```
# For example, here's several helpful packages to load in
        import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
        # Input data files are available in the "../input/" directory.
        # For example, running this (by clicking run or pressing Shift+Enter) will list the files in the input
         directory
        import os
        from glob import glob
        from random import shuffle
        import cv2
        import matplotlib.pyplot as plt
        from sklearn.model_selection import train test split
        from keras.preprocessing.image import ImageDataGenerator
        from keras.layers import Convolution1D, concatenate, SpatialDropout1D, GlobalMaxPool1D, GlobalAvgPool1D
        , Embedding, \
            Conv2D, SeparableConv1D, Add, BatchNormalization, Activation, GlobalAveragePooling2D, LeakyReLU, Fl
        from keras.layers import Dense, Input, Dropout, MaxPooling2D, Concatenate, GlobalMaxPooling2D, GlobalAv
        eragePooling2D, \
            Lambda, Multiply, LSTM, Bidirectional, PReLU, MaxPooling1D
        from keras.layers.pooling import GlobalPooling1D
        from keras.losses import mae, sparse categorical crossentropy, binary crossentropy
        from keras.models import Model
        from keras.applications.nasnet import NASNetMobile, NASNetLarge, preprocess_input
        from keras.optimizers import Adam, RMSprop
        from keras.callbacks import ModelCheckpoint, EarlyStopping, ReduceLROnPlateau
        from imgaug import augmenters as iaa
        import imgaug as ia
        print(os.listdir("../input"))
        # Any results you write to the current directory are saved as output.
In [ ]: | df train = pd.read csv("../input/train labels.csv")
        id label map = {k:v for k,v in zip(df train.id.values, df train.label.values)}
        df train.head()
In [ ]: def get id from file path(file path):
            return file path.split(os.path.sep)[-1].replace('.tif', '')
In [ ]: | labeled_files = glob('../input/train/*.tif')
        test_files = glob('../input/test/*.tif')
In [ ]: print("labeled_files size :", len(labeled_files))
        print("test_files size :", len(test_files))
In [ ]: | train, val = train test split(labeled files, test size=0.1, random state=101010)
In [ ]: def chunker(seq, size):
            return (seq[pos:pos + size] for pos in range(0, len(seq), size))
            sometimes = lambda aug: iaa.Sometimes(0.5, aug)
            seq = iaa.Sequential(
                     # apply the following augmenters to most images
                    iaa.Fliplr(0.5), # horizontally flip 50% of all images
                    iaa.Flipud(0.2), # vertically flip 20% of all images
                    sometimes (iaa. Affine (
                        scale={"x": (0.9, 1.1), "y": (0.9, 1.1)}, # scale images to 80-120% of their size, indi
        vidually per axis
                        translate percent={"x": (-0.1, 0.1), "y": (-0.1, 0.1)}, # translate by -20 to +20 perce
        nt (per axis)
                        rotate=(-10, 10), # rotate by -45 to +45 degrees
                        shear=(-5, 5), # shear by -16 to +16 degrees
                        order=[0, 1], # use nearest neighbour or bilinear interpolation (fast)
                        cval=(0, 255), # if mode is constant, use a cval between 0 and 255
                        mode=ia.ALL # use any of scikit-image's warping modes (see 2nd image from the top for e
        xamples)
                    )),
                     # execute 0 to 5 of the following (less important) augmenters per image
                     # don't execute all of them, as that would often be way too strong
                    iaa.SomeOf((0, 5),
                             sometimes(iaa.Superpixels(p replace=(0, 1.0), n segments=(20, 200))), # convert ima
        ges into their superpixel representation
                            iaa.OneOf([
                                 iaa.GaussianBlur((0, 1.0)), # blur images with a sigma between 0 and 3.0
                                 iaa. AverageBlur (k=(3, 5)), # blur image using local means with kernel sizes bet
        ween 2 and 7
                                 iaa.MedianBlur(k=(3, 5)), # blur image using local medians with kernel sizes be
        tween 2 and 7
                             ]),
                             iaa.Sharpen(alpha=(0, 1.0), lightness=(0.9, 1.1)), # sharpen images
                            iaa.Emboss(alpha=(0, 1.0), strength=(0, 2.0)), # emboss images
                             # search either for all edges or for directed edges,
                             # blend the result with the original image using a blobby mask
                            iaa.SimplexNoiseAlpha(iaa.OneOf([
                                 iaa.EdgeDetect(alpha=(0.5, 1.0)),
                                 iaa.DirectedEdgeDetect(alpha=(0.5, 1.0), direction=(0.0, 1.0)),
                             iaa.AdditiveGaussianNoise(loc=0, scale=(0.0, 0.01*255), per channel=0.5), # add gau
        ssian noise to images
                            iaa.OneOf([
                                 iaa.Dropout((0.01, 0.05), per channel=0.5), # randomly remove up to 10% of the
         pixels
                                iaa.CoarseDropout((0.01, 0.03), size percent=(0.01, 0.02), per channel=0.2),
                             ]),
                             iaa.Invert(0.01, per channel=True), # invert color channels
                             iaa.Add((-2, 2), per channel=0.5), # change brightness of images (by -10 to 10 of o
        riginal value)
                            iaa.AddToHueAndSaturation((-1, 1)), # change hue and saturation
                             # either change the brightness of the whole image (sometimes
                             # per channel) or change the brightness of subareas
                             iaa.OneOf([
                                 iaa.Multiply((0.9, 1.1), per channel=0.5),
                                 iaa.FrequencyNoiseAlpha(
                                     exponent=(-1, 0),
                                     first=iaa.Multiply((0.9, 1.1), per_channel=True),
                                     second=iaa.ContrastNormalization((0.9, 1.1))
                             ]),
                             sometimes(iaa.ElasticTransformation(alpha=(0.5, 3.5), sigma=0.25)), # move pixels 1
        ocally around (with random strengths)
                             sometimes(iaa.PiecewiseAffine(scale=(0.01, 0.05))), # sometimes move parts of the i
        mage around
                             sometimes(iaa.PerspectiveTransform(scale=(0.01, 0.1)))
                        ],
                        random order=True
                random order=True
            return seq
        def data gen(list files, id label map, batch size, augment=False):
            seq = get seq()
            while True:
                shuffle(list files)
                for batch in chunker(list files, batch size):
                    X = [cv2.imread(x)  for x  in batch]
                    Y = [id label map[get id from file path(x)] for x in batch]
                    if augment:
                        X = seq.augment images(X)
                    X = [preprocess input(x) for x in X]
                    yield np.array(X), np.array(Y)
In [ ]: def get model classif nasnet():
            inputs = Input((96, 96, 3))
            base model = NASNetMobile(include top=False, input shape=(96, 96, 3)) #, weights=None
            x = base model(inputs)
            out1 = GlobalMaxPooling2D()(x)
            out2 = GlobalAveragePooling2D()(x)
            out3 = Flatten()(x)
            out = Concatenate(axis=-1)([out1, out2, out3])
            out = Dropout(0.5)(out)
            out = Dense(1, activation="sigmoid", name="3")(out)
            model = Model(inputs, out)
            model.compile(optimizer=Adam(0.0001), loss=binary crossentropy, metrics=['acc'])
            model.summary()
            return model
In [ ]: | model = get model classif nasnet()
In [ ]: | batch size=32
        h5 path = "model.h5"
        checkpoint = ModelCheckpoint(h5 path, monitor='val acc', verbose=1, save best only=True, mode='max')
        history = model.fit generator(
            data gen(train, id label map, batch size, augment=True),
            validation_data=data_gen(val, id_label_map, batch_size),
            epochs=2, verbose=1,
            callbacks=[checkpoint],
            steps per epoch=len(train) // batch size,
            validation steps=len(val) // batch size)
        batch size=64
        history = model.fit generator(
            data gen(train, id label map, batch size, augment=True),
            validation data=data gen(val, id label map, batch size),
            epochs=6, verbose=1,
            callbacks=[checkpoint],
            steps per epoch=len(train) // batch size,
            validation steps=len(val) // batch size)
        model.load weights(h5 path)
In [ ]: | preds = []
        ids = []
In [ ]: for batch in chunker(test_files, batch_size):
            X = [preprocess_input(cv2.imread(x)) for x in batch]
            ids_batch = [get_id_from_file_path(x) for x in batch]
```

X = np.array(X)

df.head()

preds += preds_batch
ids += ids batch

In []: | df = pd.DataFrame({'id':ids, 'label':preds})

df.to_csv("baseline_nasnet.csv", index=False)

preds_batch = ((model.predict(X).ravel()*model.predict(X[:, ::-1, :, :]).ravel()*model.predict(X[:,

::-1, ::-1, :]).ravel()*model.predict(X[:, :, ::-1, :]).ravel())**0.25).tolist()

In []: | # This Python 3 environment comes with many helpful analytics libraries installed

It is defined by the kaggle/python docker image: https://github.com/kaggle/docker-python