

CNN Model 2

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In [ ]: ## Reproducibility
        ## Random seed given
        import random ## import the random library
        random.seed(10) ## Setting the seed to get the same answer no matter how many
        times and who runs the model
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In [ ]: ## Downloading specific libraries
        import numpy as np ## Library that enables linear functions
        import pandas as pd ## # Enables data processing
        import glob ## returns an array of filenames that match a pattern
        import cv2 ## helps add labels to image classifications
        import matplotlib.pyplot as plt ## library for producing figures
        from keras.preprocessing.image import ImageDataGenerator, load_img, img_to_arr
        ay, array_to_img ## importing image processing packages from
        ## keras
        from keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout ## Impo
        rt libraries for model building
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In [ ]: ## Read the train csv file
        train_dir='/kaggle/input/siim-isic-melanoma-classification/jpeg/train/' ## ass
        igning a name to the location of the train images
        train=pd.read_csv('/kaggle/input/siim-isic-melanoma-classification/train.csv')
        ## assigning a name to the location of the CSV file

        ## Read the test csv file
        test_dir='/kaggle/input/siim-isic-melanoma-classification/jpeg/test/' ## assig
        ning a name to the location of the test images
        test=pd.read_csv('/kaggle/input/siim-isic-melanoma-classification/test.csv') #
        # assigning a name to the location of the CSV file
```

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In [ ]: ## Finding the unique patient ids from train csv file
        print(f"The total patient ids are {train['patient_id'].count()}, from those th
        e unique ids are {train['patient_id'].value_counts().shape[0]} ")

        ## Finding the unique patient ids from test csv file
        print(f"The total patient ids are {test['patient_id'].count()}, from those the
        unique ids are {test['patient_id'].value_counts().shape[0]} ")
```

```
In [ ]: train['path'] = train_dir + train.image_name + ".jpg" ## adding the location o
        f the image to the row for the train data set
        train.head() ## showing the first 5 lines of the train data set, note the "pat
        h" coloumn

        test['path'] = test_dir + test.image_name + ".jpg" ## adding the location of
        the image to the row for the test data set
        test.head() ## showing the first 5 lines of the test data set, note the "pat
        h" coloumn
```

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In [ ]: ## Class Distribution
train.target.value_counts() ## Count the number of images that were classified
as malinnent or non malignant

In [ ]: df_0=train[train['target']==0].sample(1000) ## produce a data frame using 1000
images from the train data set where the target equals zero
df_1=train[train['target']==1] ## produce a data frame using all the images fr
om the test data set where the target equals 584
train=pd.concat([df_0,df_1]) ## create a new dataset using the smaller trainin
g data set
train=train.reset_index() ## making sure the new "train" data set is being use
d for the model

In [ ]: train.shape ## how many observations and variables are in the training set bei
ng used for the model

In [ ]: train.head() ## First 5 rows of the new train set

In [ ]: # we will resize the given images to 200 x 200 size images for faster processi
ng
IMG_DIM = (200, 200) ## changing the image dimensions

In [ ]: from sklearn.model_selection import train_test_split ## importing the train te
st split function
X_train, X_val, y_train, y_val = train_test_split(train, train.target, test_si
ze=0.2, random_state=42) ## taking 20% of the training data set

In [ ]: train_files = X_train.path ## Image path for the training data set
val_files = X_val.path ## Image path for the validation data set

train_imgs = [img_to_array(load_img(img, target_size=IMG_DIM)) for img in trai
n_files] ## load images using load_img function from keras
## preprocessing using the target_size function
validation_imgs = [img_to_array(load_img(img, target_size=IMG_DIM)) for img in
val_files] ## using the img_to_array will tranform the loaded image to an arra
y

train_imgs = np.array(train_imgs) ## converting the list of arrays to array fo
r the training dataset
train_labels = y_train

validation_imgs = np.array(validation_imgs) ## converting the list of arrays t
o array for the validation dataset
val_labels = y_val

print('Train dataset shape:', train_imgs.shape,
      '\tValidation dataset shape:', validation_imgs.shape)
```

```
In [ ]: ## Scale Images
## scale each image with pixel values between (0, 255) to values between (0,
1) because deep learning models work really
## well with small input values.
train_imgs_scaled = train_imgs.astype('float32')

validation_imgs_scaled = validation_imgs.astype('float32')

# divide the pixels by 255 to scale the pixels between 0 and 1
train_imgs_scaled /= 255
validation_imgs_scaled /= 255

print(train_imgs[0].shape)

array_to_img(train_imgs[0]) ## using the array_to_img function will convert th
e given array to image
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```
In [ ]: # setup basic configuration
batch_size = 30 ## indicating the total number of images passed to the model p
er iteration
num_classes = 2
epochs = 75 ## establishing the training time
input_shape = (200, 200, 3)

from keras.models import Sequential ## importing the sequential library
from keras import optimizers ## importing optimizers
```

```
In [ ]: train_datagen = ImageDataGenerator(rescale=1./255, zoom_range=0.3, rotation_ra
nge=70,
width_shift_range=0.3, height_shift_range=
0.3, shear_range=0.3,
horizontal_flip=True, fill_mode='nearest')

val_datagen = ImageDataGenerator(rescale=1./255)
```

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In [ ]: # Lets take a random image and see how transformed images actually looks
img_id = 1

img_generator = train_datagen.flow(train_imgs[img_id:img_id+1], train_labels[i
mg_id:img_id+1],
batch_size=1)

img = [next(img_generator) for i in range(0,5)]

fig, ax = plt.subplots(1,5, figsize=(16, 6))
print('Labels:', [item[1][0] for item in img])
l = [ax[i].imshow(img[i][0][0]) for i in range(0,5)]
```

```
In [ ]: import random ## import the random library
        random.seed(10) ## Setting the seed to get the same answer no matter how many
            times and who runs the model

        train_generator = train_datagen.flow(train_imgs, train_labels, batch_size=30)
        val_generator = val_datagen.flow(validation_imgs, val_labels, batch_size=20)

        input_shape = input_shape

        model = Sequential() ## creating and instance of Sequential

        model.add(Conv2D(16, kernel_size=(3, 3), activation='relu',
            input_shape=input_shape))
        model.add(MaxPooling2D(pool_size=(2, 2)))

        model.add(Conv2D(64, kernel_size=(3, 3), activation='relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))

        model.add(Conv2D(128, kernel_size=(3, 3), activation='relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))

        model.add(Conv2D(128, kernel_size=(3, 3), activation='relu'))
        model.add(MaxPooling2D(pool_size=(2, 2)))

        model.add(Flatten())
        model.add(Dense(512, activation='relu'))
        model.add(Dropout(0.3))
        model.add(Dense(512, activation='relu'))
        model.add(Dropout(0.3))
        model.add(Dense(1, activation='sigmoid'))

        model.compile(loss='binary_crossentropy',
            optimizer=optimizers.RMSprop(lr=1e-4),
            metrics=['accuracy'])

        history = model.fit_generator(train_generator, steps_per_epoch=32, epochs=75,
            validation_data=val_generator, validation_steps=
12,
            verbose=1)
```

```
In [ ]: f, (ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))
t = f.suptitle('CNN Model 2- with image argumentation', fontsize=12)
f.subplots_adjust(top=0.85, wspace=0.3)

epoch_list = list(range(1,76))
ax1.plot(epoch_list, history.history['accuracy'], label='Train Accuracy')
ax1.plot(epoch_list, history.history['val_accuracy'], label='Validation Accuracy')
ax1.set_xticks(np.arange(0, 76, 5))
ax1.set_ylabel('Accuracy Value')
ax1.set_xlabel('Epoch')
ax1.set_title('Accuracy')
l1 = ax1.legend(loc="best")

ax2.plot(epoch_list, history.history['loss'], label='Train Loss')
ax2.plot(epoch_list, history.history['val_loss'], label='Validation Loss')
ax2.set_xticks(np.arange(0, 76, 5))
ax2.set_ylabel('Loss Value')
ax2.set_xlabel('Epoch')
ax2.set_title('Loss')
l2 = ax2.legend(loc="best")
```