→ CIFAR-10

Student Name 1: Alejandro Francisco Toral

```
Student Name 2: Enrique Martín López
# Load CIFAR-10 data set
from keras.datasets import cifar10
(X_train, y_train), (X_test, y_test) = cifar10.load_data()
     Downloading data from <a href="https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz">https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz</a>
     170500096/170498071 [==========] - 11s Ous/step
# 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))
for i in range(num_classes):
    ax = fig.add_subplot(2, 5, 1 + i, xticks=[], yticks=[])
    ax.set_title(class_names[i])
    idx = np.where(y_train[:]==i)[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()
 С
        airplane
                    automobile
                                   bird
                                                cat
                                                            deer
# Data pre-processing
X train = X train.astype('float32')
X_test = X_test.astype('float32')
X_train /= 255.0
X_test /= 255.0
from keras.utils import np_utils
Y_train = np_utils.to_categorical(y_train, num_classes)
Y_test = np_utils.to_categorical(y_test, num_classes)
def plot_model_history(model_history):
    fig, axs = plt.subplots(1,2,figsize=(15,5))
    # Summarize history for accuracy
    axs[0].plot(range(1,len(model\_history.history['accuracy'])+1), model\_history.history['accuracy'])
    axs[0].plot(range(1,len(model\_history.history['val\_accuracy'])+1), model\_history.history['val\_accuracy']) \\
    axs[0].set_title('Model Accuracy')
    axs[0].set_ylabel('Accuracy')
    axs[0].set_xlabel('Epoch')
    axs[0].set_xticks(np.arange(1,len(model_history.history['accuracy'])+1,step=len(model_history.history['accuracy'])/10))
    axs[0].legend(['train', 'val'], loc='best')
    # summarize history for loss
    axs[1].plot(range(1,len(model_history.history['loss'])+1),model_history.history['loss'])
    axs[1].plot(range(1,len(model_history.history['val_loss'])+1),model_history.history['val_loss'])
    axs[1].set_title('Model Loss')
    axs[1].set_ylabel('Loss')
    axs[1].set_xlabel('Epoch')
    axs[1].set\_xticks(np.arange(1,len(model\_history.history['loss'])+1, step=len(model\_history.history['loss'])/10))
    axs[1].legend(['train', 'val'], loc='best')
    plt.show()
from keras.initializers import he_uniform
learning_rate = 0.001
layers = [2500, 1250, 675, 125, 50]
layer_activation = "relu"
n_{epochs} = 200
batch size = 128
```

```
d rate = 0.05
# Feed forward dense model
# Here it is not allowed to use convolutional layers
# You may use any regularizacion (see class slides)
from keras.models import Sequential
from keras.layers import Dense, Dropout, Activation, Flatten, BatchNormalization
from keras.regularizers import 12
model = Sequential()
model.add(Flatten(input_shape=(32, 32, 3)))
for layer in layers:
 \verb|model.add(Dense(layers[0], kernel_initializer=my\_initializer, activation=layer\_activation, kernel\_regularizer=12(0.001))) # \\
 model.add(BatchNormalization())
 model.add(Dropout(rate=d_rate))
model.add(Dense(10))
model.add(Activation('softmax'))
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['categorical_accuracy'])
model.summary()
    Model: "sequential"
     Layer (type)
                                  Output Shape
                                                             Param #
     flatten (Flatten)
                                   (None, 3072)
                                                             0
     dense (Dense)
                                   (None, 2500)
                                                             7682500
     batch normalization (BatchNo (None, 2500)
                                                             10000
     dropout (Dropout)
                                   (None, 2500)
                                                             a
     dense_1 (Dense)
                                   (None, 2500)
                                                             6252500
     batch_normalization_1 (Batch (None, 2500)
                                                             10000
     dropout_1 (Dropout)
                                   (None, 2500)
                                                             0
     dense_2 (Dense)
                                   (None, 2500)
                                                             6252500
     batch normalization 2 (Batch (None, 2500)
                                                             10000
     dropout_2 (Dropout)
                                   (None, 2500)
                                                             0
     dense_3 (Dense)
                                   (None, 2500)
                                                              6252500
     batch_normalization_3 (Batch (None, 2500)
                                                             10000
     dropout_3 (Dropout)
                                   (None, 2500)
     dense 4 (Dense)
                                                             6252500
                                   (None, 2500)
     batch_normalization_4 (Batch (None, 2500)
                                                             10000
     dropout_4 (Dropout)
                                   (None, 2500)
                                                             a
     dense_5 (Dense)
                                                             25010
                                   (None, 10)
     activation (Activation)
                                   (None, 10)
     Total params: 32,767,510
     Trainable params: 32,742,510
    Non-trainable params: 25,000
# Training
import time
start = time.time()
history = model.fit(X_train, Y_train, batch_size=batch_size, epochs=n_epochs, verbose=0, validation_data=(X_test, Y_test))
end = time.time()
print("Training MLP took " + str(end - start) + " seconds")
     Training MLP took 918.9792449474335 seconds
import pandas as pd
results = pd.DataFrame(history.history)
results.plot(figsize = (8, 5))
plt.grid(True)
plt.xlabel ("Epochs")
plt.ylabel ("Accuracy - Mean Log Loss")
```

al nango to [0 1]

my_initializer = ne_unitorm(seed=None)

```
"pic.gca().sec_yiim(0, i) " sec the vertical range to [0,i]
plt.show()
print ("Accuracy (training): ",
       round((results.categorical_accuracy.values[-1:][0])*100, 1), "%")
print ("Accuracy (development test): ",
       round((results.val_categorical_accuracy.values[-1:][0])*100, 1), "%")
        14
                                                    categorical_accuracy
                                                    val loss
       12

    val_categorical_accuracy

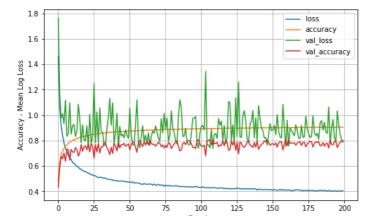
     Accuracy - Mean Log Loss
       10
         6
        4
         2
         0
                                              125
                                                     150
                                                            175
                                        100
                                                                   200
     Accuracy (training): 55.8 %
     Accuracy (development test): 42.9 %
loss, acc = model.evaluate(X_test, Y_test, verbose=0)
print('Test loss:', loss)
print('Test accuracy:', acc)
     Test loss: 1.8514587879180908
     Test accuracy: 0.42890000343322754
Definición de una red convolucional multicapa
from keras.initializers import he_uniform
learning_rate = 0.001
layer_activation = "relu"
n = 200
batch_size = 64
my_initializer = he_uniform(seed=None)
d rate = 0.2
# Convolutional Neural Network (CNN)
# Here you are allowed to use convolutional layers
# You may use also any regularizacion (see class slides)
from keras.models import Sequential
from keras.layers import Dense, Activation, Flatten
from keras.layers.convolutional import Conv2D, MaxPooling2D
import keras.backend as K
model = Sequential()
model.add(Conv2D(filters=32, kernel_size=(3, 3), padding='same', input_shape=(32, 32, 3), kernel_regularizer=12(0.0001)))
model.add(Activation('relu'))
model.add(BatchNormalization())
model.add(Dropout(rate=d rate))
model.add(MaxPooling2D(pool_size=(2, 2)))
\verb|model.add(Conv2D(filters=64, kernel\_size=(3, 3), padding='same', kernel\_regularizer=12(0.0001))||
model.add(Activation('relu'))
model.add(BatchNormalization())
model.add(Dropout(rate=d_rate))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Conv2D(filters=128, kernel_size=(3, 3), padding='same', kernel_regularizer=12(0.0001)))
model.add(Activation('relu'))
model.add(BatchNormalization())
model.add(Dropout(rate=d_rate))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Conv2D(filters=32, kernel_size=(1, 1), padding='same', kernel_regularizer=12(0.0001)))
model.add(Activation('relu'))
model.add(BatchNormalization())
model.add(Dropout(rate=d_rate))
model.add(Conv2D(filters=10, kernel_size=(4, 4), padding='valid'))
model.add(Flatten())
model add(Dense(10)
```

```
model.add(Activation('softmax'))
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model_summary()
```

Model: "sequential_3"

Layer (type)	Output	Shape	Param #
conv2d_10 (Conv2D)		32, 32, 32)	896
activation_11 (Activation)	(None,	32, 32, 32)	0
batch_normalization_13 (Batc	(None,	32, 32, 32)	128
dropout_13 (Dropout)	(None,	32, 32, 32)	0
max_pooling2d_6 (MaxPooling2	(None,	16, 16, 32)	0
conv2d_11 (Conv2D)	(None,	16, 16, 64)	18496
activation_12 (Activation)	(None,	16, 16, 64)	0
batch_normalization_14 (Batc	(None,	16, 16, 64)	256
dropout_14 (Dropout)	(None,	16, 16, 64)	0
max_pooling2d_7 (MaxPooling2	(None,	8, 8, 64)	0
conv2d_12 (Conv2D)	(None,	8, 8, 128)	73856
activation_13 (Activation)	(None,	8, 8, 128)	0
batch_normalization_15 (Batc	(None,	8, 8, 128)	512
dropout_15 (Dropout)	(None,	8, 8, 128)	0
max_pooling2d_8 (MaxPooling2	(None,	4, 4, 128)	0
conv2d_13 (Conv2D)	(None,	4, 4, 32)	4128
activation_14 (Activation)	(None,	4, 4, 32)	0
batch_normalization_16 (Batc	(None,	4, 4, 32)	128
dropout_16 (Dropout)	(None,	4, 4, 32)	0
conv2d_14 (Conv2D)	(None,	1, 1, 10)	5130
flatten_3 (Flatten)	(None,	10)	0
dense_8 (Dense)	(None,	10)	110
activation_15 (Activation)		10)	0
Total params: 103,640 Trainable params: 103,128 Non-trainable params: 512			

```
# Training
import time
start = time.time()
history = model.fit(X\_train, Y\_train, batch\_size=batch\_size, epochs=n\_epochs, verbose=0, validation\_data=(X\_test, Y\_test))
end = time.time()
print("Training MLP took " + str(end - start) + " seconds")
     Training MLP took 706.8354434967041 seconds
import pandas as pd
results = pd.DataFrame(history.history)
results.plot(figsize = (8, 5))
plt.grid(True)
plt.xlabel ("Epochs")
plt.ylabel ("Accuracy - Mean Log Loss")
\verb|#plt.gca().set_ylim(0, 1) | # set the vertical range to [0,1]|
plt.show()
print ("Accuracy (training): ",
       round((results.accuracy.values[-1:][0])*100, 1), "%")
print ("Accuracy (development test): ",
       round((results.val_accuracy.values[-1:][0])*100, 1), "%")
```



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

print('Test accuracy:', acc)

Test loss: 0.7863554358482361 Test accuracy: 0.8014000058174133