1. Download the data from here 2. Code the model to classify data like below image 3. Write your own callback function, that has to print the micro F1 score and AUC score after each epoch. 4. Save your model at every epoch if your validation accuracy is improved from previous epoch. 5. you have to decay learning based on below conditions Cond1. If your validation accuracy at that epoch is less than previous epoch accuracy, you have to decrese the learning rate by 10%. Cond2. For every 3rd epoch, decay your learning rate by 5%. 6. If you are getting any NaN values(either weigths or loss) while training, you have to terminate your training. 7. You have to stop the training if your validation accuracy is not increased in last 2 epochs. 8. Use tensorboard for every model and analyse your gradients. (you need to upload the screenshots for each model for evaluation) 9. use cross entropy as loss function 10. Try the architecture params as given below. In [111... from google.colab import drive drive.mount('/content/drive') Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True). In [59]: import tensorflow as tf import datetime import numpy as np import pandas as pd from warnings import filterwarnings filterwarnings(action='ignore', module='tensorflow') filterwarnings(action='ignore', module='tensorboard') In [60]: ## Point 1 data = pd.read_csv('/content/drive/MyDrive/Callbacks/data.csv') print (data.shape) data.head() (20000, 3)Out[60]: f2 label **0** 0.450564 1.074305 0.0 **1** 0.085632 0.967682 **2** 0.117326 0.971521 1.0 0.982179 -0.380408 **4** -0.720352 0.955850 In [61]: y = data['label'].values X = data.drop(['label'], axis=1) In [68]: np.unique(y,return_counts=True) Out[68]: (array([0., 1.]), array([10000, 10000])) In [66]: from sklearn.model selection import train test split X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, stratify=y) In [70]: print (X_train.shape) print (X_test.shape) (13400, 2)(6600, 2)In [71]: validation_data=(X_test,y_test) train_data=(X_train,y_train) Model-1 1. Use tanh as an activation for every layer except output layer. 2. use SGD with momentum as optimizer. 3. use RandomUniform(0,1) as initilizer. 3. Analyze your output and training process. import tensorflow.keras.backend as K from tensorflow.keras.callbacks import Callback from tensorflow.keras.callbacks import ModelCheckpoint from tensorflow.keras.callbacks import EarlyStopping from tensorflow.keras.callbacks import LearningRateScheduler from tensorflow.keras.callbacks import ReduceLROnPlateau from tensorflow.keras.layers import Dense from tensorflow.keras.layers import Input from tensorflow.keras.layers import Activation from tensorflow.keras.models import load_model from tensorflow.keras.models import Sequential from tensorflow.keras.models import Model from sklearn.metrics import confusion_matrix from sklearn.metrics import f1_score from sklearn.metrics import precision_score from sklearn.metrics import recall_score from sklearn.metrics import roc_auc_score In [73]: ## Point 2 def create_model(): return tf.keras.models.Sequential([tf.keras.layers.Dense(2, input_shape=(2,), kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, see tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, see tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, see tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, see tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, see tf.keras.layers.Dense(1, activation="sigmoid",name='output_layer')]) In [74]: ## Point 3 - Micro F1 Score class perf_metrics(Callback): def on_train_begin(self, logs={}): self.f1score = [] self.precisions = [] self.recalls = [] def on_epoch_end(self, epoch, logs={}): y_predict = (np.asarray(self.model.predict(X_test))).round() = y_test val_f1_temp = f1_score(y_actual,y_predict) val_precision_temp = precision_score(y_actual,y_predict) val_recall_temp = recall_score(y_actual,y_predict) self.f1score.append(val_f1_temp) self.recalls.append(val_recall_temp) self.precisions.append(val_precision_temp) print ("F1 Score : %f, Precision : %f, Recall : %f" %(val_f1_temp, val_precision_temp, val_recall_temp)) return performance_metrics = perf_metrics() In [75]: ## Point 3 AUC class auc_score(Callback): def __init__(self, training_data, validation_data): self.x_train = training_data[0] self.y_train = training_data[1] self.x_test = validation_data[0] self.y_test = validation_data[1] def on_epoch_end(self, epoch, logs={}): y_pred_train = self.model.predict_proba(self.x_train) roc_train = roc_auc_score(self.y_train, y_pred_train) y_pred_test = self.model.predict_proba(self.x_test) roc_test = roc_auc_score(self.y_test, y_pred_test) print('ROC_AUC train : %s, ROC_AUC test : %s' % (str(round(roc_train,4)), str(round(roc_test,4)))+'\n') return roc = auc_score(training_data=(X_train, y_train), validation_data=(X_test, y_test)) In [76]: ## Point 4 filepath = "/content/drive/MyDrive/Callbacks/model_save/"+"model_1"+"_weights-{epoch:02d}-{val_accuracy:.4f}.hdf5" = ModelCheckpoint(filepath=filepath, monitor='val_accuracy', verbose=1, save_best_only=True, mode='max') checkpoint In [77]: ## Point 5-Condition1 reduce_lr_accuracy = ReduceLROnPlateau(monitor='val_accuracy', factor=0.9, patience=1, verbose=1, mode='auto') In [78]: ## Point 6 class TerminateNaN(Callback): def on_epoch_end(self, epoch, logs={}): loss = logs.get('loss') if loss is not None: if np.isnan(loss) or np.isinf(loss): print("Invalid loss and terminated at epoch {}".format(epoch)) self.model.stop_training = True terminatenan = TerminateNaN() In [79]: ## Point 7 earlystop = EarlyStopping(monitor='val_accuracy', patience=2, verbose=1, mode='max') In [80]: ## Point 8 log_dir ="/content/drive/MyDrive/Callbacks/logs/fit/"+"model_1_" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S") tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir, histogram_freq=1, write_graph=True, write_grads=True) WARNING:tensorflow:`write_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback. In [81]: ## Point 5-Condition2 def scheduler(epoch, lr): # print (lr) # print (type(lr)) **if**((epoch+1)%3==0): # print (type(lr* 0.95)) **return** (lr * 0.95) else: return 1r lrschedule = LearningRateScheduler(scheduler, verbose=1) In [82]: model_1 = create_model() optimizer = tf.keras.optimizers.SGD(learning_rate=0.01, momentum=0.9, nesterov=False, name="SGD") model_1.compile(optimizer=optimizer,loss='binary_crossentropy',metrics=['accuracy']) callbacks_list = [performance_metrics,roc,checkpoint,reduce_lr_accuracy,lrschedule,terminatenan,earlystop,tensorboard_callback] model_1.fit(x=X_train,y=y_train,epochs=20,validation_data=(X_test, y_test),callbacks=[callbacks_list]) Epoch 1/20 Epoch 00001: LearningRateScheduler reducing learning rate to 0.009999999776482582. 3/419 [......] - ETA: 27s - loss: 0.7776 - accuracy: 0.4792 WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0016s vs `on_train_batch_end` time: 0.0222s). Check your callbacks. F1 Score : 0.513137, Precision : 0.497580, Recall : 0.529697 ROC_AUC train: 0.5065, ROC_AUC test: 0.4917 Epoch 00001: val_accuracy improved from -inf to 0.49742, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-01-0.4974.hdf5 Epoch 2/20 Epoch 00002: LearningRateScheduler reducing learning rate to 0.009999999776482582. /usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: UndefinedMetricWarning: Precision is ill-defined and being set to 0.0 due to n o predicted samples. Use `zero_division` parameter to control this behavior. _warn_prf(average, modifier, msg_start, len(result)) F1 Score : 0.000000, Precision : 0.000000, Recall : 0.000000 ROC_AUC train : 0.5066, ROC_AUC test : 0.4917 Epoch 00002: val_accuracy improved from 0.49742 to 0.50000, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-02-0.5000.hdf5 Epoch 3/20 Epoch 00003: LearningRateScheduler reducing learning rate to 0.009499999787658453. F1 Score: 0.581192, Precision: 0.510190, Recall: 0.675152 ROC_AUC train : 0.5066, ROC_AUC test : 0.4917 Epoch 00003: val_accuracy improved from 0.50000 to 0.51348, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-03-0.5135.hdf5 Epoch 4/20 Epoch 00004: LearningRateScheduler reducing learning rate to 0.009499999694526196. F1 Score: 0.641101, Precision: 0.523938, Recall: 0.825758 ROC_AUC train : 0.5067, ROC_AUC test : 0.4917 Epoch 00004: val_accuracy improved from 0.51348 to 0.53773, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-04-0.5377.hdf5 Epoch 5/20 Epoch 00005: LearningRateScheduler reducing learning rate to 0.009499999694526196. F1 Score: 0.575526, Precision: 0.508605, Recall: 0.662727 ROC_AUC train : 0.5648, ROC_AUC test : 0.5584 Epoch 00005: val_accuracy did not improve from 0.53773 Epoch 00005: ReduceLROnPlateau reducing learning rate to 0.008549999725073577. Epoch 6/20 Epoch 00006: LearningRateScheduler reducing learning rate to 0.008122499473392964. F1 Score: 0.646010, Precision: 0.524693, Recall: 0.840303 ROC_AUC train: 0.5827, ROC_AUC test: 0.575 Epoch 00006: val_accuracy improved from 0.53773 to 0.53955, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-06-0.5395.hdf5 Epoch 7/20 Epoch 00007: LearningRateScheduler reducing learning rate to 0.008122499100863934. F1 Score: 0.648242, Precision: 0.524121, Recall: 0.849394 ROC_AUC train : 0.5946, ROC_AUC test : 0.5858 Epoch 00007: val_accuracy did not improve from 0.53955 Epoch 00007: ReduceLROnPlateau reducing learning rate to 0.00731024919077754. Epoch 8/20 Epoch 00008: LearningRateScheduler reducing learning rate to 0.007310249377042055. F1 Score: 0.658937, Precision: 0.524176, Recall: 0.886970 ROC_AUC train : 0.5996, ROC_AUC test : 0.5984 Epoch 00008: val_accuracy improved from 0.53955 to 0.54091, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-08-0.5409.hdf5 Epoch 9/20 Epoch 00009: LearningRateScheduler reducing learning rate to 0.006944736908189952. F1 Score: 0.660384, Precision: 0.524319, Recall: 0.891818 ROC_AUC train : 0.6003, ROC_AUC test : 0.598 Epoch 00009: val_accuracy improved from 0.54091 to 0.54136, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-09-0.5414.hdf5 Epoch 10/20 Epoch 00010: LearningRateScheduler reducing learning rate to 0.006944736931473017. F1 Score : 0.660225, Precision : 0.524643, Recall : 0.890303 ROC_AUC train : 0.6005, ROC_AUC test : 0.5988 Epoch 00010: val_accuracy improved from 0.54136 to 0.54182, saving model to /content/drive/MyDrive/Callbacks/model_save/model_1_weights-10-0.5418.hdf5 Epoch 11/20 Epoch 00011: LearningRateScheduler reducing learning rate to 0.006944736931473017. F1 Score : 0.653592, Precision : 0.523949, Recall : 0.868485 ROC_AUC train : 0.6, ROC_AUC test : 0.5979 Epoch 00011: val_accuracy did not improve from 0.54182 Epoch 00011: ReduceLROnPlateau reducing learning rate to 0.006250263238325715. Epoch 12/20 Epoch 00012: LearningRateScheduler reducing learning rate to 0.005937750032171607. F1 Score: 0.656720, Precision: 0.523707, Recall: 0.880303 ROC_AUC train : 0.6018, ROC_AUC test : 0.5999 Epoch 00012: val_accuracy did not improve from 0.54182 Epoch 00012: ReduceLROnPlateau reducing learning rate to 0.005343974987044931. Epoch 00012: early stopping <tensorflow.python.keras.callbacks.History at 0x7f3862335090> In [83]: model_1.summary() Model: "sequential_18" Layer (type) Output Shape Param # _____ layer1 (Dense) (None, 2) 6 (None, 2) layer2 (Dense) 6 layer3 (Dense) (None, 2) 6 layer4 (Dense) (None, 2) 6 layer5 (Dense) (None, 2) 6 output_layer (Dense) (None, 1) 3 Total params: 33 Trainable params: 33 Non-trainable params: 0 In [84]: %reload_ext tensorboard !rm -rf ./logs/ %tensorboard --logdir='/content/drive/MyDrive/Callbacks/logs/fit/' Output hidden; open in https://colab.research.google.com to view. Model-2 1. Use relu as an activation for every layer except output layer. 2. use SGD with momentum as optimizer. use RandomUniform(0,1) as initilizer. 3. Analyze your output and training process. In [86]: def create_model2(): return tf.keras.models.Sequential([tf.keras.layers.Dense(2, input_shape=(2,),kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, seed tf.keras.layers.Dense(2, activation='relu', kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, seed tf.keras.layers.Dense(2, activation='relu', kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, seed tf.keras.layers.Dense(2, activation='relu',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, seed tf.keras.layers.Dense(2,activation='relu',kernel_initializer=tf.keras.initializers.RandomUniform(minval=0, maxval=1, seed tf.keras.layers.Dense(1,activation='sigmoid',name='output_layer')]) In [88]: filepath = "/content/drive/MyDrive/Callbacks/model_save/"+"model_2"+"_weights-{epoch:02d}-{val_accuracy:.4f}.hdf5" checkpoint = ModelCheckpoint(filepath=filepath, monitor='val_accuracy', verbose=1, save_best_only=True, mode='max') ="/content/drive/MyDrive/Callbacks/logs/fit/"+"model_2_" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S") log_dir tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir,histogram_freq=1, write_graph=True,write_grads=True) WARNING:tensorflow:`write_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback. In [89]: model_2 = create_model2() optimizer = tf.keras.optimizers.SGD(learning_rate=0.01, momentum=0.9, nesterov=False, name="SGD") model_2.compile(optimizer=optimizer, loss='binary_crossentropy', metrics=['accuracy']) callbacks_list = [checkpoint, tensorboard_callback, earlystop, reduce_lr_accuracy, lrschedule, terminatenan, performance_metrics, roc] model_2.fit(x=X_train,y=y_train,epochs=20,validation_data=(X_test, y_test),callbacks=[callbacks_list]) Epoch 1/20 Epoch 00001: LearningRateScheduler reducing learning rate to 0.009999999776482582. 3/419 [.....] - ETA: 26s - loss: 0.6647 - accuracy: 0.5729 WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0024s vs `on_train_batch_end` time: 0.0213s). Check your callbacks. Epoch 00001: val_accuracy improved from -inf to 0.53379, saving model to /content/drive/MyDrive/Callbacks/model_save/model_2_weights-01-0.5338.hdf5 F1 Score : 0.630834, Precision : 0.522145, Recall : 0.796667 ROC_AUC train : 0.5515, ROC_AUC test : 0.5388 Epoch 2/20 Epoch 00002: LearningRateScheduler reducing learning rate to 0.009999999776482582. Epoch 00002: val_accuracy improved from 0.53379 to 0.54242, saving model to /content/drive/MyDrive/Callbacks/model_save/model_2_weights-02-0.5424.hdf5 F1 Score : 0.661511, Precision : 0.524902, Recall : 0.894242 ROC_AUC train : 0.5525, ROC_AUC test : 0.5441 Epoch 3/20 Epoch 00003: LearningRateScheduler reducing learning rate to 0.009499999787658453. Epoch 00003: val_accuracy improved from 0.54242 to 0.54273, saving model to /content/drive/MyDrive/Callbacks/model_save/model_2_weights-03-0.5427.hdf5 F1 Score : 0.657279, Precision : 0.525608, Recall : 0.876970 ROC_AUC train : 0.5521, ROC_AUC test : 0.543 Epoch 4/20 Epoch 00004: LearningRateScheduler reducing learning rate to 0.009499999694526196. Epoch 00004: val_accuracy did not improve from 0.54273 Epoch 00004: ReduceLROnPlateau reducing learning rate to 0.008549999725073577. F1 Score : 0.650017, Precision : 0.524030, Recall : 0.855758 ROC_AUC train : 0.5506, ROC_AUC test : 0.5422 Epoch 5/20 Epoch 00005: LearningRateScheduler reducing learning rate to 0.008549999445676804. Epoch 00005: val_accuracy did not improve from 0.54273 Epoch 00005: ReduceLROnPlateau reducing learning rate to 0.007694999501109123. F1 Score: 0.642010, Precision: 0.524056, Recall: 0.828485 ROC_AUC train : 0.5507, ROC_AUC test : 0.5392 Epoch 00005: early stopping <tensorflow.python.keras.callbacks.History at 0x7f38620e0990> Out[89]: In [90]: model_2.summary() Model: "sequential_19" Layer (type) Output Shape Param # layer1 (Dense) (None, 2) layer2 (Dense) (None, 2) 6 layer3 (Dense) (None, 2) 6 layer4 (Dense) (None, 2) 6 layer5 (Dense) (None, 2) 6 output_layer (Dense) 3 (None, 1) _____ Total params: 33 Trainable params: 33 Non-trainable params: 0 In [91]: %reload_ext tensorboard !rm -rf ./logs/ %tensorboard --logdir='/content/drive/MyDrive/Callbacks/logs/fit/' Output hidden; open in https://colab.research.google.com to view. Model-3 1. Use relu as an activation for every layer except output layer. 2. use SGD with momentum as optimizer. use he_uniform() as initilizer. 3. Analyze your output and training process. In [92]: def create_model3(): return tf.keras.models.Sequential([tf.keras.layers.Dense(2, input_shape=(2,),kernel_initializer=tf.keras.initializers.HeUniform(),name='layer1'), tf.keras.layers.Dense(2, activation='relu',kernel_initializer=tf.keras.initializers.HeUniform(),name='layer2'), tf.keras.layers.Dense(2, activation='relu',kernel_initializer=tf.keras.initializers.HeUniform(),name='layer3'), tf.keras.layers.Dense(2, activation='relu',kernel_initializer=tf.keras.initializers.HeUniform(),name='layer4'), tf.keras.layers.Dense(2,activation='relu',kernel_initializer=tf.keras.initializers.HeUniform(),name='layer5'), tf.keras.layers.Dense(1,activation="sigmoid",name='output_layer')]) In [93]: = "/content/drive/MyDrive/Callbacks/model_save/"+"model_3"+"_weights-{epoch:02d}-{val_accuracy:.4f}.hdf5" filepath checkpoint = ModelCheckpoint(filepath=filepath, monitor='val_accuracy', verbose=1, save_best_only=True, mode='max') ="/content/drive/MyDrive/Callbacks/logs/fit/"+"model_3_" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S") tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir, histogram_freq=1, write_graph=True, write_grads=True) WARNING:tensorflow:`write_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback. In [95]: model_3 = create_model3() = tf.keras.optimizers.SGD(learning_rate=0.01, momentum=0.9, nesterov=False, name="SGD") model_3.compile(optimizer=optimizer, loss='binary_crossentropy', metrics=['accuracy']) callbacks_list = [checkpoint, tensorboard_callback, earlystop, reduce_lr_accuracy, lrschedule, terminatenan, performance_metrics, roc] model_3.fit(x=X_train,y=y_train,epochs=20,validation_data=(X_test, y_test),callbacks=[callbacks_list]) Epoch 1/20 Epoch 00001: LearningRateScheduler reducing learning rate to 0.009999999776482582. 3/419 [.....] - ETA: 20s - loss: 0.8054 - accuracy: 0.4479 WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0018s vs `on_train_batch_end` time: 0.0158s). Check your callbacks. Epoch 00001: val_accuracy improved from 0.50000 to 0.53242, saving model to /content/drive/MyDrive/Callbacks/model_save/model_3_weights-01-0.5324.hdf5 F1 Score : 0.626392, Precision : 0.521573, Recall : 0.783939 ROC_AUC train : 0.5439, ROC_AUC test : 0.535 Epoch 2/20 Epoch 00002: LearningRateScheduler reducing learning rate to 0.009999999776482582. Epoch 00002: val_accuracy improved from 0.53242 to 0.57909, saving model to /content/drive/MyDrive/Callbacks/model_save/model_3_weights-02-0.5791.hdf5 F1 Score: 0.629006, Precision: 0.562321, Recall: 0.713636 ROC_AUC train : 0.6026, ROC_AUC test : 0.6002 Epoch 3/20 Epoch 00003: LearningRateScheduler reducing learning rate to 0.009499999787658453. Epoch 00003: val_accuracy did not improve from 0.57909 Epoch 00003: ReduceLROnPlateau reducing learning rate to 0.008549999725073577. F1 Score : 0.652807, Precision : 0.547263, Recall : 0.808788 ROC_AUC train: 0.5915, ROC_AUC test: 0.5842 Epoch 4/20 Epoch 00004: LearningRateScheduler reducing learning rate to 0.008549999445676804. Epoch 00004: val_accuracy improved from 0.57909 to 0.58939, saving model to /content/drive/MyDrive/Callbacks/model_save/model_3_weights-04-0.5894.hdf5 F1 Score: 0.665432, Precision: 0.561458, Recall: 0.816667 ROC_AUC train : 0.6047, ROC_AUC test : 0.6067 Epoch 5/20 Epoch 00005: LearningRateScheduler reducing learning rate to 0.008549999445676804. Epoch 00005: val_accuracy did not improve from 0.58939 Epoch 00005: ReduceLROnPlateau reducing learning rate to 0.007694999501109123. F1 Score : 0.643603, Precision : 0.565367, Recall : 0.746970 ROC_AUC train : 0.6068, ROC_AUC test : 0.6028 Epoch 6/20 Epoch 00006: LearningRateScheduler reducing learning rate to 0.007310249703004956. Epoch 00006: val_accuracy did not improve from 0.58939 Epoch 00006: ReduceLROnPlateau reducing learning rate to 0.0065792248584330085. F1 Score : 0.632246, Precision : 0.569451, Recall : 0.710606 ROC_AUC train : 0.6086, ROC_AUC test : 0.604 Epoch 00006: early stopping <tensorflow.python.keras.callbacks.History at 0x7f385d3a9390> Out[95]: In [96]: model_3.summary() Model: "sequential_21" Layer (type) Output Shape Param # layer1 (Dense) (None, 2) 6 layer2 (Dense) (None, 2) 6 layer3 (Dense) (None, 2) 6 layer4 (Dense) (None, 2) 6 layer5 (Dense) (None, 2) 6 output_layer (Dense) (None, 1) 3 ______ Total params: 33 Trainable params: 33 Non-trainable params: 0 In [97]: %reload_ext tensorboard !rm -rf ./logs/ %tensorboard --logdir='/content/drive/MyDrive/Callbacks/logs/fit/' Output hidden; open in https://colab.research.google.com to view. Model-4 1. Try with any values to get better accuracy/f1 score. In [106... def create_model4(): return tf.keras.models.Sequential([tf.keras.layers.Dense(2, input_shape=(2,),kernel_initializer=tf.keras.initializers.GlorotNormal(seed=None),name='layer1') tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.GlorotNormal(seed=None),name='layer2'), tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.GlorotNormal(seed=None),name='layer3'), tf.keras.layers.Dense(2, activation='tanh',kernel_initializer=tf.keras.initializers.GlorotNormal(seed=None),name='layer4'), tf.keras.layers.Dense(2,activation='tanh',kernel_initializer=tf.keras.initializers.GlorotNormal(seed=None),name='layer5'), tf.keras.layers.Dense(1,activation="sigmoid",name='output_layer')]) In [107.. filepath = "/content/drive/MyDrive/Callbacks/model_save/"+"model_4"+"_weights-{epoch:02d}-{val_accuracy:.4f}.hdf5" = ModelCheckpoint(filepath=filepath, monitor='val_accuracy', verbose=1, save_best_only=True, mode='max') checkpoint ="/content/drive/MyDrive/Callbacks/logs/fit/"+"model 4 " + datetime.datetime.now().strftime("%Y%m%d-%H%M%S") log dir tensorboard_callback = tf.keras.callbacks.TensorBoard(log_dir=log_dir, histogram_freq=1, write_graph=True, write_grads=True) WARNING:tensorflow:`write_grads` will be ignored in TensorFlow 2.0 for the `TensorBoard` Callback. In [108... model_4 = create_model4() = tf.keras.optimizers.SGD(learning_rate=0.01, momentum=0.9, nesterov=False, name="SGD") model_4.compile(optimizer=optimizer, loss='binary_crossentropy', metrics=['accuracy']) callbacks_list = [checkpoint, tensorboard_callback, earlystop, reduce_lr_accuracy, lrschedule, terminatenan, performance_metrics, roc] model_4.fit(x=X_train,y=y_train,epochs=20,validation_data=(X_test, y_test),callbacks=[callbacks_list]) Epoch 1/20 Epoch 00001: LearningRateScheduler reducing learning rate to 0.009999999776482582. 3/419 [......] - ETA: 27s - loss: 0.6908 - accuracy: 0.5521 WARNING:tensorflow:Callback method `on_train_batch_end` is slow compared to the batch time (batch time: 0.0023s vs `on_train_batch_end` time: 0.0215s). Check your callbacks. Epoch 00001: val_accuracy improved from -inf to 0.50000, saving model to /content/drive/MyDrive/Callbacks/model_save/model_4_weights-01-0.5000.hdf5 F1 Score : 0.666667, Precision : 0.500000, Recall : 1.000000 ROC_AUC train : 0.5309, ROC_AUC test : 0.5213 Epoch 2/20 Epoch 00002: LearningRateScheduler reducing learning rate to 0.009999999776482582. Epoch 00002: val_accuracy improved from 0.50000 to 0.57439, saving model to /content/drive/MyDrive/Callbacks/model_save/model_4_weights-02-0.5744.hdf5 F1 Score: 0.592840, Precision: 0.568213, Recall: 0.619697 ROC_AUC train: 0.5958, ROC_AUC test: 0.5827 Epoch 3/20 Epoch 00003: LearningRateScheduler reducing learning rate to 0.009499999787658453. Epoch 00003: val_accuracy improved from 0.57439 to 0.60258, saving model to /content/drive/MyDrive/Callbacks/model_save/model_4_weights-03-0.6026.hdf5 F1 Score: 0.656900, Precision: 0.577906, Recall: 0.760909 ROC_AUC train : 0.6092, ROC_AUC test : 0.6022 Epoch 4/20 Epoch 00004: LearningRateScheduler reducing learning rate to 0.009499999694526196. Epoch 00004: val_accuracy did not improve from 0.60258 Epoch 00004: ReduceLROnPlateau reducing learning rate to 0.008549999725073577. F1 Score: 0.670635, Precision: 0.567590, Recall: 0.819394 ROC_AUC train : 0.605, ROC_AUC test : 0.5991 Epoch 5/20 Epoch 00005: LearningRateScheduler reducing learning rate to 0.008549999445676804. Epoch 00005: val_accuracy did not improve from 0.60258 Epoch 00005: ReduceLROnPlateau reducing learning rate to 0.007694999501109123. F1 Score: 0.659665, Precision: 0.576479, Recall: 0.770909 ROC_AUC train : 0.6115, ROC_AUC test : 0.6028 Epoch 00005: early stopping <tensorflow.python.keras.callbacks.History at 0x7f3853d53390> In [109. model_4.summary() Model: "sequential_26" Layer (type) Output Shape Param # layer1 (Dense) (None, 2) 6 layer2 (Dense) (None, 2) 6 layer3 (Dense) (None, 2) 6 layer4 (Dense) (None, 2) 6 layer5 (Dense) (None, 2) 6 output_layer (Dense) 3 (None, 1) Total params: 33 Trainable params: 33 Non-trainable params: 0 In [110.. %reload_ext tensorboard !rm -rf ./logs/ %tensorboard --logdir='/content/drive/MyDrive/Callbacks/logs/fit/' Output hidden; open in https://colab.research.google.com to view. In []: