Real and Stealthy Attacks on State-of-the-Art Face Recognition * Mahmood Sharif, Sruti Bhagavatula, and Lujo Bauer, CMU; Michael K. Reiter, UNC

Introduction

Machine learning (ML) is ubiquitous, enables revolutionary technologies:



Background and Prior Work

ML classifiers (e.g., in intrusion detection, cancer detection, ...) are functions from inputs to classes (or probability distributions over classes)



If ML fails:



Our research questions investigate robustness of ML algorithms:

Can attackers make ML fail? Can attacks be inconspicuous and physically realizable?

Our Approach and Results

Our focus: DNNs for state-of-the-art face recognition [2]

Attack goals:

- *Impersonation*: being classified as specific target
- Dodging: not being classified as self

Imperceptible attacks have been demonstrated that confuse deep neural networks (DNNs) [1], by solving:

> $argmin_r \left[f(x+r) - l \right]$ +*C*|*T*| misclassification imperceptibility

x is the input image; $f(\cdot)$ is the classification function (e.g., DNN); l is the desired output class; r: perturbation (or change applied to the input).



Attack generation:



Images of attacker



We create realizable, inconspicuous attacks by: 1. Limiting perturbation to eyeglass frames 2. Minimizing total variations (TV) btw. adjacent pixels 3. Minimizing "non-printability score" (NPS) 4. Increasing robustness: an attack should fool the system for more than one face image Objective for impersonation (dodging is analogous): $\underset{r}{\operatorname{argmin}} \left(\left(\sum_{\mathbf{x} \in \mathbf{Y}} |f(x + \mathbf{r}) - l| \right) + \kappa_1 T V(r) + \kappa_2 N P S(r) \right)$

* M. Sharif, S. Bhagavatula, L. Bauer, and M. K. Reiter. Accessorize to a crime: Real and stealthy attacks on state-of-the-art face recognition. CCS, 2016. [1] C. Szegedy, W. Zaremba, I. Sutskever, J. Bruna, D. Erhan, I. J. Goodfellow, and R. Fergus. Intriguing properties of neural networks. *ICLR*, 2014.

[2] O. M. Parkhi, A. Vedaldi, and A. Zisserman. Deep face recognition. BMVC, 2015.

Alg. to generate accessories



Results: fool DNN trained on 7 subjects + 3 authors



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In paper: more experiments with larger DNN*