

✓ Detecção de Casos de COVID-19 em Tomografias (versão PyTorch)

Agenda

- Carregar dataset (ImageFolder)
- Pré-processamento e augmentations
- Transfer Learning (ResNet18)
- Treino, validação, checkpoint, early stopping
- Avaliação final e plots

```
!pip install kaggle
```

```
Requirement already satisfied: kaggle in /usr/local/lib/python3.12/dist-packages (1.7.4.5)
Requirement already satisfied: bleach in /usr/local/lib/python3.12/dist-packages (from kaggle) (6.3.0)
Requirement already satisfied: certifi>=14.05.14 in /usr/local/lib/python3.12/dist-packages (from kaggle) (2025.11.12)
Requirement already satisfied: charset-normalizer in /usr/local/lib/python3.12/dist-packages (from kaggle) (3.4.4)
Requirement already satisfied: idna in /usr/local/lib/python3.12/dist-packages (from kaggle) (3.11)
Requirement already satisfied: protobuf in /usr/local/lib/python3.12/dist-packages (from kaggle) (5.29.5)
Requirement already satisfied: python-dateutil>=2.5.3 in /usr/local/lib/python3.12/dist-packages (from kaggle) (2.9.0.post0)
Requirement already satisfied: python-slugify in /usr/local/lib/python3.12/dist-packages (from kaggle) (8.0.4)
Requirement already satisfied: requests in /usr/local/lib/python3.12/dist-packages (from kaggle) (2.32.4)
Requirement already satisfied: setuptools>=21.0.0 in /usr/local/lib/python3.12/dist-packages (from kaggle) (75.2.0)
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.12/dist-packages (from kaggle) (1.17.0)
Requirement already satisfied: text-unidecode in /usr/local/lib/python3.12/dist-packages (from kaggle) (1.3)
Requirement already satisfied: tqdm in /usr/local/lib/python3.12/dist-packages (from kaggle) (4.67.1)
Requirement already satisfied: urllib3>=1.15.1 in /usr/local/lib/python3.12/dist-packages (from kaggle) (2.5.0)
Requirement already satisfied: webencodings in /usr/local/lib/python3.12/dist-packages (from kaggle) (0.5.1)
```

```
from google.colab import files
files.upload() # Envie o arquivo kaggle.json
```

Escolher arquivos Nenhum arquivo escolhido Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

```
# Cell: Imports e configuração
import os
import random

from pathlib import Path
import numpy as np
import matplotlib.pyplot as plt
from tqdm import tqdm

import torch
import torch.nn as nn
import torch.optim as optim
from torch.utils.data import DataLoader, random_split
from torchvision import transforms, datasets, models

from sklearn.metrics import classification_report, confusion_matrix

# Configurações
SEED = 42
random.seed(SEED)
np.random.seed(SEED)
torch.manual_seed(SEED)
torch.cuda.manual_seed_all(SEED)

DEVICE = torch.device("cuda" if torch.cuda.is_available() else "cpu")
print("Device:", DEVICE)
```

Device: cuda

```
# Paths e hiper-parâmetros (ajuste aqui)
# DATA_DIR = "/kaggle/input" # <- ajuste para o caminho onde você extraiu o dataset
# The dataset was unzipped to /content/covid_dataset
DATA_DIR = "/content/covid_dataset"
# Espera-se uma estrutura: DATA_DIR/class1/*.png, DATA_DIR/class2/*.png (ImageFolder)
BATCH_SIZE = 16
IMG_SIZE = 224
NUM_EPOCHS = 25
LR = 2e-4
NUM_WORKERS = 4
```

```
PATIENCE = 6 # early stopping
MODEL_SAVE_PATH = "best_model_resnet18.pt"
```

```
import zipfile
import os

# Path to the uploaded zip file
zip_path = '/content/archive (3).zip'
# Directory where the dataset will be extracted
extract_base_path = '/content/covid_dataset'

# Create the extraction directory if it doesn't exist
if not os.path.exists(extract_base_path):
    os.makedirs(extract_base_path)

# Unzip the file
print(f"Unzipping {zip_path} to {extract_base_path}...")
with zipfile.ZipFile(zip_path, 'r') as zip_ref:
    zip_ref.extractall(extract_base_path)

print("Unzipping complete.")

# Optional: List contents to verify structure and find the actual dataset root
print("Contents of extracted directory:")
for root, dirs, files in os.walk(extract_base_path):
    level = root.replace(extract_base_path, '').count(os.sep)
    indent = ' ' * 4 * (level)
    print(f'{indent}{os.path.basename(root)}/')
    subindent = ' ' * 4 * (level + 1)
    for f in files:
        if not f.startswith('.'): # Ignore hidden files
            print(f'{subindent}{f}')
    if level > 1: # Limit depth for listing to avoid long output
        break

# The actual DATA_DIR should be adjusted based on the unzipped structure.
# Assuming the zip extracts into a folder named 'CT_Scan_Dataset' like:
# /content/covid_dataset/CT_Scan_Dataset/COVID
# /content/covid_dataset/CT_Scan_Dataset/Non-COVID
# The DATA_DIR variable is updated in cell 'nPLXoMoN7YCI' to reflect this.
```

Unzipping /content/archive (3).zip to /content/covid_dataset...

Unzipping complete.

Contents of extracted directory:

covid_dataset/

 COVID/

 Covid (512).png

 Covid (583).png

 Covid (931).png

 Covid (613).png

 Covid (522).png

 Covid (149).png

 Covid (177).png

 Covid (449).png

 Covid (704).png

 Covid (96).png

 Covid (1155).png

 Covid (1090).png

 Covid (1063).png

 Covid (926).png

 Covid (1024).png

 Covid (341).png

 Covid (79).png

 Covid (468).png

 Covid (820).png

 Covid (314).png

 Covid (738).png

 Covid (1011).png

 Covid (216).png

 Covid (989).png

 Covid (723).png

 Covid (202).png

 Covid (655).png

 Covid (941).png

 Covid (450).png

 Covid (1137).png

 Covid (1031).png

 Covid (81).png

 Covid (132).png

 Covid (602).png

 Covid (313).png

 Covid (409).png

 Covid (169).png

 Covid (726).png

 Covid (546).png

```
Covid (357).png
Covid (636).png
Covid (558).png
Covid (11).png
Covid (348).png
Covid (710).png
Covid (337).png
Covid (311).png
Covid (83).png
Covid (89).png
Covid (706).png
Covid (330).png
Covid (342).png
```

```
# Augmentations e transforms
train_transform = transforms.Compose([
    transforms.Resize((IMG_SIZE, IMG_SIZE)),
    transforms.RandomHorizontalFlip(),
    transforms.RandomRotation(10),
    transforms.ColorJitter(brightness=0.1, contrast=0.1),
    transforms.ToTensor(),
    transforms.Normalize(mean=[0.485,0.456,0.406], std=[0.229,0.224,0.225])
])

val_transform = transforms.Compose([
    transforms.Resize((IMG_SIZE, IMG_SIZE)),
    transforms.ToTensor(),
    transforms.Normalize(mean=[0.485,0.456,0.406], std=[0.229,0.224,0.225])
])

# Carregar todo dataset via ImageFolder
full_dataset = datasets.ImageFolder(root=DATA_DIR, transform=train_transform)
class_names = full_dataset.classes
num_classes = len(class_names)
print("Classes:", class_names, "Num classes:", num_classes)

# Split train / val (ex.: 80/20)
val_ratio = 0.2
n_total = len(full_dataset)
n_val = int(n_total * val_ratio)
n_train = n_total - n_val

train_dataset, val_dataset = random_split(
    full_dataset,
    [n_train, n_val],
    generator=torch.Generator().manual_seed(SEED)
)

# Aplica a transformação de validação ao conjunto de validação
# Isso é necessário porque `random_split` herda a transformação do `full_dataset`
# que é `train_transform`. Precisamos aplicar `val_transform` ao `val_dataset`.
val_dataset.dataset.transform = val_transform

print(f"Tamanho train: {len(train_dataset)}, val: {len(val_dataset)}")

train_loader = DataLoader(
    train_dataset,
    batch_size=BATCH_SIZE,
    shuffle=True,
    num_workers=NUM_WORKERS,
    pin_memory=True
)

val_loader = DataLoader(
    val_dataset,
    batch_size=BATCH_SIZE,
    shuffle=False,
    num_workers=NUM_WORKERS,
    pin_memory=True
)

Classes: ['COVID', 'non-COVID'] Num classes: 2
Tamanho train: 1985, val: 496
/usr/local/lib/python3.12/dist-packages/torch/utils/data/dataloader.py:627: UserWarning: This DataLoader will create 4 workers
warnings.warn()
```

```
# Model (Transfer Learning com ResNet18)
def build_model(num_classes):
    model = models.resnet18(weights=models.ResNet18_Weights.DEFAULT)
    # Congelar todos os parâmetros do modelo pré-treinado
    for param in model.parameters():
        param.requires_grad = False
```

```

# Substituir a camada final (fully connected) para nosso número de classes
num_ftrs = model.fc.in_features
model.fc = nn.Linear(num_ftrs, num_classes)
model = model.to(DEVICE)
return model

model = build_model(num_classes)
print(model)

# Loss e Optimizer
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.fc.parameters(), lr=LR) # Apenas camada final treinável

# Scheduler para ajuste do Learning Rate
scheduler = optim.lr_scheduler.ReduceLROnPlateau(
    optimizer,
    mode='min',
    factor=0.1,
    patience=PATIENCE // 2
)

Downloading: "https://download.pytorch.org/models/resnet18-f37072fd.pth" to /root/.cache/torch/hub/checkpoints/resnet18-f37072fd.pth
100%|██████████| 44.7M/44.7M [00:00<00:00, 106MB/s]
ResNet(
  (conv1): Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2), padding=(3, 3), bias=False)
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
  (relu): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel_size=3, stride=2, padding=1, dilation=1, ceil_mode=False)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
    )
    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
    )
  )
  (layer2): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 128, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (downsample): Sequential(
        (0): Conv2d(64, 128, kernel_size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      )
    )
    (1): BasicBlock(
      (conv1): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
    )
  )
  (layer3): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(128, 256, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      (downsample): Sequential(
        (0): Conv2d(128, 256, kernel_size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running_stats=True)
      )
    )
    (1): BasicBlock(
      (conv1): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False)
    )
  )
)
```

```

def train_one_epoch(model, loader, criterion, optimizer, device):
    model.train()
    running_loss = 0.0
    running_corrects = 0
    total = 0

    for inputs, labels in tqdm(loader, desc="Train", leave=False):

```

```

for inputs, labels in tqdm(loader, desc="Train", leave=False):
    inputs = inputs.to(device)
    labels = labels.to(device)

    optimizer.zero_grad()
    outputs = model(inputs)
    loss = criterion(outputs, labels)
    loss.backward()
    optimizer.step()

    running_loss += loss.item() * inputs.size(0)
    _, preds = torch.max(outputs, 1)
    running_corrects += torch.sum(preds == labels.data).item()
    total += inputs.size(0)

epoch_loss = running_loss / total
epoch_acc = running_corrects / total
return epoch_loss, epoch_acc

```

```

def validate(model, loader, criterion, device):
    model.eval()
    running_loss = 0.0
    running_corrects = 0
    total = 0
    all_preds = []
    all_labels = []

    with torch.no_grad():
        for inputs, labels in tqdm(loader, desc="Val", leave=False):
            inputs = inputs.to(device)
            labels = labels.to(device)
            outputs = model(inputs)
            loss = criterion(outputs, labels)

            running_loss += loss.item() * inputs.size(0)
            _, preds = torch.max(outputs, 1)
            running_corrects += torch.sum(preds == labels.data).item()
            total += inputs.size(0)

            all_preds.extend(preds.cpu().numpy().tolist())
            all_labels.extend(labels.cpu().numpy().tolist())

    epoch_loss = running_loss / total
    epoch_acc = running_corrects / total
    return epoch_loss, epoch_acc, all_preds, all_labels

```

```

best_val_loss = float('inf')
best_val_acc = 0.0
patience_counter = 0

history = {
    'train_loss': [], 'train_acc': [],
    'val_loss': [], 'val_acc': []
}

for epoch in range(1, NUM_EPOCHS + 1):
    print(f"Epoch {epoch}/{NUM_EPOCHS}")
    train_loss, train_acc = train_one_epoch(model, train_loader, criterion, optimizer, DEVICE)
    val_loss, val_acc, _, _ = validate(model, val_loader, criterion, DEVICE)
    scheduler.step(val_loss) # opcional

    history['train_loss'].append(train_loss)
    history['train_acc'].append(train_acc)
    history['val_loss'].append(val_loss)
    history['val_acc'].append(val_acc)

    print(f" train_loss: {train_loss:.4f} train_acc: {train_acc:.4f}")
    print(f" val_loss: {val_loss:.4f} val_acc: {val_acc:.4f}")

    # Checkpoint by val_loss (mudar se preferir val_acc)
    if val_loss < best_val_loss:
        best_val_loss = val_loss
        best_val_acc = val_acc
        torch.save({
            'model_state_dict': model.state_dict(),
            'optimizer_state_dict': optimizer.state_dict(),
            'class_names': class_names
        }, MODEL_SAVE_PATH)
        print(" Saved best model.")
        patience_counter = 0
    else:
        patience_counter += 1
        if patience_counter == PATIENCE_THRESHOLD:
            print("Patience reached, stopping training")
            break

```

```

    patience_counter += 1
    print(f" Patience: {patience_counter}/{PATIENCE}")

    if patience_counter >= PATIENCE:
        print("Early stopping triggered.")
        break

    print("Treino finalizado. Best val_loss:", best_val_loss, "best val_acc:", best_val_acc)

```

```

Epoch 1/25
train_loss: 0.6174 train_acc: 0.6650
val_loss: 0.5535 val_acc: 0.7258
Saved best model.
Epoch 2/25
train_loss: 0.5246 train_acc: 0.7476
val_loss: 0.4785 val_acc: 0.7843
Saved best model.
Epoch 3/25
train_loss: 0.4648 train_acc: 0.8091
val_loss: 0.4396 val_acc: 0.8125
Saved best model.
Epoch 4/25
train_loss: 0.4297 train_acc: 0.8081
val_loss: 0.4083 val_acc: 0.8306
Saved best model.
Epoch 5/25
train_loss: 0.4118 train_acc: 0.8302
val_loss: 0.3915 val_acc: 0.8387
Saved best model.
Epoch 6/25
train_loss: 0.3956 train_acc: 0.8252
val_loss: 0.3704 val_acc: 0.8488
Saved best model.
Epoch 7/25
train_loss: 0.3698 train_acc: 0.8453
val_loss: 0.3637 val_acc: 0.8589
Saved best model.
Epoch 8/25
train_loss: 0.3671 train_acc: 0.8458
val_loss: 0.3478 val_acc: 0.8508
Saved best model.
Epoch 9/25
train_loss: 0.3618 train_acc: 0.8524
val_loss: 0.3726 val_acc: 0.8327
Patience: 1/6
Epoch 10/25
train_loss: 0.3546 train_acc: 0.8448
val_loss: 0.3310 val_acc: 0.8710
Saved best model.
Epoch 11/25
train_loss: 0.3326 train_acc: 0.8690
val_loss: 0.3269 val_acc: 0.8609
Saved best model.
Epoch 12/25
train_loss: 0.3492 train_acc: 0.8484
val_loss: 0.3237 val_acc: 0.8609
Saved best model.
Epoch 13/25
train_loss: 0.3352 train_acc: 0.8625
val_loss: 0.3184 val_acc: 0.8710
Saved best model.
Epoch 14/25
train_loss: 0.3168 train_acc: 0.8741
val_loss: 0.3115 val_acc: 0.8810
Saved best model.
Epoch 15/25
train_loss: 0.3205 train_acc: 0.8615

```

```

# Carregar melhor checkpoint
checkpoint = torch.load(MODEL_SAVE_PATH, map_location=DEVICE)
model.load_state_dict(checkpoint['model_state_dict'])
model.to(DEVICE)
model.eval()

# Rodar validação final para obter previsões
val_loss, val_acc, preds, labels = validate(model, val_loader, criterion, DEVICE)
print("Val loss:", val_loss, "Val acc:", val_acc)

# Relatório
print("\nClassification Report:")
print(classification_report(labels, preds, target_names=class_names))

# Matriz de confusão
cm = confusion_matrix(labels, preds)
print("Confusion Matrix:\n", cm)

```

Val loss: 0.27112853094454736 Val acc: 0.9032258064516129

```
Classification Report:
precision    recall   f1-score   support
COVID         0.93     0.86     0.89      231
non-COVID     0.88     0.94     0.91      265

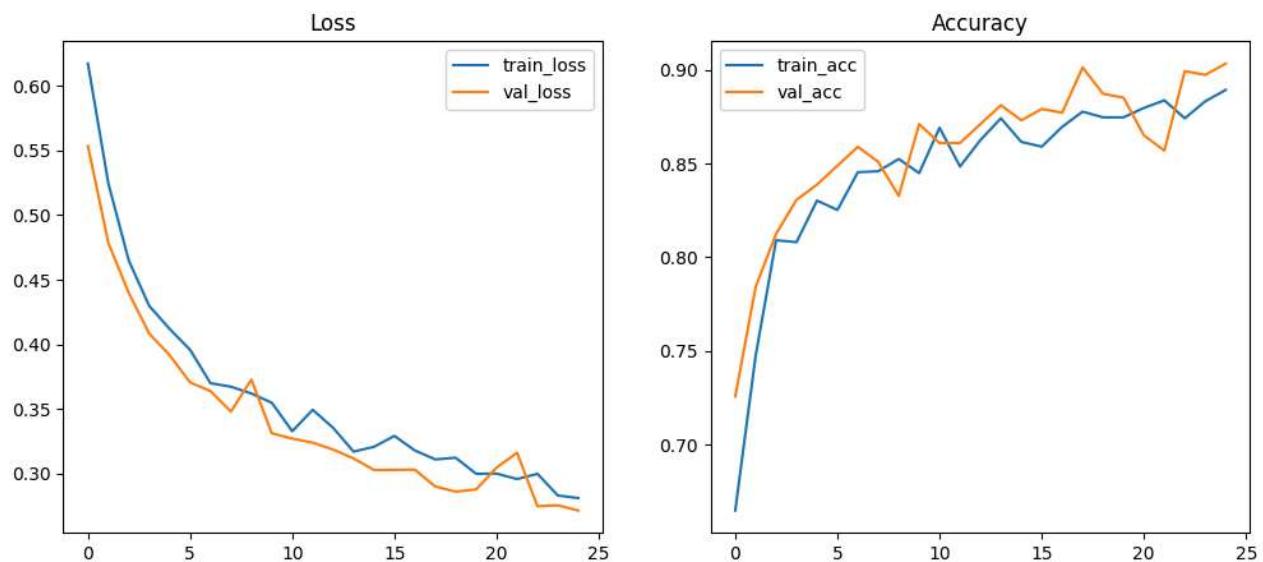
accuracy          0.90
macro avg       0.91     0.90     0.90      496
weighted avg    0.90     0.90     0.90      496
```

```
Confusion Matrix:
[[198  33]
 [ 15 250]]
```

```
# Plots simples
plt.figure(figsize=(12,5))
plt.subplot(1,2,1)
plt.plot(history['train_loss'], label='train_loss')
plt.plot(history['val_loss'], label='val_loss')
plt.legend()
plt.title('Loss')

plt.subplot(1,2,2)
plt.plot(history['train_acc'], label='train_acc')
plt.plot(history['val_acc'], label='val_acc')
plt.legend()
plt.title('Accuracy')

plt.show()
```



```
# Salvar classes em arquivo para referência futura
import json
with open("class_names.json", "w") as f:
    json.dump(class_names, f, ensure_ascii=False, indent=2)
print("Class names saved to class_names.json")

print("Pronto. Para testar previsões individuais, carregue imagem, aplique val_transform e execute model(img.unsqueeze(0)).")

Class names saved to class_names.json
Pronto. Para testar previsões individuais, carregue imagem, aplique val_transform e execute model(img.unsqueeze(0)).
```

Comece a programar ou gere código com IA.

