

# Model Description

In [1]:

```
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.nn.init as init

from torch.autograd import Variable

def _weights_init(m):
    classname = m.__class__.__name__
    #print(classname)
    if isinstance(m, nn.Linear) or isinstance(m, nn.Conv2d):
        init.kaiming_normal_(m.weight)

class LambdaLayer(nn.Module):
    def __init__(self, lambd):
        super(LambdaLayer, self).__init__()
        self.lambd = lambd

    def forward(self, x):
        return self.lambd(x)

class BasicBlock(nn.Module):
    expansion = 1

    def __init__(self, in_planes, planes, stride=1, option='A'):
        super(BasicBlock, self).__init__()
        self.conv1 = nn.Conv2d(in_planes, planes, kernel_size=3, stride=stride, padding=
1, bias=False)
        self.bn1 = nn.BatchNorm2d(planes)
        self.conv2 = nn.Conv2d(planes, planes, kernel_size=3, stride=1, padding=1, bias=
False)
        self.bn2 = nn.BatchNorm2d(planes)

        self.shortcut = nn.Sequential()
        if stride != 1 or in_planes != planes:
            if option == 'A':
                """
                For CIFAR10 ResNet paper uses option A.
                """
                self.shortcut = LambdaLayer(lambda x:
                    F.pad(x[:, :, ::2, ::2], (0, 0, 0, 0, plane
s//4, planes//4), "constant", 0))
            elif option == 'B':
                self.shortcut = nn.Sequential(
                    nn.Conv2d(in_planes, self.expansion * planes, kernel_size=1, stride
=stride, bias=False),
                    nn.BatchNorm2d(self.expansion * planes)
                )

    def forward(self, x):
        out = F.relu(self.bn1(self.conv1(x)))
        out = self.bn2(self.conv2(out))
        out += self.shortcut(x)
        out = F.relu(out)
        return out

class ResNet(nn.Module):
    def __init__(self, block, num_blocks, num_classes=100):
        super(ResNet, self).__init__()
        self.in_planes = 16
```

```

        self.conv1 = nn.Conv2d(3, 16, kernel_size=3, stride=1, padding=1, bias=False)
        self.bn1 = nn.BatchNorm2d(16)
        self.layer1 = self._make_layer(block, 16, num_blocks[0], stride=1)
        self.layer2 = self._make_layer(block, 32, num_blocks[1], stride=2)
        self.layer3 = self._make_layer(block, 64, num_blocks[2], stride=2)
        self.linear = nn.Linear(64, num_classes)

    def apply(_weights_init)

def _make_layer(self, block, planes, num_blocks, stride):
    strides = [stride] + [1]*(num_blocks-1)
    layers = []
    for stride in strides:
        layers.append(block(self.in_planes, planes, stride))
        self.in_planes = planes * block.expansion

    return nn.Sequential(*layers)

def forward(self, x):
    out = F.relu(self.bn1(self.conv1(x)))
    out = self.layer1(out)
    out = self.layer2(out)
    out = self.layer3(out)
    out = F.avg_pool2d(out, out.size()[3])
    out = out.view(out.size(0), -1)
    out = self.linear(out)
    return out

def resnet20(num_classes=100):
    return ResNet(BasicBlock, [3, 3, 3], num_classes=num_classes)

```

In [2]:

```
model = resnet20(num_classes=100)
```

In [3]:

```
print(model)
```

```

ResNet(
  (conv1): Conv2d(3, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
  (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv2): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (shortcut): Sequential()
    )
    (1): BasicBlock(
      (conv1): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv2): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (shortcut): Sequential()
    )
    (2): BasicBlock(
      (conv1): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (conv2): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)

```

```

se)
    (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): Sequential()
    )
)
(layer2): Sequential(
  (0): BasicBlock(
    (conv1): Conv2d(16, 32, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): LambdaLayer()
  )
  (1): BasicBlock(
    (conv1): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): Sequential()
  )
  (2): BasicBlock(
    (conv1): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): Sequential()
  )
)
(layer3): Sequential(
  (0): BasicBlock(
    (conv1): Conv2d(32, 64, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): LambdaLayer()
  )
  (1): BasicBlock(
    (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (shortcut): Sequential()
  )
  (2): BasicBlock(
    (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr
ue)
    (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=Fal
se)
    (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=Tr

```

```

ue)
    (shortcut): Sequential()
    )
)
(linear): Linear(in_features=64, out_features=100, bias=True)
)

```

In [6]:

```
from torchsummary import summary
```

In [10]:

```
summary(model, (3, 224, 224))
```

Layer (type)	Output Shape	Param #
=====		
Conv2d-1	[-1, 16, 224, 224]	432
BatchNorm2d-2	[-1, 16, 224, 224]	32
Conv2d-3	[-1, 16, 224, 224]	2,304
BatchNorm2d-4	[-1, 16, 224, 224]	32
Conv2d-5	[-1, 16, 224, 224]	2,304
BatchNorm2d-6	[-1, 16, 224, 224]	32
BasicBlock-7	[-1, 16, 224, 224]	0
Conv2d-8	[-1, 16, 224, 224]	2,304
BatchNorm2d-9	[-1, 16, 224, 224]	32
Conv2d-10	[-1, 16, 224, 224]	2,304
BatchNorm2d-11	[-1, 16, 224, 224]	32
BasicBlock-12	[-1, 16, 224, 224]	0
Conv2d-13	[-1, 16, 224, 224]	2,304
BatchNorm2d-14	[-1, 16, 224, 224]	32
Conv2d-15	[-1, 16, 224, 224]	2,304
BatchNorm2d-16	[-1, 16, 224, 224]	32
BasicBlock-17	[-1, 16, 224, 224]	0
Conv2d-18	[-1, 32, 112, 112]	4,608
BatchNorm2d-19	[-1, 32, 112, 112]	64
Conv2d-20	[-1, 32, 112, 112]	9,216
BatchNorm2d-21	[-1, 32, 112, 112]	64
LambdaLayer-22	[-1, 32, 112, 112]	0
BasicBlock-23	[-1, 32, 112, 112]	0
Conv2d-24	[-1, 32, 112, 112]	9,216
BatchNorm2d-25	[-1, 32, 112, 112]	64
Conv2d-26	[-1, 32, 112, 112]	9,216
BatchNorm2d-27	[-1, 32, 112, 112]	64
BasicBlock-28	[-1, 32, 112, 112]	0
Conv2d-29	[-1, 32, 112, 112]	9,216
BatchNorm2d-30	[-1, 32, 112, 112]	64
Conv2d-31	[-1, 32, 112, 112]	9,216
BatchNorm2d-32	[-1, 32, 112, 112]	64
BasicBlock-33	[-1, 32, 112, 112]	0
Conv2d-34	[-1, 64, 56, 56]	18,432
BatchNorm2d-35	[-1, 64, 56, 56]	128
Conv2d-36	[-1, 64, 56, 56]	36,864
BatchNorm2d-37	[-1, 64, 56, 56]	128
LambdaLayer-38	[-1, 64, 56, 56]	0
BasicBlock-39	[-1, 64, 56, 56]	0
Conv2d-40	[-1, 64, 56, 56]	36,864
BatchNorm2d-41	[-1, 64, 56, 56]	128
Conv2d-42	[-1, 64, 56, 56]	36,864
BatchNorm2d-43	[-1, 64, 56, 56]	128
BasicBlock-44	[-1, 64, 56, 56]	0
Conv2d-45	[-1, 64, 56, 56]	36,864
BatchNorm2d-46	[-1, 64, 56, 56]	128
Conv2d-47	[-1, 64, 56, 56]	36,864
BatchNorm2d-48	[-1, 64, 56, 56]	128
BasicBlock-49	[-1, 64, 56, 56]	0
Linear-50	[-1, 100]	6,500
=====		

Total params: 275,572

Trainable params: 275,572

Non-trainable params: 0

Non trainable params: 0

-----  
Input size (MB): 0.57

Forward/backward pass size (MB): 177.63

Params size (MB): 1.05

Estimated Total Size (MB): 179.25  
-----

In [ ]: