

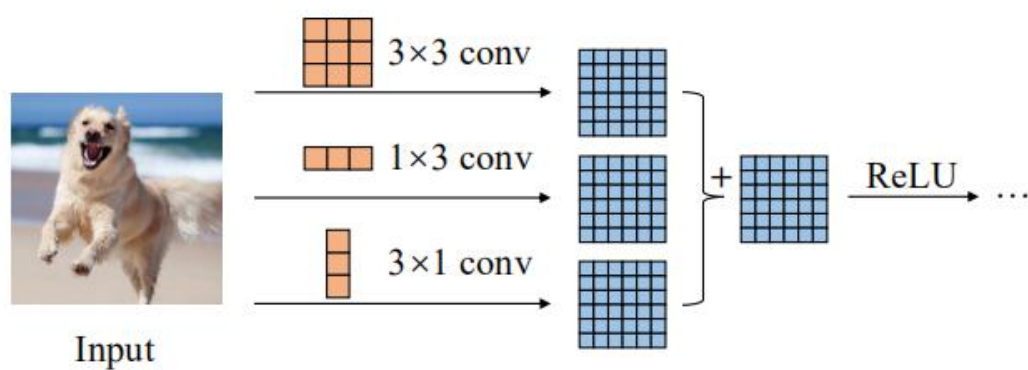


非对称卷积(Asymmetric Convolution)
异构卷积(Heterogeneous Convolution)
八度卷积(Octave Convolution)

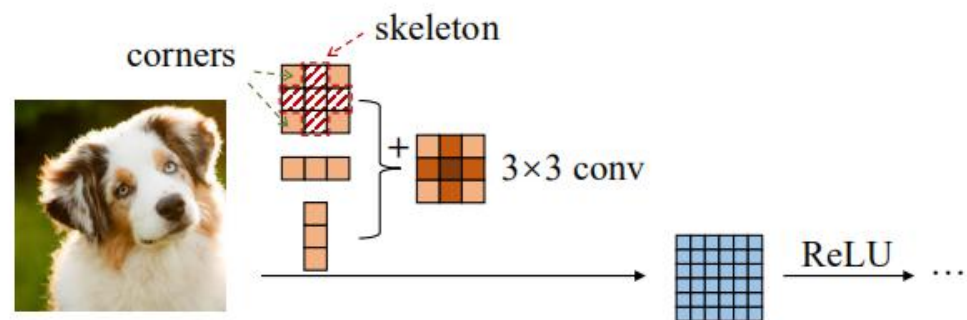


非对称卷积(Asymmetric Convolution)

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Training-time ACNet

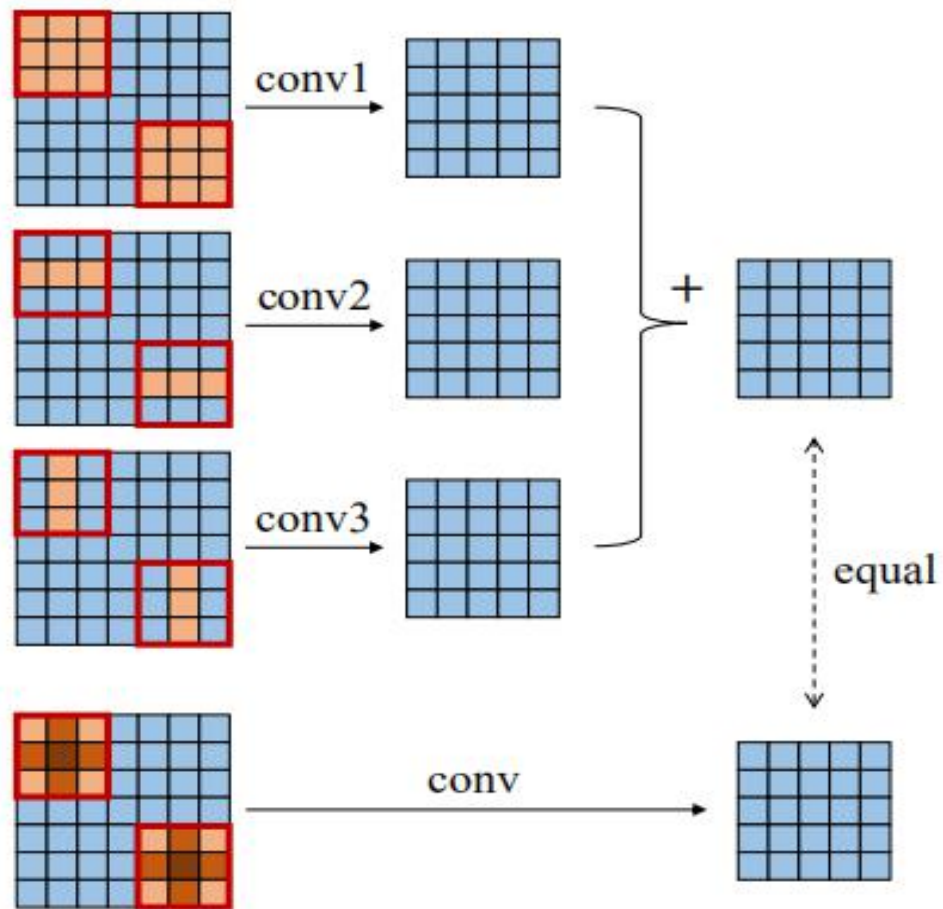


Deployed model



非对称卷积(Asymmetric Convolution)

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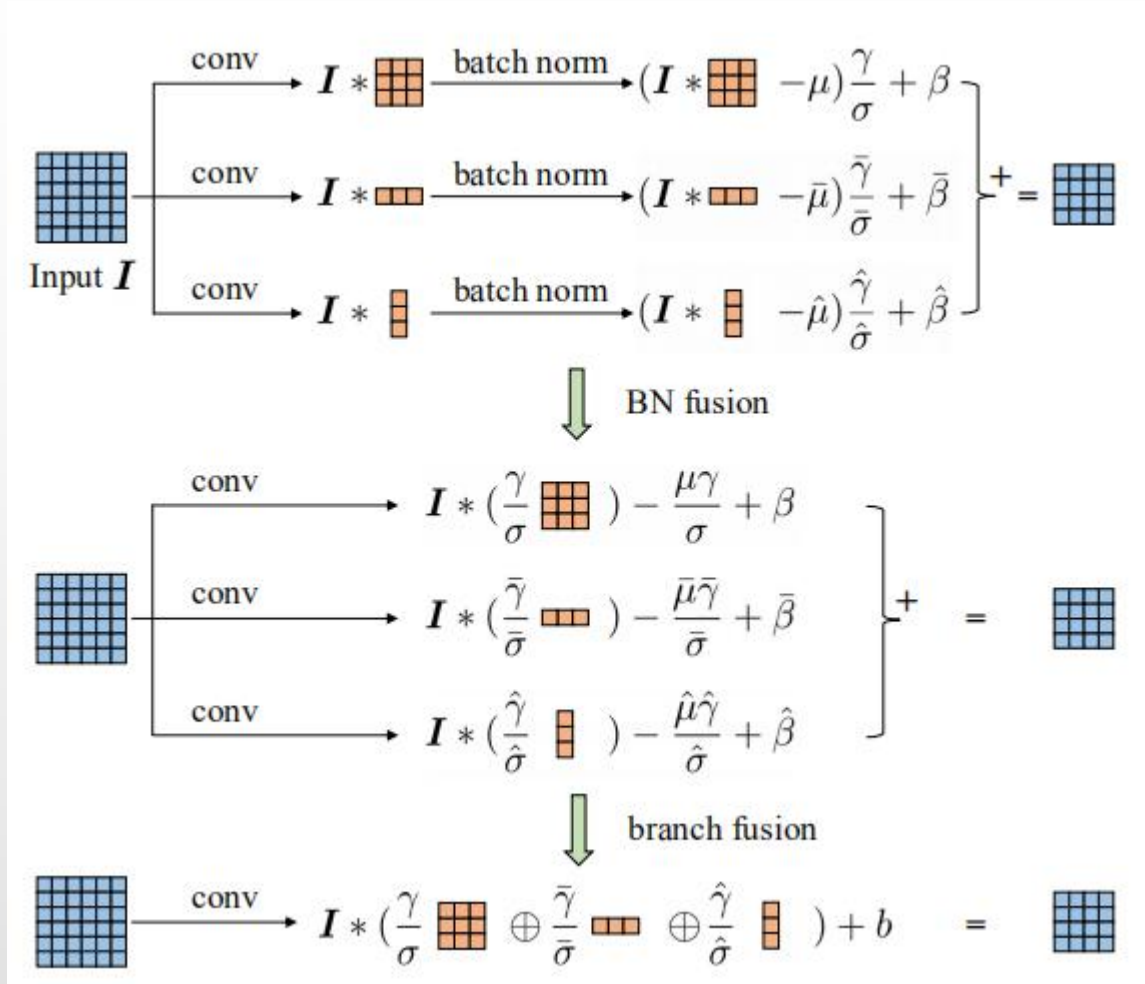


利用卷积的可加性



非对称卷积(Asymmetric Convolution)

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非对称卷积(Asymmetric Convolution)

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0.826	0.860	0.833
0.850	0.935	0.867
0.821	0.854	0.827

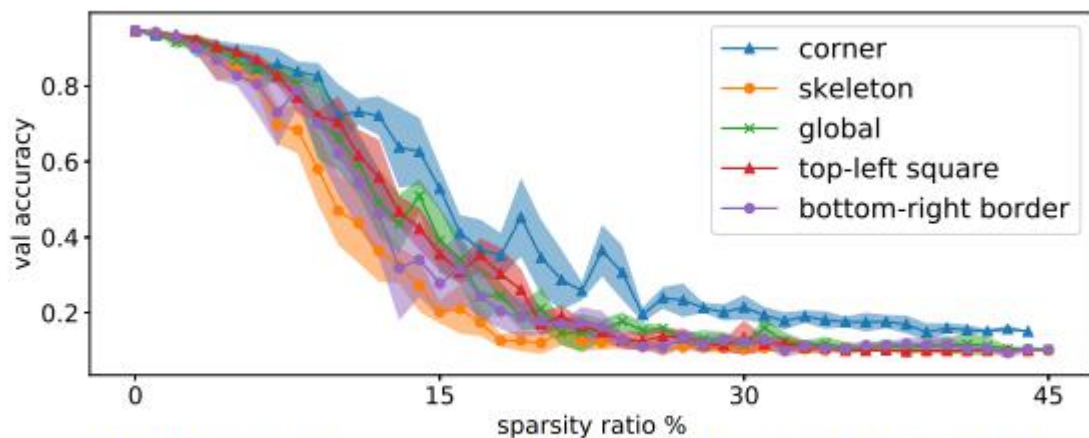
(a) Normal.

0.399	0.771	0.399
0.674	1.000	0.666
0.396	0.753	0.393

(b) ACNet, skeleton.

0.595	0.585	0.810
0.591	0.594	0.793
0.814	0.806	0.970

(c) ACNet, border.



(c) ACNet with the asymmetric kernels added to the border.



非对称卷积(Asymmetric Convolution)

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```
import torch.nn as nn
import torch.nn.init as init
from custom_layers.crop_layer import CropLayer
# source: https://github.com/DingXiaoH/ACNet/blob/master/acnet/acb.py

class ACBlock(nn.Module):

    def __init__(self, in_channels, out_channels, kernel_size, stride=1, padding=0):
        super(ACBlock, self).__init__()
        self.square_conv = nn.Conv2d(in_channels=in_channels, out_channels=out_channels,
                                       kernel_size=(kernel_size, kernel_size), stride=stride,
                                       padding=padding)
        self.square_bn = nn.BatchNorm2d(num_features=out_channels, affine=use_affine)
        self.ver_conv = nn.Conv2d(in_channels=in_channels, out_channels=out_channels, kernel_size=(kernel_size, 1),
                                   stride=stride, padding=padding)
        self.hor_conv = nn.Conv2d(in_channels=in_channels, out_channels=out_channels, kernel_size=(1, kernel_size),
                                   stride=stride, padding=padding)
        self.ver_bn = nn.BatchNorm2d(num_features=out_channels, affine=use_affine)
        self.hor_bn = nn.BatchNorm2d(num_features=out_channels, affine=use_affine)
```

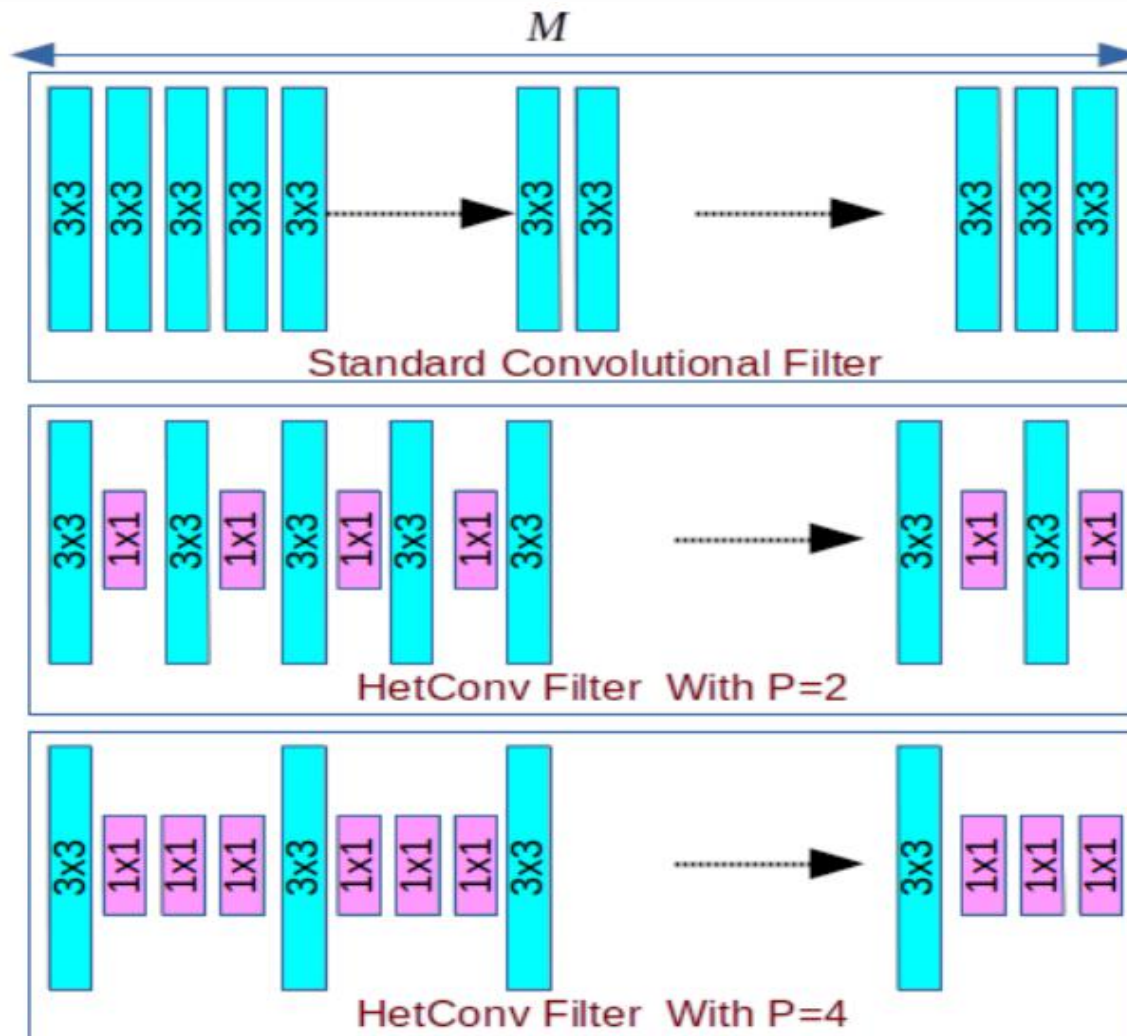


```
def forward(self, input):  
    if self.deploy:  
        return self.fused_conv(input)  
    else:  
        square_outputs = self.square_conv(input)  
        square_outputs = self.square_bn(square_outputs)  
        vertical_outputs = self.ver_conv(vertical_outputs)  
        vertical_outputs = self.ver_bn(vertical_outputs)  
        horizontal_outputs = self.hor_conv(horizontal_outputs)  
        horizontal_outputs = self.hor_bn(horizontal_outputs)  
        result = square_outputs + vertical_outputs + horizontal_outputs  
        return result
```



异构卷积(Heterogeneous Convolution)

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异构卷积(Heterogeneous Convolution)

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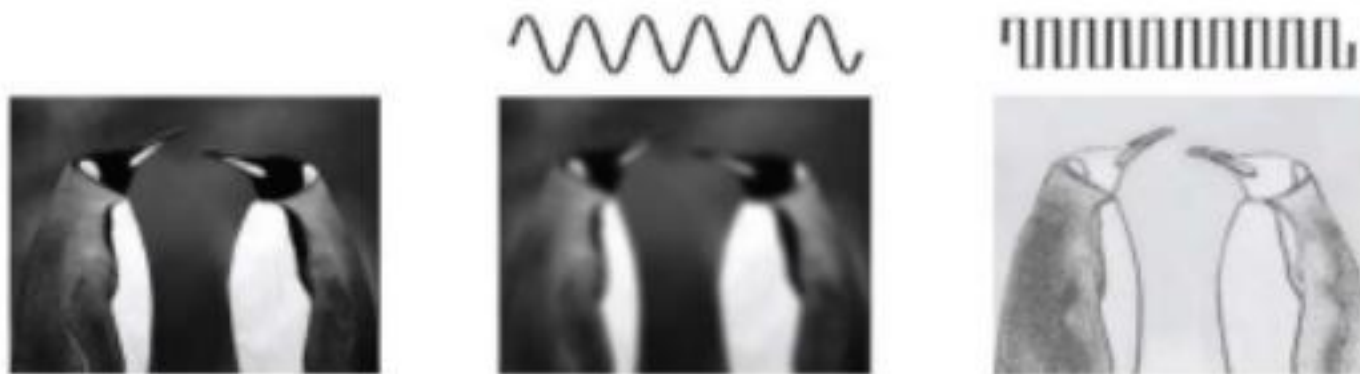
```
class HetConv(nn.Module):
    def __init__(self, in_channels, out_channels, p):
        super(HetConv, self).__init__()
        # Groupwise Convolution
        self.gwc = nn.Conv2d(in_channels, out_channels, kernel_size=3, padding=1, groups=p, bias=False)
        # Pointwise Convolution
        self.pwc = nn.Conv2d(in_channels, out_channels, kernel_size=1, bias=False)

    def forward(self, x):
        return self.gwc(x) + self.pwc(x)
```

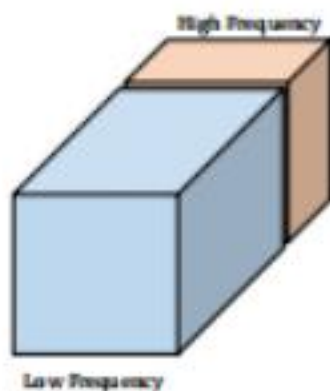


八度卷积(Octave Convolution)

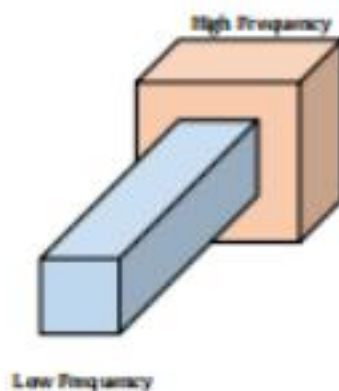
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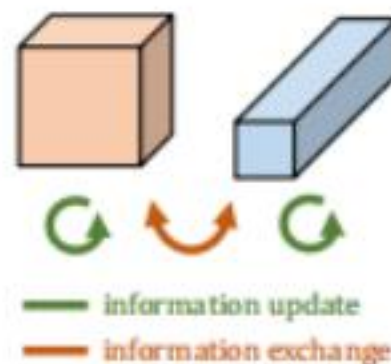
(a) Separating the low and high spatial frequency signal [1, 10].



(b)



(c)

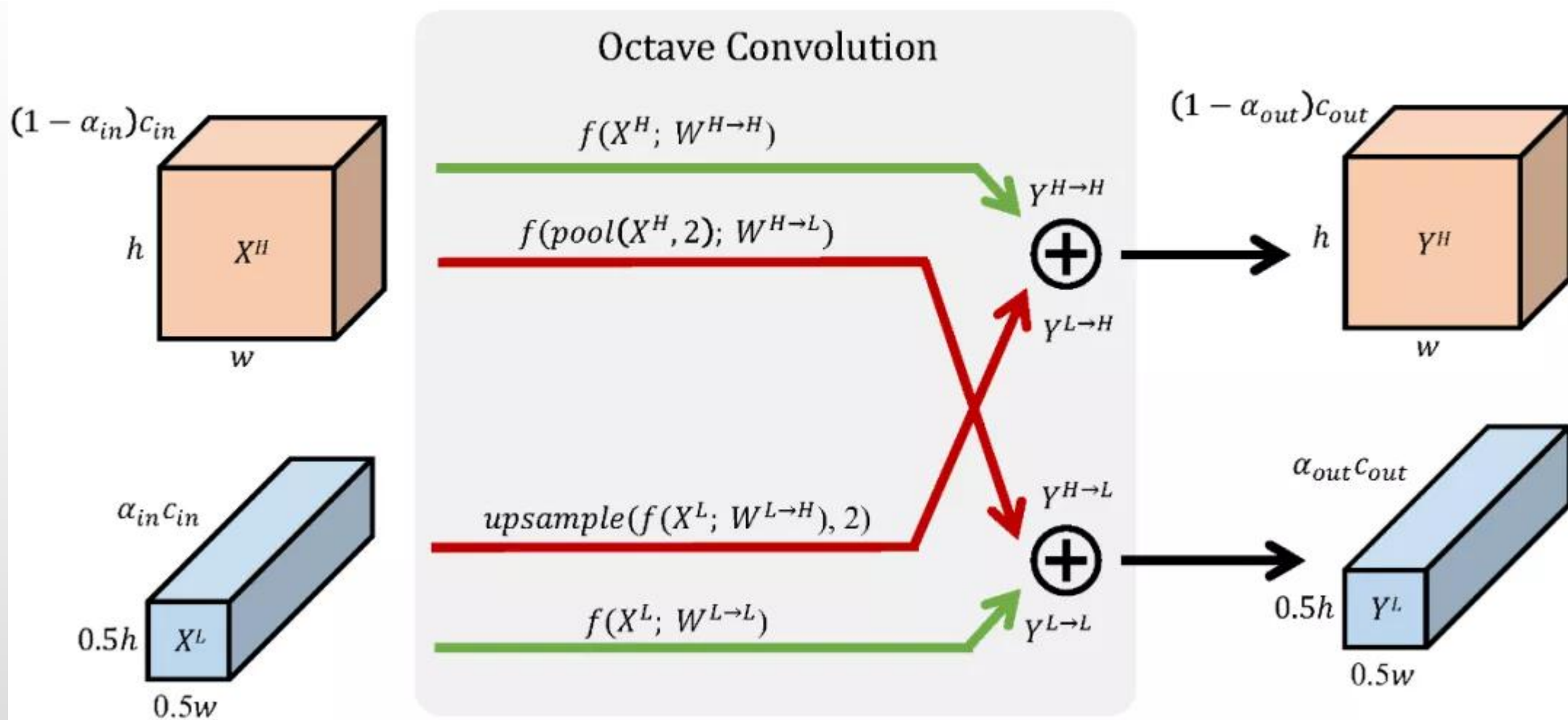


(d)



八度卷积(Octave Convolution)

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```
class OctaveConv(nn.Module):
```

```
    def __init__(self, in_channels, out_channels, kernel_size, alpha=0.5, stride=1, padding=1, dilation=1,
                  groups=1, bias=False):
```

```
        super(OctaveConv, self).__init__()
```

```
        kernel_size = kernel_size[0]
```

```
        self.h2g_pool = nn.AvgPool2d(kernel_size=(2, 2), stride=2)
```

```
        self.upsample = torch.nn.Upsample(scale_factor=2, mode='nearest')
```

```
        self.stride = stride
```

```
        self.l2l = torch.nn.Conv2d(int(alpha * in_channels), int(alpha * out_channels),
                                     kernel_size, 1, padding, dilation, groups, bias)
```

```
        self.l2h = torch.nn.Conv2d(int(alpha * in_channels), out_channels - int(alpha * out_channels),
                                     kernel_size, 1, padding, dilation, groups, bias)
```

```
        self.h2l = torch.nn.Conv2d(in_channels - int(alpha * in_channels), int(alpha * out_channels),
                                     kernel_size, 1, padding, dilation, groups, bias)
```

```
        self.h2h = torch.nn.Conv2d(in_channels - int(alpha * in_channels),
                                     out_channels - int(alpha * out_channels),
                                     kernel_size, 1, padding, dilation, groups, bias)
```




八度卷积(Octave Convolution)

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```
def forward(self, x):
    X_h, X_l = x
    if self.stride == 2:
        X_h, X_l = self.h2g_pool(X_h), self.h2g_pool(X_l)
        X_h2l = self.h2g_pool(X_h)
        X_h2h = self.h2h(X_h)
        X_l2h = self.l2h(X_l)
        X_l2l = self.l2l(X_l)
        X_h2l = self.h2l(X_h2l)

        X_l2h = self.upsample(X_l2h)
        X_h = X_l2h + X_h2h
        X_l = X_h2l + X_l2l
    return X_h, X_l
```



Tokens-to-Token ViT: Training Vision Transformers from Scratch on ImageNet

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感谢大家！