Practical 11

```
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
import matplotlib.pyplot as plt
# Load and preprocess the CIFAR10 dataset
(train images, train labels), (test images, test labels) = datasets.cifar10.load data()
train images, test images = train images / 255.0, test images / 255.0
# Define the CNN model
model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10))
# Compile the model
model.compile(optimizer='adam',
              {\tt loss=tf.keras.losses.SparseCategoricalCrossentropy(from\_logits=True)}\ ,
              metrics=['accuracy'])
# Train the model
history = model.fit(train images, train labels, epochs=10,
                    validation_data=(test_images, test_labels))
# Evaluate the model
test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)
print(f"Test accuracy: {test_acc}")
# Plot training history (optional)
plt.plot(history.history['accuracy'], label='accuracy')
plt.plot(history.history['val_accuracy'], label = 'val_accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim([0.5, 1])
plt.legend(loc='lower right')
plt.show()
```

```
Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
170498071/170498071
                                          4s Ous/step
/usr/local/lib/python3.10/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_
 super().__init__(activity_regularizer=activity_regularizer, **kwargs)
Epoch 1/10
                              - 80s 50ms/step - accuracy: 0.3526 - loss: 1.7454 - val_accuracy: 0.5313 - val_loss: 1.2987
1563/1563
Epoch 2/10
1563/1563
                               77s 46ms/step - accuracy: 0.5758 - loss: 1.1994 - val_accuracy: 0.6293 - val_loss: 1.0616
Epoch 3/10
1563/1563 -
                               83s 47ms/step - accuracy: 0.6423 - loss: 1.0168 - val_accuracy: 0.6550 - val_loss: 0.9724
Epoch 4/10
1563/1563
                               80s 46ms/step - accuracy: 0.6752 - loss: 0.9240 - val_accuracy: 0.6784 - val_loss: 0.9329
Epoch 5/10
                               71s 46ms/step - accuracy: 0.7077 - loss: 0.8337 - val_accuracy: 0.7002 - val_loss: 0.8751
1563/1563
Epoch 6/10
1563/1563
                                73s 47ms/step - accuracy: 0.7294 - loss: 0.7701 - val_accuracy: 0.6954 - val_loss: 0.9079
Epoch 7/10
1563/1563
                                84s 48ms/step - accuracy: 0.7483 - loss: 0.7219 - val_accuracy: 0.7021 - val_loss: 0.8801
1563/1563
                                80s 47ms/step - accuracy: 0.7676 - loss: 0.6647 - val_accuracy: 0.6877 - val_loss: 0.9188
Epoch 9/10
                               · 76s 49ms/step - accuracy: 0.7811 - loss: 0.6213 - val_accuracy: 0.7107 - val_loss: 0.8640
1563/1563
Epoch 10/10
1563/1563 — 72s 46ms/step - accuracy: 0.7952 - loss: 0.5811 - val_accuracy: 0.7145 - val_loss: 0.8668
313/313 - 4s - 11ms/step - accuracy: 0.7145 - loss: 0.8668
Test accuracy: 0.7145000100135803
   0.9
   0.8
 Accuracy
    0.7
   0.6
                                                             accuracy
                                                             val_accuracy
    0.5
                                                   6
                                                                8
                                       Epoch
```

Object detection using CNN

```
# Load and preprocess an image (replace with your own image)
img_path = '/content/download.jpg'  # Replace with the path to your image
img = cv2.imread(img_path)
img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
img = cv2.resize(img, (32, 32))  # Resize to CIFAR-10 input size
img = img / 255.0  # Normalize
```

Make a prediction

```
img_array = np.expand_dims(img, axis=0)
predictions = model.predict(img_array)
predicted_class = np.argmax(predictions[0])
```

Display the result

print(f"Predicted class: {class_names[predicted_class]}")

