

Carga de datos

```
In [64]: # Load CIFAR-10 data set
from keras.datasets import cifar10
(X_train, y_train), (X_test, y_test) = cifar10.load_data()

In [65]: # Show examples from each class
import numpy as np
import matplotlib.pyplot as plt

num_classes = len(np.unique(y_train))
class_names = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']
fig = plt.figure(figsize=(8,3))
fig = plt.figure(figsize=(8,3))
for i in range(num_classes):
    ax = fig.add_subplot(2, 5, 1 + i, ticks=[])
    ax.set_title(class_names[i])
    idx = np.where(y_train[i]==1)[0]
    features_idx = X_train[idx,:]
    rnd_img = np.random.randint(features_idx.shape[0])
    im = np.transpose(features_idx[rnd_img,:], (0, 1, 2))
    plt.imshow(im)
plt.show()

airplane      automobile      bird      cat      deer
dog      frog      horse      ship      truck
```

```
In [66]: # Data pre-processing
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
X_train /= 255.0
X_test /= 255.0

Y_train = np_utils.to_categorical(y_train, num_classes)
Y_test = np_utils.to_categorical(y_test, num_classes)
```

Utilidades

En esta sección se ejecutan todas las funciones que contienen las utilidades que se emplearán en la fase de experimentación

Dibujo de gráficas

La primera función sirve para dibujar las gráficas con el historial de entrenamiento

```
In [67]: def fig_model_history(model, history):
fig, axes = plt.subplots(1,2,figsize=(15,5))
# Summarize history for accuracy
axes[0].plot(range(1,len(model.history.history['accuracy']))+1,model.history.history['accuracy'])
axes[0].plot(range(1,len(model.history.history['val_accuracy']))+1,model.history.history['val_accuracy'])
axes[0].set_title('Model Accuracy')
axes[0].set_ylabel('Accuracy')
axes[0].set_xlabel('Epoch')
axes[0].set_xticks(np.arange(1,len(model.history.history['accuracy']))+1,step=len(model.history.history['accuracy']/10))
axes[0].legend(['train', 'val'], loc='best')
# Summarize history for loss
axes[1].plot(range(1,len(model.history.history['loss']))+1,model.history.history['loss'])
axes[1].plot(range(1,len(model.history.history['val_loss']))+1,model.history.history['val_loss'])
axes[1].set_title('Model Loss')
axes[1].set_ylabel('Loss')
axes[1].set_xlabel('Epoch')
axes[1].set_xticks(np.arange(1,len(model.history.history['loss']))+1,step=len(model.history.history['loss']/10))
axes[1].legend(['train', 'val'], loc='best')
plt.show()
```

Creación de modelos

A continuación se definen las funciones necesarias para la creación de los modelos que se utilizarán en la experimentación.

```
In [70]: from tensorflow.keras import Sequential
from tensorflow.keras.layers import Dense, Dropout, Activation, Flatten, BatchNormalization, Input, Conv2D, MaxPooling2D
from tensorflow.keras.layers import experimental

def build_model_v1(dout=True, bn=True):
model = Sequential([experimental.preprocessing.Rescaling(1./255, input_shape=(32, 32, 3)),
experimental.preprocessing.RandomRotation(0.2),
experimental.preprocessing.RandomZoom(0.2),
experimental.preprocessing.RandomFlip("horizontal")])

model.add(Flatten())
model.add(Dense(128))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(128))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(10))
model.add(Activation('softmax'))
return model

def build_model_v2(dout=True, bn=True):
model = Sequential([experimental.preprocessing.Rescaling(1./255, input_shape=(32, 32, 3)),
experimental.preprocessing.RandomRotation(0.2),
experimental.preprocessing.RandomZoom(0.2),
experimental.preprocessing.RandomFlip("horizontal")])

model.add(Flatten())
model.add(Dense(128))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(64))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(32))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(10))
model.add(Activation('softmax'))
return model

def build_model_v3(dout=True, bn=True):
model = Sequential([experimental.preprocessing.Rescaling(1./255, input_shape=(32, 32, 3)),
experimental.preprocessing.RandomRotation(0.2),
experimental.preprocessing.RandomZoom(0.2),
experimental.preprocessing.RandomFlip("horizontal")])

model.add(Flatten())
model.add(Dense(1024))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(64))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(32))
if bn:
model.add(BatchNormalization())
if dout:
model.add(Dropout(0.3))
model.add(Activation('relu'))
model.add(Dense(10))
model.add(Activation('softmax'))
return model

def create_block(block_input, n_filtros, bn, bn_before):
x = block_input
for i in range(2):
x = Conv2D(n_filtros, kernel_size=(3,3), padding="same")(x)
x = BatchNormalization()(x)
x = Activation('relu')(x)
if bn and not bn_before:
x = BatchNormalization()(x)
return x

def build_conv(dout=True, bn=True, bn_before=True, extra_layer=False): # De https://www.kaggle.com/mah
tabshah/autoencoder-as-feature-extractor-cifar10
autoencoder = Input((32,32,3))
x = net_input
#x = experimental.preprocessing.Rescaling(1./255)(x)
#x = experimental.preprocessing.RandomRotation(0.2)(x)
#x = experimental.preprocessing.RandomZoom(0.2)(x)
#x = experimental.preprocessing.RandomFlip("horizontal")(x)

block1 = create_block(x, 32, bn, bn_before)
x = MaxPooling2D(pool_size=(2, 2))(block1)
if dout:
x = Dropout(0.2)(x)

block2 = create_block(x, 64, bn, bn_before)
x = MaxPooling2D(pool_size=(2, 2))(block2)
if dout:
x = Dropout(0.3)(x)

block3 = create_block(x, 128, bn, bn_before)
x = MaxPooling2D(pool_size=(2, 2))(block3)
if dout:
x = Dropout(0.3)(x)

x = Flatten()(x)
if extra_layer:
x = Dense(2048)(x)
output = Dense(10, activation='softmax')(x)
return Model(net_input, output)

Training MLP took 232.05204224586487 seconds
```

Experimentación

Modelo fnnn 1

Incluyendo capas de dropout y batch normalization

```
In [81]: model = build_model_v1(bn=False)
```

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

```
# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()
```

```
loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)
```

```
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")
```

Model: "sequential_19"

Layer (type)	Output Shape	Param #
rescaling_41 (Rescaling)	(None, 32, 32, 3)	0
random_rotation_41 (RandomRo	(None, 32, 32, 3)	0
random_zoom_41 (RandomZoom)	(None, 32, 32, 3)	0
random_flip_41 (RandomFlip)	(None, 32, 32, 3)	0
flatten_52 (Flatten)	(None, 3072)	0
dense_133 (Dense)	(None, 128)	393344
batch_normalization_210 (Bat	(None, 128)	512
dropout_107 (Dropout)	(None, 128)	0
activation_281 (Activation)	(None, 128)	0
dense_134 (Dense)	(None, 128)	16512
batch_normalization_211 (Bat	(None, 128)	512
dropout_108 (Dropout)	(None, 128)	0
activation_282 (Activation)	(None, 128)	0
dense_135 (Dense)	(None, 10)	1290

activation_283 (Activation) (None, 10) 0

Trainable params: 411,658

Non-trainable params: 512

Epoch 1/50

391/391 [=====] - 7s 12ms/step - loss: 2.1124 - accuracy: 0.2479 - val_loss: 2.1803 - val_accuracy: 0.1545

Epoch 2/50

391/391 [=====] - 4s 11ms/step - loss: 1.8516 - accuracy: 0.3370 - val_loss: 1.7572 - val_accuracy: 0.3831

Epoch 3/50

391/391 [=====] - 4s 11ms/step - loss: 1.7905 - accuracy: 0.3585 - val_loss: 1.7231 - val_accuracy: 0.3890

Epoch 4/50

391/391 [=====] - 4s 11ms/step - loss: 1.7552 - accuracy: 0.3715 - val_loss: 1.7870 - val_accuracy: 0.3521

Epoch 5/50

391/391 [=====] - 4s 11ms/step - loss: 1.7296 - accuracy: 0.3805 - val_loss: 1.6960 - val_accuracy: 0.4189

Epoch 6/50

391/391 [=====] - 4s 11ms/step - loss: 1.7142 - accuracy: 0.3857 - val_loss: 1.6931 - val_accuracy: 0.3944

Epoch 7/50

391/391 [=====] - 4s 11ms/step - loss: 1.6877 - accuracy: 0.3963 - val_loss: 1.5979 - val_accuracy: 0.4074

Epoch 8/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 9/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 10/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 11/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 12/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 13/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 14/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 15/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 16/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 17/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 18/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 19/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 20/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 21/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 22/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 23/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 24/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 25/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 26/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 27/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 28/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 29/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 30/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 31/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 32/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 33/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 34/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 35/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 36/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 37/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 38/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 39/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 40/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 41/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 42/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 43/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 44/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Epoch 45/50

391/391 [=====] - 4s 11ms/step - loss: 1.6628 - accuracy: 0.4064 - val_loss: 1.6491 - val_accuracy: 0.4123

Epoch 46/50

391/391 [=====] - 4s 11ms/step - loss: 1.6640 - accuracy: 0.4064 - val_loss: 1.6258 - val_accuracy: 0.4137

Epoch 47/50

391/391 [=====] - 5s 12ms/step - loss: 1.6543 - accuracy: 0.4132 - val_loss: 1.5921 - val_accuracy: 0.4393

Epoch 48/50

391/391 [=====] - 5s 12ms/step - loss: 1.6494 - accuracy: 0.4085 - val_loss: 1.6541 - val_accuracy: 0.4100

Epoch 49/50

391/391 [=====] - 5s 12ms/step - loss: 1.6427 - accuracy: 0.4140 - val_loss: 1.6791 - val_accuracy: 0.4074

Epoch 50/50

391/391 [=====] - 4s 11ms/step - loss: 1.6824 - accuracy: 0.3973 - val_loss: 1.6791 - val_accuracy: 0.3814

Test loss: 1.590074062347412

Test accuracy: 0.4352999256134033

Training MLP took 213.21921491622925 seconds

Con capa de batch normalization pero sin dropout

```
In [82]: model = build_model_v1(bn=False)
```

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

```
# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()
```

```
loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)
```

```
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")
```

Model: "sequential_20"

Layer (type)	Output Shape	Param #
rescaling_42 (Rescaling)	(None, 32, 32, 3)	0
random_rotation_42 (RandomRo	(None, 32, 32, 3)	0
random_zoom_42 (RandomZoom)	(None, 32, 32, 3)	0
random_flip_42 (RandomFlip)	(None, 32, 32, 3)	0
flatten_53 (Flatten)	(None, 3072)	0
dense_136 (Dense)	(None, 128)	393344
dropout_109 (Dropout)	(None, 128)	0
activation_284 (Activation)	(None, 128)	0
dense_137 (Dense)	(None, 128)	16512
dropout_110 (Dropout)	(None, 128)	0
activation_285 (Activation)	(None, 128)	0
dense_138 (Dense)	(None, 10)	1290

activation_286 (Activation) (None, 10) 0

Total params: 411,146

Trainable params: 411,146

Non-trainable params: 0

Epoch 1/50

391/391 [=====] - 6s 11ms/step - loss: 2.2251 - accuracy: 0.1540 - val_loss: 1.9946 - val_accuracy: 0.2706

Epoch 2/50

391/391 [=====] - 4s 11ms/step - loss: 2.0302 - accuracy: 0.2526 - val_loss: 1.8931 - val_accuracy: 0.2959

Epoch 3/50

391/391 [=====] - 4s 11ms/step - loss: 1.9816 - accuracy: 0.2756 - val_loss: 1.8931 - val_accuracy: 0.3150

Epoch 4/50

391/391 [=====] - 4s 11ms/step - loss: 1.9575 - accuracy: 0.2860 - val_loss: 1.8931 - val_accuracy: 0.3264

Epoch 5/50

391/391 [=====] - 4s 11ms/step - loss: 1.9385 - accuracy: 0.2971 - val_loss: 1.8931 - val_accuracy: 0.3389

Epoch 6/50

In (83):

```
model = build_model_v1(dou=False)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_21"
```

Layer (type)	Output Shape	Param #
rescaling_43 (Rescaling)	(None, 32, 32, 3)	0
random_rotation_43 (RandomRo	(None, 32, 32, 3)	0
random_zoom_43 (RandomZoom)	(None, 32, 32, 3)	0
Random_flip_43 (RandomFlip)	(None, 32, 32, 3)	0
flatten_54 (Flatten)	(None, 3072)	0
dense_139 (Dense)	(None, 128)	393344
batch_normalization_212 (Bat	(None, 128)	512
activation_287 (Activation)	(None, 128)	0
dense_140 (Dense)	(None, 128)	16512
batch_normalization_213 (Bat	(None, 128)	512
activation_288 (Activation)	(None, 128)	0
dense_141 (Dense)	(None, 10)	1290
activation_289 (Activation)	(None, 10)	0
Total params: 412,170		
Trainable params: 411,658		
Non-trainable params: 512		

Epoch 1/50

391/391 [=====] - 6s 12ms/step - loss: 1.9234 - accuracy: 0.3122 - val_loss: 2.1350 - val_accuracy: 0.1626

Epoch 2/50

391/391 [=====] - 4s 11ms/step - loss: 1.7169 - accuracy: 0.3871 - val_loss: 1.7279 - val_accuracy: 0.3872

Epoch 3/50

391/391 [=====] - 4s 11ms/step - loss: 1.6423 - accuracy: 0.4151 - val_loss: 1.7905 - val_accuracy: 0.3611

Epoch 4/50

391/391 [=====] - 4s 10ms/step - loss: 1.6144 - accuracy: 0.4225 - val_loss: 1.8699 - val_accuracy: 0.3954

Epoch 5/50

391/391 [=====] - 4s 11ms/step - loss: 1.5861 - accuracy: 0.4263 - val_loss: 1.7116 - val_accuracy: 0.3863

Epoch 6/50

391/391 [=====] - 4s 11ms/step - loss: 1.5616 - accuracy: 0.4418 - val_loss: 1.7279 - val_accuracy: 0.3939

Epoch 7/50

391/391 [=====] - 4s 11ms/step - loss: 1.5421 - accuracy: 0.4542 - val_loss: 1.8166 - val_accuracy: 0.3763

Epoch 8/50

391/391 [=====] - 4s 11ms/step - loss: 1.5202 - accuracy: 0.4580 - val_loss: 1.6220 - val_accuracy: 0.4110

Epoch 9/50

391/391 [=====] - 4s 10ms/step - loss: 1.5067 - accuracy: 0.4641 - val_loss: 1.6088 - val_accuracy: 0.4293

Epoch 10/50

391/391 [=====] - 4s 10ms/step - loss: 1.4960 - accuracy: 0.4649 - val_loss: 1.6254 - val_accuracy: 0.3513

Epoch 11/50

391/391 [=====] - 4s 11ms/step - loss: 1.4843 - accuracy: 0.4725 - val_loss: 1.6506 - val_accuracy: 0.4207

Epoch 12/50

391/391 [=====] - 4s 11ms/step - loss: 1.4820 - accuracy: 0.4730 - val_loss: 1.6452 - val_accuracy: 0.4148

Epoch 13/50

391/391 [=====] - 5s 12ms/step - loss: 1.4624 - accuracy: 0.4779 - val_loss: 1.5649 - val_accuracy: 0.4462

Epoch 14/50

391/391 [=====] - 4s 11ms/step - loss: 1.4630 - accuracy: 0.4804 - val_loss: 1.5116 - val_accuracy: 0.4492

Epoch 15/50

391/391 [=====] - 4s 11ms/step - loss: 1.4600 - accuracy: 0.4788 - val_loss: 1.5731 - val_accuracy: 0.4538

Epoch 16/50

391/391 [=====] - 4s 11ms/step - loss: 1.4301 - accuracy: 0.4876 - val_loss: 1.6210 - val_accuracy: 0.3612

Epoch 17/50

391/391 [=====] - 4s 11ms/step - loss: 1.4311 - accuracy: 0.4918 - val_loss: 1.4840 - val_accuracy: 0.4725

Epoch 18/50

391/391 [=====] - 4s 11ms/step - loss: 1.4194 - accuracy: 0.4960 - val_loss: 1.6745 - val_accuracy: 0.4079

Epoch 19/50

391/391 [=====] - 4s 11ms/step - loss: 1.4211 - accuracy: 0.4911 - val_loss: 1.5235 - val_accuracy: 0.3761

Epoch 20/50

391/391 [=====] - 4s 11ms/step - loss: 1.4182 - accuracy: 0.4960 - val_loss: 1.6799 - val_accuracy: 0.4024

Epoch 22/50

391/391 [=====] - 4s 11ms/step - loss: 1.4091 - accuracy: 0.4974 - val_loss: 1.4783 - val_accuracy: 0.4656

Epoch 23/50

391/391 [=====] - 4s 11ms/step - loss: 1.4012 - accuracy: 0.5042 - val_loss: 1.4709 - val_accuracy: 0.4719

Epoch 24/50

391/391 [=====] - 4s 11ms/step - loss: 1.4017 - accuracy: 0.5000 - val_loss: 1.5116 - val_accuracy: 0.4746

Epoch 25/50

391/391 [=====] - 4s 11ms/step - loss: 1.3912 - accuracy: 0.5070 - val_loss: 1.5814 - val_accuracy: 0.4652

Epoch 26/50

391/391 [=====] - 4s 11ms/step - loss: 1.3903 - accuracy: 0.5041 - val_loss: 1.4819 - val_accuracy: 0.4785

Epoch 27/50

391/391 [=====] - 5s 12ms/step - loss: 1.3928 - accuracy: 0.5064 - val_loss: 1.5074 - val_accuracy: 0.4688

Epoch 28/50

391/391 [=====] - 5s 12ms/step - loss: 1.3827 - accuracy: 0.5080 - val_loss: 1.4250 - val_accuracy: 0.4909

Epoch 29/50

391/391 [=====] - 4s 11ms/step - loss: 1.3746 - accuracy: 0.5099 - val_loss: 1.4783 - val_accuracy: 0.4677

Epoch 30/50

391/391 [=====] - 4s 11ms/step - loss: 1.3811 - accuracy: 0.5070 - val_loss: 1.5235 - val_accuracy: 0.4523

Epoch 31/50

391/391 [=====] - 4s 11ms/step - loss: 1.3773 - accuracy: 0.5105 - val_loss: 1.5393 - val_accuracy: 0.4584

Epoch 32/50

391/391 [=====] - 4s 11ms/step - loss: 1.3745 - accuracy: 0.5123 - val_loss: 1.4501 - val_accuracy: 0.4611

Epoch 33/50

391/391 [=====] - 4s 10ms/step - loss: 1.3776 - accuracy: 0.5087 - val_loss: 1.4746 - val_accuracy: 0.4746

Epoch 34/50

391/391 [=====] - 4s 10ms/step - loss: 1.3683 - accuracy: 0.5143 - val_loss: 1.4650 - val_accuracy: 0.4794

Epoch 35/50

391/391 [=====] - 4s 10ms/step - loss: 1.3633 - accuracy: 0.5138 - val_loss: 1.4470 - val_accuracy: 0.4866

Epoch 36/50

391/391 [=====] - 4s 10ms/step - loss: 1.3588 - accuracy: 0.5178 - val_loss: 1.4259 - val_accuracy: 0.4914

Epoch 37/50

391/391 [=====] - 4s 10ms/step - loss: 1.3517 - accuracy: 0.5178 - val_loss: 1.4252 - val_accuracy: 0.4913

Epoch 38/50

391/391 [=====] - 4s 10ms/step - loss: 1.3554 - accuracy: 0.5192 - val_loss: 1.4500 - val_accuracy: 0.4809

Epoch 39/50

391/391 [=====] - 4s 10ms/step - loss: 1.3498 - accuracy: 0.5184 - val_loss: 1.4787 - val_accuracy: 0.4900

Epoch 40/50

391/391 [=====] - 4s 10ms/step - loss: 1.3510 - accuracy: 0.5178 - val_loss: 1.4893 - val_accuracy: 0.4940

Epoch 41/50

391/391 [=====] - 4s 10ms/step - loss: 1.3420 - accuracy: 0.5212 - val_loss: 1.4709 - val_accuracy: 0.4540

Epoch 42/50

391/391 [=====] - 4s 10ms/step - loss: 1.3429 - accuracy: 0.5249 - val_loss: 1.4731 - val_accuracy: 0.4761

Epoch 43/50

391/391 [=====] - 4s 10ms/step - loss: 1.3426 - accuracy: 0.5236 - val_loss: 1.4736 - val_accuracy: 0.4939

Epoch 44/50

391/391 [=====] - 4s 10ms/step - loss: 1.3406 - accuracy: 0.5227 - val_loss: 1.4245 - val_accuracy: 0.4929

Epoch 45/50

391/391 [=====] - 4s 10ms/step - loss: 1.3393 - accuracy: 0.5254 - val_loss: 1.4495 - val_accuracy: 0.4908

Epoch 46/50

391/391 [=====] - 4s 10ms/step - loss: 1.3294 - accuracy: 0.5305 - val_loss: 1.4571 - val_accuracy: 0.4873

Epoch 47/50

391/391 [=====] - 4s 10ms/step - loss: 1.3261 - accuracy: 0.5311 - val_loss: 1.4783 - val_accuracy: 0.4931

Epoch 48/50

391/391 [=====] - 4s 11ms/step - loss: 1.3199 - accuracy: 0.5295 - val_loss: 1.4823 - val_accuracy: 0.4945

Epoch 49/50

391/391 [=====] - 4s 11ms/step - loss: 1.3106 - accuracy: 0.5357 - val_loss: 1.4709 - val_accuracy: 0.5058

Test loss: 1.396878957748413

Test accuracy: 0.505800087738037

Model Accuracy

Model Loss

Training MLP took 211.6930109178162 seconds

Modelo ffnn 2

Incluyendo capas de dropout y batch normalization

In (84):

```
model = build_model_v2()
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_22"
```

Layer (type)	Output Shape	Param #
rescaling_44 (Rescaling)	(None, 32, 32, 3)	0
random_rotation_44 (RandomRo	(None, 32, 32, 3)	0
random_zoom_44 (RandomZoom)	(None, 32, 32, 3)	0
random_flip_44 (RandomFlip)	(None, 32, 32, 3)	0
flatten_55 (Flatten)	(None, 3072)	0
dense_142 (Dense)	(None, 128)	393344
batch_normalization_214 (Bat	(None, 128)	512
dropout_111 (Dropout)	(None, 128)	0
activation_290 (Activation)	(None, 128)	0
dense_143 (Dense)	(None, 64)	8256
batch_normalization_215 (Bat	(None, 64)	256
dropout_112 (Dropout)	(None, 64)	0
activation_291 (Activation)	(None, 64)	0
dense_144 (Dense)	(None, 32)	2080
batch_normalization_216 (Bat	(None, 32)	128
dropout_113 (Dropout)	(None, 32)	0
activation_292 (Activation)	(None, 32)	0
dense_145 (Dense)	(None, 16)	528
batch_normalization_217 (Bat	(None, 16)	64
dropout_114 (Dropout)	(None, 16)	0
activation_293 (Activation)	(None, 16)	0
dense_146 (Dense)	(None, 10)	170
activation_294 (Activation)	(None, 10)	0
Total params: 405,338		
Trainable params: 404,858		
Non-trainable params: 480		

Epoch 1/50

391/391 [=====] - 7s 12ms/step - loss: 2.3825 - accuracy: 0.1520 - val_loss: 2.1350 - val_accuracy: 0.1653

Epoch 2/50

391/391 [=====] - 5s 12ms/step - loss: 2.0833 - accuracy: 0.2269 - val_loss: 1.9428 - val_accuracy: 0.3226

Epoch 3/50

391/391 [=====] - 5s 12ms/step - loss: 2.0009 - accuracy: 0.2584 - val_loss: 1.9000 - val_accuracy: 0.3268

Epoch 4/50

391/391 [=====] - 5s 12ms/step - loss: 1.9595 - accuracy: 0.2791 - val_loss: 1.8451 - val_accuracy: 0.3354

Epoch 5/50

391/391 [=====] - 5s 12ms/step - loss: 1.9350 - accuracy: 0.2950 - val_loss: 1.8699 - val_accuracy: 0.3296

Epoch 6/50

391/391 [=====] - 5s 12ms/step - loss: 1.9025 - accuracy: 0.3109 - val_loss: 1.8146 - val_accuracy: 0.3555

Epoch 7/50

391/391 [=====] - 5s 12ms/step - loss: 1.8985 - accuracy: 0.3144 - val_loss: 1.8071 - val_accuracy: 0.3508

Epoch 8/50

391/391 [=====] - 4s 11ms/step - loss: 1.8793 - accuracy: 0.3263 - val_loss: 1.8409 - val_accuracy: 0.3278

Epoch 9/50

391/391 [=====] - 5s 12ms/step - loss: 1.8683 - accuracy: 0.3316 - val_loss: 1.8313 - val_accuracy: 0.3758

Epoch 10/50

391/391 [=====] - 4s 11ms/step - loss: 1.8663 - accuracy: 0.3304 - val_loss: 1.8699 - val_accuracy: 0.3485

Epoch 11/50

391/391 [=====] - 5s 12ms/step - loss: 1.8484 - accuracy: 0.3397 - val_loss: 1.8071 - val_accuracy: 0.3451

Epoch 12/50

391/391 [=====] - 5s 12ms/step - loss: 1.8514 - accuracy: 0.3360 - val_loss: 1.7534 - val_accuracy: 0.3796

Epoch 13/50

391/391 [=====] - 5s 12ms/step - loss: 1.8373 - accuracy: 0.3406 - val_loss: 1.6759 - val_accuracy: 0.4100

Epoch 14/50

391/391 [=====] - 4s 11ms/step - loss: 1.8323 - accuracy: 0.3470 - val_loss: 1.7545 - val_accuracy: 0.3734

Epoch 15/50

391/391 [=====] - 4s 11ms/step - loss: 1.8299 - accuracy: 0.3468 - val_loss: 1.7351 - val_accuracy: 0.3764

Epoch 16/50

391/391 [=====] - 4s 11ms/step - loss: 1.8179 - accuracy: 0.3494 - val_loss: 1.6991 - val_accuracy: 0.3705

Epoch 17/50

391/391 [=====] - 4s 11ms/step - loss: 1.8150 - accuracy: 0.3516 - val_loss: 1.6759 - val_accuracy: 0.3751

Epoch 18/50

391/391 [=====] - 4s 11ms/step - loss: 1.8104 - accuracy: 0.3514 - val_loss: 1.6941 - val_accuracy: 0.3909

Epoch 19/50

391/391 [=====] - 4s 11ms/step - loss: 1.8191 - accuracy: 0.3559 - val_loss: 1.6531 - val_accuracy: 0.4152

Epoch 20/50

391/391 [=====] - 4s 11ms/step - loss: 1.8071 - accuracy: 0.3547 - val_loss: 1.6506 - val_accuracy: 0.4120

Epoch 21/50

391/391 [=====] - 5s 12ms/step - loss: 1.7971 - accuracy: 0.3580 - val_loss: 1.6424 - val_accuracy: 0.4120

Epoch 22/50

391/391 [=====] - 5s 12ms/step - loss: 1.7956 - accuracy: 0.3611 - val_loss: 1.6941 - val_accuracy: 0.4010

Epoch 23/50

391/391 [=====] - 5s 12ms/step - loss: 1.7956 - accuracy: 0.3611 - val_loss: 1.6911 - val_accuracy: 0.3933

Epoch 24/50

391/391 [=====] - 5s 12ms/step - loss: 1.7942 - accuracy: 0.3571 - val_loss: 1.6725 - val_accuracy: 0.4065

Epoch 25/50

391/391 [=====] - 5s 12ms/step - loss: 1.7805 - accuracy: 0.3553 - val_loss: 1.6929 - val_accuracy: 0.4108

Epoch 26/50

391/391 [=====] - 4s 11ms/step - loss: 1.7848 - accuracy: 0.3652 - val_loss: 1.6812 - val_accuracy: 0.3957

Epoch 27/50

391/391 [=====] - 4s 11ms/step - loss: 1.7917 - accuracy: 0.3677 - val_loss: 1.6799 - val_accuracy: 0.4028

Epoch 28/50

391/391 [=====] - 4s 11ms/step - loss: 1.7805 - accuracy: 0.3675 - val_loss: 1.6941 - val_accuracy: 0.4241

Epoch 29/50

391/391 [=====] - 4s 11ms/step - loss: 1.7823 - accuracy: 0.3664 - val_loss: 1.6531 - val_accuracy: 0.4051

Epoch 30/50

391/391 [=====] - 4s 11ms/step - loss: 1.7826 - accuracy: 0.3671 - val_loss: 1.6610 - val_accuracy: 0.4061

Epoch 31/50

391/391 [=====] - 4s 11ms/step - loss: 1.7866 - accuracy: 0.3637 - val_loss: 1.6468 - val_accuracy: 0.4176

Epoch 32/50

391/391 [=====] - 4s 11ms/step - loss: 1.7851 - accuracy: 0.3638 - val_loss: 1.6154 - val_accuracy: 0.4265

Epoch 33/50

391/391 [=====] - 4s 11ms/step - loss: 1.7721 - accuracy: 0.3741 - val_loss: 1.6301 - val_accuracy: 0.4179

Epoch 34/50

391/391 [=====] - 4s 11ms/step - loss: 1.7782 - accuracy: 0.3686 - val_loss: 1.6146 - val_accuracy: 0.4263

Epoch 35/50

391/391 [=====] - 4s 11ms/step - loss: 1.7756 - accuracy: 0.3659 - val_loss: 1.6214 - val_accuracy: 0.4290

Epoch 36/50

391/391 [=====] - 5s 12ms/step - loss: 1.7785 - accuracy: 0.3688 - val_loss: 1.6593 - val_accuracy: 0.4426

Epoch 37/50

391/391 [=====] - 4s 11ms/step - loss: 1.7779 - accuracy: 0.3713 - val_loss: 1.6531 - val_accuracy: 0.3723

Epoch 38/50

391/391 [=====] - 4s 11ms/step - loss: 1.7728 - accuracy: 0.3717 - val_loss: 1.6531 - val_accuracy: 0.4395

Epoch 39/50

391/391 [=====] - 4s 11ms/step - loss: 1.7671 - accuracy: 0.3772 - val_loss: 1.6563 - val_accuracy: 0.4165

Epoch 40/50

391/391 [=====] - 4s 11ms/step - loss: 1.7704 - accuracy: 0.3717 - val_loss: 1.6267 - val_accuracy: 0.4085

Epoch 41/50

391/391 [=====] - 4s 11ms/step - loss: 1.7636 - accuracy: 0.3735 - val_loss: 1.6873 - val_accuracy: 0.3987

Epoch 42/50

391/391 [=====] - 4s 11ms/step - loss: 1.7647 - accuracy: 0.3727 - val_loss: 1.7267 - val_accuracy: 0.3771

Epoch 43/50

391/391 [=====] - 4s 11ms/step - loss: 1.7619 - accuracy: 0.3746 - val_loss: 1.9568 - val_accuracy: 0.4407

Epoch 44/50

391/391 [=====] - 4s 11ms/step - loss: 1.7580 - accuracy: 0.3770 - val_loss: 1.6775 - val_accuracy: 0.4078

Epoch 45/50

391/391 [=====] - 5s 12ms/step - loss: 1.7555 - accuracy: 0.3784 - val_loss: 1.6709 - val_accuracy: 0.4206

Epoch 46/50

391/391 [=====] - 4s 11ms/step - loss: 1.7561 - accuracy: 0.3757 - val_loss: 1.6531 - val_accuracy: 0.4467

Epoch 47/50

391/391 [=====] - 4s 11ms/step - loss: 1.7671 - accuracy: 0.3766 - val_loss: 1.6530 - val_accuracy: 0.4182

Epoch 48/50

391/391 [=====] - 4s 11ms/step - loss: 1.7608 - accuracy: 0.3772 - val_loss: 1.5990 - val_accuracy: 0.4375

Epoch 49/50

391/391 [=====] - 4s 11ms/step - loss: 1.7620 - accuracy: 0.3776 - val_loss: 1.6012 - val_accuracy: 0.4356

Epoch 50/50

391/391 [=====] - 4s 11ms/step - loss: 1.7647 - accuracy: 0.3727 - val_loss: 1.6235 - val_accuracy: 0.4352

Epoch 51/50

391/391 [=====] - 4s 11ms/step - loss: 1.7626 - accuracy: 0.3758 - val_loss: 1.6079 - val_accuracy: 0.4411

Test loss: 1.6078776121139526

Test accuracy: 0.4411000133514404

Model Accuracy

Model Loss

Incluyendo capas de dropout, pero sin batch normalization

In (85):

```
Model = build_model_v2(bn=False)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_23"
```

Layer (type)	Output Shape	Param #
rescaling_45 (Rescaling)	(None, 32, 32, 3)	0
random_rotation_45 (RandomRo	(None, 32, 32, 3)	0
random_zoom_45 (RandomZoom)	(None, 32, 32, 3)	0
random_flip_45 (RandomFlip)	(None, 32, 32, 3)	0
flatten_56 (Flatten)	(None, 3072)	0
dense_147 (Dense)	(None, 128)	393344
dropout_115 (Dropout)	(None, 128)	0
activation_295 (Activation)	(None, 128)	0
dense_148 (Dense)	(None, 64)	8256
dropout_116 (Dropout)	(None, 64)	0
activation_296 (Activation)	(None, 64)	0
dense_149 (Dense)	(None, 32)	2080
dropout_117 (Dropout)	(None, 32)	0
activation_297 (Activation)	(None, 32)	0
dense_150 (Dense)	(None, 16)	528
batch_normalization_218 (Bat	(None, 16)	64
dropout_118 (Dropout)	(None, 16)	0
activation_298 (Activation)	(None, 16)	0
dense_151 (Dense)	(None, 10)	170
activation_299 (Activation)	(None, 10)	0
Total params: 404,442		
Trainable params: 404,410		
Non-trainable params: 32		

Epoch 1/50

391/391 [=====] - 5s 11ms/step - loss: 2.2792 - accuracy: 0.1217 - val_loss: 2.2186 - val_accuracy: 0.1692

Epoch 2/50

391/391 [=====] - 4s 11ms/step - loss: 2.1467 - accuracy: 0.1720 - val_loss: 2.0550 - val_accuracy: 0.2187

Epoch 3/50

391/391 [=====] - 4s 11ms/step - loss: 2.1180 - accuracy: 0.1869 - val_loss: 2.0271 - val_accuracy: 0.2329

Epoch 4/50

391/391 [=====] - 4s 10ms/step - loss: 2.1050 - accuracy: 0.1940 - val_loss: 2.0420 - val_accuracy: 0.2361

Epoch 5/50

391/391 [=====] - 4s 11ms/step - loss: 2.0797 - accuracy: 0.2034 - val_loss: 1.9697 - val_accuracy: 0.2356

Epoch 6/50

391/391 [=====] - 4s 11ms/step - loss: 2.0771 - accuracy: 0.2094 - val_loss: 1.9779 - val_accuracy: 0.2177

Epoch 7/50

391/391 [=====] - 4s 11ms/step - loss: 2.0673 - accuracy: 0.2124 - val_loss: 1.9779 - val_accuracy: 0.2462

Epoch 8/50

391/391 [=====] - 4s 11ms/step - loss: 2.0573 - accuracy: 0.2239 - val_loss: 1.9699 - val_accuracy: 0.2565

Epoch 9/50

391/391 [=====] - 4s 11ms/step - loss: 2.0450 - accuracy: 0.2216 - val_loss: 1.9540 - val_accuracy: 0.2088

Epoch 10/50

391/391 [=====] - 4s 11ms/step - loss: 2.0489 - accuracy: 0.2213 - val_loss: 1.9572 - val_accuracy: 0.2490

Epoch 11/50

391/391 [=====] - 4s 10ms/step - loss: 2.03


```
[86]: model = build_model2(debug=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

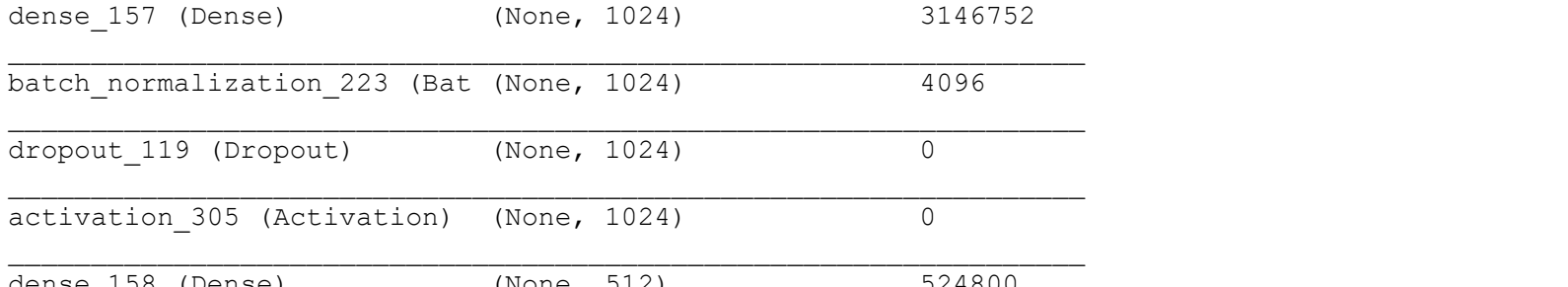
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_24"

Layer (type) Output Shape Param #
-----
rescaling_46 (Rescaling) (None, 32, 32, 3) 0
random_rotation_46 (RandomRo (None, 32, 32, 3) 0
random_zoom_46 (RandomZoom) (None, 32, 32, 3) 0
random_flip_46 (RandomFlip) (None, 32, 32, 3) 0
flatten_57 (Flatten) (None, 3072) 0
dense_152 (Dense) (None, 128) 393344
batch_normalization_219 (Bat (None, 128) 512
activation_300 (Activation) (None, 128) 0
dense_153 (Dense) (None, 64) 8256
batch_normalization_220 (Bat (None, 64) 256
activation_301 (Activation) (None, 64) 0
dense_154 (Dense) (None, 32) 2080
batch_normalization_221 (Bat (None, 32) 128
activation_302 (Activation) (None, 32) 0
dense_155 (Dense) (None, 16) 528
batch_normalization_222 (Bat (None, 16) 64
activation_303 (Activation) (None, 16) 0
dense_156 (Dense) (None, 10) 170
activation_304 (Activation) (None, 10) 0
-----
Total params: 405,338
Trainable params: 404,858
Non-trainable params: 480

Epoch 1/50
391/391 [=====] - 6s 11ms/step - loss: 2.1057 - accuracy: 0.2487 - val_loss: 2.4896 - val_accuracy: 0.2438
Epoch 2/50
391/391 [=====] - 4s 11ms/step - loss: 1.7854 - accuracy: 0.3653 - val_loss: 1.6742 - val_accuracy: 0.3718
Epoch 3/50
391/391 [=====] - 4s 11ms/step - loss: 1.7093 - accuracy: 0.3885 - val_loss: 1.4591 - val_accuracy: 0.3901
Epoch 4/50
391/391 [=====] - 4s 11ms/step - loss: 1.6701 - accuracy: 0.4074 - val_loss: 1.4219 - val_accuracy: 0.4228
Epoch 5/50
391/391 [=====] - 4s 11ms/step - loss: 1.6275 - accuracy: 0.4207 - val_loss: 1.4316 - val_accuracy: 0.3341
Epoch 6/50
391/391 [=====] - 4s 11ms/step - loss: 1.6275 - accuracy: 0.4207 - val_loss: 1.4660 - val_accuracy: 0.3836
Epoch 7/50
391/391 [=====] - 4s 11ms/step - loss: 1.6095 - accuracy: 0.4265 - val_loss: 1.7610 - val_accuracy: 0.3596
Epoch 8/50
391/391 [=====] - 4s 11ms/step - loss: 1.5829 - accuracy: 0.4350 - val_loss: 1.6460 - val_accuracy: 0.4528
Epoch 9/50
391/391 [=====] - 4s 11ms/step - loss: 1.5652 - accuracy: 0.4403 - val_loss: 1.7040 - val_accuracy: 0.3943
Epoch 10/50
391/391 [=====] - 4s 11ms/step - loss: 1.5524 - accuracy: 0.4435 - val_loss: 1.6781 - val_accuracy: 0.4082
Epoch 11/50
391/391 [=====] - 4s 11ms/step - loss: 1.5375 - accuracy: 0.4513 - val_loss: 1.5916 - val_accuracy: 0.4262
Epoch 12/50
391/391 [=====] - 4s 11ms/step - loss: 1.5362 - accuracy: 0.4537 - val_loss: 1.4896 - val_accuracy: 0.3750
Epoch 13/50
391/391 [=====] - 4s 11ms/step - loss: 1.5210 - accuracy: 0.4581 - val_loss: 1.4742 - val_accuracy: 0.4142
Epoch 14/50
391/391 [=====] - 4s 11ms/step - loss: 1.5050 - accuracy: 0.4631 - val_loss: 1.4246 - val_accuracy: 0.4228
Epoch 15/50
391/391 [=====] - 4s 11ms/step - loss: 1.5006 - accuracy: 0.4666 - val_loss: 1.4246 - val_accuracy: 0.4210
Epoch 16/50
391/391 [=====] - 4s 11ms/step - loss: 1.4893 - accuracy: 0.4699 - val_loss: 1.4520 - val_accuracy: 0.4087
Epoch 17/50
391/391 [=====] - 4s 11ms/step - loss: 1.4783 - accuracy: 0.4736 - val_loss: 1.6710 - val_accuracy: 0.4142
Epoch 18/50
391/391 [=====] - 4s 11ms/step - loss: 1.4683 - accuracy: 0.4689 - val_loss: 1.4384 - val_accuracy: 0.4179
Epoch 19/50
391/391 [=====] - 4s 11ms/step - loss: 1.4624 - accuracy: 0.4716 - val_loss: 1.4781 - val_accuracy: 0.4082
Epoch 20/50
391/391 [=====] - 4s 11ms/step - loss: 1.4520 - accuracy: 0.4736 - val_loss: 1.5520 - val_accuracy: 0.4483
Epoch 21/50
391/391 [=====] - 4s 11ms/step - loss: 1.4783 - accuracy: 0.4736 - val_loss: 1.4896 - val_accuracy: 0.4386
Epoch 22/50
391/391 [=====] - 4s 11ms/step - loss: 1.4698 - accuracy: 0.4772 - val_loss: 1.4574 - val_accuracy: 0.4432
Epoch 23/50
391/391 [=====] - 4s 11ms/step - loss: 1.4621 - accuracy: 0.4790 - val_loss: 1.4571 - val_accuracy: 0.4571
Epoch 24/50
391/391 [=====] - 4s 11ms/step - loss: 1.4616 - accuracy: 0.4800 - val_loss: 1.4710 - val_accuracy: 0.4710
Epoch 25/50
391/391 [=====] - 4s 11ms/step - loss: 1.4585 - accuracy: 0.4817 - val_loss: 1.5140 - val_accuracy: 0.4592
Epoch 26/50
391/391 [=====] - 4s 11ms/step - loss: 1.4539 - accuracy: 0.4838 - val_loss: 1.4413 - val_accuracy: 0.4441
Epoch 27/50
391/391 [=====] - 4s 11ms/step - loss: 1.4391 - accuracy: 0.4851 - val_loss: 1.4791 - val_accuracy: 0.4723
Epoch 28/50
391/391 [=====] - 4s 11ms/step - loss: 1.4426 - accuracy: 0.4871 - val_loss: 1.4545 - val_accuracy: 0.4757
Epoch 29/50
391/391 [=====] - 4s 11ms/step - loss: 1.4403 - accuracy: 0.4881 - val_loss: 1.4846 - val_accuracy: 0.4742
Epoch 30/50
391/391 [=====] - 4s 11ms/step - loss: 1.4326 - accuracy: 0.4912 - val_loss: 1.4391 - val_accuracy: 0.4693
Epoch 31/50
391/391 [=====] - 4s 11ms/step - loss: 1.4231 - accuracy: 0.4939 - val_loss: 1.4591 - val_accuracy: 0.4534
Epoch 32/50
391/391 [=====] - 4s 11ms/step - loss: 1.4210 - accuracy: 0.4952 - val_loss: 1.4646 - val_accuracy: 0.4646
Epoch 33/50
391/391 [=====] - 4s 11ms/step - loss: 1.4276 - accuracy: 0.4905 - val_loss: 1.4413 - val_accuracy: 0.4896
Epoch 34/50
391/391 [=====] - 4s 11ms/step - loss: 1.4201 - accuracy: 0.4977 - val_loss: 1.4591 - val_accuracy: 0.4783
Epoch 35/50
391/391 [=====] - 4s 11ms/step - loss: 1.4116 - accuracy: 0.4971 - val_loss: 1.5702 - val_accuracy: 0.4562
Epoch 36/50
391/391 [=====] - 4s 11ms/step - loss: 1.4144 - accuracy: 0.4938 - val_loss: 1.4701 - val_accuracy: 0.4835
Epoch 37/50
391/391 [=====] - 4s 11ms/step - loss: 1.4166 - accuracy: 0.4955 - val_loss: 1.4340 - val_accuracy: 0.4867
Epoch 38/50
391/391 [=====] - 4s 11ms/step - loss: 1.4161 - accuracy: 0.4938 - val_loss: 1.5670 - val_accuracy: 0.4469
Epoch 39/50
391/391 [=====] - 4s 11ms/step - loss: 1.3791 - accuracy: 0.5062 - val_loss: 1.4698 - val_accuracy: 0.4981
Epoch 40/50
391/391 [=====] - 4s 11ms/step - loss: 1.3852 - accuracy: 0.5078 - val_loss: 1.4742 - val_accuracy: 0.4360
Epoch 41/50
391/391 [=====] - 4s 11ms/step - loss: 1.3843 - accuracy: 0.5086 - val_loss: 1.4742 - val_accuracy: 0.4603
Epoch 42/50
391/391 [=====] - 4s 11ms/step - loss: 1.3971 - accuracy: 0.5033 - val_loss: 1.4627 - val_accuracy: 0.4627
Epoch 43/50
391/391 [=====] - 4s 11ms/step - loss: 1.3971 - accuracy: 0.5033 - val_loss: 1.4719 - val_accuracy: 0.4780
Epoch 44/50
391/391 [=====] - 4s 11ms/step - loss: 1.3854 - accuracy: 0.5080 - val_loss: 1.4719 - val_accuracy: 0.4780
Epoch 45/50
391/391 [=====] - 4s 11ms/step - loss: 1.3854 - accuracy: 0.5080 - val_loss: 1.4671 - val_accuracy: 0.4277
Epoch 46/50
391/391 [=====] - 4s 11ms/step - loss: 1.3946 - accuracy: 0.5038 - val_loss: 1.4442 - val_accuracy: 0.4914
Epoch 47/50
391/391 [=====] - 4s 11ms/step - loss: 1.3858 - accuracy: 0.5107 - val_loss: 1.4442 - val_accuracy: 0.4914
Epoch 48/50
391/391 [=====] - 4s 11ms/step - loss: 1.3791 - accuracy: 0.5062 - val_loss: 1.5670 - val_accuracy: 0.4469
Epoch 49/50
391/391 [=====] - 4s 11ms/step - loss: 1.3791 - accuracy: 0.5062 - val_loss: 1.4698 - val_accuracy: 0.4981
Epoch 50/50
391/391 [=====] - 4s 11ms/step - loss: 1.3852 - accuracy: 0.5078 - val_loss: 1.4742 - val_accuracy: 0.4360
Test accuracy: 0.50050023500208

Model Accuracy
```



Training MLP took 215.06497025489807 seconds

Modelo fNN 3

Incluyendo capas de dropout y batch normalization

```
[87]: model = build_model_v3()

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

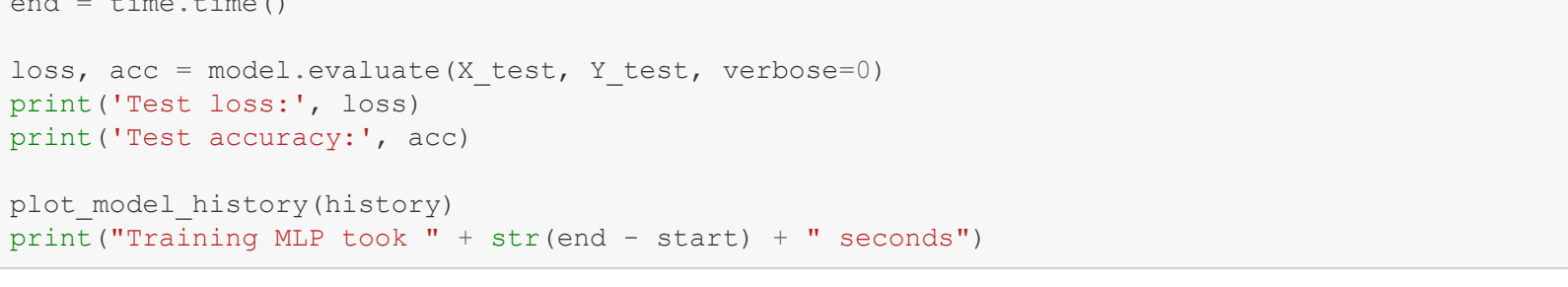
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_25"

Layer (type) Output Shape Param #
-----
rescaling_47 (Rescaling) (None, 32, 32, 3) 0
random_rotation_47 (RandomRo (None, 32, 32, 3) 0
random_zoom_47 (RandomZoom) (None, 32, 32, 3) 0
random_flip_47 (RandomFlip) (None, 32, 32, 3) 0
flatten_58 (Flatten) (None, 3072) 0
dense_157 (Dense) (None, 1024) 3146752
batch_normalization_223 (Bat (None, 1024) 4096
dropout_119 (Dropout) (None, 1024) 0
activation_305 (Activation) (None, 1024) 0
dense_158 (Dense) (None, 512) 524800
batch_normalization_224 (Bat (None, 512) 2048
dropout_120 (Dropout) (None, 512) 0
activation_306 (Activation) (None, 512) 0
dense_159 (Dense) (None, 64) 28232
batch_normalization_225 (Bat (None, 64) 256
dropout_121 (Dropout) (None, 64) 0
activation_307 (Activation) (None, 64) 0
dense_160 (Dense) (None, 64) 4160
batch_normalization_226 (Bat (None, 64) 256
dropout_122 (Dropout) (None, 64) 0
activation_308 (Activation) (None, 64) 0
dense_161 (Dense) (None, 10) 650
activation_309 (Activation) (None, 10) 0
-----
Total params: 3,715,850
Trainable params: 3,712,522
Non-trainable params: 3,328

Epoch 1/50
391/391 [=====] - 13s 28ms/step - loss: 2.2159 - accuracy: 0.2143 - val_loss: 2.1688 - val_accuracy: 0.1888
Epoch 2/50
391/391 [=====] - 11s 29ms/step - loss: 1.8941 - accuracy: 0.3218 - val_loss: 2.0414 - val_accuracy: 0.2656
Epoch 3/50
391/391 [=====] - 14s 36ms/step - loss: 1.8134 - accuracy: 0.3512 - val_loss: 1.9998 - val_accuracy: 0.3148
Epoch 4/50
391/391 [=====] - 13s 34ms/step - loss: 1.7600 - accuracy: 0.3718 - val_loss: 1.7318 - val_accuracy: 0.3712
Epoch 5/50
391/391 [=====] - 13s 34ms/step - loss: 1.7371 - accuracy: 0.3819 - val_loss: 1.7210 - val_accuracy: 0.3821
Epoch 6/50
391/391 [=====] - 14s 36ms/step - loss: 1.7185 - accuracy: 0.3896 - val_loss: 1.7242 - val_accuracy: 0.3856
Epoch 7/50
391/391 [=====] - 13s 34ms/step - loss: 1.7042 - accuracy: 0.3995 - val_loss: 1.7460 - val_accuracy: 0.3749
Epoch 8/50
391/391 [=====] - 13s 34ms/step - loss: 1.6759 - accuracy: 0.4114 - val_loss: 1.7244 - val_accuracy: 0.3812
Epoch 9/50
391/391 [=====] - 13s 34ms/step - loss: 1.6687 - accuracy: 0.4137 - val_loss: 1.6188 - val_accuracy: 0.4234
Epoch 10/50
391/391 [=====] - 13s 34ms/step - loss: 1.6568 - accuracy: 0.4155 - val_loss: 1.7107 - val_accuracy: 0.3752
Epoch 11/50
391/391 [=====] - 13s 34ms/step - loss: 1.6396 - accuracy: 0.4184 - val_loss: 1.6734 - val_accuracy: 0.4057
Epoch 12/50
391/391 [=====] - 13s 33ms/step - loss: 1.6309 - accuracy: 0.4246 - val_loss: 1.6481 - val_accuracy: 0.4214
Epoch 13/50
391/391 [=====] - 13s 34ms/step - loss: 1.6185 - accuracy: 0.4275 - val_loss: 1.6303 - val_accuracy: 0.3593
Epoch 14/50
391/391 [=====] - 13s 34ms/step - loss: 1.6098 - accuracy: 0.4335 - val_loss: 1.6170 - val_accuracy: 0.4289
Epoch 15/50
391/391 [=====] - 13s 33ms/step - loss: 1.5914 - accuracy: 0.4404 - val_loss: 1.6463 - val_accuracy: 0.4037
Epoch 16/50
391/391 [=====] - 13s 34ms/step - loss: 1.5981 - accuracy: 0.4347 - val_loss: 1.6648 - val_accuracy: 0.4092
Epoch 17/50
391/391 [=====] - 13s 33ms/step - loss: 1.5894 - accuracy: 0.4405 - val_loss: 1.5725 - val_accuracy: 0.4373
Epoch 18/50
391/391 [=====] - 13s 34ms/step - loss: 1.5837 - accuracy: 0.4420 - val_loss: 1.5585 - val_accuracy: 0.4383
Epoch 19/50
391/391 [=====] - 13s 34ms/step - loss: 1.5761 - accuracy: 0.4449 - val_loss: 1.6843 - val_accuracy: 0.4017
Epoch 20/50
391/391 [=====] - 13s 33ms/step - loss: 1.5706 - accuracy: 0.4467 - val_loss: 1.7176 - val_accuracy: 0.3816
Epoch 21/50
391/391 [=====] - 13s 33ms/step - loss: 1.5658 - accuracy: 0.4520 - val_loss: 1.5337 - val_accuracy: 0.4577
Epoch 22/50
391/391 [=====] - 14s 37ms/step - loss: 1.5567 - accuracy: 0.4529 - val_loss: 1.5748 - val_accuracy: 0.4602
Epoch 23/50
391/391 [=====] - 13s 34ms/step - loss: 1.5502 - accuracy: 0.4595 - val_loss: 1.5732 - val_accuracy: 0.4464
Epoch 24/50
391/391 [=====] - 13s 34ms/step - loss: 1.5589 - accuracy: 0.4458 - val_loss: 1.5739 - val_accuracy: 0.4746
Epoch 25/50
391/391 [=====] - 13s 34ms/step - loss: 1.5412 - accuracy: 0.4559 - val_loss: 1.4859 - val_accuracy: 0.4656
Epoch 26/50
391/391 [=====] - 13s 34ms/step - loss: 1.5360 - accuracy: 0.4605 - val_loss: 1.6088 - val_accuracy: 0.4194
Epoch 27/50
391/391 [=====] - 13s 34ms/step - loss: 1.5262 - accuracy: 0.4609 - val_loss: 1.4650 - val_accuracy: 0.4812
Epoch 28/50
391/391 [=====] - 13s 34ms/step - loss: 1.5376 - accuracy: 0.4571 - val_loss: 1.4367 - val_accuracy: 0.4957
Epoch 29/50
391/391 [=====] - 13s 34ms/step - loss: 1.5153 - accuracy: 0.4649 - val_loss: 1.4398 - val_accuracy: 0.4918
Epoch 30/50
391/391 [=====] - 13s 34ms/step - loss: 1.5129 - accuracy: 0.4655 - val_loss: 1.4528 - val_accuracy: 0.4789
Epoch 31/50
391/391 [=====] - 14s 35ms/step - loss: 1.5233 - accuracy: 0.4621 - val_loss: 1.4393 - val_accuracy: 0.4674
Epoch 32/50
391/391 [=====] - 13s 34ms/step - loss: 1.5084 - accuracy: 0.4676 - val_loss: 1.4592 - val_accuracy: 0.4461
Epoch 33/50
391/391 [=====] - 13s 34ms/step - loss: 1.5024 - accuracy: 0.4756 - val_loss: 1.4715 - val_accuracy: 0.4715
Epoch 34/50
391/391 [=====] - 13s 34ms/step - loss: 1.4991 - accuracy: 0.4708 - val_loss: 1.4431 - val_accuracy: 0.4930
Epoch 35/50
391/391 [=====] - 13s 34ms/step - loss: 1.5013 - accuracy: 0.4741 - val_loss: 1.5860 - val_accuracy: 0.4322
Epoch 36/50
391/391 [=====] - 14s 35ms/step - loss: 1.4999 - accuracy: 0.4743 - val_loss: 1.5117 - val_accuracy: 0.4552
Epoch 37/50
391/391 [=====] - 14s 35ms/step - loss: 1.4823 - accuracy: 0.4768 - val_loss: 1.4163 - val_accuracy: 0.4990
Epoch 38/50
391/391 [=====] - 12s 32ms/step - loss: 1.4855 - accuracy: 0.4809 - val_loss: 1.4070 - val_accuracy: 0.5105
Epoch 39/50
391/391 [=====] - 12s 30ms/step - loss: 1.4685 - accuracy: 0.4831 - val_loss: 1.4366 - val_accuracy: 0.4545
Epoch 40/50
391/391 [=====] - 12s 31ms/step - loss: 1.4682 - accuracy: 0.4853 - val_loss: 1.4593 - val_accuracy: 0.4848
Epoch 41/50
391/391 [=====] - 12s 32ms/step - loss: 1.4761 - accuracy: 0.4835 - val_loss: 1.4592 - val_accuracy: 0.4778
Epoch 42/50
391/391 [=====] - 12s 30ms/step - loss: 1.4694 - accuracy: 0.4869 - val_loss: 1.4431 - val_accuracy: 0.4819
Epoch 43/50
391/391 [=====] - 12s 30ms/step - loss: 1.4724 - accuracy: 0.4860 - val_loss: 1.4464 - val_accuracy: 0.5000
Epoch 44/50
391/391 [=====] - 11s 29ms/step - loss: 1.4697 - accuracy: 0.4823 - val_loss: 1.4464 - val_accuracy: 0.4856
Epoch 45/50
391/391 [=====] - 12s 30ms/step - loss: 1.4728 - accuracy: 0.4873 - val_loss: 1.5912 - val_accuracy: 0.4298
Epoch 46/50
391/391 [=====] - 12s 31ms/step - loss: 1.4607 - accuracy: 0.4915 - val_loss: 1.3820 - val_accuracy: 0.5132
Epoch 47/50
391/391 [=====] - 12s 31ms/step - loss: 1.4477 - accuracy: 0.4897 - val_loss: 1.4750 - val_accuracy: 0.4680
Epoch 48/50
391/391 [=====] - 14s 35ms/step - loss: 1.4509 - accuracy: 0.4942 - val_loss: 1.4824 - val_accuracy: 0.4973
Epoch 49/50
391/391 [=====] - 14s 36ms/step - loss: 1.4437 - accuracy: 0.4976 - val_loss: 1.4350 - val_accuracy: 0.4900
Epoch 50/50
391/391 [=====] - 14s 36ms/step - loss: 1.4442 - accuracy: 0.4944 - val_loss: 1.4150 - val_accuracy: 0.4762
Test loss: 1.4149885177612305
Test accuracy: 0.4873999532699585

Model Accuracy
```



Training MLP took 651.9230682849884 seconds

Incluyendo capas de dropout, pero sin batch normalization

```
[88]: model = build_model_v3(bn=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_26"

Layer (type) Output Shape Param #
-----
rescaling_48 (Rescaling) (None, 32, 32, 3) 0
random_rotation_48 (RandomRo (None, 32, 32, 3) 0
random_zoom_48 (RandomZoom) (None, 32, 32, 3) 0
random_flip_48 (RandomFlip) (None, 32, 32, 3) 0
flatten_59 (Flatten) (None, 3072) 0
dense_162 (Dense) (None, 1024) 3146752
dropout_123 (Dropout) (None, 1024) 0
activation_310 (Activation) (None, 1024) 0
dense_163 (Dense) (None, 512) 524800
dropout_124 (Dropout) (None, 512) 0
activation_311 (Activation) (None, 512) 0
dense_164 (Dense) (None, 64) 28232
batch_normalization_227 (Bat (None, 64) 256
dropout_125 (Dropout) (None, 64) 0
activation_312 (Activation) (None, 64) 0
dense_165 (Dense) (None, 64) 4160
dropout_126 (Dropout) (None, 64) 0
dense_166 (Dense) (None, 10) 650
activation_314 (Activation) (None, 10) 0
-----
Total params: 3,709,450
Trainable params: 3,709,322
Non-trainable params: 128

Epoch 1/50
391/391 [=====] - 14s 33ms/step - loss: 2.1934 - accuracy: 0.1755 - val_loss: 1.4891 - val_accuracy: 0.2533
Epoch 2/50
391/391 [=====] - 12s 32ms/step - loss: 2.0218 - accuracy: 0.2613 - val_loss: 1.4580 - val_accuracy: 0.3304
Epoch 3/50
391/391 [=====] - 13s 32ms/step - loss: 1.9841 - accuracy: 0.2770 - val_loss: 1.8752 - val_accuracy: 0.3262
Epoch 4/50
391/391 [=====] - 12s 32ms/step - loss: 1.9665 - accuracy: 0.2867 - val_loss: 1.8180 - val_accuracy: 0.3522
Epoch 5/50
391/391 [=====] - 13s 32ms/step - loss: 1.9473 - accuracy: 0.2976 - val_loss: 1.8344 - val_accuracy: 0.3435
Epoch 6/50
391/391 [=====] - 12s 32ms/step - loss: 1.9272 - accuracy: 0.3049 - val_loss: 1.7966 - val_accuracy: 0.3523
Epoch 7/50
391/391 [=====] - 12s 32ms/step - loss: 1.9169 - accuracy: 0.3085 - val_loss: 1.8701 - val_accuracy: 0.3707
Epoch 8/50
391/391 [=====] - 12s 32ms/step - loss: 1.9092 - accuracy: 0.3115 - val_loss: 1.6711 - val_accuracy: 0.4062
Epoch 9/50
391/391 [=====] - 12s 30ms/step - loss: 1.8914 - accuracy: 0.3197 - val_loss: 1.6580 - val_accuracy: 0.4064
Epoch 10/50
391/391 [=====] - 13s 32ms/step - loss: 1.8926 - accuracy: 0.3216 - val_loss: 1.6402 - val_accuracy: 0.4105
Epoch 11/50
391/391 [=====] - 11s 28ms/step - loss: 1.8728 - accuracy: 0.3306 - val_loss: 1.6175 - val_accuracy: 0.3851
Epoch 12/50
391/391 [=====] - 12s 31ms/step - loss: 1.8704 - accuracy: 0.3320 - val_loss: 1.7233 - val_accuracy: 0.3844
Epoch 13/50
391/391 [=====] - 11s 29ms/step - loss: 1.8613 - accuracy: 0.3292 - val_loss: 1.7106 - val_accuracy: 0.3886
Epoch 14/50
391/391 [=====] - 11s 28ms/step - loss: 1.8588 - accuracy: 0.3355 - val_loss: 1.7069 - val_accuracy: 0.3968
Epoch 15/50
391/391 [=====] - 11s 28ms/step - loss: 1.8495 - accuracy: 0.3385 - val_loss: 1.6933 - val_accuracy: 0.3922
Epoch 16/50
391/391 [=====] - 11s 28ms/step - loss: 1.8561 - accuracy: 0.3401 - val_loss: 1.6413 - val_accuracy: 0.4016
Epoch 17/50
391/391 [=====] - 11s 27ms/step - loss: 1.8446 - accuracy: 0.3352 - val_loss: 1.6711 - val_accuracy: 0.4062
Epoch 18/50
391/391 [=====] - 11s 28ms/step - loss: 1.8440 - accuracy: 0.3425 - val_loss: 1.6580 - val_accuracy: 0.4064
Epoch 19/50
391/391 [=====] - 10s 27ms/step - loss: 1.8366 - accuracy: 0.3453 - val_loss: 1.6402 - val_accuracy: 0.4105
Epoch 20/50
391/391 [=====] - 11s 27ms/step - loss: 1.8229 - accuracy: 0.3546 - val_loss: 1.6580 - val_accuracy: 0.4162
Epoch 21/50
391/391 [=====] - 11s 28ms/step - loss: 1.8271 - accuracy: 0.3473 - val_loss: 1.6445 - val_accuracy: 0.4164
Epoch 22/50
391/391 [=====] - 11s 28ms/step - loss: 1.8217 - accuracy: 0.3509 - val_loss: 1.6491 - val_accuracy: 0.4159
Epoch 23/50
391/391 [=====] - 11s 28ms/step - loss: 1.8178 - accuracy: 0.3479 - val_loss: 1.6525 - val_accuracy: 0.4124
Epoch 24/50
391/391 [=====] - 11s 28ms/step - loss: 1.8246 - accuracy: 0.3488 - val_loss: 1.6621 - val_accuracy: 0.4066
Epoch 25/50
391/391 [=====] - 10s 27ms/step - loss: 1.8155 - accuracy: 0.3522 - val_loss: 1.6449 - val_accuracy: 0.4133
Epoch 26/50
391/391 [=====] - 10s 26ms/step - loss: 1.8150 - accuracy: 0.3523 - val_loss: 1.6413 - val_accuracy: 0.4155
Epoch 27/50
391/391 [=====] - 10s 27ms/step - loss: 1.8093 - accuracy: 0.3565 - val_loss: 1.6460 - val_accuracy: 0.4212
Epoch 28/50
391/391 [=====] - 10s 27ms/step - loss: 1.8122 - accuracy: 0.3547 - val_loss: 1.6402 - val_accuracy: 0.4187
Epoch 29/50
391/391 [=====] - 13s 33ms/step - loss: 1.7966 - accuracy: 0.3590 - val_loss: 1.6325 - val_accuracy: 0.4244
Epoch 30/50
391/391 [=====] - 12s 32ms/step - loss: 1.8058 - accuracy: 0.3579 - val_loss: 1.6285 - val_accuracy: 0.4176
Epoch 31/50
391/391 [=====] - 15s 38ms/step - loss: 1.7968 - accuracy: 0.3578 - val_loss: 1.6260 - val_accuracy: 0.4206
Epoch 32/50
391/391 [=====] - 12s 31ms/step - loss: 1.7974 - accuracy: 0.3577 - val_loss: 1.6249 - val_accuracy: 0.4212
Epoch 33/50
391/391 [=====] - 12s 31ms/step - loss: 1.7909 - accuracy: 0.3609 - val_loss: 1.6338 - val_accuracy: 0.4184
Epoch 34/50
391/391 [=====] - 11s 29ms/step - loss: 1.7955 - accuracy: 0.3600 - val_loss: 1.6150 - val_accuracy: 0.4209
Epoch 35/50
391/391 [=====] - 12s 32ms/step - loss: 1.7810 - accuracy: 0.3657 - val_loss: 1.6410 - val_accuracy: 0.4086
Epoch 36/50
391/391 [=====] - 13s 33ms/step - loss: 1.7822 - accuracy: 0.3624 - val_loss: 1.6070 - val_accuracy: 0.4284
Epoch 37/50
391/391 [=====] - 12s 30ms/step - loss: 1.7846 - accuracy: 0.3621 - val_loss: 1.6070 - val
```



```
[89]: model = build_model(dout=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=128, epochs=50, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

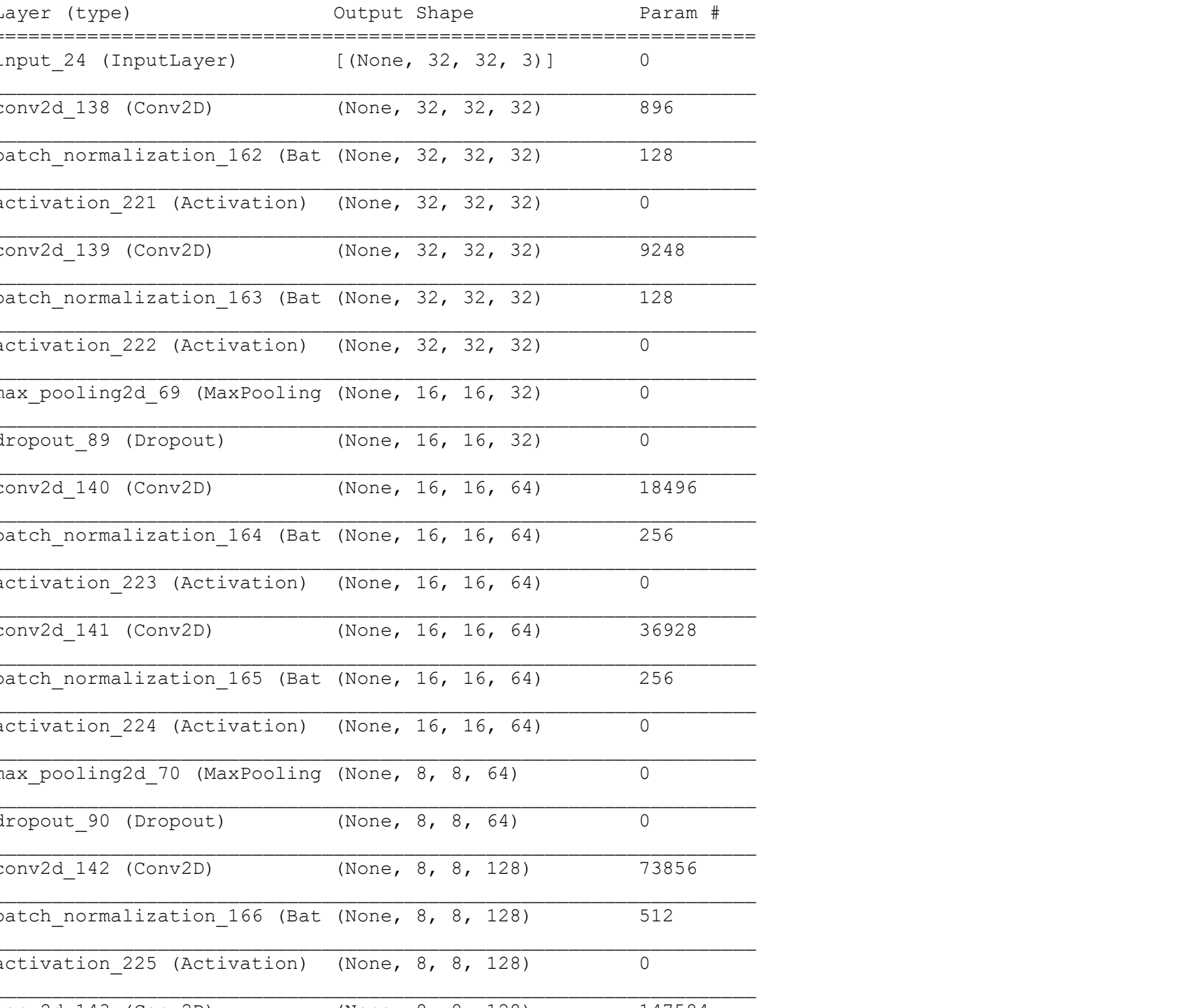
loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "sequential_2"

Layer (type) Output Shape Param #
-----
rescaling_49 (Rescaling) (None, 32, 32, 3) 0
random_rotation_49 (RandomRo (None, 32, 32, 3) 0
random_zoom_49 (RandomZoom) (None, 32, 32, 3) 0
random_flip_49 (RandomFlip) (None, 32, 32, 3) 0
flatten_60 (Flatten) (None, 3072) 0
dense_167 (Dense) (None, 1024) 3146752
batch_normalization_228 (Bat (None, 1024) 4096
activation_315 (Activation) (None, 1024) 0
dense_168 (Dense) (None, 512) 524800
batch_normalization_229 (Bat (None, 512) 2048
activation_316 (Activation) (None, 512) 0
dense_169 (Dense) (None, 64) 2832
batch_normalization_230 (Bat (None, 64) 256
activation_317 (Activation) (None, 64) 0
dense_170 (Dense) (None, 64) 4160
batch_normalization_231 (Bat (None, 64) 256
activation_318 (Activation) (None, 64) 0
dense_171 (Dense) (None, 10) 650
activation_319 (Activation) (None, 10) 0
Total params: 3,715,850
Trainable params: 3,712,522
Non-trainable params: 3,328

Epoch 1/50
391/391 [=====] - 14s 31ms/step - loss: 1.9528 - accuracy: 0.3035 - val_loss: 1.1712 - val_accuracy: 0.1804
391/391 [=====] - 12s 31ms/step - loss: 1.6937 - accuracy: 0.3977 - val_loss: 1.1508 - val_accuracy: 0.3264
Epoch 2/50
391/391 [=====] - 12s 30ms/step - loss: 1.6341 - accuracy: 0.4172 - val_loss: 1.1461 - val_accuracy: 0.3638
Epoch 3/50
391/391 [=====] - 11s 29ms/step - loss: 1.5783 - accuracy: 0.4393 - val_loss: 1.1508 - val_accuracy: 0.3690
Epoch 4/50
391/391 [=====] - 11s 29ms/step - loss: 1.5403 - accuracy: 0.4482 - val_loss: 1.1609 - val_accuracy: 0.3225
Epoch 5/50
391/391 [=====] - 12s 30ms/step - loss: 1.5164 - accuracy: 0.4602 - val_loss: 1.6943 - val_accuracy: 0.4009
Epoch 6/50
391/391 [=====] - 11s 29ms/step - loss: 1.4926 - accuracy: 0.4673 - val_loss: 1.5998 - val_accuracy: 0.4264
Epoch 7/50
391/391 [=====] - 11s 27ms/step - loss: 1.4730 - accuracy: 0.4749 - val_loss: 1.6631 - val_accuracy: 0.4104
Epoch 8/50
391/391 [=====] - 11s 28ms/step - loss: 1.4589 - accuracy: 0.4791 - val_loss: 1.6444 - val_accuracy: 0.4217
Epoch 9/50
391/391 [=====] - 12s 30ms/step - loss: 1.4403 - accuracy: 0.4844 - val_loss: 1.7135 - val_accuracy: 0.3957
Epoch 10/50
391/391 [=====] - 11s 28ms/step - loss: 1.4300 - accuracy: 0.4905 - val_loss: 1.7145 - val_accuracy: 0.3799
Epoch 11/50
391/391 [=====] - 11s 28ms/step - loss: 1.4132 - accuracy: 0.4937 - val_loss: 1.7145 - val_accuracy: 0.3866
Epoch 12/50
391/391 [=====] - 11s 29ms/step - loss: 1.3863 - accuracy: 0.5066 - val_loss: 1.5490 - val_accuracy: 0.4450
Epoch 13/50
391/391 [=====] - 11s 28ms/step - loss: 1.3832 - accuracy: 0.5074 - val_loss: 1.5704 - val_accuracy: 0.4373
Epoch 14/50
391/391 [=====] - 11s 29ms/step - loss: 1.3662 - accuracy: 0.5120 - val_loss: 1.6023 - val_accuracy: 0.4325
Epoch 15/50
391/391 [=====] - 11s 27ms/step - loss: 1.3428 - accuracy: 0.5226 - val_loss: 1.6943 - val_accuracy: 0.4009
Epoch 16/50
391/391 [=====] - 11s 28ms/step - loss: 1.3364 - accuracy: 0.5247 - val_loss: 1.5998 - val_accuracy: 0.4264
Epoch 17/50
391/391 [=====] - 11s 28ms/step - loss: 1.3333 - accuracy: 0.5238 - val_loss: 1.4995 - val_accuracy: 0.4678
Epoch 18/50
391/391 [=====] - 12s 30ms/step - loss: 1.3213 - accuracy: 0.5293 - val_loss: 1.5561 - val_accuracy: 0.4525
Epoch 19/50
391/391 [=====] - 11s 29ms/step - loss: 1.3104 - accuracy: 0.5338 - val_loss: 1.4950 - val_accuracy: 0.4249
Epoch 20/50
391/391 [=====] - 12s 30ms/step - loss: 1.2918 - accuracy: 0.5416 - val_loss: 1.4693 - val_accuracy: 0.5065
Epoch 21/50
391/391 [=====] - 12s 30ms/step - loss: 1.2784 - accuracy: 0.5444 - val_loss: 1.4495 - val_accuracy: 0.4462
Epoch 22/50
391/391 [=====] - 11s 29ms/step - loss: 1.2815 - accuracy: 0.5438 - val_loss: 1.6009 - val_accuracy: 0.4328
Epoch 23/50
391/391 [=====] - 12s 30ms/step - loss: 1.2605 - accuracy: 0.5499 - val_loss: 1.6085 - val_accuracy: 0.4363
Epoch 24/50
391/391 [=====] - 12s 30ms/step - loss: 1.2601 - accuracy: 0.5520 - val_loss: 1.4156 - val_accuracy: 0.5018
Epoch 25/50
391/391 [=====] - 12s 30ms/step - loss: 1.2529 - accuracy: 0.5520 - val_loss: 1.4333 - val_accuracy: 0.4889
Epoch 26/50
391/391 [=====] - 11s 28ms/step - loss: 1.2445 - accuracy: 0.5566 - val_loss: 1.4670 - val_accuracy: 0.4866
Epoch 27/50
391/391 [=====] - 11s 27ms/step - loss: 1.2412 - accuracy: 0.5563 - val_loss: 1.5708 - val_accuracy: 0.4601
Epoch 28/50
391/391 [=====] - 11s 28ms/step - loss: 1.2279 - accuracy: 0.5659 - val_loss: 1.4013 - val_accuracy: 0.5041
Epoch 29/50
391/391 [=====] - 12s 30ms/step - loss: 1.2107 - accuracy: 0.5684 - val_loss: 1.4051 - val_accuracy: 0.4658
Epoch 30/50
391/391 [=====] - 11s 29ms/step - loss: 1.2016 - accuracy: 0.5730 - val_loss: 1.4955 - val_accuracy: 0.4791
Epoch 31/50
391/391 [=====] - 11s 29ms/step - loss: 1.1965 - accuracy: 0.5712 - val_loss: 1.4208 - val_accuracy: 0.4999
Epoch 32/50
391/391 [=====] - 11s 27ms/step - loss: 1.1895 - accuracy: 0.5748 - val_loss: 1.5174 - val_accuracy: 0.4778
Epoch 33/50
391/391 [=====] - 11s 27ms/step - loss: 1.1899 - accuracy: 0.5826 - val_loss: 1.4141 - val_accuracy: 0.5039
Epoch 34/50
391/391 [=====] - 11s 29ms/step - loss: 1.1731 - accuracy: 0.5836 - val_loss: 1.4455 - val_accuracy: 0.4886
Epoch 35/50
391/391 [=====] - 11s 29ms/step - loss: 1.1685 - accuracy: 0.5862 - val_loss: 1.4505 - val_accuracy: 0.5003
Epoch 36/50
391/391 [=====] - 11s 29ms/step - loss: 1.1588 - accuracy: 0.5866 - val_loss: 1.4700 - val_accuracy: 0.5115
Epoch 37/50
391/391 [=====] - 11s 29ms/step - loss: 1.1592 - accuracy: 0.5875 - val_loss: 1.5561 - val_accuracy: 0.5107
Epoch 38/50
391/391 [=====] - 11s 28ms/step - loss: 1.1542 - accuracy: 0.5869 - val_loss: 1.4485 - val_accuracy: 0.4817
Epoch 39/50
391/391 [=====] - 11s 28ms/step - loss: 1.1383 - accuracy: 0.5948 - val_loss: 1.3815 - val_accuracy: 0.5161
Epoch 40/50
391/391 [=====] - 11s 28ms/step - loss: 1.1360 - accuracy: 0.5951 - val_loss: 1.4301 - val_accuracy: 0.4979
Epoch 41/50
391/391 [=====] - 11s 27ms/step - loss: 1.1401 - accuracy: 0.5959 - val_loss: 1.4054 - val_accuracy: 0.5114
Epoch 42/50
391/391 [=====] - 11s 27ms/step - loss: 1.1322 - accuracy: 0.5999 - val_loss: 1.3681 - val_accuracy: 0.5176
Epoch 43/50
391/391 [=====] - 12s 29ms/step - loss: 1.1160 - accuracy: 0.6040 - val_loss: 1.3482 - val_accuracy: 0.5286
Epoch 44/50
391/391 [=====] - 12s 30ms/step - loss: 1.1136 - accuracy: 0.6040 - val_loss: 1.4412 - val_accuracy: 0.4991
Epoch 45/50
391/391 [=====] - 12s 30ms/step - loss: 1.1061 - accuracy: 0.6093 - val_loss: 1.4950 - val_accuracy: 0.5114
Epoch 46/50
391/391 [=====] - 12s 29ms/step - loss: 1.0997 - accuracy: 0.6086 - val_loss: 1.1382 - val_accuracy: 0.5086
Epoch 47/50
391/391 [=====] - 12s 30ms/step - loss: 1.0955 - accuracy: 0.6101 - val_loss: 1.1426 - val_accuracy: 0.5018
Epoch 48/50
391/391 [=====] - 11s 28ms/step - loss: 1.0872 - accuracy: 0.6107 - val_loss: 1.1486 - val_accuracy: 0.4913
Epoch 49/50
391/391 [=====] - 11s 28ms/step - loss: 1.0918 - accuracy: 0.6091 - val_loss: 1.3809 - val_accuracy: 0.5127
Test loss: 1.380945533163452
Test accuracy: 0.5127000212669373
```



Training MLP took 566.1155211925507 seconds

Red convolucional

En la experimentación con la red neuronal hay muchas variables que afectan al desempeño:

- Incluir capa de dropout o no
- Incluir capa de batch normalization antes o después de la activación, o simplemente no incluirla
- Incluir una capa extra o no. Para no perder eficiencia del modelo se comprobó su desempeño con casi todas las combinaciones de configuraciones posibles (se excluyeron aquellas en las que no hay capa de dropout ni capa de batch normalization). Para entender en cada sesión de esta experimentación que configuración tiene el modelo entrenado se incluye en el título de la sección los valores que toman cada una de las 3 variables mencionadas.

droppout=si, batch_normalization=despues, capa_extra=no

```
In [71]: import keras.backend as K
import time

model = build_conv1(dout=True, bn=True, bn_before=True, extra_layer=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

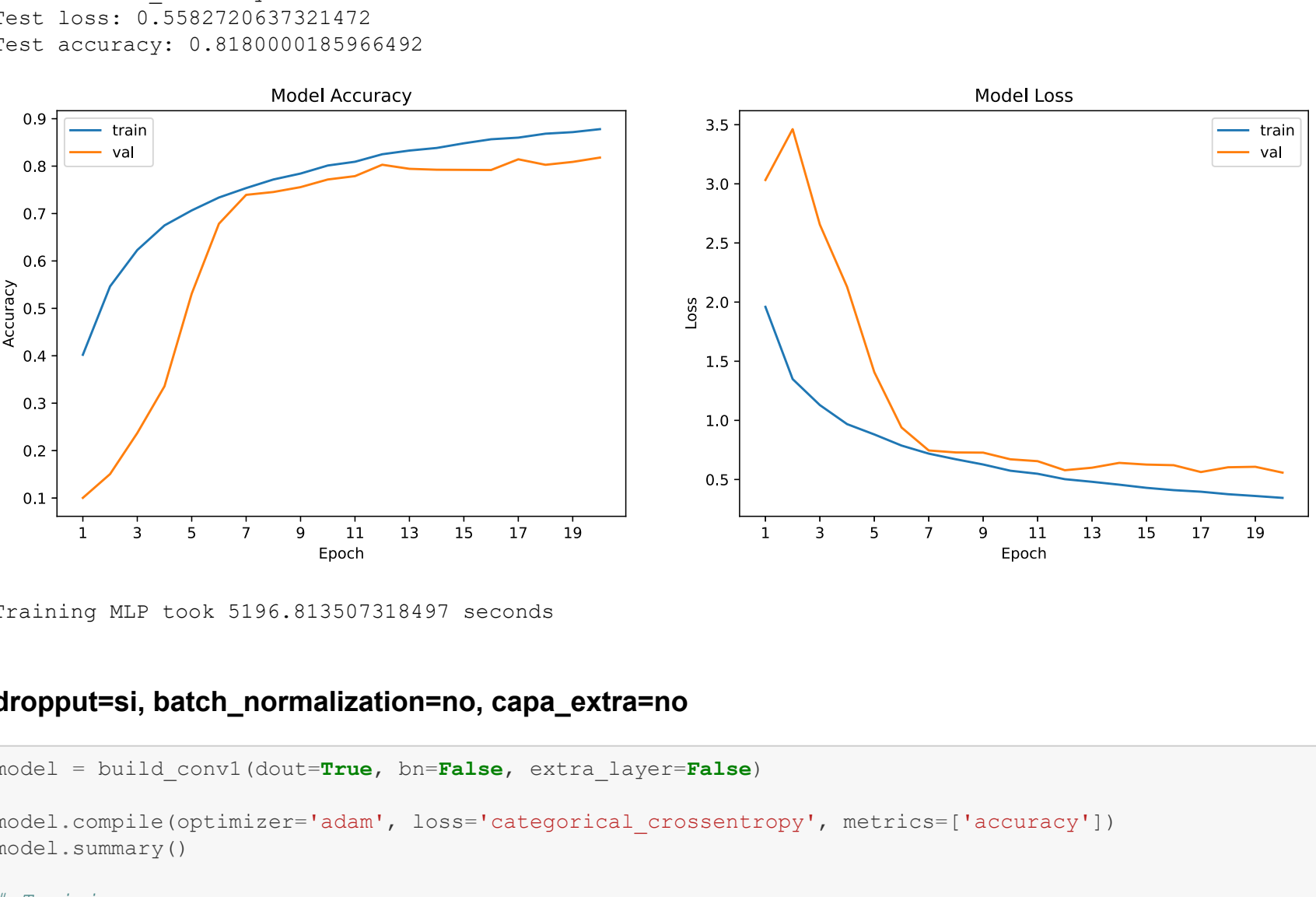
loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_16"

Layer (type) Output Shape Param #
-----
input_24 (InputLayer) [(None, 32, 32, 3)] 0
conv2d_138 (Conv2D) (None, 32, 32, 32) 896
batch_normalization_162 (Bat (None, 32, 32, 32) 128
activation_221 (Activation) (None, 32, 32, 32) 0
conv2d_139 (Conv2D) (None, 32, 32, 32) 9248
batch_normalization_163 (Bat (None, 32, 32, 32) 128
activation_222 (Activation) (None, 32, 32, 32) 0
max_pooling2d_69 (MaxPooling (None, 16, 16, 32) 0
dropout_89 (Dropout) (None, 16, 16, 32) 0
conv2d_140 (Conv2D) (None, 16, 16, 64) 18496
batch_normalization_164 (Bat (None, 16, 16, 64) 256
activation_223 (Activation) (None, 16, 16, 64) 0
conv2d_141 (Conv2D) (None, 16, 16, 64) 36928
batch_normalization_165 (Bat (None, 16, 16, 64) 256
activation_224 (Activation) (None, 16, 16, 64) 0
max_pooling2d_70 (MaxPooling (None, 8, 8, 64) 0
dropout_90 (Dropout) (None, 8, 8, 64) 0
conv2d_142 (Conv2D) (None, 8, 8, 128) 73856
batch_normalization_166 (Bat (None, 8, 8, 128) 512
activation_225 (Activation) (None, 8, 8, 128) 0
conv2d_143 (Conv2D) (None, 8, 8, 128) 147584
batch_normalization_167 (Bat (None, 8, 8, 128) 512
activation_226 (Activation) (None, 8, 8, 128) 0
max_pooling2d_71 (MaxPooling (None, 4, 4, 128) 0
dropout_91 (Dropout) (None, 4, 4, 128) 0
flatten_42 (Flatten) (None, 2048) 0
dense_118 (Dense) (None, 10) 20490
Total params: 309,290
Trainable params: 308,394
Non-trainable params: 896

Epoch 1/20
391/391 [=====] - 200s 2s/step - loss: 2.2669 - accuracy: 0.2967 - val_loss: 3.2269 - val_accuracy: 0.1000
Epoch 2/20
391/391 [=====] - 198s 2s/step - loss: 1.3478 - accuracy: 0.5176 - val_loss: 3.3123 - val_accuracy: 0.1019
Epoch 3/20
391/391 [=====] - 202s 2s/step - loss: 1.1044 - accuracy: 0.6053 - val_loss: 2.1212 - val_accuracy: 0.2130
Epoch 4/20
391/391 [=====] - 202s 2s/step - loss: 0.9317 - accuracy: 0.6689 - val_loss: 0.8420 - val_accuracy: 0.3528
Epoch 5/20
391/391 [=====] - 217s 2s/step - loss: 0.8334 - accuracy: 0.7067 - val_loss: 1.1226 - val_accuracy: 0.6032
Epoch 6/20
391/391 [=====] - 199s 2s/step - loss: 0.7579 - accuracy: 0.7356 - val_loss: 0.8420 - val_accuracy: 0.6987
Epoch 7/20
391/391 [=====] - 221s 2s/step - loss: 0.7091 - accuracy: 0.7491 - val_loss: 0.7804 - val_accuracy: 0.7270
Epoch 8/20
391/391 [=====] - 235s 2s/step - loss: 0.6303 - accuracy: 0.7774 - val_loss: 0.6236 - val_accuracy: 0.7832
Epoch 9/20
391/391 [=====] - 241s 2s/step - loss: 0.5702 - accuracy: 0.7987 - val_loss: 0.7510 - val_accuracy: 0.7440
Epoch 10/20
391/391 [=====] - 236s 2s/step - loss: 0.5464 - accuracy: 0.8071 - val_loss: 0.6423 - val_accuracy: 0.7774
Epoch 11/20
391/391 [=====] - 271s 3s/step - loss: 0.5281 - accuracy: 0.8150 - val_loss: 0.7127 - val_accuracy: 0.7545
Epoch 12/20
391/391 [=====] - 268s 3s/step - loss: 0.4971 - accuracy: 0.8242 - val_loss: 0.7120 - val_accuracy: 0.7544
Epoch 13/20
391/391 [=====] - 280s 3s/step - loss: 0.4763 - accuracy: 0.8326 - val_loss: 0.5715 - val_accuracy: 0.8021
Epoch 14/20
391/391 [=====] - 279s 3s/step - loss: 0.4498 - accuracy: 0.8425 - val_loss: 0.5715 - val_accuracy: 0.8021
Epoch 15/20
391/391 [=====] - 266s 3s/step - loss: 0.4450 - accuracy: 0.8440 - val_loss: 0.5804 - val_accuracy: 0.8035
Epoch 16/20
391/391 [=====] - 263s 3s/step - loss: 0.4218 - accuracy: 0.8517 - val_loss: 0.6617 - val_accuracy: 0.7790
Epoch 17/20
391/391 [=====] - 235s 2s/step - loss: 0.4030 - accuracy: 0.8573 - val_loss: 0.5206 - val_accuracy: 0.8230
Epoch 18/20
391/391 [=====] - 212s 2s/step - loss: 0.3915 - accuracy: 0.8628 - val_loss: 0.5773 - val_accuracy: 0.8064
Epoch 19/20
391/391 [=====] - 251s 3s/step - loss: 0.3756 - accuracy: 0.8710 - val_loss: 0.8900 - val_accuracy: 0.8028
Test loss: 0.580036699718811
Test accuracy: 0.8027999997138977
```



Training MLP took 4681.244248628616 seconds

droppout=si, batch_normalization=antes, capa_extra=no

```
In [72]: model = build_conv1(dout=True, bn=True, bn_before=False, extra_layer=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_17"

Layer (type) Output Shape Param #
-----
input_25 (InputLayer) [(None, 32, 32, 3)] 0
conv2d_144 (Conv2D) (None, 32, 32, 32) 896
activation_227 (Activation) (None, 32, 32, 32) 0
batch_normalization_168 (Bat (None, 32, 32, 32) 128
conv2d_145 (Conv2D) (None, 32, 32, 32) 9248
batch_normalization_169 (Bat (None, 32, 32, 32) 128
max_pooling2d_72 (MaxPooling (None, 16, 16, 32) 0
dropout_92 (Dropout) (None, 16, 16, 32) 0
conv2d_146 (Conv2D) (None, 16, 16, 64) 18496
activation_229 (Activation) (None, 16, 16, 64) 0
batch_normalization_170 (Bat (None, 16, 16, 64) 256
conv2d_147 (Conv2D) (None, 16, 16, 64) 36928
activation_230 (Activation) (None, 16, 16, 64) 0
batch_normalization_171 (Bat (None, 16, 16, 64) 256
conv2d_148 (Conv2D) (None, 8, 8, 64) 0
dropout_93 (Dropout) (None, 8, 8, 64) 0
conv2d_149 (Conv2D) (None, 8, 8, 128) 73856
activation_231 (Activation) (None, 8, 8, 128) 0
batch_normalization_172 (Bat (None, 8, 8, 128) 512
conv2d_149 (Conv2D) (None, 8, 8, 128) 147584
activation_232 (Activation) (None, 8, 8, 128) 0
batch_normalization_173 (Bat (None, 8, 8, 128) 512
max_pooling2d_74 (MaxPooling (None, 4, 4, 128) 0
dropout_94 (Dropout) (None, 4, 4, 128) 0
flatten_43 (Flatten) (None, 2048) 0
dense_119 (Dense) (None, 10) 20490
Total params: 309,290
Trainable params: 308,394
Non-trainable params: 896

Epoch 1/20
391/391 [=====] - 255s 3s/step - loss: 2.4937 - accuracy: 0.3267 - val_loss: 3.4623 - val_accuracy: 0.1004
Epoch 2/20
391/391 [=====] - 235s 2s/step - loss: 1.4170 - accuracy: 0.5266 - val_loss: 3.4623 - val_accuracy: 0.1506
Epoch 3/20
391/391 [=====] - 189s 2s/step - loss: 1.1674 - accuracy: 0.6113 - val_loss: 2.6553 - val_accuracy: 0.2371
Epoch 4/20
391/391 [=====] - 243s 3s/step - loss: 0.9957 - accuracy: 0.6623 - val_loss: 1.2293 - val_accuracy: 0.3358
Epoch 5/20
391/391 [=====] - 252s 3s/step - loss: 0.8984 - accuracy: 0.7006 - val_loss: 1.4080 - val_accuracy: 0.5310
Epoch 6/20
391/391 [=====] - 252s 3s/step - loss: 0.7943 - accuracy: 0.7332 - val_loss: 0.8598 - val_accuracy: 0.6785
Epoch 7/20
391/391 [=====] - 278s 3s/step - loss: 0.7224 - accuracy: 0.7516 - val_loss: 0.7458 - val_accuracy: 0.7394
Epoch 8/20
391/391 [=====] - 288s 3s/step - loss: 0.6597 - accuracy: 0.7767 - val_loss: 0.7289 - val_accuracy: 0.7454
Epoch 9/20
391/391 [=====] - 285s 3s/step - loss: 0.6204 - accuracy: 0.7887 - val_loss: 0.7578 - val_accuracy: 0.7557
Epoch 10/20
391/391 [=====] - 293s 3s/step - loss: 0.5828 - accuracy: 0.7989 - val_loss: 0.7120 - val_accuracy: 0.7718
Epoch 11/20
391/391 [=====] - 276s 3s/step - loss: 0.5412 - accuracy: 0.8112 - val_loss: 0.7260 - val_accuracy: 0.7789
Epoch 12/20
391/391 [=====] - 287s 3s/step - loss: 0.4978 - accuracy: 0.8264 - val_loss: 0.5788 - val_accuracy: 0.8029
Epoch 13/20
391/391 [=====] - 278s 3s/step - loss: 0.4654 - accuracy: 0.8382 - val_loss: 0.6000 - val_accuracy: 0.7943
Epoch 14/20
391/391 [=====] - 280s 3s/step - loss: 0.4471 - accuracy: 0.8407 - val_loss: 0.6407 - val_accuracy: 0.7924
Epoch 15/20
391/391 [=====] - 298s 3s/step - loss: 0.4220 - accuracy: 0.8510 - val_loss: 0.6264 - val_accuracy: 0.7921
Epoch 16/20
391/391 [=====] - 279s 3s/step - loss: 0.3933 - accuracy: 0.8613 - val_loss: 0.6122 - val_accuracy: 0.7918
Epoch 17/20
391/391 [=====] - 217s 2s/step - loss: 0.3859 - accuracy: 0.8627 - val_loss: 0.5633 - val_accuracy: 0.8144
Epoch 18/20
391/391 [=====] - 194s 2s/step - loss: 0.3670 - accuracy: 0.8717 - val_loss: 0.5425 - val_accuracy: 0.8028
Epoch 19/20
391/391 [=====] - 225s 2s/step - loss: 0.3513 - accuracy: 0.8734 - val_loss: 0.5809 - val_accuracy: 0.8089
Epoch 20/20
391/391 [=====] - 292s 3s/step - loss: 0.3290 - accuracy: 0.8823 - val_loss: 0.5893 - val_accuracy: 0.8100
Test loss: 0.5382720637321472
Test accuracy: 0.818000185966492
```



Training MLP took 2428.5012600421906 seconds

droppout=no, batch_normalization=despues, capa_extra=no

```
In [73]: model = build_conv1(dout=True, bn=False, extra_layer=False)

model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_18"

Layer (type) Output Shape Param #
-----
input_26 (InputLayer) [(None, 32, 32, 3)] 0
conv2d_150 (Conv2D) (None, 32, 32, 32) 896
activation_233 (Activation) (None, 32, 32, 32) 0
conv2d_151 (Conv2D) (None, 32, 32, 32) 9248
max_pooling2d_75 (MaxPooling (None, 16, 16, 32) 0
dropout_95 (Dropout) (None, 16, 16, 32) 0
conv2d_152 (Conv2D) (None, 16, 16, 64) 18496
activation_235 (Activation) (None, 16, 16, 64) 0
conv2d_153 (Conv2D) (None, 16, 16, 64) 36928
activation_236 (Activation) (None, 16, 16, 64) 0
max_pooling2d_76 (MaxPooling (None, 8, 8, 64) 0
dropout_96 (Dropout) (None, 8, 8, 64) 0
conv2d_154 (Conv2D) (None, 8, 8, 128) 73856
activation_237 (Activation) (None, 8, 8, 128) 0
conv2d_155 (Conv2D) (None, 8, 8, 128) 147584
activation_238 (Activation) (None, 8, 8, 128) 0
max_pooling2d_77 (MaxPooling (None, 4, 4, 128) 0
dropout_97 (Dropout) (None, 4, 4, 128) 0
flatten_44 (Flatten) (None, 2048) 0
dense_120 (Dense) (None, 10) 20490
Total params: 307,498
Trainable params: 307,498
Non-trainable params: 0

Epoch 1/20
391/391 [=====] - 147s 1s/step - loss: 2.0608 - accuracy: 0.2252 - val_loss: 1.6926 - val_accuracy: 0.3968
Epoch 2/20
391/391 [=====] - 134s 1s/step - loss: 1.5636 - accuracy: 0.4263 - val_loss: 1.6926 - val_accuracy: 0.5592
Epoch 3/20
391/391 [=====] - 113s 1s/step - loss: 1.2286 - accuracy: 0.5622 - val_loss: 1.0183 - val_accuracy: 0.6178
Epoch 4/20
391/391 [=====] - 113s 1s/step - loss: 1.1101 - accuracy: 0.6048 - val_loss: 0.5955 - val_accuracy: 0.6559
Epoch 5/20
391/391 [=====] - 135s 1s/step - loss: 1.0144 - accuracy: 0.6447 - val_loss: 0.9149 - val_accuracy: 0.6789
Epoch 6/20
391/391 [=====] - 131s 1s/step - loss: 0.9321 - accuracy: 0.6761 - val_loss: 0.8488 - val_accuracy: 0.7062
Epoch 7/20
391/391 [=====] - 132s 1s/step - loss: 0.8879 - accuracy: 0.6923 - val_loss: 0.8342 - val_accuracy: 0.7048
Epoch 8/20
391/391 [=====] - 125s 1s/step - loss: 0.8301 - accuracy: 0.7095 - val_loss: 0.7961 - val_accuracy: 0.7243
Epoch 9/20
391/391 [=====] - 122s 1s/step - loss: 0.7804 - accuracy: 0.7271 - val_loss: 0.7893 - val_accuracy: 0.7259
Epoch 10/20
391/391 [=====] - 96s 983ms/step - loss: 0.7427 - accuracy: 0.7417 - val_loss: 0.7893 - val_accuracy: 0.7403
Epoch 11/20
391/391 [=====] - 115s 1s/step - loss: 0.7080 - accuracy: 0.7497 - val_loss: 0.7425 - val_accuracy: 0.7562
Epoch 12/20
391/391 [=====] - 126s 1s/step - loss: 0.6706 - accuracy: 0.7661 - val_loss: 0.7842 - val_accuracy: 0.7638
Epoch 13/20
391/391 [=====] - 122s 1s/step - loss: 0.6425 - accuracy: 0.7736 - val_loss: 0.6924 - val_accuracy: 0.7587
Epoch 14/20
391/391 [=====] - 126s 1s/step - loss: 0.6227 - accuracy: 0.7818 - val_loss: 0.7030 - val_accuracy: 0.7587
Epoch 15/20
391/391 [=====] - 98s 998ms/step - loss: 0.5937 - accuracy: 0.7892 - val_loss: 0.6607 - val_accuracy: 0.7551
Epoch 16/20
391/391 [=====] - 106s 1s/step - loss: 0.5715 - accuracy: 0.8012 - val_loss: 0.6319 - val_accuracy: 0.7816
Epoch 17/20
391/391 [=====] - 122s 1s/step - loss: 0.5447 - accuracy: 0.8093 - val_loss: 0.6191 - val_accuracy: 0.7846
Epoch 18/20
391/391 [=====] - 124s 1s/step - loss: 0.5357 - accuracy: 0.8124 - val_loss: 0.6191 - val_accuracy: 0.7828
Epoch 19/20
391/391 [=====] - 123s 1s/step - loss: 0.5237 - accuracy: 0.8179 - val_loss: 0.6191 - val_accuracy: 0.7828
Epoch 20/20
391/391 [=====] - 123s 1s/step - loss: 0.5237 - accuracy: 0.8179 - val_loss: 0.6191 - val_accuracy: 0.7828
Test loss: 0.592851996421
```


In [74]:

```
model = build_conv1(dout=False, bn=True, bn_before=True, extra_layer=False)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_19"
```

Layer (type)	Output Shape	Param #
input_27 (InputLayer)	[(None, 32, 32, 3)]	0
conv2d_156 (Conv2D)	(None, 32, 32, 32)	896
batch_normalization_174 (Bat	(None, 32, 32, 32)	128
activation_239 (Activation)	(None, 32, 32, 32)	0
conv2d_157 (Conv2D)	(None, 32, 32, 32)	9248
batch_normalization_175 (Bat	(None, 32, 32, 32)	128
activation_240 (Activation)	(None, 32, 32, 32)	0
max_pooling2d_78 (MaxPooling	(None, 16, 16, 32)	0
conv2d_158 (Conv2D)	(None, 16, 16, 64)	18496
batch_normalization_176 (Bat	(None, 16, 16, 64)	256
activation_241 (Activation)	(None, 16, 16, 64)	0
conv2d_159 (Conv2D)	(None, 16, 16, 64)	36928
batch_normalization_179 (Bat	(None, 16, 16, 64)	256
activation_242 (Activation)	(None, 16, 16, 64)	0
max_pooling2d_79 (MaxPooling	(None, 8, 8, 64)	0
conv2d_160 (Conv2D)	(None, 8, 8, 128)	73856
batch_normalization_178 (Bat	(None, 8, 8, 128)	512
activation_243 (Activation)	(None, 8, 8, 128)	0
conv2d_161 (Conv2D)	(None, 8, 8, 128)	147584
batch_normalization_179 (Bat	(None, 8, 8, 128)	512
activation_244 (Activation)	(None, 8, 8, 128)	0
max_pooling2d_80 (MaxPooling	(None, 4, 4, 128)	0
flatten_45 (Flatten)	(None, 2048)	0
dense_121 (Dense)	(None, 10)	20490
Total params: 309,290		
Trainable params: 308,394		
Non-trainable params: 896		
Epoch 1/20		
98/98 [=====]	- 221s 2s/step - loss: 1.8586 - accuracy: 0.3896 - val_loss: 2.9792 - val_accuracy: 0.1191	
Epoch 2/20	98/98 [=====]	- 208s 2s/step - loss: 0.9776 - accuracy: 0.6560 - val_loss: 3.2510 - val_accuracy: 0.1434
Epoch 3/20	98/98 [=====]	- 245s 3s/step - loss: 0.7136 - accuracy: 0.7515 - val_loss: 2.4966 - val_accuracy: 0.2265
Epoch 4/20	98/98 [=====]	- 198s 2s/step - loss: 0.5858 - accuracy: 0.7988 - val_loss: 1.6289 - val_accuracy: 0.4418
Epoch 5/20	98/98 [=====]	- 252s 3s/step - loss: 0.4734 - accuracy: 0.8427 - val_loss: 1.3341 - val_accuracy: 0.5786
Epoch 6/20	98/98 [=====]	- 248s 3s/step - loss: 0.3810 - accuracy: 0.8733 - val_loss: 0.4897 - val_accuracy: 0.7135
Epoch 7/20	98/98 [=====]	- 258s 3s/step - loss: 0.3097 - accuracy: 0.9031 - val_loss: 0.7926 - val_accuracy: 0.7167
Epoch 8/20	98/98 [=====]	- 212s 2s/step - loss: 0.2379 - accuracy: 0.9280 - val_loss: 0.7981 - val_accuracy: 0.7533
Epoch 9/20	98/98 [=====]	- 210s 2s/step - loss: 0.1753 - accuracy: 0.9532 - val_loss: 0.9216 - val_accuracy: 0.7326
Epoch 10/20	98/98 [=====]	- 194s 2s/step - loss: 0.1350 - accuracy: 0.9666 - val_loss: 0.9641 - val_accuracy: 0.7307
Epoch 11/20	98/98 [=====]	- 195s 2s/step - loss: 0.1034 - accuracy: 0.9783 - val_loss: 0.8441 - val_accuracy: 0.7437
Epoch 12/20	98/98 [=====]	- 245s 3s/step - loss: 0.0654 - accuracy: 0.9903 - val_loss: 0.8028 - val_accuracy: 0.7687
Epoch 13/20	98/98 [=====]	- 248s 3s/step - loss: 0.0443 - accuracy: 0.9951 - val_loss: 0.7981 - val_accuracy: 0.7681
Epoch 14/20	98/98 [=====]	- 243s 2s/step - loss: 0.0269 - accuracy: 0.9985 - val_loss: 0.8990 - val_accuracy: 0.7570
Epoch 15/20	98/98 [=====]	- 241s 2s/step - loss: 0.0147 - accuracy: 1.0000 - val_loss: 0.7981 - val_accuracy: 0.7939
Epoch 16/20	98/98 [=====]	- 247s 3s/step - loss: 0.0084 - accuracy: 1.0000 - val_loss: 0.7981 - val_accuracy: 0.7919
Epoch 17/20	98/98 [=====]	- 263s 3s/step - loss: 0.0057 - accuracy: 1.0000 - val_loss: 0.7981 - val_accuracy: 0.7976
Epoch 18/20	98/98 [=====]	- 243s 2s/step - loss: 0.0047 - accuracy: 1.0000 - val_loss: 0.7981 - val_accuracy: 0.8004
Epoch 19/20	98/98 [=====]	- 238s 2s/step - loss: 0.0039 - accuracy: 1.0000 - val_loss: 0.7981 - val_accuracy: 0.8004
Epoch 20/20	98/98 [=====]	- 229s 2s/step - loss: 0.0033 - accuracy: 1.0000 - val_loss: 0.8019 - val_accuracy: 0.7982
Test loss: 0.80192625261353		
Test accuracy: 0.798200011233569		

dropputno, batch_normalization=antes, capa_extra=no

In [75]:

```
model = build_conv1(dout=False, bn=True, bn_before=False, extra_layer=False)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

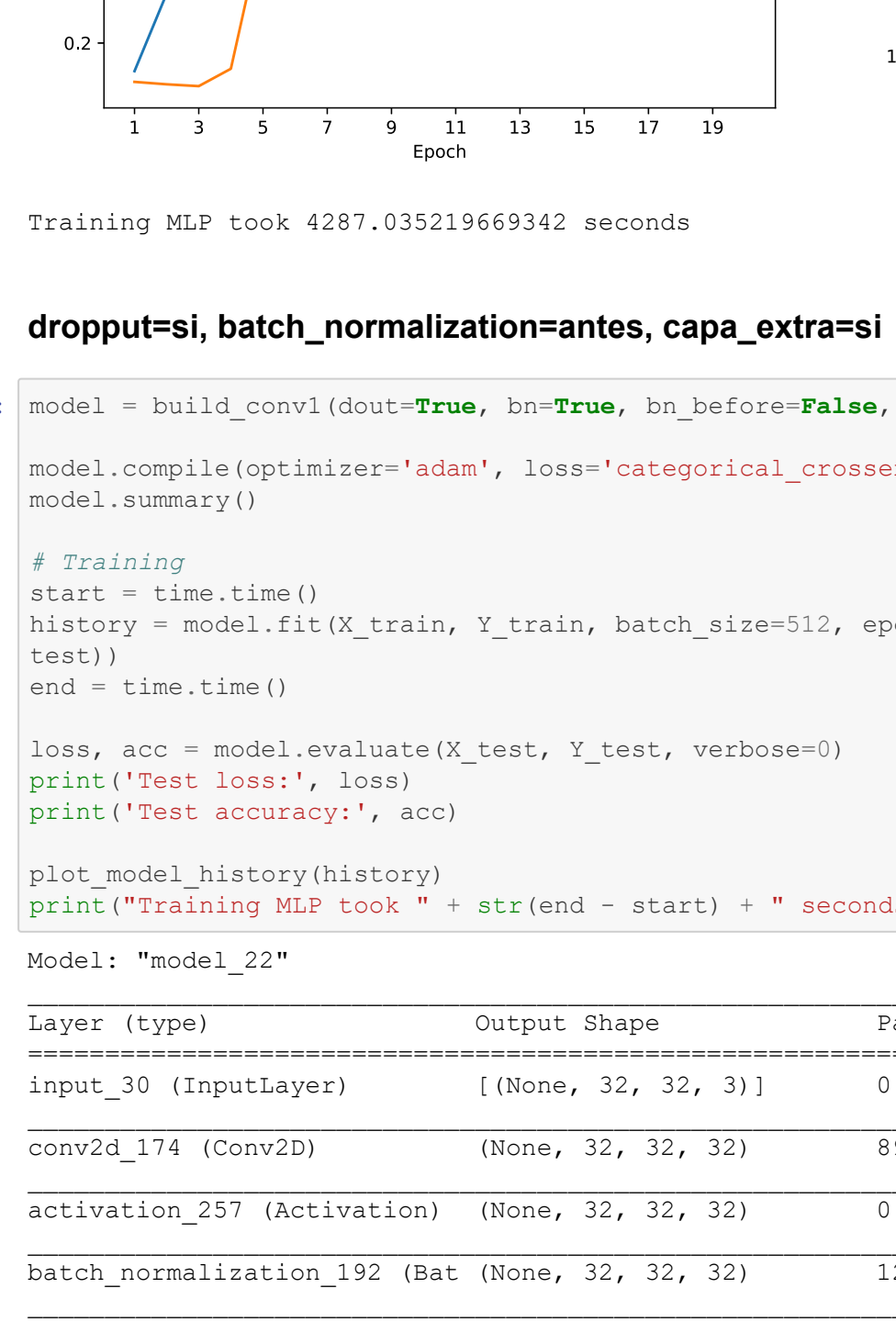
# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_20"
```

Layer (type)	Output Shape	Param #
input_28 (InputLayer)	[(None, 32, 32, 3)]	0
conv2d_162 (Conv2D)	(None, 32, 32, 32)	896
activation_245 (Activation)	(None, 32, 32, 32)	0
batch_normalization_180 (Bat	(None, 32, 32, 32)	128
conv2d_163 (Conv2D)	(None, 32, 32, 32)	9248
activation_246 (Activation)	(None, 32, 32, 32)	0
batch_normalization_181 (Bat	(None, 32, 32, 32)	128
max_pooling2d_81 (MaxPooling	(None, 16, 16, 32)	0
conv2d_164 (Conv2D)	(None, 16, 16, 64)	18496
activation_247 (Activation)	(None, 16, 16, 64)	0
batch_normalization_182 (Bat	(None, 16, 16, 64)	256
conv2d_165 (Conv2D)	(None, 16, 16, 64)	36928
activation_248 (Activation)	(None, 16, 16, 64)	0
batch_normalization_183 (Bat	(None, 16, 16, 64)	256
max_pooling2d_82 (MaxPooling	(None, 8, 8, 64)	0
conv2d_166 (Conv2D)	(None, 8, 8, 128)	73856
activation_249 (Activation)	(None, 8, 8, 128)	0
batch_normalization_184 (Bat	(None, 8, 8, 128)	512
conv2d_167 (Conv2D)	(None, 8, 8, 128)	147584
activation_250 (Activation)	(None, 8, 8, 128)	0
batch_normalization_185 (Bat	(None, 8, 8, 128)	512
max_pooling2d_83 (MaxPooling	(None, 4, 4, 128)	0
flatten_46 (Flatten)	(None, 2048)	0
dense_122 (Dense)	(None, 10)	20490
Total params: 309,290		
Trainable params: 308,394		
Non-trainable params: 896		
Epoch 1/20	98/98 [=====]	- 245s 2s/step - loss: 1.8790 - accuracy: 0.4127 - val_loss: 2.8086 - val_accuracy: 0.1024
Epoch 2/20	98/98 [=====]	- 236s 2s/step - loss: 0.9649 - accuracy: 0.6610 - val_loss: 3.2265 - val_accuracy: 0.1224
Epoch 3/20	98/98 [=====]	- 243s 2s/step - loss: 0.7243 - accuracy: 0.7515 - val_loss: 1.9383 - val_accuracy: 0.3648
Epoch 4/20	98/98 [=====]	- 240s 2s/step - loss: 0.5609 - accuracy: 0.8097 - val_loss: 1.5081 - val_accuracy: 0.4625
Epoch 5/20	98/98 [=====]	- 236s 2s/step - loss: 0.4444 - accuracy: 0.8482 - val_loss: 1.0935 - val_accuracy: 0.6285
Epoch 6/20	98/98 [=====]	- 235s 2s/step - loss: 0.3418 - accuracy: 0.8867 - val_loss: 0.7981 - val_accuracy: 0.6906
Epoch 7/20	98/98 [=====]	- 241s 2s/step - loss: 0.2485 - accuracy: 0.9182 - val_loss: 0.7981 - val_accuracy: 0.7090
Epoch 8/20	98/98 [=====]	- 245s 2s/step - loss: 0.1726 - accuracy: 0.9493 - val_loss: 0.7981 - val_accuracy: 0.7263
Epoch 9/20	98/98 [=====]	- 237s 2s/step - loss: 0.1228 - accuracy: 0.9650 - val_loss: 0.9338 - val_accuracy: 0.7430
Epoch 10/20	98/98 [=====]	- 241s 2s/step - loss: 0.0869 - accuracy: 0.9758 - val_loss: 0.9674 - val_accuracy: 0.7448
Epoch 11/20	98/98 [=====]	- 236s 2s/step - loss: 0.0662 - accuracy: 0.9818 - val_loss: 1.0267 - val_accuracy: 0.7423
Epoch 12/20	98/98 [=====]	- 238s 2s/step - loss: 0.0550 - accuracy: 0.9846 - val_loss: 1.0928 - val_accuracy: 0.7366
Epoch 13/20	98/98 [=====]	- 240s 2s/step - loss: 0.0417 - accuracy: 0.9901 - val_loss: 1.1042 - val_accuracy: 0.7522
Epoch 14/20	98/98 [=====]	- 243s 2s/step - loss: 0.0337 - accuracy: 0.9917 - val_loss: 1.1785 - val_accuracy: 0.7419
Epoch 15/20	98/98 [=====]	- 250s 3s/step - loss: 0.0313 - accuracy: 0.9921 - val_loss: 1.1365 - val_accuracy: 0.7465
Epoch 16/20	98/98 [=====]	- 252s 3s/step - loss: 0.0367 - accuracy: 0.9898 - val_loss: 0.7981 - val_accuracy: 0.7317
Epoch 17/20	98/98 [=====]	- 242s 2s/step - loss: 0.0600 - accuracy: 0.9805 - val_loss: 0.7981 - val_accuracy: 0.7099
Epoch 18/20	98/98 [=====]	- 246s 3s/step - loss: 0.1466 - accuracy: 0.9478 - val_loss: 1.5094 - val_accuracy: 0.7137
Epoch 19/20	98/98 [=====]	- 235s 2s/step - loss: 0.1208 - accuracy: 0.9562 - val_loss: 1.4203 - val_accuracy: 0.7233
Epoch 20/20	98/98 [=====]	- 239s 2s/step - loss: 0.0669 - accuracy: 0.9764 - val_loss: 1.1963 - val_accuracy: 0.7581
Test loss: 1.196294188494507		
Test accuracy: 0.758098732017517		



Training MLP took 4822.8211834430695 seconds

dropputsi, batch_normalization=despues, capa_extra=si

In [76]:

```
model = build_conv1(dout=True, bn=True, bn_before=True, extra_layer=True)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_21"
```

Layer (type)	Output Shape	Param #
input_29 (InputLayer)	[(None, 32, 32, 3)]	0
conv2d_168 (Conv2D)	(None, 32, 32, 32)	896
batch_normalization_186 (Bat	(None, 32, 32, 32)	128
activation_251 (Activation)	(None, 32, 32, 32)	0
conv2d_169 (Conv2D)	(None, 32, 32, 32)	9248
batch_normalization_187 (Bat	(None, 32, 32, 32)	128
activation_252 (Activation)	(None, 32, 32, 32)	0
max_pooling2d_84 (MaxPooling	(None, 16, 16, 32)	0
dropout_98 (Dropout)	(None, 16, 16, 32)	0
conv2d_170 (Conv2D)	(None, 16, 16, 64)	18496
batch_normalization_188 (Bat	(None, 16, 16, 64)	256
activation_253 (Activation)	(None, 16, 16, 64)	0
conv2d_171 (Conv2D)	(None, 16, 16, 64)	36928
batch_normalization_189 (Bat	(None, 16, 16, 64)	256
activation_254 (Activation)	(None, 16, 16, 64)	0
max_pooling2d_85 (MaxPooling	(None, 8, 8, 64)	0
dropout_99 (Dropout)	(None, 8, 8, 64)	0
conv2d_172 (Conv2D)	(None, 8, 8, 128)	73856
batch_normalization_190 (Bat	(None, 8, 8, 128)	512
activation_255 (Activation)	(None, 8, 8, 128)	0
conv2d_173 (Conv2D)	(None, 8, 8, 128)	147584
batch_normalization_191 (Bat	(None, 8, 8, 128)	512
activation_256 (Activation)	(None, 8, 8, 128)	0
max_pooling2d_86 (MaxPooling	(None, 4, 4, 128)	0
dropout_100 (Dropout)	(None, 4, 4, 128)	0
flatten_47 (Flatten)	(None, 2048)	0
dense_123 (Dense)	(None, 2048)	4196352
dense_124 (Dense)	(None, 10)	20490
Total params: 4,505,642		
Trainable params: 4,504,746		
Non-trainable params: 896		
Epoch 1/20	98/98 [=====]	- 273s 3s/step - loss: 13.6607 - accuracy: 0.1343 - val_loss: 2.1091 - val_accuracy: 0.1408
Epoch 2/20	98/98 [=====]	- 261s 3s/step - loss: 2.0823 - accuracy: 0.2440 - val_loss: 2.7084 - val_accuracy: 0.1371
Epoch 3/20	98/98 [=====]	- 270s 3s/step - loss: 1.7005 - accuracy: 0.3808 - val_loss: 3.8148 - val_accuracy: 0.1343
Epoch 4/20	98/98 [=====]	- 274s 3s/step - loss: 1.5201 - accuracy: 0.4470 - val_loss: 3.3705 - val_accuracy: 0.1608
Epoch 5/20	98/98 [=====]	- 265s 3s/step - loss: 1.3712 - accuracy: 0.5041 - val_loss: 3.0748 - val_accuracy: 0.3803
Epoch 6/20	98/98 [=====]	- 259s 3s/step - loss: 1.2398 - accuracy: 0.5528 - val_loss: 1.8380 - val_accuracy: 0.3897
Epoch 7/20	98/98 [=====]	- 257s 3s/step - loss: 1.1107 - accuracy: 0.6027 - val_loss: 1.8236 - val_accuracy: 0.4052
Epoch 8/20	98/98 [=====]	- 282s 3s/step - loss: 1.0239 - accuracy: 0.6383 - val_loss: 1.4599 - val_accuracy: 0.5443
Epoch 9/20	98/98 [=====]	- 272s 3s/step - loss: 0.9458 - accuracy: 0.6649 - val_loss: 0.7981 - val_accuracy: 0.6523
Epoch 10/20	98/98 [=====]	- 192s 2s/step - loss: 0.8886 - accuracy: 0.6879 - val_loss: 0.7981 - val_accuracy: 0.5987
Epoch 11/20	98/98 [=====]	- 168s 2s/step - loss: 0.8295 - accuracy: 0.7051 - val_loss: 0.8470 - val_accuracy: 0.7031
Epoch 12/20	98/98 [=====]	- 171s 2s/step - loss: 0.8065 - accuracy: 0.7172 - val_loss: 0.9333 - val_accuracy: 0.6670
Epoch 13/20	98/98 [=====]	- 180s 2s/step - loss: 0.7627 - accuracy: 0.7294 - val_loss: 0.8081 - val_accuracy: 0.7158
Epoch 14/20	98/98 [=====]	- 176s 2s/step - loss: 0.7176 - accuracy: 0.7498 - val_loss: 0.8537 - val_accuracy: 0.7009
Epoch 15/20	98/98 [=====]	- 178s 2s/step - loss: 0.6719 - accuracy: 0.7645 - val_loss: 0.7981 - val_accuracy: 0.7202
Epoch 16/20	98/98 [=====]	- 162s 2s/step - loss: 0.6642 - accuracy: 0.7685 - val_loss: 0.7981 - val_accuracy: 0.7241
Epoch 17/20	98/98 [=====]	- 162s 2s/step - loss: 0.6221 - accuracy: 0.7817 - val_loss: 0.7981 - val_accuracy: 0.7351
Epoch 18/20	98/98 [=====]	- 161s 2s/step - loss: 0.6209 - accuracy: 0.7811 - val_loss: 0.7981 - val_accuracy: 0.7104
Epoch 19/20	98/98 [=====]	- 162s 2s/step - loss: 0.5965 - accuracy: 0.7901 - val_loss: 0.7981 - val_accuracy: 0.7538
Epoch 20/20	98/98 [=====]	- 160s 2s/step - loss: 0.5690 - accuracy: 0.7971 - val_loss: 0.7555 - val_accuracy: 0.7380
Test loss: 0.754526136450195		
Test accuracy: 0.737999756813049		



Training MLP took 4287.035219669342 seconds

dropput=si, batch_normalization=antes, capa_extra=si

In [77]:

```
model = build_conv1(dout=True, bn=True, bn_before=False, extra_layer=True)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)

plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_22"
```

Layer (type)	Output Shape	Param #
input_30 (InputLayer)	[(None, 32, 32, 3)]	0
conv2d_174 (Conv2D)	(None, 32, 32, 32)	896
activation_257 (Activation)	(None, 32, 32, 32)	0
batch_normalization_192 (Bat	(None, 32, 32, 32)	128
conv2d_175 (Conv2D)	(None, 32, 32, 32)	9248
activation_258 (Activation)	(None, 32, 32, 32)	0
batch_normalization_193 (Bat	(None, 32, 32, 32)	128
max_pooling2d_87 (MaxPooling	(None, 16, 16, 32)	0
dropout_101 (Dropout)	(None, 16, 16, 32)	0
conv2d_176 (Conv2D)	(None, 16, 16, 64)	18496
activation_259 (Activation)	(None, 16, 16, 64)	0
conv2d_177 (Conv2D)	(None, 16, 16, 64)	36928
activation_260 (Activation)	(None, 16, 16, 64)	0
batch_normalization_195 (Bat	(None, 16, 16, 64)	256
max_pooling2d_88 (MaxPooling	(None, 8, 8, 64)	0
dropout_102 (Dropout)	(None, 8, 8, 64)	0
conv2d_178 (Conv2D)	(None, 8, 8, 128)	73856
activation_261 (Activation)	(None, 8, 8, 128)	0
batch_normalization_196 (Bat	(None, 8, 8, 128)	512
conv2d_179 (Conv2D)	(None, 8, 8, 128)	147584
activation_262 (Activation)	(None, 8, 8, 128)	0
batch_normalization_197 (Bat	(None, 8, 8, 128)	512
max_pooling2d_89 (MaxPooling	(None, 4, 4, 128)	0
dropout_103 (Dropout)	(None, 4, 4, 128)	0
flatten_48 (Flatten)	(None, 2048)	0
dense_125 (Dense)	(None, 2048)	4196352
dense_126 (Dense)	(None, 10)	20490
Total params: 4,505,642		
Trainable params: 4,504,746		
Non-trainable params: 896		
Epoch 1/20	98/98 [=====]	- 80s 813ms/step - loss: 10.4629 - accuracy: 0.2396 - val_loss: 2.3159 - val_accuracy: 0.1082
Epoch 2/20		


```
[79]: model = build_conv1(dout=False, bn=True, bn_before=True, extra_layer=True)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_24"

Layer (type) Output Shape Param #
-----
input_32 (InputLayer) [(None, 32, 32, 3)] 0
conv2d_186 (Conv2D) (None, 32, 32, 32) 896
batch_normalization_198 (Batch Normalization) (None, 32, 32, 32) 128
activation_269 (Activation) (None, 32, 32, 32) 0
conv2d_187 (Conv2D) (None, 32, 32, 32) 9248
batch_normalization_199 (Batch Normalization) (None, 32, 32, 32) 128
activation_270 (Activation) (None, 32, 32, 32) 0
max_pooling2d_93 (Max Pooling) (None, 16, 16, 32) 0
conv2d_188 (Conv2D) (None, 16, 16, 64) 18496
batch_normalization_200 (Batch Normalization) (None, 16, 16, 64) 256
activation_271 (Activation) (None, 16, 16, 64) 0
conv2d_189 (Conv2D) (None, 16, 16, 64) 36928
batch_normalization_201 (Batch Normalization) (None, 16, 16, 64) 256
activation_272 (Activation) (None, 16, 16, 64) 0
max_pooling2d_94 (Max Pooling) (None, 8, 8, 128) 0
conv2d_190 (Conv2D) (None, 8, 8, 128) 73856
batch_normalization_202 (Batch Normalization) (None, 8, 8, 128) 512
activation_273 (Activation) (None, 8, 8, 128) 0
conv2d_191 (Conv2D) (None, 8, 8, 128) 147584
batch_normalization_203 (Batch Normalization) (None, 8, 8, 128) 512
activation_274 (Activation) (None, 8, 8, 128) 0
max_pooling2d_95 (Max Pooling) (None, 4, 4, 128) 0
flatten_50 (Flatten) (None, 2048) 0
dense_129 (Dense) (None, 2048) 4196352
dense_130 (Dense) (None, 10) 20490
Total params: 4,505,642
Trainable params: 4,504,746
Non-trainable params: 896

Epoch 1/20
98/98 [=====] - 153s 2s/step - loss: 11.6732 - accuracy: 0.1837 - val_loss: 2.7736 - val_accuracy: 0.1015
Epoch 2/20
98/98 [=====] - 151s 2s/step - loss: 1.6263 - accuracy: 0.4026 - val_loss: 2.7520 - val_accuracy: 0.1241
Epoch 3/20
98/98 [=====] - 151s 2s/step - loss: 1.3786 - accuracy: 0.4932 - val_loss: 2.7520 - val_accuracy: 0.1110
Epoch 4/20
98/98 [=====] - 152s 2s/step - loss: 1.2295 - accuracy: 0.5521 - val_loss: 1.6728 - val_accuracy: 0.3672
Epoch 5/20
98/98 [=====] - 151s 2s/step - loss: 1.0704 - accuracy: 0.6176 - val_loss: 1.4516 - val_accuracy: 0.4111
Epoch 6/20
98/98 [=====] - 151s 2s/step - loss: 0.9543 - accuracy: 0.6568 - val_loss: 1.4516 - val_accuracy: 0.5076
Epoch 7/20
98/98 [=====] - 151s 2s/step - loss: 0.8655 - accuracy: 0.6946 - val_loss: 1.2261 - val_accuracy: 0.5837
Epoch 8/20
98/98 [=====] - 152s 2s/step - loss: 0.7584 - accuracy: 0.7312 - val_loss: 1.0182 - val_accuracy: 0.6628
Epoch 9/20
98/98 [=====] - 151s 2s/step - loss: 0.6803 - accuracy: 0.7604 - val_loss: 0.8593 - val_accuracy: 0.6901
Epoch 10/20
98/98 [=====] - 151s 2s/step - loss: 0.6333 - accuracy: 0.7780 - val_loss: 0.9634 - val_accuracy: 0.6783
Epoch 11/20
98/98 [=====] - 151s 2s/step - loss: 0.5610 - accuracy: 0.8023 - val_loss: 0.8875 - val_accuracy: 0.7015
Epoch 12/20
98/98 [=====] - 151s 2s/step - loss: 0.5118 - accuracy: 0.8199 - val_loss: 0.9123 - val_accuracy: 0.6802
Epoch 13/20
98/98 [=====] - 151s 2s/step - loss: 0.4629 - accuracy: 0.8366 - val_loss: 0.8995 - val_accuracy: 0.7119
Epoch 14/20
98/98 [=====] - 151s 2s/step - loss: 0.4090 - accuracy: 0.8558 - val_loss: 0.9193 - val_accuracy: 0.7224
Epoch 15/20
98/98 [=====] - 151s 2s/step - loss: 0.3783 - accuracy: 0.8669 - val_loss: 0.9170 - val_accuracy: 0.6887
Epoch 16/20
98/98 [=====] - 153s 2s/step - loss: 0.3286 - accuracy: 0.8822 - val_loss: 1.3201 - val_accuracy: 0.6567
Epoch 17/20
98/98 [=====] - 151s 2s/step - loss: 0.2712 - accuracy: 0.9029 - val_loss: 1.1920 - val_accuracy: 0.7464
Epoch 18/20
98/98 [=====] - 151s 2s/step - loss: 0.2691 - accuracy: 0.9043 - val_loss: 1.2469 - val_accuracy: 0.6970
Epoch 19/20
98/98 [=====] - 151s 2s/step - loss: 0.2079 - accuracy: 0.9255 - val_loss: 0.9756 - val_accuracy: 0.7471
Epoch 20/20
98/98 [=====] - 152s 2s/step - loss: 0.1505 - accuracy: 0.9472 - val_loss: 1.0386 - val_accuracy: 0.7528
Test loss: 1.0385514497756958
Test accuracy: 0.7527999877929688
```

Training MLP took 3025.1649990081787 seconds

dropput=no, batch_normalization=antes, capa_extra=si

```
In [80]: model = build_conv1(dout=False, bn=True, bn_before=False, extra_layer=True)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.summary()

# Training
start = time.time()
history = model.fit(X_train, Y_train, batch_size=512, epochs=20, verbose=1, validation_data=(X_test, Y_test))
end = time.time()

loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print("Test loss:", loss)
print("Test accuracy:", acc)
plot_model_history(history)
print("Training MLP took " + str(end - start) + " seconds")

Model: "model_25"

Layer (type) Output Shape Param #
-----
input_33 (InputLayer) [(None, 32, 32, 3)] 0
conv2d_192 (Conv2D) (None, 32, 32, 32) 896
activation_275 (Activation) (None, 32, 32, 32) 0
batch_normalization_204 (Batch Normalization) (None, 32, 32, 32) 128
conv2d_193 (Conv2D) (None, 32, 32, 32) 9248
activation_276 (Activation) (None, 32, 32, 32) 0
batch_normalization_205 (Batch Normalization) (None, 32, 32, 32) 128
max_pooling2d_96 (Max Pooling) (None, 16, 16, 32) 0
conv2d_194 (Conv2D) (None, 16, 16, 64) 18496
activation_277 (Activation) (None, 16, 16, 64) 0
batch_normalization_206 (Batch Normalization) (None, 16, 16, 64) 256
conv2d_195 (Conv2D) (None, 16, 16, 64) 36928
activation_278 (Activation) (None, 16, 16, 64) 0
batch_normalization_207 (Batch Normalization) (None, 16, 16, 64) 256
max_pooling2d_97 (Max Pooling) (None, 8, 8, 64) 0
conv2d_196 (Conv2D) (None, 8, 8, 128) 73856
activation_279 (Activation) (None, 8, 8, 128) 0
batch_normalization_208 (Batch Normalization) (None, 8, 8, 128) 512
conv2d_197 (Conv2D) (None, 8, 8, 128) 147584
activation_280 (Activation) (None, 8, 8, 128) 0
batch_normalization_209 (Batch Normalization) (None, 8, 8, 128) 512
max_pooling2d_98 (Max Pooling) (None, 4, 4, 128) 0
flatten_51 (Flatten) (None, 2048) 0
dense_131 (Dense) (None, 2048) 4196352
dense_132 (Dense) (None, 10) 20490
Total params: 4,505,642
Trainable params: 4,504,746
Non-trainable params: 896

Epoch 1/20
98/98 [=====] - 147s 1s/step - loss: 7.9319 - accuracy: 0.2507 - val_loss: 4.2513 - val_accuracy: 0.1004
Epoch 2/20
98/98 [=====] - 146s 1s/step - loss: 2.0367 - accuracy: 0.4284 - val_loss: 3.3177 - val_accuracy: 0.1027
Epoch 3/20
98/98 [=====] - 146s 1s/step - loss: 1.8096 - accuracy: 0.5220 - val_loss: 2.6041 - val_accuracy: 0.2519
Epoch 4/20
98/98 [=====] - 147s 1s/step - loss: 1.4932 - accuracy: 0.5818 - val_loss: 1.9135 - val_accuracy: 0.3423
Epoch 5/20
98/98 [=====] - 147s 1s/step - loss: 1.1822 - accuracy: 0.6403 - val_loss: 1.4302 - val_accuracy: 0.4954
Epoch 6/20
98/98 [=====] - 147s 1s/step - loss: 0.9767 - accuracy: 0.6797 - val_loss: 1.0093 - val_accuracy: 0.6437
Epoch 7/20
98/98 [=====] - 146s 1s/step - loss: 0.8111 - accuracy: 0.7254 - val_loss: 0.9738 - val_accuracy: 0.6536
Epoch 8/20
98/98 [=====] - 148s 2s/step - loss: 0.6800 - accuracy: 0.7660 - val_loss: 0.8995 - val_accuracy: 0.6822
Epoch 9/20
98/98 [=====] - 147s 1s/step - loss: 0.5964 - accuracy: 0.7966 - val_loss: 0.8193 - val_accuracy: 0.7203
Epoch 10/20
98/98 [=====] - 146s 1s/step - loss: 0.5053 - accuracy: 0.8236 - val_loss: 0.9613 - val_accuracy: 0.6865
Epoch 11/20
98/98 [=====] - 146s 1s/step - loss: 0.4186 - accuracy: 0.8536 - val_loss: 0.8573 - val_accuracy: 0.7262
Epoch 12/20
98/98 [=====] - 147s 2s/step - loss: 0.3336 - accuracy: 0.8825 - val_loss: 0.9650 - val_accuracy: 0.7152
Epoch 13/20
98/98 [=====] - 147s 1s/step - loss: 0.2809 - accuracy: 0.8993 - val_loss: 1.0707 - val_accuracy: 0.7222
Epoch 14/20
98/98 [=====] - 147s 1s/step - loss: 0.2322 - accuracy: 0.9177 - val_loss: 1.0346 - val_accuracy: 0.7356
Epoch 15/20
98/98 [=====] - 147s 1s/step - loss: 0.1896 - accuracy: 0.9313 - val_loss: 1.3116 - val_accuracy: 0.7073
Epoch 16/20
98/98 [=====] - 147s 2s/step - loss: 0.1824 - accuracy: 0.9345 - val_loss: 1.3514 - val_accuracy: 0.7199
Epoch 17/20
98/98 [=====] - 147s 1s/step - loss: 0.1401 - accuracy: 0.9492 - val_loss: 0.8573 - val_accuracy: 0.7350
Epoch 18/20
98/98 [=====] - 147s 1s/step - loss: 0.1277 - accuracy: 0.9544 - val_loss: 1.4452 - val_accuracy: 0.7261
Epoch 19/20
98/98 [=====] - 147s 1s/step - loss: 0.1238 - accuracy: 0.9564 - val_loss: 1.4452 - val_accuracy: 0.7348
Epoch 20/20
98/98 [=====] - 147s 1s/step - loss: 0.1322 - accuracy: 0.9539 - val_loss: 1.7509 - val_accuracy: 0.7199
Test loss: 1.7509175539016724
Test accuracy: 0.719900120162964
```

Training MLP took 2935.25500008296967 seconds

In [] :