```
In [1]: 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
 In [2]: def load_file(path):
            return np.load(path).astype(np.float32)
 In [3]: 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]),
In [13]: train_dataset = torchvision.datasets.DatasetFolder(
             r"C:\Users\kisha\OneDrive\Desktop\Projects\MedAI\AI-IN-MEDICAL-MATERIALS\04-Pneumonia-ClassificationProcessed/train/",
            loader=load_file, extensions="npy", transform=train_transforms)
         val_dataset = torchvision.datasets.DatasetFolder(
            r"C:\Users\kisha\OneDrive\Desktop\Projects\MedAI\AI-IN-MEDICAL-MATERIALS\04-Pneumonia-ClassificationProcessed/val/",
            loader=load_file, extensions="npy", transform=val_transforms)
In [30]: 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, 20000)
                x_ray, label = train_dataset[random_index]
                axis[i][j].imshow(x_ray[0], cmap="bone")
                axis[i][j].set_title(f"Label:{label}")
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 In [6]: batch_size = 64
         num_workers = 0
         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)
         print(f"There are {len(train_dataset)} train images and {len(val_dataset)} val images")
        There are 24000 train images and 2684 val images
 In [7]: np.unique(train_dataset.targets, return_counts=True), np.unique(val_dataset.targets, return_counts=True)
 Out[7]: ((array([0, 1]), array([18593, 5407], dtype=int64)),
          (array([0, 1]), array([2079, 605], dtype=int64)))
 In [8]: class PneumoniaModel(pl.LightningModule):
            def __init__(self, weight=1):
                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)
                self.optimizer = torch.optim.Adam(self.model.parameters(), lr=1e-4)
                self.loss_fn = torch.nn.BCEWithLogitsLoss(pos_weight=torch.tensor([weight]))
                # simple accuracy computation
                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() # Convert label to float (just needed for loss computation)
                 pred = self(x_ray)[:,0] # Prediction: Make sure prediction and label have same shape
                #while using pytorch lightening we don't need to write self.forward
                #our prediction are of shape (batchsize,1) but our lebels are only of shape batchsize
                 #hence remove second dimension from prediction by slicing
                loss = self.loss_fn(pred, label) # Compute the loss
                # Log loss and batch accuracy
                 self.log("Train Loss", loss)
                 self.log("Step Train Acc", self.train_acc(torch.sigmoid(pred), label.int()))
                return loss
            def on_train_epoch_end(self):
                # After one epoch compute the whole train_data accuracy
                self.log("Train Acc", self.train_acc.compute())
             def validation_step(self, batch, batch_idx):
                 # Same steps as in the training_step
                x_ray, label = batch
                label = label.float()
                pred = self(x_ray)[:,0] # to make sure prediction and label have same shape
                loss = self.loss_fn(pred, label)
                # Log validation metrics
                self.log("Val Loss", loss)
                self.log("Step Val Acc", self.val_acc(torch.sigmoid(pred), label.int()))
                return loss
            def on_validation_epoch_end(self):
                 self.log("Val Acc", self.val_acc.compute())
            def configure_optimizers(self):
                 #Caution! You always need to return a list here (just pack your optimizer into one :))
                return [self.optimizer]
 In [9]: model = PneumoniaModel()
In [10]: # Create the checkpoint callback
         checkpoint_callback = ModelCheckpoint(
            monitor='Val Acc',
            save_top_k=10,
            mode='max')
In [11]: # Create the trainer
         # Change the gpus parameter to the number of available gpus on your system. Use 0 for CPU training
         #here provide whole path to your logs directory
         trainer = pl.Trainer(devices=[0], accelerator="gpu", logger=TensorBoardLogger(save_dir=r"C:\Users\kisha\OneDrive\Desktop\Projects\MedAI\AI-IN-MEDICAL-MATERIALS\04-Pneumonia-Classification/logs"), log_every_n_steps=1,
                             callbacks=checkpoint_callback,
                             max_epochs=15)
        GPU available: True (cuda), used: True
        TPU available: False, using: 0 TPU cores
        IPU available: False, using: 0 IPUs
        HPU available: False, using: 0 HPUs
In [12]: trainer.fit(model, train_loader, val_loader)
        You are using a CUDA device ('NVIDIA GeForce RTX 3050 Laptop GPU') that has Tensor Cores. To properly utilize them, you should set `torch.set_float32_matmul_precision('medium' | 'high')` which will trade-off precision for performance. For more details, read https://pyto
        rch.org/docs/stable/generated/torch.set_float32_matmul_precision.html#torch.set_float32_matmul_precision
        LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES: [0]
         | Name | Type
                                        | Params
        0 | model | ResNet
                                     | 11.2 M
       1 | loss_fn | BCEWithLogitsLoss | 0
       2 | train_acc | BinaryAccuracy | 0
       3 | val_acc | BinaryAccuracy | 0
        -----
       11.2 M Trainable params
                 Non-trainable params
       11.2 M Total params
       44.683 Total estimated model params size (MB)
        Sanity Checking: |
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        C:\Users\kisha\anaconda3\envs\kishan\lib\site-packages\pytorch_lightning\trainer\connectors\data_connector.py:441: The 'val_dataloader' does not have many workers which may be a bottleneck. Consider increasing the value of the `num_workers` argument` to `num_workers=15`
        in the `DataLoader` to improve performance.
        C:\Users\kisha\anaconda3\envs\kishan\lib\site-packages\pytorch_lightning\trainer\connector.py:441: The 'train_dataloader' does not have many workers which may be a bottleneck. Consider increasing the value of the `num_workers` argument` to `num_workers=1
       5` in the `DataLoader` to improve performance.
        Training: |
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        Validation:
        `Trainer.fit` stopped: `max_epochs=15` reached.
In [14]: device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
         #We evaluate to calculate accuracy, precision, recall and confusion matrix
         # Use strict=False, otherwise we would want to match the pos_weight which is not necessary
         #write path correctly
         model = PneumoniaModel.load_from_checkpoint(r"C:\Users\kisha\OneDrive\Desktop\Projects\MedAI\AI-IN-MEDICAL-MATERIALS\04-Pneumonia-Classification\weights\weights_3.ckpt")
         model.eval()
         model.to(device);
        C:\Users\kisha\anaconda3\envs\kishan\lib\site-packages\pytorch_lightning\utilities\migration.py:208: You have multiple `ModelCheckpoint, but we found state keys that would end up colliding with each other after an upgrade, w
        hich means we can't differentiate which of your checkpoint callbacks needs which states. At least one of your `ModelCheckpoint` callbacks will not be able to reload the state.
        Lightning automatically upgraded your loaded checkpoint from v1.3.4 to v2.2.1. To apply the upgrade to your files permanently, run `python -m pytorch_lightning.utilities.upgrade_checkpoint C:\Users\kisha\OneDrive\Desktop\Projects\MedAI\AI-IN-MEDICAL-MATERIALS\04-Pneumon
        ia-Classification\weights\weights_3.ckpt
In [15]: 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()
                      | 0/2684 [00:00<?, ?it/s]
In [20]: | acc = torchmetrics.Accuracy(task="binary")(preds, labels)
         precision = torchmetrics.Precision(task="binary")(preds, labels)
         recall = torchmetrics.Recall(task="binary")(preds, labels)
         cm = torchmetrics.ConfusionMatrix(num_classes=2, task="binary")(preds, labels)
         cm_threshed = torchmetrics.ConfusionMatrix(num_classes=2, threshold=0.25,task="binary")(preds, labels)
         print(f"Val Accuracy: {acc}")
         print(f"Val Precision: {precision}")
         print(f"Val Recall: {recall}")
         print(f"Confusion Matrix:\n {cm}")
         print(f"Confusion Matrix 2:\n {cm_threshed}")
         #recall is much larger, this means our model rarely misses a case of pneumonia
         #however low precision means lots of False positive
         #we can rerun the complete evaluation with the model which was not trained with weighted loss
         #this time it will fails to detect many amount of pneumonias
        Val Accuracy: 0.7652757167816162
        Val Precision: 0.48819640278816223
        Val Recall: 0.8545454740524292
        Confusion Matrix:
        tensor([[1537, 542],
                [ 88, 517]])
        Confusion Matrix 2:
         tensor([[1135, 944],
               [ 23, 582]])
In [32]: fig, axis = plt.subplots(2, 2, figsize=(4, 4))
         for i in range(2):
            for j in range(2):
                rnd_idx = np.random.randint(0, len(preds))
                axis[i][j].imshow(val_dataset[rnd_idx][0][0], cmap="bone")
                axis[i][j].set_title(f"Pred:{int(preds[rnd_idx] > 0.5)}, Label:{labels[rnd_idx]}")
                axis[i][j].axis("off")
         Pred:1, Label:1
                              Pred:1, Label:1
```

Pred:1, Label:0

Pred:0, Label:0

In []:
In []: