

Baidu

Team Members: Darren Yang, Xincheng Li, Jay Xiong



Problem

With increasing amount of valuable and privacy-sensitive digital assets uploaded on the social media, online booking website and et al, image CAPTCHAs are being widely used across the Internet to defend against abusive programs¹ and economics of cyber-crimes that rely on large-scale automation. However, via deep learning technology, a CAPTCHA breaker is capable of compromising most of the CAPTCHAs on the market²,³. What's more, the arm race between defender and CAPTCHA breaker results in CAPTCHAs too complex for humans making traditional CAPTCHA schemas only a hassle for the Internet users.

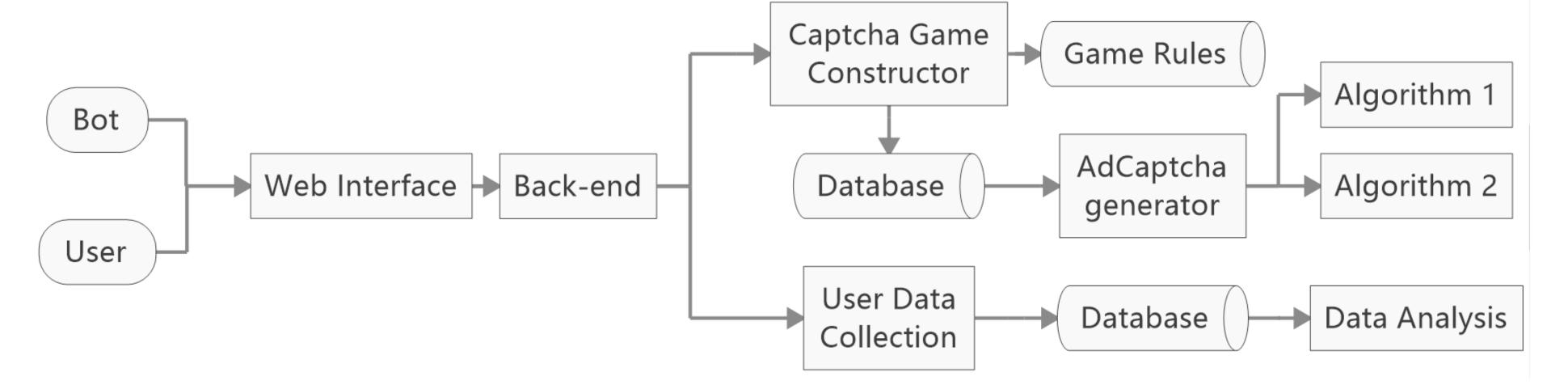
Competitor Analysis of CAPTCHAs

| | Text | Behavior | Image | Our AdCaptcha |
|--------------|---|------------------------------|--------------------------------|--|
| For Computer | Easy | Moderate | Hard | Hard |
| For Human | Moderate | Easy | Hard | Easy |
| | Type the two words: OF CAPTCHA OF CAPTCHA | | Select all images with cars | Select all images with tailed frog Click verify once there are none left Progress: 1 / 50 |
| | I'm not a robot | reCAPTCHA Privacy - Terms | C O O Verify Report a problem | |

Solution

To make CAPTCHAs more effective in screening bots and more friendly to human users, we proposed AdCaptcha (Adversarial CAPTCHA). AdCaptcha's schema comes from deep learning attacking algorithms which differs from traditional ones whose security schemas are produced by a set of fixed rules. The technology could exploit the maximum effectiveness of a security schema that could be imposed on a CAPTCHA image. We used C&W⁴ attacking algorithm to build Adversarial images on ImageNet⁵ data set, and we prototyped a system with API that works on both human and bot for HCI testing.

System Architecture



References:

[1] "Towards Evaluating the Security of Real-World Deployed ... - NESA." https://nesa.zju.edu.cn/download/Towards%20Evaluating%20the%20Security%20of%20Real-World%20Deployed%20Image.pdf. Accessed 2 Dec. 2018.

[2] "Yet Another Text Captcha Solver: A Generative Adversarial Network" http://www.lancaster.ac.uk/staff/wangz3/publication/ccs18/. Accessed 2 Dec. 2018

[3] "I'm not a human: Breaking the Google reCAPTCHA - Black Hat." https://www.blackhat.com/docs/asia-16/materials/asia-16-Sivakorn-Im-Not-a-Human-Breaking-the-Google-reCAPTCHA-wp.pdf. Accessed 2 Dec. 2018.
[4] "Towards Evaluating the Robustness of Neural Networks." Accessed December 4, 2018. https://arxiv.org/abs/1608.04644.

[5] "ImageNet Large Scale Visual Recognition Competition (ILSVRC)." Accessed December 4, 2018. http://image-net.org/challenges/LSVRC/.

Evaluation

We conducted two groups of tests on Google Vision API and human participants. Independent variables are generated adversarial images and original images from imagenet dataset. Captcha recognition error rate and time are used for metrics. In addition, NASA TLS is used for difficulty measurement on human test. Results show that for Google Vision API, adversarial images have higher error rate, while for human original and adversarial images don't have significant difference on error rate, time and difficulties. This means adversarial captcha can defend captcha cracker better, and is as easy as normal images for human.

