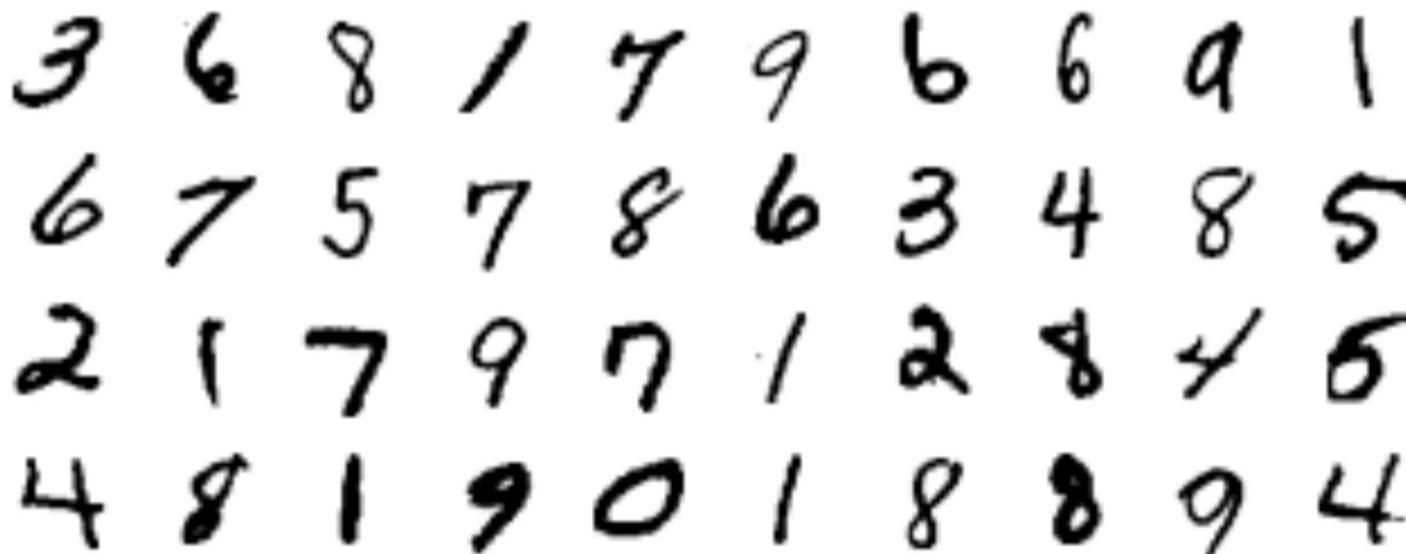


Image classification with Scattering and Dictionary learning

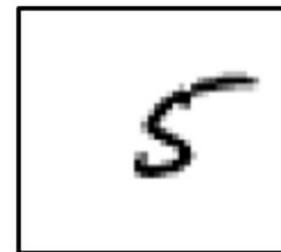
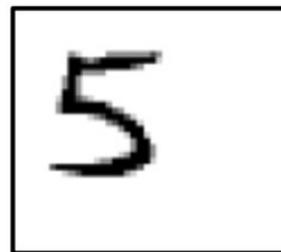
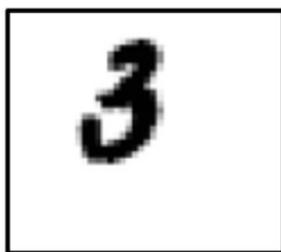
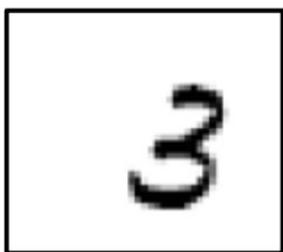
- <https://arxiv.org/abs/1910.03561>
- J. Zarka, L. Thiry, T. Angles, S. Mallat
- Pytorch code soon published on
<https://github.com/kymatio/kymatio>

Digits classification

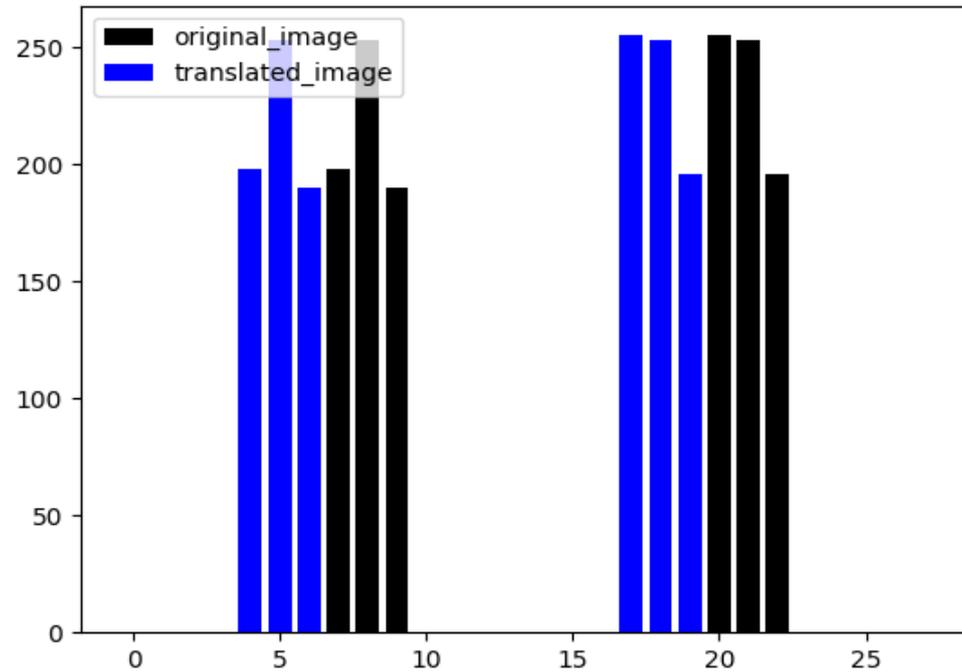
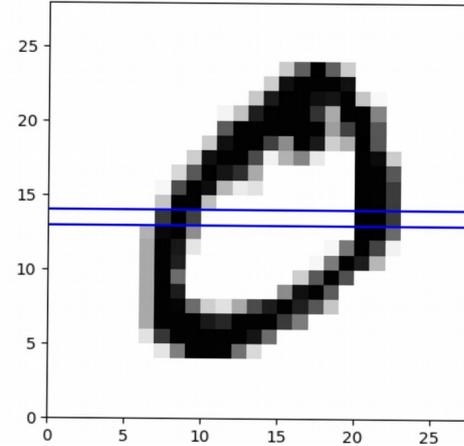
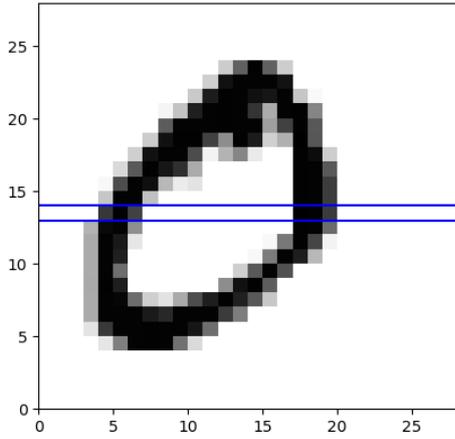
- MNIST database



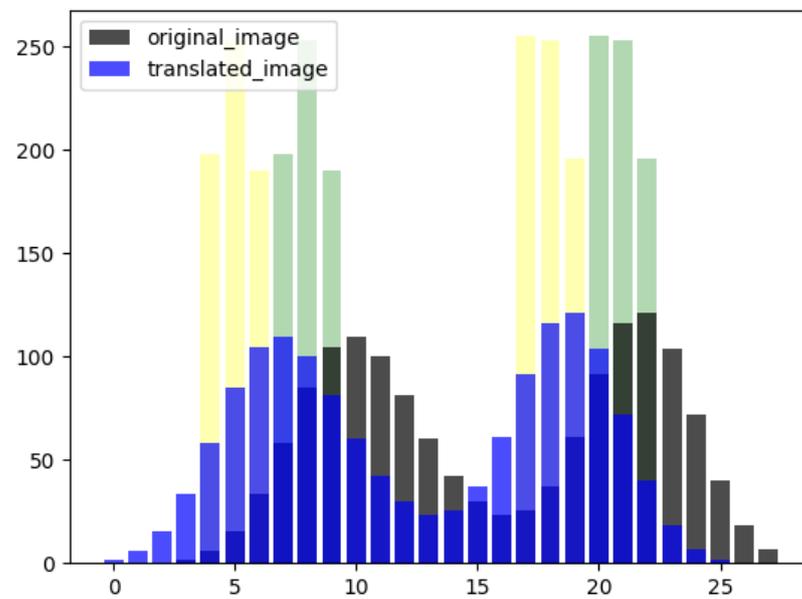
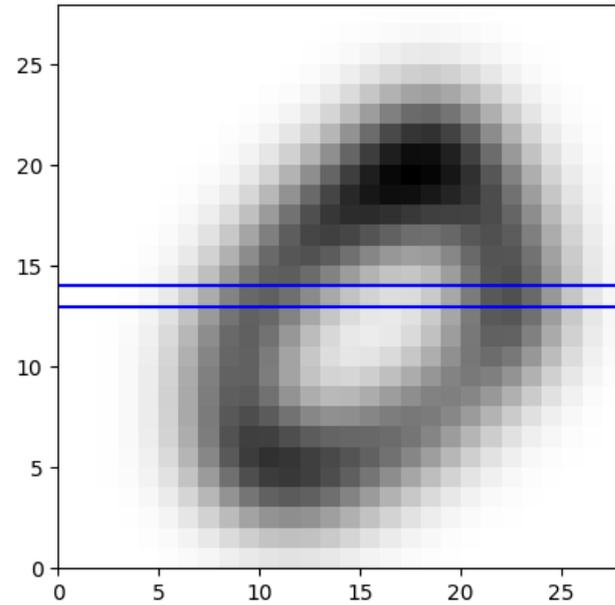
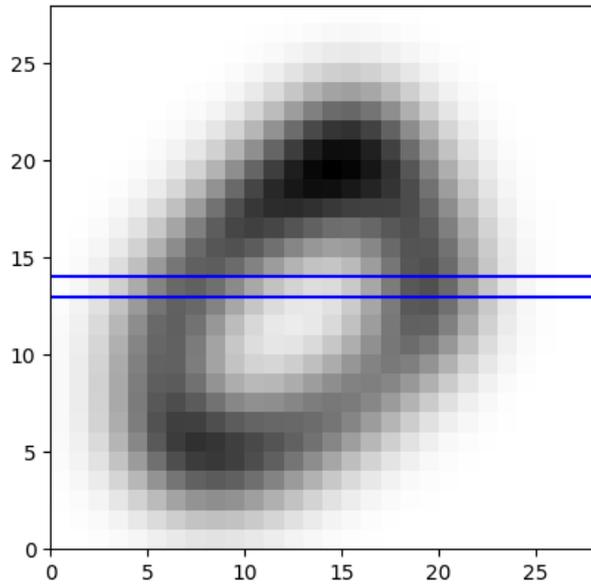
- Invariance to translations, stability to deformations



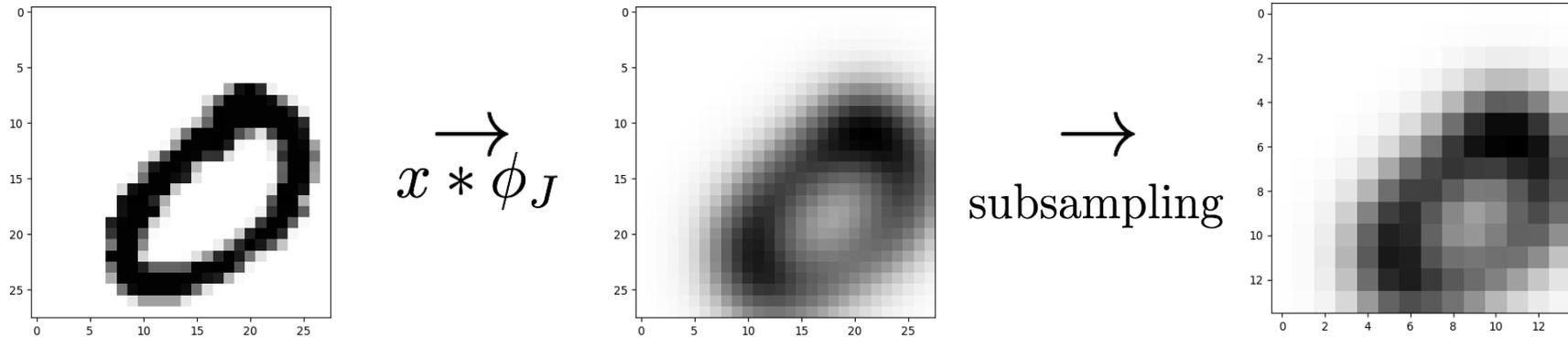
l_2 metric Instability to translations



Averaging



Stability to geometric transformations

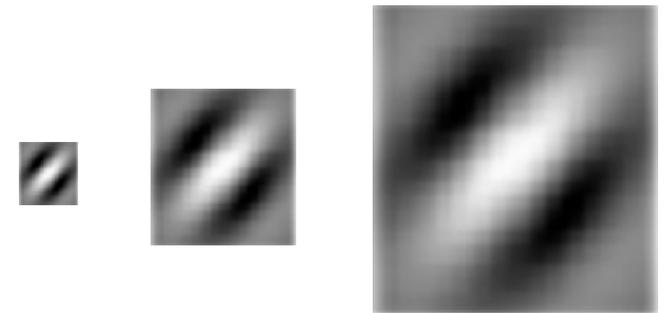
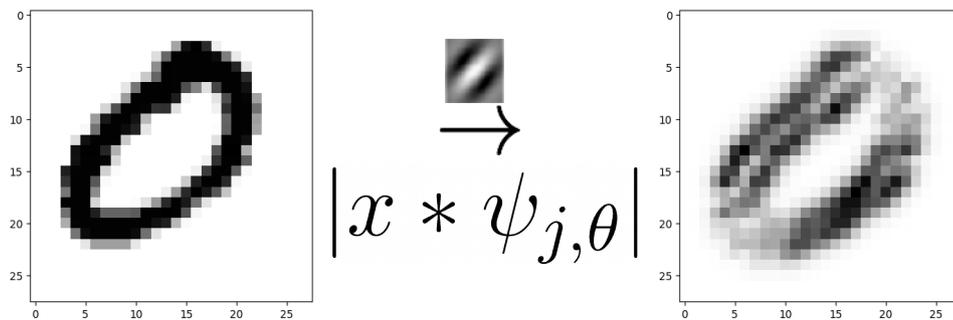


Convolution with Gaussian kernel ϕ_J :

- stable to geometric deformations
- dimensionality reduction via subsampling
- lots of details are lost

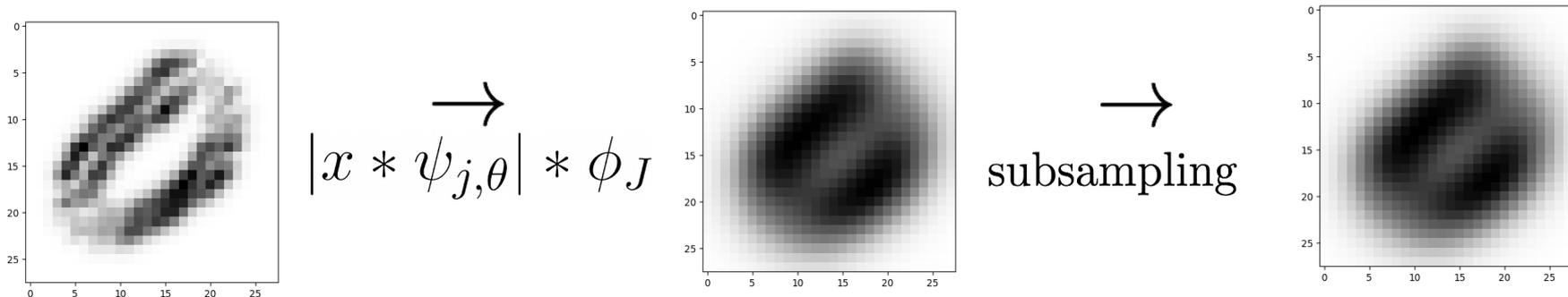
Preserving signal information

Recover information lost in averaging



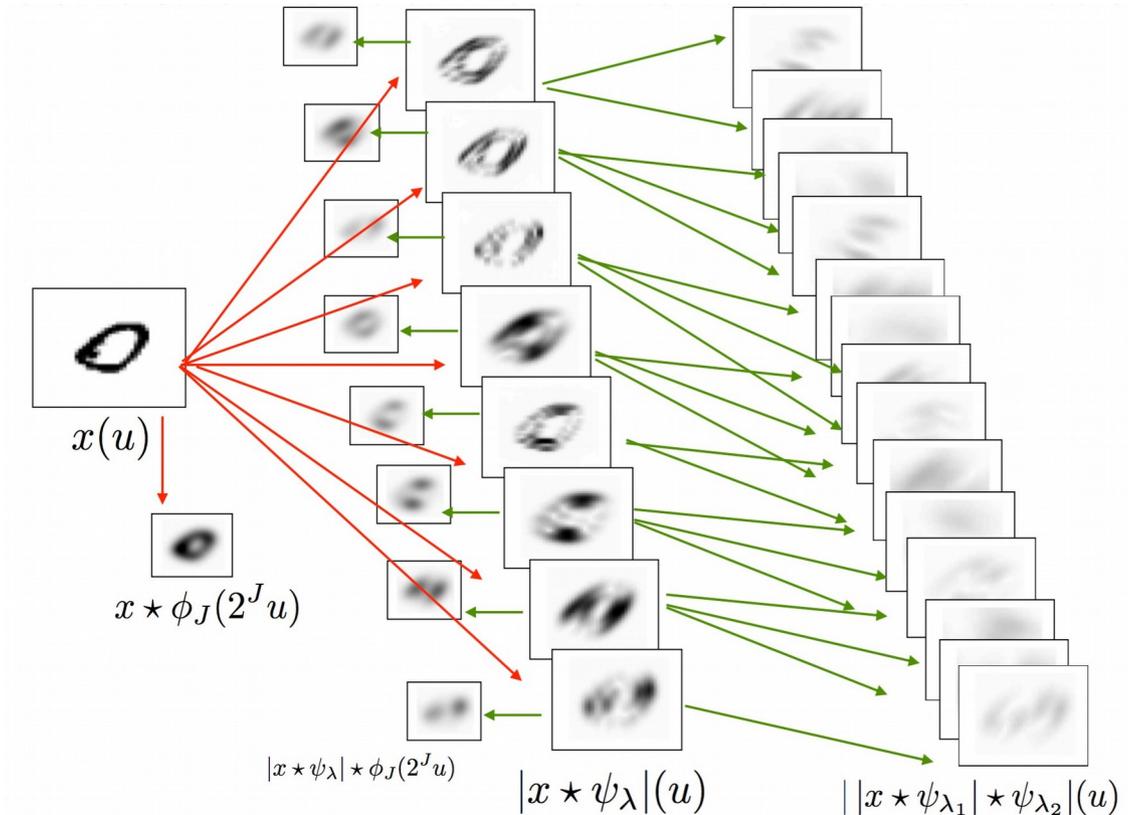
Gabor wavelets $\psi_{j,\theta}$

Stability to geometric transformations



Scattering transform

Mallat (2011), Mallat, Bruna (2012)

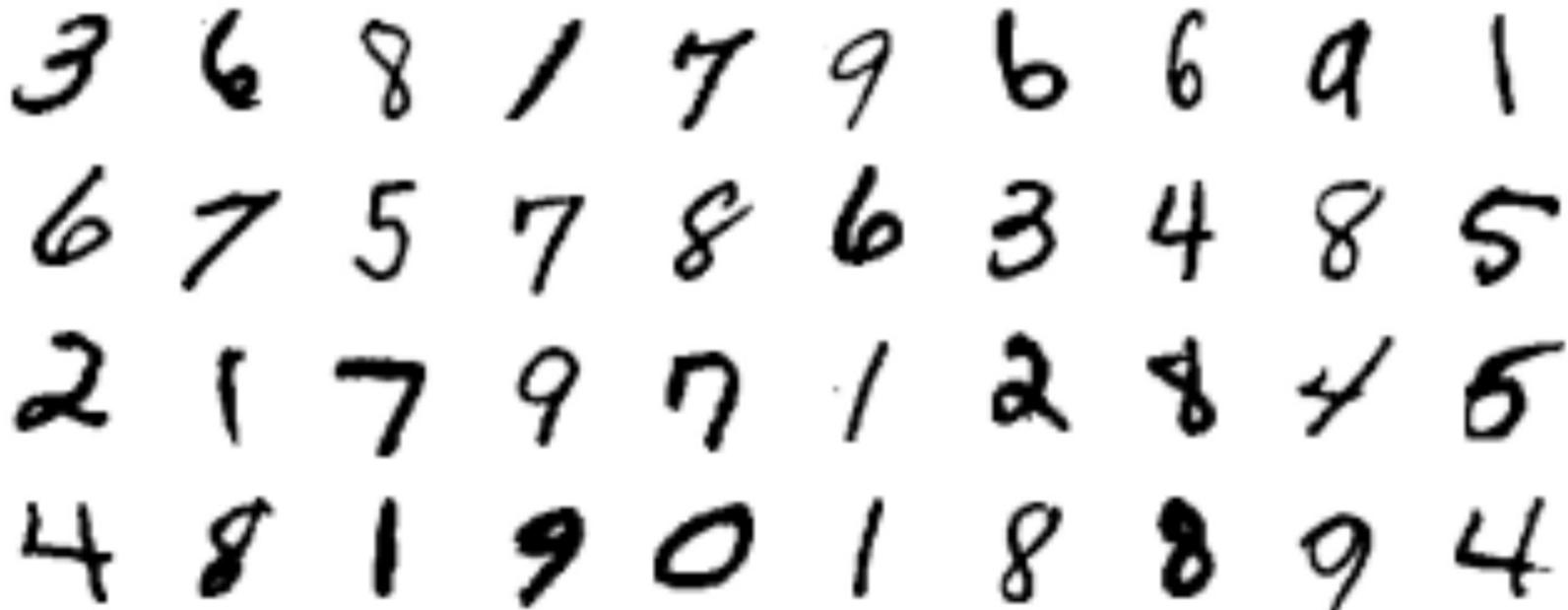


Theorem

$$\|Sx_\tau - Sx\| \leq K \|x\| \|\nabla \tau\|_\infty$$

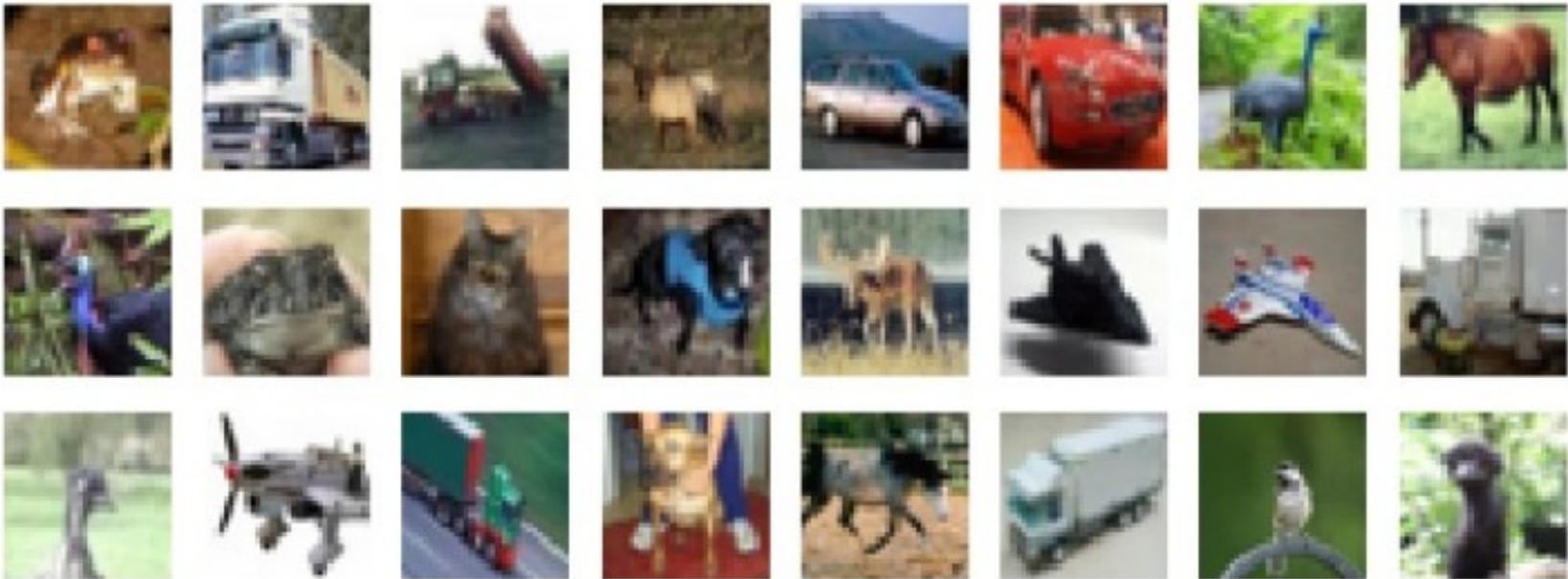
Scattering vs Deep ConvNets

Dataset	Scattering Transform	AlexNet	ResNet
MNIST 28 ² digit images 10 classes	>99 %	>99 %	>99 %



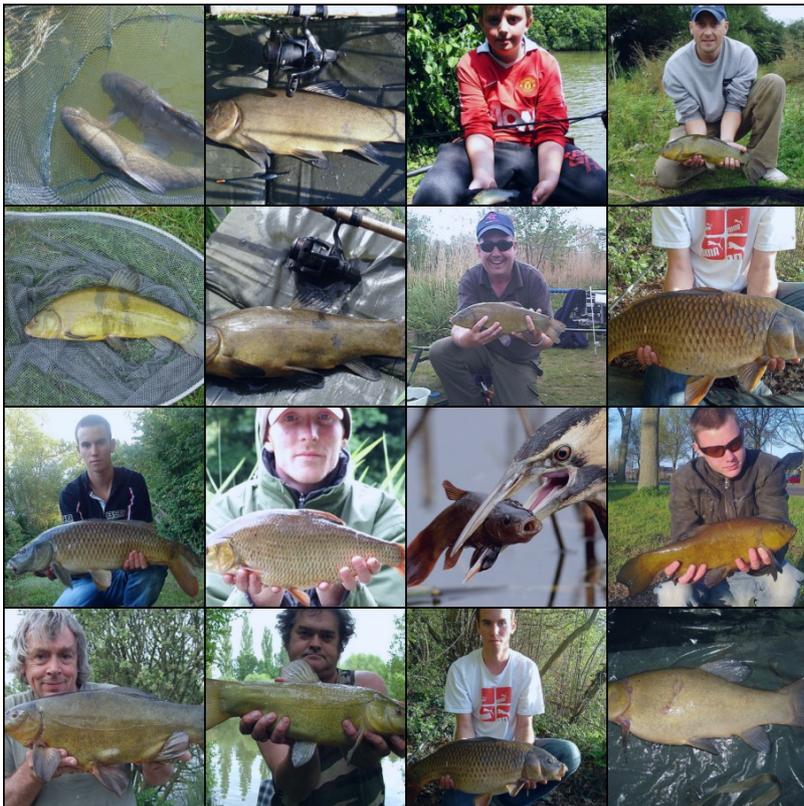
Scattering vs Deep ConvNets

Dataset	Scattering Transform	AlexNet	ResNet
CIFAR-10 32 ² object images 10 classes	84.7 %	89.1 %	95.5 %



Scattering vs Deep ConvNets

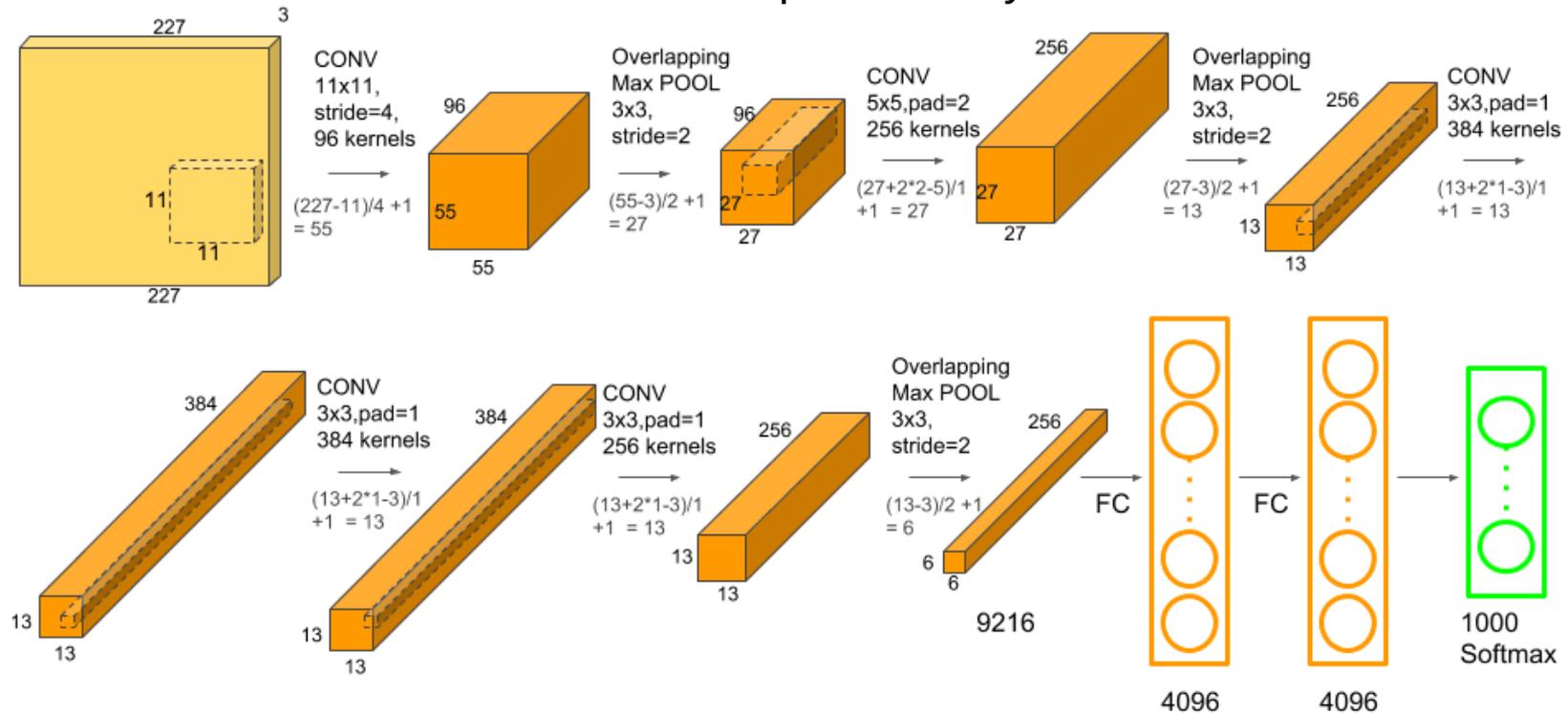
Dataset	Scattering Transform	AlexNet	ResNet
ImageNet 224 ² object images 1000 classes	61.4 %	79.1 %	94.2 %



AlexNet

Krizhevsky et al. 2012

79.1 % top5 accuracy



First layer learned filters

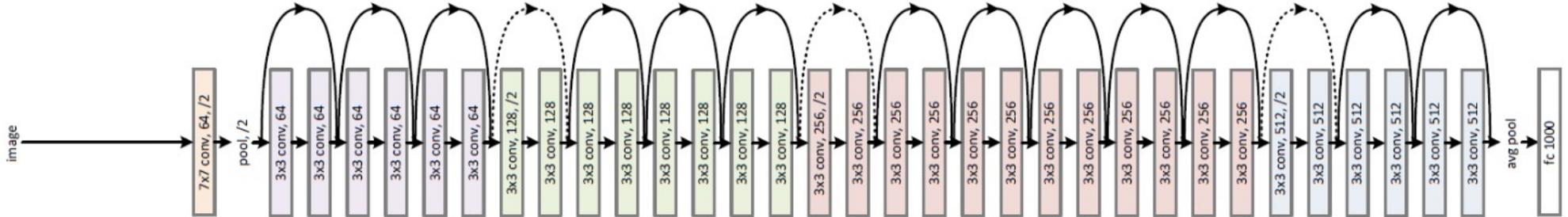


ResNet

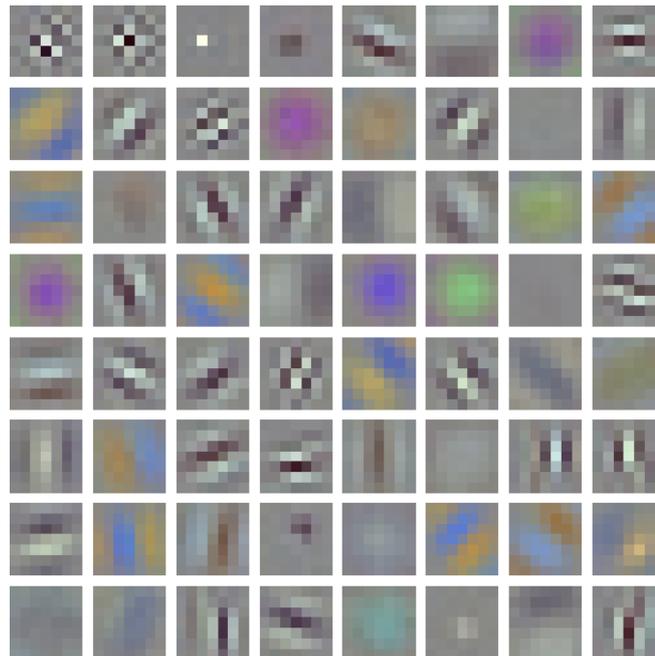
He et al. 2016

94.2 % top5 accuracy

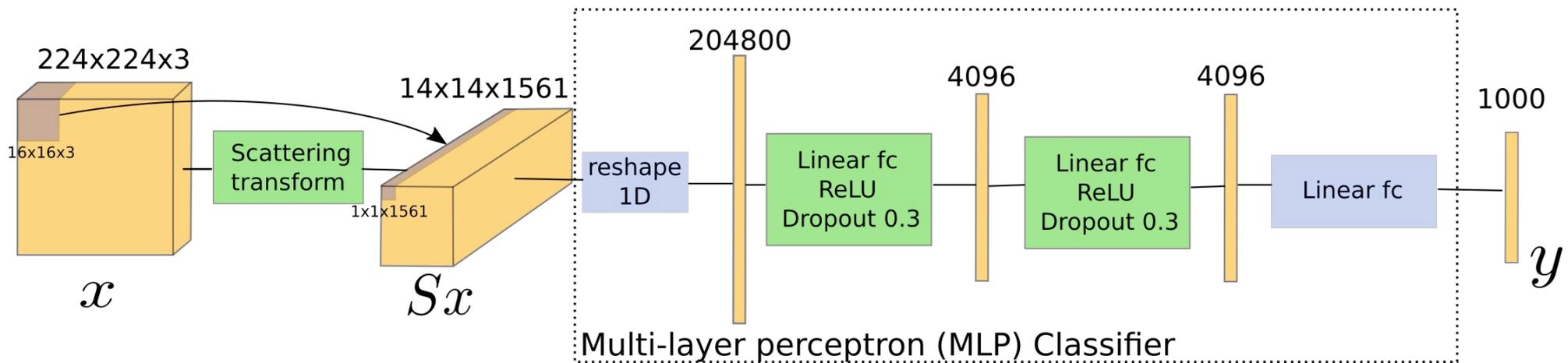
- skip connections
- up to 152 convolutional layers



First layer learned filters



Scattering : ImageNet classification



- Scattering transform has 16×16 receptive field
- Same MLP Classifier as AlexNet
- 38.1 % top1, 61.5 % top5

Research directions

What are in the convolutional layers of a Deep Networks ?

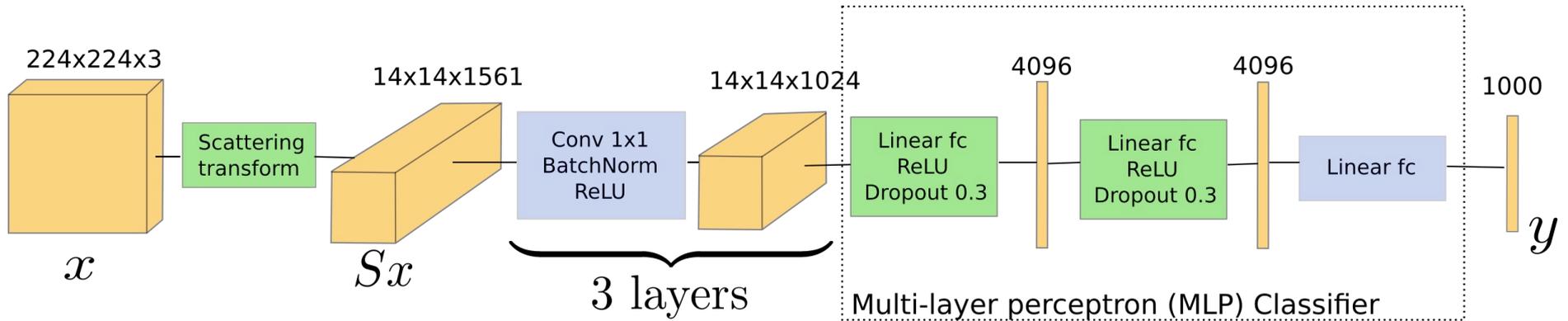
→ Visualizing and Understanding Convolutional Networks, Zeiler, Fergus 2014

What's needed to fill the gap between Scattering and DeepNets ?

→ Scaling the Scattering Transform: Deep Hybrid Networks, Oyallon et al. 2017

Shared Local Encoder

Oyallon et al. (2017)

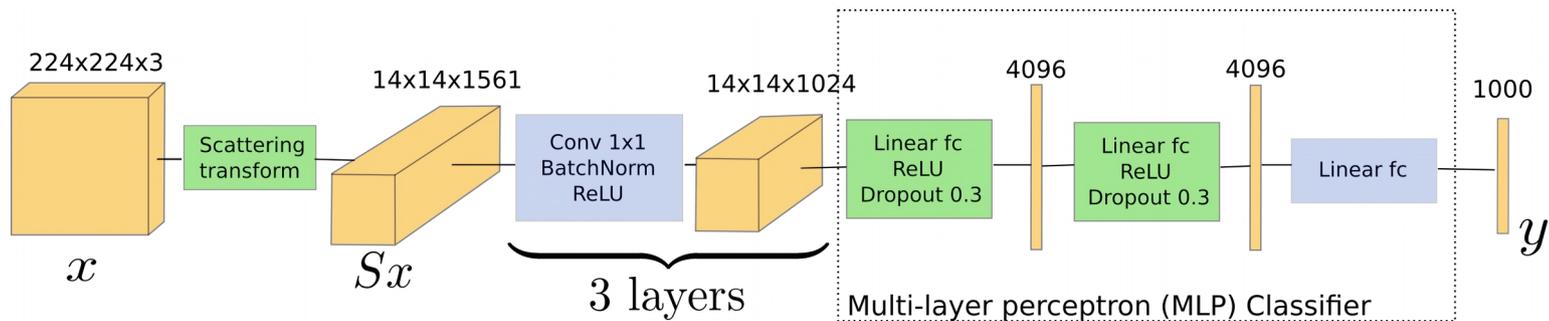


- Scattering transform
- 3 layers 1×1 convolution with ReLU and BatchNorm
- MLP Classifier

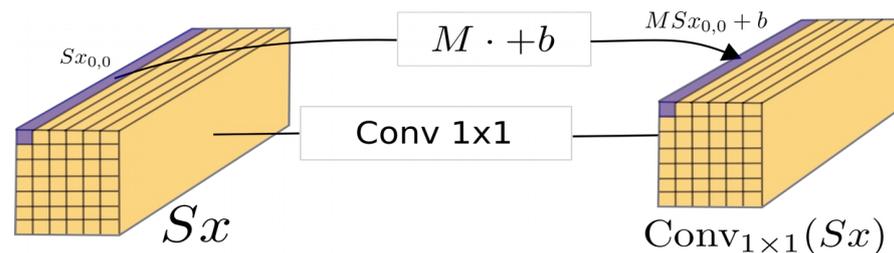
Performance on Imagenet 2012 :

- 57.8 % top1
- 79.6 % top5

l_1 sparse coding of Scattering



Convolution 1x1



Non negative sparse coding of Scattering

$$\alpha_*(D, \lambda, Sx) = \underset{\alpha \geq 0}{\operatorname{argmin}} \mathcal{L}(\alpha), \quad \mathcal{L}(\alpha) = \|D\alpha - Sx\|_2^2 + \lambda \|\alpha\|_1$$

ISTA-like algorithm

$$\alpha_{n+1} = \operatorname{ReLU} \left(\left(Id - \frac{1}{L} D^T D \right) \alpha_n + \frac{1}{L} D^T Sx - \frac{\lambda}{L} \right)$$

Supervised dictionary learning + LISTA

Mairal et al. (2008), Gregor and Lecun (2011)

SDL

$$\min_{D, \alpha, \theta} \mathcal{C}(y, f(\alpha, Sx, \theta)) + \lambda_0 \|D\alpha - Sx\|_2^2 + \lambda_1 \|\alpha\|_1$$

LISTA

$$\alpha_0 = 0, \alpha_{n+1} = \text{st}_{\lambda_{n+1}}(VSx + W\alpha_n), n = 0 \dots N - 1$$

Performance

64% top1, 86% top5

l1 Convergence

$$\frac{\mathcal{L}(\alpha_N)}{\mathcal{L}(\alpha_*)} = 4.3$$

Task Driven dictionary learning + ISTC

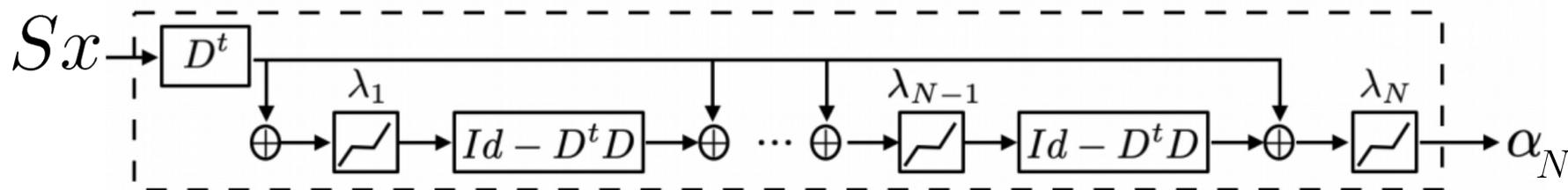
Mairal et al. (2011), ours

TDDL

$$\min_{D, \theta, \lambda} \mathcal{C}(y, f(\alpha_*(D, \lambda, Sx), \theta))$$

ISTC

$$\alpha_0 = 0, \quad \alpha_{n+1} = \text{st}_{\lambda_\infty \gamma^n} \left((Id - D^T D) \alpha_n + D^T Sx \right), \quad \lambda_\infty \gamma^N = \lambda$$



Theorem

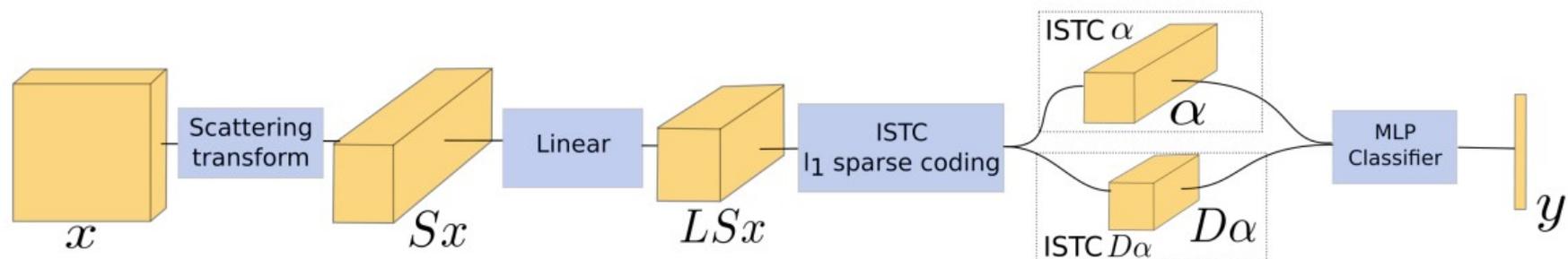
- s support size of α_*
- $\mu = \max_{m \neq m'} \langle D_m, D_{m'} \rangle$

If $s\mu \leq 1/2$ and $2s\mu < \gamma < 1$:

$$\|\alpha_n - \alpha_*\|_\infty \leq K \gamma^n$$

Scattering + ISTC classification

Architecture



ALISTA

$$\alpha_0 = 0, \alpha_{n+1} = \text{st}_{\lambda_\infty \gamma^n} \left((Id - W^T D) \alpha_n + W^T Sx \right), \langle W_m, D_m \rangle = 1$$

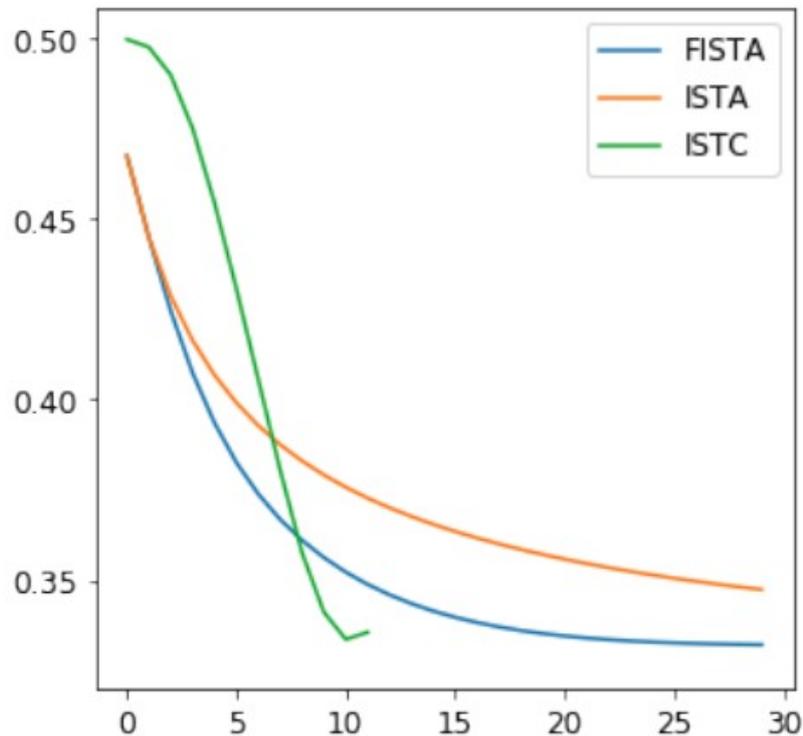
Results

l_1 algo	ISTC α	ISTC $D\alpha$	ALISTA
Top 1	58.0	55.4	61.9
Top 5	80.9	78.3	84.3

Convergence analysis

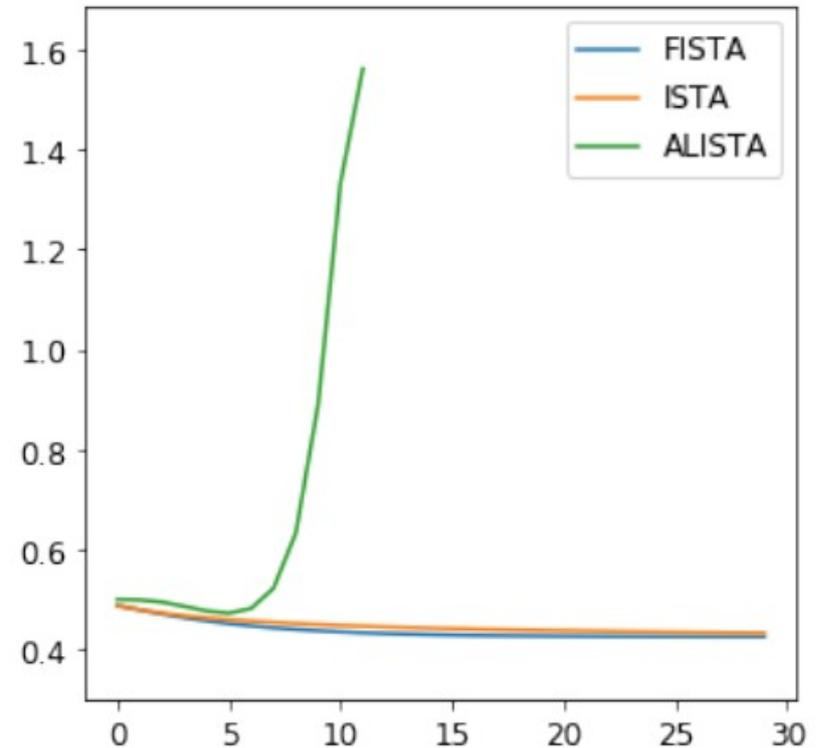
ISTC

$$\frac{\mathcal{L}(\alpha_N)}{\mathcal{L}(\alpha_*)} = 1.01$$



ALISTA

$$\frac{\mathcal{L}(\alpha_N)}{\mathcal{L}(\alpha_*)} = 3.8$$



Comments

- Improvement of 20% over Scattering alone
- Large factor λ_* , reconstruction error $\|Sx - D\alpha\|/\|Sx\| = 0.5$
- Hard to reconstruct the original image from α
- Classification works with the reconstruction $D\alpha$
55 % top1, 78 % top5
- Atoms D_m are in Scattering space, can not be visualised like usual dictionary atoms
- ISTC can be replaced by FISTA (requires classifier retraining)
- Still far from ResNet performance
→ what's in LISTA with SDL ?

Questions ?