

```

# pip install torch torchvision torchmetrics pytorch-lightning tqdm
# numpy matplotlib tensorboard tensorboardX

import torch
import torchvision
from torchvision import transforms
import torchmetrics
import pytorch_lightning as pl
from pytorch_lightning.callbacks import ModelCheckpoint
from pytorch_lightning.loggers import TensorBoardLogger
from tqdm.notebook import tqdm
import numpy as np
import matplotlib.pyplot as plt

# Load dataset

def load_file(path):
    return np.load(path).astype(np.float32)

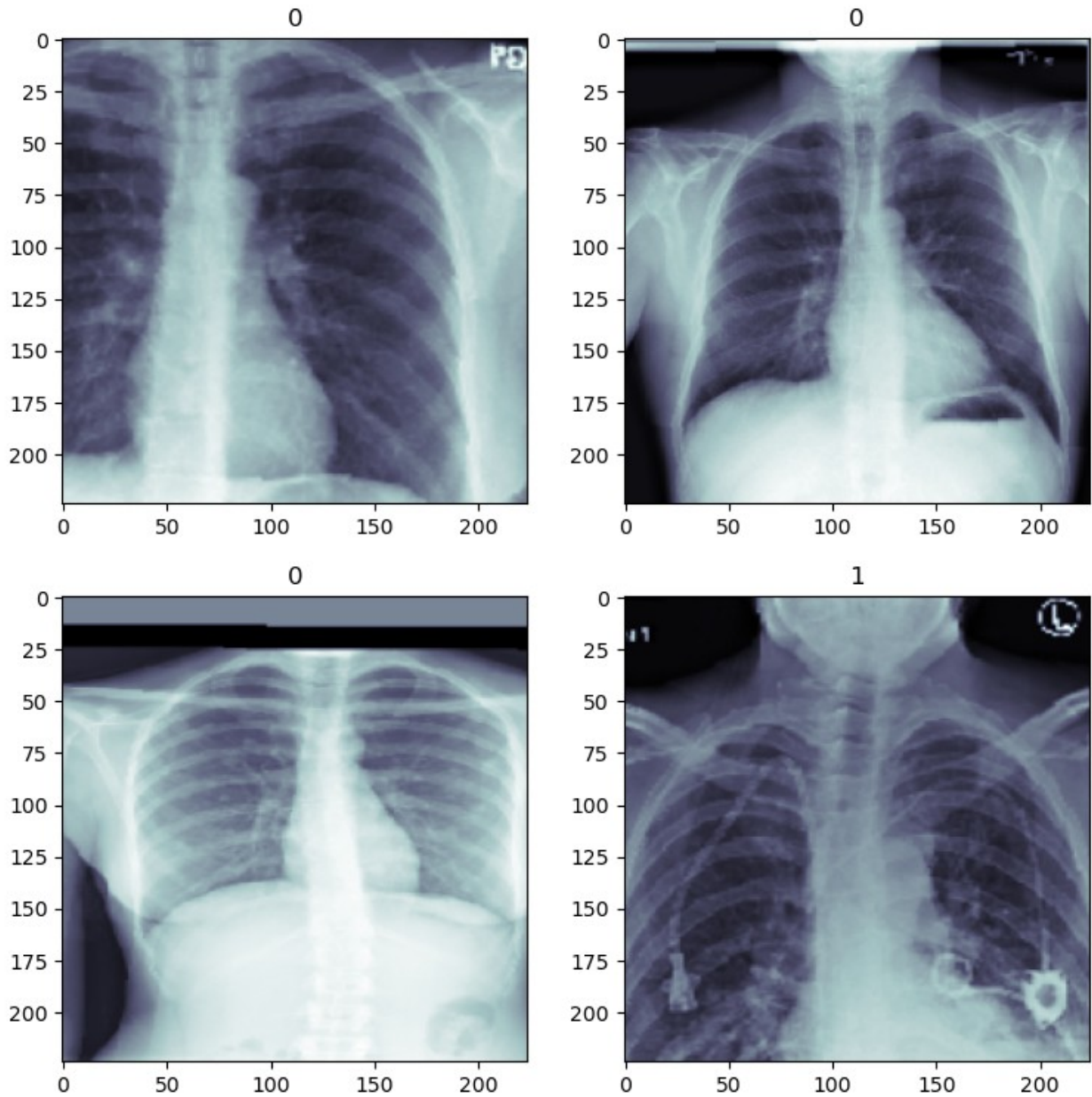
train_transforms = transforms.Compose([
    transforms.ToTensor(),
    transforms.Normalize(0.49, 0.248),
    transforms.RandomAffine(degrees=(-5, 5), translate=(0, 0.05),
scale=(0.9, 1.1)),
    transforms.RandomResizedCrop((224,224), scale=(0.35,1))
])

val_transforms = transforms.Compose([
    transforms.ToTensor(),
    transforms.Normalize(0.49, 0.248),
])

train_dataset = torchvision.datasets.DatasetFolder("processed/train/",
loader=load_file, extensions="npy", transform=train_transforms)
val_dataset = torchvision.datasets.DatasetFolder("processed/val/",
loader=load_file, extensions="npy", transform=val_transforms)

fig, axis = plt.subplots(2, 2, figsize=(9,9))
for i in range(2):
    for j in range(2):
        random_index = np.random.randint(0,24000)
        x_ray, label = train_dataset[random_index]
        axis[i][j].imshow(x_ray[0], cmap="bone")
        axis[i][j].set_title(label)

```



```
batch_size = 64
num_workers = 4
```

```
train_loader = torch.utils.data.DataLoader(train_dataset,
batch_size=batch_size, num_workers=num_workers, shuffle=True)
val_loader = torch.utils.data.DataLoader(val_dataset,
batch_size=batch_size, num_workers=num_workers, shuffle=False)
```

```
np.unique(train_dataset.targets, return_counts=True) #there are aprox
three times more cases without Phneumonia than cases with Phneumonia,
this is an unbalaced dataset
```

```
(array([0, 1]), array([18593, 5407]))
```

`torchvision.models.resnet18()` *#see the structure of the previously trained model i will adapt*

```
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)
            (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)
        )
    )
    (layer4): Sequential(
        (0): BasicBlock(

```

```

        (conv1): Conv2d(256, 512, kernel_size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
        (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (relu): ReLU(inplace=True)
        (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
        (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (downsample): Sequential(
          (0): Conv2d(256, 512, kernel_size=(1, 1), stride=(2, 2),
bias=False)
          (1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        )
      )
    (1): BasicBlock(
      (conv1): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    )
  )
  (avgpool): AdaptiveAvgPool2d(output_size=(1, 1))
  (fc): Linear(in_features=512, out_features=1000, bias=True)
)

```

Train model

```

class PneumoniaModel(pl.LightningModule):
    def __init__(self):
        super().__init__()
        self.model = torchvision.models.resnet18()
        self.model.conv1 = torch.nn.Conv2d(1, 64, kernel_size=(7, 7),
stride=(2, 2), padding=(3, 3), bias=False)
        self.model.fc = torch.nn.Linear(in_features=512,
out_features=1, bias=True)

        self.optimizer = torch.optim.Adam(self.model.parameters(),
lr=1e-4)
        self.loss_fn =
torch.nn.BCEWithLogitsLoss(pos_weight=torch.tensor([3]))

        self.train_acc = torchmetrics.Accuracy(task="binary")
        self.val_acc = torchmetrics.Accuracy(task="binary")

```

```

def forward(self, data):
    pred = self.model(data)
    return pred

def training_step(self, batch, batch_idx):
    x_ray, label = batch
    label = label.float()
    pred = self(x_ray)[: ,0] # forward
    loss = self.loss_fn(pred, label)

    self.log("Train Logs", loss)
    self.log("Step Train ACC", self.train_acc(torch.sigmoid(pred),
label.int()))

    return loss

def on_train_epoch_end(self):
    self.log("Train ACC", self.train_acc.compute())

def validation_step(self, batch, batch_idx):
    x_ray, label = batch
    label = label.float()
    pred = self(x_ray)[: ,0] # forward
    loss = self.loss_fn(pred, label)

    self.log("Val Logs", loss)
    val_acc = self.val_acc(torch.sigmoid(pred), label.int())
    self.log("Val ACC", val_acc, prog_bar=True, on_epoch=True)
    return loss

def on_val_epoch_end(self):
    self.log("Val ACC", self.val_acc.compute())

def configure_optimizers(self):
    return [self.optimizer]

model = PneumoniaModel()

checkpoint_callback = ModelCheckpoint(
    monitor= "Val ACC",
    save_top_k = 10,
    mode = "max",)

trainer = pl.Trainer(
    accelerator="gpu",
    devices=1,
    logger=TensorBoardLogger(save_dir="./logs"),
    log_every_n_steps=1,
    callbacks=checkpoint_callback,

```

```
max_epochs=35
)
```

```
GPU available: True (cuda), used: True
TPU available: False, using: 0 TPU cores
HPU available: False, using: 0 HPUs
```

```
trainer.fit(model, train_loader, val_loader)
```

```
LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES: [0]
```

	Name	Type	Params	Mode
0	model	ResNet	11.2 M	train
1	loss_fn	BCEWithLogitsLoss	0	train
2	train_acc	BinaryAccuracy	0	train
3	val_acc	BinaryAccuracy	0	train

```
11.2 M    Trainable params
0         Non-trainable params
11.2 M    Total params
44.683    Total estimated model params size (MB)
71        Modules in train mode
0         Modules in eval mode
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "bc05d77fbf95461ba30d16ef9dfe7c79", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

```
{"model_id": "", "version_major": 2, "version_minor": 0}
```

[illegible]


```

PneumoniaModel(
  (model): ResNet(
    (conv1): Conv2d(1, 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)
      (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)
    )
  )
  (layer4): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(256, 512, kernel_size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)

```

```

        (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
        (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (downsample): Sequential(
          (0): Conv2d(256, 512, kernel_size=(1, 1), stride=(2, 2),
bias=False)
          (1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        )
      )
      (1): BasicBlock(
        (conv1): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
        (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
        (relu): ReLU(inplace=True)
        (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
        (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      )
      (avgpool): AdaptiveAvgPool2d(output_size=(1, 1))
      (fc): Linear(in_features=512, out_features=1, bias=True)
    )
    (loss_fn): BCEWithLogitsLoss()
    (train_acc): BinaryAccuracy()
    (val_acc): BinaryAccuracy()
  )

preds = []
labels = []

with torch.no_grad():
    for data, label in tqdm(val_dataset):
        data = data.to(device).float().unsqueeze(0)
        pred = torch.sigmoid(model(data)[0]).cpu()
        preds.append(pred)
        labels.append(label)
preds = torch.tensor(preds)
labels = torch.tensor(labels).int()

{"model_id": "de76db55ae974a7a9dfd99b246e198db", "version_major": 2, "version_minor": 0}

import pandas as pd
from torchmetrics import Accuracy, Precision, Recall, ConfusionMatrix

acc_metric = Accuracy(task="binary")

```

```
precision_metric = Precision(task="binary")
recall_metric    = Recall(task="binary")
cm_metric        = ConfusionMatrix(task="binary", num_classes=2)
```

```
acc      = acc_metric(preds, labels).item()
precision = precision_metric(preds, labels).item()
recall   = recall_metric(preds, labels).item()
cm       = cm_metric(preds, labels).numpy()
```

```
df = pd.DataFrame({
    "Accuracy":      [acc],
    "Precision":     [precision],
    "Recall":        [recall],
    "Confusion Mat.": [cm.tolist()]
})
```

```
df
```

	Accuracy	Precision	Recall	Confusion Mat.
0	0.805514	0.544768	0.834711	[[1657, 422], [100, 505]]

```
# https://www.kaggle.com/competitions/rsna-pneumonia-detection-challenge
```

```
import pathlib
import pydicom
import numpy as np
import cv2
import pandas as pd
import matplotlib.pyplot as plt
from tqdm.notebook import tqdm
import os
```

```
labels =
pd.read_csv("./rsna-pneumonia-detection-challenge/stage_2_train_labels.csv")
labels.head(6)
```

	patientId	x	y	width	height
Target					
0	0004cfab-14fd-4e49-80ba-63a80b6bddd6	NaN	NaN	NaN	NaN
0					
1	00313ee0-9eaa-42f4-b0ab-c148ed3241cd	NaN	NaN	NaN	NaN
0					
2	00322d4d-1c29-4943-afc9-b6754be640eb	NaN	NaN	NaN	NaN
0					
3	003d8fa0-6bf1-40ed-b54c-ac657f8495c5	NaN	NaN	NaN	NaN
0					
4	00436515-870c-4b36-a041-de91049b9ab4	264.0	152.0	213.0	379.0
1					
5	00436515-870c-4b36-a041-de91049b9ab4	562.0	152.0	256.0	453.0
1					

```
labels = labels.drop_duplicates("patientId")
```

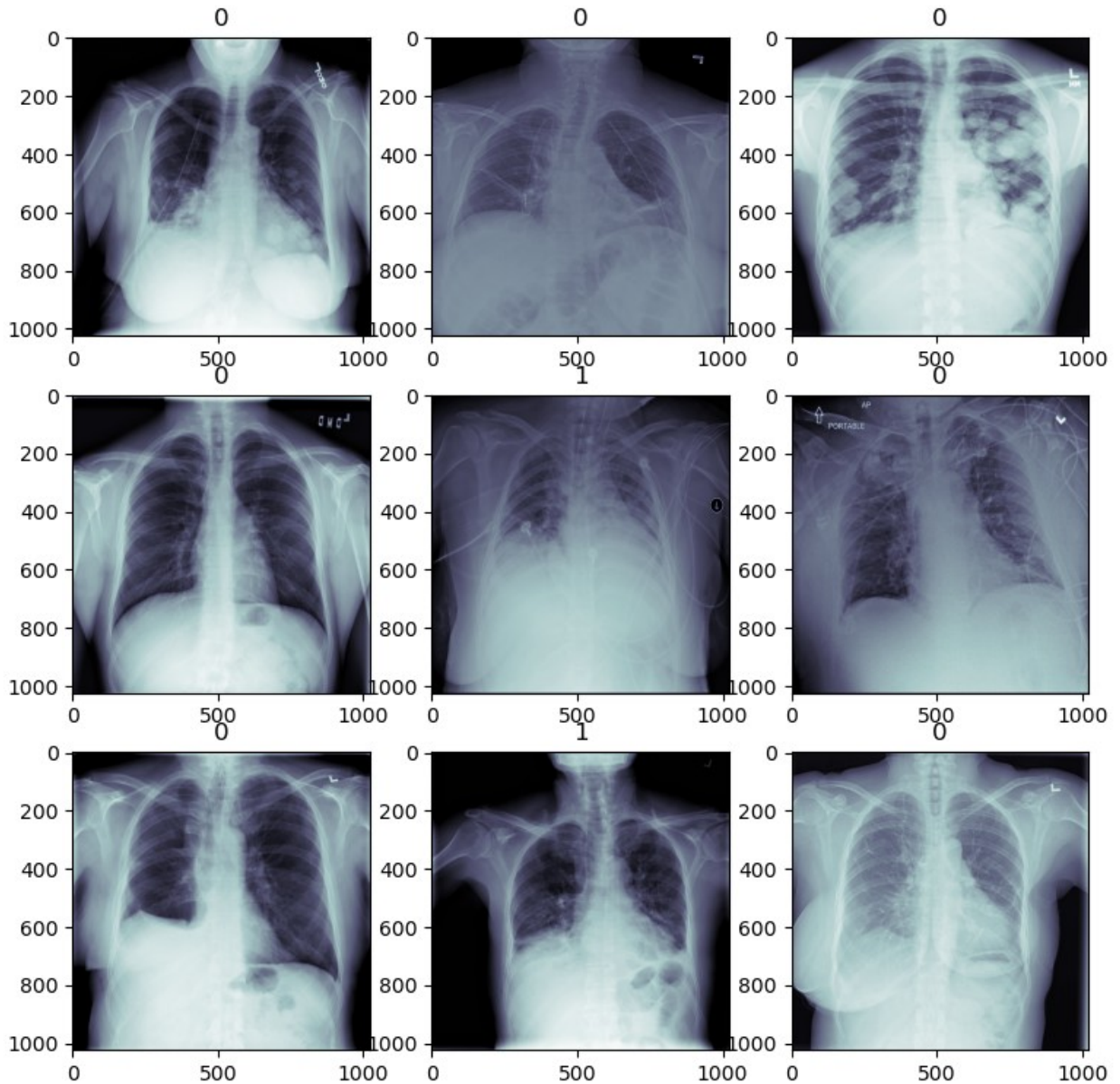
```
ROOT_PATH =
pathlib.Path("./rsna-pneumonia-detection-challenge/stage_2_train_images/")
```

```
SAVE_PATH = pathlib.Path("./processed")
```

```
fig, axis = plt.subplots(3, 3, figsize=(9, 9))
c = 0
```

```
for i in range(3):
    for j in range(3):
        patient_id = labels.patientId.iloc[c]
        dcm_path = ROOT_PATH/patient_id
        dcm_path = dcm_path.with_suffix(".dcm")
        dcm = pydicom.dcmread(dcm_path).pixel_array
        label = labels["Target"].iloc[c]
        axis[i][j].imshow(dcm, cmap="bone")
        axis[i][j].set_title(label)
```

```
c+=1
```



```
sums, sums_squared = 0, 0

for c, patient_id in enumerate(tqdm(labels.patientId)):
    patient_id = labels.patientId.iloc[c]
    dcm_path = ROOT_PATH/patient_id
    dcm_path = dcm_path.with_suffix(".dcm")
    dcm = pydicom.dcmread(dcm_path).pixel_array / 255
    dcm_array = cv2.resize(dcm, (224,224)).astype(np.float16)
```

```

label = labels.Target.iloc[c]
train_or_val = "train" if c < 24000 else "val"
current_save_path = SAVE_PATH/train_or_val/str(label)
current_save_path.mkdir(parents=True, exist_ok=True)
np.save(current_save_path/patient_id, dcm_array)
normalizer = 224*224
if train_or_val == "train":
    sums += np.sum(dcm_array) / normalizer
    sums_squared += (dcm_array ** 2).sum() / normalizer

{"model_id": "4db04e85af3646d1ad7af2640238b35e", "version_major": 2, "version_minor": 0}

mean = sums / 24000
std = np.sqrt((sums_squared / 24000) - mean**2)
print(mean, std)

0.4903962485384803 0.24795070634161256

```