

## Q1 Augmentation Implementation

2 Points

Implement augmentation by finishing train\_tfm in the code with image size of your choice.  
Copy your train\_tfm code here, for example :

```
train_tfm = transforms.Compose([
    # Resize the image into a fixed shape (height = width = 128)
    transforms.Resize((128, 128)),
    # You need to add some transforms here.

    transforms.ToTensor(),
])
```

```
train_tfm = transforms.Compose([
    #new type
    transforms.RandomResizedCrop((128,128)),
    transforms.RandomHorizontalFlip(),
    transforms.GaussianBlur(7,3),
    transforms.ToTensor(),
    transforms.Normalize([0.485, 0.456, 0.406],[0.229, 0.224, 0.225])
])
```

## Q2 Residual\_Model Implementation

2 Points

Implement Residual Connections in the Residual\_Model, following the graph in the slides.  
Copy your Residual\_Model code and paste it here, for example:

```
from torch import nn
class Residual_Network(nn.Module):
    def __init__(self):
        super(Residual_Network, self).__init__()

        self.cnn_layer1 = nn.Sequential(
            nn.Conv2d(3, 64, 3, 1, 1),
            nn.BatchNorm2d(64),
        )

        self.cnn_layer2 = nn.Sequential(
            nn.Conv2d(64, 64, 3, 1, 1),
            nn.BatchNorm2d(64),
        )

        self.cnn_layer3 = nn.Sequential(
            nn.Conv2d(64, 128, 3, 2, 1),
            nn.BatchNorm2d(128),
        )

        self.cnn_layer4 = nn.Sequential(
            nn.Conv2d(128, 128, 3, 1, 1),
            nn.BatchNorm2d(128),
        )

        self.cnn_layer5 = nn.Sequential(
            nn.Conv2d(128, 256, 3, 2, 1),
            nn.BatchNorm2d(256),
```

```

    )
    self.cnn_layer6 = nn.Sequential(
        nn.Conv2d(256, 256, 3, 1, 1),
        nn.BatchNorm2d(256),
    )
    self.fc_layer = nn.Sequential(
        nn.Linear(256* 32* 32, 256),
        nn.ReLU(),
        nn.Linear(256, 11)
    )
    self.relu = nn.ReLU()

def forward(self, x):
    # input (x): [batch_size, 3, 128, 128]
    # output: [batch_size, 11]

    # Extract features by convolutional layers.
    x1 = self.cnn_layer1(x)

    x1 = self.relu(x1)

    x2 = self.cnn_layer2(x1)

    x2 = self.relu(x2)

    x3 = self.cnn_layer3(x2)

    x3 = self.relu(x3)

    x4 = self.cnn_layer4(x3)

    x4 = self.relu(x4)

    x5 = self.cnn_layer5(x4)

    x5 = self.relu(x5)

    x6 = self.cnn_layer6(x5)

    x6 = self.relu(x6)

    xout = x6.flatten(1)

    xout = self.fc_layer(xout)
    return xout

```

```

class Classifier(nn.Module):
    def __init__(self):
        super(Classifier, self).__init__()
        # torch.nn.Conv2d(in_channels, out_channels, kernel_size, stride, padding)
        # torch.nn.MaxPool2d(kernel_size, stride, padding)
        # input 維度 [3, 128, 128]
        dRate = 0.25
        self.cnn = nn.Sequential(
            nn.Conv2d(3, 64, 3, 1, 1), # [64, 128, 128]
            nn.BatchNorm2d(64),
            nn.ReLU(),
            nn.Dropout(dRate),
            nn.MaxPool2d(2, 2, 0), # [64, 64, 64]

            nn.Conv2d(64, 128, 3, 1, 1), # [128, 64, 64]
            nn.BatchNorm2d(128),
            nn.ReLU(),

```

```

nn.Dropout(dRate),
nn.MaxPool2d(2, 2, 0), # [128, 32, 32]

nn.Conv2d(128, 256, 3, 1, 1), # [256, 32, 32]
nn.BatchNorm2d(256),
nn.ReLU(),
nn.Dropout(dRate),
nn.MaxPool2d(2, 2, 0), # [256, 16, 16]

nn.Conv2d(256, 512, 3, 1, 1), # [512, 16, 16]
nn.BatchNorm2d(512),
nn.ReLU(),
nn.Dropout(dRate),
nn.MaxPool2d(2, 2, 0), # [512, 8, 8]

nn.Conv2d(512, 512, 3, 1, 1), # [512, 8, 8]
nn.BatchNorm2d(512),
nn.ReLU(),
nn.Dropout(dRate),
nn.MaxPool2d(2, 2, 0), # [512, 4, 4]
)
self.fc = nn.Sequential(
    nn.Linear(512*4*4, 1024),
    nn.ReLU(),
    nn.Linear(1024, 512),
    nn.ReLU(),
    nn.Linear(512, 11)
)

def forward(self, x):
    out = self.cnn(x)
    origin = nn.Sequential(
        nn.Conv2d(3,512,3,1,1),
        nn.MaxPool2d(32, 32, 0))(x)
    out = out+origin
    out = out.view(out.size()[0], -1)
    return self.fc(out)

```

## HW3

● GRADED

STUDENT

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TOTAL POINTS

**3 / 4 pts**

QUESTION 1

Augmentation Implementation

2 / 2 pts

QUESTION 2

Residual\_Model Implementation

1 / 2 pts