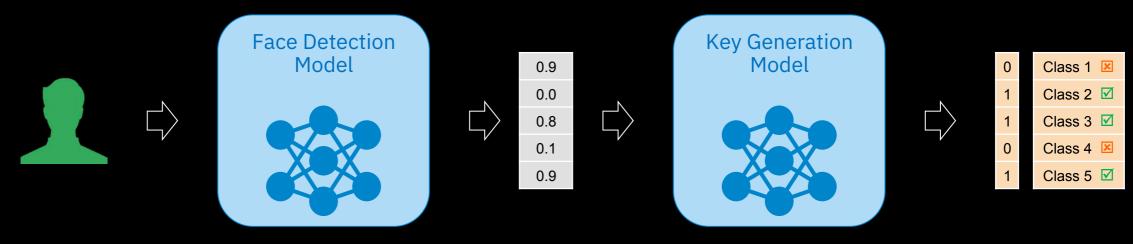


Key generation





Key generation



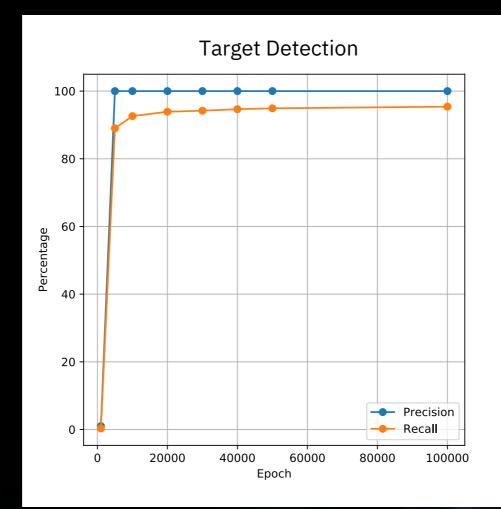
Target

High-dimensional facial features

Key



Analysis of the key generation model



Dataset: Labeled Faces in the Wild (LFW) http://vis-www.cs.umass.edu/lfw/



DeepLocker – AI-powered concealment

1 Target Class Concealment

Does not reveal **what** it is looking for (e.g., faces, organization, or a completely obscure object specific to the target environment)

3 Malicious Intent Concealment

Payload is fully encrypted concealing **how** the final attack is executed

2 Target Instance Concealment

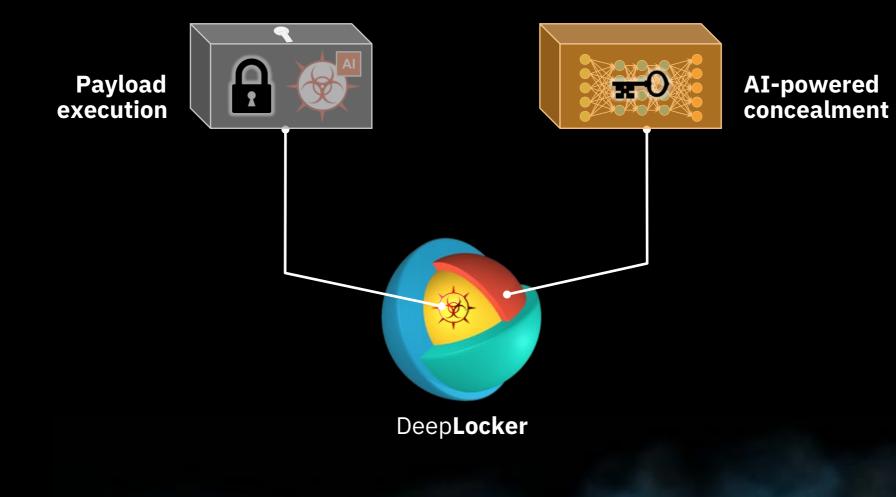
If the target class is an individual, it does not reveal **who** it is looking for



Attacking DeepLocker – AI Lock Picking



Ways to counter

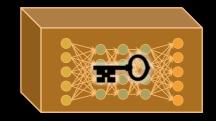




Ways to counter

Payload execution





AI-powered concealment

Code attestation



01 1Q

Block sensor access

Host-based monitoring

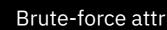
Brute-force key

Deceptive resources

Code analysis

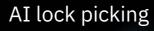


AI usage monitoring



Brute-force attributes

Deceptive attributes





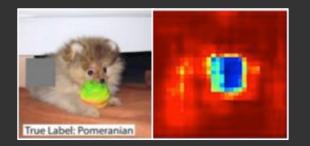




Reverse engineering AI models

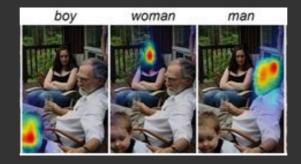
Partial occlusion

Occlude a portion of the image to see how the embedding is affected (deconvnet) [1]



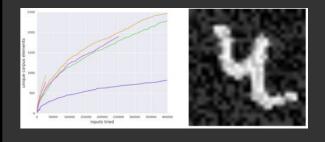
Neural attention model

Heatmap using the degree of excitation of neurons in each layer (excitation backprop) [2]



Debug neural networks

Fuzzing for neural networks (coverage-guided fuzzing) [3]



M. Zeiler and R. Fergus, "Visualizing and understanding convolutional networks," ECCV 2014
J. Zhang et. al., "Top-down neural attention by excitation backprop," ECCV 2016
A. Odena and I. Goodfellow, "TensorFuzz: Debugging neural networks with coverage-guided fuzzing," arXiv 2018



Takeaways

Rapid democratization of AI has made AI-powered attacks an imminent threat

DeepLocker is a demonstration of the potential of **AI-embedded attacks**

Current defenses will become obsolete and **new defenses are needed**



Thank you

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