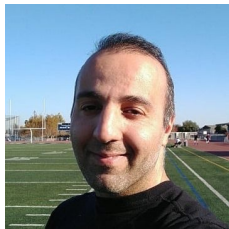


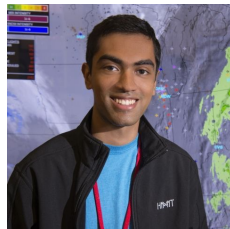
# Mind the Pad – CNNs can Develop Blind Spots



Bilal  
Alsallakh



Narine  
Kokhlikyan



Vivek  
Miglani



Jun  
Yuan



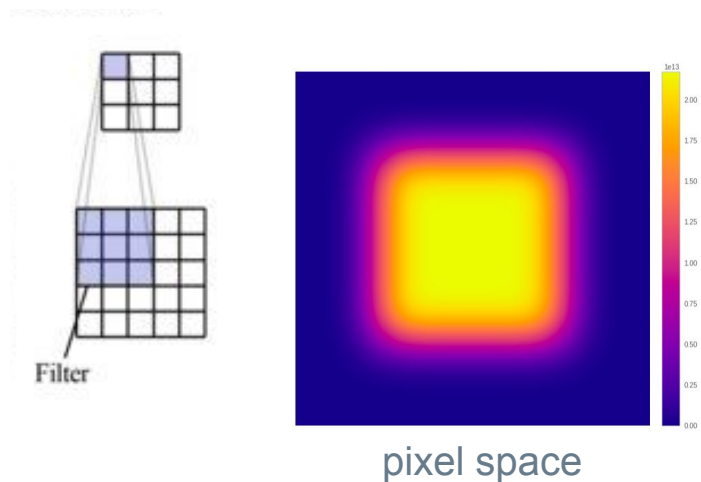
Orion  
Reblitz-Richardson

FACEBOOK

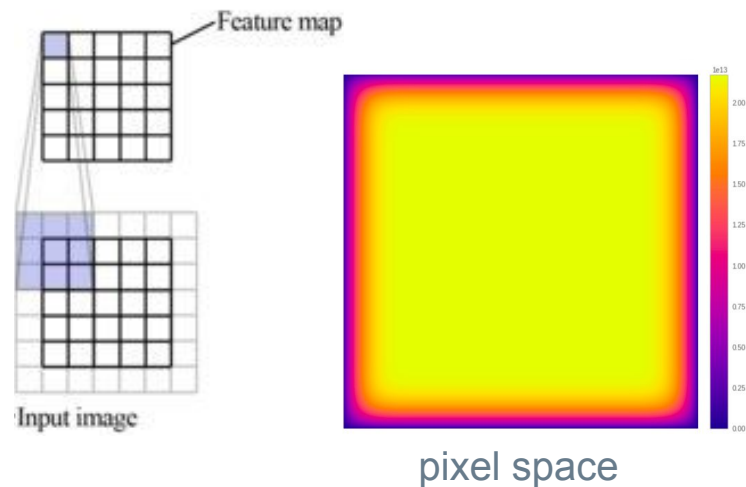


# Why Padding?

No Padding



SAME 0-Padding



Color represents the number of times an input pixel is utilized by VGG-19

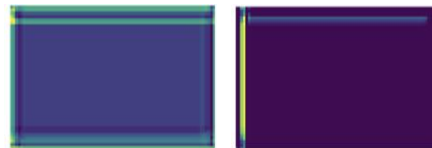
# Problems with Zero Padding

0	0	0	0	0	0
0	35	19	25	6	0
0	13	22	16	53	0
0	4	3	7	10	0
0	9	8	1	3	0
0	0	0	0	0	0

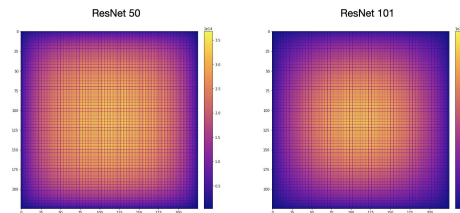
It can **skew the learnt filters** if applied unevenly.



It can **induce line artifacts** in feature maps.



It can **incur mild foveation** in the pixel space.



# Problems with Zero Padding

0	0	0	0	0	0
0	35	19	25	6	0
0	13	22	16	53	0
0	4	3	7	10	0
0	9	8	1	3	0
0	0	0	0	0	0

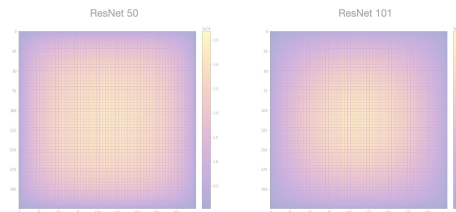
It can **skew the learnt filters** if applied unevenly.



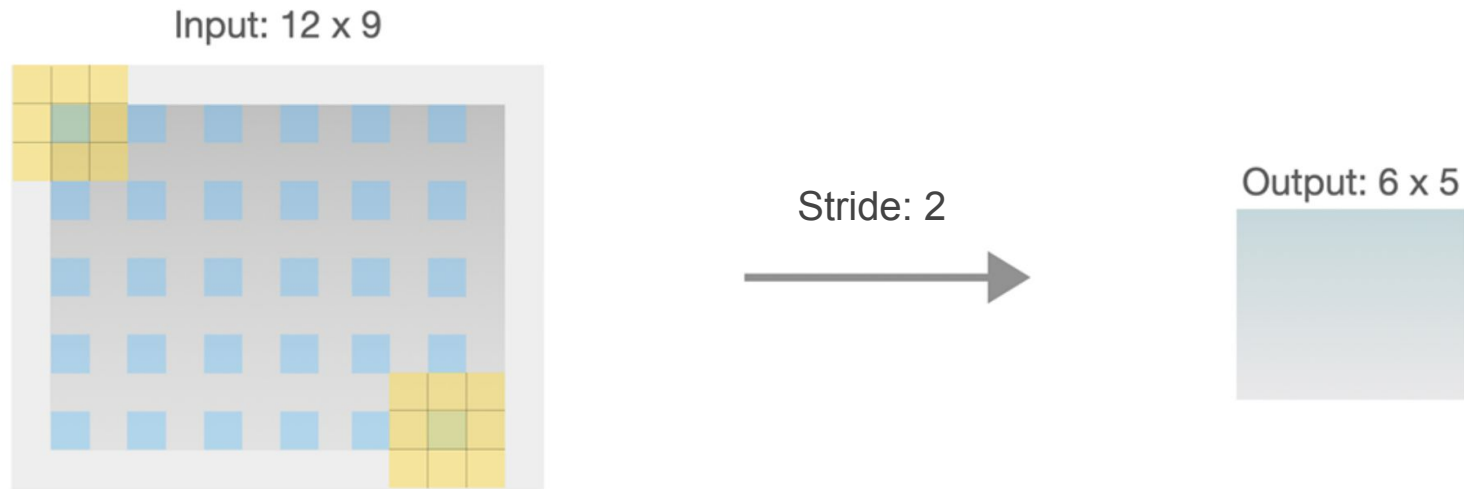
It can **induce line artifacts** in feature maps.



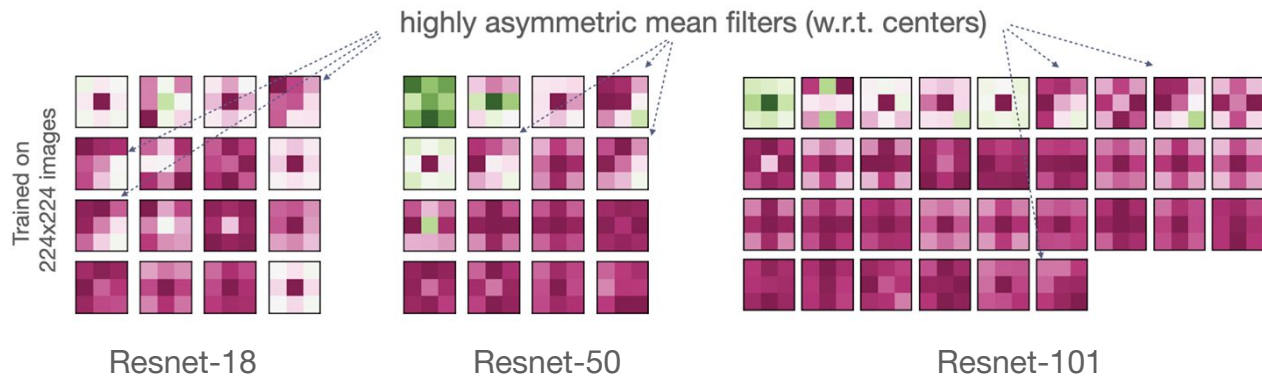
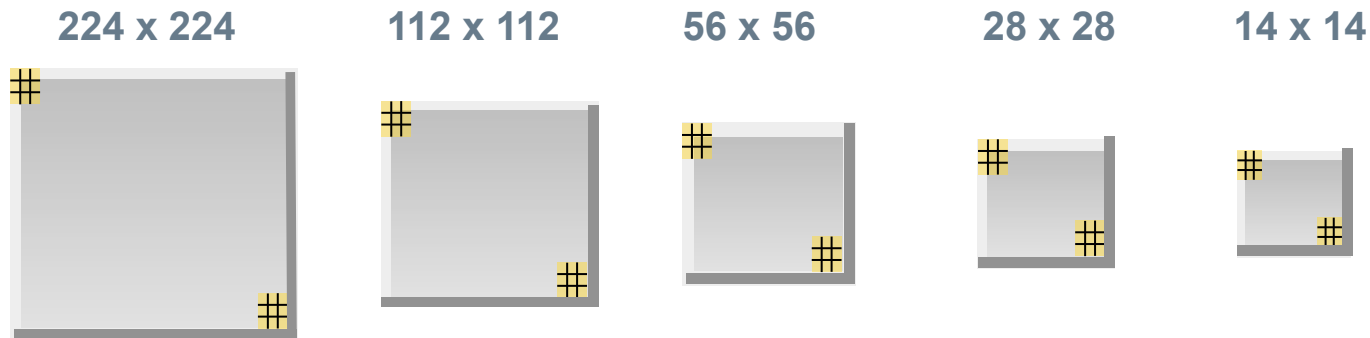
It can **incur mild foveation** in the pixel space.



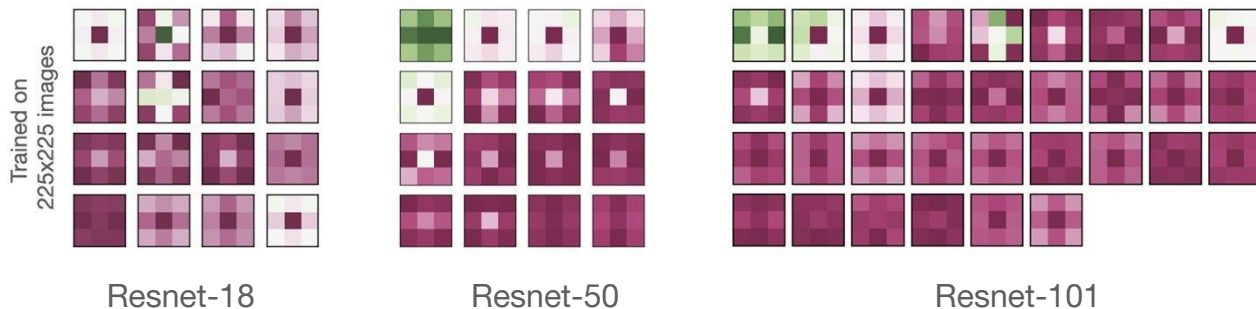
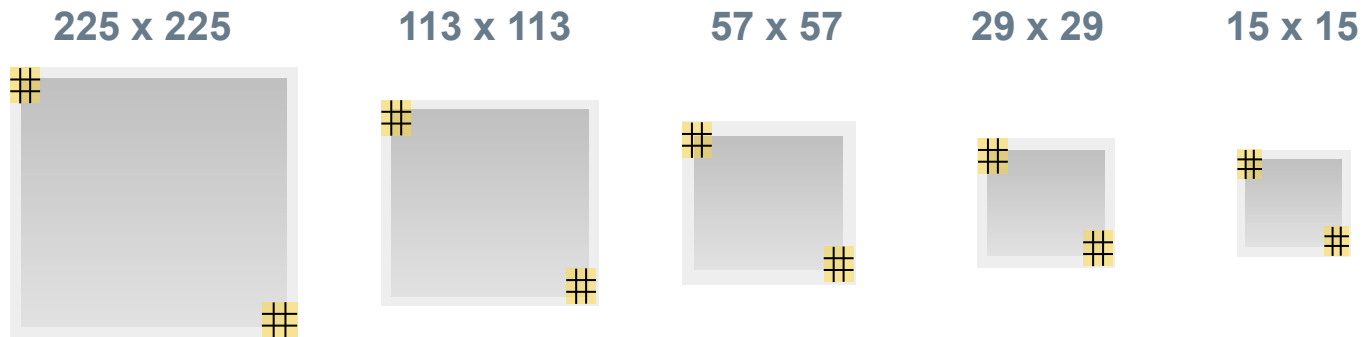
**Strided convolution** might consume 0-padding unevenly:



## Uneven padding $\Rightarrow$ skewed filters:



Even padding  $\Rightarrow$  symmetric mean filters:



## Symmetric filters $\Rightarrow$ higher accuracy (and shift invariance)

Top-1 (and Top-5) Accuracy on ImageNet

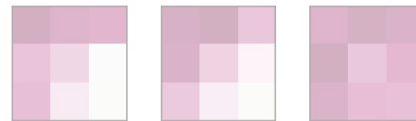
Input Size	MobileNet	ResNet-18	ResNet-34	ResNet-50	ResNet-101
224 $\times$ 224	68.19 (88.44)	69.93 (89.22)	73.30 (91.42)	75.65 (92.47)	77.37 (93.56)
225 $\times$ 225	68.80 (88.78)	70.27 (89.52)	73.72 (91.58)	76.01 (92.90)	77.67 (93.81)



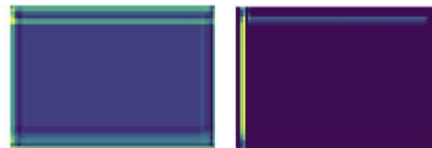
# Problems with Zero Padding

0	0	0	0	0	0
0	35	19	25	6	0
0	13	22	16	53	0
0	4	3	7	10	0
0	9	8	1	3	0
0	0	0	0	0	0

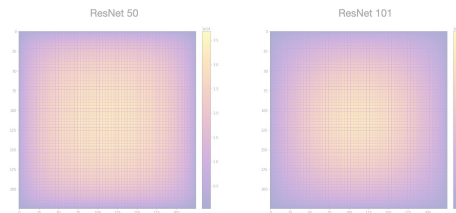
It can **skew the learnt filters** if applied unevenly.

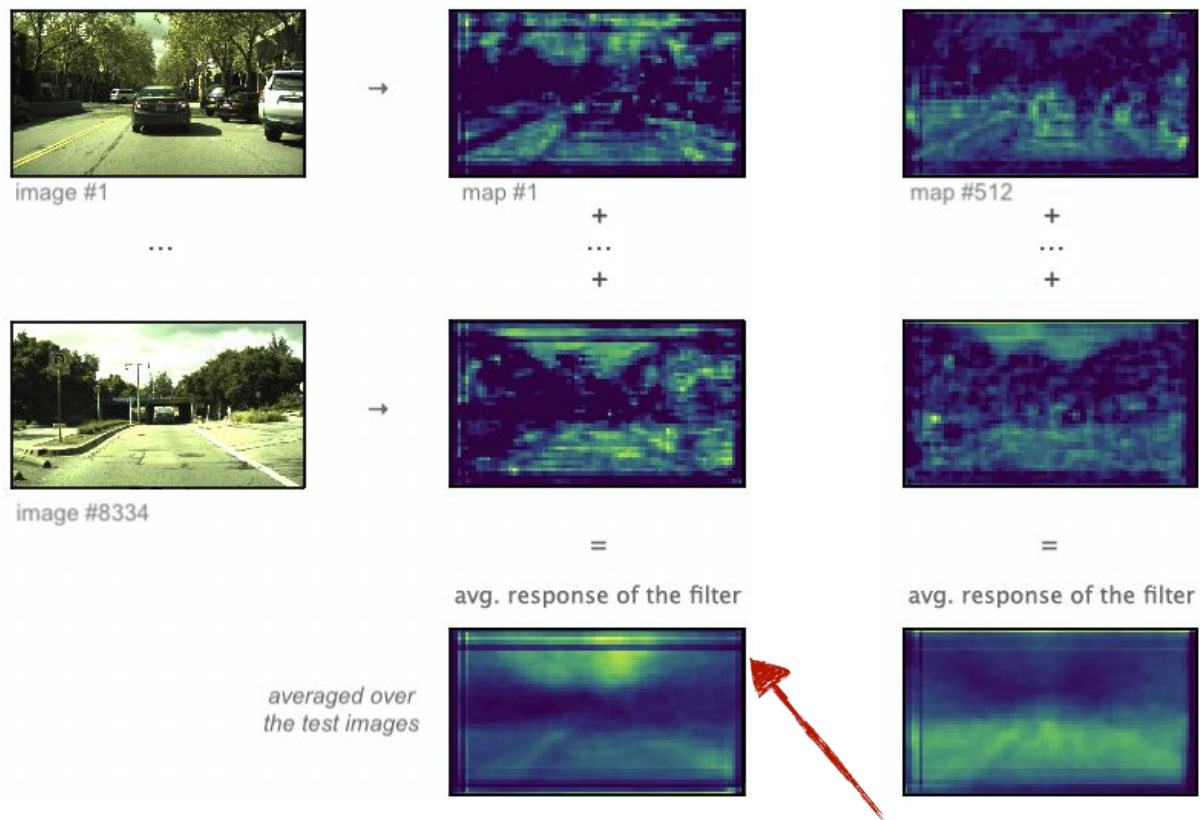


It can **induce line artifacts** in feature maps.



It can **incur mild foveation** in the pixel space.

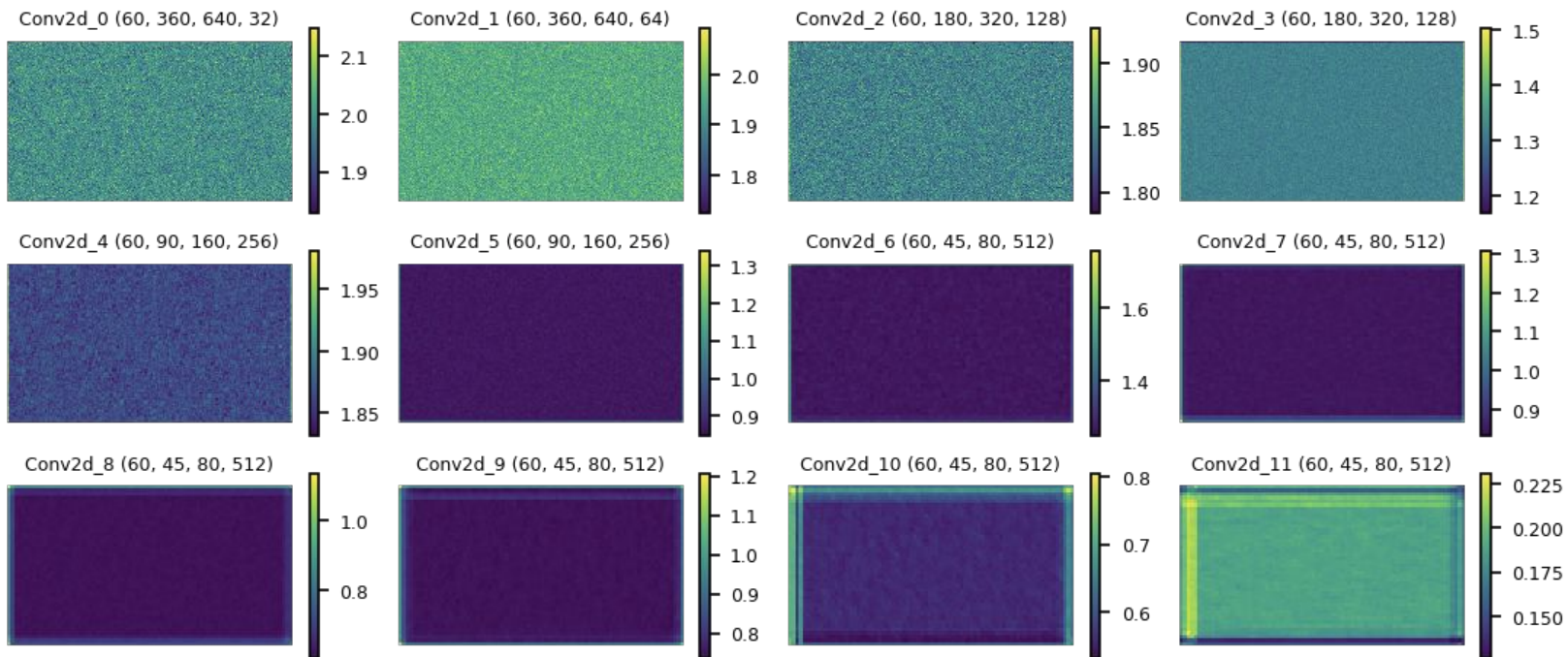




**Artificially suppressed activation at certain locations**

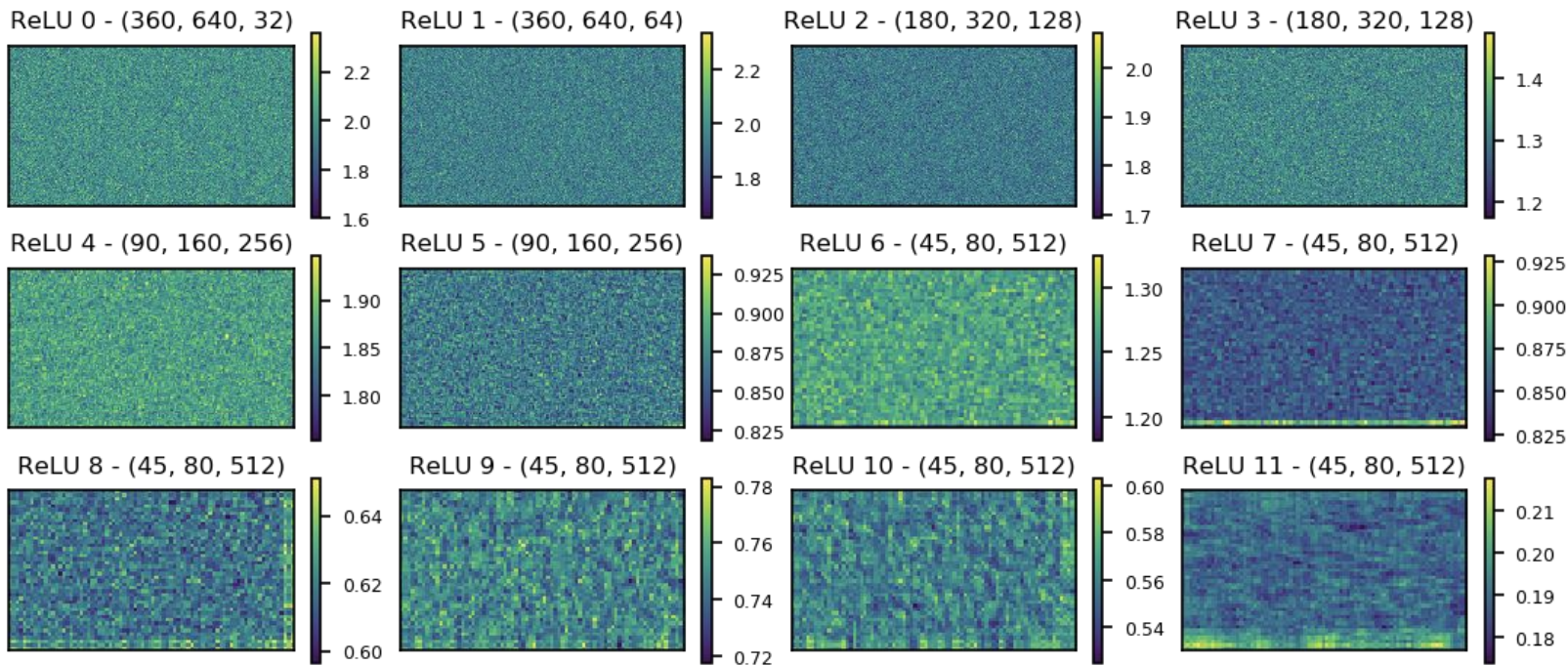
# Feature maps under 0-padding

Mean filter response per layer (with random inputs)

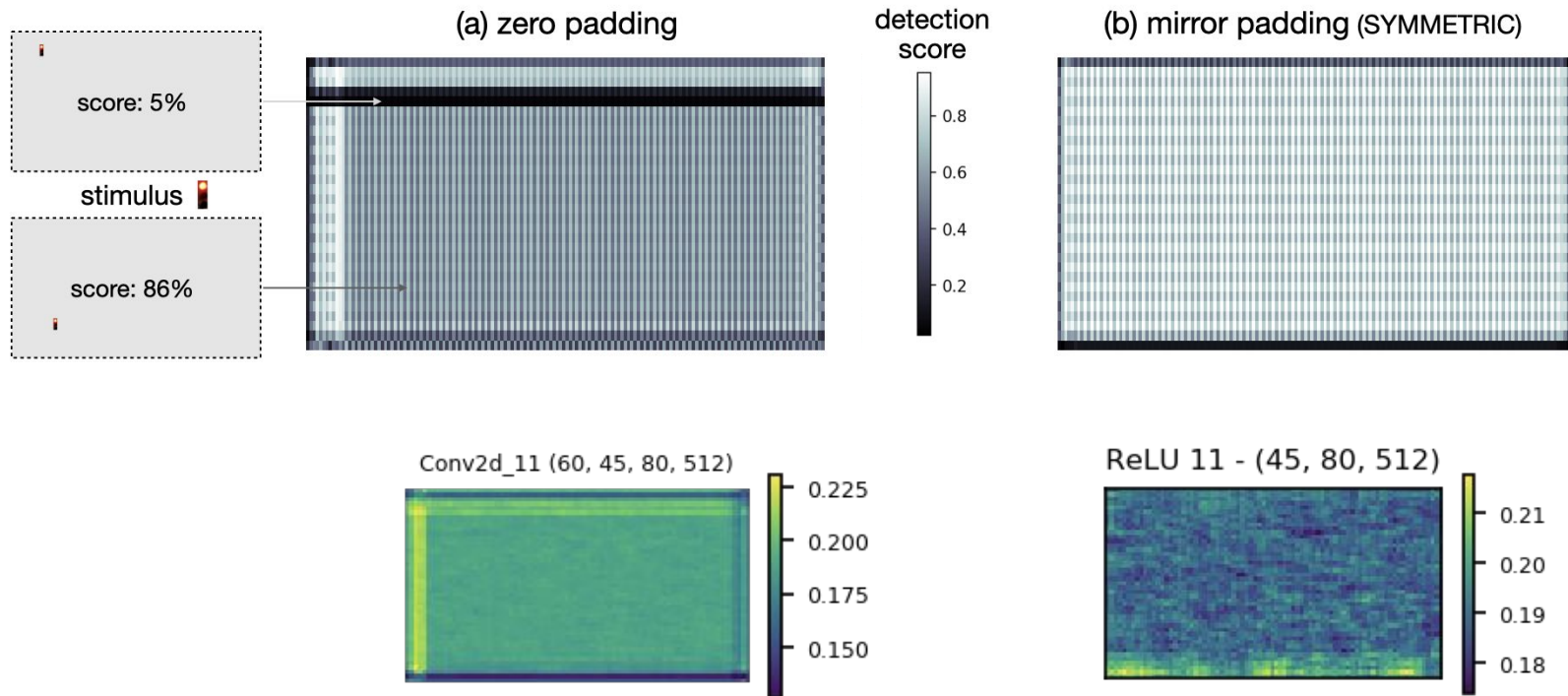


# Feature maps under mirror padding

Mean filter response per layer (with random inputs)



# Stimulus Detectability at Different Locations





## Mitigated artifacts $\Rightarrow$ higher accuracy (and shift invariance)



44% (shifted upwards)



7% (baseline)



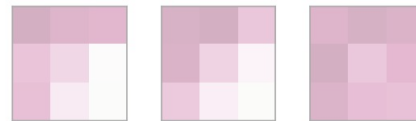
82% (shifted downwards)

Average Precision (AP)	AP@.20IOU	AP@.50IOU	AP@.75IOU
Zero Padding	80.24%	49.58%	3.7%
Mirror Padding	83.20%	57%	8.44%

# Problems with Zero Padding

0	0	0	0	0	0
0	35	19	25	6	0
0	13	22	16	53	0
0	4	3	7	10	0
0	9	8	1	3	0
0	0	0	0	0	0

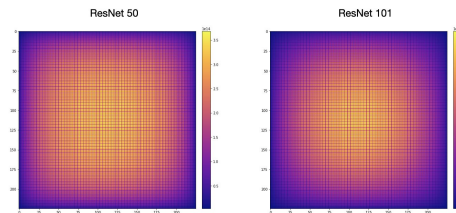
It can **skew the learnt filters** if applied unevenly.



It can **induce line artifacts** in feature maps.

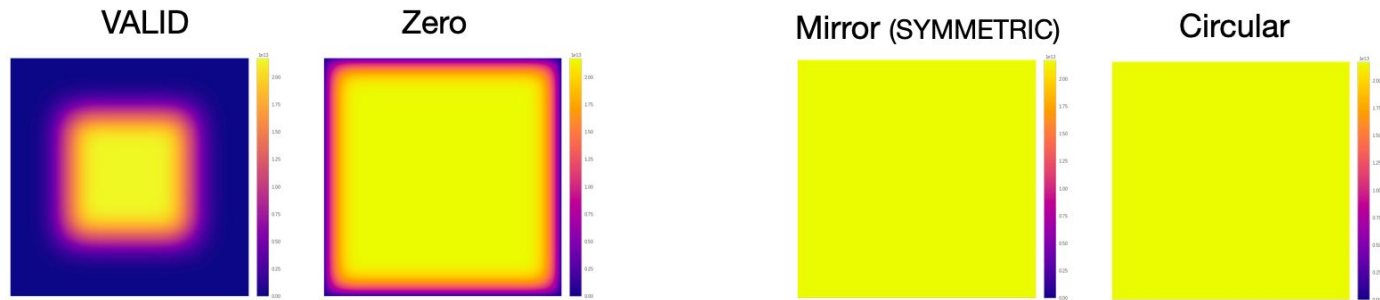


It can **incur mild foveation** in the pixel space.

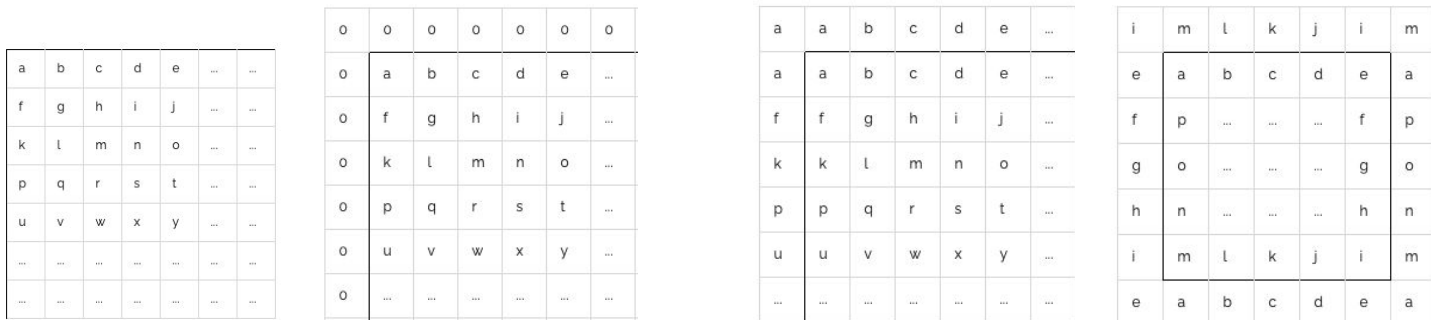


# SAME 0-padding mildly marginalizes the periphery:

Input Space (512x512) of VGG-19



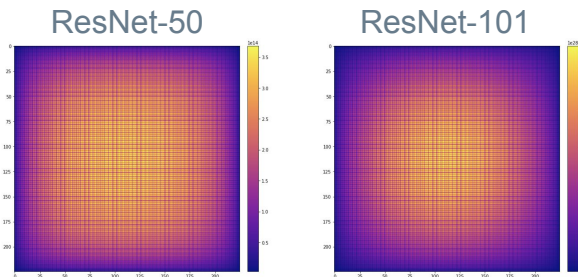
Illustrations on small inputs and one layer



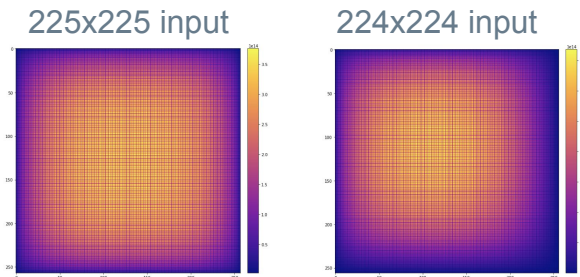


# SAME 0-padding mildly marginalizes the periphery:

Deeper CNN => more “foveation”

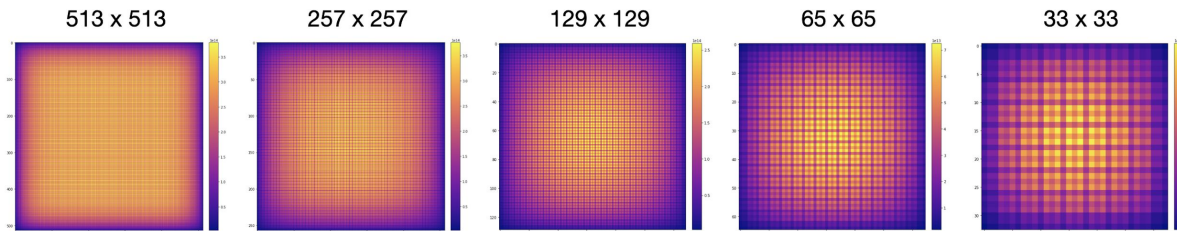


Uneven padding => skewed effects



Boundary marginalization is more extensive in smaller inputs (relative to input size):

computed for  
ResNet-50



## Summary

Zero padding can:

- **Skew the learned filters** during downsampling.

Consider eliminating uneven padding.

Consider rigid kernels (maxpool, avgpool).

- **Induce line artifacts** in the feature maps.
- **Marginalize the periphery** in the pixel space.

Circular or symmetric mirror padding:

- Mitigate these issues.

Might fit your task better - worth trying.

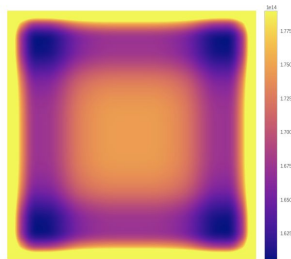


[mind-the-pad.github.io](https://mind-the-pad.github.io)

## Backup Slides

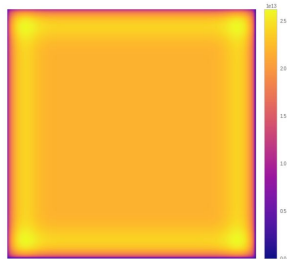
# Other Padding Schemes Can Also Incur “Foveation” Effect Under Different Padding Schemes

Replicate (5x5 kernel)



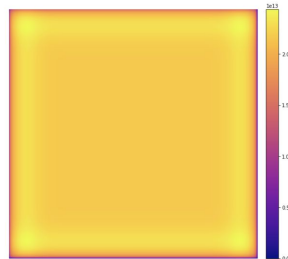
a	a	a	b	c	d	e
a	a	a	b	c	d	e
a	a	a	b	c	d	e
f	f	f	g	h	i	j
k	k	k	l	m	n	o
p	p	p	q	r	s	t
u	u	u	v	w	x	y

Mirror (REFLECT)



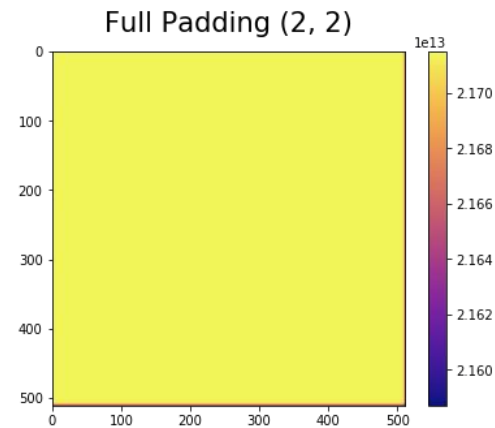
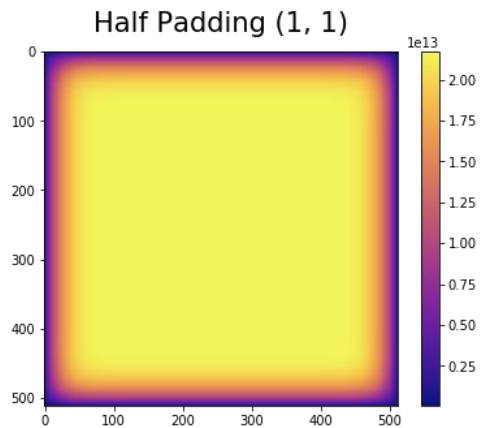
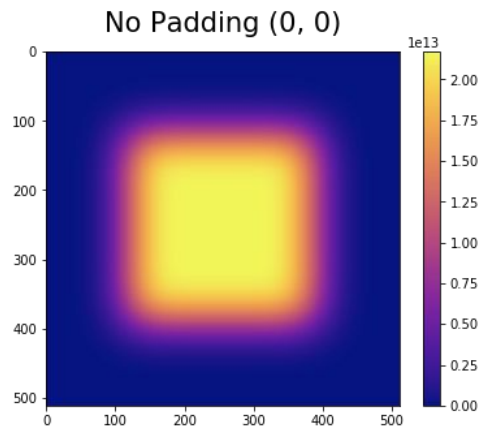
g	f	g	h	i	j
b	a	b	c	d	e
g	f	g	h	i	j
l	k	l	m	n	o
q	p	q	r	s	t
v	u	v	w	x	y

PartialConv



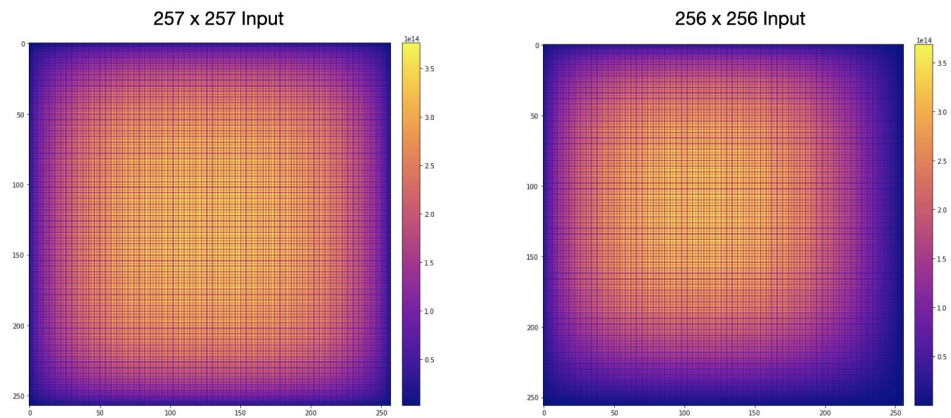
	a	b	c	d	e
	f	g	h	i	j
	k	l	m	n	o
	p	q	r	s	t
	u	v	w	x	y

# Full Padding



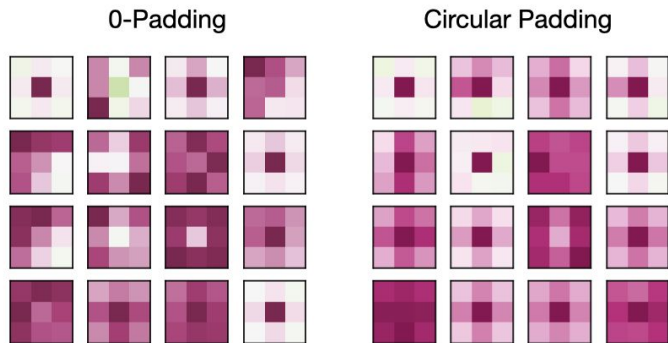
# Uneven Application of Padding

Color represents the number of time an input pixel is utilized in ResNet-50

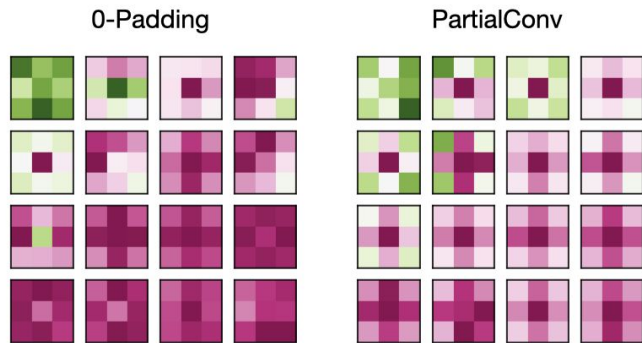


# Padding Mode and Symmetry of Average Filter

Input size is 224x224, leading to uneven padding at every downsampling layer



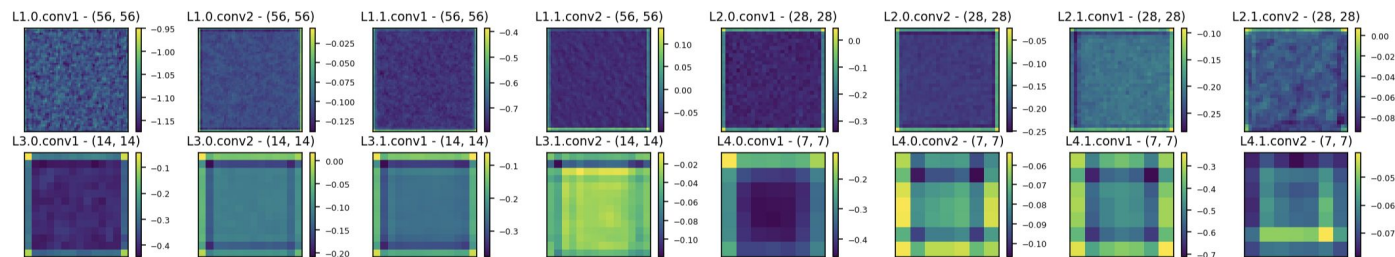
ResNet-18



ResNet-50

# Feature-Maps in ResNet-18

## 0-Padding



## Mirror (SYMMETRIC)

