

# Adversarially Robust Generalization Requires More Data



**Ludwig Schmidt**



**Shibani Santurkar**



**Dimitris Tsipras**



**Kunal Talwar**

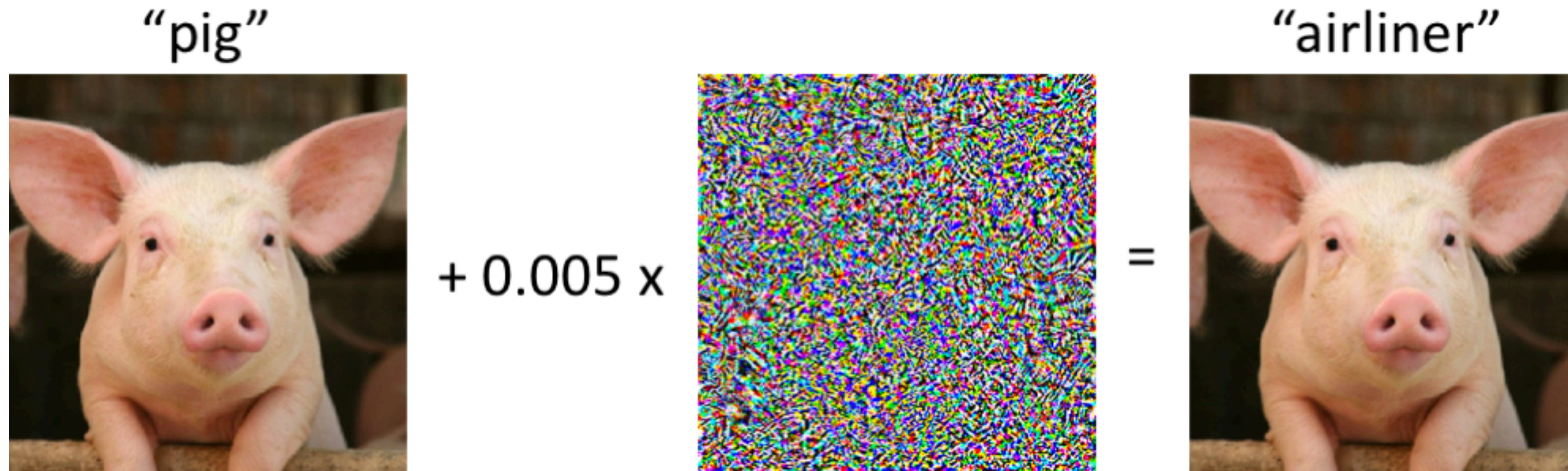


**Aleksander Mądry**



Poster #31

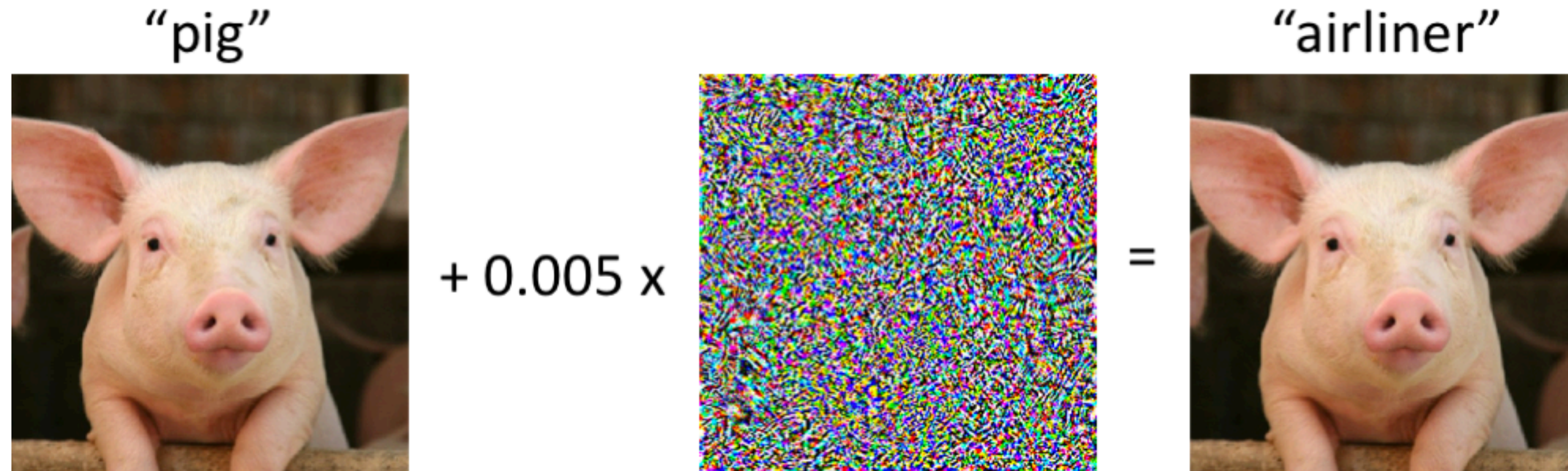
# Adversarial Examples



[Szegedy, Zaremba, Sutskever, Bruna, Erhan, Goodfellow, Fergus, 2013]

[Biggio, Corona, Maiorca, Nelson, Srndic, Laskov, Giacinto, Roli, 2013]

# Adversarial Examples



[Szegedy, Zaremba, Sutskever, Bruna, Erhan, Goodfellow, Fergus, 2013]

[Biggio, Corona, Maiorca, Nelson, Srndic, Laskov, Giacinto, Roli, 2013]

What makes adversarial examples a hard problem?

➔ This paper: perspective on **sample complexity**

# Standard vs Robust Generalization

“Standard” Generalization:  $\mathbb{E}_{x,y \sim \mathcal{D}} [\text{loss}(f(x), y)]$

# Standard vs Robust Generalization

“Standard” Generalization:  $\mathbb{E}_{x, y \sim \mathcal{D}} [\text{loss}(f(x), y)]$

Adversarially robust generalization:  $\mathbb{E}_{x, y \sim \mathcal{D}} \left[ \max_{x' \in P(x)} \text{loss}(f(x'), y) \right]$

Perturbation set: small  $\ell_\infty$ -perturbations, rotations, translations, ...

# Standard vs Robust Generalization

“Standard” Generalization:  $\mathbb{E}_{x, y \sim \mathcal{D}} [\text{loss}(f(x), y)]$

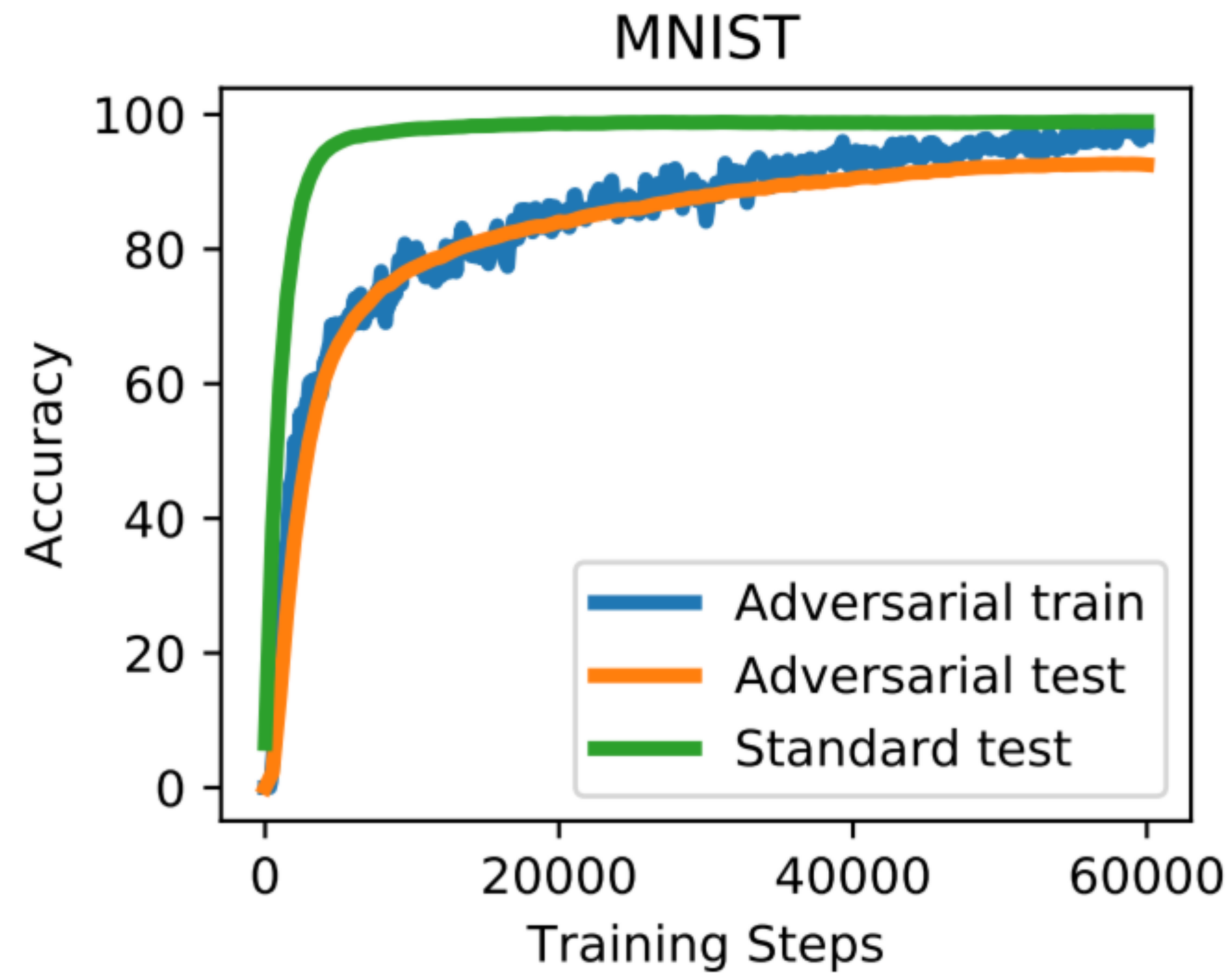
Adversarially robust generalization:  $\mathbb{E}_{x, y \sim \mathcal{D}} \left[ \max_{x' \in P(x)} \text{loss}(f(x'), y) \right]$

Perturbation set: small  $\ell_\infty$ -perturbations, rotations, translations, ...

How do these two notions of generalization compare?

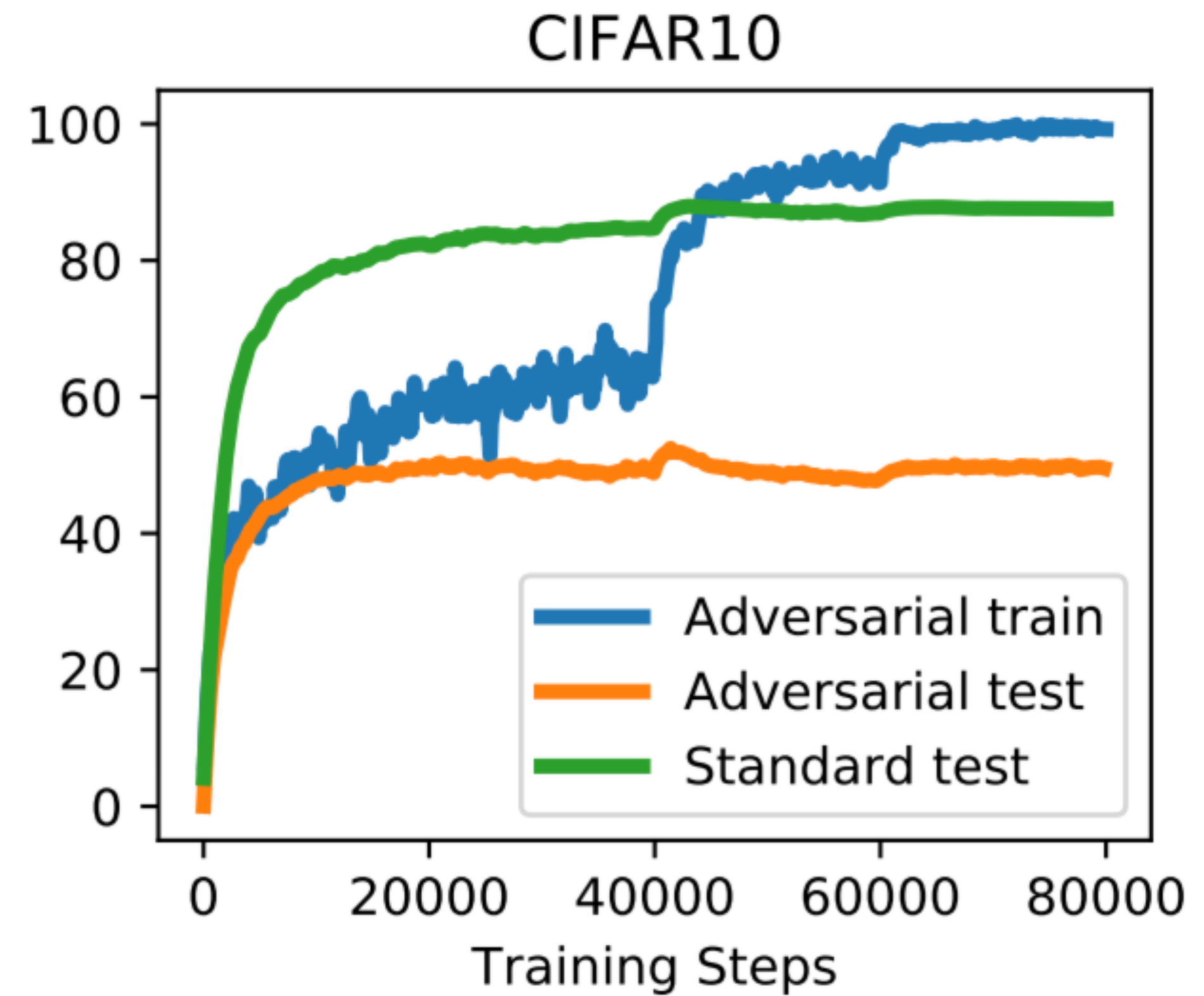
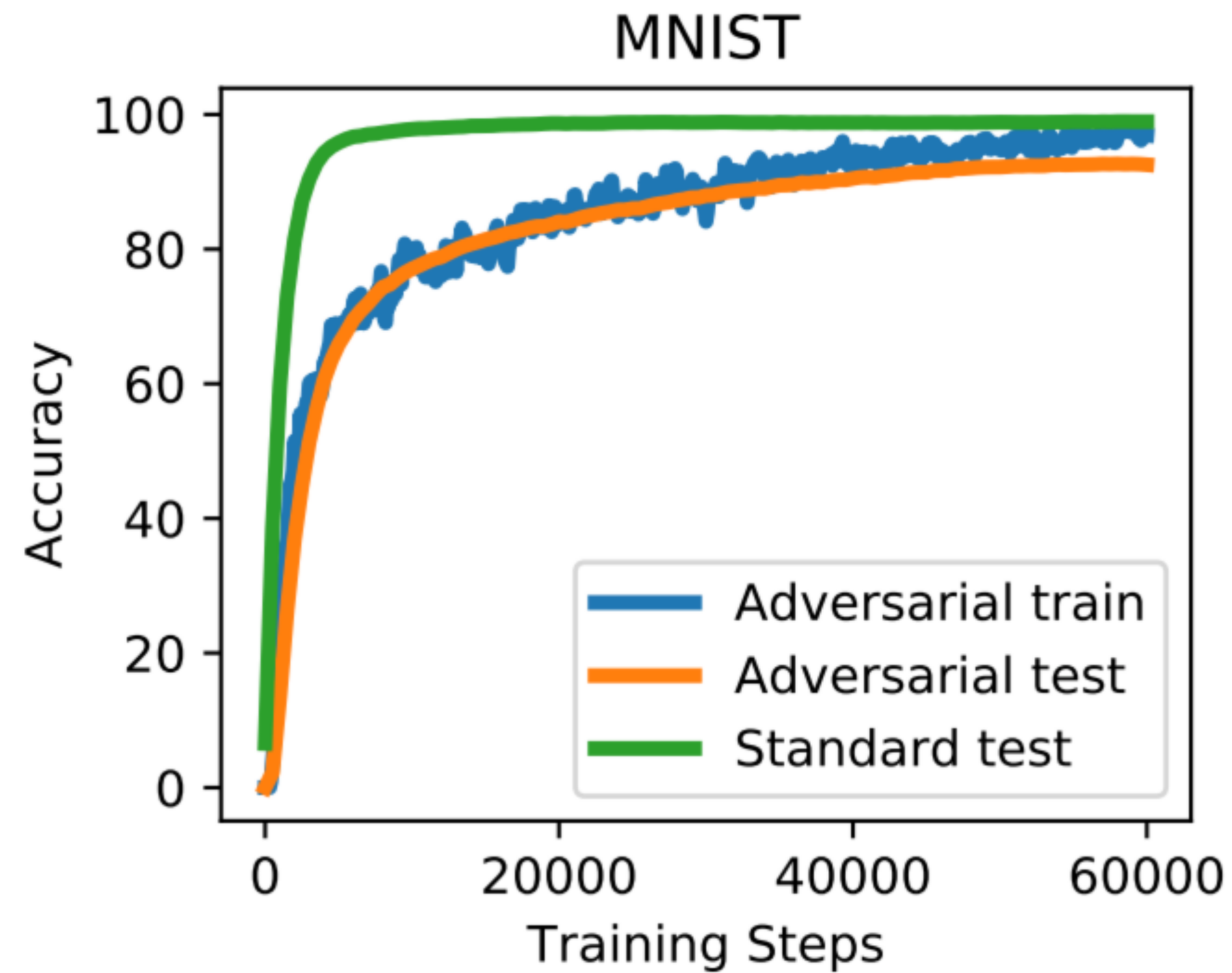
# State Of The Art in $\ell_\infty$ -Robustness

Robust optimization as in [\[Madry, Makelov, Schmidt, Tsipras, Vladu, 2017\]](#):



# State Of The Art in $\ell_\infty$ -Robustness

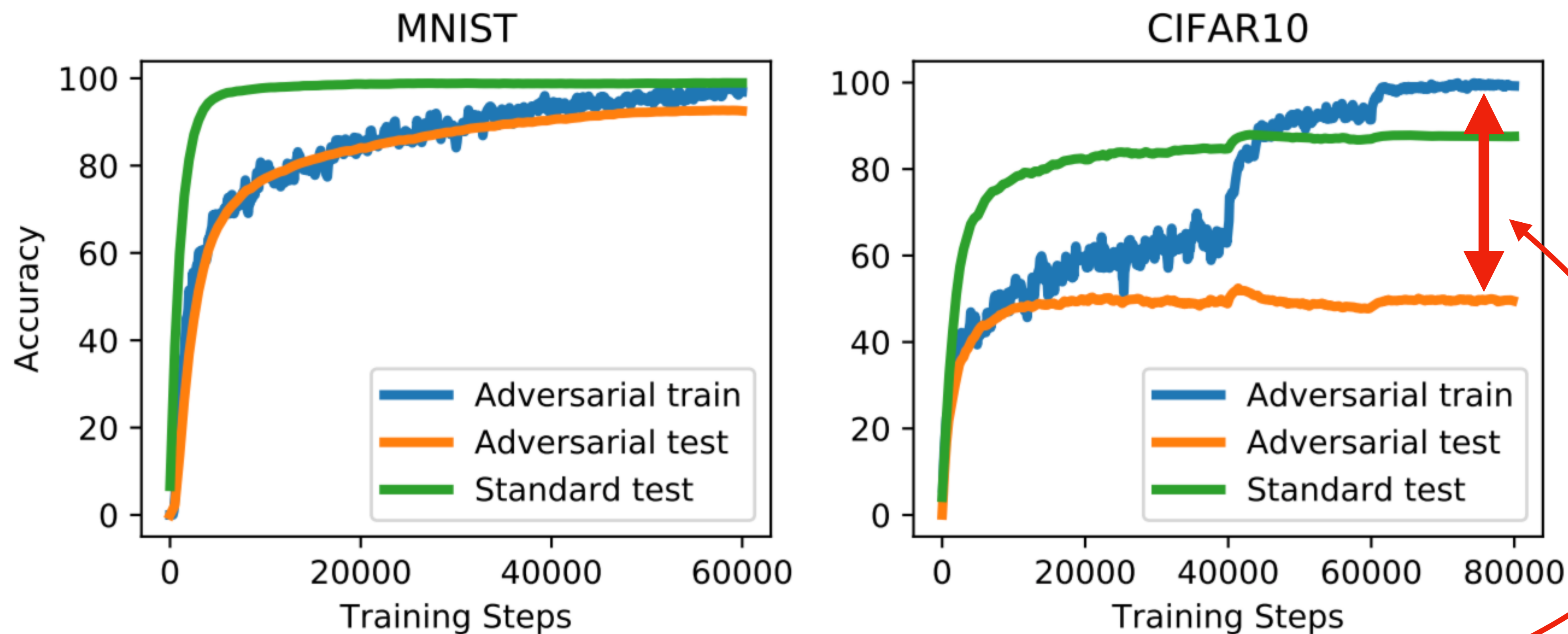
Robust optimization as in [\[Madry, Makelov, Schmidt, Tsipras, Vladu, 2017\]](#):





# State Of The Art in $\ell_\infty$ -Robustness

Robust optimization as in [Madry, Makelov, Schmidt, Tsipras, Vladu, 2017]:



Optimization succeeds in both cases, but the model **overfits** on CIFAR-10.

# Robust Generalization

Main question: Does robust generalization require more data?

# Robust Generalization

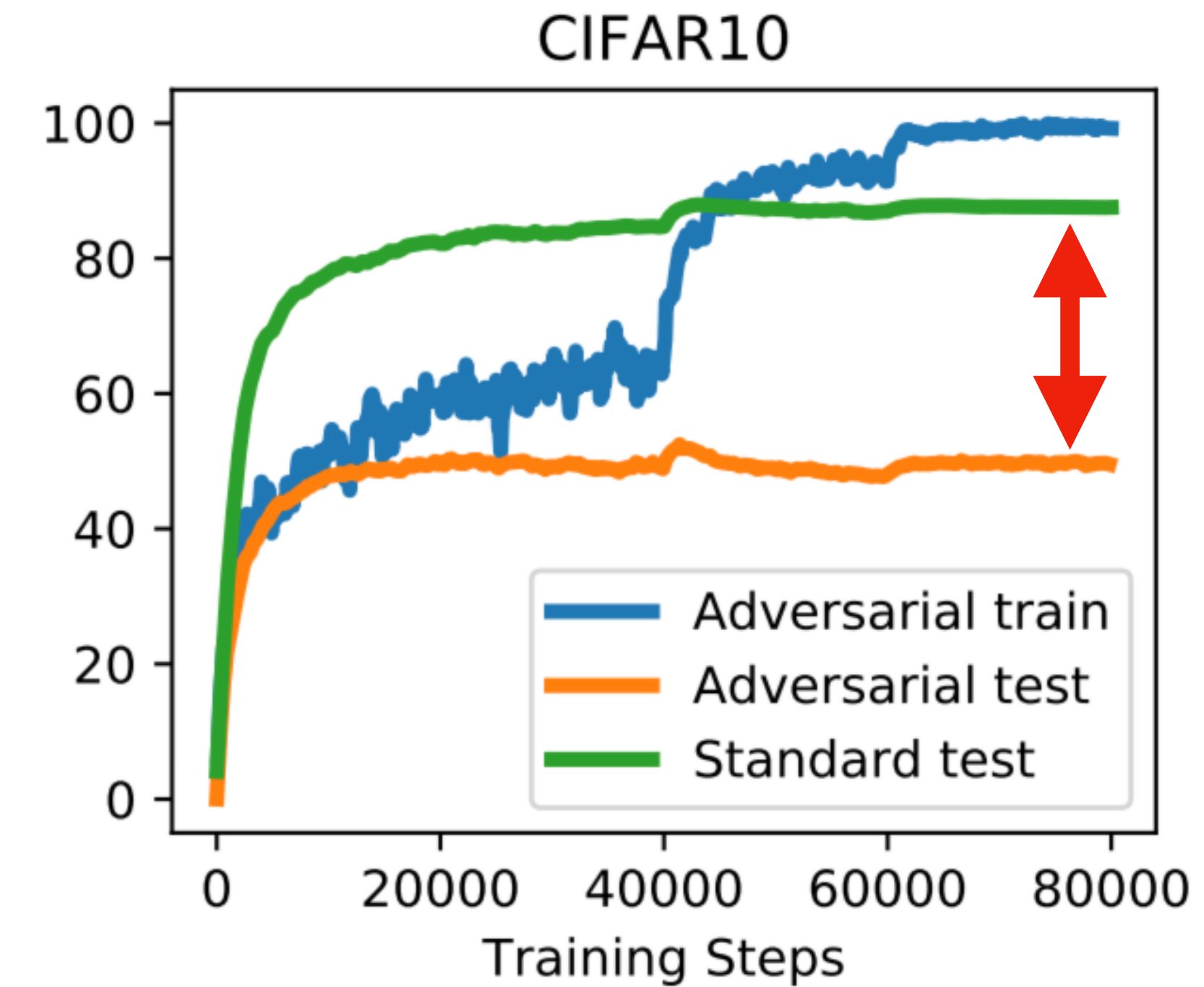
Main question: Does robust generalization require more data?

**Theorem** (informal): There is a natural distribution over points in  $\mathbb{R}^d$  with the following property:  
Learning an  $\ell_\infty$ -robust classifier for this distribution requires  $\sqrt{d}$  times more samples than learning a non-robust classifier.

# Conclusions

## Further results

- An alternative data model for MNIST
- Experiments on MNIST, CIFAR-10, SVHN



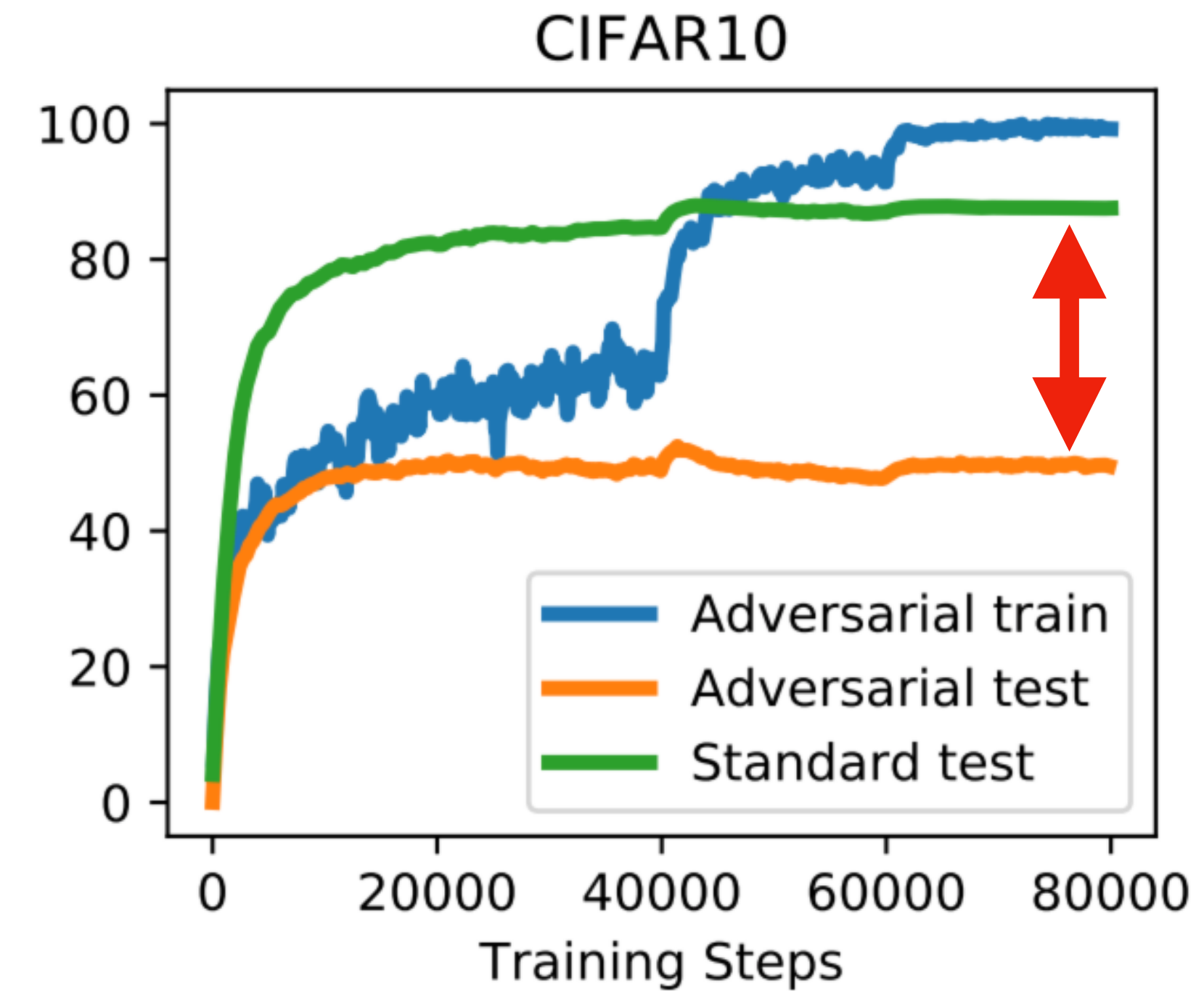
# Conclusions

## Further results

- An alternative data model for MNIST
- Experiments on MNIST, CIFAR-10, SVHN

## Main takeaways

- **Sample complexity** can be an obstacle for adv. robustness
- Need to improve priors encoded in models?
- Many phenomena not yet understood theoretically



Poster #31

[gradient-science.org](http://gradient-science.org)