warnings.filterwarnings("ignore") 🔋 import ImageDataGenerator .keras.utils import plot\_model from PIL import Image from keras.layers import GlobalAveragePooling2D, Dense, Conv2D, BatchNormalization, Dropout from **keras** import backend as K from keras.models import Model,load\_model ork.ops import disable\_eager\_execution from keras.regularizers import 12 %load\_ext tensorboard The tensorboard extension is already loaded. To reload it, use: %reload\_ext tensorboard from google.colab import drive drive.mount('/content/drive') Mounted at /content/drive vith open('/content/drive//My Drive/Steel\_Detection /data.pkl','rb') as f: train=pickle load(f) train head() rle\_1 rle\_2 rle\_3 rle\_4 defect stratify defect\_1 defect\_2 defect\_3 defect\_4 total\_defects image\_id 29102 12 29346 24 **0** 0002cc93b.jpg 29602 24 29858 24 30114 24 3... **1** 00031f466.jpg **2** 000418bfc.jpg **3** 000789191.jpg 18661 28 18863 82 4 0007a71bf.jpg 19091 110 19347 110 19603 11... ]: train shape (y\_true, y\_pred): true\_positives=K.sum(K.round(K.clip(y\_true\*y\_pred,0,1))) #calculates number of true possible\_positives=K.sum(K.round(K.clip(y\_true,0,1))) predicted\_positives=K.sum(K.round(K.clip(y\_pred,0,1))) precision=true\_positives/(predicted\_positives +K.epsilon()) recall=true\_positives/(possible\_positives+K.epsilon()) f1\_val=2\*(precision\*recall)/(precision+recall+K.epsilon()) return f1\_val x\_train, x\_test=train\_test\_split(train, test\_size=0.10, stratify=train['stratify'], random\_st ate=0) x\_train, x\_val=train\_test\_split(x\_train, test\_size=0.20, stratify=x\_train['stratify'], random \_state=0) x\_train=x\_train[['image\_id','defect']] x\_val=x\_val[['image\_id','defect']] x\_test=x\_test[['image\_id','defect']] print("x\_train {}".format(x\_train.shape)," x\_val {}".format(x\_val.shape)," x\_test {}".fo rmat(x\_test shape)) x\_train (9048, 2) x\_val (2263, 2) x\_test (1257, 2) Data augmentation is used here to increase the amount of data by adding slightly modified copies of already existing data also during EDA we have seen that data is highly imbalanced. So it is advisable to do at least Data augmentation. train\_datagen=ImageDataGenerator(rescale=1./255, shear\_range=0.1, zoom\_range=0.1, horizontal\_flip=True, vertical\_flip=True, rotation\_range=60) val\_datagen=ImageDataGenerator(rescale=1./255) train\_folder\_path= train\_image\_generator=train\_datagen\_flow\_from\_dataframe(dataframe=x\_train\_astype(str), directory=train\_folder\_path, x\_col="image\_id", y\_col="defect", seed=42, shuffle=True, batch\_size=32, class\_mode="binary", target\_size=(256,512)) val\_image\_generator=val\_datagen.flow\_from\_dataframe(dataframe=x\_val.astype(str), directory=train\_folder\_path, x\_col="image\_id", y\_col="defect", batch\_size=32, seed=42, shuffle=True, class\_mode="binary", target\_size=(256,512)) Found 9048 validated image filenames belonging to 2 classes. Found 2263 validated image filenames belonging to 2 classes. Flattening is No brainer and it simply converts a multi-dimensional object to one-dimensional by rearranging the elements. • While GlobalAveragePooling is a methodology used for better representation of your vector. It can be 1D/2D/3D.It uses a parser window which moves across the object and pools the data by averaging it (GlobalAveragePooling) or picking max value (GlobalMaxPooling). • Batch normalization, it is a process to make neural networks faster and more stable through adding extra layers in a deep neural network. The new layer performs the standardizing and normalizing operations on the input of a layer coming from a previous layer. A typical neural network is trained using a collected set of input data called batch. Similarly, the normalizing process in batch normalization takes place in batches, not as a single input. • Large neural nets trained on relatively small datasets can overfit the training data. By dropping a unit out, we mean temporarily removing it from the network, along with all its incoming and outgoing connections we can overcome overfitting problem. • Used pre-trained Xception() model by keras without fully-connected layer at the top of the network and later freeze the pre-trained model. base\_model=tf.keras.applications.xception.Xception(input\_shape=(256,512,3),include\_top=50 base\_model.trainable=False m=base\_model output m=GlobalAveragePooling2D()(m) m=Dense(1024, activation='relu')(m) m=BatchNormalization()(m) #https://www.analyticsvidhya.com/blog/2021/03/introduction-to m=Dropout(0.3)(m) m=Dense(512, activation='relu')(m) m=BatchNormalization()(m) m= Dropout(0.3)(m) m=Dense(64, activation='relu')(m) m=BatchNormalization()(m) m=Dropout(0.3)(m) output=Dense(1,activation='sigmoid')(m) #Binary Classification thus sigmoid is used model=Model(inputs=base\_model.input,outputs=output) model.\_name="Binary\_Classification\_Model model.summary() Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/xception/xception\_weights\_tf\_dim\_ordering\_t f\_kernels\_notop.h5 83689472/83683744 [============] - 1s Ous/step 83697664/83683744 [============ ] - 1s Ous/step Model: "Binary\_Classification\_Model" Output Shape Layer (type) input\_1 (InputLayer) input\_1[0][0] block1\_conv1[0][0] block1\_conv1\_act (Activation) (None, 127, 255, 32) 0 block1\_conv1\_bn[0][0] block1\_conv1\_act[0][0] block1\_conv2\_bn (BatchNormaliza (None, 125, 253, 64) 256 block1\_conv2[0][0] block1\_conv2\_bn[0][0] block2\_sepconv1 (SeparableConv2 (None, 125, 253, 128 8768 block1\_conv2\_act[0][0] block2\_sepconv1\_bn (BatchNormal (None, 125, 253, 128 512 block2\_sepconv1[0][0] block2\_sepconv1\_bn[0][0] block2\_sepconv2 (SeparableConv2 (None, 125, 253, 128 17536 block2\_sepconv2\_act[0][0] block2\_sepconv2\_bn (BatchNormal (None, 125, 253, 128 512 block2\_sepconv2[0][0] block1\_conv2\_act[0][0] block2\_pool (MaxPooling2D) block2\_sepconv2\_bn[0][0] conv2d[0][0] add (Add) block2\_pool[0][0] batch\_normalization[0][0] add[0][0] block3\_sepconv1 (SeparableConv2 (None, 63, 127, 256) 33920 block3\_sepconv1\_act[0][0] block3\_sepconv1\_bn (BatchNormal (None, 63, 127, 256) 1024 block3\_sepconv1[0][0] block3\_sepconv2\_act (Activation (None, 63, 127, 256) 0 block3\_sepconv1\_bn[0][0] block3\_sepconv2 (SeparableConv2 (None, 63, 127, 256) 67840 block3\_sepconv2\_act[0][0] block3\_sepconv2\_bn (BatchNormal (None, 63, 127, 256) 1024 block3\_sepconv2[0][0] (None, 32, 64, 256) 32768 add[0][0] block3\_pool (MaxPooling2D) (None, 32, 64, 256) 0 block3\_sepconv2\_bn[0][0] batch\_normalization\_1 (BatchNor (None, 32, 64, 256) 1024 conv2d\_1[0][0] add\_1 (Add) (None, 32, 64, 256) 0 block3\_pool[0][0] batch\_normalization\_1[0][0] block4\_sepconv1\_act (Activation (None, 32, 64, 256) 0 add\_1[0][0] block4\_sepconv1 (SeparableConv2 (None, 32, 64, 728) 188672 block4\_sepconv1\_act[0][0] block4\_sepconv1[0][0] block4\_sepconv1\_bn[0][0] block4\_sepconv2 (SeparableConv2 (None, 32, 64, 728) 536536 block4\_sepconv2\_act[0][0] block4\_sepconv2[0][0] add\_1[0][0] (None, 16, 32, 728) 186368 block4\_pool (MaxPooling2D) block4\_sepconv2\_bn[0][0] batch\_normalization\_2 (BatchNor (None, 16, 32, 728) 2912 conv2d\_2[0][0] add\_2 (Add) block4\_pool[0][0] batch\_normalization\_2[0][0] block5\_sepconv1\_act (Activation (None, 16, 32, 728) 0 add\_2[0][0] block5\_sepconv1\_act[0][0] block5\_sepconv1[0][0] block5\_sepconv1\_bn[0][0] block5\_sepconv2\_act[0][0] block5\_sepconv2[0][0] block5\_sepconv2\_bn[0][0] block5\_sepconv3 (SeparableConv2 (None, 16, 32, 728) 536536 block5\_sepconv3\_act[0][0] block5\_sepconv3\_bn (BatchNormal (None, 16, 32, 728) 2912 block5\_sepconv3[0][0] block5\_sepconv3\_bn[0][0] add\_3 (Add) add\_2[0][0] add\_3[0][0] block6\_sepconv1\_act (Activation (None, 16, 32, 728) 0 block6\_sepconv1 (SeparableConv2 (None, 16, 32, 728) 536536 block6\_sepconv1\_act[0][0] block6\_sepconv1[0][0] block6\_sepconv2 (SeparableConv2 (None, 16, 32, 728) 536536 block6\_sepconv2\_act[0][0] block6\_sepconv2[0][0] block6\_sepconv2\_bn[0][0] block6\_sepconv3\_act[0][0] block6\_sepconv3\_bn (BatchNormal (None, 16, 32, 728) 2912 block6\_sepconv3[0][0] add\_4 (Add) block6\_sepconv3\_bn[0][0] add\_3[0][0] add\_4[0][0] block7\_sepconv1\_act[0][0] block7\_sepconv1[0][0] block7\_sepconv1\_bn[0][0] block7\_sepconv2\_act[0][0] block7\_sepconv2[0][0] block7\_sepconv2\_bn[0][0] block7\_sepconv3\_act[0][0] block7\_sepconv3\_bn (BatchNormal (None, 16, 32, 728) 2912 block7\_sepconv3[0][0] add\_5 (Add) block7\_sepconv3\_bn[0][0] add\_4[0][0] add\_5[0][0] block8\_sepconv1\_act[0][0] block8\_sepconv1[0][0] block8\_sepconv1\_bn[0][0] block8\_sepconv2\_act[0][0] block8\_sepconv2[0][0] block8\_sepconv2\_bn[0][0] block8\_sepconv3 (SeparableConv2 (None, 16, 32, 728) 536536 block8\_sepconv3\_act[0][0] block8\_sepconv3\_bn (BatchNormal (None, 16, 32, 728) 2912 block8\_sepconv3[0][0] block8\_sepconv3\_bn[0][0] add\_6 (Add) add\_5[0][0] add\_6[0][0] block9\_sepconv1\_act[0][0] block9\_sepconv1[0][0] block9\_sepconv1\_bn[0][0] block9\_sepconv2\_act[0][0] block9\_sepconv2\_bn (BatchNormal (None, 16, 32, 728) 2912 block9\_sepconv2[0][0] block9\_sepconv3\_act (Activation (None, 16, 32, 728) 0 block9\_sepconv2\_bn[0][0] block9\_sepconv3\_act[0][0] block9\_sepconv3\_bn (BatchNormal (None, 16, 32, 728) 2912 block9\_sepconv3[0][0] block9\_sepconv3\_bn[0][0] add\_7 (Add) add\_6[0][0] block10\_sepconv1\_act (Activatio (None, 16, 32, 728) 0 add\_7[0][0] block10\_sepconv1 (SeparableConv (None, 16, 32, 728) 536536 block10\_sepconv1\_act[0][0] block10\_sepconv1\_bn (BatchNorma (None, 16, 32, 728) 2912 block10\_sepconv1[0][0] block10\_sepconv2\_act (Activatio (None, 16, 32, 728) 0 block10\_sepconv1\_bn[0][0] block10\_sepconv2 (SeparableConv (None, 16, 32, 728) 536536 block10\_sepconv2\_act[0][0] block10\_sepconv2\_bn (BatchNorma (None, 16, 32, 728) 2912 block10\_sepconv2[0][0] block10\_sepconv3\_act (Activatio (None, 16, 32, 728) 0 block10\_sepconv2\_bn[0][0] block10\_sepconv3 (SeparableConv (None, 16, 32, 728) 536536 block10\_sepconv3\_act[0][0] block10\_sepconv3\_bn (BatchNorma (None, 16, 32, 728) 2912 block10\_sepconv3[0][0] add\_8 (Add) block10\_sepconv3\_bn[0][0] add\_7[0][0] add\_8[0][0] block11\_sepconv1 (SeparableConv (None, 16, 32, 728) 536536 block11\_sepconv1\_act[0][0] block11\_sepconv1\_bn (BatchNorma (None, 16, 32, 728) 2912 block11\_sepconv1[0][0] block11\_sepconv1\_bn[0][0] block11\_sepconv2 (SeparableConv (None, 16, 32, 728) 536536 block11\_sepconv2\_act[0][0] block11\_sepconv2\_bn (BatchNorma (None, 16, 32, 728) 2912 block11\_sepconv2[0][0] block11\_sepconv3\_act (Activatio (None, 16, 32, 728) 0 block11\_sepconv2\_bn[0][0] block11\_sepconv3 (SeparableConv (None, 16, 32, 728) 536536 block11\_sepconv3\_act[0][0] block11\_sepconv3\_bn (BatchNorma (None, 16, 32, 728) 2912 block11\_sepconv3[0][0] add\_9 (Add) block11\_sepconv3\_bn[0][0] add\_8[0][0] add\_9[0][0] block12\_sepconv1 (SeparableConv (None, 16, 32, 728) 536536 block12\_sepconv1\_act[0][0] block12\_sepconv1[0][0] block12\_sepconv1\_bn[0][0] block12\_sepconv2 (SeparableConv (None, 16, 32, 728) 536536 block12\_sepconv2\_act[0][0] block12\_sepconv2\_bn (BatchNorma (None, 16, 32, 728) 2912 block12\_sepconv2[0][0] block12\_sepconv3\_act (Activatio (None, 16, 32, 728) 0 block12\_sepconv2\_bn[0][0] block12\_sepconv3\_act[0][0] block12\_sepconv3 (SeparableConv (None, 16, 32, 728) 536536 block12\_sepconv3[0][0] block12\_sepconv3\_bn (BatchNorma (None, 16, 32, 728) 2912 add\_10 (Add) block12\_sepconv3\_bn[0][0] add\_9[0][0] block13\_sepconv1 (SeparableConv (None, 16, 32, 728) 536536 block13\_sepconv1\_act[0][0] block13\_sepconv1[0][0] block13\_sepconv1\_bn[0][0] block13\_sepconv2 (SeparableConv (None, 16, 32, 1024) 752024 block13\_sepconv2\_act[0][0] block13\_sepconv2\_bn (BatchNorma (None, 16, 32, 1024) 4096 block13\_sepconv2[0][0] add\_10[0][0] block13\_pool (MaxPooling2D) (None, 8, 16, 1024) 0 block13\_sepconv2\_bn[0][0] conv2d\_3[0][0] block13\_pool[0][0] batch\_normalization\_3[0][0] block14\_sepconv1 (SeparableConv (None, 8, 16, 1536) 1582080 add\_11[0][0] block14\_sepconv1[0][0] block14\_sepconv1\_bn[0][0] block14\_sepconv2 (SeparableConv (None, 8, 16, 2048) 3159552 block14\_sepconv1\_act[0][0] block14\_sepconv2\_bn (BatchNorma (None, 8, 16, 2048) 8192 block14\_sepconv2[0][0] block14\_sepconv2\_act (Activatio (None, 8, 16, 2048) 0 block14\_sepconv2\_bn[0][0] global\_average\_pooling2d (Globa (None, 2048) block14\_sepconv2\_act[0][0] global\_average\_pooling2d[0][0] (None, 1024) 2098176 4096 dense[0][0] (None, 1024) batch\_normalization\_4[0][0] dropout (Dropout) 524800 dropout[0][0] batch\_normalization\_5 (BatchNor (None, 512) 2048 dense\_1[0][0] dropout\_1 (Dropout) batch\_normalization\_5[0][0] dropout\_1[0][0] batch\_normalization\_6 (BatchNor (None, 64) dense\_2[0][0] (None, 64) batch\_normalization\_6[0][0] dense\_3 (Dense) dropout\_2[0][0] Total params: 23,523,753 Trainable params: 2,659,073 Non-trainable params: 20,864,680 log\_dir=os.path.join("logs", datetime.datetime.now().strftime("%Y%m%d-%H%M%S")) tensorboard=tf.keras.callbacks.TensorBoard(log\_dir=log\_dir,histogram\_freq=1,write\_graph=1 rue, write\_grads=True) checkpoint\_filepath='/content/drive//My\_Drive/Steel\_Detection\_/binary\_Xception\_2.h5' model\_checkpoint\_callback=tf.keras.callbacks.ModelCheckpoint(filepath=checkpoint\_filepath , monitor='val\_f1\_score', mode='max', save\_best\_only=True) model.compile(optimizer='Adam',loss='binary\_crossentropy',metrics=["acc",f1\_score]) callback=[model\_checkpoint\_callback, tensorboard] history=model fit\_generator(train\_image\_generator, validation\_data=val\_image\_generator, epo chs=20, verbose=1, callbacks=callback) WARNING:tensorflow:`write\_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback. val\_acc: 0.8401 - val\_f1\_score: 0.8538 al\_acc: 0.8423 - val\_f1\_score: 0.8423 Epoch 3/20 al\_acc: 0.8595 - val\_f1\_score: 0.8629 al\_acc: 0.8472 - val\_f1\_score: 0.8656 al\_acc: 0.8785 - val\_f1\_score: 0.8814 al\_acc: 0.8772 - val\_f1\_score: 0.8879 al\_acc: 0.8706 - val\_f1\_score: 0.8809 Epoch 8/20 al\_acc: 0.8741 - val\_f1\_score: 0.8712 Epoch 9/20 al\_acc: 0.8750 - val\_f1\_score: 0.8739 al\_acc: 0.8962 - val\_f1\_score: 0.9015 Epoch 11/20 al\_acc: 0.8847 - val\_f1\_score: 0.8922 al\_acc: 0.8847 - val\_f1\_score: 0.8884 Epoch 13/20 al\_acc: 0.8834 - val\_f1\_score: 0.8874 al\_acc: 0.8679 - val\_f1\_score: 0.8694 Epoch 15/20 al\_acc: 0.8940 - val\_f1\_score: 0.8990 Epoch 16/20 al\_acc: 0.8891 - val\_f1\_score: 0.8933 al\_acc: 0.8838 - val\_f1\_score: 0.8862 al\_acc: 0.8838 - val\_f1\_score: 0.8949 al\_acc: 0.8869 - val\_f1\_score: 0.8880 Epoch 20/20 %tensorboard --logdir logs model=load\_model('/content/drive//My Drive/Steel\_Detection /binary\_Xception\_2.h5',custom\_ objects={'f1\_score':f1\_score}) train\_image\_generator=val\_datagen\_flow\_from\_dataframe(dataframe=x\_train\_astype(str), directory=train\_folder\_path, x\_col="image\_id", y\_col="defect", batch\_size=32, shuffle=False, class\_mode="binary", target\_size=(256,512)) val\_image\_generator=val\_datagen\_flow\_from\_dataframe(dataframe=x\_val\_astype(str), directory=train\_folder\_path, x\_col="image\_id", y\_col="defect", batch\_size=32, shuffle=False, class\_mode="binary", target\_size=(256,512)) test\_image\_generator=val\_datagen.flow\_from\_dataframe(dataframe=x\_test.astype(str), directory=train\_folder\_path, x\_col="image\_id", y\_col="defect", batch\_size=32, shuffle=False, class\_mode="binary", target\_size=(256,512)) Found 9048 validated image filenames belonging to 2 classes. Found 2263 validated image filenames belonging to 2 classes. Found 1257 validated image filenames belonging to 2 classes. print(model.evaluate(train\_image\_generator, verbose=1)) print("="\*100) print(model.evaluate(val\_image\_generator, verbose=1)) print("="\*100) print(model.evaluate(test\_image\_generator,verbose=1)) Training Dataset: [0.22544366121292114, 0.9016121625900269, 0.9064911007881165] Validation Dataset: [0.23688536882400513, 0.9037102460861206, 0.9095268845558167] [0.24440200626850128, 0.8958664536476135, 0.9009650945663452] • One can observe the value of loss and metrics are similar for train, validation and test datasets thus the model is not over-fitting. f1 score of all datasets is above 0.90. Overall model is having good performance on train, validation and test datasets.

from matplotlib import pyplot as plt

from sklearn.model\_selection import train\_test\_split