### Convolutional neural networks

#### **Practice Session 9**

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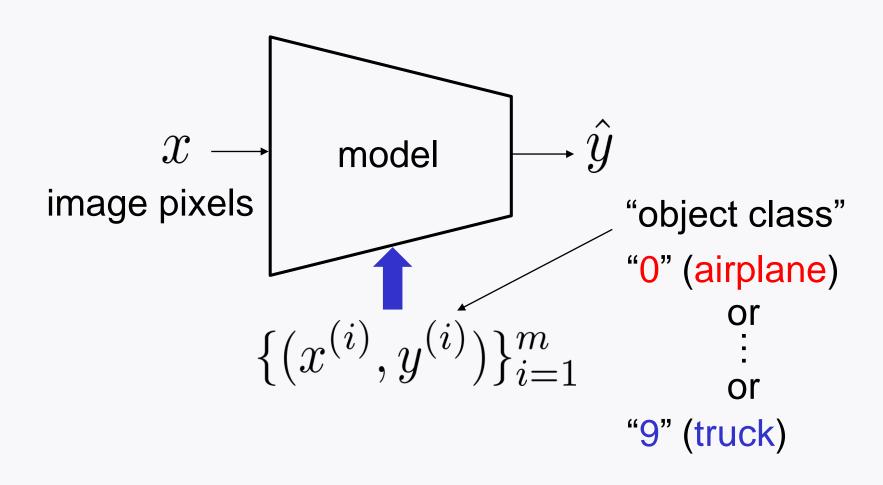
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### **Outline**

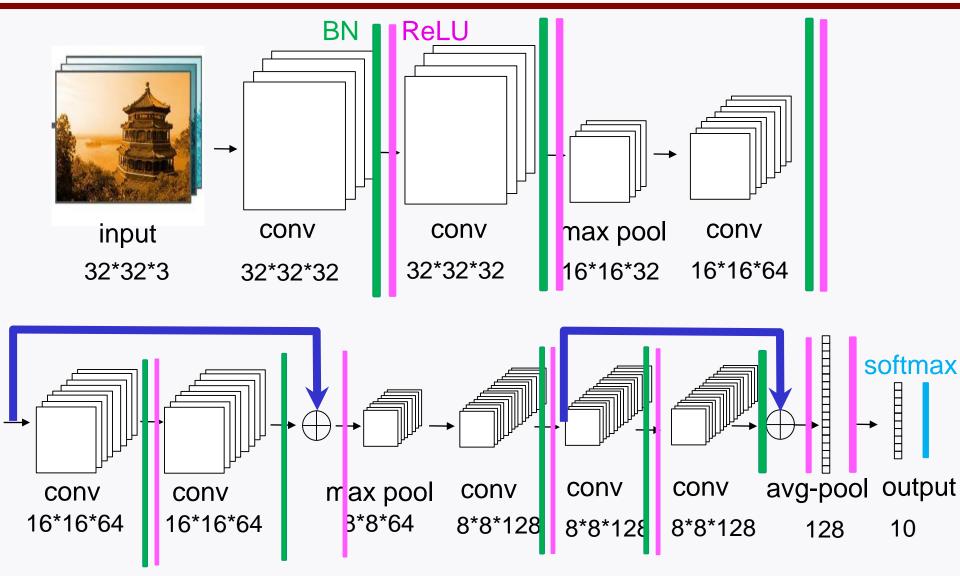
Will implement the simplified ResNet.

Task: Image recognition (CIFAR10)

## Image recognition



## Simplified ResNet



### Two ways to construct a model

### 1. Sequential:

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense
Model_NN = Sequential()
Model_NN.add(Flatten(input_shape=(28,28)))
Model_NN.add(Dense(128, activation='relu'))
Model_NN.add(Dense(10, activation='softmax'))
```

## Two ways to construct a model

#### 2. Model:

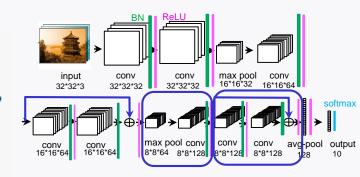
```
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Input
inputs = Input(shape=(28,28))
flatten = Flatten()(inputs)
hidden = Dense(128, activation='relu')(flatten)
outputs = Dense(10, activation='softmax')(hidden)
model_NN = Model(inputs = inputs, outputs = outputs)
```

```
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input
from tensorflow.keras.layers import BatchNormalization
from tensorflow.keras.layers import ReLU
from tensorflow.keras.layers import Conv2D, MaxPool2D
from tensorflow.keras.layers import Add, AveragePooling2D
inputs = Input(shape=(32,32,3))
x = Conv2D(32, kernel\_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(inputs)
x = BatchNormalization()(x)
x = ReLU()(x)
x = Conv2D(32, kernel\_size=(3,3), strides=(1,1),
            padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = ReLU()(x)
```

```
x = MaxPool2D(pool_size=(2,2), strides=(2,2),
               padding='valid')(x)
x = Conv2D(64, kernel_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = ReLU()(x)
# 1st skip connection
skip = x
x = Conv2D(64, kernel\_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = ReLU()(x)
x = Conv2D(64, kernel\_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = Add()([x, skip])
x = ReLU()(x)
```

```
input conv 32*32*3 32*32*32 max pool conv 16*16*32 16*16*64 softmax pool conv 16*16*64 le*8*8*128 8*8*128 8*8*128 10
```

```
x = MaxPool2D(pool_size=(2,2), strides=(2,2),
              padding='valid')(x)
x = Conv2D(128, kernel_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = ReLU()(x)
# 2nd skip connection
skip = x
x = Conv2D(128, kernel\_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = ReLU()(x)
x = Conv2D(128, kernel\_size=(3,3), strides=(1,1),
           padding='same', use_bias=False)(x)
x = BatchNormalization()(x)
x = Add()([x, skip])
x = ReLU()(x)
```



```
# Average Pooling
x = AveragePooling2D(pool_size=(8,8))(x)
x = Flatten()(x)
x = ReLU()(x)
outputs = Dense(10, activation='softmax')(x)
```

```
input 32*32*3 32*32 32*32 max pool conv 16*16*32 16*16*64 softmax pool conv 16*16*64 8*8*128 8*8*128 8*8*128 10
```

```
Model_resnet = Model(inputs = inputs, outputs = outputs)
```

## **Compile**

# **Training & evaluation**

```
# training
Model_resnet.fit(X_train,y_train,epochs=20)

# Evaluation
test_performance = Model_resnet.evaluate(X_test,y_test)
print(test_performance)
```

[0.7550032138824463, 0.8246999979019165]