This Assignment was done by:

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RSNA Pneumonia Detection

1- The dataset provided in two folder:

- train_images.zip
- test_images.zip.

2- train.csv:

- the training set. Contains patientIds and bounding box / target information.
- patientld: A patientld. Each patientld corresponds to a unique image.
- x:- the upper-left x coordinate of the bounding box.
- y:- the upper-left y coordinate of the bounding box.
- width:-the width of the bounding box.
- height:- the height of the bounding box.
- Target :- the binary Target, indicating whether this sample has evidence of pneumonia

First we view the data

```
In [ ]: import numpy as np
   import matplotlib.pyplot as plt
   import pandas as pd
   import pydicom
   from PIL import Image
   import os
   from ultralytics import YOLO
```

```
In [ ]: df = pd.read_csv('dataset/train_labels.csv')
    df.head()
```

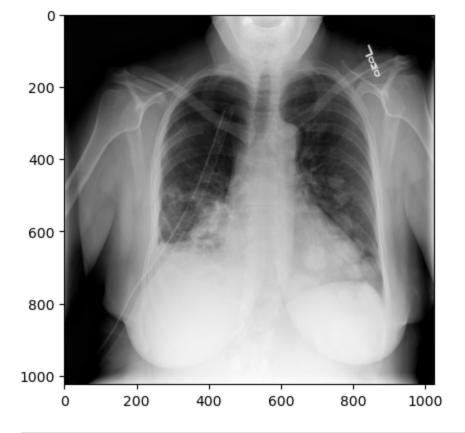
```
0004cfab-14fd-4e49-80ba-63a80b6bddd6
                                                  NaN
                                                         NaN
                                                                NaN
                                                                       NaN
                                                                                 0
            00313ee0-9eaa-42f4-b0ab-c148ed3241cd
                                                                       NaN
                                                                                 0
                                                   NaN
                                                         NaN
                                                                NaN
         2 00322d4d-1c29-4943-afc9-b6754be640eb
                                                                NaN
                                                                       NaN
                                                                                 0
                                                  NaN
                                                         NaN
         3
             003d8fa0-6bf1-40ed-b54c-ac657f8495c5
                                                   NaN
                                                                       NaN
                                                         NaN
                                                                NaN
                                                                                 0
         4 00436515-870c-4b36-a041-de91049b9ab4
                                                 264.0
                                                        152.0
                                                                      379.0
                                                                                  1
                                                               213.0
In [ ]: | # replace nans with 0
         df.fillna(0, inplace=True)
         df.head()
                                                           y width height Target
Out[]:
                                       patientId
                                                     Х
            0004cfab-14fd-4e49-80ba-63a80b6bddd6
                                                   0.0
                                                          0.0
                                                                 0.0
                                                                         0.0
                                                                                 0
            00313ee0-9eaa-42f4-b0ab-c148ed3241cd
                                                   0.0
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                                                                 0.0
                                                                         0.0
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         2 00322d4d-1c29-4943-afc9-b6754be640eb
                                                   0.0
                                                          0.0
                                                                 0.0
                                                                         0.0
                                                                                 0
             003d8fa0-6bf1-40ed-b54c-ac657f8495c5
         3
                                                   0.0
                                                          0.0
                                                                 0.0
                                                                         0.0
                                                                                 0
         4 00436515-870c-4b36-a041-de91049b9ab4 264.0 152.0
                                                               213.0
                                                                      379.0
                                                                                  1
In [ ]: df.describe()
Out[]:
                           Х
                                         У
                                                   width
                                                                height
                                                                              Target
         count 30227.000000 30227.000000 30227.000000 30227.000000
         mean
                  124.561683
                                115.960962
                                               69.060575
                                                            104.084825
                                                                            0.316108
           std
                  216.326397
                                190.012883
                                              106.910496
                                                            176.932152
                                                                            0.464963
                    0.000000
                                  0.000000
                                                0.000000
                                                              0.000000
                                                                            0.000000
           min
          25%
                    0.000000
                                  0.000000
                                                0.000000
                                                              0.000000
                                                                            0.000000
          50%
                    0.000000
                                  0.000000
                                                0.000000
                                                              0.000000
                                                                            0.000000
          75%
                  193.000000
                                231.000000
                                              169.000000
                                                            188.000000
                                                                            1.000000
                  835.000000
                                881.000000
                                                            942.000000
                                                                            1.000000
                                              528.000000
          max
In [ ]: # Plotting the first image
         path to image = df['patientId'][0]
         # Load the .dcm image
         ds = pydicom.dcmread(f'dataset/train images/{path to image}.dcm')
         # Access image data
         image = ds.pixel array
         # Display the image
         plt.imshow(image, cmap='gray')
         plt.show()
```

patientId

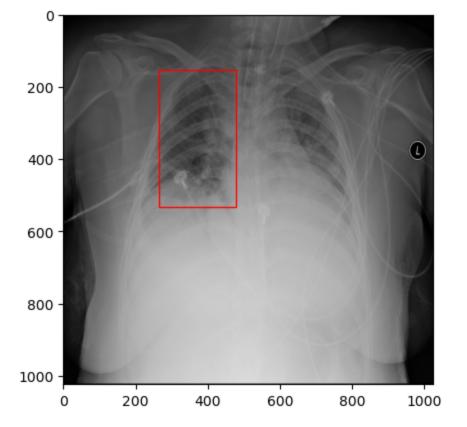
Χ

width height Target

Out[]:



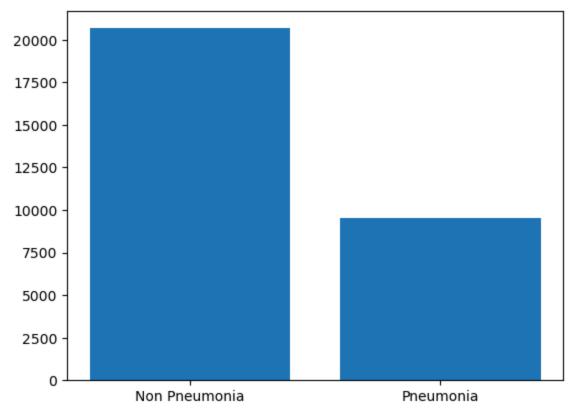
```
In [ ]: # Plotting the first image with bounding box
        # Load the .dcm image
        path_to_image = df['patientId'][4]
        ds = pydicom.dcmread(f'dataset/train images/{path to image}.dcm')
        # Access image data
        image = ds.pixel array
        # Display the image
        plt.imshow(image, cmap='gray')
        # Add bounding box
        # Get the target values
        x = df['x'][4]
        y = df['y'][4]
        width = df['width'][4]
        height = df['height'][4]
        # Create a Rectangle patch
        rect = plt.Rectangle((x, y), width, height, edgecolor='r', facecolor='none')
        # Add the patch to the Axes
        plt.gca().add_patch(rect)
        plt.show()
```



```
In [ ]: # compare between the number of images with pneumonia and without pneumonia
    print(df['Target'].value_counts())
    plt.bar(['Non Pneumonia', 'Pneumonia'], df['Target'].value_counts());
```

Target 0 20672 1 9555

Name: count, dtype: int64



In []: # count the .dcm files in the dataset/train_images folder
print(len(os.listdir('dataset/train_images')))

```
In []: # check if any of the patientId in the train_labels.csv file are in the dataset/test_ima
    test_images = os.listdir('dataset/test_images')
    for patientId in df['patientId']:
        if f'{patientId}.dcm' in test_images:
            print(patientId)

In []: df.shape

Out[]: (30227, 6)

In []: # count the number of unique patients
    df['patientId'].nunique()

Out[]: 26684

In []: print(len(os.listdir('dataset/test_images')))
    3000
```

• Nan values are replaced with zeros

- number of images with non-pneumonia are double the images with pneumonia (imbalanced data)
- number of unique patients in the dataframe is equal to the number of training images (no redundancy)

Convert the .dcm images to a format that YOLOv8 can work with

- First, we extract the image data from .dcm file
- Second, we will Normalize the pixel values to be from 0 255 (leading to faster convergence)
- we then convert it to jpg format

```
In []: # convert the .dcm files to .jpg files

def convert_dcm_to_jpg(input_folder, output_folder):
    for file in os.listdir(input_folder):
        if os.path.isfile(f'{output_folder}/{file.split(".")[0]}.jpg'):
            continue
        ds = pydicom.dcmread(f'{input_folder}/{file}')
        img = ds.pixel_array
        # Normalize the pixel values to the range [0, 255]
        image_normalized = (img - np.min(img)) / (np.max(img) - np.min(img)) * 255
        image_normalized = image_normalized.astype(np.uint8) # Convert to unsigned 8-bi

# Convert the numpy array to a PIL image
        image = Image.fromarray(image_normalized)

filename = file.split('.')[0]
        image.save(f'{output_folder}/{filename}.jpg')
In []: convert dcm to jpg('dataset/train images', 'dataset/images/train')
```

```
In [ ]: convert_dcm_to_jpg('dataset/train_images', 'dataset/images/train')
In [ ]: convert_dcm_to_jpg('dataset/test_images', 'dataset/images/test')
```

Normalizing the labels (YOLO Format)

```
In []: def normalize_bboxs(df):
    img_size = 1024

# normalize the bounding boxes
    df['width'] = df['width'] / img_size
    df['height'] = df['height'] / img_size

# Centering X , Y coordinates and normalize them
    df['x'] = df['x'] / img_size + df['width'] / 2
    df['y'] = df['y'] / img_size + df['height'] / 2

def denormalize_bbox(rx, ry, rw, rh):
    img_size = 1024

    x = (rx-rw/2)*img_size
    y = (ry-rh/2)*img_size
    w = rw*img_size
    h = rh*img_size
    return x, y, w, h
```

Out[]:		patientId	х	у	width	height	Target
	0	0004cfab-14fd-4e49-80ba-63a80b6bddd6	0.000000	0.000000	0.000000	0.000000	0
	1	00313ee0-9eaa-42f4-b0ab-c148ed3241cd	0.000000	0.000000	0.000000	0.000000	0
	2	00322d4d-1c29-4943-afc9-b6754be640eb	0.000000	0.000000	0.000000	0.000000	0
	3	003d8fa0-6bf1-40ed-b54c-ac657f8495c5	0.000000	0.000000	0.000000	0.000000	0
	4	00436515-870c-4b36-a041-de91049b9ab4	0.361816	0.333496	0.208008	0.370117	1

Preparing The labels folder

```
In [ ]: # Converting the df to .txt files per patientId each patientId in separate file (as ther
        def convert df to txt files(df, output folder):
            if not os.path.exists(output folder):
                os.makedirs(output folder)
            for patientId in df['patientId'].unique():
                patient df = df[df['patientId'] == patientId].copy()
                if patient_df['Target'].values[0] == 0:
                    # if the patient doesn't have pneumonia, output empty txt file
                    with open(f'{output folder}/{patientId}.txt', 'w') as f:
                        f.close()
                    continue
                patient df.drop(columns='patientId', inplace=True)
                # getting the target column to the first column
                cols = patient df.columns.tolist()
                cols = cols[-1:] + cols[:-1]
                patient df = patient df[cols]
```

Splitting the data into train_1, train_2, train_3 and validation (Hardware constraints)

```
In [ ]: ## split the data into train 1 , train 2 , train 3 and validation sets and without shuff
        # get the unique patientIds
        patientIds = df['patientId'].unique()
        train 1 = patientIds[:9000]
        train 2 = patientIds[9000:18000]
        train 3 = patientIds[18000:24000]
        validation = patientIds[24000:]
        # get the patientIds for each set
        train 1 df = df[df['patientId'].isin(train 1)].copy()
        train 2 df = df[df['patientId'].isin(train 2)].copy()
        train 3 df = df[df['patientId'].isin(train 3)].copy()
        validation df = df[df['patientId'].isin(validation)].copy()
        # convert the train_1_df, train_2_df, train_3_df and validation_df to .txt files
        # convert df to txt files(train 1 df, 'dataset/labels/train 1')
        # convert df to txt files(train 2 df, 'dataset/labels/train 2')
        # convert df to txt files(train 3 df, 'dataset/labels/train 3')
        # convert df to txt files(validation df, 'dataset/labels/validation')
        # copy the images to the respective folders
        def copy_images_to_folder(df, input_folder, output folder):
            if not os.path.exists(output folder):
                os.makedirs(output folder)
            for patientId in df['patientId'].unique():
                os.system(f'cp {input folder}/{patientId}.jpg {output folder}/{patientId}.jpg')
        # copy_images_to_folder(train_1_df, 'dataset/images/train', 'dataset/images/train_1')
        # copy_images_to_folder(train_2_df, 'dataset/images/train', 'dataset/images/train_2')
        # copy images to folder(train 3 df, 'dataset/images/train', 'dataset/images/train 3')
        # copy images to folder(validation df, 'dataset/images/train', 'dataset/images/validatio
```

Training The YOLOv8 Algorithm

- we ran each part (train_1 , train_2 , train_3) of the data for two epochs to achieve better results
- at the end of each part we get the best.pt from the previous one for the next training session
- that lead to better losses, precision and recall

```
In [ ]: model = Y0L0("dataset/best3.pt")
    res = model.train(data="data.yaml", epochs=1) # one final epoch to finish training
```

New https://pypi.org/project/ultralytics/8.2.19 available $\stackrel{\textcircled{\tiny 4}}{}$ Update with 'pip install -U ultralytics'

Ultralytics YOLOv8.2.18 ₹ Python-3.11.5 torch-2.2.2 CUDA:0 (NVIDIA GeForce GTX 1650, 390 4MiB)

engine/trainer: task=detect, mode=train, model=dataset/best3.pt, data=data.yaml, epochs= 1, time=None, patience=100, batch=16, imgsz=640, save=True, save period=-1, cache=False, device=None, workers=8, project=None, name=train, exist ok=False, pretrained=True, optimi zer=auto, verbose=True, seed=0, deterministic=True, single cls=False, rect=False, cos lr= False, close mosaic=10, resume=False, amp=True, fraction=1.0, profile=False, freeze=None, multi scale=False, overlap mask=True, mask ratio=4, dropout=0.0, val=True, split=val, sav e json=False, save hybrid=False, conf=None, iou=0.7, max det=300, half=False, dnn=False, plots=True, source=None, vid stride=1, stream buffer=False, visualize=False, augment=Fals e, agnostic nms=False, classes=None, retina masks=False, embed=None, show=False, save fra mes=False, save txt=False, save conf=False, save crop=False, show labels=True, show conf= True, show boxes=True, line width=None, format=torchscript, keras=False, optimize=False, int8=False, dynamic=False, simplify=False, opset=None, workspace=4, nms=False, lr0=0.01, lrf=0.01, momentum=0.937, weight decay=0.0005, warmup epochs=3.0, warmup momentum=0.8, wa rmup bias lr=0.1, box=7.5, cls=0.5, dfl=1.5, pose=12.0, kobj=1.0, label smoothing=0.0, nbs=64, hsv h=0.015, hsv s=0.7, hsv v=0.4, degrees=0.0, translate=0.1, scale=0.5, shear=0. 0, perspective=0.0, flipud=0.0, fliplr=0.5, bgr=0.0, mosaic=1.0, mixup=0.0, copy paste=0. 0, auto augment=randaugment, erasing=0.4, crop fraction=1.0, cfg=None, tracker=botsort.ya ml, save dir=runs/detect/train

• =	•				
nts	from	n	params	module	argume
0 6, 3, 2]	-1	1	464	ultralytics.nn.modules.conv.Conv	[3, 1
1	-1	1	4672	ultralytics.nn.modules.conv.Conv	[16, 3
2, 3, 2] 2	-1	1	7360	ultralytics.nn.modules.block.C2f	[32, 3
2, 1, True] 3	-1	1	18560	ultralytics.nn.modules.conv.Conv	[32, 6
4, 3, 2] 4	-1	2	49664	ultralytics.nn.modules.block.C2f	[64, 6
4, 2, True] 5	-1	1	73984	ultralytics.nn.modules.conv.Conv	[64, 1
28, 3, 2] 6	-1	2	197632	ultralytics.nn.modules.block.C2f	[128,
128, 2, True] 7	-1	1	295424	ultralytics.nn.modules.conv.Conv	[128,
256, 3, 2] 8	-1	1	460288	ultralytics.nn.modules.block.C2f	[256,
256, 1, True] 9	-1	1	164608	ultralytics.nn.modules.block.SPPF	[256,
256, 5] 10	-1	1	Θ	torch.nn.modules.upsampling.Upsample	[None,
2, 'nearest'] 11 12	[-1, 6] -1	1 1	0 148224	ultralytics.nn.modules.conv.Concat ultralytics.nn.modules.block.C2f	[1] [384,
128, 1] 13	-1	1	0	torch.nn.modules.upsampling.Upsample	
2, 'nearest']					[None,
14 15	[-1, 4] -1	1 1	0 37248	ultralytics.nn.modules.conv.Concat ultralytics.nn.modules.block.C2f	[1] [192,
64, 1] 16	-1	1	36992	ultralytics.nn.modules.conv.Conv	[64, 6
4, 3, 2] 17	[-1, 12]	1	0	ultralytics.nn.modules.conv.Concat	[1]
18 128, 1]	-1	1	123648	ultralytics.nn.modules.block.C2f	[192,
19 128, 3, 2]	- 1	1	147712	ultralytics.nn.modules.conv.Conv	[128,

```
20
                                   0 ultralytics.nn.modules.conv.Concat
                [-1, 9] 1
                                                                                    [1]
 21
                     -1 1
                              493056 ultralytics.nn.modules.block.C2f
                                                                                    [384,
256, 1]
22
           [15, 18, 21] 1
                              751507 ultralytics.nn.modules.head.Detect
                                                                                   [1, [6
4, 128, 256]]
Model summary: 225 layers, 3011043 parameters, 3011027 gradients, 8.2 GFLOPs
Transferred 355/355 items from pretrained weights
TensorBoard: Start with 'tensorboard --logdir runs/detect/train', view at http://localhos
t:6006/
Freezing layer 'model.22.dfl.conv.weight'
AMP: running Automatic Mixed Precision (AMP) checks with YOLOv8n...
AMP: checks passed <a></a>
train: Scanning /media/mahmoud/31A5D0D7034B3BD5/General/FCAI/8th Term/Assignment 1/datase
t/labels/train 1.cache... 9000 images, 6932 backgrounds, 0 corrupt: 100%
0/9000 [00:00<?, ?it/s]
val: Scanning /media/mahmoud/31A5D0D7034B3BD5/General/FCAI/8th Term/Assignment 1/dataset/
labels/validation.cache... 2684 images, 2079 backgrounds, 0 corrupt: 100%
4/2684 [00:00<?, ?it/s]
Plotting labels to runs/detect/train/labels.jpg...
optimizer: 'optimizer=auto' found, ignoring 'lr0=0.01' and 'momentum=0.937' and determini
ng best 'optimizer', 'lr0' and 'momentum' automatically...
optimizer: AdamW(lr=0.002, momentum=0.9) with parameter groups 57 weight(decay=0.0), 64 w
eight(decay=0.0005), 63 bias(decay=0.0)
TensorBoard: model graph visualization added 🔽
Image sizes 640 train, 640 val
Using 8 dataloader workers
Logging results to runs/detect/train
Starting training for 1 epochs...
```

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size	
1/1	2.31G	2.032	2.694	1.866	1	640:	100%
563/563	[08:51<00	0:00, 1.06i	t/s]				
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95):
100%	84/84	[00:59<00:00	, 1.42it/s]			
	all	2684	981	0.153	0.243	0.108	0.0331

1 epochs completed in 0.167 hours.

Optimizer stripped from runs/detect/train/weights/last.pt, 6.2MB Optimizer stripped from runs/detect/train/weights/best.pt, 6.2MB

Validating runs/detect/train/weights/best.pt...

Model summary (fused): 168 layers, 3005843 parameters, 0 gradients, 8.1 GFLOPs

```
Class Images Instances Box(P R mAP50 mAP50-95):
100%| 84/84 [00:51<00:00, 1.62it/s]
all 2684 981 0.157 0.224 0.109 0.0337
```

Speed: 0.4ms preprocess, 17.1ms inference, 0.0ms loss, 0.9ms postprocess per image Results saved to **runs/detect/train**

Predicting Bounding Boxes

```
In []: model = Y0L0("best3.pt")
    results = model('dataset/images/test',conf=0.1,stream=True)

def get_id_from_path(path):
    return os.path.splitext(os.path.basename(path))[0]
```

```
for result in results:
    id = get_id_from_path(result.path)
    with open(f'dataset/predictions/{id}.txt', 'w') as f:
        if len(result.xywh) == 0:
            f.write(f'{id}\n')
        else:
            with open(f'dataset/predictions/{id}.txt', 'w') as f:

            x, y, w, h = denormalize_bbox(*result.xywh.tolist())
            confidence = result.conf
            f.write(f'{id} {confidence} {x} {y} {w} {h} \n')
```