AdVersarial: Perceptual Ad Blocking meets Adversarial Machine Learning

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The Future of Ad-Blocking



Towards Computer Vision for Ad-Blocking

Why not detect ad-disclosures programmatically?





New arms race on HTML obfuscation E.g., Facebook vs uBlockOrigin: <u>https://github.com/uBlockOrigin/uAssets/issues/3367</u> >1 year, >275 comments, and counting... Exact image matching is not enough

Perceptual Ad-Blocking

- Ad Highlighter [Storey et al., 2017]
 - > Visually detects ad-disclosures
 - > Traditional computer vision techniques
 - > Similar techniques deployed in Adblock Plus
- Sentinel by Adblock Plus [Paraska, 2018]
 - > Locates ads in Facebook screenshots using neural networks
- Percival by Brave [Din et al., 2019]
 - > Neural network embedded in Chromium's rendering pipeline









Perceptual Ad-Blocking



How Secure is Perceptual Ad-Blocking?



The Current State of ML

ML works well on average

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ML works well on adversarial data

Adversarial Examples



Szegedy et al., 2014 Goodfellow et al., 2015

What's the Threat Model?



(Eykholt et al. 2017)



(Eykholt **et al.** 2018)





What's the Threat Model?

Is there an adversary?

Are there no simpler attacks?

- Misclassified clean examples?
- Attacks that affect human perception too?

White-box access to the model?

Or query access / access to training data?









Unless the answer to all these questions is Yes, adversarial examples are likely not the most relevant threat

Adversarial Examples for Perceptual Ad-Blockers



Ad-Block Evasion

- Goal: Make ads unrecognizable by ad-blocker
- Adversary = Website publisher



Evasion: Universal Transparent Overlay

Web publisher perturbs every rendered pixel





Use HTML tiling to minimize perturbation size (20 KB)

- > 100% success rate on 20 webpages not used to create the overlay
- The attack is universal: the overlay is computed once and works for all (or most) websites
- Attack can be made stealthier without relying on CSS

Ad-Block Detection

- <u>Goal</u>: Trigger ad-blocker on "honeypot" content
 - > Detect ad-blocking in client-side JavaScript or on server
 - > Applicability of these attacks depends on ad-blocker type



- Adversary = Website publisher
 - > Use client-side JavaScript to detect DOM changes

Detection: Perturb fixed page layout

Publisher adds honeypot in page-region with fixed layout > *E.g.*, page header



With honeypot header

New Threats: Privilege Abuse

Ad-block evasion & detection is a well-known arms race. But there's more!



What happened?

- Object detector model generates box predictions from full page inputs
- *Content from one user can affect predictions anywhere on page*
- Model's segmentation is not aligned with web-security boundaries

Defense Strategies



The Most Challenging Threat Model for ML

- Adversary has white-box access to ad-blocker
- Adversary can exploit False Negatives and False Positives in classification pipeline
- Adversary prepares attacks offline against attacks in real-time in the user's browser
- Adversary can take part in crowd-sourced data collection for training the ad-blocker

Take Away

- Emulating human detection of ads *could be* the end-game for ad-blockers
 - > But very hard (impossible?) with current computer vision techniques
- Perceptual ad-blockers must survive an extremely strong threat model
 - > This threat model perfectly aligns with white-box adversarial examples
 - > Will we soon see adversarial examples used by real-world adversaries?
- More in the paper
 - > Unified architecture + attacks for all perceptual ad-blocker designs
 - > Similar attacks for non-Web ad-blockers (e.g., Adblock Radio)



📮 ftramer / **ad-versarial**

- Train a page-based ad-blocker
- Download pre-trained models
- Attack demos

Research Impact



How does a Perceptual Ad-Blocker Work?



- Element-based (e.g., find all tags) [Storey et al. 2017]
- Frame-based (segment rendered webpage into "frames" as in Percival)
- Page-based (unsegmented screenshots à-la-Sentinel)

Building a Page-Based Ad-Blocker

We trained a neural network to detect ads on news websites from all G20 nations



Video taken from 5 websites not used during training

Defense Strategies

- Obfuscate the ad-blocker?
 - > It isn't hard to create adversarial examples for black-box classifiers
- Randomize the ad-blocker?
 - > Adversarial examples robust to random transformations / multiple models
- Pro-actively retrain the model? (Adversarial training)
 - > *New arms-race:* The adversary finds new attacks and ad-blocker re-trains
 - > Mounting a new attack is much easier than updating the model
 - > On-going research: so far the adversary always wins!