```
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.datasets import cifar10
from tensorflow.keras.applications import VGG16
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import SGD
from tensorflow.keras.utils import to_categorical
# Load and preprocess the CIFAR-10 dataset
(X_train, y_train), (X_test, y_test) = cifar10.load_data()
X_train = X_train / 255.0
X_{\text{test}} = X_{\text{test}} / 255.0
y_train = to_categorical(y_train, num_classes=10)
y_test = to_categorical(y_test, num_classes=10)
# Load a pre-trained model (VGG16 in this example)
base_model = VGG16(weights='imagenet', include_top=False, input_shape=(32, 32, 3))
# Freeze the pre-trained model layers
for layer in base model.layers:
  layer.trainable = False
# Add custom classification layers on top of the pre-trained model
x = Flatten()(base_model.output)
x = Dense(256, activation='relu')(x)
x = Dense(128, activation='relu')(x)
output = Dense(10, activation='softmax')(x)
# Create the final model
model = Model(inputs=base_model.input, outputs=output)
# Compile the model
model.compile(optimizer=SGD(learning rate=0.001, momentum=0.9), loss='categorical crossentropy', metrics=['accuracy'])
# Train the model
model.fit(X_train, y_train, batch_size=32, epochs=10, validation_data=(X_test, y_test))
⇒ Epoch 1/10
     1563/1563
                                   - 793s 507ms/step - accuracy: 0.3306 - loss: 1.9110 - val_accuracy: 0.4659 - val_loss: 1.4934
     Epoch 2/10
                                  - 777s 491ms/step - accuracy: 0.5031 - loss: 1.4125 - val accuracy: 0.5211 - val loss: 1.3653
     1563/1563
     Epoch 3/10
     1563/1563
                                  – 822s 504ms/step - accuracy: 0.5366 - loss: 1.3269 - val_accuracy: 0.5367 - val_loss: 1.3182
     Epoch 4/10
     1563/1563
                                  - 802s 504ms/step - accuracy: 0.5514 - loss: 1.2837 - val_accuracy: 0.5591 - val_loss: 1.2663
     Epoch 5/10
     1563/1563
                                  - 797s 500ms/step - accuracy: 0.5649 - loss: 1.2406 - val_accuracy: 0.5511 - val_loss: 1.2794
     Epoch 6/10
     1563/1563
                                  – 765s 489ms/step - accuracy: 0.5718 - loss: 1.2217 - val_accuracy: 0.5587 - val_loss: 1.2577
     Epoch 7/10
     1563/1563
                                  - 765s 490ms/step - accuracy: 0.5813 - loss: 1.1975 - val accuracy: 0.5784 - val loss: 1.2089
     Epoch 8/10
     1563/1563
                                   - 767s 491ms/step - accuracy: 0.5881 - loss: 1.1750 - val_accuracy: 0.5731 - val_loss: 1.2154
     Epoch 9/10
     1563/1563
                                   - 808s 494ms/step - accuracy: 0.5925 - loss: 1.1640 - val_accuracy: 0.5782 - val_loss: 1.1967
     Epoch 10/10
                                   - 821s 506ms/step - accuracy: 0.5948 - loss: 1.1520 - val_accuracy: 0.5810 - val_loss: 1.1976
     1563/1563
     <keras.src.callbacks.history.History at 0x7fc4da507400>
# Evaluate the model
loss, accuracy = model.evaluate(X_test, y_test)
print("Test Loss:", loss)
print("Test Accuracy:", accuracy)
→ 313/313 ·
                                 - 129s 411ms/step - accuracy: 0.5847 - loss: 1.1934
     Test Loss: 1.1976414918899536
     Test Accuracy: 0.5809999704360962
import numpy as np
# Make Predictions on New Data
# Predict class for a single test image
img = X_test[1] # Take one image from the test set
img = np.expand_dims(img, axis=0) # Expand dimensions to fit the model input shape
```

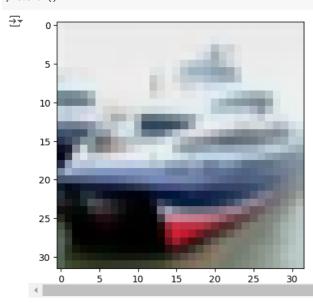
```
# Get prediction
pred = model.predict(img)
predicted_class = np.argmax(pred, axis=1)
```

```
→ 1/1 — 0s 295ms/step
```

```
# Map predicted class index to the class label
class_labels = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']
print(f"Predicted class: {class_labels[predicted_class[0]]}")
```

→ Predicted class: ship

```
import matplotlib.pyplot as plt
img = X_test[1]
#img = img.squeeze(axis=0)
plt.imshow(img)
plt.show()
```



Start coding or generate with AI.