

Making Convolutional Networks Shift-Invariant Again

Richard Zhang
Adobe Research



Example classifications



86.7



P(correct class)



69.2



P(correct class)

Deep Networks are not Shift-Invariant



46.3



P(correct class)



18.0



P(correct class)

Deep Networks are not Shift-Invariant



46.3



P(correct class)



18.0



P(correct class)

Azulay and Weiss. Why do deep convolutional networks generalize so poorly to small image transformations? In ArXiv, 2018.
Engstrom, Tsipras, Schmidt, Madry. Exploring the Landscape of Spatial Robustness. In ICML, 2019.

Why is shift-invariance lost?

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“Convolutions are **shift-equivariant**”

Why is shift-invariance lost?

“Convolutions are shift-equivariant”

“Pooling builds up shift-invariance”

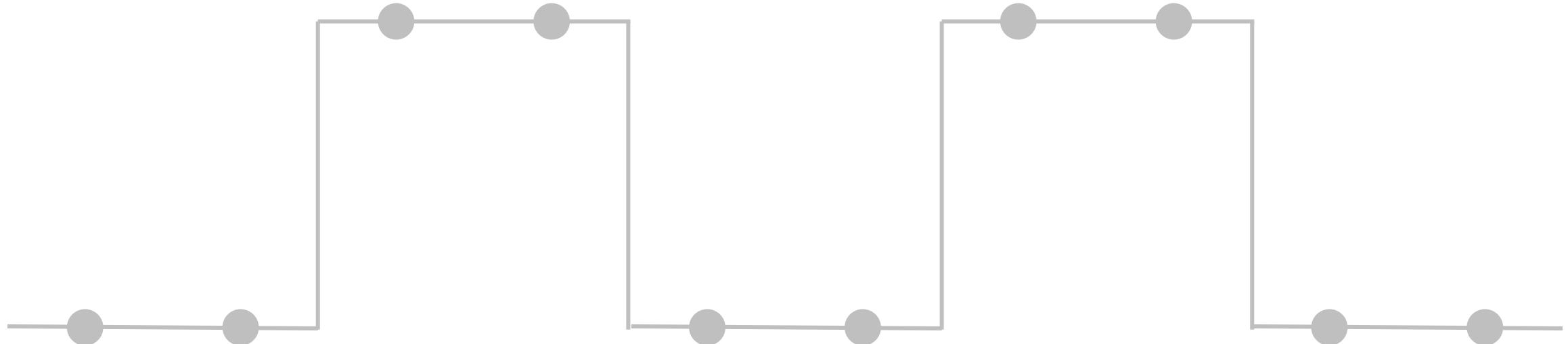
Why is shift-invariance lost?

“Convolutions are **shift-equivariant**”

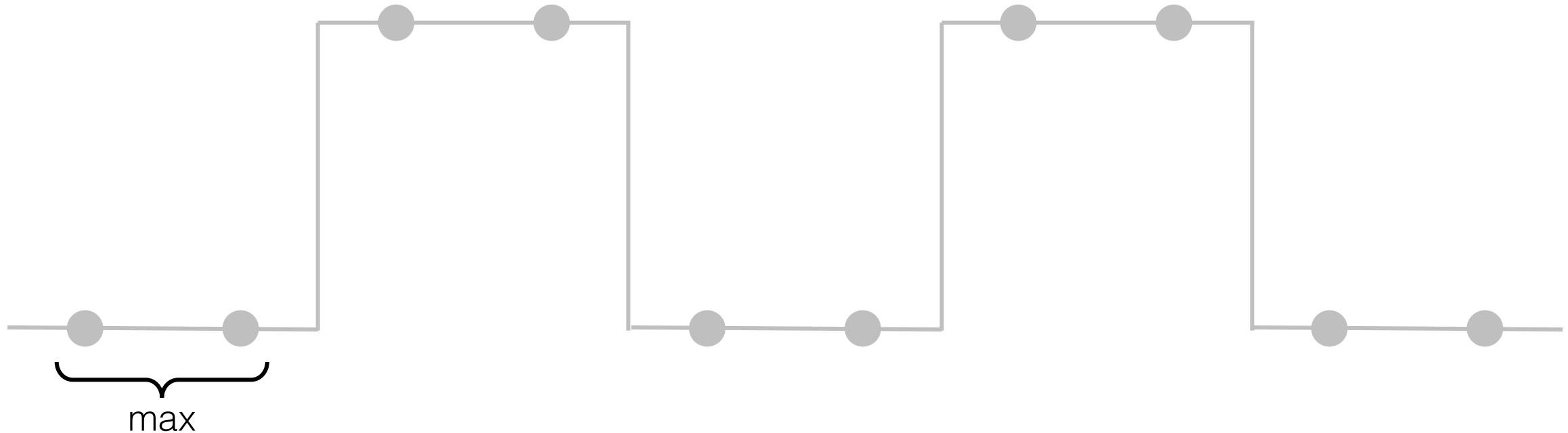
“Pooling builds up shift-invariance”

...but striding ignores Nyquist sampling theorem
and **aliases**

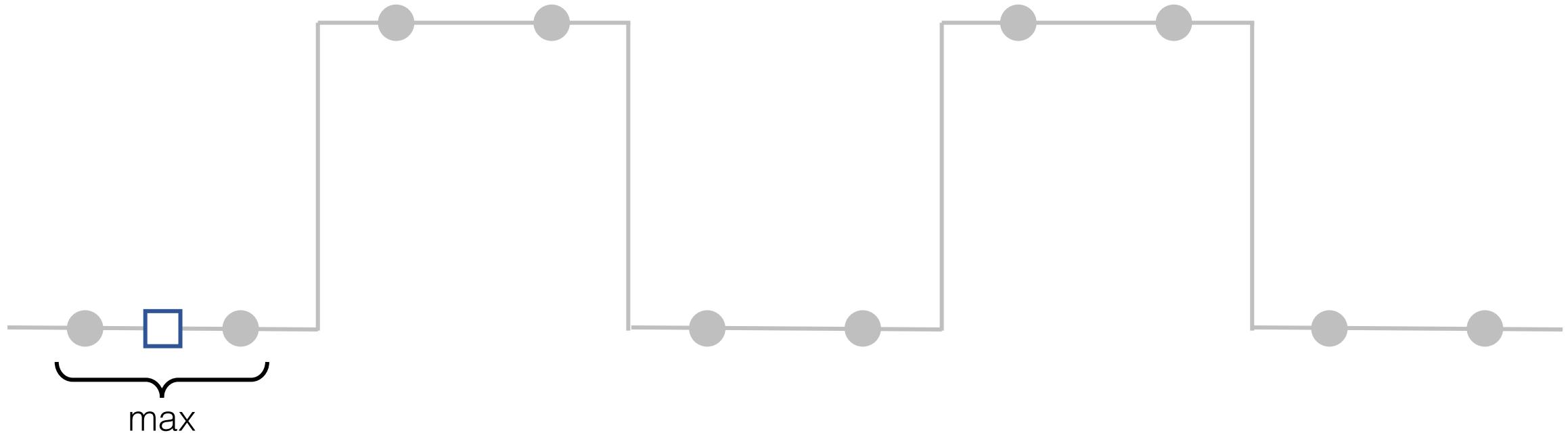
Re-examining Max-Pooling



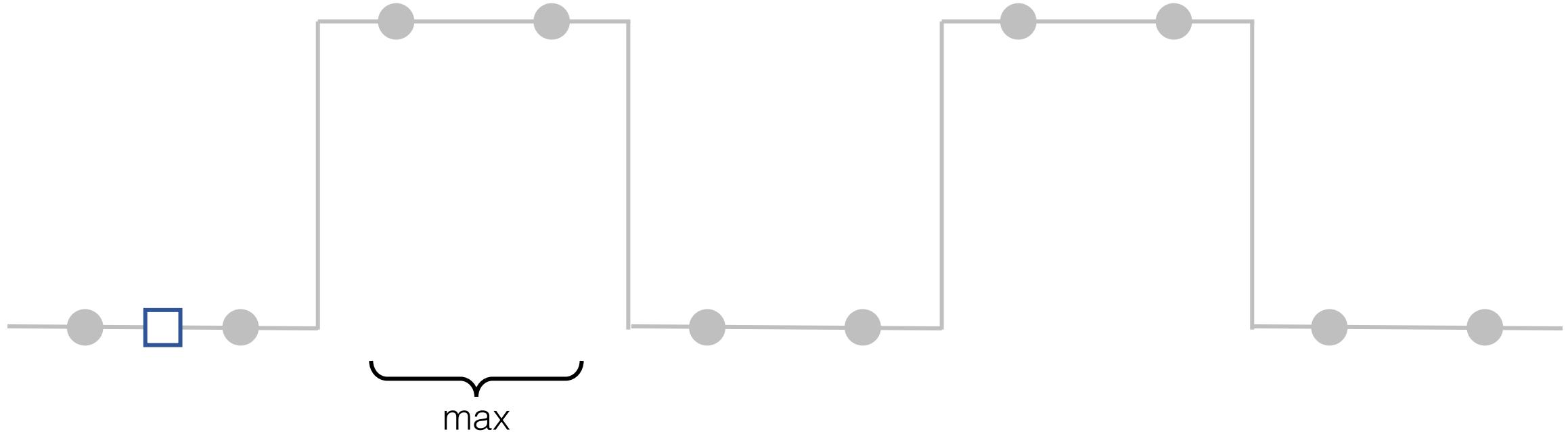
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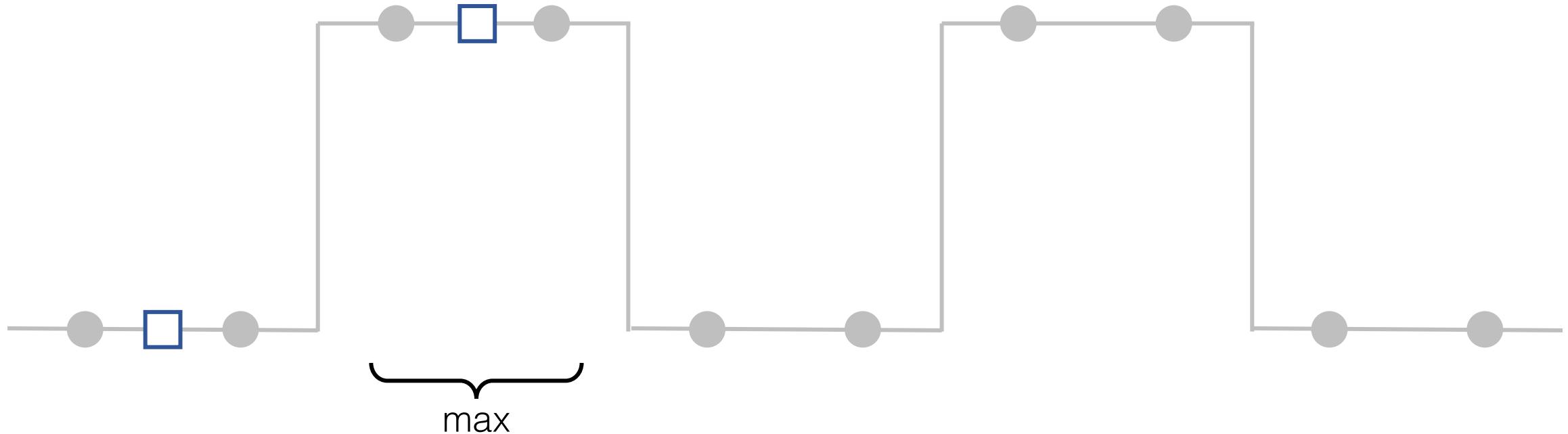
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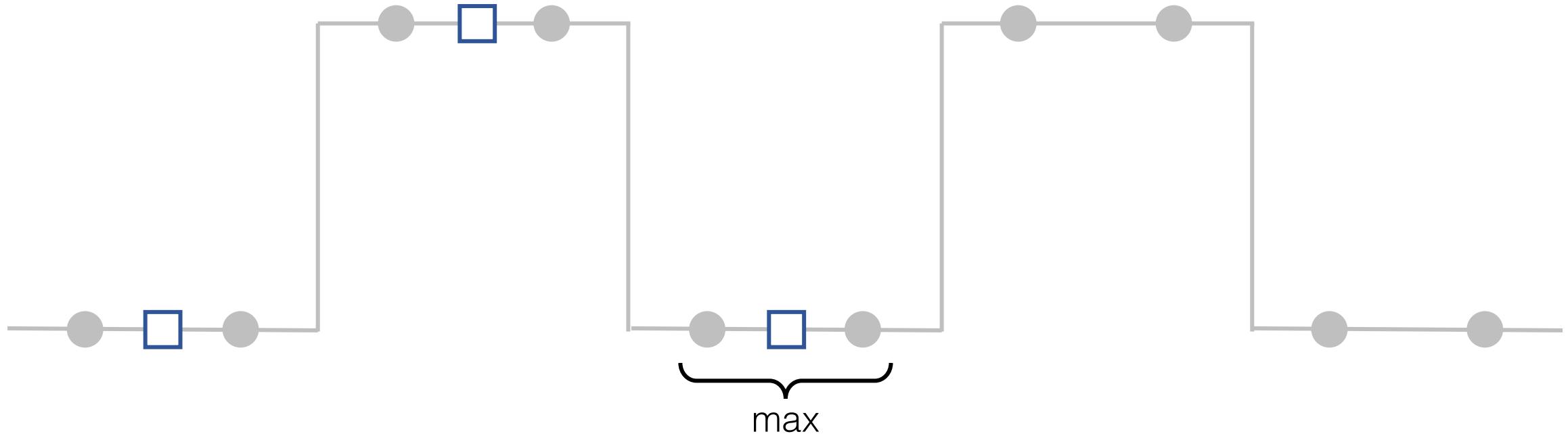
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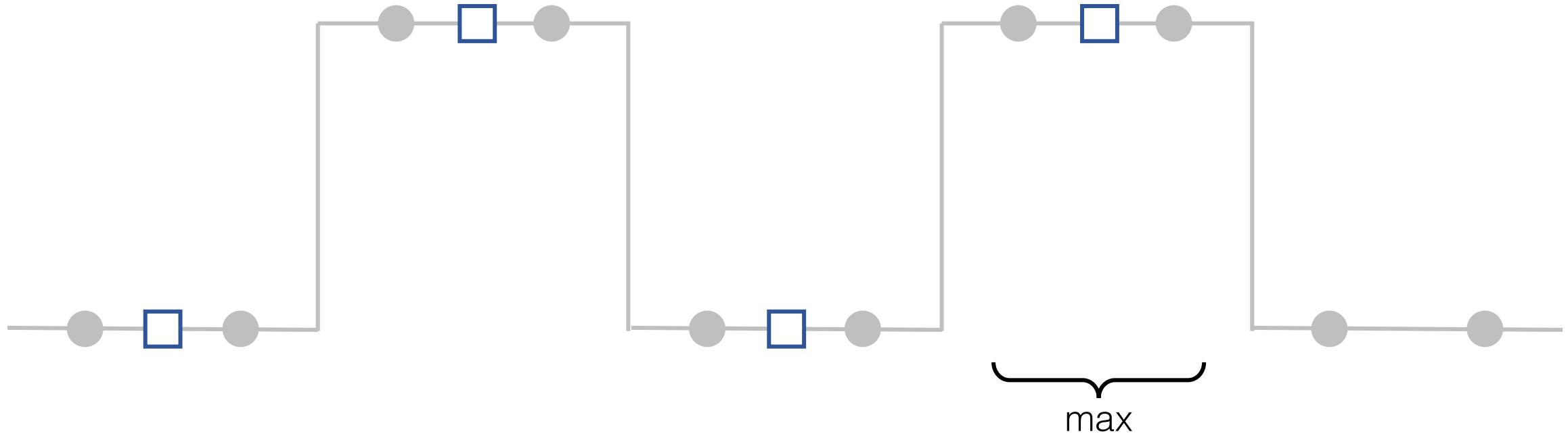
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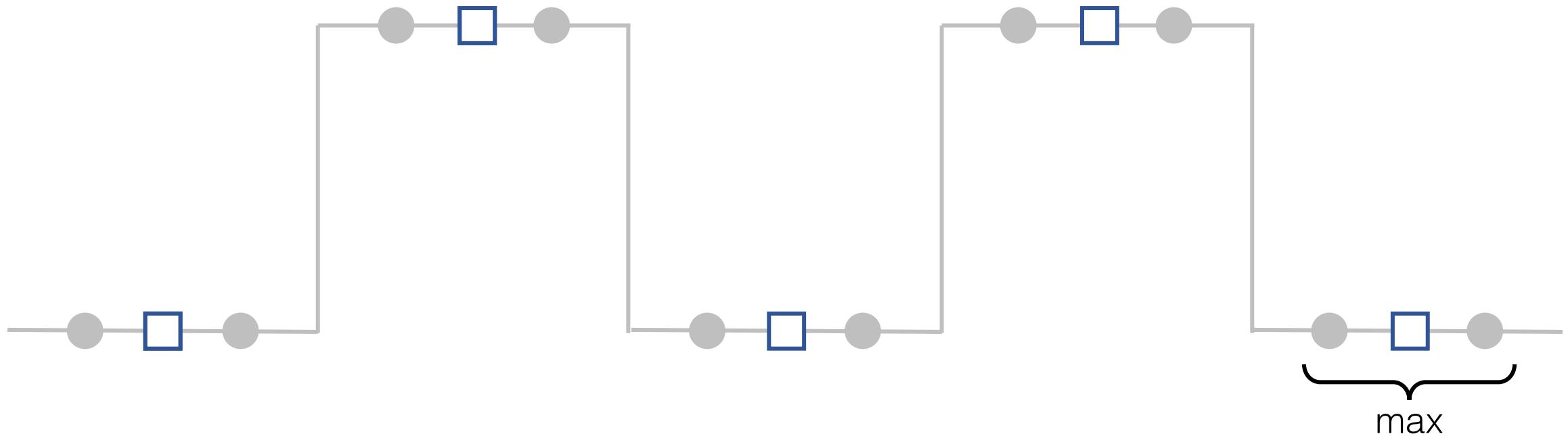
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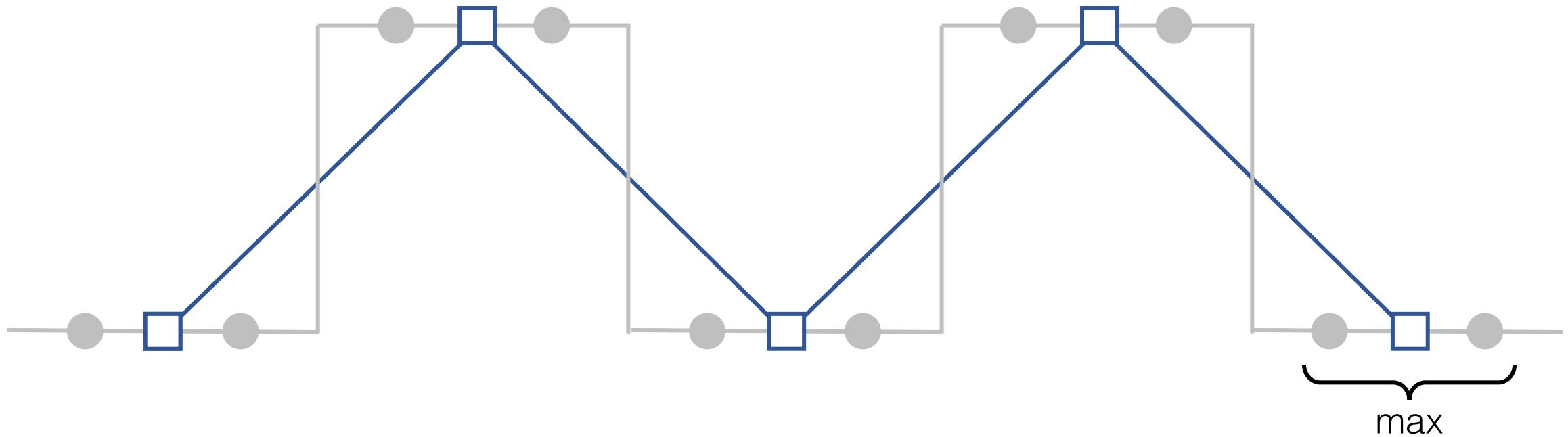
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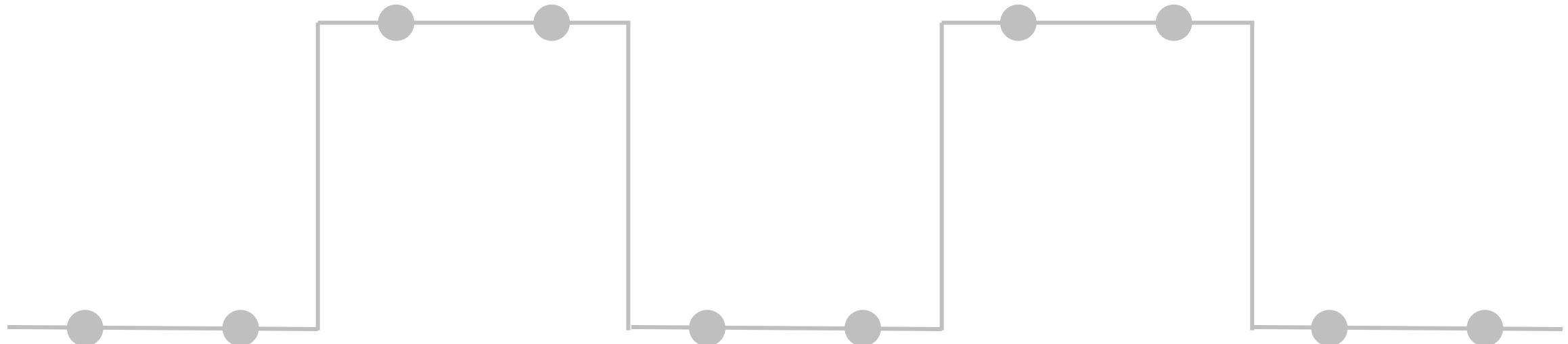
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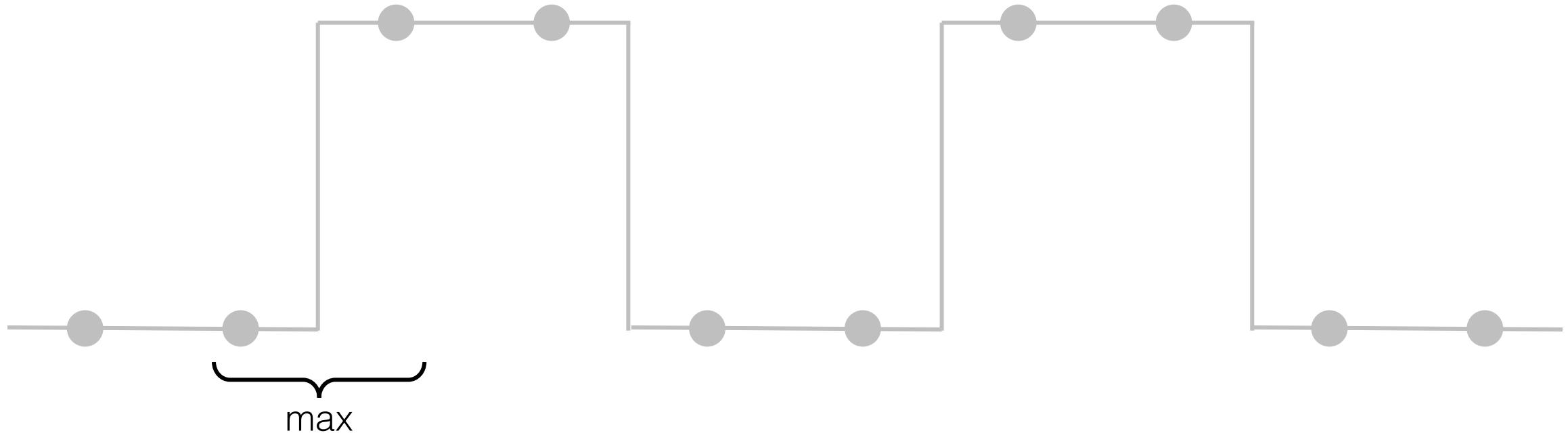
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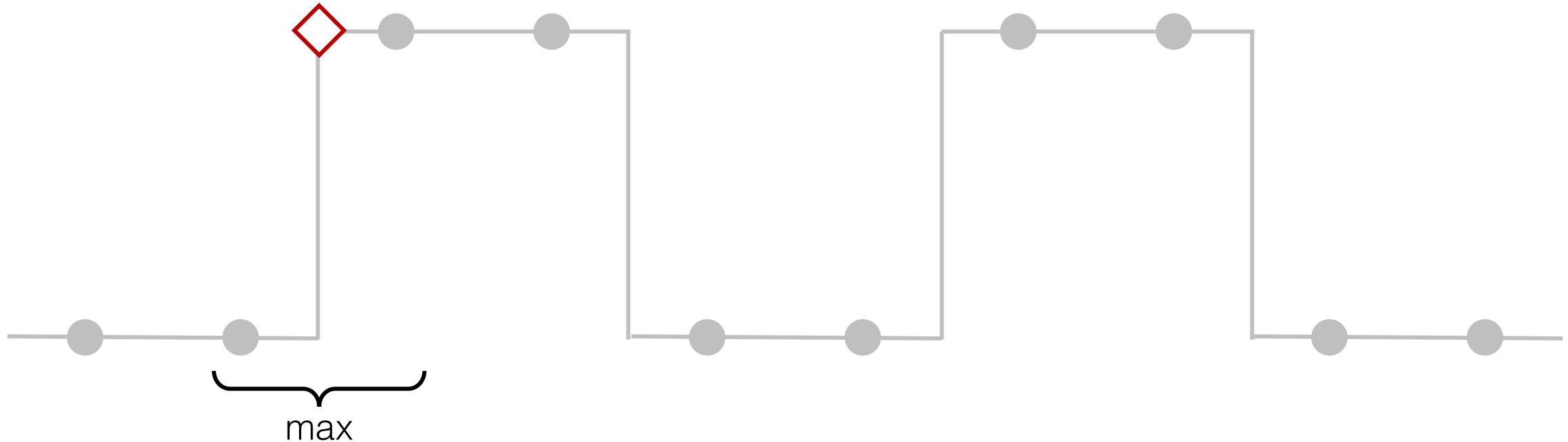
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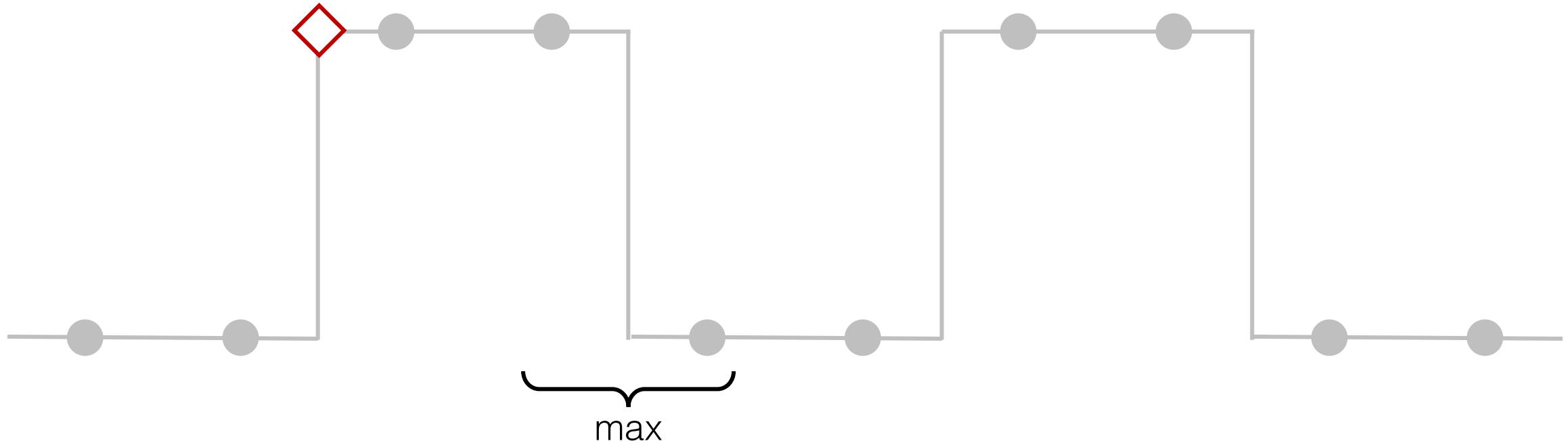
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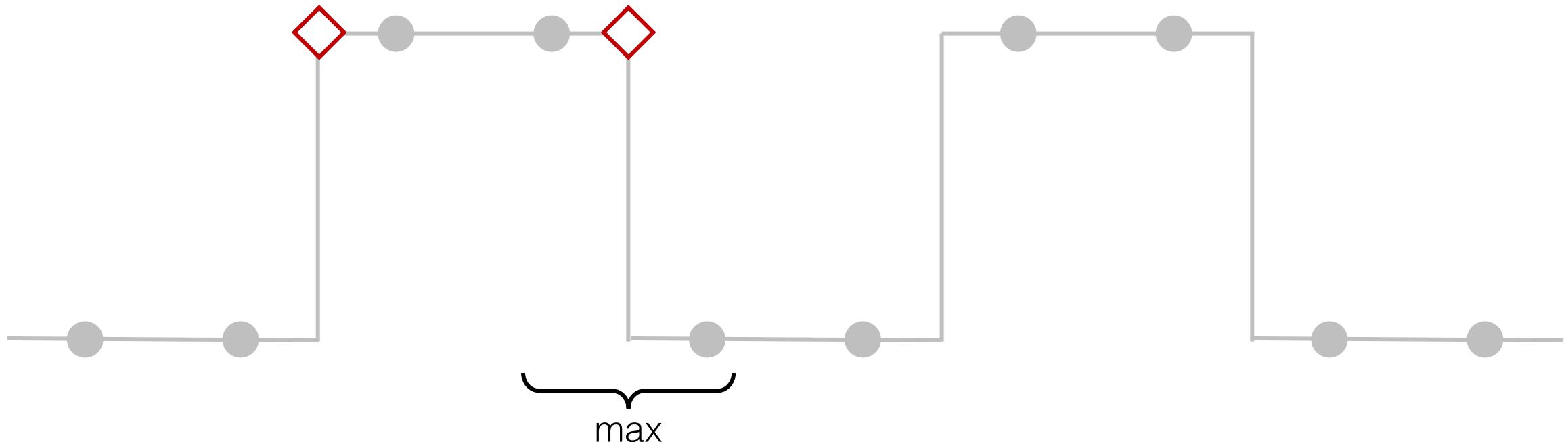
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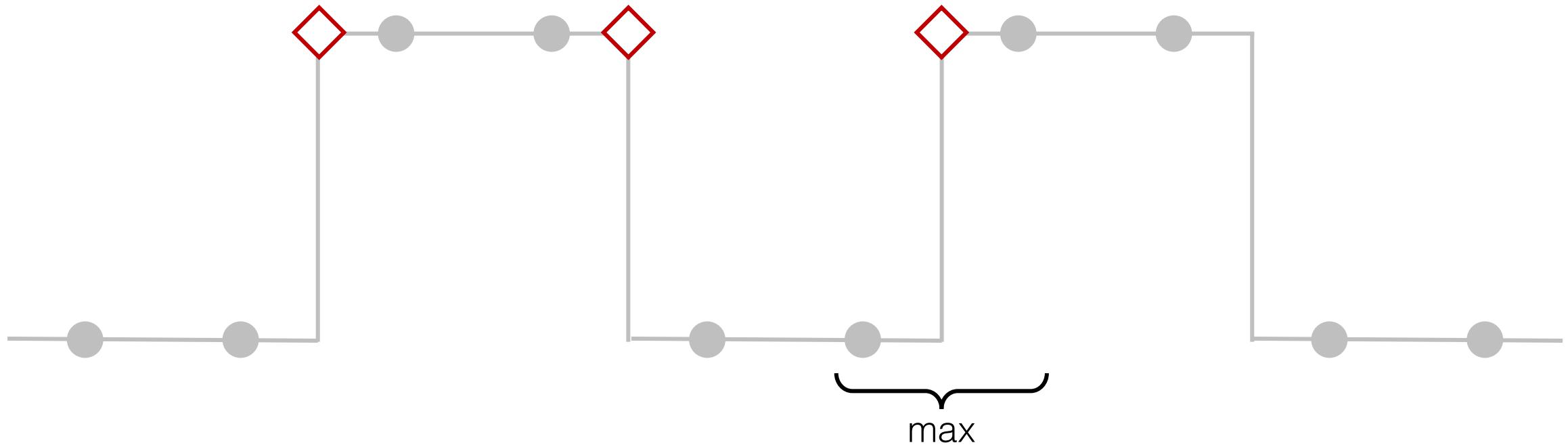
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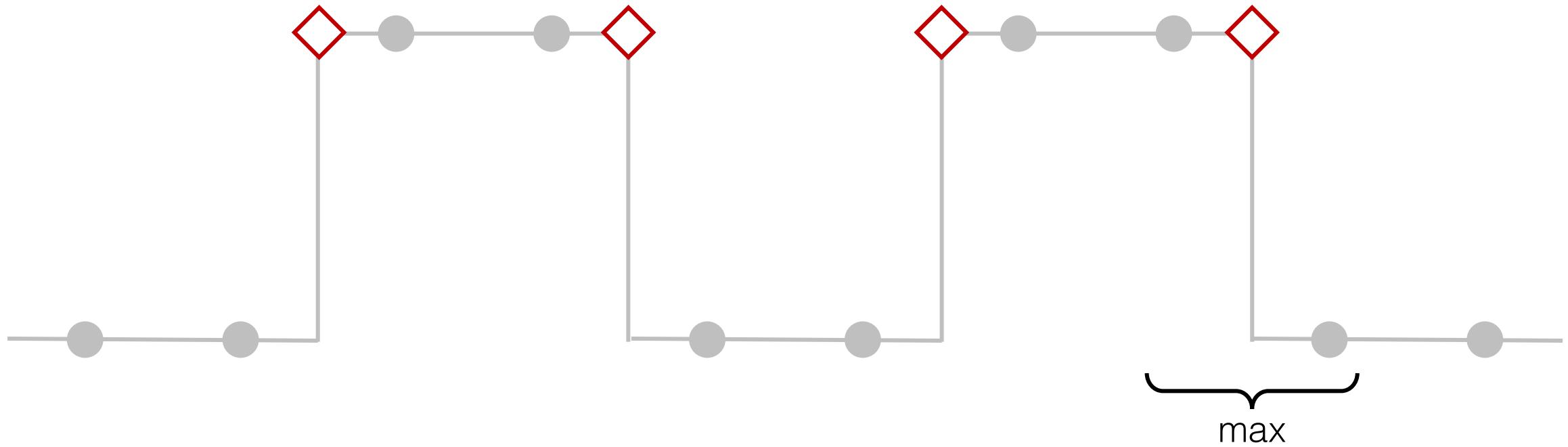
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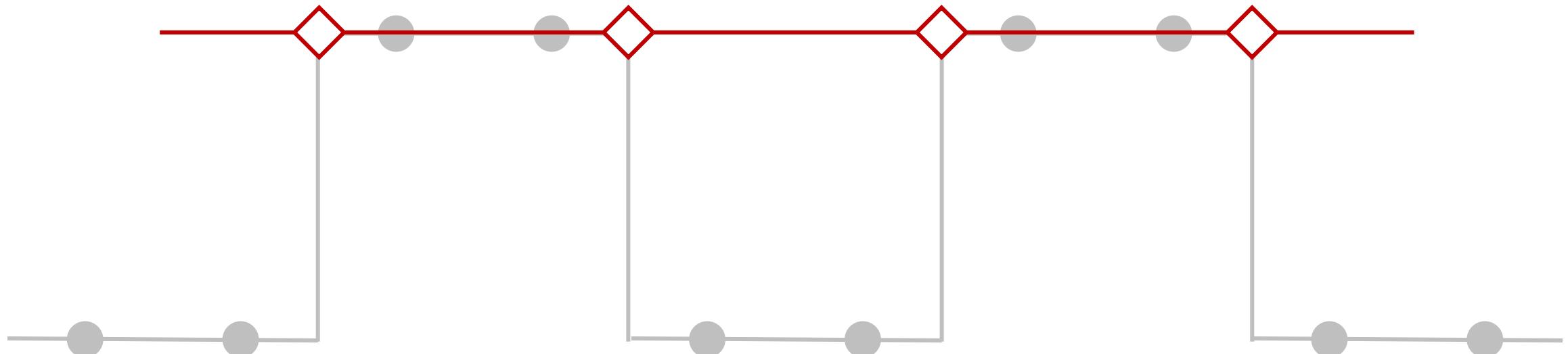
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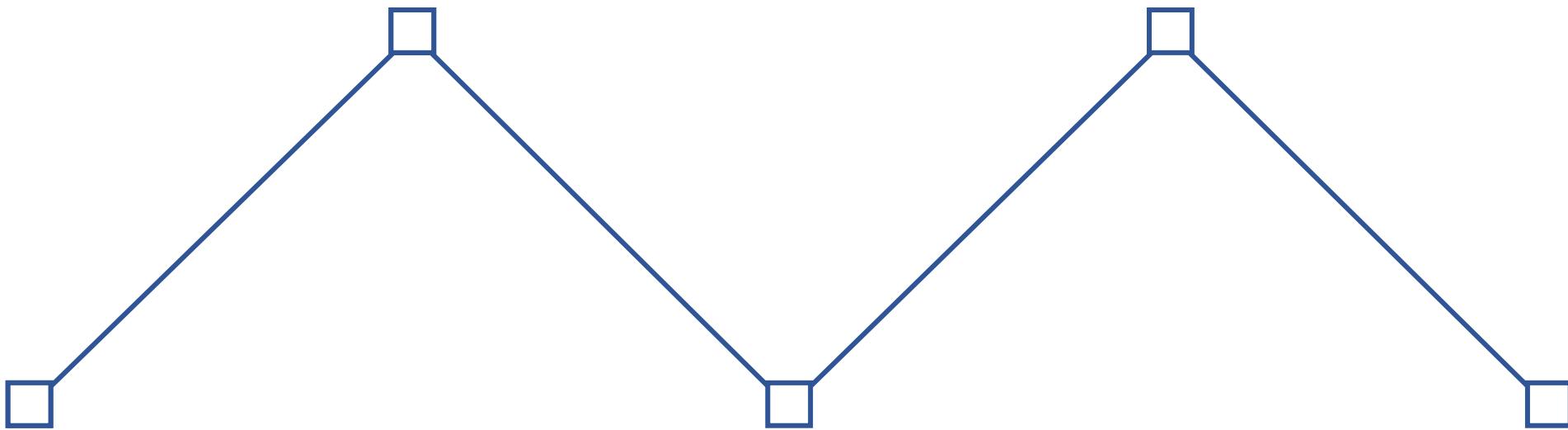
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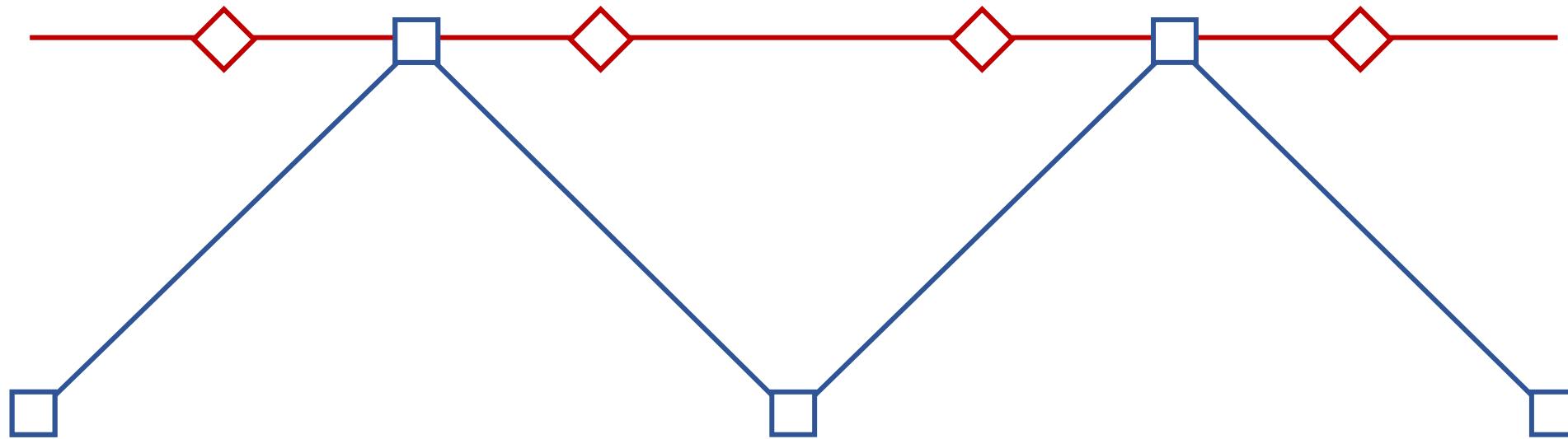
Re-examining Max-Pooling



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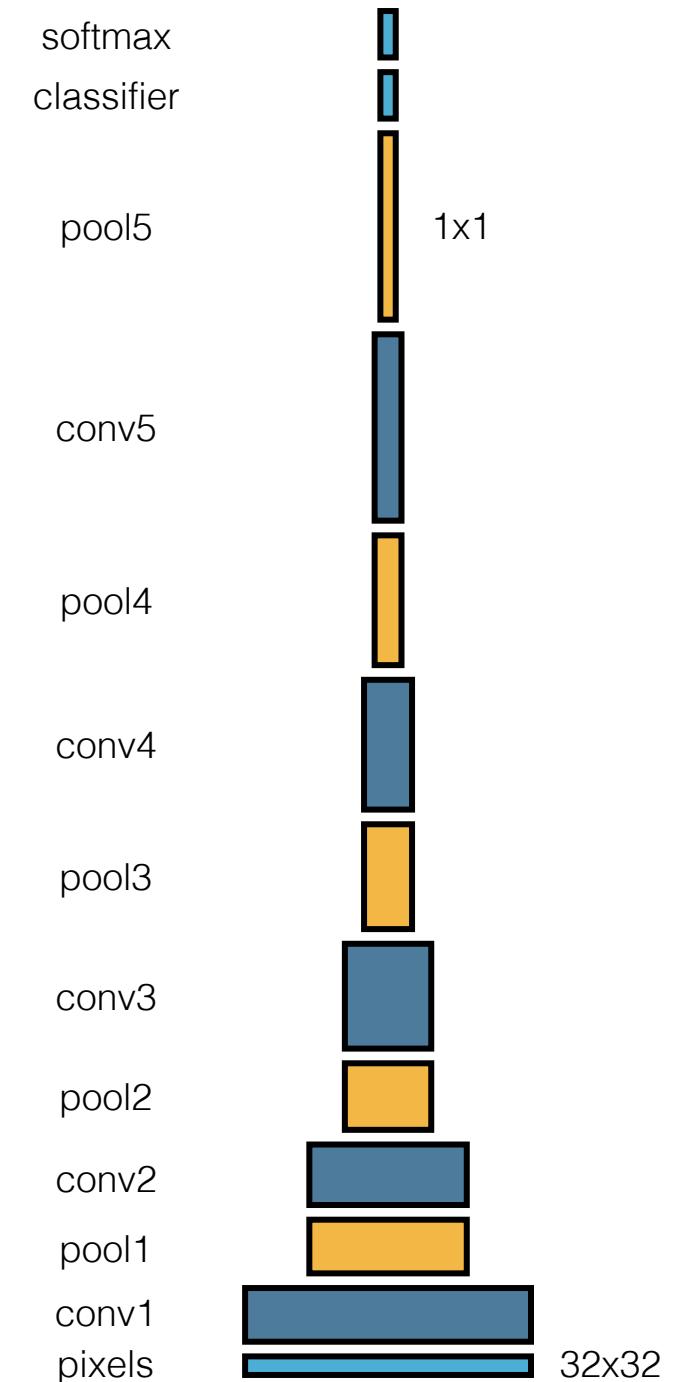
Re-examining Max-Pooling



Max-pooling breaks shift-equivariance

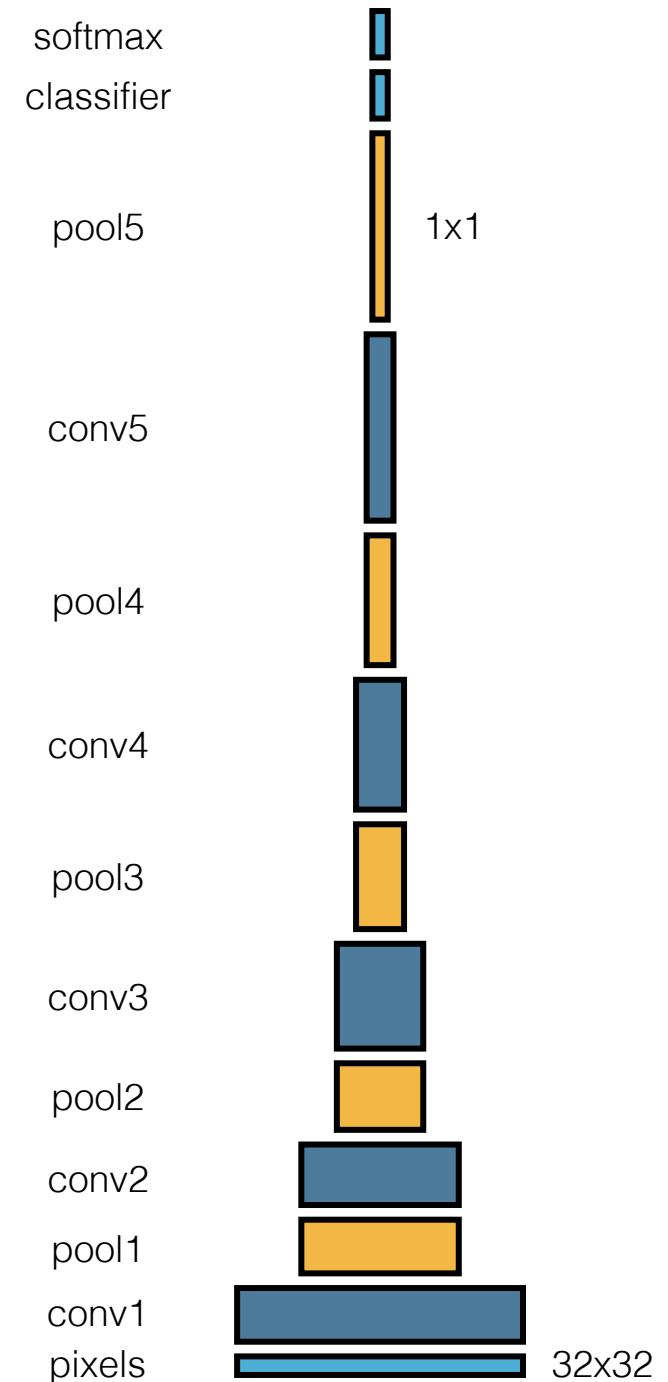
Shift-equivariance in VGG

- CIFAR
- VGG network
 - 5 max-pools

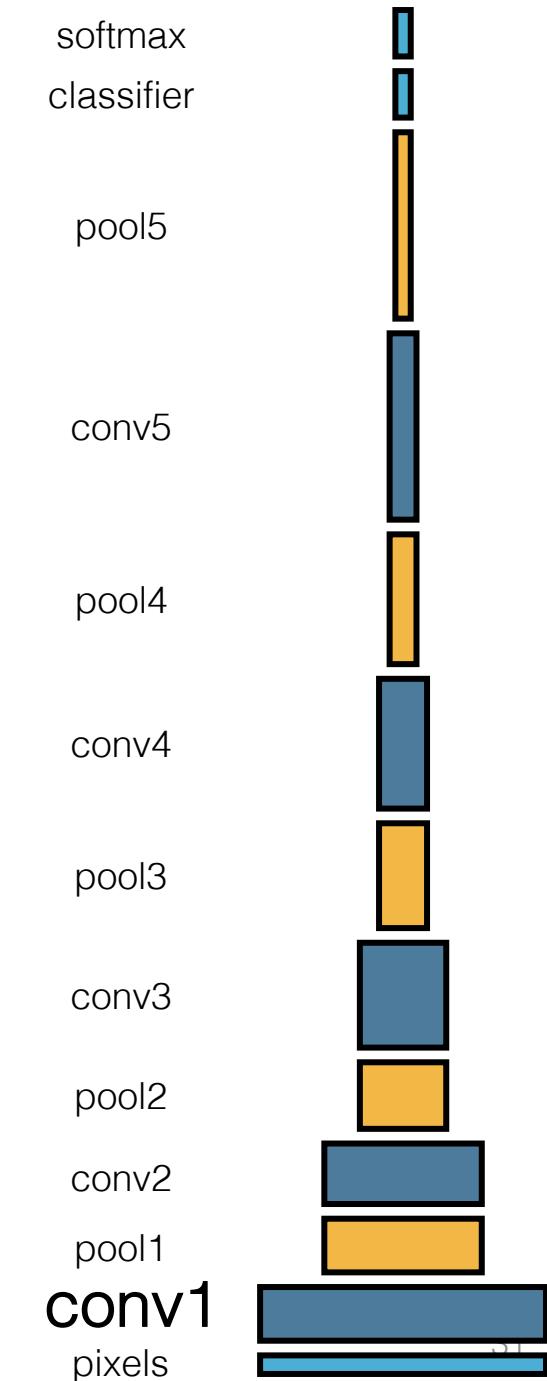
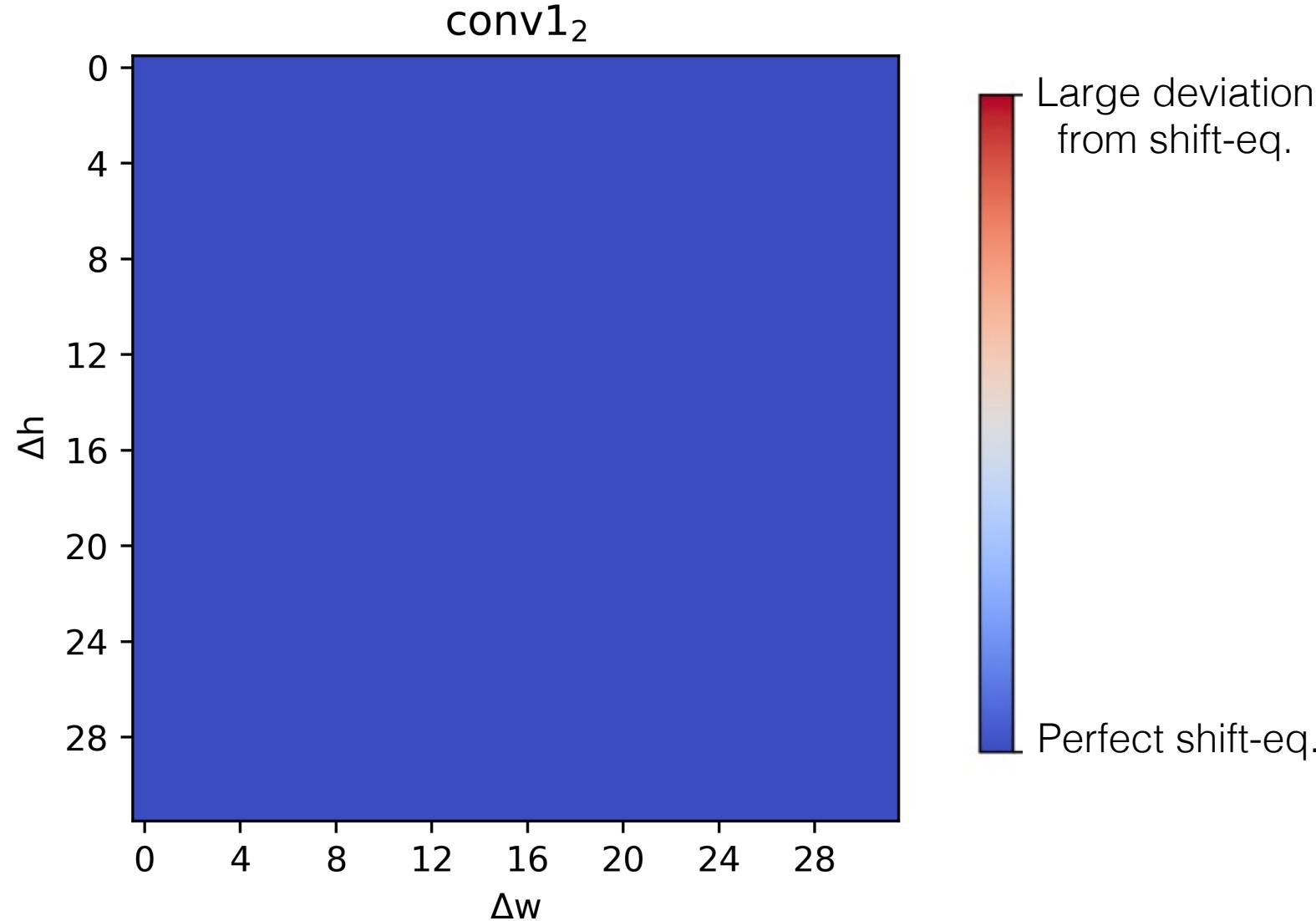


Shift-equivariance in VGG

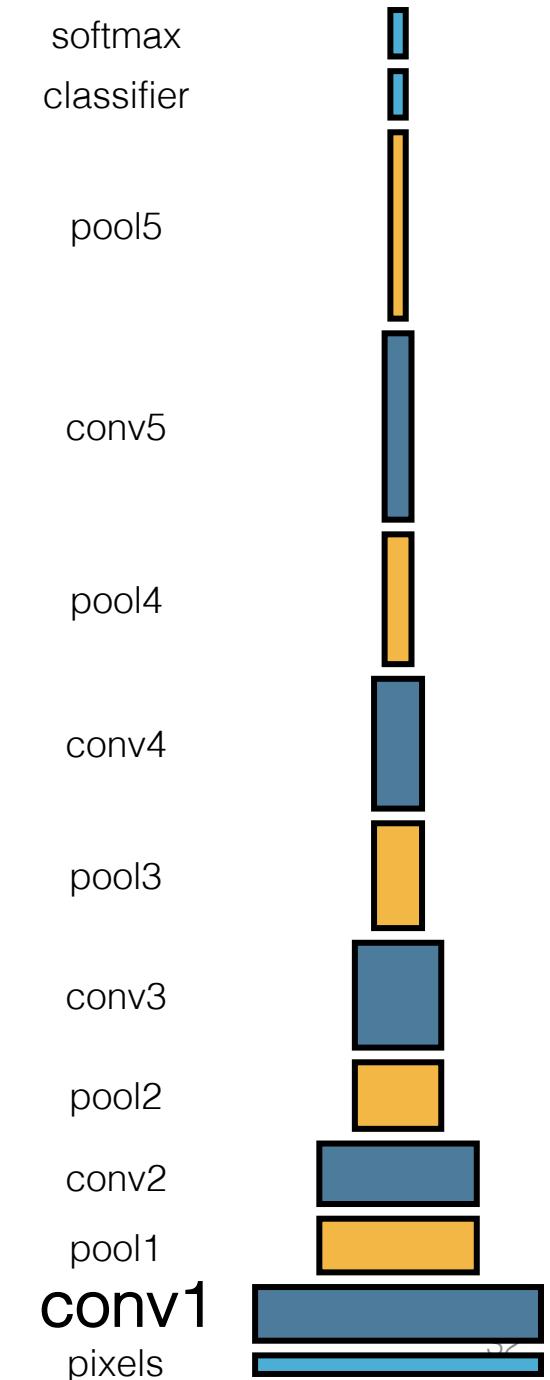
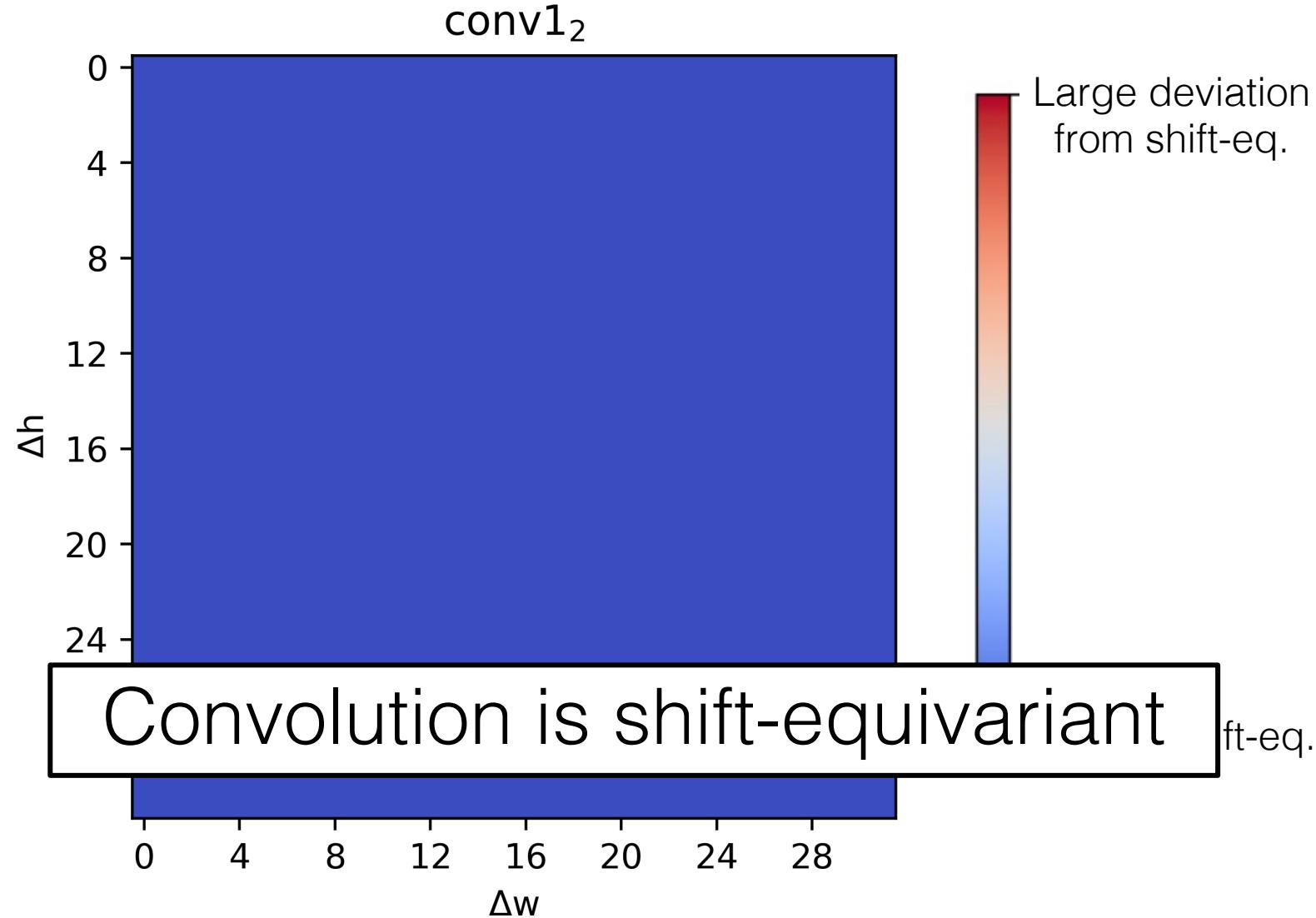
- CIFAR
- VGG network
 - 5 max-pools
- Test shift-equivariance condition
 - $\text{dist}(F(\text{Shift}_{\Delta h, \Delta w}(X)), \text{Shift}_{\Delta h, \Delta w}(F(X)))$



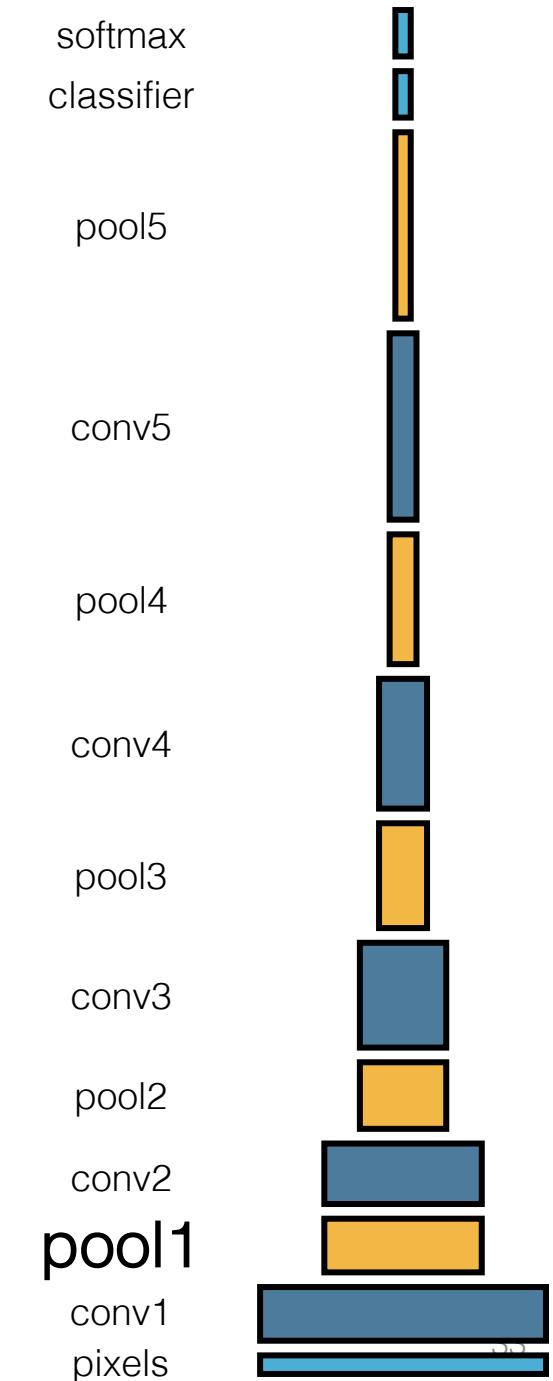
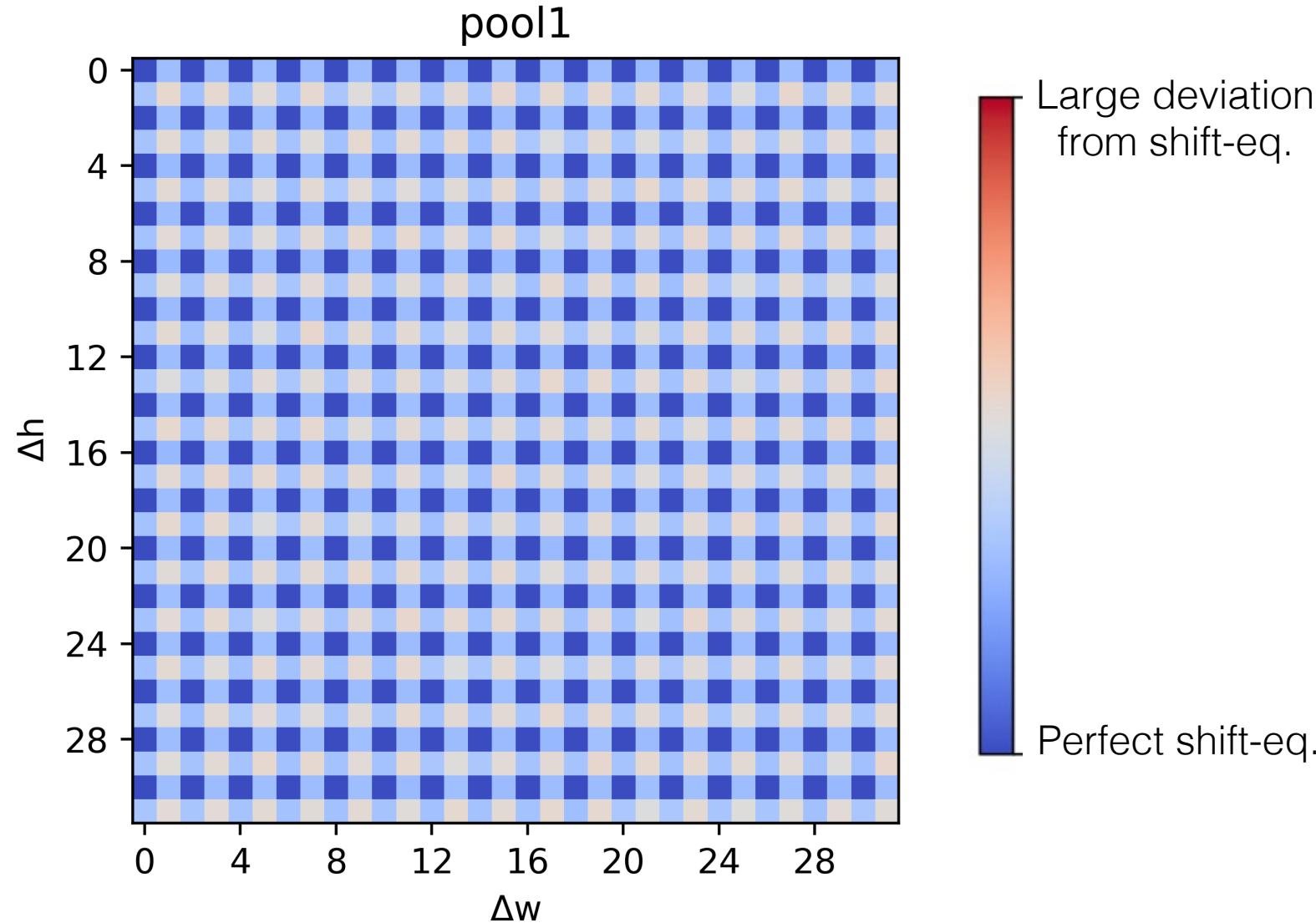
Shift-equivariance, per layer



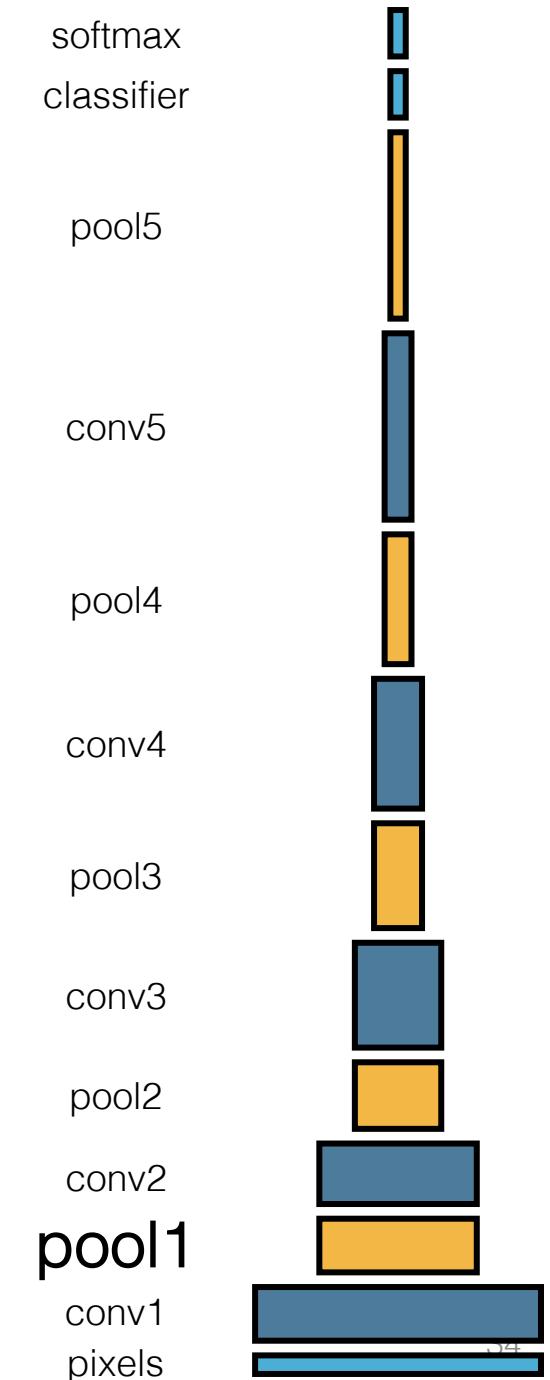
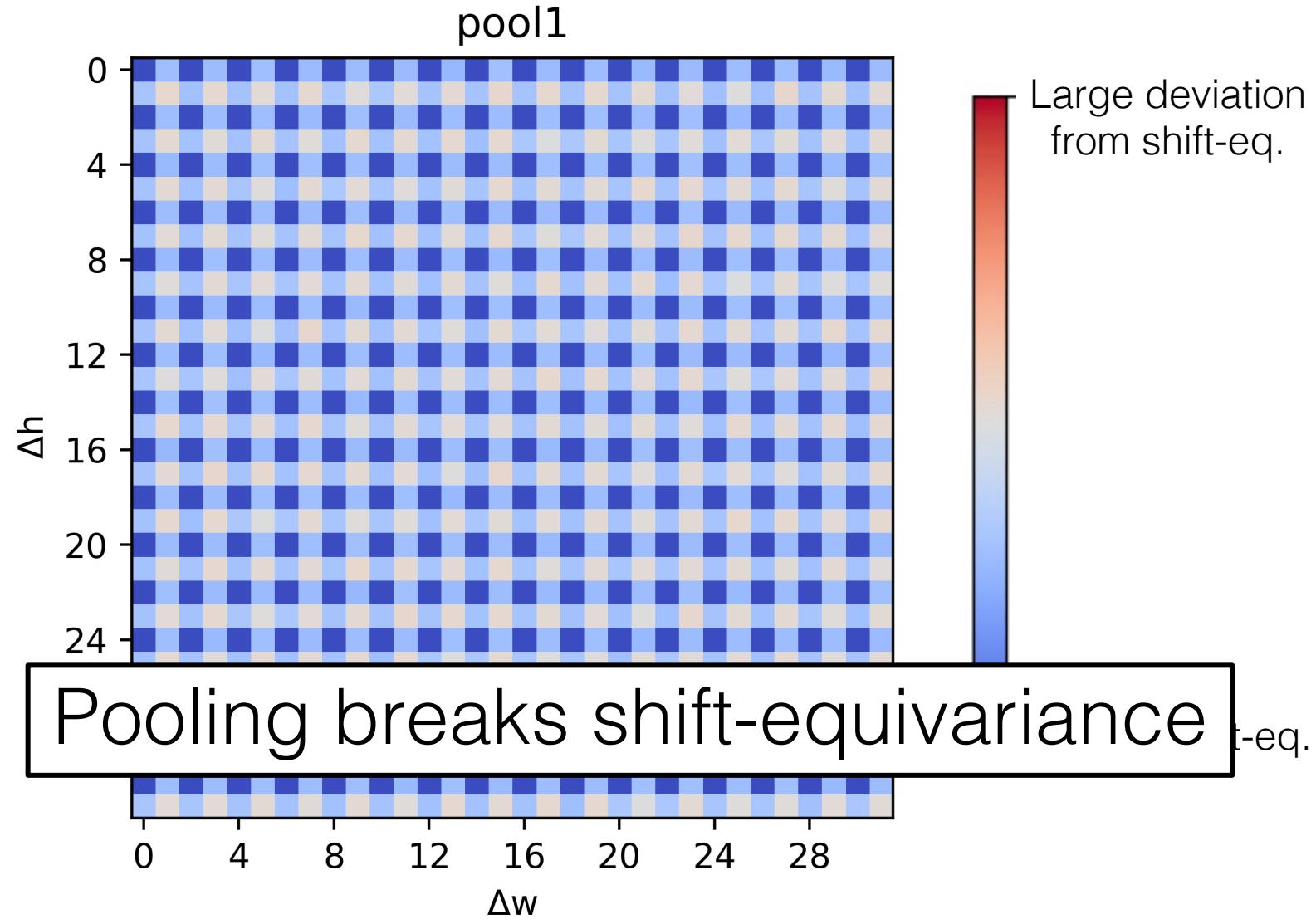
Shift-equivariance, per layer



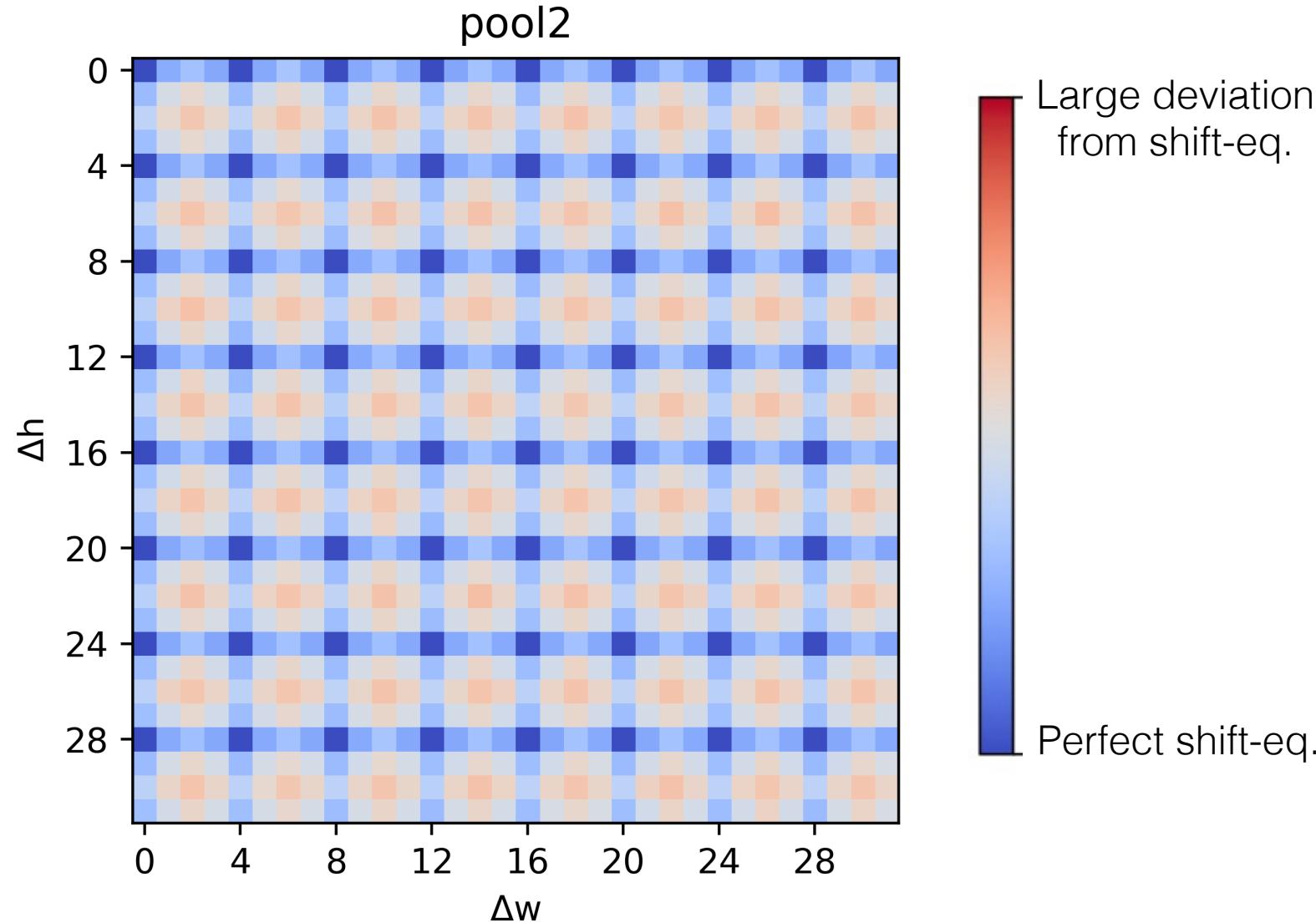
Shift-equivariance, per layer



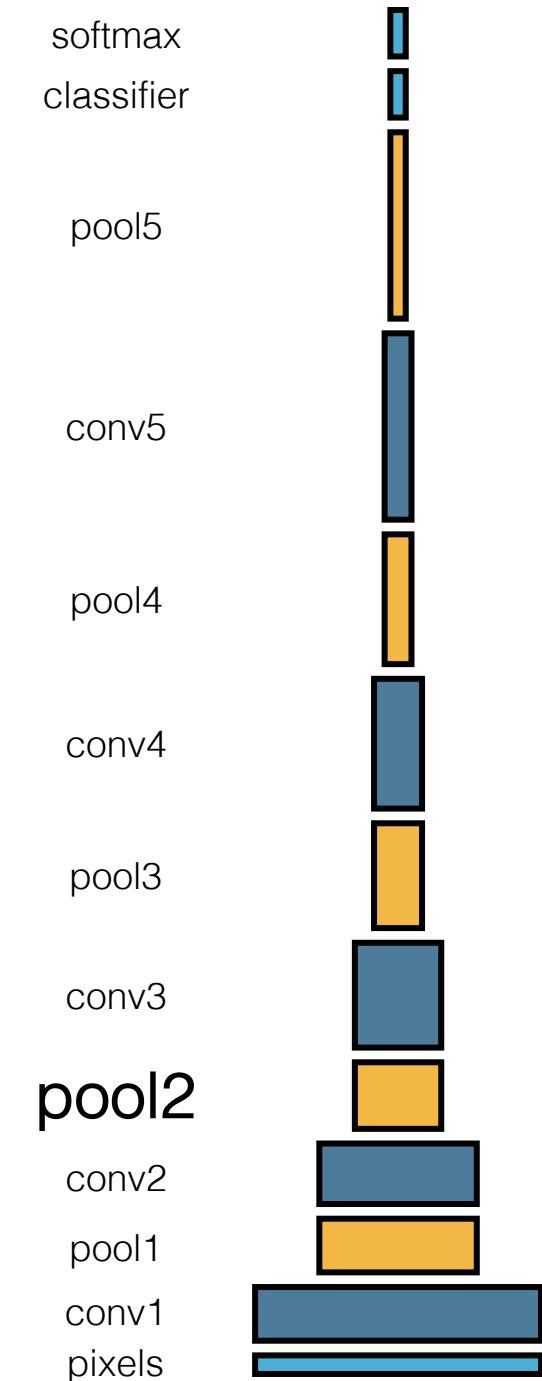
Shift-equivariance, per layer



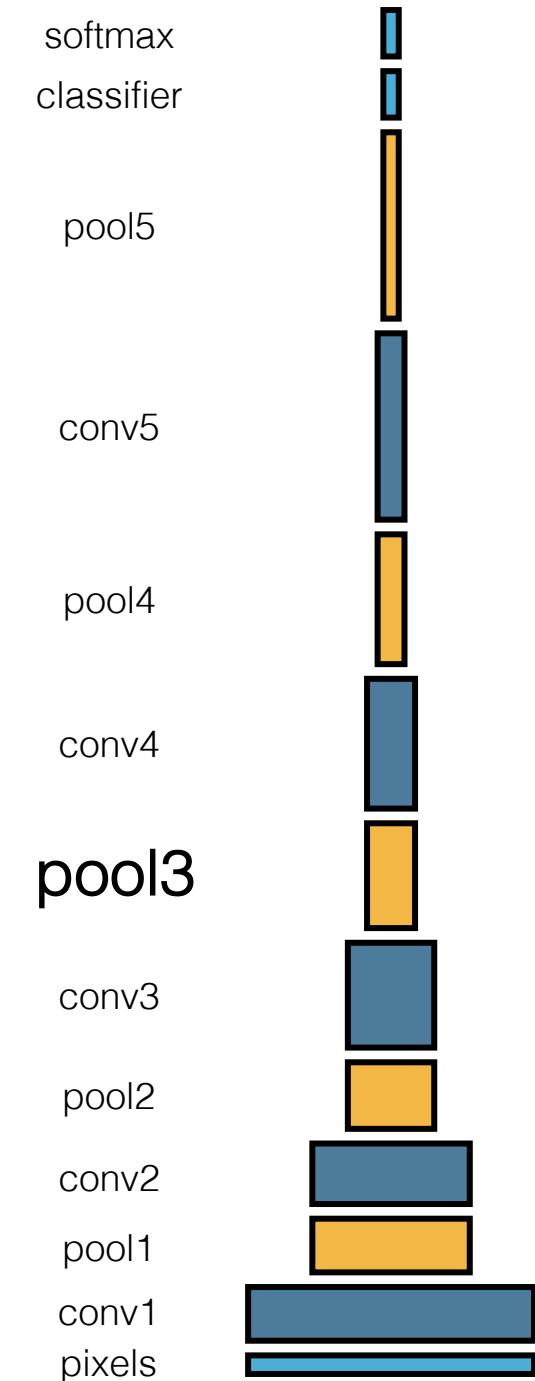
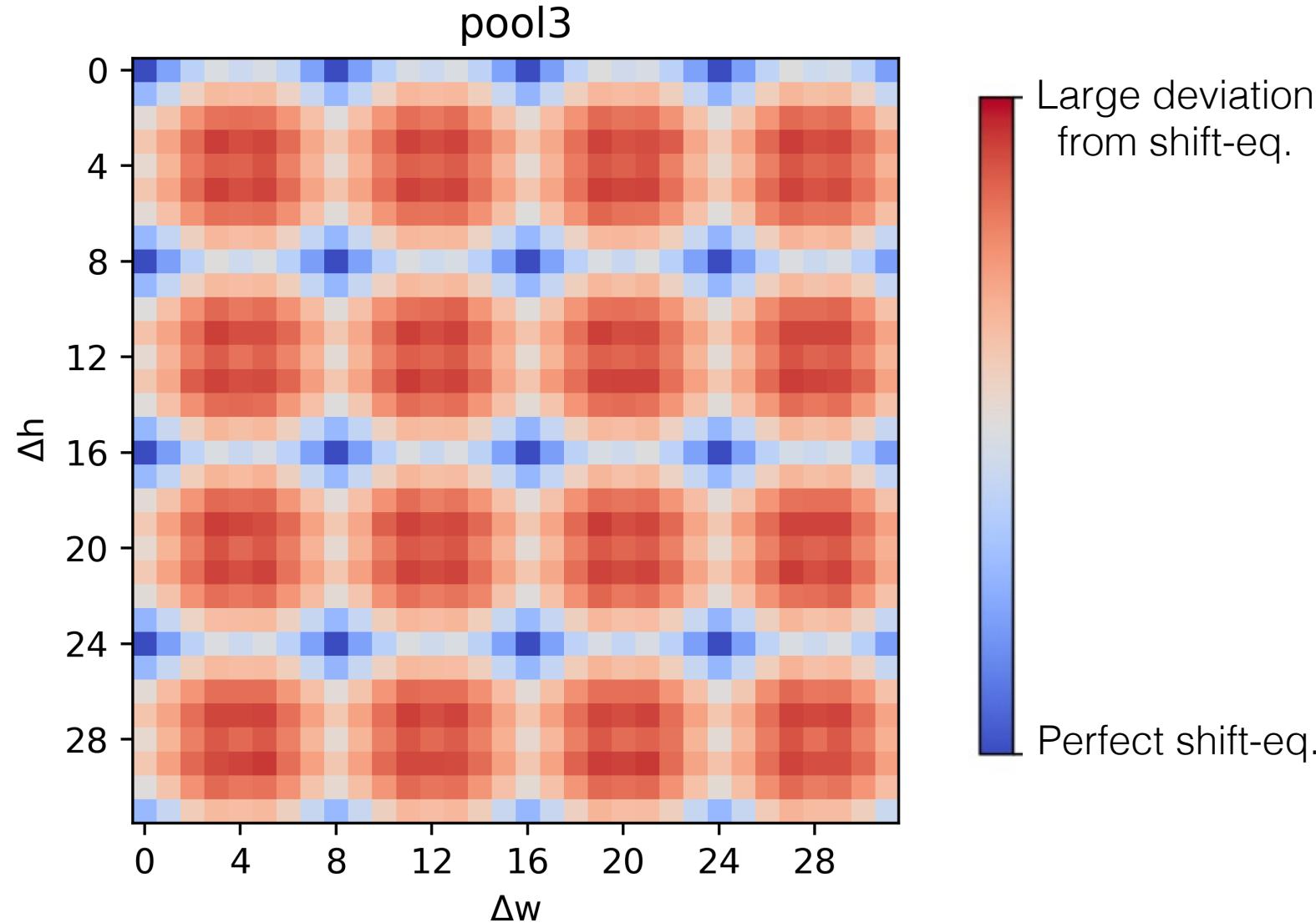
Shift-equivariance, per layer



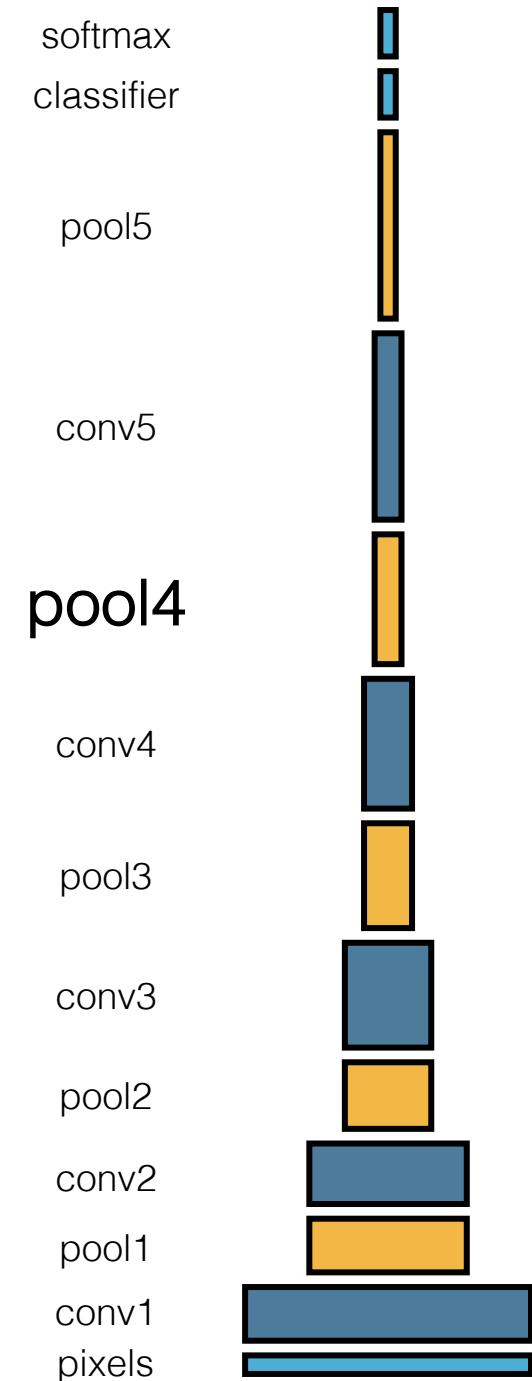
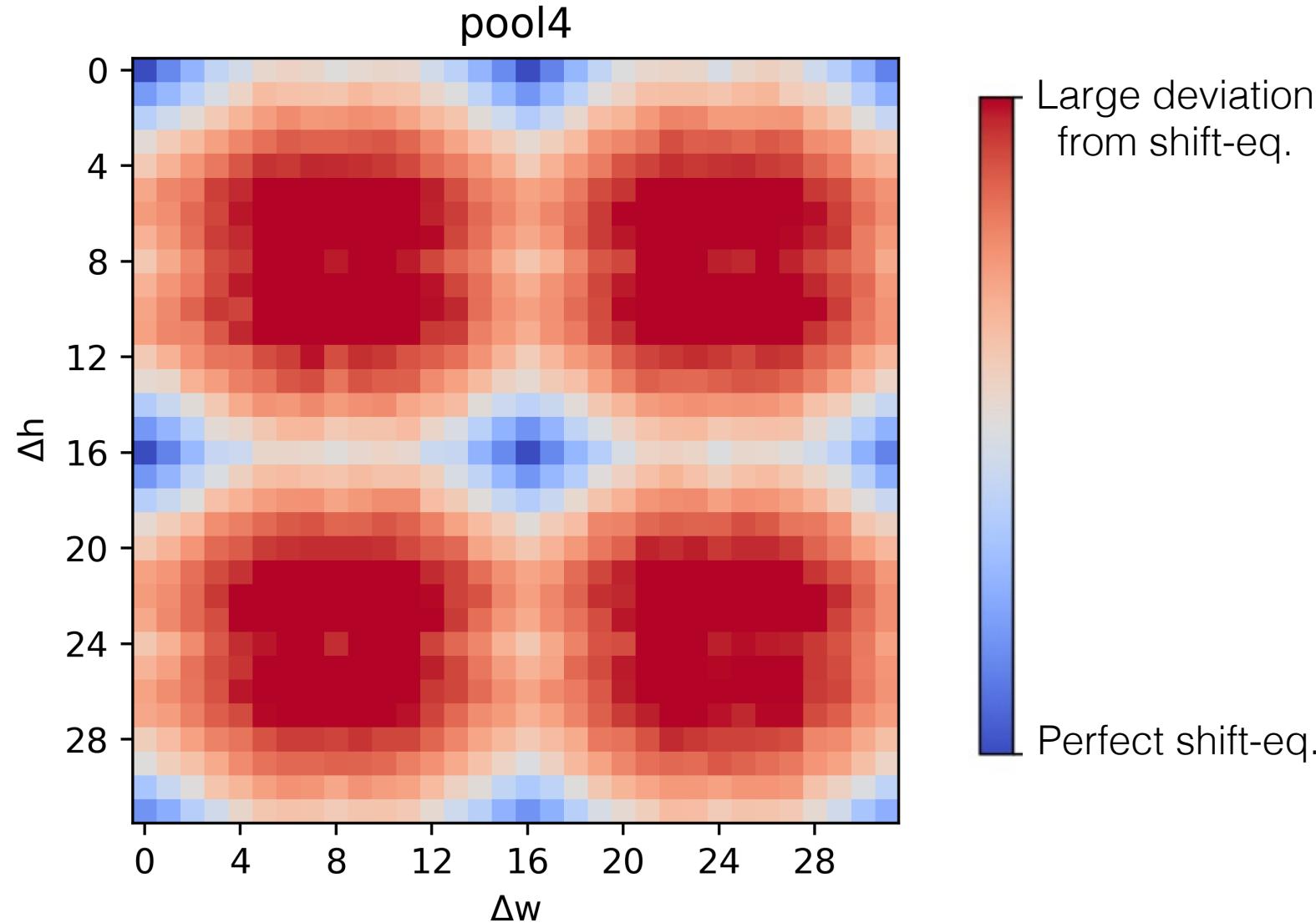
Large deviation
from shift-eq.
Perfect shift-eq.



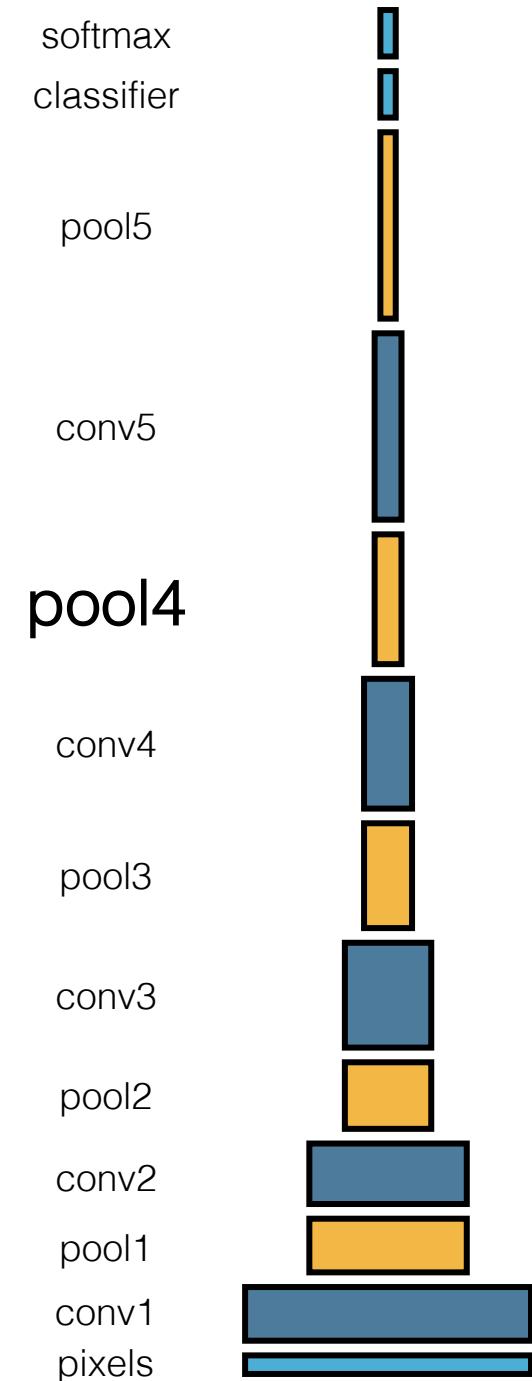
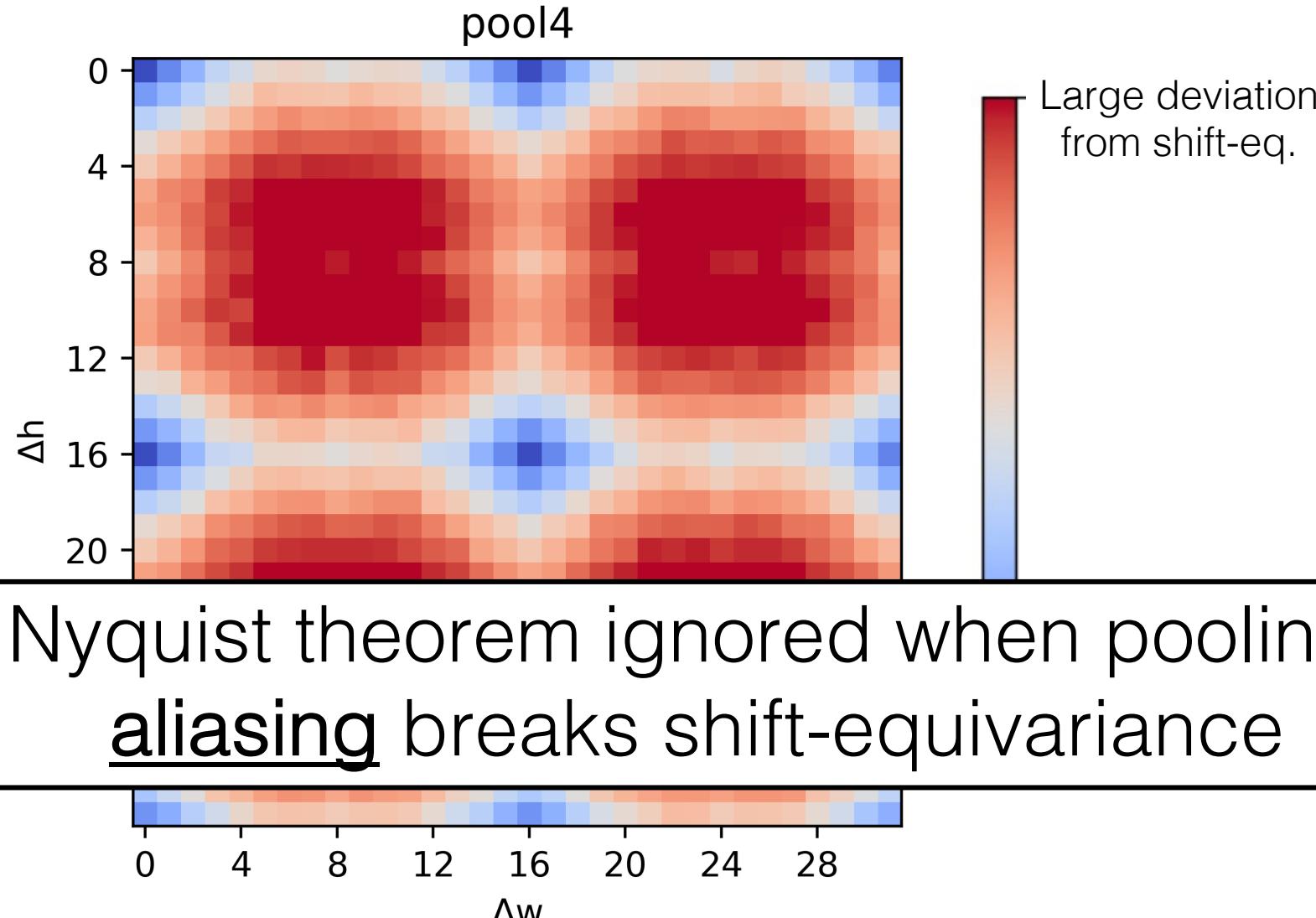
Shift-equivariance, per layer



Shift-equivariance, per layer



Shift-equivariance, per layer



Alternative downsampling methods

- Blur+subsample
 - Antialiasing in signal processing; image processing; graphics

Alternative downsampling methods

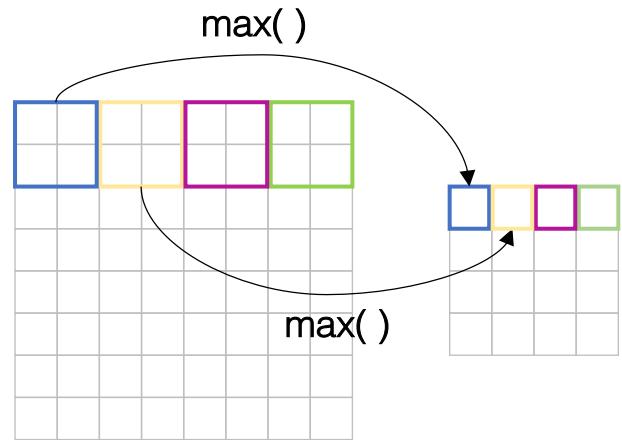
- Blur+subsample
 - Antialiasing in signal processing; image processing; graphics
- Max-pooling
 - Performs better in deep learning applications [Scherer 2010]

Alternative downsampling methods

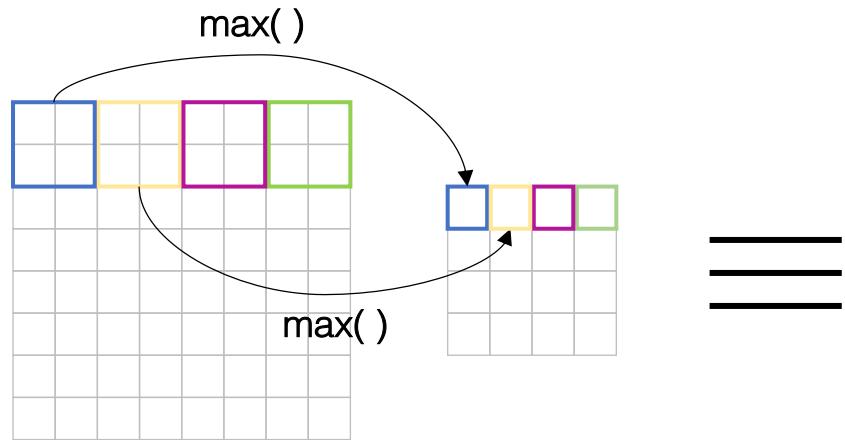
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Reconcile antialiasing with max-pooling

Baseline
(MaxPool)

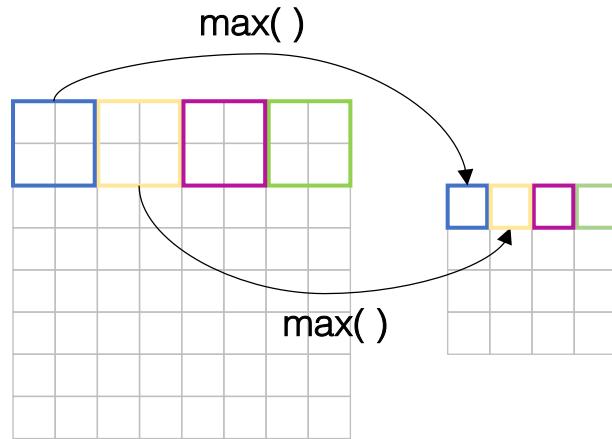


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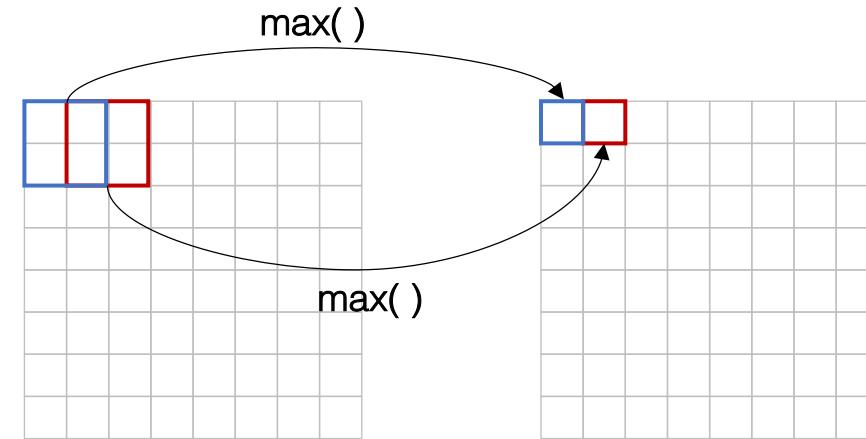
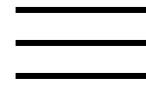


heavy aliasing

Baseline
(MaxPool)

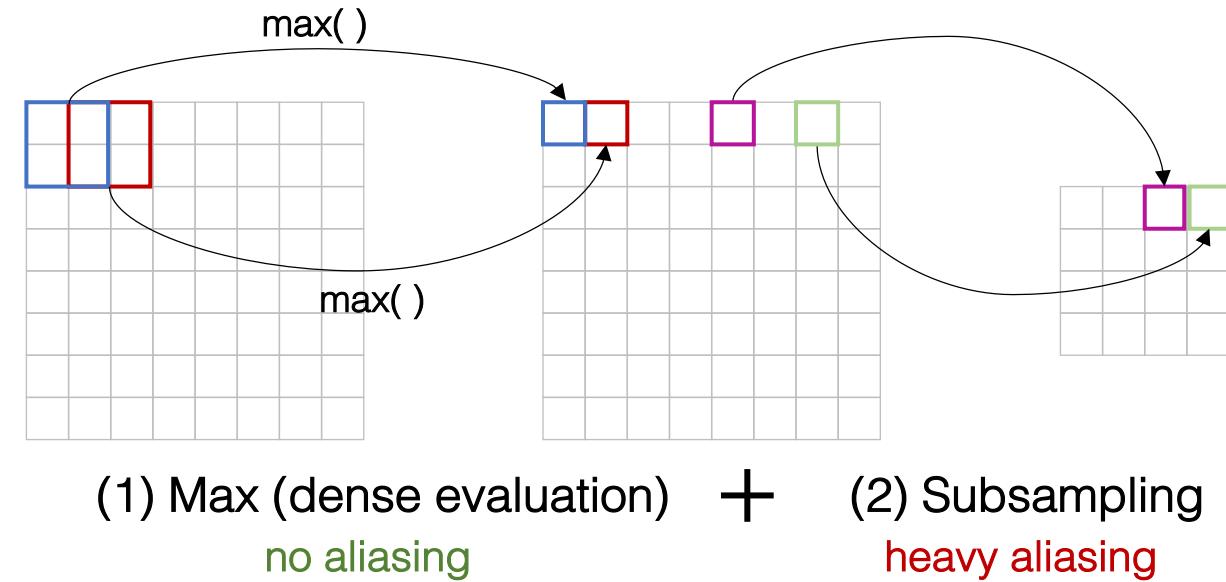
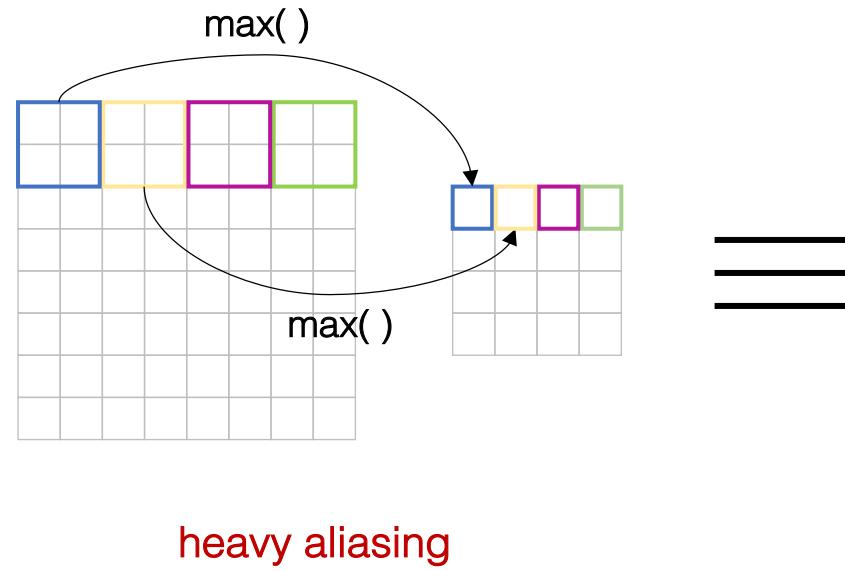


heavy aliasing

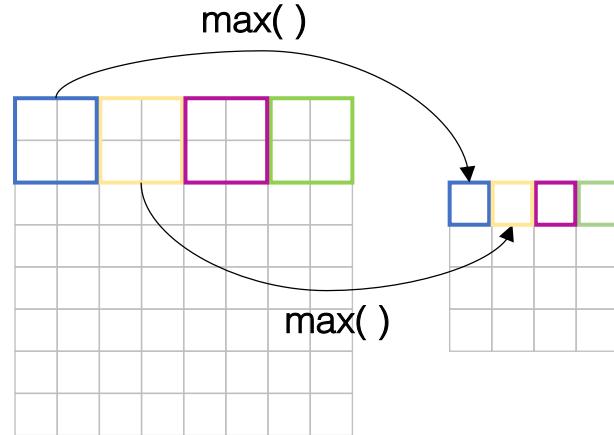


(1) Max (dense evaluation)
no aliasing

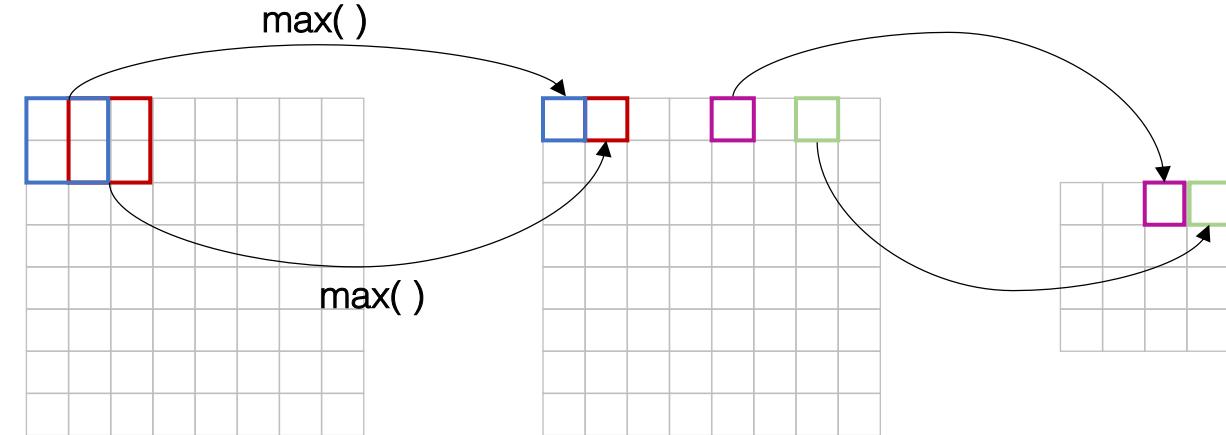
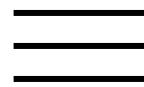
Baseline (MaxPool)



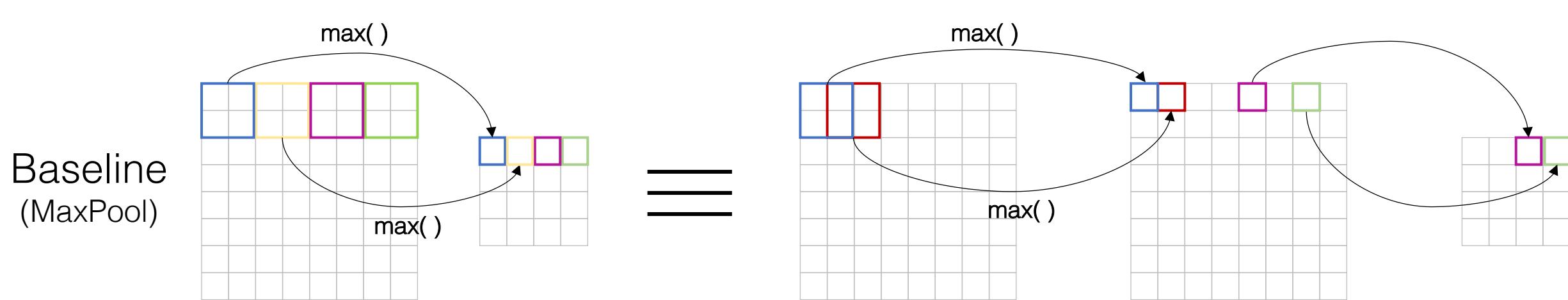
Baseline
(MaxPool)



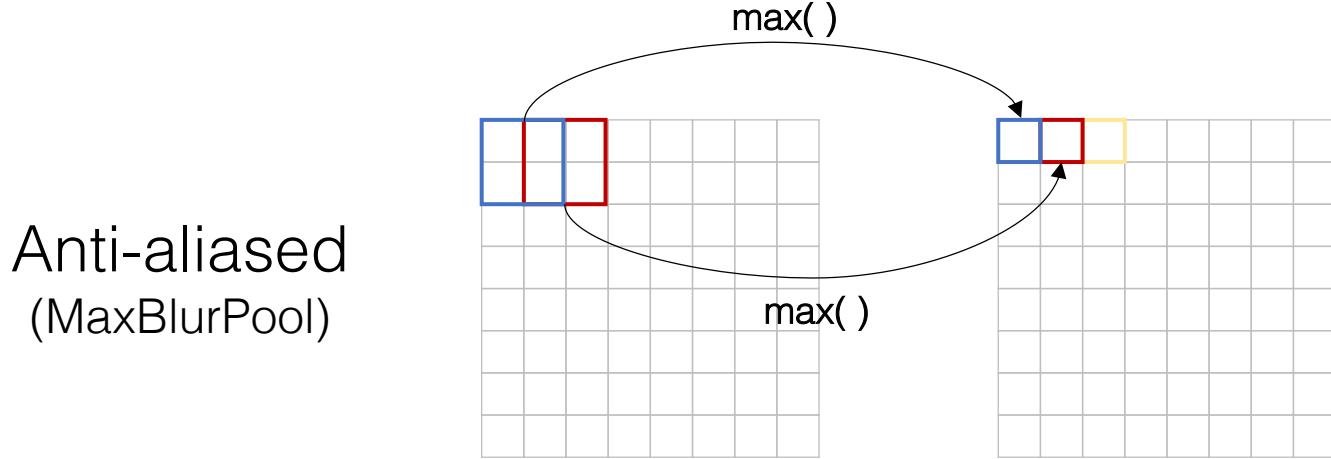
heavy aliasing



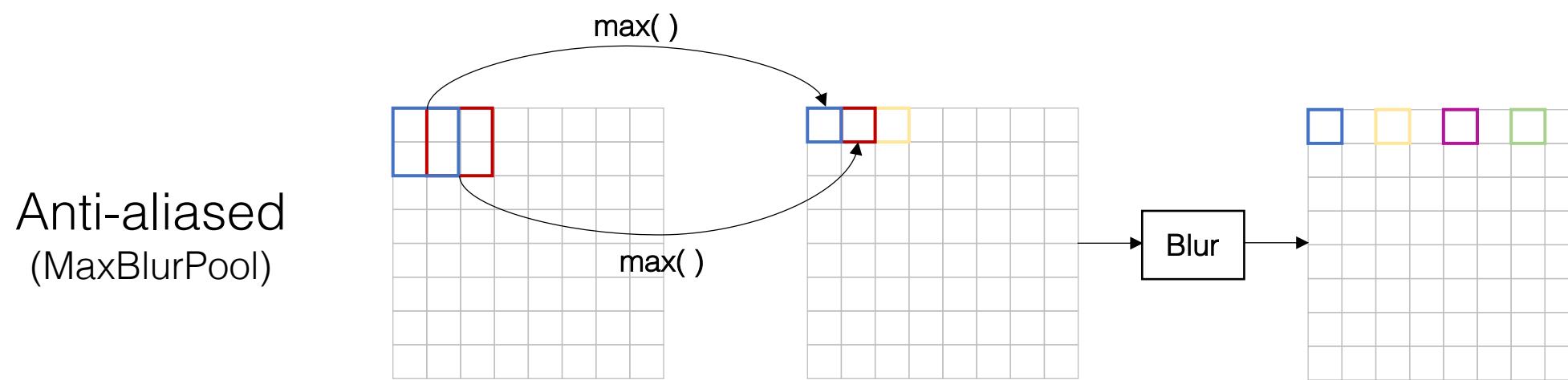
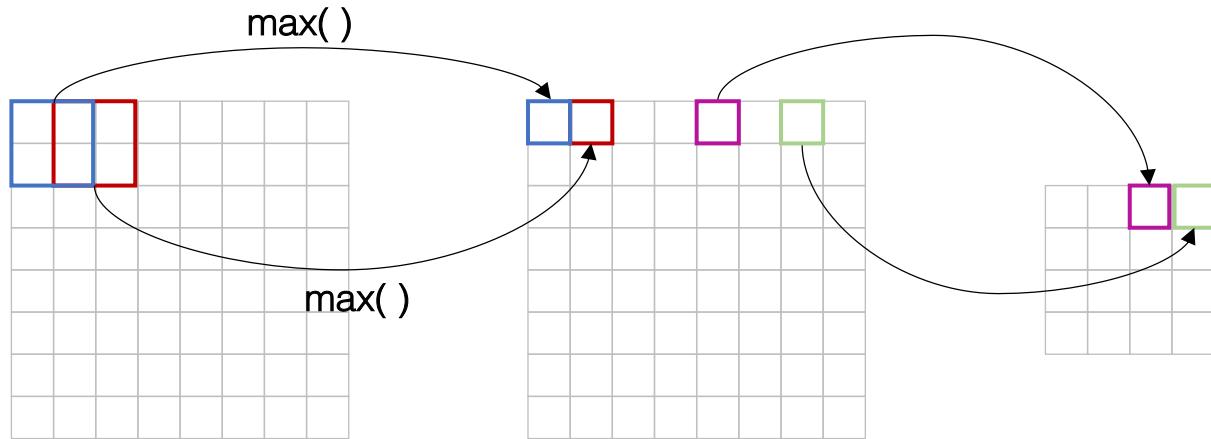
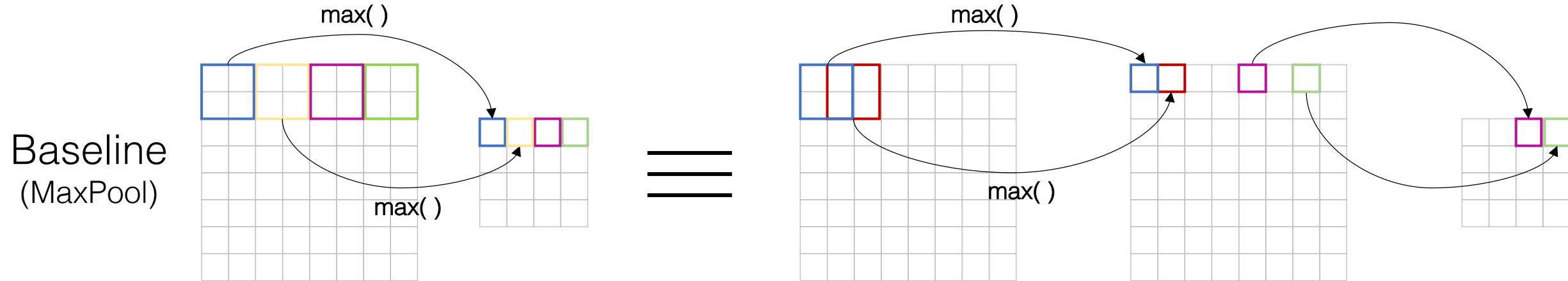
(1) Max (dense evaluation) + (2) Subsampling
no aliasing heavy aliasing



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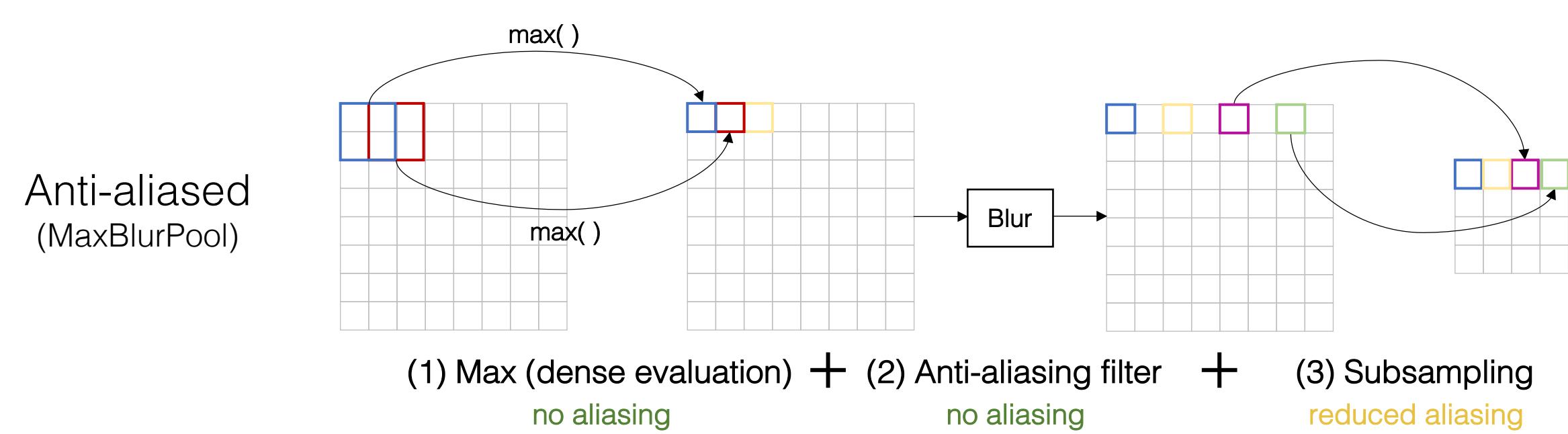
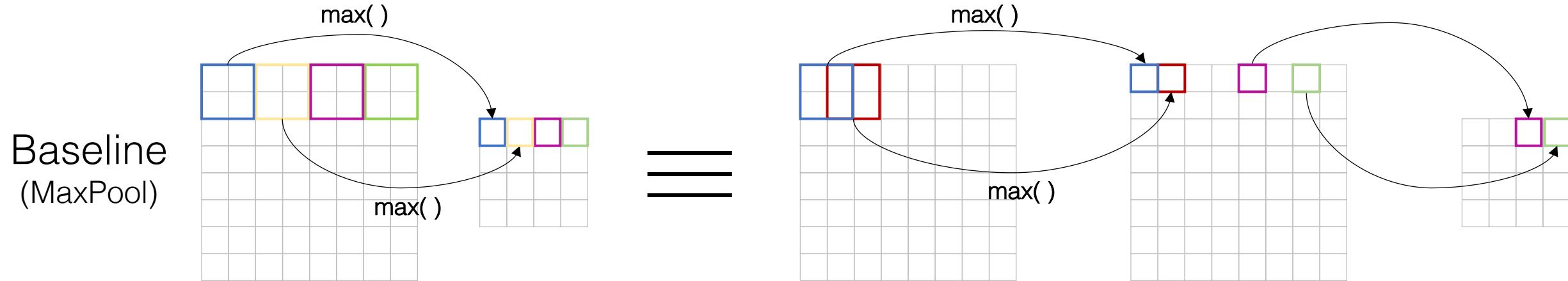
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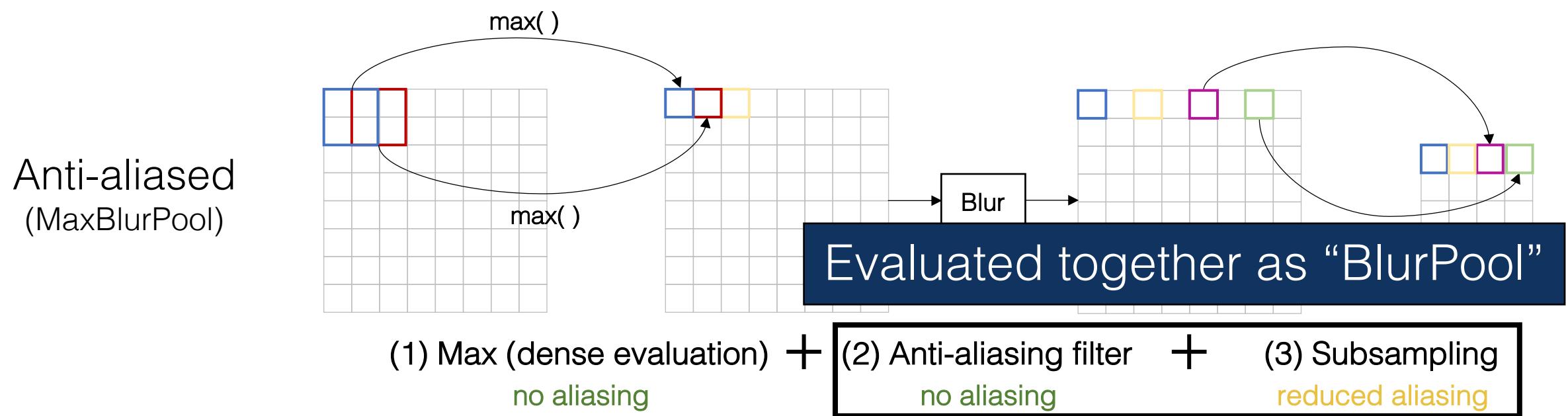
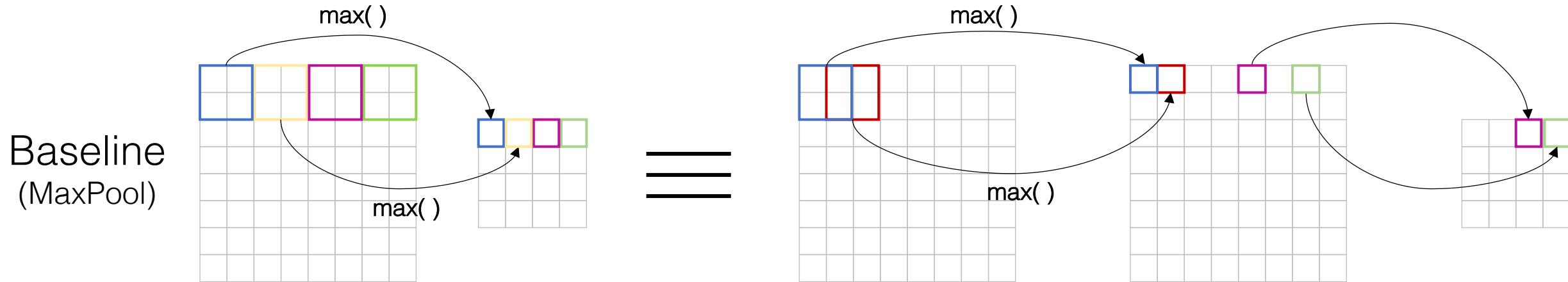


(1) Max (dense evaluation) + (2) Anti-aliasing filter

no aliasing

no aliasing





Antialiasing any downsampling layer

- Max Pool
 - VGG, Alexnet

Antialiasing any downsampling layer

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 - VGG, Alexnet
- Strided Convolution
 - Resnet, MobileNetv2

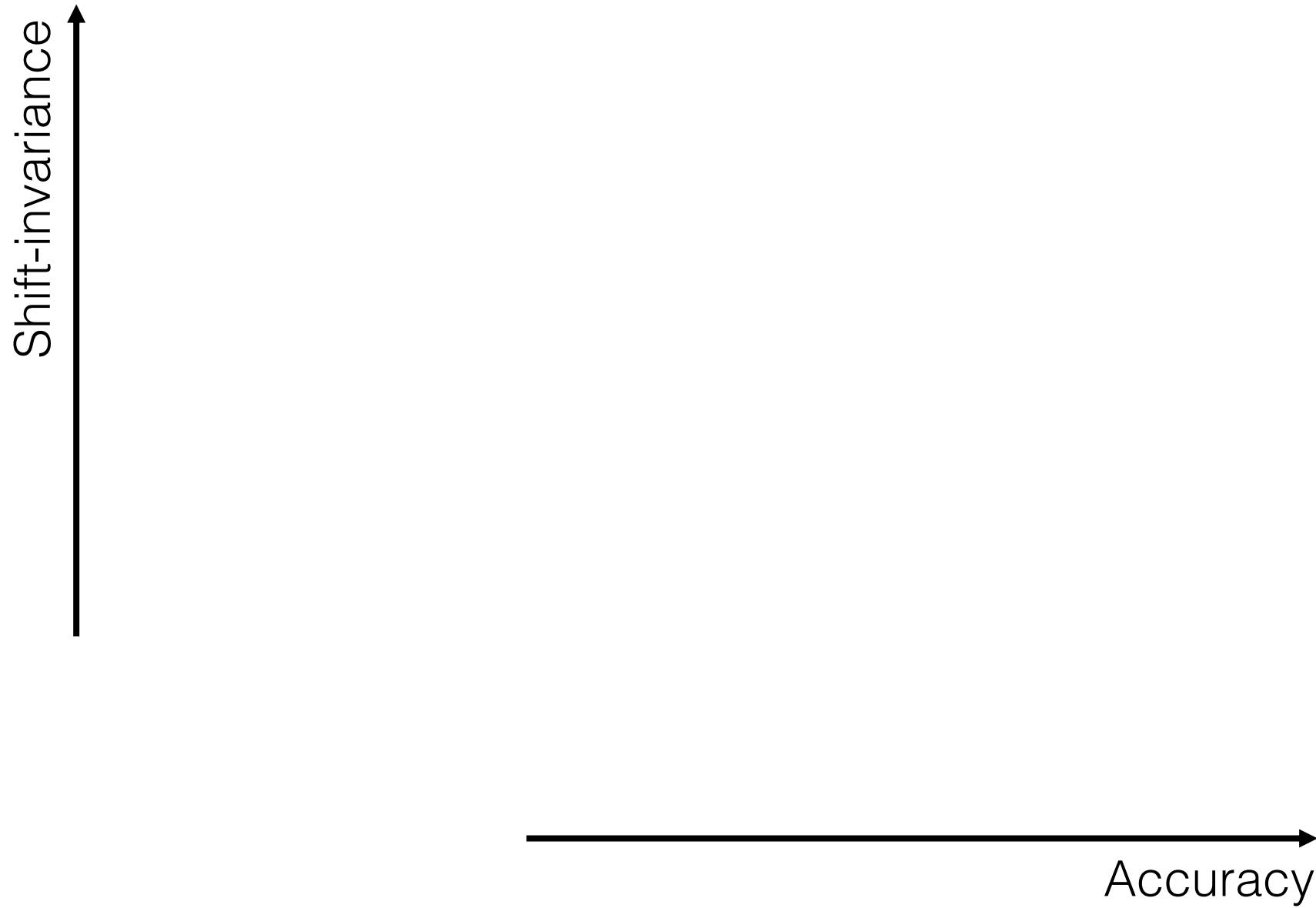
Antialiasing any downsampling layer

- Max Pool
 - VGG, Alexnet
- Strided Convolution
 - Resnet, MobileNetv2
- Average Pool
 - DenseNet

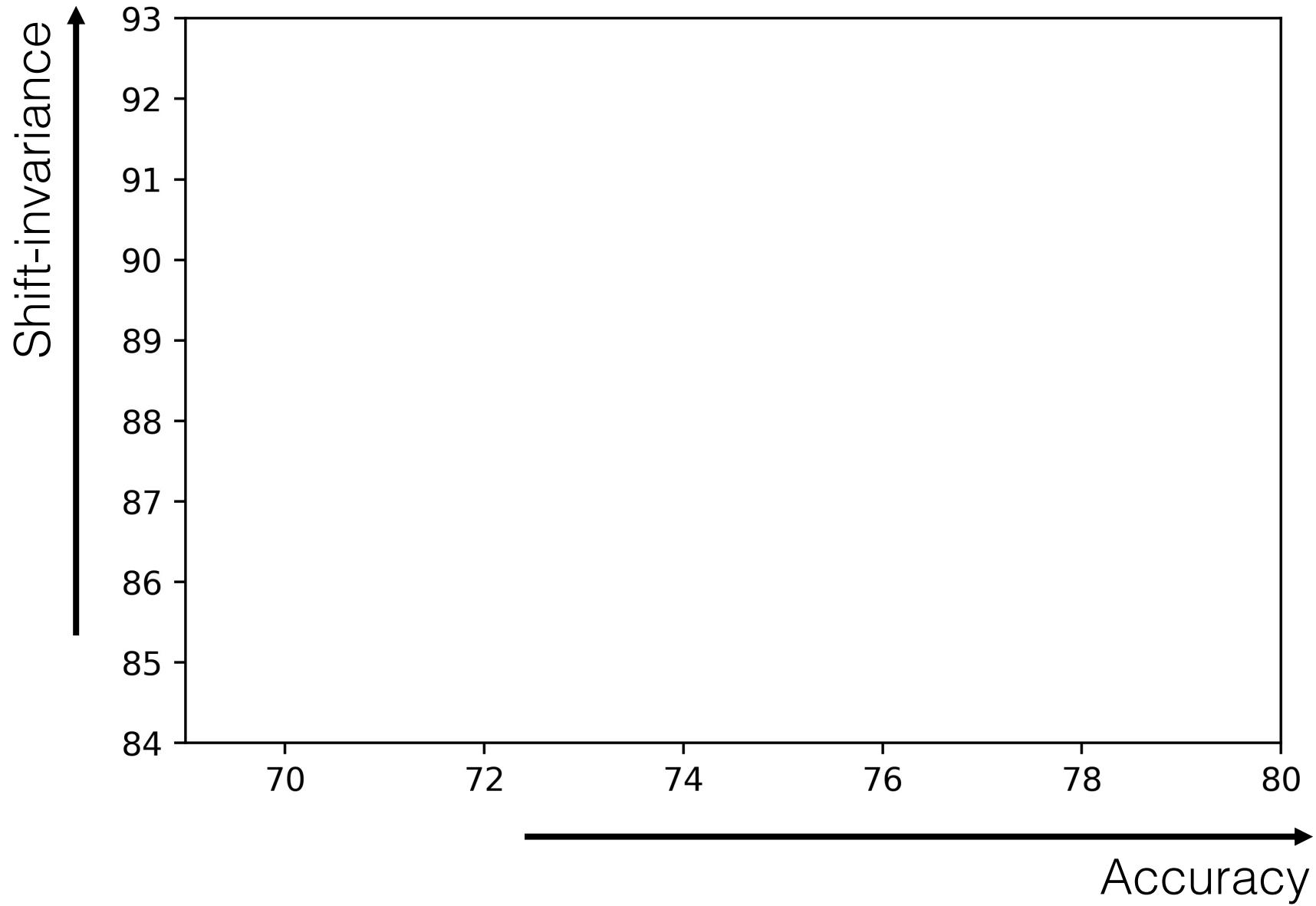
ImageNet



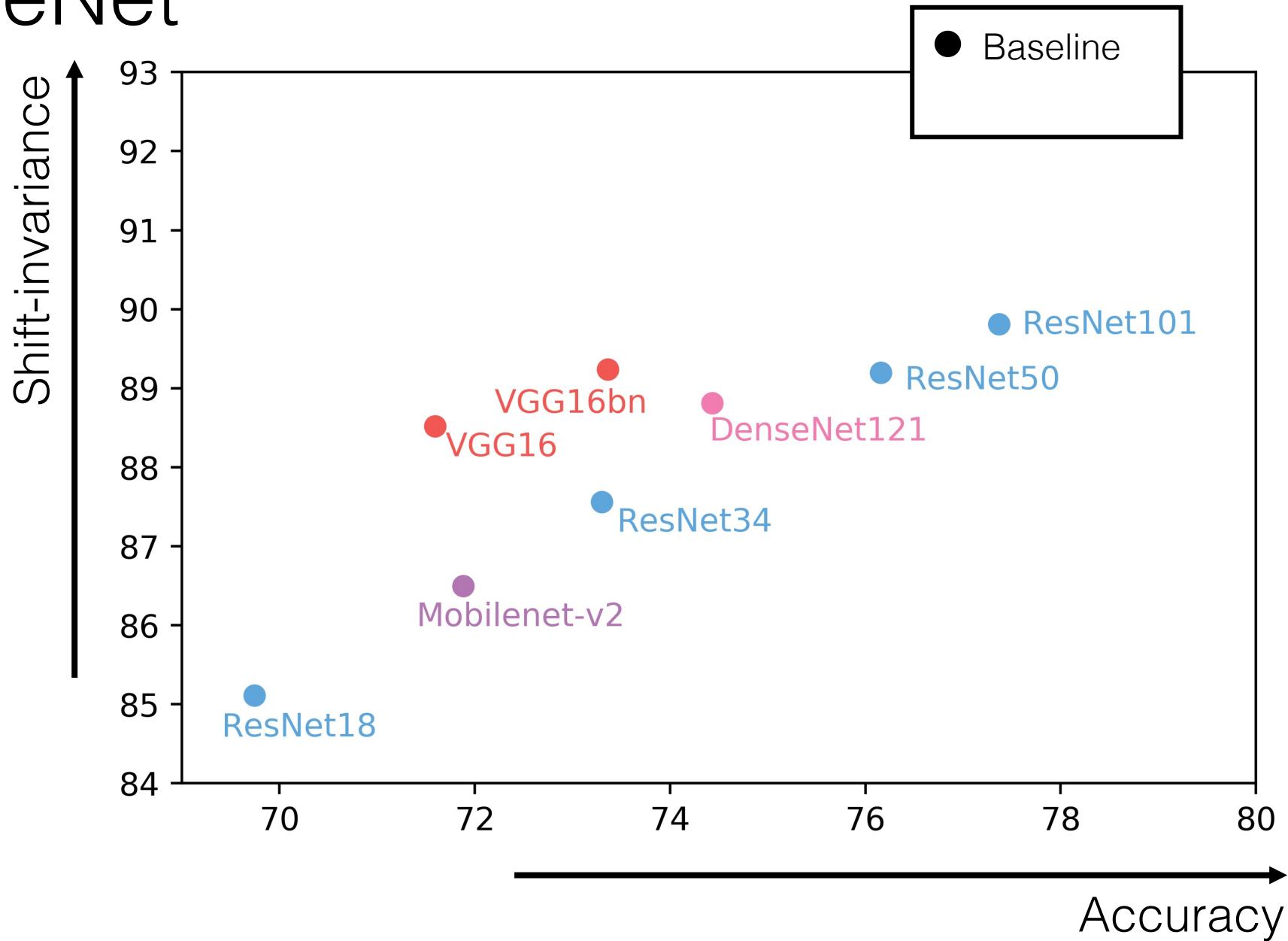
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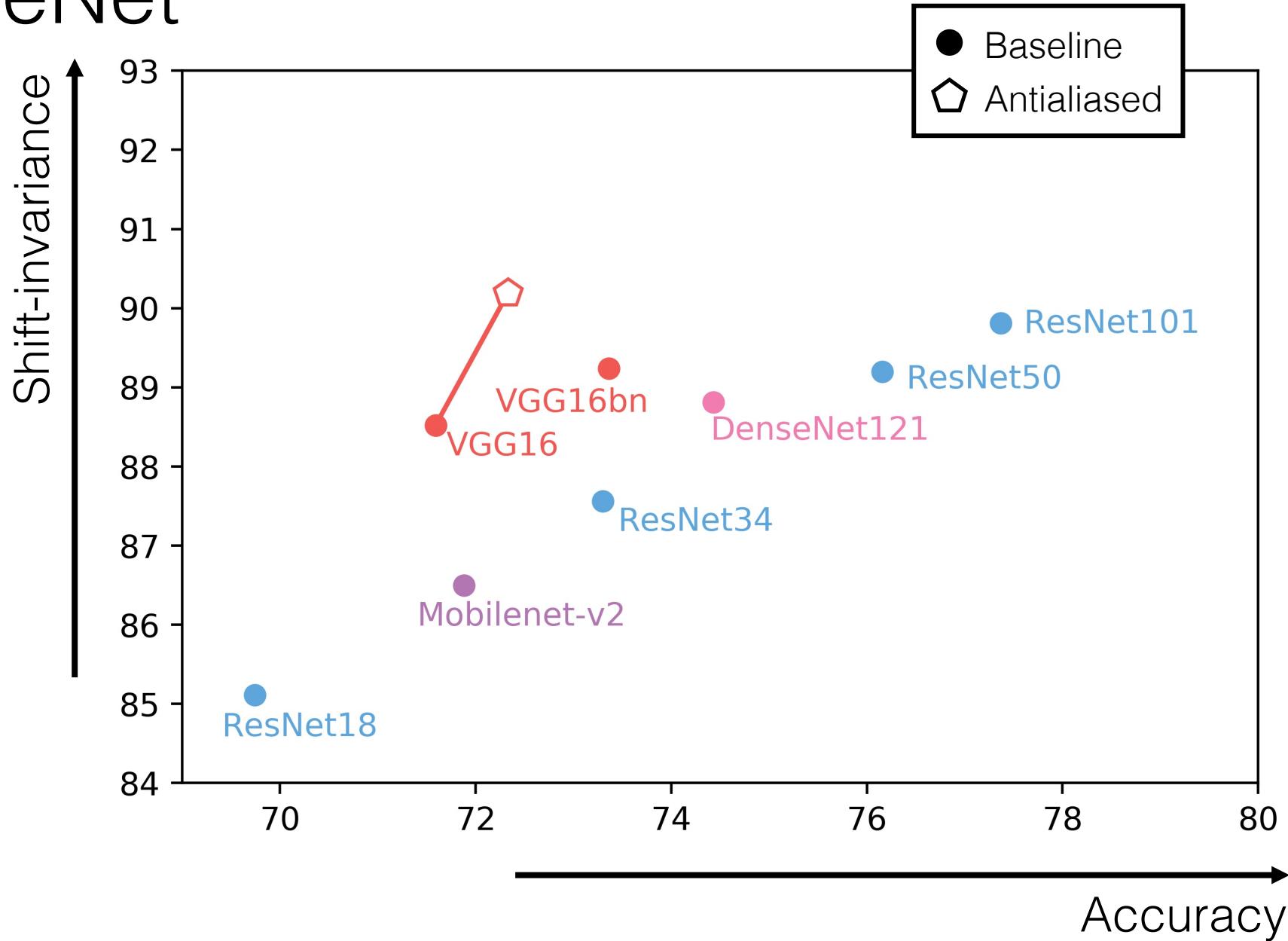
ImageNet



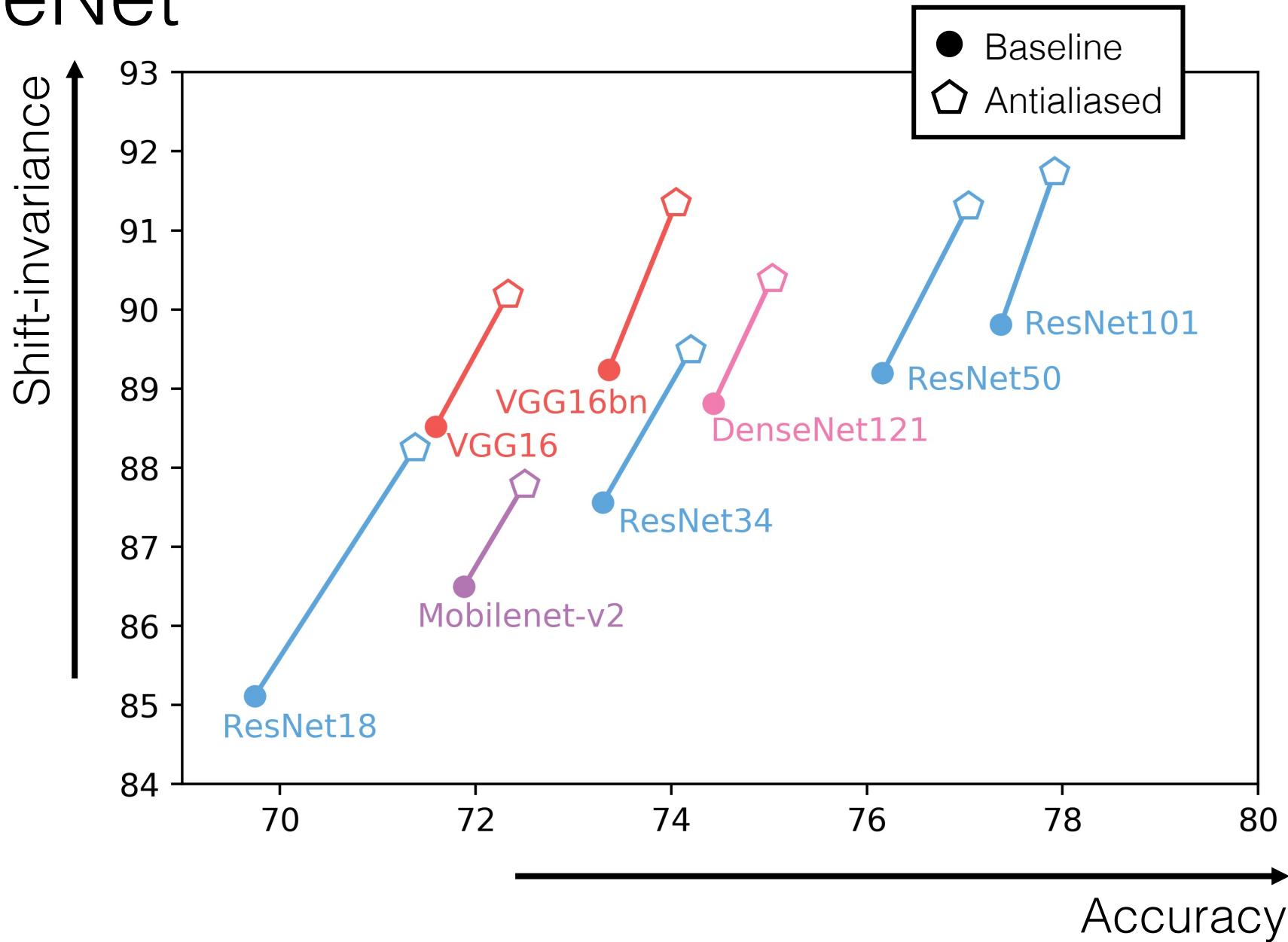
ImageNet



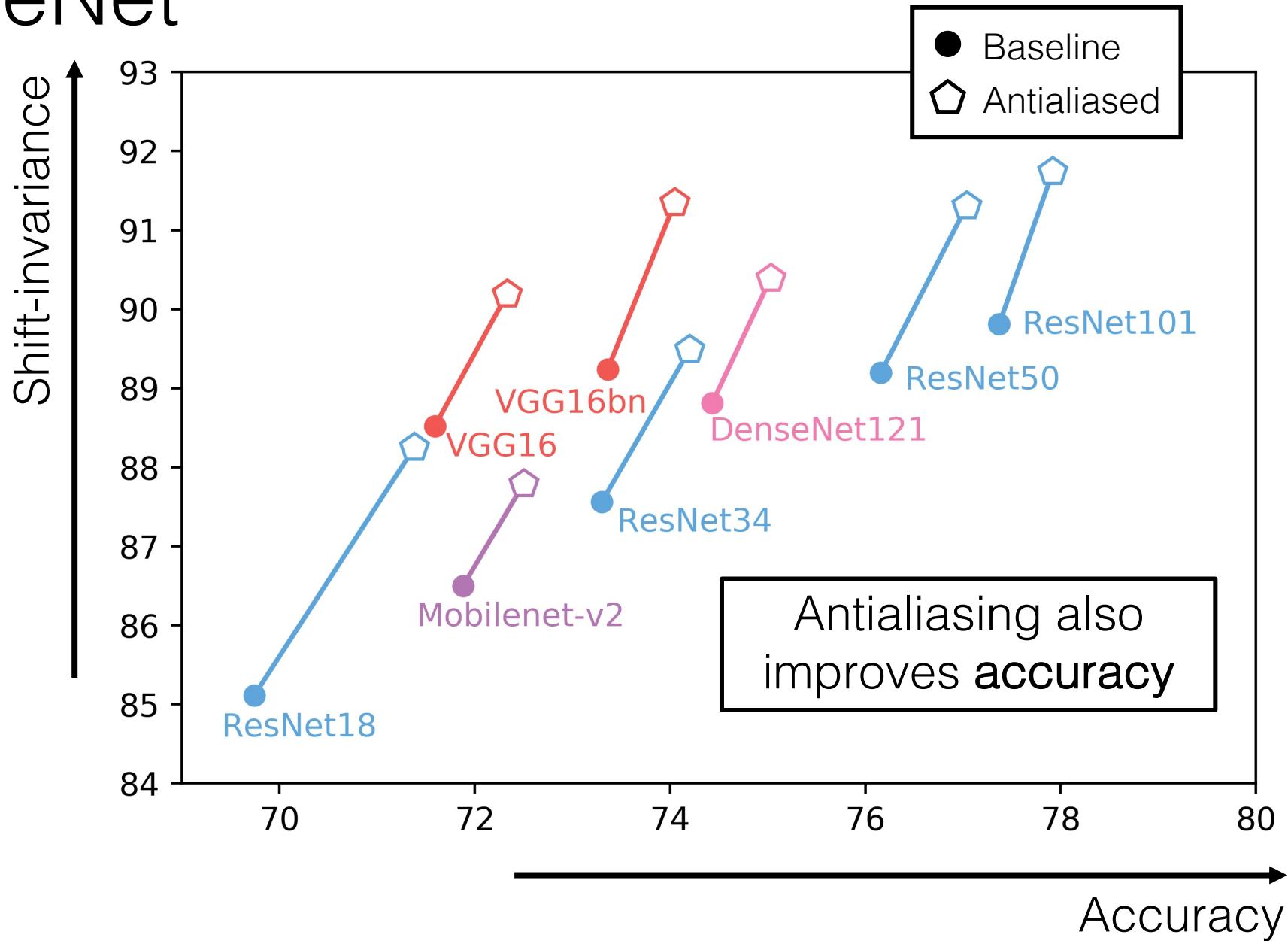
ImageNet



ImageNet



ImageNet



Discussion

Striding aliases (`stride=2`)

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+ Improved shift-equivariance

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Antialiasing code, pretrained models

<https://richzhang.github.io/antialiased-cnns/>

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Add antialiasing filter

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Thank you!