

图像分类 ResNet

PyTorch版

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环境配置

1. 驱动+CUDA11.8+cuDNN8.6.0

2. Python 3.8.12 (conda)

3. PyTorch 2.1.2

v2.1.2

Conda

OSX

```
# conda
conda install pytorch==2.1.2 torchvision==0.16.2 torchaudio==2.1.2 -c pytorch
```

Linux and Windows

```
# CUDA 11.8
conda install pytorch==2.1.2 torchvision==0.16.2 torchaudio==2.1.2 pytorch-cuda=11.8 -c pytorch -c nvidia
# CUDA 12.1
conda install pytorch==2.1.2 torchvision==0.16.2 torchaudio==2.1.2 pytorch-cuda=12.1 -c pytorch -c nvidia
# CPU Only
conda install pytorch==2.1.2 torchvision==0.16.2 torchaudio==2.1.2 cpuonly -c pytorch
```

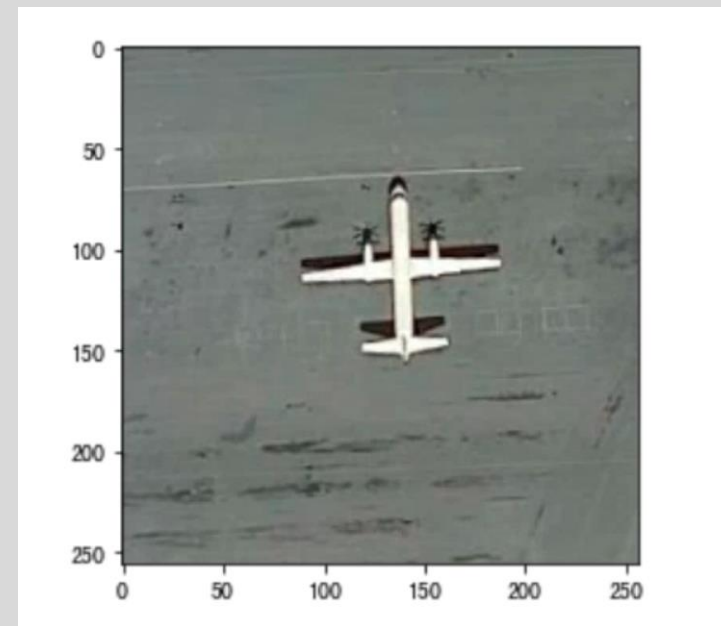
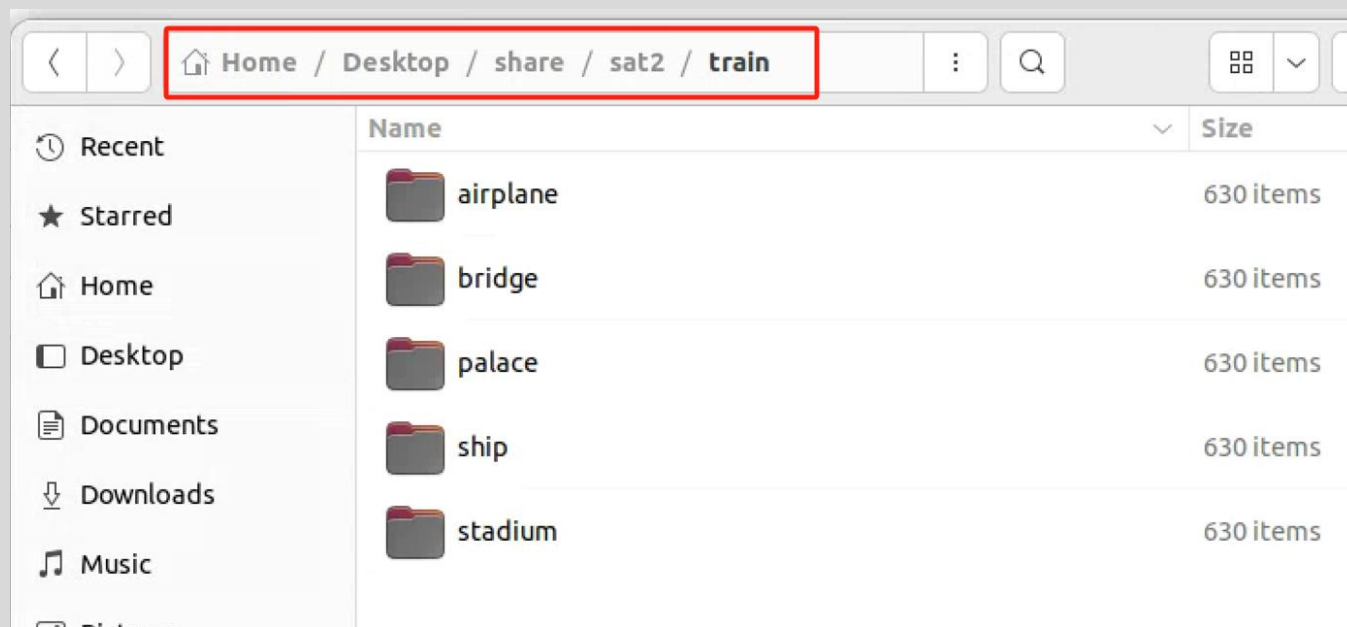
Wheel

OSX

```
pip install torch==2.1.2 torchvision==0.16.2 torchaudio==2.1.2
```

<https://pytorch.org/get-started/previous-versions/>

数据集结构



包含**31500**张遥感图像（45类*700张），**256x256**像素的彩色图。

本次使用其中的**5**类，划分每类**630**张为训练集，**70**张为测试集。

Pytorch 数据处理

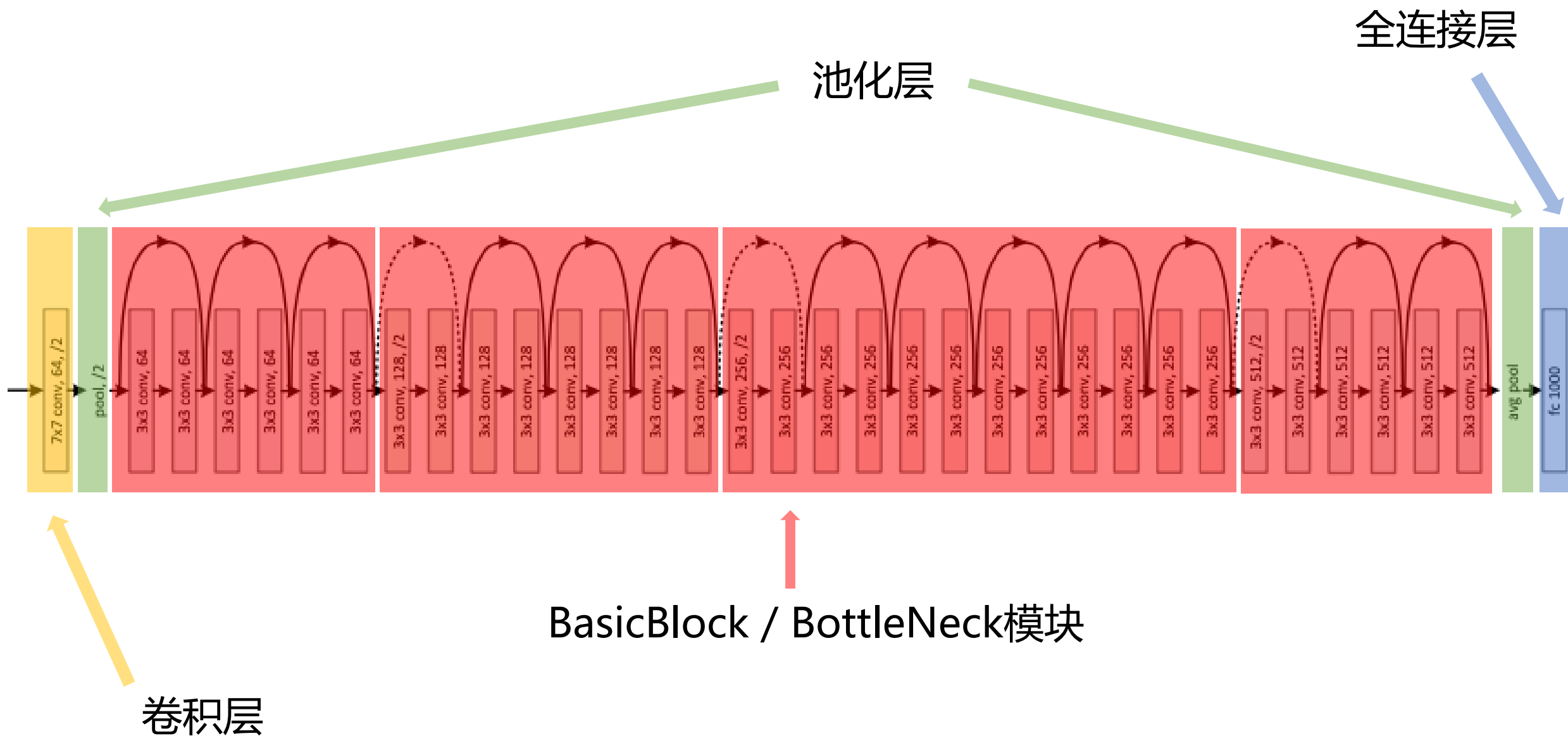
1. 按文件夹读取数据：

```
train_dir = '../sat2/train'    datasets.ImageFolder(train_dir, .....
```

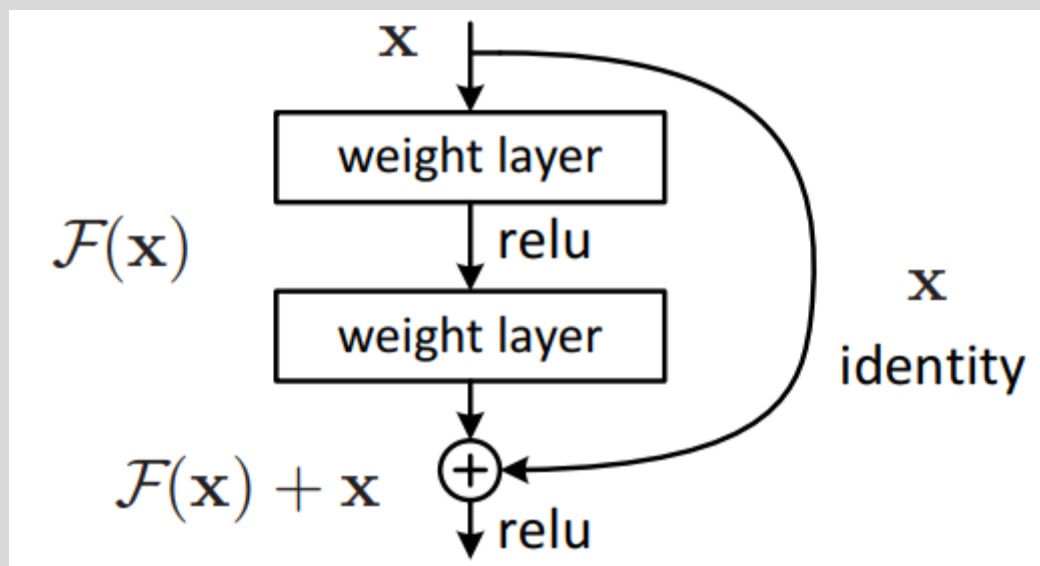
2. 图像调整（尺寸调整、格式转换）： transforms.Compose

3. 数据迭代（批次数量、随机）： DataLoader

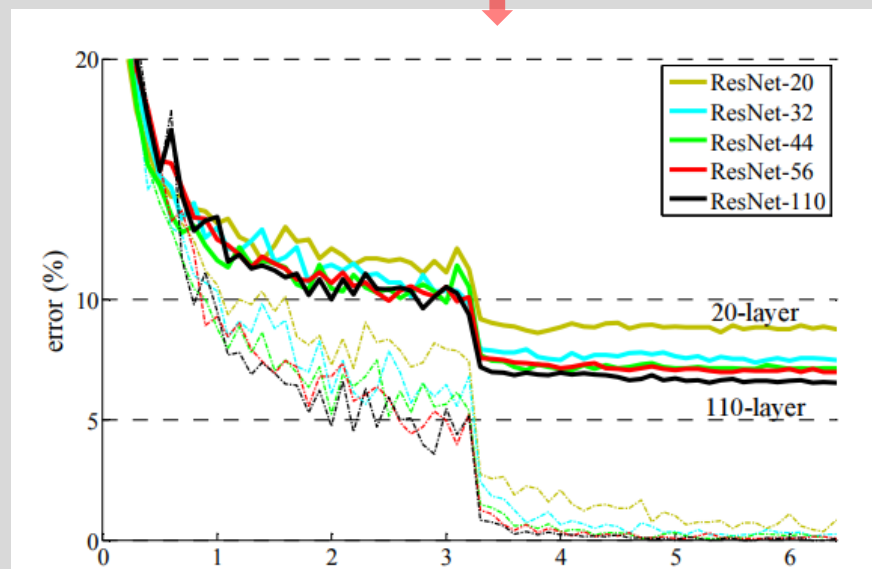
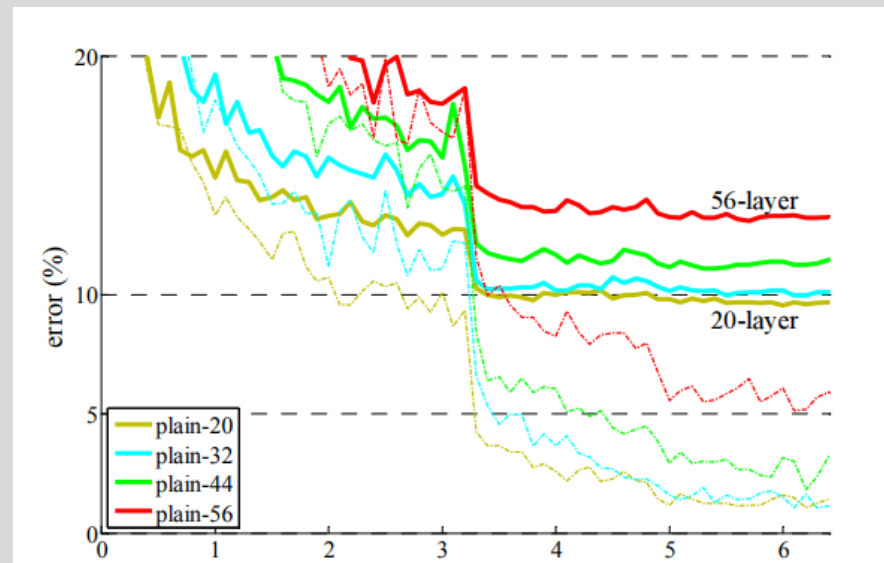
ResNet 结构



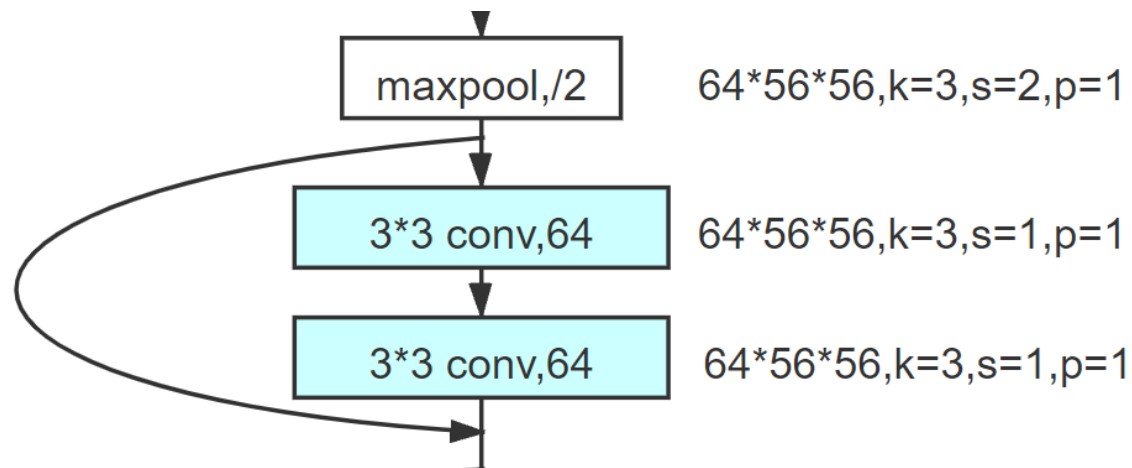
捷径分支



输出为 $H(x) = F(x) + x$,
权重层实际上是学习一种残差映射: $F(x) = H(x) - x$



BasicBlock 残差模块

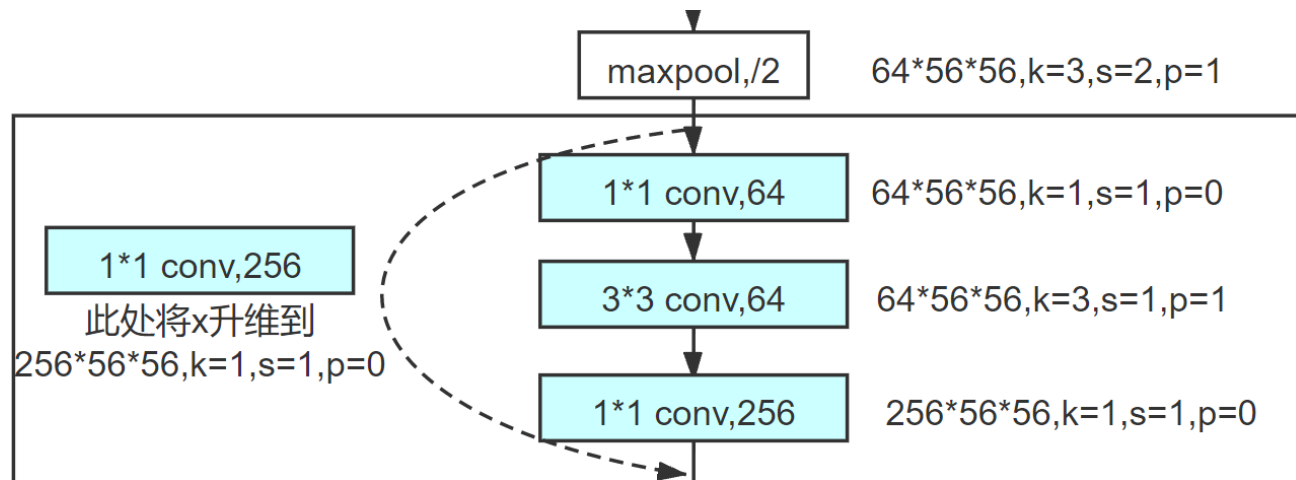


```
class BasicBlock(nn.Module):
    expansion = 1 # 输出通道不变

    def __init__(self, in_channels, out_channels, stride=1, downsample=None):
        super().__init__()
        self.conv1 = nn.Conv2d(in_channels, out_channels, kernel_size=3, stride=stride, padding=1, bias=False)
        self.bn1 = nn.BatchNorm2d(out_channels)
        self.relu = nn.ReLU(inplace=True)
        self.conv2 = nn.Conv2d(out_channels, out_channels, kernel_size=3, stride=1, padding=1, bias=False)
        self.bn2 = nn.BatchNorm2d(out_channels)
        self.downsample = downsample # 用于匹配维度

    def forward(self, x):
        identity = x if self.downsample is None else self.downsample(x)
        out = self.relu(self.bn1(self.conv1(x)))
        out = self.bn2(self.conv2(out))
        out += identity
        out = self.relu(out)
        return out
```


BottleNeck 残差模块



```
class BottleNeck(nn.Module):
    expansion = 4 # 输出通道为输入的4倍

    def __init__(self, in_channels, out_channels, stride=1, downsample=None):
        super().__init__()
        self.conv1 = nn.Conv2d(in_channels, out_channels, kernel_size=1, stride=1, bias=False)
        self.bn1 = nn.BatchNorm2d(out_channels)
        self.conv2 = nn.Conv2d(out_channels, out_channels, kernel_size=3, stride=stride, padding=1, bias=False)
        self.bn2 = nn.BatchNorm2d(out_channels)
        self.conv3 = nn.Conv2d(out_channels, out_channels * 4, kernel_size=1, stride=1, bias=False)
        self.bn3 = nn.BatchNorm2d(out_channels * 4)
        self.relu = nn.ReLU(inplace=True)
        self.downsample = downsample # 用于匹配维度

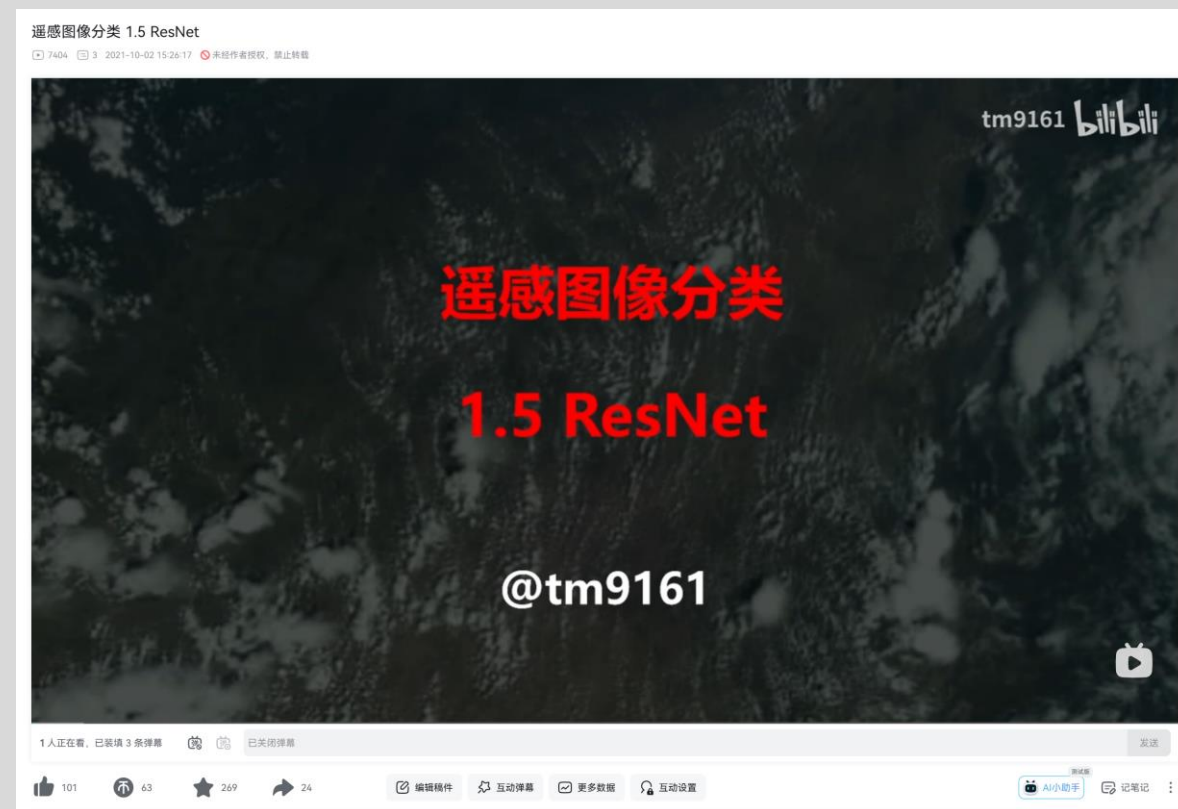
    def forward(self, x):
        identity = x if self.downsample is None else self.downsample(x)
        out = self.relu(self.bn1(self.conv1(x)))
        out = self.relu(self.bn2(self.conv2(out)))
        out = self.bn3(self.conv3(out))
        out += identity
        out = self.relu(out)
        return out
```


| layer name | output size | 18-layer | 34-layer | 50-layer | 101-layer | 152-layer |
|------------|------------------|-----------------------------------------------------------------------------|-----------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------|
| conv1 | 112×112 | 7×7 , 64, stride 2 | | | | |
| conv2_x | 56×56 | 3×3 max pool, stride 2 | | | | |
| | | $\begin{bmatrix} 3 \times 3, 64 \\ 3 \times 3, 64 \end{bmatrix} \times 2$ | $\begin{bmatrix} 3 \times 3, 64 \\ 3 \times 3, 64 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 64 \\ 3 \times 3, 64 \\ 1 \times 1, 256 \end{bmatrix} \times 3$ |
| conv3_x | 28×28 | $\begin{bmatrix} 3 \times 3, 128 \\ 3 \times 3, 128 \end{bmatrix} \times 2$ | $\begin{bmatrix} 3 \times 3, 128 \\ 3 \times 3, 128 \end{bmatrix} \times 4$ | $\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 4$ | $\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 4$ | $\begin{bmatrix} 1 \times 1, 128 \\ 3 \times 3, 128 \\ 1 \times 1, 512 \end{bmatrix} \times 8$ |
| conv4_x | 14×14 | $\begin{bmatrix} 3 \times 3, 256 \\ 3 \times 3, 256 \end{bmatrix} \times 2$ | $\begin{bmatrix} 3 \times 3, 256 \\ 3 \times 3, 256 \end{bmatrix} \times 6$ | $\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 6$ | $\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 23$ | $\begin{bmatrix} 1 \times 1, 256 \\ 3 \times 3, 256 \\ 1 \times 1, 1024 \end{bmatrix} \times 36$ |
| conv5_x | 7×7 | $\begin{bmatrix} 3 \times 3, 512 \\ 3 \times 3, 512 \end{bmatrix} \times 2$ | $\begin{bmatrix} 3 \times 3, 512 \\ 3 \times 3, 512 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$ | $\begin{bmatrix} 1 \times 1, 512 \\ 3 \times 3, 512 \\ 1 \times 1, 2048 \end{bmatrix} \times 3$ |
| | 1×1 | average pool, 1000-d fc, softmax | | | | |
| FLOPs | | 1.8×10^9 | 3.6×10^9 | 3.8×10^9 | 7.6×10^9 | 11.3×10^9 |

Downsampling is performed by conv3 1, conv4 1, and conv5 1 with a stride of 2



<https://www.bilibili.com/video/BV1n7Viz9EXQ>



<https://www.bilibili.com/video/BV1Ru411f7Y6/>

图像分类数据集下载: <https://pan.baidu.com/s/1FCOUIhplI7aObASpBwNIhw>

提取码: yqvb