# FlexConv: Continuous Kernel Convolutions With Differentiable Kernel Sizes

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\* equal contribution

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## **Abstract**

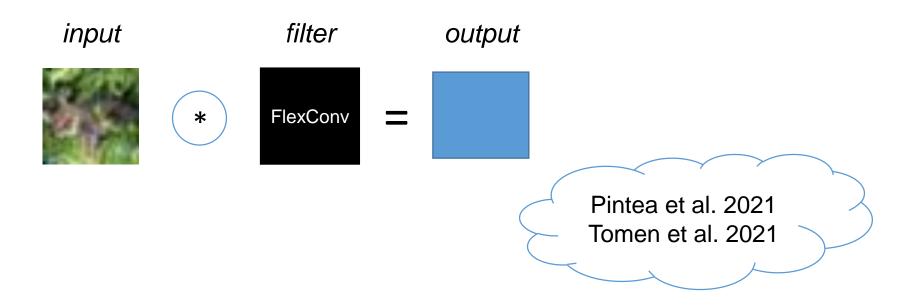
FlexConv replaces convolutions

#### **Contributions**

- 1. Learns kernel size
- 2. High bandwidth kernel
- 3. Adaptable to data resolution

#### Method

## Learning kernel size



Silvia L Pintea, Nergis Tomen, Stanley F Goes, Marco Loog, and Jan C van Gemert. Resolution learning in deep convolutional networks using scale-space theory. arXiv preprint arXiv:2106.03412, 2021.

Nergis Tomen, Silvia-Laura Pintea, and Jan Van Gemert. Deep continuous networks. In Marina Meila and Tong Zhang (eds.), Proceedings of the 38th International Conference on Machine Learning, volume 139 of Proceedings of Machine Learning Research, pp. 10324–10335. PMLR, 18–24 Jul 2021.

## Method FlexConv Fathony et al. 2021 Sitzmann et al. 2020 0 MLP

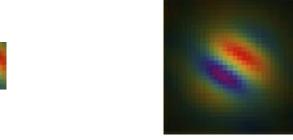
Rizal Fathony, Anit Kumar Sahu, Devin Willmott, and J Zico Kolter. Multiplicative filter networks. In International Conference on Learning Representations, 2021.

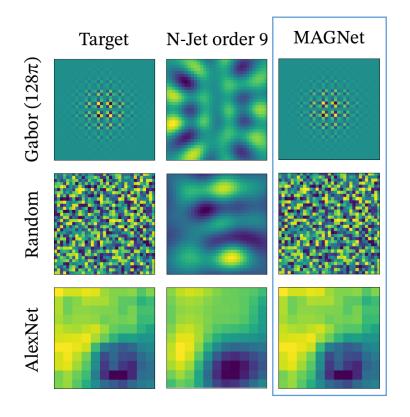
Vincent Sitzmann, Julien Martel, Alexander Bergman, David Lindell, and Gordon Wetzstein. Implicit neural representations with periodic activation functions. Advances in Neural Information Processing Systems, 33, 2020

#### Contribution

## High bandwidth kernels

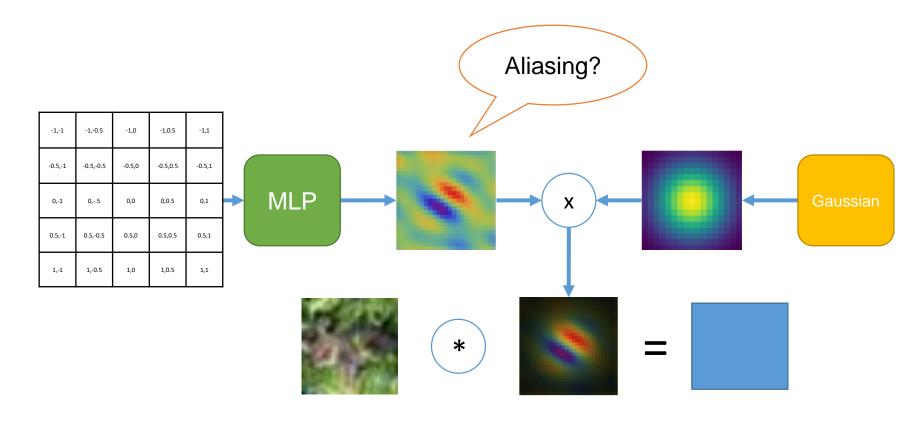
Related works FlexConv





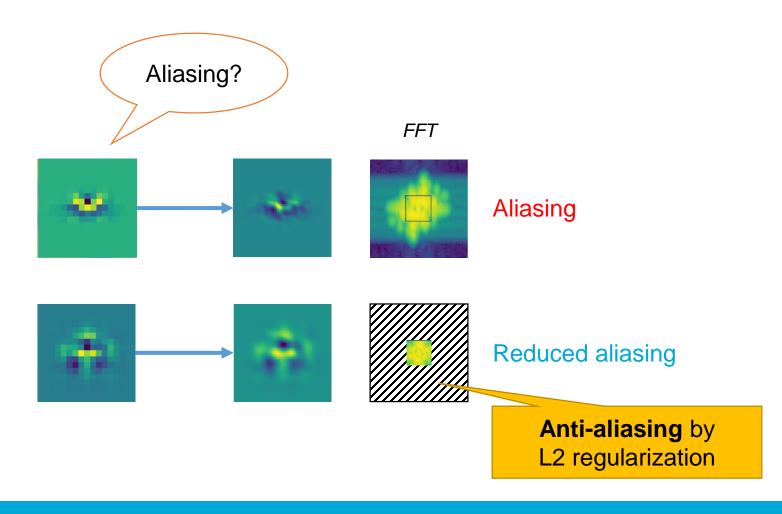
### Contribution

## Adapting resolution



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## Results

#### Upscaled image classification

CIZE	CIFAR-10 Acc.		
SIZE	16 px	$\Delta_{16px}$ 32 px	
0.66м	$85.8 \pm 0.2$	$-31.6 \pm 1.3$	
0.67м	<b>86.4</b> ± 0.4	-34.4 ± 14.3	
0.67M	<b>86.5</b> ± 0.1	-3.8 ± 2.0 -3.3 ± 0.3	
	0.67м	SIZE $16 \text{ px}$ $0.66\text{M}$ $85.8 \pm 0.2$ $0.67\text{M}$ $86.4 \pm 0.4$ $0.67\text{M}$ $86.5 \pm 0.1$	

#### Sequence classification

MODEL	SIZE	CT	SC	SC_RAW
GRU-ODE	89K	96.2	44.8	~10.0
GRU- $\Delta t$	89K	97.8	20.0	~10.0
GRU-D	89K	95.9	23.9	~10.0
ODE-RNN	89K	97.1	93.2	~10.0
NCDE	89K	98.8	88.5	~10.0
FlexTCN-2	105sk	99.53	97.10	88.03
FlexTCN-4	239K	99.53	97.73	90.45
FlexTCN-6	373K	99.53	97.67	91.73

#### Image classification

Model	SIZE	CIFAR-10 Acc.	TIME (SEC/EPOCH)
CIFARResNet-44	0.66м	92.9*†	22
DCN- $\sigma^{ji}$	0.47M	$89.7 \pm 0.3*$	-
N-Jet-CIFARResNet32	0.52M	$92.3 \pm 0.3*$	-
N-Jet-ALLCNN	1.07M	$92.5 \pm 0.1*$	-
FlexNet-16	0.67м	$92.2 \pm 0.1$	127

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