# Ejercicio 2

October 30, 2022

# 1 2 (30 puntos) Perceptrón multi-capa para detección de glaucoma en imágenes de fondo de ojo

1. Modifique el perceptrón multi-capa y la red convolucional implementadas en el codigo base provisto. Explique las modificaciones necesarias para implementar que ambos modelos funcionen para el objetivo especificado.

## 1.1 Imports

```
[]: from __future__ import print_function
     import argparse
     import torch
     import torch.nn as nn
     import torch.nn.functional as F
     import torch.optim as optim
     from torchvision import datasets, transforms
     from torch.optim.lr_scheduler import StepLR
     import numpy as np
     import torch
     import torchvision
     import matplotlib.pyplot as plt
     from time import time
     from torchvision import datasets, transforms
     from torchvision.io import read image
     from torch import nn, optim
     from matplotlib import cm
     from matplotlib import style
     from locale import normalize
     from matplotlib.pyplot import imshow
     %matplotlib widget
     style.use('default')
```

#### 1.2 Parámetros

```
[]: | lr = 0.01 # Learning rate empleado para la función de optimización SGD
    epochs = 15 # Cantidad de iteraciones que se ejecutaran por cada intento
    attempts = 10 # Cantidad de intentos totales que se ejecutaran (en cada uno de L
      →los cuales se creará el modelo, entrenará y obtendrán datos de efectividad)
    img size = 227 # Tamaño asignado para el width y height de las imágenes para
      ⇔entrenar y probar.
    batch_size = 32 # Cantidad de imágenes que serán procesadas por lote.
    criterion = nn.NLLLoss() # Función de perdida que se calculará individualmente
      ⇒para cada iteración y que se suma al total del intento
    ruta_base = r'.\\DatosDePrueba\\' # Ruta de la computadora a partir de la cualu
     ⇔se accede a los datos de prueba
    device = torch.device('cuda' if torch.cuda.is_available() else 'cpu') #_J
      →Dispositivo en el que se ejecutará
    transform = transforms.Compose([transforms.ToTensor(), transforms.
      Resize((img_size, img_size))]) # Transformaciones aplicadas a los sets de
      →entrenamiento y pruebas (todas las imágenes se ajustan al cuadrado img_size)
```

## 1.3 Cargado de datos de entrenamiento y prueba

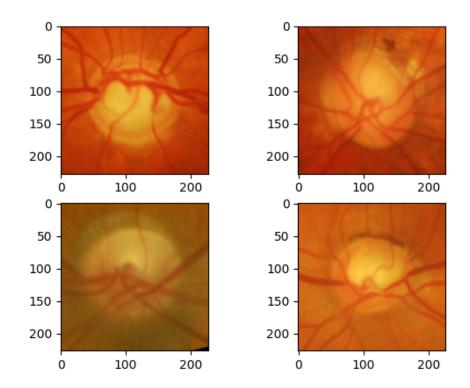
```
[]: def load data(nom carpeta): # Carqa imágenes desde el folder según el nombre
      ⇒que se indique
         set = torchvision.datasets.ImageFolder(ruta_base + nom_carpeta, transform = __
      →transform)
         loader = torch.utils.data.DataLoader(set, batch_size = batch_size, shuffle_
      →= True)
         return loader
     def load_test_train_data(): # Carga imágenes de entrenamiento y pruebas
         trainloader = load_data('train')
         testloader = load_data('test')
         return trainloader, testloader
     trainloader, testloader = load_test_train_data()
     dataiter = iter(trainloader)
     images, labels = dataiter.next()
     print('[Batch_size, channels, img_size, img_size] =,', images.mT.shape)
     fig, axs = plt.subplots(2,2)
     axs[0,0].imshow(images[0].T.numpy().squeeze(), cmap='gray_r')
     axs[0,1].imshow(images[1].T.numpy().squeeze(), cmap='gray_r')
     axs[1,0].imshow(images[2].T.numpy().squeeze(), cmap='gray_r')
     axs[1,1].imshow(images[3].T.numpy().squeeze(), cmap='gray_r')
```

[Batch\_size, channels, img\_size, img\_size] =, torch.Size([32, 3, 227, 227])

C:\Users\XPC\AppData\Local\Temp\ipykernel\_46556\2524731925.py:19: UserWarning: The use of `x.T` on tensors of dimension other than 2 to reverse their shape is deprecated and it will throw an error in a future release. Consider `x.mT` to transpose batches of matricesor `x.permute(\*torch.arange(x.ndim - 1, -1, -1))` to reverse the dimensions of a tensor. (Triggered internally at ..\aten\src\ATen\native\TensorShape.cpp:2985.)

axs[0,0].imshow(images[0].T.numpy().squeeze(), cmap='gray\_r')

[]: <matplotlib.image.AxesImage at 0x20747706e60>



## 1.4 Creando el modelo del perceptrón multicapa

```
[]: def create_MLP_model(verbose = True): # Creación del modelo
    model = nn.Sequential(
        nn.Linear(img_size*img_size*3,128), # Empleando como entrada lasu
        -dimensiones de la imagen (w * h) multiplicado por los 3 canales de color queu
        -se están utilizando => 154587 = 227 * 227 * 3
        nn.Sigmoid(),
        nn.Linear(128,64),
        nn.Sigmoid(),
        nn.Linear(64,2),
```

```
nn.LogSoftmax(dim=1))
  if (verbose): print("Running on device: ", device)
  model.to(device)
  return model
mlp_model = create_MLP_model()
print("MLP model")
print(mlp_model)
Running on device: cuda
MLP model
Sequential(
  (0): Linear(in_features=154587, out_features=128, bias=True)
  (1): Sigmoid()
  (2): Linear(in_features=128, out_features=64, bias=True)
  (3): Sigmoid()
  (4): Linear(in_features=64, out_features=2, bias=True)
  (5): LogSoftmax(dim=1)
)
```

#### 1.5 Entrenando el modelo

```
[]: def train_model(model, criterion, epochs, lr, verbose = True):
         start = time()
         epochs_list = []
         running_loss_list= []
         optimizer = optim.SGD(model.parameters(), lr= lr, momentum=0.9)
         for e in range(epochs):
             running_loss = 0
             for images, labels in trainloader: # Ejecución de la iteración por lotes
                 images = images.to(device) # Se preparan las imágenes del lote enu
      ⇔el device usado
                 labels = labels.to(device)
                 images = images.view(images.shape[0], -1)
                 optimizer.zero_grad()
                 output = model(images) # Aplicando el modelo para cada batch
                 loss = criterion(output, labels) # Y calculando el error asociado
                 loss.backward() # Aprendizaje por backpropagation
                 optimizer.step() # weights optimization
                 running_loss += loss.item() # Se suma la pérdida para el total de_
      →la iteración
             else:
```

```
if (verbose): print("Epoch {} - Training loss: {}".format(e, userunning_loss/len(trainloader)))

if (verbose): print("\nTraining Time (in minutes) = ",(time()-start)/60)
    return model

print("Training MLP model")
print("Device", next(mlp_model.parameters()).is_cuda)

mlp_model = train_model(mlp_model, criterion, epochs, lr)
```

```
Training MLP model
Device True
Epoch 0 - Training loss: 0.6811960376799107
Epoch 1 - Training loss: 0.6621619202196598
Epoch 2 - Training loss: 0.6019255146384239
Epoch 3 - Training loss: 0.5343579966574907
Epoch 4 - Training loss: 0.45502882823348045
Epoch 5 - Training loss: 0.3947829157114029
Epoch 6 - Training loss: 0.47907054238021374
Epoch 7 - Training loss: 0.378512155264616
Epoch 8 - Training loss: 0.334895103238523
Epoch 9 - Training loss: 0.296166880056262
Epoch 10 - Training loss: 0.2786378934979439
Epoch 11 - Training loss: 0.30313558131456375
Epoch 12 - Training loss: 0.35534085519611835
Epoch 13 - Training loss: 0.2813700754195452
Epoch 14 - Training loss: 0.2712722350843251
```

## 1.6 Probando el modelo

```
[]: def test_model_mlp(testloader, model, verbose = True):
    correct_rate, false_negative_rate, all_count = 0, 0, 0

    for images,labels in testloader:
        images = images.to(device) # Se preparan las imágenes del lote en elu
        device usado
        labels = labels.to(device)

        for i in range(len(labels)): # se itera sobre los índices de targets
        img = images[i].view(1, img_size*img_size*3)
        with torch.no_grad():
            logps = model(img)
            ps = torch.exp(logps)
            probab = list(ps.cpu().numpy()[0])
```

```
pred_label = probab.index(max(probab)) # Se obtiene el target predicho_
  →de la iteración actual
        true_label = labels.cpu().numpy()[i] # Se obtiene el target correcto de_
  →la iteración actual
        if (true_label == pred_label): correct_rate += 1 # Predicción correcta_
  ⇔si iqual a target
        else:
          if (pred_label == 0): false_negative_rate += 1 # Falso negativo si_
  ⇔predicción es 0 pero correcto 1
        all_count += 1
    if (verbose):
      print("Images Tested =", all_count)
      print("Correct Tests =", correct_rate)
      print("False Positive Tests =", (all_count - correct_rate) -__
  →false_negative_rate)
      print("False Negative Tests =", false_negative_rate)
      print("\nModel Accuracy (Average) =", np.round((correct_rate/
  ⇔all_count)*100,4),"%")
    return correct_rate, false_negative_rate, all_count
print("Testing MLP model")
res = test_model_mlp(testloader, mlp_model)
Testing MLP model
```

```
Images Tested = 210

Correct Tests = 120

False Positive Tests = 90

False Negative Tests = 0

Model Accuracy (Average) = 57.1429 %
```

## 1.7 Ejecuciones consecutivas de entrenamiento y pruebas

a) Entrene el perceptrón multi-capa usando el conjunto de datos completo para entrenamiento y validación. Calibre los hiper-parámetros necesarios para obtener los mejores resultados posibles y reportelos. Ejecute el entrenamiento 10 veces por 15 épocas por corrida, y reporte la tasa de aciertos tasa de aciertos, falsos positivos y falsos negativos promedio y su desviación estándar para esas 10 corridas.

```
[]: print('{} training and testing attempts:\n'.format(attempts))
   aciertos = []

for i in range(attempts):
```

```
mlp_model = create_MLP_model(False)
    mlp_model = train_model(mlp_model, criterion, epochs, lr, False)
    correct_rate, false_negative_rate, all_count = test_model_mlp(testloader,_u
 →mlp_model, False)
    aciertos.append(correct_rate)
    print('Results of attempt #{}:'.format(i))
    print('\tCorrect: {}\tFalse Positive Tests: {}\tFalse Negative Tests: 
  →{}\tModel Accuracy: {}\n'.format(correct_rate, (all_count - correct_rate) -
 false negative rate, false negative rate, correct_rate/all_count))
resultados = torch.tensor(aciertos)
print('Average correct tests: {}'.format(torch.mean(resultados.double()).
 →item()))
print('Standard deviation: {}\n'.format(torch.std(resultados.double()).item()))
10 training and testing attempts:
Results of attempt #0:
       Correct: 189 False Positive Tests: 13
                                                       False Negative Tests: 8
Model Accuracy: 0.9
Results of attempt #1:
       Correct: 180
                       False Positive Tests: 25
                                                       False Negative Tests: 5
Model Accuracy: 0.8571428571428571
Results of attempt #2:
       Correct: 187
                     False Positive Tests: 15
                                                       False Negative Tests: 8
Model Accuracy: 0.8904761904761904
Results of attempt #3:
       Correct: 187
                       False Positive Tests: 4 False Negative Tests: 19
Model Accuracy: 0.8904761904761904
Results of attempt #4:
       Correct: 165
                       False Positive Tests: 42
                                                       False Negative Tests: 3
Model Accuracy: 0.7857142857142857
Results of attempt #5:
       Correct: 180
                                                       False Negative Tests: 6
                       False Positive Tests: 24
Model Accuracy: 0.8571428571428571
Results of attempt #6:
       Correct: 185
                       False Positive Tests: 11
                                                      False Negative Tests: 14
Model Accuracy: 0.8809523809523809
Results of attempt #7:
```

Correct: 161 False Positive Tests: 46 False Negative Tests: 3

Model Accuracy: 0.766666666666667

Results of attempt #8:

Correct: 188 False Positive Tests: 4 False Negative Tests: 18

Model Accuracy: 0.8952380952380953

Results of attempt #9:

Correct: 153 False Positive Tests: 57 False Negative Tests: 0

Model Accuracy: 0.7285714285714285

Average correct tests: 177.5

Standard deviation: 12.99786307223016

## 2 Ejecución de corridas sobre el modelo MLP empleando distintos LR cada dos corridas

#### 2.1 LR 0.1

## 2.1.1 Intento 1

## Training MLP model

- Epoch 0 Training loss: 0.7369163818657398
- Epoch 1 Training loss: 0.6828029863536358
- Epoch 2 Training loss: 0.6817193254828453
- Epoch 3 Training loss: 0.6662973240017891
- Epoch 4 Training loss: 0.6442956253886223
- Epoch 5 Training loss: 0.7077872566878796
- Epoch 6 Training loss: 0.6701737195253372
- Epoch 7 Training loss: 0.6801279932260513
- Epoch 9 Training loss: 0.6868789345026016
- Epoch 11 Training loss: 0.6495210751891136
- Epoch 12 Training loss: 0.6918781399726868
- $\bullet \;\; \text{Epoch } 13$  Training loss: 0.7025960497558117
- Epoch 14 Training loss: 0.6821348257362843

Training Time (in minutes) = 0.6270199020703634

## Testing MLP model

- Images Tested = 210
- Correct Tests = 105
- False Positive Tests = 105
- False Negative Tests = 0

Model Accuracy (Average) = 50.0 %

#### 2.1.2 Intento 2

### Training MLP model

- Epoch 0 Training loss: 0.6815511770546436
- Epoch 1 Training loss: 0.7036696895956993
- Epoch 2 Training loss: 0.6777684912085533
- Epoch 3 Training loss: 0.5605829171836376
- Epoch 4 Training loss: 0.6949820667505264
- Epoch 5 Training loss: 0.6932078413665295
- Epoch 6 Training loss: 0.6773754023015499
- Epoch 7 Training loss: 0.7096707485616207
- Epoch 8 Training loss: 0.6872112117707729
- Epoch 9 Training loss: 0.6869790367782116
- Epoch 10 Training loss: 0.6942282542586327
- Epoch 11 Training loss: 0.7144105024635792
- Epoch 12 Training loss: 0.6969177834689617
- Epoch 13 Training loss: 0.722785871475935
- Epoch 14 Training loss: 0.7197191417217255

Training Time (in minutes) = 0.6199134389559428

## Testing MLP model

- Images Tested = 210
- Correct Tests = 105
- False Positive Tests = 105
- False Negative Tests = 0

Model Accuracy (Average) = 50.0 %

#### 2.2 LR 0.01

#### **2.2.1** Intento 1

- Epoch 0 Training loss: 0.6827975213527679
- Epoch 1 Training loss: 0.6605164743959904
- Epoch 2 Training loss: 0.6082503236830235
- Epoch 3 Training loss: 0.5294231809675694
- Epoch 4 Training loss: 0.4830891713500023
- Epoch 5 Training loss: 0.4102689679712057
- Epoch 6 Training loss: 0.37855312693864107
- Epoch 7 Training loss: 0.3209534380584955
- Epoch 8 Training loss: 0.31282537151128054
- Epoch 9 Training loss: 0.28880161326378584
- Epoch 10 Training loss: 0.2564556635916233
- Epoch 11 Training loss: 0.2673725155182183
- Epoch 12 Training loss: 0.260069502517581
- Epoch 13 Training loss: 0.2765366407111287

• Epoch 14 - Training loss: 0.29861110355705023

Training Time (in minutes) = 0.61580970287323

#### Testing MLP model

- Images Tested = 210
- Correct Tests = 185
- False Positive Tests = 22
- False Negative Tests = 3

Model Accuracy (Average) = 88.0952 %

#### 2.2.2 Intento 2

## Training MLP model

- Epoch 0 Training loss: 0.7057710886001587
  - Epoch 1 Training loss: 0.6802574507892132
  - Epoch 2 Training loss: 0.6727069728076458
  - Epoch 3 Training loss: 0.6631590500473976
  - Epoch 4 Training loss: 0.6375782452523708
  - Epoch 5 Training loss: 0.5831635408103466
  - Epoch 6 Training loss: 0.5207257680594921
  - Epoch 7 Training loss: 0.4437313415110111
  - Epoch 8 Training loss: 0.36804841458797455
  - Epoch 9 Training loss: 0.3788630049675703
  - Epoch 10 Training loss: 0.3542727045714855
  - Epoch 11 Training loss: 0.34089745953679085
  - Epoch 12 Training loss: 0.3315946366637945
  - Epoch 13 Training loss: 0.28785437252372503
  - Epoch 14 Training loss: 0.2747528455220163

Training Time (in minutes) = 0.6155428210894267

#### Testing MLP model

- Images Tested = 210
- Correct Tests = 167
- False Positive Tests = 0
- False Negative Tests = 43

Model Accuracy (Average) = 79.5238 %

## 2.3 LR 0.03

#### 2.3.1 Intento 1

- Epoch 0 Training loss: 0.7078320719301701
- Epoch 1 Training loss: 0.6655924059450626
- Epoch 2 Training loss: 0.5964057557284832

- Epoch 3 Training loss: 0.5449526645243168
- Epoch 4 Training loss: 0.5642602499574423
- Epoch 5 Training loss: 0.4499690402299166
- Epoch 6 Training loss: 0.4313625395298004
- Epoch 7 Training loss: 0.41177461203187704
- Epoch 8 Training loss: 0.3985733538866043
- Epoch 9 Training loss: 0.45222795754671097
- Epoch 10 Training loss: 0.4886812958866358
- Epoch 11 Training loss: 0.63541179895401
- Epoch 12 Training loss: 0.6751604080200195
- Epoch 13 Training loss: 0.6382348835468292
- Epoch 14 Training loss: 0.6463134028017521

## Testing MLP model

- Images Tested = 210
- Correct Tests = 126
- False Positive Tests = 84
- False Negative Tests = 0

Model Accuracy (Average) = 60.0 %

#### 2.3.2 Intento 2

#### Training MLP model

- Epoch 0 Training loss: 0.6665529124438763
- Epoch 1 Training loss: 0.6589692607522011
- Epoch 2 Training loss: 0.6091506369411945
- Epoch 3 Training loss: 0.4977915268391371
- Epoch 4 Training loss: 0.4174402579665184
- Epoch 5 Training loss: 0.3881151517853141
- Epoch 6 Training loss: 0.4580058716237545
- Epoch 8 Training loss: 0.39315566048026085
- Epoch 9 Training loss: 0.5275244060903788
- Epoch 10 Training loss: 0.44969186559319496
- Epoch 11 Training loss: 0.5123887434601784
- Epoch 12 Training loss: 0.4593819808214903
- Epoch 13 Training loss: 0.6415568217635155
- Epoch 14 Training loss: 0.6992569081485271

Training Time (in minutes) = 0.6200714071591695

#### Testing MLP model

- Images Tested = 210
- Correct Tests = 117
- False Positive Tests = 93

• False Negative Tests = 0

Model Accuracy (Average) = 55.7143 %

#### 2.4 LR 0.001

#### 2.4.1 Intento 1

## Training MLP model

- Epoch 0 Training loss: 0.6840277686715126
- Epoch 1 Training loss: 0.6728365309536457
- Epoch 2 Training loss: 0.6657264605164528
- Epoch 3 Training loss: 0.661868829280138
- Epoch 4 Training loss: 0.655780304223299
- Epoch 5 Training loss: 0.6464584358036518
- Epoch 6 Training loss: 0.6451276578009129
- Epoch 7 Training loss: 0.6388522908091545
- Epoch 8 Training loss: 0.6317578107118607
- Epoch 10 Training loss: 0.6149414479732513
- Epoch 11 Training loss: 0.6064675077795982
- Epoch 12 Training loss: 0.5985792241990566
- Epoch 13 Training loss: 0.5939729511737823
- Epoch 14 Training loss: 0.586756456643343

Training Time (in minutes) = 0.6210965474446615

#### Testing MLP model

- Images Tested = 210
- Correct Tests = 129
- False Positive Tests = 79
- False Negative Tests = 2

Model Accuracy (Average) = 61.4286 %

#### 2.4.2 Intento 2

- Epoch 0 Training loss: 0.6771115027368069
- Epoch 1 Training loss: 0.6740257628262043
- Epoch 2 Training loss: 0.6744579784572124
- Epoch 3 Training loss: 0.6677091605961323
- Epoch 4 Training loss: 0.6612282618880272
- Epoch 5 Training loss: 0.6544647552073002
- Epoch 6 Training loss: 0.6521985307335854
- Epoch 7 Training loss: 0.6469268091022968
- Epoch 8 Training loss: 0.6415448822081089
- Epoch 9 Training loss: 0.637713972479105
- Epoch 10 Training loss: 0.6335229761898518

- Epoch 11 Training loss: 0.6256400793790817
- Epoch 12 Training loss: 0.6197583377361298
- Epoch 13 Training loss: 0.6133669465780258
- Epoch 14 Training loss: 0.6117302030324936

## Testing MLP model

- Images Tested = 210
- Correct Tests = 113
- False Positive Tests = 97
- False Negative Tests = 0

Model Accuracy (Average) = 53.8095 %

#### 2.5 LR 0.003

#### 2.5.1 Intento 1

#### Training MLP model

- Epoch 0 Training loss: 0.6874632090330124
- Epoch 1 Training loss: 0.6731723919510841
- Epoch 2 Training loss: 0.6625744327902794
- Epoch 3 Training loss: 0.6509069390594959
- Epoch 4 Training loss: 0.6418970599770546
- Epoch 5 Training loss: 0.626440305262804
- Epoch 6 Training loss: 0.6111289225518703
- Epoch 7 Training loss: 0.5899031274020672
- Epoch 8 Training loss: 0.5702399350702763
- Epoch 9 Training loss: 0.5409582275897264
- Epoch 10 Training loss: 0.5159615613520145
- Epoch 11 Training loss: 0.48713919520378113
- Epoch 12 Training loss: 0.45356206968426704
- Epoch 13 Training loss: 0.4310105964541435
- Epoch 14 Training loss: 0.4148829597979784

Training Time (in minutes) = 0.6126735369364421

## Testing MLP model

- Images Tested = 210
- Correct Tests = 155
- False Positive Tests = 52
- False Negative Tests = 3

Model Accuracy (Average) = 73.8095 %

#### 2.5.2 Intento 2

- Epoch 0 Training loss: 0.6764406636357307
- Epoch 1 Training loss: 0.6625492610037327
- Epoch 2 Training loss: 0.6434317827224731
- Epoch 3 Training loss: 0.6279043667018414
- Epoch 4 Training loss: 0.6049855165183544
- Epoch 5 Training loss: 0.583499364554882
- Epoch 6 Training loss: 0.5568369124084711
- Epoch 7 Training loss: 0.5259818211197853
- Epoch 8 Training loss: 0.4958261363208294
- Epoch 9 Training loss: 0.4689131695777178
- Epoch 10 Training loss: 0.43894270434975624
- Epoch 11 Training loss: 0.43525825813412666
- Epoch 12 Training loss: 0.398719334974885
- Epoch 13 Training loss: 0.3785032369196415
- Epoch 14 Training loss: 0.3820067588239908

## Testing MLP model

- Images Tested = 210
- Correct Tests = 179
- False Positive Tests = 13
- False Negative Tests = 18

Model Accuracy (Average) = 85.2381 %

# 3 Ejecución de corridas sobre el modelo MLP empleando los mismos LR en todas las corridas (LR = 0.01)

## 3.1 Ejecución #1

### 3.1.1 Training MLP model

- Epoch 0 Training loss: 0.6755483783781528
- Epoch 1 Training loss: 0.6295309439301491
- Epoch 2 Training loss: 0.5783903300762177
- Epoch 3 Training loss: 0.5343762598931789
- Epoch 4 Training loss: 0.4437193498015404
- Epoch 5 Training loss: 0.39533296041190624
- Epoch 6 Training loss: 0.41546942479908466
- Epoch 7 Training loss: 0.3631477542221546
- Epoch 8 Training loss: 0.36750431172549725
- Epoch 9 Training loss: 0.3276412822306156
- Epoch 10 Training loss: 0.3217639746144414
- Epoch 11 Training loss: 0.31924678198993206
- Epoch 12 Training loss: 0.3261508094146848
- Epoch 13 Training loss: 0.26522853458300233
- Epoch 14 Training loss: 0.27376265823841095

#### 3.1.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 184
- False Positive Tests = 20
- False Negative Tests = 6

Model Accuracy (Average) = 87.619 %

## 3.2 Ejecución #2

#### 3.2.1 Training MLP model

- Epoch 0 Training loss: 0.6789938472211361
- Epoch 1 Training loss: 0.6559620313346386
- Epoch 2 Training loss: 0.5979436337947845
- Epoch 3 Training loss: 0.5186452232301235
- Epoch 4 Training loss: 0.4913574047386646
- Epoch 5 Training loss: 0.4588260278105736
- Epoch 6 Training loss: 0.42552139796316624
- Epoch 7 Training loss: 0.38591246493160725
- Epoch 8 Training loss: 0.3318267595022917
- Epoch 9 Training loss: 0.3677603965625167
- Epoch 10 Training loss: 0.3285478949546814
- Epoch 11 Training loss: 0.30944768711924553
- Epoch 12 Training loss: 0.31867357436567545
- Epoch 13 Training loss: 0.2513004643842578
- Epoch 14 Training loss: 0.32883313158527017

Training Time (in minutes) = 0.6374959429105123

#### 3.2.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 184
- False Positive Tests = 20
- False Negative Tests = 6

Model Accuracy (Average) = 87.619 %

## 3.3 Ejecución #3

#### 3.3.1 Training MLP model

- Epoch 0 Training loss: 0.669500358402729
- Epoch 1 Training loss: 0.6638954095542431
- Epoch 2 Training loss: 0.6090108752250671
- Epoch 3 Training loss: 0.5524321272969246
- Epoch 4 Training loss: 0.46861656941473484

- Epoch 5 Training loss: 0.4160280302166939
- Epoch 6 Training loss: 0.43497278820723295
- Epoch 7 Training loss: 0.3823196701705456
- Epoch 8 Training loss: 0.33440618123859167
- Epoch 9 Training loss: 0.29762012232095003
- Epoch 10 Training loss: 0.4215814843773842
- Epoch 11 Training loss: 0.35535851027816534
- Epoch 12 Training loss: 0.38565926626324654
- Epoch 13 Training loss: 0.3213418032974005
- Epoch 14 Training loss: 0.28836515452712774

## 3.3.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 177
- False Positive Tests = 28
- False Negative Tests = 5

Model Accuracy (Average) = 84.2857 %

#### 3.4 Ejecución #4

#### 3.4.1 Training MLP model

- Epoch 0 Training loss: 0.6817897818982601
- Epoch 1 Training loss: 0.6486177258193493
- Epoch 2 Training loss: 0.6065827794373035
- Epoch 3 Training loss: 0.5542502943426371
- Epoch 4 Training loss: 0.4871502108871937
- Epoch 5 Training loss: 0.4139289911836386
- Epoch 6 Training loss: 0.43559644743800163
- Epoch 7 Training loss: 0.3611491918563843
- Epoch 8 Training loss: 0.3386910930275917
- Epoch 9 Training loss: 0.33063480723649263
- Epoch 10 Training loss: 0.28561106510460377
- Epoch 11 Training loss: 0.35864088498055935
- Epoch 12 Training loss: 0.3085365677252412
- Epoch 13 Training loss: 0.28754513058811426
- Epoch 14 Training loss: 0.27902613021433353

Training Time (in minutes) = 0.6235944430033366

#### 3.4.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 188
- False Positive Tests = 6
- False Negative Tests = 16

## 3.5 Ejecución #5

#### 3.5.1 Training MLP model

- Epoch 0 Training loss: 0.6733688861131668
- Epoch 1 Training loss: 0.6364881433546543
- Epoch 2 Training loss: 0.5911890938878059
- Epoch 3 Training loss: 0.5003748796880245
- Epoch 4 Training loss: 0.4436192847788334
- Epoch 5 Training loss: 0.4071769081056118
- Epoch 6 Training loss: 0.36832852475345135
- Epoch 7 Training loss: 0.31542853731662035
- Epoch 8 Training loss: 0.31395438965409994
- Epoch 9 Training loss: 0.3593072723597288
- Epoch 10 Training loss: 0.3414111491292715
- Epoch 11 Training loss: 0.366707980632782
- Epoch 12 Training loss: 0.3180706249549985
- Epoch 13 Training loss: 0.29311301838606596
- Epoch 14 Training loss: 0.3368922136723995

Training Time (in minutes) = 0.6206583380699158

## 3.5.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 156
- False Positive Tests = 53
- False Negative Tests = 1

Model Accuracy (Average) = 74.2857 %

#### 3.6 Ejecución #6

#### 3.6.1 Training MLP model

- Epoch 0 Training loss: 0.6783614419400692
- Epoch 1 Training loss: 0.6365665383636951
- Epoch 2 Training loss: 0.5871519818902016
- Epoch 3 Training loss: 0.5118771959096193
- Epoch 4 Training loss: 0.44583754427731037
- Epoch 5 Training loss: 0.3815526831895113
- Epoch 6 Training loss: 0.3625989994034171
- Epoch 7 Training loss: 0.42644291929900646
- Epoch 8 Training loss: 0.3171516880393028
- Epoch 9 Training loss: 0.2905137366615236
- Epoch 10 Training loss: 0.3491507964208722
- Epoch 11 Training loss: 0.4224322587251663
- Epoch 12 Training loss: 0.31965601071715355
- Epoch 13 Training loss: 0.28112538903951645

• Epoch 14 - Training loss: 0.2759061469696462

Training Time (in minutes) = 0.6020392537117004

#### 3.6.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 164
- False Positive Tests = 45
- False Negative Tests = 1

Model Accuracy (Average) = 78.0952 %

## 3.7 Ejecución #7

#### 3.7.1 Training MLP model

- Epoch 0 Training loss: 0.6865210831165314
- Epoch 1 Training loss: 0.6606931053102016
- Epoch 2 Training loss: 0.6039097309112549
- Epoch 3 Training loss: 0.5431239902973175
- Epoch 4 Training loss: 0.5155778210610151
- Epoch 5 Training loss: 0.5216354019939899
- Epoch 6 Training loss: 0.39762562699615955
- Epoch 7 Training loss: 0.380113935098052
- Epoch 8 Training loss: 0.33804269321262836
- Epoch 9 Training loss: 0.3554385444149375
- Epoch 10 Training loss: 0.3892012722790241
- Epoch 11 Training loss: 0.35064855590462685
- Epoch 12 Training loss: 0.3019036818295717
- Epoch 13 Training loss: 0.318946554325521
- Epoch 14 Training loss: 0.2749772281385958

Training Time (in minutes) = 0.6188109715779623

#### 3.7.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 163
- False Positive Tests = 45
- False Negative Tests = 2

Model Accuracy (Average) = 77.619 %

## 3.8 Ejecución #8

#### 3.8.1 Training MLP model

- Epoch 0 Training loss: 0.7105191797018051
- Epoch 1 Training loss: 0.6804268285632133
- Epoch 2 Training loss: 0.6639746502041817
- Epoch 3 Training loss: 0.6529141366481781

- Epoch 4 Training loss: 0.6171780154109001
- Epoch 5 Training loss: 0.557680094614625
- Epoch 6 Training loss: 0.5138429421931505
- Epoch 7 Training loss: 0.43288153782486916
- Epoch 8 Training loss: 0.37753714621067047
- Epoch 9 Training loss: 0.3594362251460552
- Epoch 10 Training loss: 0.40580929070711136
- Epoch 11 Training loss: 0.3357595829293132
- Epoch 12 Training loss: 0.4155731126666069
- Epoch 13 Training loss: 0.3880708534270525
- Epoch 14 Training loss: 0.3696711454540491

## 3.8.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 133
- False Positive Tests = 77
- False Negative Tests = 0

Model Accuracy (Average) = 63.3333%

#### 3.9 Ejecución #9

## 3.9.1 Training MLP model

- Epoch 0 Training loss: 0.6786065921187401
- Epoch 1 Training loss: 0.6465523429214954
- Epoch 2 Training loss: 0.5971381552517414
- Epoch 3 Training loss: 0.5270331036299467
- Epoch 4 Training loss: 0.4790476132184267
- Epoch 5 Training loss: 0.39014657586812973
- Epoch 6 Training loss: 0.41562835685908794
- Epoch 7 Training loss: 0.343689332716167
- Epoch 8 Training loss: 0.3210268598049879
- Epoch 9 Training loss: 0.3434343272820115
- Epoch 10 Training loss: 0.2874893303960562
- Epoch 11 Training loss: 0.3127787606790662
- Epoch 12 Training loss: 0.2764781713485718
- Epoch 13 Training loss: 0.288813638035208
- Epoch 14 Training loss: 0.2772562876343727

Training Time (in minutes) = 0.6265958666801452

#### 3.9.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 175
- False Positive Tests = 0
- False Negative Tests = 35

## 3.10 Ejecución #10

#### 3.10.1 Training MLP model

- Epoch 0 Training loss: 0.6788728050887585
- Epoch 1 Training loss: 0.644656702876091
- Epoch 2 Training loss: 0.6026467494666576
- Epoch 3 Training loss: 0.5293423216789961
- Epoch 4 Training loss: 0.43763526156544685
- Epoch 5 Training loss: 0.3918335437774658
- Epoch 6 Training loss: 0.36272476986050606
- Epoch 7 Training loss: 0.33194719441235065
- Epoch 8 Training loss: 0.39527758676558733
- Epoch 9 Training loss: 0.31878712866455317
- Epoch 10 Training loss: 0.32153183966875076
- Epoch 11 Training loss: 0.33105915877968073
- Epoch 12 Training loss: 0.2790641039609909
- Epoch 13 Training loss: 0.28298106137663126
- Epoch 14 Training loss: 0.27733859745785594

Training Time (in minutes) = 0.6084364891052246

#### 3.10.2 Testing MLP model

- Images Tested = 210
- Correct Tests = 188
- False Positive Tests = 3
- False Negative Tests = 19

Model Accuracy (Average) = 89.5238 %