#### Convolutional neural networks

#### **Practice Session 8**

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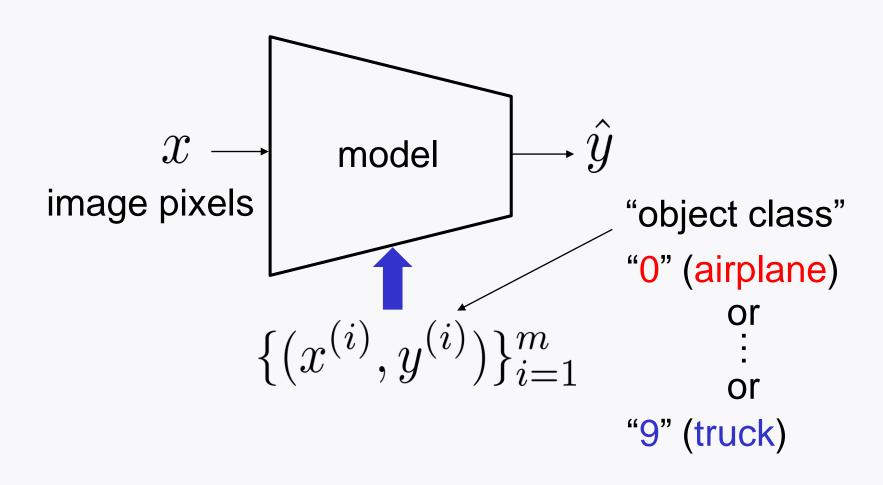
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### **Outline**

Will implement the simplified AlexNet.

Task: Image recognition (CIFAR10)

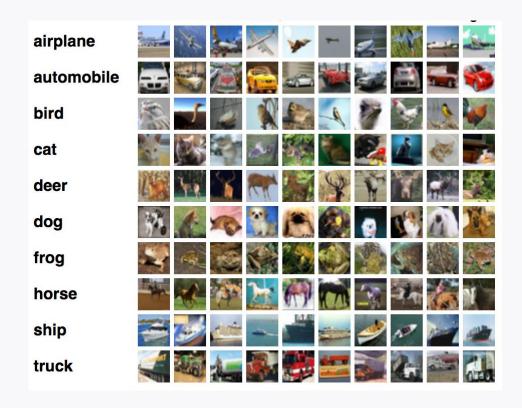
## Image recognition



#### CIFAR10

#### Canadian Institute For Advanced Research

10 classes:



$$m = 50,000$$
  $m_{\text{test}} = 10,000$ 

#### CIFAR10

**Data:** RGB scale image pixels  $\in \mathbb{R}^{32 \times 32 \times 3}$ 

**Label:** human-annotated labels  $\in \{0, 1, \dots, 9\}$  (10 classes)

Example:

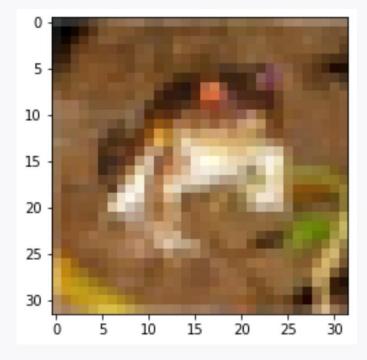


# **Loading CIFAR10**

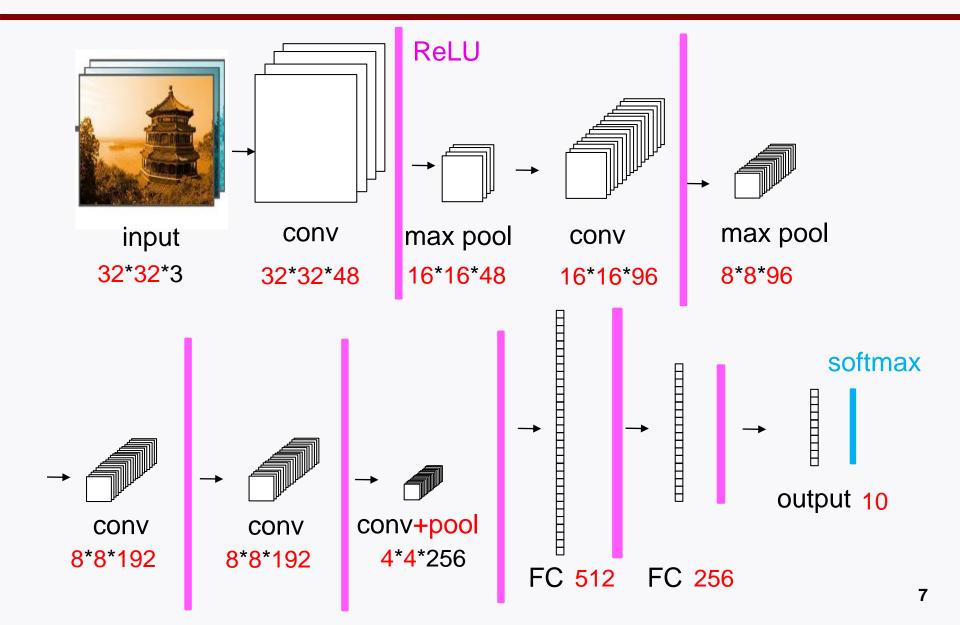
```
from tensorflow.keras.datasets import cifar10
(X_train, y_train), (X_test, y_test) = cifar10.load_data()
print(X_train.shape)
                                (50000, 32, 32, 3)
print(X_test.shape)
                                (10000, 32, 32, 3)
print(y_train.shape)
                                (50000, 1)
                                (10000, 1)
print(y_test.shape)
# normalization
X_{train}, X_{test} = X_{train}/255.0, X_{test}/255.0
```

### **Data visualization**

```
import matplotlib.pyplot as plt
plt.figure()
plt.imshow(X_train[0])
plt.show()
```



### Simplified AlexNet

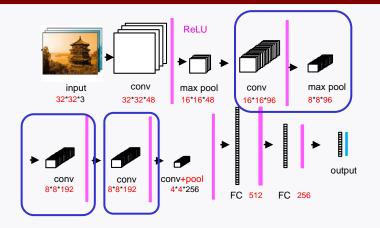


# Simplified AlexNet: Tensorflow coding

```
ReLU
from keras.models import Sequential
from keras.layers import Flatten, Dense
                                                         max pool
                                                     32*32*48
                                                         16*16*48
from keras.layers import Conv2D
from keras.layers import MaxPool2D
model_alexnet = Sequential()
#1st stack: [conv]+[ReLU]+[pool]
model_alexnet.add(Conv2D(input_shape = (32,32,3),
                          kernel_size = (3,3),
                          strides=(1,1),
                          filters=48.
                          padding='same',
                          activation='relu'))
model_alexnet.add(MaxPool2D(pool_size = (2,2),
                              strides = (2,2).
                              padding = 'valid'))
```

# Simplified AlexNet: Tensorflow coding

```
#2nd stack: [conv]+[ReLU]+[pool]
model_alexnet.add(Conv2D(kernel_size = (3,3),
                         strides=(1,1).
                         filters=96.
                         padding='same',
                         activation='relu'))
model_alexnet.add(MaxPool2D(pool_size = (2,2),
                            strides = (2.2).
                            padding = 'valid'))
# 3rd stack: [conv]+[ReLU]
model_alexnet.add(Conv2D(kernel_size = (3,3),
                         strides=(1,1),
                         filters=192.
                         padding='same',
                         activation='relu'))
# 4th stack: [conv]+[ReLU]
model_alexnet.add(Conv2D(kernel_size = (3,3),
                         strides=(1,1).
                          filters=192.
                         padding='same',
                         activation='relu'))
```



# Simplified AlexNet: Tensorflow coding

```
# 5th stack: [conv]+[ReLU]+[pool]
model_alexnet.add(Conv2D(kernel_size = (3,3),
                         strides=(1,1),
                         filters=256.
                         padding='same',
                         activation='relu'))
model_alexnet.add(MaxPool2D(pool_size = (2,2),
                            strides = (2,2).
                            padding = 'valid'))
# 6th layer: 512 Fully connected
model_alexnet.add(Flatten())
model_alexnet.add(Dense(512,activation='relu'))
# 7th layer: 256 fully connected
model_alexnet.add(Dense(256,activation='relu'))
#8th layer: 10 output
model_alexnet.add(Dense(10,activation='softmax'))
```

```
input conv max pool conv max pool 8*8*96

conv conv conv 8*8*192

conv 8*8*192

ReLU

max pool conv max pool 8*8*96

conv pool 4*4*256

FC 512 FC 256
```

## **Compile**

## **Training & evaluation**

```
# Training
model_alexnet.fit(X_train,y_train,epochs=20)

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

[1.6277161836624146, 0.7396000027656555]
```

#### Look ahead

Will implement simplified ResNet in the context of image recognition (CIFAR10).