

Making Targeted Evasion Attacks Effective and Efficient



PRISM_R Demo on Google Cloud Vision

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Targeted Black-box Evasion Against Realistic, General APIs:

- PRISM: A novel black-box attack using substitute models
- An agile adversary can achieve better effectiveness/efficiency by switching through methods
- Demonstrated against real-life API: Google Cloud Vision

Targeted evasion against realistic, general APIs

- Targeted DNN trained with 1000s of classes
- How to change API response to target class *y*:

$$y' \leftarrow API(x')$$
$$||x - x'||_{\infty} < \epsilon$$

- Partial Information API: responds with top-k results
- Modifications very small, e.g. $\epsilon = 5\%$ (12.75 / 255)

Approach 1: Transferability attacks: Ens

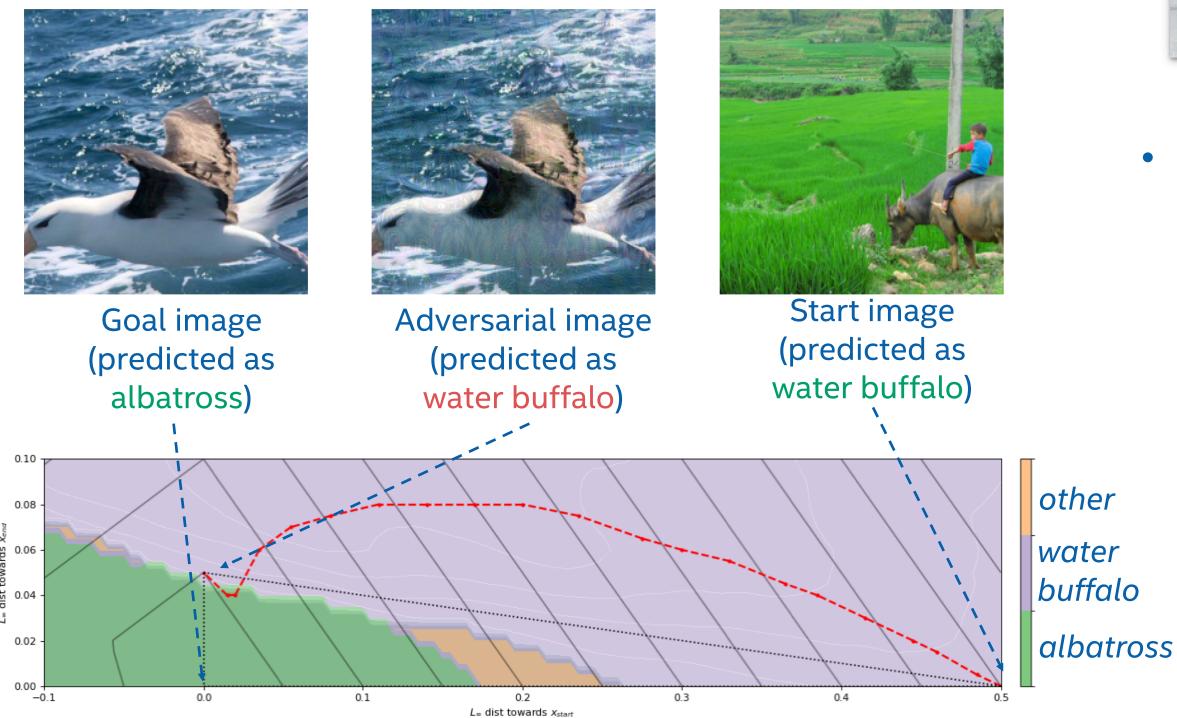
- Adversarial examples created on *ensembles*, e.g. using MIFGSM [1]
- Efficient attacks: first query may succeed
- Targeted evasion ineffective with imperceptible modifications [3]

Approach 2: Query-only methods for PI API: QO

- Finite-difference methods for estimating gradients, e.g. NES [2]
- Start with image of target class *y*
- Effective: any target model attackable with almost 100% success
- Inefficient under PI API 10,000s 100,000s of queries per sample

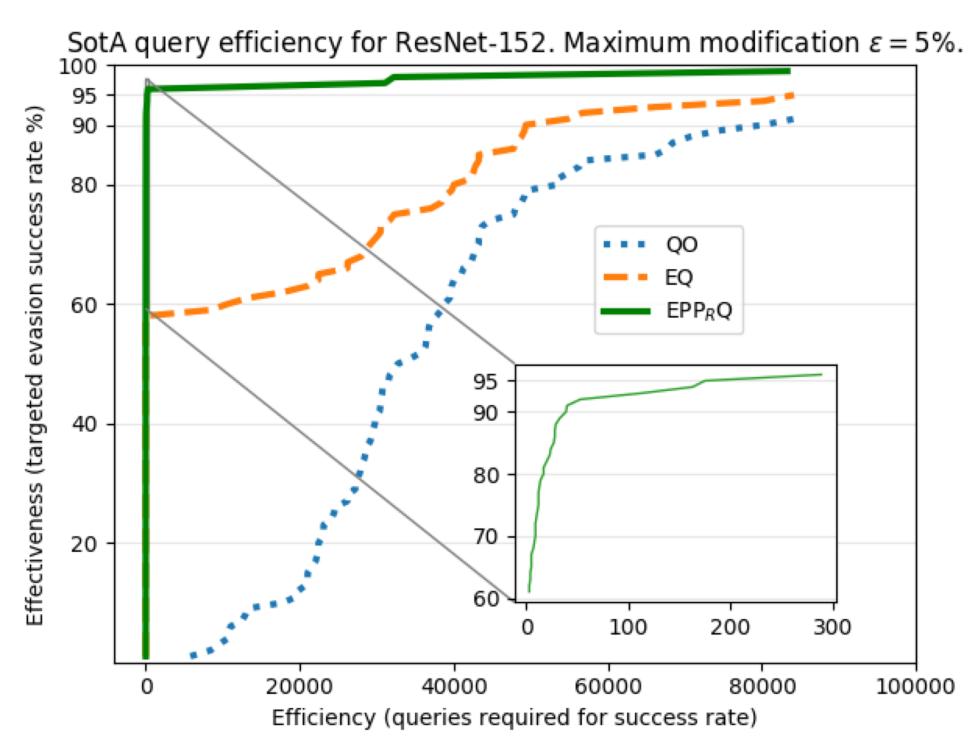
Our Approach: PRISM

- Start with image of target class *y*
- Gradient estimation via **Ens**
- **PRISM**_R: randomized variant
- Effective: similar success rate as **QO**
- Efficient: three orders of magnitude faster than **QO**



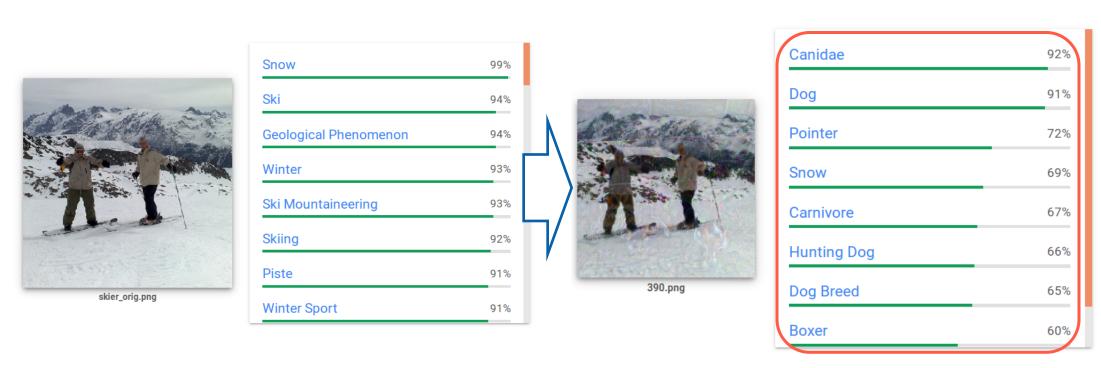
Agile Adversary

- Can analyze efficiency of methods and use these pareto-optimally
- Some Ens then QO more efficient than simple QO
- Significant efficiency/effectiveness improvement with paretooptimal order: Ens, PRISM, PRISM_R, QO: EPP_RQ

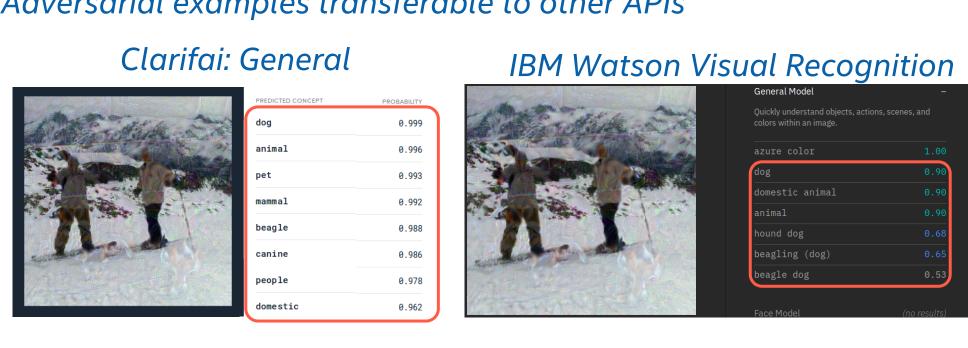


Applicable against real-life APIs

- Demo on Google Cloud Vision
- Decreases effort from ~20,000 queries [2] to ~400-1000 queries



Adversarial examples transferable to other APIs



- [1] Dong et al. Boosting adversarial attacks with momentum. CVPR'18
- [2] Ilyas et al. Black-box adversarial attacks with limited information and queries. ICML'18.
- [3] https://github.com/dongyp13/Targeted-Adversarial-Attack