

```
In [1]: import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
```

```
In [2]: import numpy as np
```

```
In [3]: data = np.load('cifar-10-python.npz')
```

```
In [4]: filenames = ["x_train","y_train","x_test","y_test"]
nps = []
for filename in filenames:
    nps.append(data[filename])
train_images,train_labels,test_images,test_labels = nps
```

Shuffling, resizing images

```
In [5]: from tensorflow.keras.utils import to_categorical
train_images, test_images = train_images / 255.0, test_images / 255.0
y_train_cat = to_categorical(train_labels,10)
y_test_cat = to_categorical(test_labels,10)
```

```
In [6]: train_images
```

```
Out[6]: array([[[[0.23137255, 0.24313725, 0.24705882],  
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```

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```

In [7]: `import matplotlib.pyplot as plt`

In [8]: `class_names = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'truck', 'ship']
plt.figure(figsize=(10,10))
for i in range(10):
 plt.subplot(5,5,i+1)
 plt.xticks([])
 plt.yticks([])
 # plt.grid(False)
 plt.imshow(train_images[i])
 plt.xlabel(class_names[train_labels[i]])
plt.show()`



```
In [9]: from tensorflow.keras.layers import BatchNormalization, Dropout
```

```
In [10]: model = models.Sequential()

# Conv block 1
model.add(layers.Conv2D(32, (3, 3), padding='same', activation='relu', input_shape=)
model.add(BatchNormalization())
model.add(layers.Conv2D(32, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))
model.add(Dropout(0.25))

# Conv block 2
model.add(layers.Conv2D(64, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(layers.Conv2D(64, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))
model.add(Dropout(0.25))

# Conv block 3
model.add(layers.Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(layers.Conv2D(128, (3, 3), padding='same', activation='relu'))
model.add(BatchNormalization())
model.add(layers.MaxPooling2D((2, 2)))
model.add(Dropout(0.25))

# Fully connected Layers
model.add(layers.Flatten())
model.add(layers.Dense(512, activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.5))
model.add(layers.Dense(10, activation='softmax'))
```

```
C:\Users\SMIT DESHMUKH\AppData\Local\Programs\Python\Python313\Lib\site-packages\keras\src\layers\convolutional\base_conv.py:113: UserWarning: Do not pass an `input_shape` / `input_dim` argument to a layer. When using Sequential models, prefer using an `Input(shape)` object as the first layer in the model instead.
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
In [11]: model.compile(optimizer='adam',
                     loss='categorical_crossentropy',
                     metrics=['accuracy'])
```

```
In [12]: epochs=15
history=model.fit(train_images, y_train_cat,
                  validation_data=(test_images,y_test_cat),epochs=epochs)
```

```
Epoch 1/15
1563/1563 258s 159ms/step - accuracy: 0.4881 - loss: 1.5081 - val_accuracy: 0.6154 - val_loss: 1.1117
Epoch 2/15
1563/1563 236s 151ms/step - accuracy: 0.6552 - loss: 0.9796 - val_accuracy: 0.6983 - val_loss: 0.8738
Epoch 3/15
1563/1563 242s 155ms/step - accuracy: 0.7089 - loss: 0.8386 - val_accuracy: 0.7030 - val_loss: 0.8835
Epoch 4/15
1563/1563 235s 150ms/step - accuracy: 0.7349 - loss: 0.7617 - val_accuracy: 0.7005 - val_loss: 0.8646
Epoch 5/15
1563/1563 249s 159ms/step - accuracy: 0.7664 - loss: 0.6788 - val_accuracy: 0.7663 - val_loss: 0.6815
Epoch 6/15
1563/1563 253s 154ms/step - accuracy: 0.7868 - loss: 0.6197 - val_accuracy: 0.7983 - val_loss: 0.5865
Epoch 7/15
1563/1563 242s 155ms/step - accuracy: 0.8004 - loss: 0.5841 - val_accuracy: 0.7814 - val_loss: 0.6456
Epoch 8/15
1563/1563 273s 162ms/step - accuracy: 0.8136 - loss: 0.5377 - val_accuracy: 0.8005 - val_loss: 0.5910
Epoch 9/15
1563/1563 258s 159ms/step - accuracy: 0.8269 - loss: 0.5024 - val_accuracy: 0.8236 - val_loss: 0.5349
Epoch 10/15
1563/1563 250s 160ms/step - accuracy: 0.8376 - loss: 0.4709 - val_accuracy: 0.8243 - val_loss: 0.5203
Epoch 11/15
1563/1563 258s 157ms/step - accuracy: 0.8473 - loss: 0.4389 - val_accuracy: 0.8347 - val_loss: 0.4918
Epoch 12/15
1563/1563 267s 160ms/step - accuracy: 0.8546 - loss: 0.4170 - val_accuracy: 0.8323 - val_loss: 0.5073
Epoch 13/15
1563/1563 250s 160ms/step - accuracy: 0.8638 - loss: 0.3898 - val_accuracy: 0.8378 - val_loss: 0.4838
Epoch 14/15
1563/1563 259s 158ms/step - accuracy: 0.8693 - loss: 0.3757 - val_accuracy: 0.8332 - val_loss: 0.5053
Epoch 15/15
1563/1563 252s 161ms/step - accuracy: 0.8759 - loss: 0.3519 - val_accuracy: 0.8536 - val_loss: 0.4580
```

```
In [13]: test_loss,test_acc = model.evaluate(test_images,y_test_cat)
print("loss %.3f"%test_loss)
print("acc %.3f"%test_acc)
```

```
313/313 14s 46ms/step - accuracy: 0.8536 - loss: 0.4580
loss 0.458
acc 0.854
```

```
In [14]: predicted_values = model.predict(test_images)
```

```
313/313 14s 43ms/step
```

```
In [15]: predicted_values.shape
```

```
Out[15]: (10000, 10)
```

```
In [16]: import random
n = random.randint(0,9999)
plt.figure(figsize=(10,10))
plt.imshow(test_images[n])
plt.yticks([])
plt.xticks([])
plt.grid(False)
plt.title(class_names[np.argmax(predicted_values[n])])
```

```
Out[16]: Text(0.5, 1.0, 'dog')
```

dog



```
In [17]: plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
```

```
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```

