Algorithm Description

General Description

- Intended Use
 - To be used by Emergency Room physician to diagnose patient with pneumonia before x-rays are formally reported by a radiologist.
- Indications for Use
 - o To be used on adults. Chest Xray from AP or PA view.
- Device Limitations
 - o Not to be used on children.
- Clinical Impact of Performance
 - Speedier diagnosis of pneumonia. Reduced time pressures on radiologist to report CXRs.

Algorithm Design and Function

- Algorithm Flowchart
 - EDA -> Data Cleaning -> Creating test and training data -> Model-building ->
 Training -> Performance statistics -> Predicting pneumonia using created models and weights.
- DICOM Checking Steps
 - Line 1 11 File -> Inference.ipynb
- Pre-processing Steps

```
# This function takes the numpy array output by check_dicom and
# runs the appropriate pre-processing needed for our model input
def preprocess_image(img,img_mean,img_std,img_size):
    image = (img-img_mean)/img_std
    proc_img = resize(image, img_size)

return proc_img
```

CNN Architecture

```
#Loads the VGG 16 model and freezes all but the last CNN layer, returns the new model
def load_pretrained_model():

    model = VGG16(include_top=True, weights='imagenet')
    transfer_layer = model.get_layer('block5_pool')
    vgg_model = Model(inputs = model.input, outputs = transfer_layer.output)
    for layer in vgg_model.layers[0:17]:
        layer.trainable = False

return vgg_model
```

```
def build_my_model(vgg_model):
    my_model = Sequential()
    my_model.add(vgg_model)
    my_model.add(Flatten())
    my_model.add(Dropout(0.5))
    my_model.add(Dense(1024, activation = 'relu'))
    my_model.add(Dropout(0.5))
    my_model.add(Dense(512, activation = 'relu'))
    my_model.add(Dropout(0.5))
    my_model.add(Dense(256, activation = 'relu'))
    my_model.add(Dense(1, activation = 'sigmoid'))
    optimizer = Adam(lr = .0001)
    loss = 'binary_crossentropy'
    metrics = ['binary_accuracy']
    my_model.compile(optimizer = optimizer, loss = loss, metrics = metrics)
    return my_model
```

Algorithm Training

- Parameters:
 - Types of augmentation used during training

```
def my_image_augmentation():
    my_idg = ImageDataGenerator(rescale = 1. / 255.0,
                                horizontal_flip = True,
                                vertical_flip = False,
                                height_shift_range = 0.1,
                                width_shift_range = 0.1,
                                rotation range = 20,
                                shear_range = 0.1,
                                zoom_range = 0.1
    return my_idg
```

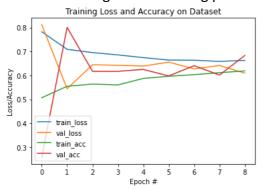
- Batch size
- Optimizer learning rate

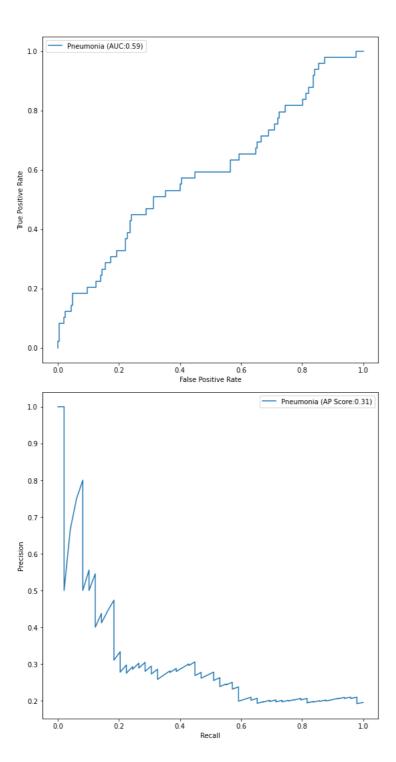
```
def make_train_gen(vargs):
    ## Create the actual generators using the output of my image augmentation for
    ## your training data
    ## This generator uses a batch size of 32
    idg = my_image_augmentation()
    train_gen = idg.flow_from_dataframe(dataframe=vargs,
                                          directory=None,
                                          x_col = 'path',
y_col = 'Pneumonia Class',
                                          class_mode = 'binary'
                                          target_size = (224, 224),
                                          batch_size = 32)
```

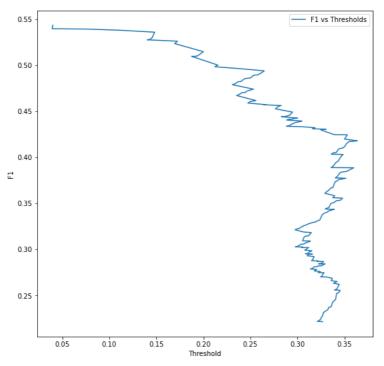
- Layers of pre-existing architecture that were frozen
- Layers of pre-existing architecture that were fine-tuned
- Layers added to pre-existing architecture

```
#Loads the VGG 16 model and freezes all but the last CNN layer, returns the new model
def load_pretrained_model():
    model = VGG16(include_top=True, weights='imagenet')
    transfer_layer = model.get_layer('block5_pool')
    vgg_model = Model(inputs = model.input, outputs = transfer_layer.output)
    for layer in vgg_model.layers[0:17]:
        layer.trainable = False
    return vgg_model
#This model has 4 Dense layers and uses a dropout of 0.5 and is added on top of the
#VGG 16 model with all but its last CNN layer frozen
def build_my_model(vgg_model):
    my_model = Sequential()
    my_model.add(vgg_model)
    my_model.add(Flatten())
    my_model.add(Dropout(0.5))
    my_model.add(Dense(1024, activation = 'relu'))
my_model.add(Dropout(0.5))
    my_model.add(Dense(512, activation = 'relu'))
    my_model.add(Dropout(0.5))
    my_model.add(Dense(256, activation = 'relu'))
    my_model.add(Dense(1, activation = 'sigmoid'))
    optimizer = Adam(lr = .0001)
    loss = 'binary_crossentropy'
    metrics = ['binary_accuracy']
    my_model.compile(optimizer = optimizer, loss = loss, metrics = metrics)
    return my_model
```

• Insert algorithm training performance visualization





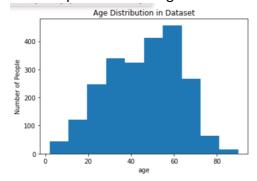


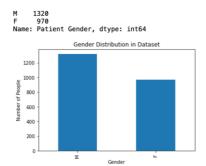
• Final Threshold and Explanation

```
# Threshold of 0.35 will result in F1 score of 0.32
threshold_select = 0.35
index = (np.abs(threshold_2 - threshold_select)).argmin()
precision = prec[index]
rec = recall[index]
```

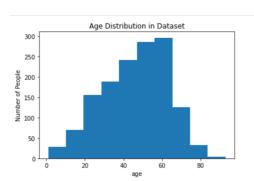
<u>Databases</u> (For the below, include visualizations as they are useful and relevant)

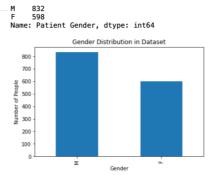
• Description of Training Dataset





• Description of Validation Dataset





Ground Truth

• Labels from the NIH data sample – using label defined by radiologist.

FDA Validation Plan

- Patient Population Description for FDA Validation Dataset
 - o Adults 18 and over who presents with symptoms of Pneumonia
- Ground Truth Acquisition Methodology
 - o Labels captured using NLP from experienced radiologist.