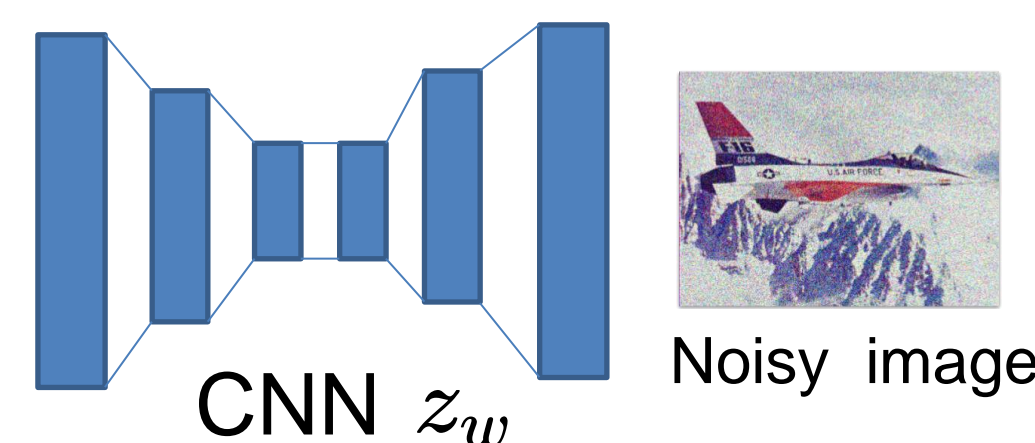


Understanding CNN denoisers

- How do CNNs perform denoising?
- Implicit learning bias?
- Deep image prior phenomenon [1]

$$\arg \min_w \|z_w - y\|^2$$



1. Only access a noisy image y
2. Trainable weights $w \gg$ training pixels in y
3. Early stopping provides very good denoising?!

Neural tangent kernel (NTK) perspective

Assuming:

1. **Overparameterization**: channels of hidden layers $c \rightarrow \infty$
2. Standard iid **initialization** of weights (e.g., He, LeCun)
3. **Gradient descent** with learning rate $\eta \propto c^{-1}$

- Weights move little from initialization $\|w^t - w^0\|_2 \approx \mathcal{O}(c^{-0.5})$
- The network's output dynamics are given in closed form by

$$z^{t+1} = z^t + \eta \Theta(y - z^t)$$

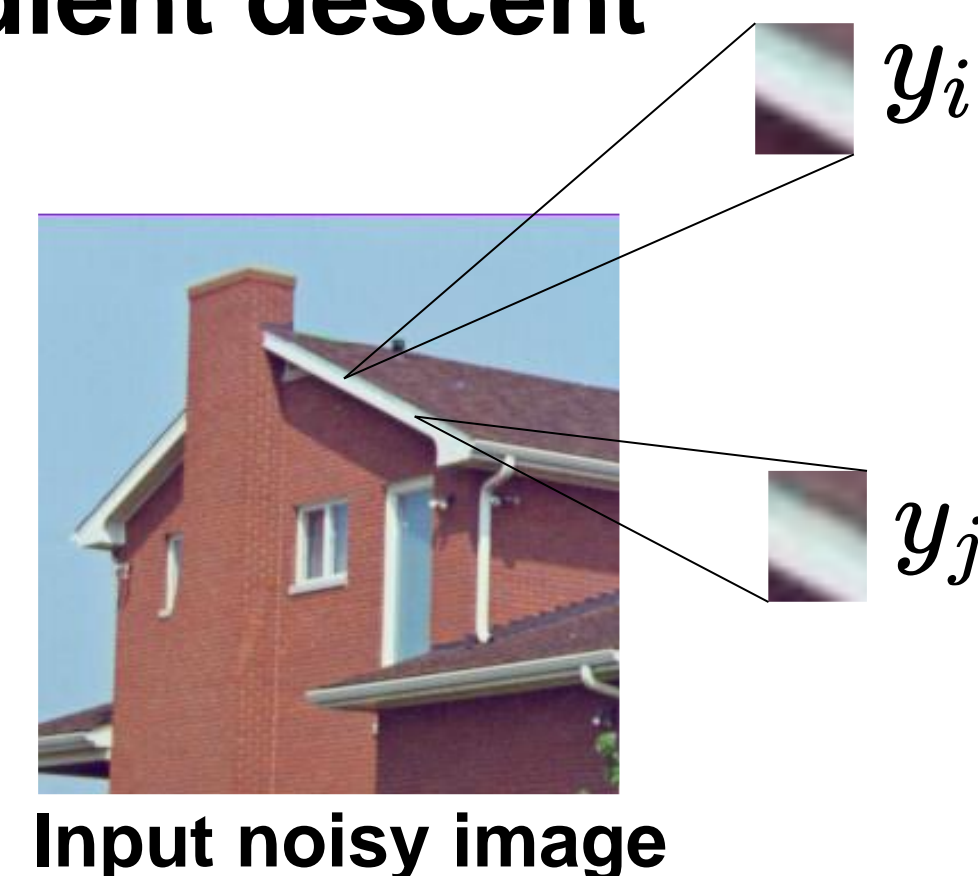
where the NTK [2] $\eta \Theta = \eta \frac{\delta z}{\delta w} \left(\frac{\delta z}{\delta w} \right)^\top \Big|_{w=w^0}$

- Stays **constant** throughout training
- Available in **closed form**
- Only **depends on network's input** and architecture

1. Input noisy image + gradient descent

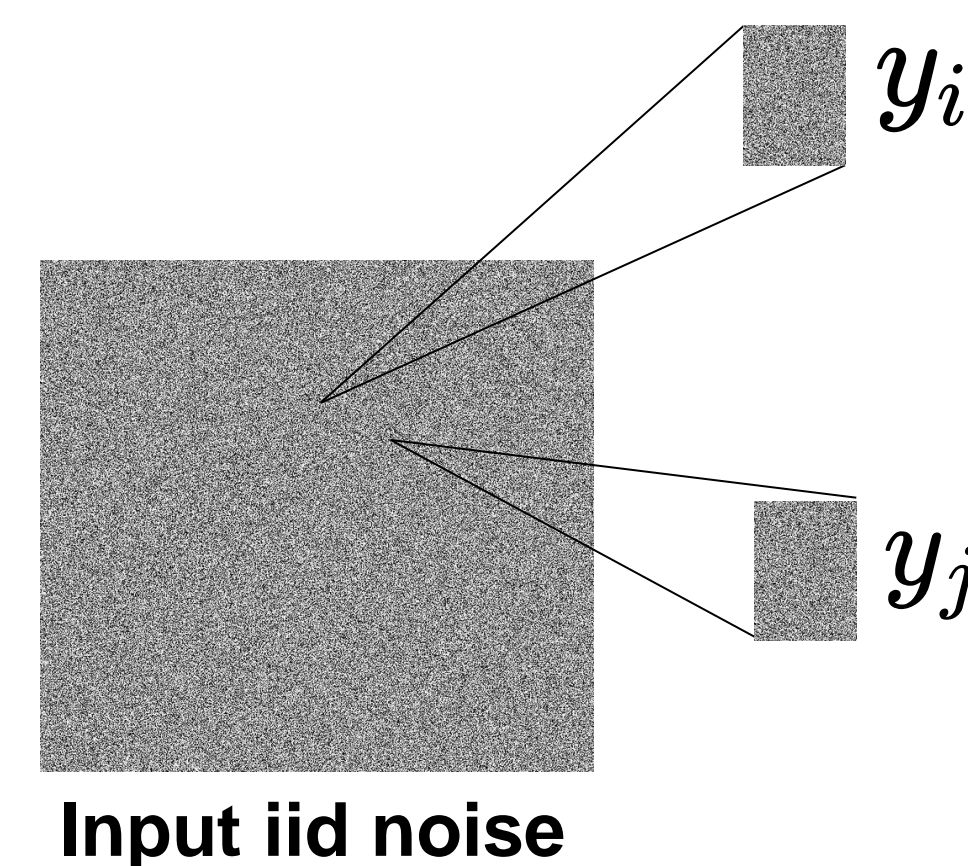
Patch similarity function

$$[\eta \Theta]_{i,j} = k(y_i, y_j)$$



- NTK = **non-local filter**
- CNN optimization = **iterative filtering** (twicing)
- Early stopping controls bias-variance tradeoff
- CNNs can be replaced by **closed form NTK** + Nyström

2. Input iid noise + Adam



- Deep image prior setting
- NTK = simple **low-pass filter**

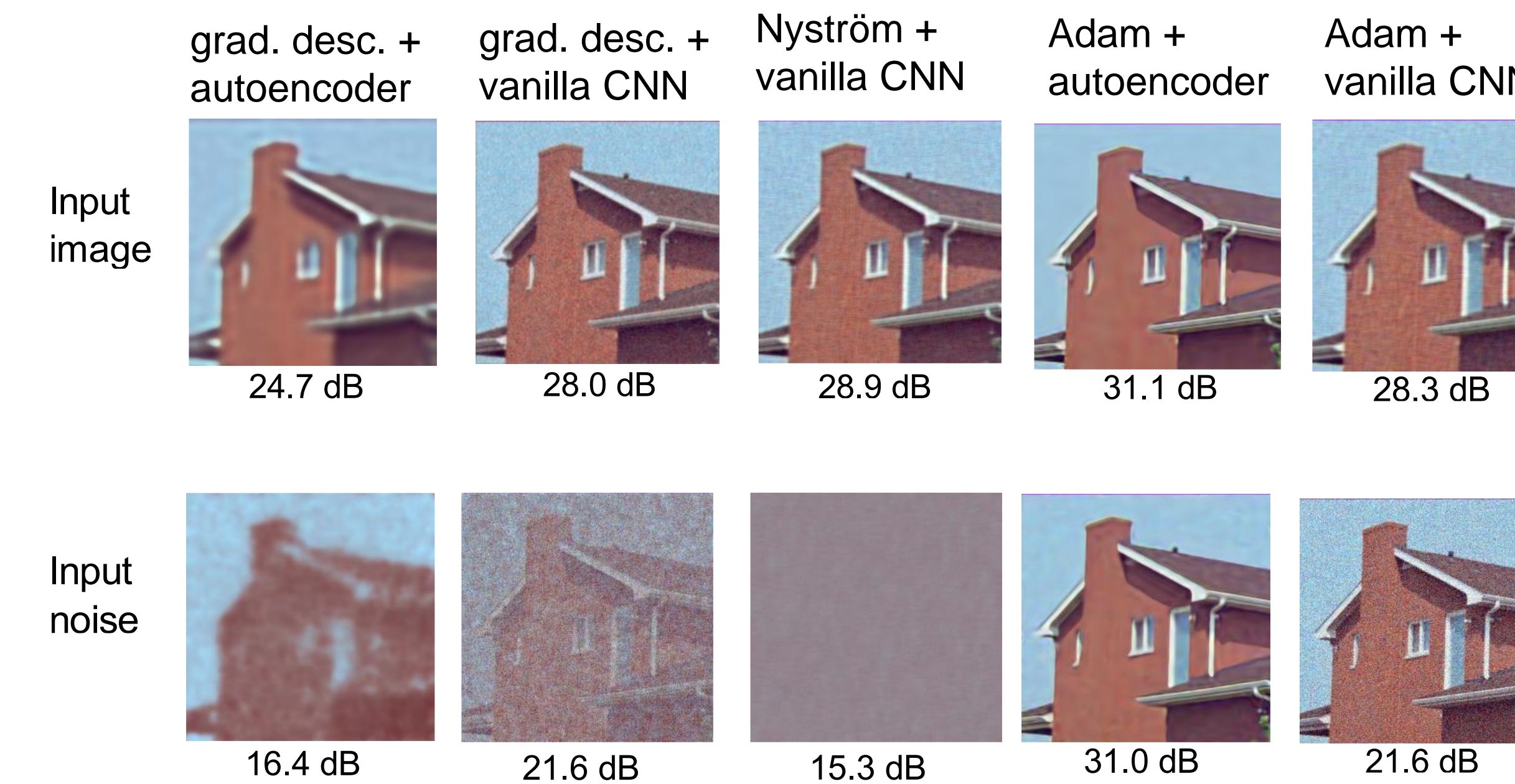
Fixed NTK **cannot explain** good performance!

- Weights **move significantly** from initialization $\|w^t - w^0\|_2 \approx \mathcal{O}(1)$
- Tangent kernel adapts throughout training:

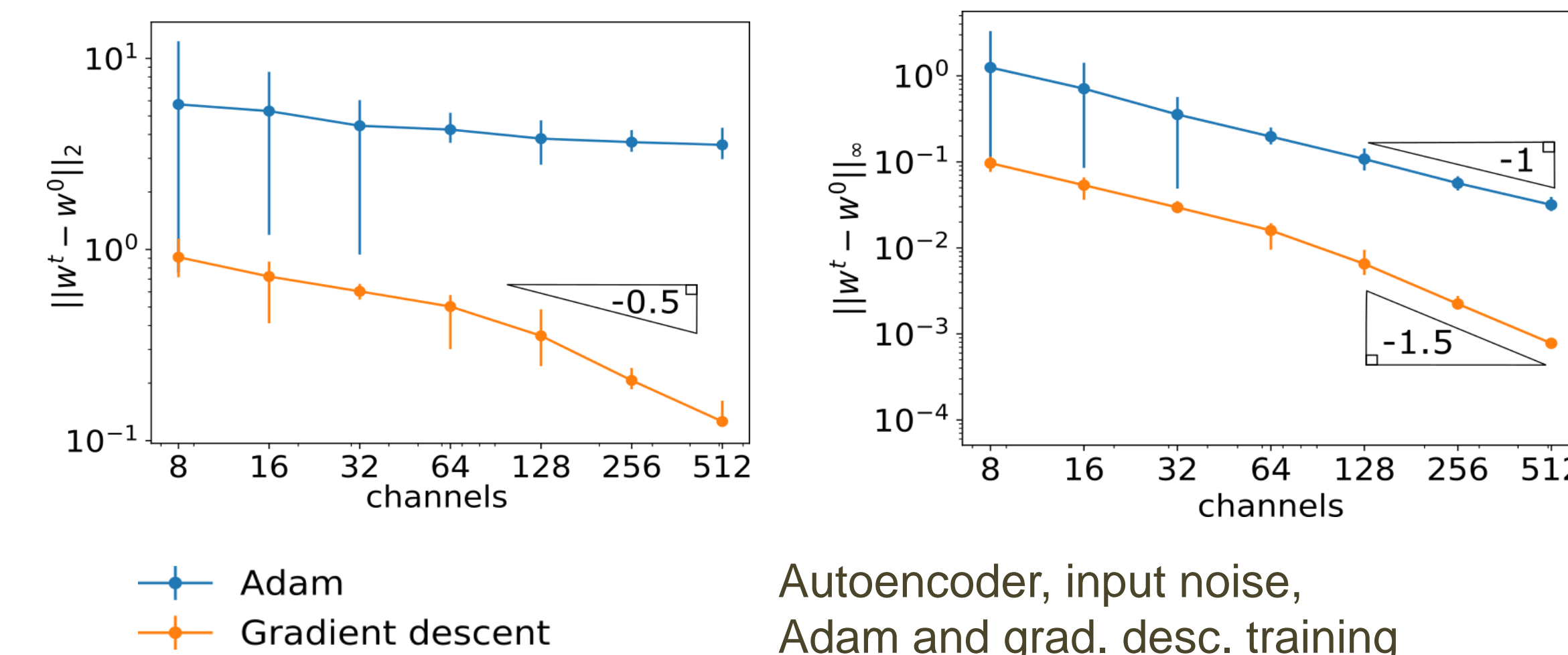
$$z^{t+1} = z^t + \eta \Theta^t(y - z^t)$$

Experiments:

Denoising performance



Change of weights throughout training



Summary

- CNNs possess a very strong bias towards clean images
- NTK gives the associated patch similarity function
- Effective degrees of freedom \ll parameters in the network
- Optimizer plays key role in implicit bias

[1] Ulyanov et al. "Deep image prior", CVPR 2018

[2] Jacot et al. "Neural tangent kernel: Convergence and generalization in neural networks", NIPS 2018

