Assignment_3_Ahmad_Sayeb

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0.0.1 AIDI 1002 Assignment 3

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```
[167]: # ------ Libraries-----
      import pandas as pd
      import numpy as np
      # ----- Plotting-----
      import matplotlib.pyplot as plt
      import seaborn as sns
      # ----- Sklearn libraries-----
      from sklearn.model_selection import train_test_split
      from sklearn.preprocessing import LabelEncoder
      from sklearn.preprocessing import OneHotEncoder
      from sklearn.preprocessing import MinMaxScaler
      from sklearn.metrics import f1_score, accuracy_score, recall_score,
       ⇔precision score
      from sklearn.metrics import roc_curve,roc_auc_score
      from imblearn.over_sampling import SMOTE
      # ----- Keras--
      from tensorflow import keras
      from tensorflow.keras.models import Sequential
      from tensorflow.keras.layers import Dense, Dropout
      from tensorflow.keras.callbacks import EarlyStopping
      from keras import regularizers
      # -----Tensor-flow support-----
      import tensorflow addons as tfa
      # -----Warnings-----
      #removing cuda warnings for gpu
      import warnings
      warnings.filterwarnings('ignore')
```

0.0.3 All the Necessary Functions

```
[246]: def load_csv(path: str) -> 'dataframe':

Loads csv file into dataframe

path: path to the file
```

```
df = pd.read_csv(path)
   return df
def num_col_nan(df: 'dataframe'):
   replaces nan value in numerical columns with mode
    df: dataframe
   numerics = ['int16', 'int32', 'int64', 'float16', 'float32', 'float64']
   col_num = df.select_dtypes(include=numerics)
   for col in col_num:
        df[col].replace([np.nan], df[col].mode()[0], inplace=True)
def cat_col_nan(df: 'dataframe'):
   replace nan value in categorical column with None string
    df: dataframe
    111
    categorical = ['object']
   cat_columns = df.select_dtypes(include=categorical)
   for col in cat_columns:
        df[col].replace([np.nan], 'None', inplace=True)
def encoder(df: 'dataframe', train=True, only_label=False) -> 'dataframe':
    label encoding categorical data that are indep variable
    and one hot encode target variables
    df: dataframe
   label = LabelEncoder()
    categorical = ['object']
   cat_col = df.select_dtypes(include=categorical)
    if train and only_label:
       target_only_label = cat_col[['Var_1']]
       cat_col.drop(columns=['Var_1'], inplace=True)
       for col in cat_col:
            df[col] = label.fit_transform(df[col])
        df.drop(columns=['Var_1'], inplace=True)
        return pd.concat([df, target_only_label], axis=1)
```

```
if train and not only_label:
        target = cat_col[['Var_1']]
        cat_col.drop(columns=['Var_1'], inplace=True)
    for col in cat_col:
        df[col] = label.fit_transform(df[col])
    if train and not only label:
        target = pd.get_dummies(target)
        df.drop(columns=['Var_1'], inplace=True)
        return pd.concat([df, target], axis=1)
    if not train and not only_label:
        return df
def replacing_classes(row: 'object'):
    if row != 'Cat_4' and row != 'Cat_6':
        return 'Other'
    else:
        return row
def drop_cols(cols: list, df: 'dataframe'):
    drops specified columns
    col: list of columns
    df: dataframe
    111
    print(f'dropping {cols}...')
    df.drop(columns=cols, inplace=True)
def pre_processing(df: 'dataframe', train=True, only_label=False):
    performs nan value replacement and encoding categorical values
    try:
        if train:
            print('dropping columns...')
            drop_cols(['ID', 'Segmentation','Age'], df)
        if not train:
            print('dropping ID column')
            drop_cols(['ID','Var_1','Age'], df)
        print('replacing numerical nans with mode...')
        num_col_nan(df)
        print('replacing categorical nans with None string...')
        cat_col_nan(df)
```

```
if train:
            print('changing anything otehr than cat_6 and cat_4 to other...')
            df['Var_1'] = df['Var_1'].apply(replacing_classes)
       print('label encoding categorical data...')
        df = encoder(df, train, only_label)
       print('\033[1m' + 'SUCCESSFULLY PERFORMED PREPROCESSING' + '\033[0m')
       return df
    except Exception as e:
       print('error occurred in pre-processing')
       print(e)
       return False
def train_validation_split(val_size: float, df: 'dataframe'):
    111
    splits dataframe into train and validation and SHUFFLES
    test_size: size of the validation
    df: dataframe
    # Shuffle is set to true
   # Stratify is set to true
   df_dep = df[['Var_1_Cat_4','Var_1_Cat_6', 'Var_1_Other']]
   df_indep = df.drop(columns=['Var_1_Cat_4','Var_1_Cat_6', 'Var_1_Other'])
   X_train, X_valid, y_train, y_valid = train_test_split(
                                                        df_indep,
                                                        df_dep,
                                                        test_size=val_size,
                                                        shuffle=True,
 stratify=df_dep[['Var_1_Cat_4', 'Var_1_Cat_6', 'Var_1_Other']]
   return X_train, X_valid, y_train, y_valid
def min_max_scaler(df: 'dataframe'):
    normalize numerical data
   df: dataframe
    scaler = MinMaxScaler()
    col_num = ['Work_Experience', 'Family_Size']
   df[col_num] = scaler.fit_transform(df[col_num])
def fit_model(X_train: 'dataframe',
```

```
y_train: 'dataframe',
              X_val: 'dataframe',
              y_val: 'dataframe'
    111
    This builds and trains the model
    X_trian: training data
    111
    model = Sequential()
    model.add(Dense(512, input_shape=(7,)))
    model.add(Dense(256, activation='relu'))
    model.add(Dense(128))
    # dropout for regularization
    model.add(Dropout(0.2))
    model.add(Dense(64, activation='relu'))
    model.add(Dropout(0.35))
    model.add(Dense(16, activation='relu'))
    model.add(Dropout(0.5))
    model.add(Dense(8, activation='relu'))
    model.add(Dense(3, activation='softmax'))
    # metric is accuracy and f1_score
    model.compile(loss='categorical_crossentropy', optimizer='adam',__
 ometrics=[tfa.metrics.F1Score(average='micro', num_classes=3), 'accuracy'])
    print(model.summary())
    X_train_array = X_train.values
    y_train_array = y_train.values
    X_val_array = X_val.values
    y_val_array = y_val.values
    # callback to stop training if reverse learning happens
    # Patience is set to 200, stop wont happen before 200 epochs
    callbacks = EarlyStopping(monitor='val_loss', mode='min', patience=200)
    history = model.fit(X_train_array,
                        y_train_array,
                        epochs=500,
                        verbose=1,
                        validation_data=(X_val_array, y_val_array),
                       callbacks=[callbacks])
    return history, model
# Link: https://github.com/vinyluis/Articles/blob/main/
 -ROC%20Curve%20and%20ROC%20AUC/ROC%20Curve%20-%20Multiclass.ipynb
def get_all_roc_coordinates(y_real, y_proba):
    111
    Calculates all the ROC Curve coordinates (tpr and fpr) by considering each \sqcup
 ⇒point as a threshold for the prediction of the class.
```

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Arqs:
        y_real: The list or series with the real classes.
        y proba: The array with the probabilities for each class, obtained by \Box
 \neg using the `.predict_proba()` method.
    Returns:
        tpr_list: The list of TPRs representing each threshold.
        fpr_list: The list of FPRs representing each threshold.
    tpr_list = [0]
    fpr_list = [0]
    for i in range(len(y_proba)):
        threshold = y_proba[i]
        y_pred = y_proba >= threshold
        tpr, fpr = calculate_tpr_fpr(y_real, y_pred)
        tpr_list.append(tpr)
        fpr_list.append(fpr)
    return tpr_list, fpr_list
# Link: https://github.com/vinyluis/Articles/blob/main/
 →ROC%20Curve%20and%20ROC%20AUC/ROC%20Curve%20-%20Multiclass.ipynb
def plot_roc_curve(tpr, fpr, scatter = True, ax = None):
    111
    Plots the ROC Curve by using the list of coordinates (tpr and fpr).
    Args:
        tpr: The list of TPRs representing each coordinate.
        fpr: The list of FPRs representing each coordinate.
        scatter: When True, the points used on the calculation will be plotted \sqcup
 \rightarrow with the line (default = True).
    111
    if ax == None:
        plt.figure(figsize = (5, 5))
        ax = plt.axes()
    if scatter:
        sns.scatterplot(x = fpr, y = tpr, ax = ax)
    sns.lineplot(x = fpr, y = tpr, ax = ax)
    sns.lineplot(x = [0, 1], y = [0, 1], color = 'green', ax = ax)
    plt.xlim(-0.05, 1.05)
    plt.ylim(-0.05, 1.05)
    plt.xlabel("False Positive Rate")
    plt.ylabel("True Positive Rate")
```

```
# Link: https://qithub.com/vinyluis/Articles/blob/main/
 ~ROC%20Curve%20and%20ROC%20AUC/ROC%20Curve%20-%20Multiclass.ipynb
def calculate_tpr_fpr(y_real, y_pred):
    Calculates the True Positive Rate (tpr) and the True Negative Rate (fpr)_{\sqcup}
 ⇒based on real and predicted observations
    Arqs:
        y_real: The list or series with the real classes
        y_pred: The list or series with the predicted classes
    Returns:
        tpr: The True Positive Rate of the classifier
        fpr: The False Positive Rate of the classifier
    # Calculates the confusion matrix and recover each element
    cm = confusion_matrix(y_real, y_pred)
    TN = cm[0, 0]
    FP = cm[0, 1]
    FN = cm[1, 0]
    TP = cm[1, 1]
    # Calculates tpr and fpr
    tpr = TP/(TP + FN) # sensitivity - true positive rate
    fpr = 1 - TN/(TN+FP) # 1-specificity - false positive rate
    return tpr, fpr
def apply_smote(df: 'dataframe') -> 'dataframe':
    This function applies SMOTE and return dataframe
    X = df.loc[:, df.columns != "Var 1"]
    y = df[['Var_1']]
    sm = SMOTE(sampling_strategy='auto', k_neighbors=1, random_state=100)
    X_res, y_res = sm.fit_resample(X, y)
    return pd.concat([X_res, y_res], axis=1)
```

0.0.4 All Steps without Dealing Minority with Minority Class

Applying the preprocessing function to the train data. Applying min_max_scaler brought no change to the accuracy, f1_score or recall. Since most of the columns are categorical, normalization only apply to work experience column and it will reduce it importance significantly as the other columns are values between 0 to 4 because of label encoding. This is the reason that I avoided normalization. We also train the model with this pipeline. test_validation_split shuffles the data and uses stratify to include all types of classes in validation set.

[248]: history, model = train_pipeline()

dropping columns...

dropping ['ID', 'Segmentation', 'Age']...

replacing numerical nans with mode...

replacing categorical nans with None string...

changing anything otehr than cat_6 and cat_4 to other...

label encoding categorical data...

SUCCESSFULLY PERFORMED PREPROCESSING

Model: "sequential_25"

Layer (type)	Output Shape	Param #
dense_175 (Dense)	(None, 512)	4096
dense_176 (Dense)	(None, 256)	131328
dense_177 (Dense)	(None, 128)	32896
dropout_75 (Dropout)	(None, 128)	0
dense_178 (Dense)	(None, 64)	8256
dropout_76 (Dropout)	(None, 64)	0
dense_179 (Dense)	(None, 16)	1040
dropout_77 (Dropout)	(None, 16)	0
dense_180 (Dense)	(None, 8)	136

```
______
Total params: 177,779
Trainable params: 177,779
Non-trainable params: 0
        ______
None
Epoch 1/500
f1 score: 0.6356 - accuracy: 0.6354 - val_loss: 0.8709 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 2/500
f1_score: 0.6480 - accuracy: 0.6480 - val_loss: 0.8568 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 3/500
227/227 [============ ] - Os 1ms/step - loss: 0.8686 -
f1_score: 0.6488 - accuracy: 0.6488 - val_loss: 0.8576 - val_f1_score: 0.6493 -
val accuracy: 0.6493
Epoch 4/500
f1_score: 0.6485 - accuracy: 0.6485 - val_loss: 0.8652 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 5/500
f1_score: 0.6494 - accuracy: 0.6494 - val_loss: 0.8515 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 6/500
f1_score: 0.6502 - accuracy: 0.6502 - val_loss: 0.8530 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 7/500
227/227 [============ ] - Os 2ms/step - loss: 0.8565 -
f1 score: 0.6502 - accuracy: 0.6502 - val loss: 0.8519 - val f1 score: 0.6493 -
val_accuracy: 0.6493
Epoch 8/500
f1_score: 0.6540 - accuracy: 0.6540 - val_loss: 0.8526 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 9/500
227/227 [============ ] - Os 1ms/step - loss: 0.8520 -
f1_score: 0.6528 - accuracy: 0.6528 - val_loss: 0.8510 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 10/500
f1_score: 0.6535 - accuracy: 0.6535 - val_loss: 0.8474 - val_f1_score: 0.6481 -
val_accuracy: 0.6481
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Epoch 11/500
f1_score: 0.6529 - accuracy: 0.6529 - val_loss: 0.8539 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 12/500
f1_score: 0.6509 - accuracy: 0.6509 - val_loss: 0.8442 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 13/500
227/227 [============ ] - Os 1ms/step - loss: 0.8540 -
f1 score: 0.6513 - accuracy: 0.6513 - val_loss: 0.8442 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 14/500
f1_score: 0.6531 - accuracy: 0.6531 - val_loss: 0.8444 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 15/500
f1_score: 0.6543 - accuracy: 0.6543 - val_loss: 0.8474 - val_f1_score: 0.6468 -
val_accuracy: 0.6468
Epoch 16/500
f1_score: 0.6538 - accuracy: 0.6538 - val_loss: 0.8483 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 17/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8514 -
f1_score: 0.6513 - accuracy: 0.6513 - val_loss: 0.8522 - val_f1_score: 0.6481 -
val_accuracy: 0.6481
Epoch 18/500
f1_score: 0.6507 - accuracy: 0.6507 - val_loss: 0.8478 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 19/500
f1_score: 0.6551 - accuracy: 0.6551 - val_loss: 0.8524 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 20/500
f1_score: 0.6542 - accuracy: 0.6542 - val_loss: 0.8465 - val_f1_score: 0.6468 -
val_accuracy: 0.6468
Epoch 21/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8483 -
f1_score: 0.6547 - accuracy: 0.6547 - val_loss: 0.8571 - val_f1_score: 0.6468 -
val_accuracy: 0.6468
Epoch 22/500
f1_score: 0.6528 - accuracy: 0.6528 - val_loss: 0.8535 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
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Epoch 23/500
f1_score: 0.6564 - accuracy: 0.6564 - val_loss: 0.8546 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 24/500
f1_score: 0.6547 - accuracy: 0.6547 - val_loss: 0.8570 - val_f1_score: 0.6468 -
val_accuracy: 0.6468
Epoch 25/500
227/227 [============ ] - Os 2ms/step - loss: 0.8466 -
f1 score: 0.6539 - accuracy: 0.6539 - val_loss: 0.8517 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 26/500
f1_score: 0.6550 - accuracy: 0.6550 - val_loss: 0.8492 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 27/500
f1_score: 0.6558 - accuracy: 0.6558 - val_loss: 0.8518 - val_f1_score: 0.6481 -
val_accuracy: 0.6481
Epoch 28/500
f1_score: 0.6540 - accuracy: 0.6540 - val_loss: 0.8480 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 29/500
227/227 [============ ] - Os 2ms/step - loss: 0.8460 -
f1_score: 0.6549 - accuracy: 0.6549 - val_loss: 0.8518 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 30/500
f1_score: 0.6539 - accuracy: 0.6539 - val_loss: 0.8476 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 31/500
227/227 [============ ] - Os 2ms/step - loss: 0.8448 -
f1_score: 0.6557 - accuracy: 0.6557 - val_loss: 0.8452 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 32/500
f1_score: 0.6543 - accuracy: 0.6543 - val_loss: 0.8437 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 33/500
227/227 [============ ] - Os 2ms/step - loss: 0.8421 -
f1_score: 0.6521 - accuracy: 0.6521 - val_loss: 0.8503 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 34/500
f1_score: 0.6557 - accuracy: 0.6557 - val_loss: 0.8499 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
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Epoch 35/500
f1_score: 0.6562 - accuracy: 0.6562 - val_loss: 0.8517 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 36/500
f1_score: 0.6551 - accuracy: 0.6551 - val_loss: 0.8443 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 37/500
227/227 [============ ] - Os 2ms/step - loss: 0.8423 -
f1 score: 0.6576 - accuracy: 0.6576 - val_loss: 0.8467 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 38/500
f1_score: 0.6554 - accuracy: 0.6554 - val_loss: 0.8516 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 39/500
f1_score: 0.6550 - accuracy: 0.6550 - val_loss: 0.8455 - val_f1_score: 0.6506 -
val accuracy: 0.6506
Epoch 40/500
f1_score: 0.6543 - accuracy: 0.6543 - val_loss: 0.8407 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 41/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8429 -
f1_score: 0.6551 - accuracy: 0.6551 - val_loss: 0.8398 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 42/500
f1_score: 0.6584 - accuracy: 0.6584 - val_loss: 0.8420 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 43/500
f1 score: 0.6561 - accuracy: 0.6561 - val loss: 0.8459 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 44/500
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8460 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 45/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8401 -
f1_score: 0.6553 - accuracy: 0.6553 - val_loss: 0.8500 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 46/500
f1_score: 0.6558 - accuracy: 0.6558 - val_loss: 0.8480 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
```

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Epoch 47/500
f1_score: 0.6562 - accuracy: 0.6562 - val_loss: 0.8446 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 48/500
f1_score: 0.6558 - accuracy: 0.6558 - val_loss: 0.8438 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 49/500
227/227 [============ ] - Os 2ms/step - loss: 0.8374 -
f1 score: 0.6565 - accuracy: 0.6565 - val_loss: 0.8459 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 50/500
f1_score: 0.6569 - accuracy: 0.6569 - val_loss: 0.8516 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 51/500
f1_score: 0.6560 - accuracy: 0.6560 - val_loss: 0.8508 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 52/500
f1_score: 0.6565 - accuracy: 0.6565 - val_loss: 0.8482 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 53/500
227/227 [============ ] - Os 2ms/step - loss: 0.8375 -
f1_score: 0.6558 - accuracy: 0.6558 - val_loss: 0.8436 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 54/500
f1_score: 0.6558 - accuracy: 0.6558 - val_loss: 0.8514 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 55/500
227/227 [============ ] - 1s 2ms/step - loss: 0.8433 -
f1 score: 0.6560 - accuracy: 0.6560 - val loss: 0.8529 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 56/500
f1_score: 0.6571 - accuracy: 0.6571 - val_loss: 0.8483 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 57/500
227/227 [============ ] - 1s 3ms/step - loss: 0.8334 -
f1_score: 0.6579 - accuracy: 0.6579 - val_loss: 0.8499 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 58/500
f1_score: 0.6571 - accuracy: 0.6571 - val_loss: 0.8411 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
```

```
Epoch 59/500
f1_score: 0.6565 - accuracy: 0.6565 - val_loss: 0.8463 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 60/500
f1_score: 0.6580 - accuracy: 0.6580 - val_loss: 0.8501 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 61/500
227/227 [============ ] - Os 2ms/step - loss: 0.8419 -
f1 score: 0.6560 - accuracy: 0.6560 - val_loss: 0.8472 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 62/500
f1_score: 0.6562 - accuracy: 0.6562 - val_loss: 0.8408 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 63/500
f1_score: 0.6567 - accuracy: 0.6567 - val_loss: 0.8480 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 64/500
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8399 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 65/500
227/227 [============ ] - 1s 3ms/step - loss: 0.8366 -
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8440 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 66/500
f1_score: 0.6569 - accuracy: 0.6569 - val_loss: 0.8494 - val_f1_score: 0.6481 -
val_accuracy: 0.6481
Epoch 67/500
227/227 [============ ] - Os 2ms/step - loss: 0.8379 -
f1 score: 0.6575 - accuracy: 0.6575 - val loss: 0.8502 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 68/500
f1_score: 0.6586 - accuracy: 0.6586 - val_loss: 0.8426 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 69/500
227/227 [============ ] - Os 2ms/step - loss: 0.8346 -
f1_score: 0.6580 - accuracy: 0.6580 - val_loss: 0.8413 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 70/500
f1_score: 0.6569 - accuracy: 0.6569 - val_loss: 0.8496 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 71/500
f1_score: 0.6583 - accuracy: 0.6583 - val_loss: 0.8490 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 72/500
f1_score: 0.6582 - accuracy: 0.6582 - val_loss: 0.8475 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 73/500
227/227 [============ ] - 1s 2ms/step - loss: 0.8340 -
f1 score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8409 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 74/500
f1_score: 0.6587 - accuracy: 0.6587 - val_loss: 0.8433 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 75/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8456 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 76/500
f1_score: 0.6578 - accuracy: 0.6578 - val_loss: 0.8475 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 77/500
227/227 [============ ] - 1s 2ms/step - loss: 0.8380 -
f1_score: 0.6565 - accuracy: 0.6565 - val_loss: 0.8411 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 78/500
f1_score: 0.6582 - accuracy: 0.6582 - val_loss: 0.8425 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 79/500
f1 score: 0.6596 - accuracy: 0.6596 - val loss: 0.8399 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 80/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8399 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 81/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8324 -
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8448 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 82/500
f1_score: 0.6576 - accuracy: 0.6576 - val_loss: 0.8456 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 83/500
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8439 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 84/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8413 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 85/500
227/227 [============ ] - Os 2ms/step - loss: 0.8318 -
f1 score: 0.6578 - accuracy: 0.6578 - val_loss: 0.8461 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 86/500
f1_score: 0.6580 - accuracy: 0.6580 - val_loss: 0.8416 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 87/500
f1_score: 0.6569 - accuracy: 0.6569 - val_loss: 0.8424 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 88/500
f1_score: 0.6579 - accuracy: 0.6579 - val_loss: 0.8388 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 89/500
227/227 [============ ] - Os 2ms/step - loss: 0.8317 -
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8456 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 90/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8429 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 91/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8335 -
f1 score: 0.6579 - accuracy: 0.6579 - val loss: 0.8454 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 92/500
f1_score: 0.6589 - accuracy: 0.6589 - val_loss: 0.8461 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 93/500
227/227 [============ ] - Os 2ms/step - loss: 0.8301 -
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8406 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 94/500
f1_score: 0.6572 - accuracy: 0.6572 - val_loss: 0.8372 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 95/500
f1_score: 0.6579 - accuracy: 0.6579 - val_loss: 0.8410 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 96/500
f1_score: 0.6582 - accuracy: 0.6582 - val_loss: 0.8463 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 97/500
227/227 [============ ] - Os 2ms/step - loss: 0.8306 -
f1 score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8410 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 98/500
f1_score: 0.6583 - accuracy: 0.6583 - val_loss: 0.8486 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 99/500
f1_score: 0.6578 - accuracy: 0.6578 - val_loss: 0.8519 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 100/500
f1_score: 0.6569 - accuracy: 0.6569 - val_loss: 0.8450 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 101/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8341 -
f1_score: 0.6590 - accuracy: 0.6590 - val_loss: 0.8360 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 102/500
f1_score: 0.6561 - accuracy: 0.6561 - val_loss: 0.8439 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 103/500
227/227 [============ ] - Os 2ms/step - loss: 0.8281 -
f1 score: 0.6601 - accuracy: 0.6601 - val loss: 0.8464 - val f1 score: 0.6543 -
val_accuracy: 0.6543
Epoch 104/500
f1_score: 0.6582 - accuracy: 0.6582 - val_loss: 0.8433 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 105/500
227/227 [============ ] - 1s 2ms/step - loss: 0.8292 -
f1_score: 0.6586 - accuracy: 0.6586 - val_loss: 0.8368 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 106/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8462 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
```

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Epoch 107/500
f1_score: 0.6580 - accuracy: 0.6580 - val_loss: 0.8410 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 108/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8437 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 109/500
227/227 [============ ] - Os 2ms/step - loss: 0.8300 -
f1 score: 0.6582 - accuracy: 0.6582 - val_loss: 0.8456 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 110/500
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8388 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 111/500
f1_score: 0.6587 - accuracy: 0.6587 - val_loss: 0.8420 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 112/500
f1_score: 0.6594 - accuracy: 0.6594 - val_loss: 0.8453 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 113/500
227/227 [============ ] - Os 2ms/step - loss: 0.8305 -
f1_score: 0.6573 - accuracy: 0.6573 - val_loss: 0.8440 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 114/500
f1_score: 0.6598 - accuracy: 0.6598 - val_loss: 0.8396 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 115/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8288 -
f1 score: 0.6586 - accuracy: 0.6586 - val loss: 0.8374 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 116/500
f1_score: 0.6593 - accuracy: 0.6593 - val_loss: 0.8444 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 117/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8310 -
f1_score: 0.6593 - accuracy: 0.6593 - val_loss: 0.8464 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 118/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8412 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 119/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8461 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 120/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8405 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 121/500
227/227 [============ ] - Os 2ms/step - loss: 0.8307 -
f1 score: 0.6594 - accuracy: 0.6594 - val_loss: 0.8456 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 122/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8447 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 123/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8433 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 124/500
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8366 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 125/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8291 -
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8434 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 126/500
f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8345 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 127/500
227/227 [============ ] - Os 2ms/step - loss: 0.8278 -
f1 score: 0.6584 - accuracy: 0.6584 - val loss: 0.8369 - val f1 score: 0.6543 -
val_accuracy: 0.6543
Epoch 128/500
f1_score: 0.6587 - accuracy: 0.6587 - val_loss: 0.8368 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 129/500
f1_score: 0.6601 - accuracy: 0.6601 - val_loss: 0.8393 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 130/500
f1_score: 0.6593 - accuracy: 0.6593 - val_loss: 0.8386 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 131/500
f1_score: 0.6589 - accuracy: 0.6589 - val_loss: 0.8404 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 132/500
f1_score: 0.6604 - accuracy: 0.6604 - val_loss: 0.8310 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 133/500
227/227 [============ ] - Os 2ms/step - loss: 0.8301 -
f1 score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8369 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 134/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8377 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 135/500
f1_score: 0.6584 - accuracy: 0.6584 - val_loss: 0.8452 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 136/500
f1_score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8430 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 137/500
227/227 [============ ] - Os 2ms/step - loss: 0.8277 -
f1_score: 0.6590 - accuracy: 0.6590 - val_loss: 0.8391 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 138/500
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8450 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 139/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8504 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 140/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8457 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 141/500
227/227 [========== ] - Os 2ms/step - loss: 0.8254 -
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8464 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 142/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8487 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 143/500
f1_score: 0.6583 - accuracy: 0.6583 - val_loss: 0.8407 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 144/500
f1_score: 0.6593 - accuracy: 0.6593 - val_loss: 0.8459 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 145/500
227/227 [============ ] - Os 2ms/step - loss: 0.8290 -
f1 score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8435 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 146/500
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8424 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 147/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8431 - val_f1_score: 0.6506 -
val accuracy: 0.6506
Epoch 148/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8528 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 149/500
227/227 [============ ] - Os 2ms/step - loss: 0.8282 -
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8510 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 150/500
f1_score: 0.6580 - accuracy: 0.6580 - val_loss: 0.8409 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 151/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8238 -
f1 score: 0.6601 - accuracy: 0.6601 - val loss: 0.8417 - val f1 score: 0.6543 -
val_accuracy: 0.6543
Epoch 152/500
f1_score: 0.6604 - accuracy: 0.6604 - val_loss: 0.8422 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 153/500
227/227 [============ ] - Os 2ms/step - loss: 0.8276 -
f1_score: 0.6584 - accuracy: 0.6584 - val_loss: 0.8449 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 154/500
f1_score: 0.6586 - accuracy: 0.6586 - val_loss: 0.8443 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
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Epoch 155/500
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8423 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 156/500
f1_score: 0.6598 - accuracy: 0.6598 - val_loss: 0.8406 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 157/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8230 -
f1 score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8394 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 158/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8427 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 159/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8485 - val_f1_score: 0.6506 -
val accuracy: 0.6506
Epoch 160/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8462 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 161/500
227/227 [============= ] - 1s 2ms/step - loss: 0.8252 -
f1_score: 0.6601 - accuracy: 0.6601 - val_loss: 0.8434 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 162/500
f1_score: 0.6587 - accuracy: 0.6587 - val_loss: 0.8410 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 163/500
227/227 [============ ] - Os 2ms/step - loss: 0.8249 -
f1 score: 0.6596 - accuracy: 0.6596 - val loss: 0.8401 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 164/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8415 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 165/500
227/227 [============ ] - Os 2ms/step - loss: 0.8261 -
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8334 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 166/500
f1_score: 0.6576 - accuracy: 0.6576 - val_loss: 0.8451 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 167/500
f1_score: 0.6589 - accuracy: 0.6589 - val_loss: 0.8384 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 168/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8400 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 169/500
227/227 [============ ] - Os 2ms/step - loss: 0.8236 -
f1 score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8377 - val_f1_score: 0.6617 -
val_accuracy: 0.6617
Epoch 170/500
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8390 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 171/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8428 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 172/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8385 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 173/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8265 -
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8411 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 174/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8398 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 175/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8237 -
f1 score: 0.6596 - accuracy: 0.6596 - val loss: 0.8433 - val f1 score: 0.6493 -
val_accuracy: 0.6493
Epoch 176/500
f1_score: 0.6590 - accuracy: 0.6590 - val_loss: 0.8363 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 177/500
227/227 [============ ] - 1s 2ms/step - loss: 0.8225 -
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8409 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 178/500
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8365 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
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Epoch 179/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8348 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 180/500
f1_score: 0.6624 - accuracy: 0.6624 - val_loss: 0.8369 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 181/500
227/227 [============ ] - Os 2ms/step - loss: 0.8227 -
f1 score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8388 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 182/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8353 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 183/500
f1_score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8356 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 184/500
f1_score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8447 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 185/500
227/227 [=========== ] - Os 1ms/step - loss: 0.8194 -
f1_score: 0.6651 - accuracy: 0.6651 - val_loss: 0.8373 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 186/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8359 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 187/500
227/227 [============ ] - Os 2ms/step - loss: 0.8217 -
f1 score: 0.6605 - accuracy: 0.6605 - val loss: 0.8407 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 188/500
f1_score: 0.6594 - accuracy: 0.6594 - val_loss: 0.8404 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 189/500
227/227 [============ ] - Os 2ms/step - loss: 0.8255 -
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8367 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 190/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8364 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
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Epoch 191/500
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8369 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 192/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8336 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 193/500
227/227 [============ ] - Os 2ms/step - loss: 0.8214 -
f1 score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8382 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 194/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8371 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 195/500
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8436 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 196/500
f1_score: 0.6601 - accuracy: 0.6601 - val_loss: 0.8477 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 197/500
f1_score: 0.6568 - accuracy: 0.6568 - val_loss: 0.8466 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 198/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8413 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 199/500
227/227 [============ ] - Os 2ms/step - loss: 0.8218 -
f1 score: 0.6586 - accuracy: 0.6586 - val loss: 0.8466 - val f1 score: 0.6493 -
val_accuracy: 0.6493
Epoch 200/500
f1_score: 0.6634 - accuracy: 0.6634 - val_loss: 0.8462 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 201/500
227/227 [========== ] - Os 2ms/step - loss: 0.8236 -
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8418 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 202/500
f1_score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8461 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 203/500
f1_score: 0.6608 - accuracy: 0.6608 - val_loss: 0.8433 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 204/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8441 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 205/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8216 -
f1 score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8450 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 206/500
f1_score: 0.6601 - accuracy: 0.6601 - val_loss: 0.8418 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 207/500
f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8403 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 208/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8375 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 209/500
227/227 [============ ] - Os 2ms/step - loss: 0.8193 -
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8432 - val_f1_score: 0.6580 -
val_accuracy: 0.6580
Epoch 210/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8472 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 211/500
f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8380 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 212/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8441 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 213/500
227/227 [========== ] - Os 2ms/step - loss: 0.8205 -
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8407 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 214/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8401 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 215/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8448 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 216/500
f1_score: 0.6602 - accuracy: 0.6602 - val_loss: 0.8406 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 217/500
f1 score: 0.6608 - accuracy: 0.6608 - val_loss: 0.8470 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 218/500
f1_score: 0.6598 - accuracy: 0.6598 - val_loss: 0.8409 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 219/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8421 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 220/500
f1_score: 0.6598 - accuracy: 0.6598 - val_loss: 0.8444 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 221/500
227/227 [============ ] - Os 2ms/step - loss: 0.8250 -
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8465 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 222/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8422 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 223/500
f1 score: 0.6601 - accuracy: 0.6601 - val loss: 0.8406 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 224/500
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8421 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 225/500
227/227 [============ ] - Os 2ms/step - loss: 0.8216 -
f1_score: 0.6604 - accuracy: 0.6604 - val_loss: 0.8421 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 226/500
f1_score: 0.6600 - accuracy: 0.6600 - val_loss: 0.8392 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 227/500
f1_score: 0.6586 - accuracy: 0.6586 - val_loss: 0.8432 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 228/500
f1_score: 0.6629 - accuracy: 0.6629 - val_loss: 0.8474 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 229/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8202 -
f1 score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8454 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 230/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8432 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 231/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8428 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 232/500
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8432 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 233/500
227/227 [============ ] - Os 2ms/step - loss: 0.8156 -
f1_score: 0.6629 - accuracy: 0.6629 - val_loss: 0.8474 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 234/500
f1_score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8395 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 235/500
227/227 [============ ] - Os 2ms/step - loss: 0.8208 -
f1 score: 0.6593 - accuracy: 0.6593 - val loss: 0.8389 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 236/500
f1_score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8443 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 237/500
227/227 [========== ] - Os 2ms/step - loss: 0.8200 -
f1_score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8464 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 238/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8407 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 239/500
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8381 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 240/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8391 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 241/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8239 -
f1 score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8449 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 242/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8451 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 243/500
f1_score: 0.6631 - accuracy: 0.6631 - val_loss: 0.8406 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 244/500
f1_score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8484 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 245/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8238 -
f1_score: 0.6629 - accuracy: 0.6629 - val_loss: 0.8413 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 246/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8476 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 247/500
f1 score: 0.6626 - accuracy: 0.6626 - val loss: 0.8395 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 248/500
f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8450 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 249/500
227/227 [========== ] - Os 2ms/step - loss: 0.8190 -
f1_score: 0.6637 - accuracy: 0.6637 - val_loss: 0.8421 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 250/500
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8374 - val_f1_score: 0.6580 -
val_accuracy: 0.6580
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Epoch 251/500
f1_score: 0.6583 - accuracy: 0.6583 - val_loss: 0.8421 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 252/500
f1_score: 0.6590 - accuracy: 0.6590 - val_loss: 0.8440 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 253/500
227/227 [============ ] - Os 2ms/step - loss: 0.8198 -
f1 score: 0.6608 - accuracy: 0.6608 - val_loss: 0.8415 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 254/500
f1_score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8446 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 255/500
f1_score: 0.6596 - accuracy: 0.6596 - val_loss: 0.8398 - val_f1_score: 0.6530 -
val accuracy: 0.6530
Epoch 256/500
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8415 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 257/500
227/227 [============ ] - Os 2ms/step - loss: 0.8193 -
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8460 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 258/500
f1_score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8430 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 259/500
f1 score: 0.6586 - accuracy: 0.6586 - val loss: 0.8382 - val f1 score: 0.6506 -
val_accuracy: 0.6506
Epoch 260/500
f1_score: 0.6598 - accuracy: 0.6598 - val_loss: 0.8329 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 261/500
227/227 [========== ] - Os 2ms/step - loss: 0.8203 -
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8421 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 262/500
f1_score: 0.6638 - accuracy: 0.6638 - val_loss: 0.8370 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
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Epoch 263/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8377 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 264/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8398 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 265/500
f1 score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8433 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 266/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8365 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 267/500
f1_score: 0.6591 - accuracy: 0.6591 - val_loss: 0.8365 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 268/500
f1_score: 0.6609 - accuracy: 0.6609 - val_loss: 0.8363 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 269/500
227/227 [============ ] - Os 2ms/step - loss: 0.8237 -
f1_score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8370 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 270/500
f1_score: 0.6642 - accuracy: 0.6642 - val_loss: 0.8429 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 271/500
227/227 [============ ] - Os 2ms/step - loss: 0.8211 -
f1 score: 0.6624 - accuracy: 0.6624 - val loss: 0.8450 - val f1 score: 0.6530 -
val_accuracy: 0.6530
Epoch 272/500
f1_score: 0.6622 - accuracy: 0.6622 - val_loss: 0.8437 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 273/500
227/227 [========== ] - Os 2ms/step - loss: 0.8221 -
f1_score: 0.6633 - accuracy: 0.6633 - val_loss: 0.8394 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 274/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8441 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
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Epoch 275/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8397 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 276/500
f1_score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8360 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 277/500
227/227 [============ ] - Os 2ms/step - loss: 0.8206 -
f1 score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8382 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 278/500
f1_score: 0.6637 - accuracy: 0.6637 - val_loss: 0.8356 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 279/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8469 - val_f1_score: 0.6555 -
val accuracy: 0.6555
Epoch 280/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8461 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 281/500
227/227 [============ ] - Os 2ms/step - loss: 0.8176 -
f1_score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8386 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 282/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8376 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 283/500
227/227 [============ ] - Os 2ms/step - loss: 0.8207 -
f1 score: 0.6594 - accuracy: 0.6594 - val loss: 0.8358 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 284/500
f1_score: 0.6622 - accuracy: 0.6622 - val_loss: 0.8402 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 285/500
227/227 [========== ] - Os 2ms/step - loss: 0.8194 -
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8383 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 286/500
f1_score: 0.6602 - accuracy: 0.6602 - val_loss: 0.8363 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
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Epoch 287/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8447 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 288/500
f1_score: 0.6633 - accuracy: 0.6633 - val_loss: 0.8495 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 289/500
227/227 [============ ] - Os 2ms/step - loss: 0.8207 -
f1_score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8396 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 290/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8391 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 291/500
f1_score: 0.6604 - accuracy: 0.6604 - val_loss: 0.8362 - val_f1_score: 0.6543 -
val accuracy: 0.6543
Epoch 292/500
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8353 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 293/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8224 -
f1_score: 0.6579 - accuracy: 0.6579 - val_loss: 0.8374 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 294/500
f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8430 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 295/500
f1_score: 0.6618 - accuracy: 0.6618 - val_loss: 0.8374 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 296/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8353 - val_f1_score: 0.6493 -
val_accuracy: 0.6493
Epoch 297/500
227/227 [============ ] - Os 2ms/step - loss: 0.8193 -
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8371 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 298/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8340 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
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Epoch 299/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8379 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 300/500
f1_score: 0.6597 - accuracy: 0.6597 - val_loss: 0.8369 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 301/500
227/227 [============ ] - Os 2ms/step - loss: 0.8183 -
f1 score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8328 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 302/500
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8392 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 303/500
f1_score: 0.6629 - accuracy: 0.6629 - val_loss: 0.8395 - val_f1_score: 0.6530 -
val accuracy: 0.6530
Epoch 304/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8379 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 305/500
227/227 [=========== ] - Os 2ms/step - loss: 0.8229 -
f1_score: 0.6623 - accuracy: 0.6623 - val_loss: 0.8435 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 306/500
f1_score: 0.6624 - accuracy: 0.6624 - val_loss: 0.8404 - val_f1_score: 0.6568 -
val_accuracy: 0.6568
Epoch 307/500
f1_score: 0.6615 - accuracy: 0.6615 - val_loss: 0.8392 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 308/500
f1_score: 0.6604 - accuracy: 0.6604 - val_loss: 0.8397 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 309/500
227/227 [============ ] - Os 2ms/step - loss: 0.8178 -
f1_score: 0.6611 - accuracy: 0.6611 - val_loss: 0.8447 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 310/500
f1_score: 0.6622 - accuracy: 0.6622 - val_loss: 0.8320 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
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Epoch 311/500
f1_score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8464 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 312/500
f1_score: 0.6626 - accuracy: 0.6626 - val_loss: 0.8366 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 313/500
227/227 [============ ] - Os 2ms/step - loss: 0.8151 -
f1 score: 0.6631 - accuracy: 0.6631 - val_loss: 0.8438 - val_f1_score: 0.6518 -
val_accuracy: 0.6518
Epoch 314/500
f1_score: 0.6620 - accuracy: 0.6620 - val_loss: 0.8385 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 315/500
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8398 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 316/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8348 - val_f1_score: 0.6506 -
val_accuracy: 0.6506
Epoch 317/500
227/227 [============ ] - Os 2ms/step - loss: 0.8169 -
f1_score: 0.6612 - accuracy: 0.6612 - val_loss: 0.8438 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
Epoch 318/500
f1_score: 0.6613 - accuracy: 0.6613 - val_loss: 0.8326 - val_f1_score: 0.6555 -
val_accuracy: 0.6555
Epoch 319/500
227/227 [============ ] - Os 2ms/step - loss: 0.8210 -
f1 score: 0.6608 - accuracy: 0.6608 - val loss: 0.8346 - val f1 score: 0.6518 -
val_accuracy: 0.6518
Epoch 320/500
f1_score: 0.6605 - accuracy: 0.6605 - val_loss: 0.8397 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 321/500
227/227 [========== ] - Os 2ms/step - loss: 0.8193 -
f1_score: 0.6629 - accuracy: 0.6629 - val_loss: 0.8466 - val_f1_score: 0.6530 -
val_accuracy: 0.6530
Epoch 322/500
f1_score: 0.6616 - accuracy: 0.6616 - val_loss: 0.8385 - val_f1_score: 0.6543 -
val_accuracy: 0.6543
```

```
f1_score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8405 - val_f1_score: 0.6530 -
    val_accuracy: 0.6530
    Epoch 324/500
    f1_score: 0.6594 - accuracy: 0.6594 - val_loss: 0.8414 - val_f1_score: 0.6543 -
    val_accuracy: 0.6543
    Epoch 325/500
    227/227 [============ ] - Os 2ms/step - loss: 0.8151 -
    f1 score: 0.6627 - accuracy: 0.6627 - val_loss: 0.8438 - val_f1_score: 0.6506 -
    val_accuracy: 0.6506
    Epoch 326/500
    f1_score: 0.6648 - accuracy: 0.6648 - val_loss: 0.8458 - val_f1_score: 0.6493 -
    val_accuracy: 0.6493
    Epoch 327/500
    f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8417 - val_f1_score: 0.6506 -
    val accuracy: 0.6506
    Epoch 328/500
    f1_score: 0.6607 - accuracy: 0.6607 - val_loss: 0.8396 - val_f1_score: 0.6530 -
    val_accuracy: 0.6530
    Epoch 329/500
    227/227 [============ ] - Os 2ms/step - loss: 0.8179 -
    f1_score: 0.6622 - accuracy: 0.6622 - val_loss: 0.8435 - val_f1_score: 0.6518 -
    val_accuracy: 0.6518
    Epoch 330/500
    f1_score: 0.6619 - accuracy: 0.6619 - val_loss: 0.8513 - val_f1_score: 0.6518 -
    val_accuracy: 0.6518
    Epoch 331/500
    227/227 [============ ] - Os 2ms/step - loss: 0.8273 -
    f1 score: 0.6598 - accuracy: 0.6598 - val loss: 0.8419 - val f1 score: 0.6543 -
    val_accuracy: 0.6543
    Epoch 332/500
    f1_score: 0.6624 - accuracy: 0.6624 - val_loss: 0.8350 - val_f1_score: 0.6530 -
    val_accuracy: 0.6530
[249]: def test_pipeline(model):
        '''performs all the necessary steps for the testing
        dataset
        df_test = load_csv('archive/test.csv')
        df_test = pre_processing(df_test, train=False)
```

Epoch 323/500

```
y_pred = model.predict(df_test.values)
          y_test = load_csv('archive/test.csv')[['Var_1']]
          y_test['Var_1'] = y_test['Var_1'].apply(replacing_classes)
          y_test = pd.get_dummies(y_test)
           # putting predictions in a list
           # as the index of max probability
          pred = list()
          for i in range(len(y_pred)):
              pred.append(np.argmax(y_pred[i]))
           # putting test target variables in a list
          test = list()
          y_test_arr = y_test.values
          for i in range(len(y_test_arr)):
               test.append(np.argmax(y_test_arr[i]))
          def num_to_class(x):
               if x == 0:
                  return "Var_1_Cat_4"
               if x == 1:
                  return "Var_1_Cat_6"
               if x == 2:
                  return "Var_1_Other"
          return (list(map(lambda x: num to class(x), pred)),
                   list(map(lambda x: num_to_class(x), test)),
                  y_pred,
                   y_test,
                  df_test)
[250]: pred, test, y_pred, y_test, X_test = test_pipeline(model)
      dropping ID column
      dropping ['ID', 'Var_1', 'Age']...
      replacing numerical nans with mode...
      replacing categorical nans with None string...
      label encoding categorical data...
      SUCCESSFULLY PERFORMED PREPROCESSING
      83/83 [========= ] - 0s 547us/step
[251]: X test
[251]:
            Gender Ever_Married Graduated Profession Work_Experience \
      0
                 0
                                2
                                          2
                                                       2
                                                                      0.0
      1
                 1
                                2
                                          2
                                                       5
                                                                      8.0
      2
                 0
                                2
                                          0
                                                       9
                                                                      0.0
      3
                 1
                                2
                                           0
                                                       4
                                                                     11.0
      4
                 0
                                0
                                           0
                                                       8
                                                                      1.0
```

•••	•••		•••	•••	
2622	1	0	0	5	9.0
2623	0	0	2	1	1.0
2624	0	0	2	3	1.0
2625	1	2	2	4	1.0
2626	0	0	2	5	9.0

	Spending_Score	Family_Size
0	2	1.0
1	0	4.0
2	2	1.0
3	1	2.0
4	2	4.0
•••	•••	•••
2622	2	4.0
2623	2	1.0
2624	2	2.0
2625	1	5.0
2626	2	3.0

[2627 rows x 7 columns]

```
f1_score for the test data is: 0.2849736839965477 recall for the test data is: 0.5258282267330525 precision score for the test data is: 0.34608733916751366 accuracy for the test data is: 0.6417967263037686
```

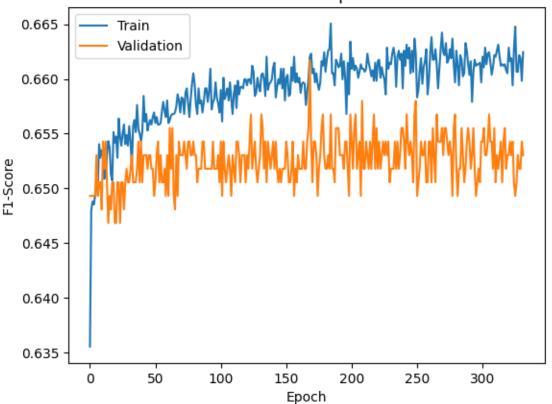
recall: recall is the true positives over false negatives and true positives. In multi class the true positives mean all the value for the class A that was correctly identified. False negative means all the predictions that belonged to class A but incorrectly classified as B or C. the recall would be average recall of the recalls for all three classes.

precision: precision is the true positives over true positives and false positives. false positive are the values that belong to class B or C but was predicted incorrectly as A. precision would be the average of the precision for the three classes.

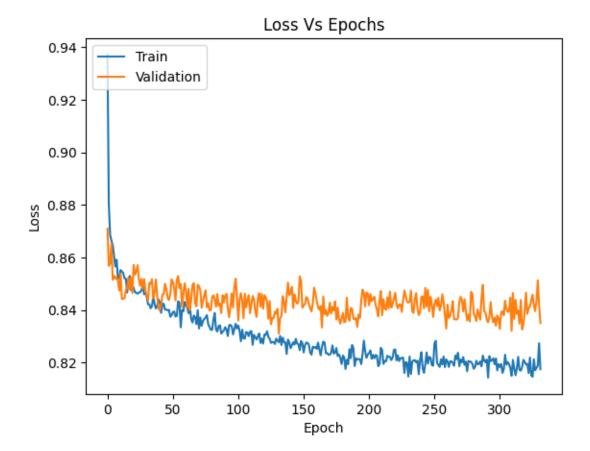
fl_score: it is the harmonic mean that punishes the extreme values. it means that if recall is high then recall would be punished and the harmonic mean would be more in the side of precision. It is a good metric for imbalance data. it is 2 times average precision time average recall over average recall and average precision.

```
[253]: plt.plot(history.history['f1_score'])
   plt.plot(history.history['val_f1_score'])
   plt.title('F1-Score VS Epochs')
   plt.ylabel('F1-Score')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Validation'], loc='upper left')
   plt.show()
```

F1-Score VS Epochs



```
[254]: plt.plot(history.history['loss'])
   plt.plot(history.history['val_loss'])
   plt.title('Loss Vs Epochs')
   plt.ylabel('Loss')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Validation'], loc='upper left')
   plt.show()
```

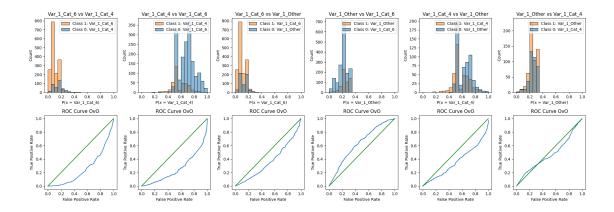


Model is definitely learning in training set. you can see that loss is decreasing and f1_score is increasing. However Because of small number of data (although it was shuffled and stratified) model doesn't reach its best form. The early stopage was applied and the model stopped after 332 epoch of 500 total epochs.

```
# Gets the class
       comb = classes_combinations[i]
      c1 = comb[0]
      c2 = comb[1]
      c1_index = class_list.index(c1)
      title = c1 + "vs" + c2
       # Prepares an auxiliar dataframe to help with the plots
      df aux = df test.copy()
      df_aux['class'] = test
      df_aux['prob'] = y_pred[:, c1_index]
       # Slices only the subset with both classes
      df_{aux} = df_{aux}[(df_{aux}['class'] == c1) | (df_{aux}['class'] == c2)]
      df_aux['class'] = [1 if y == c1 else 0 for y in df_aux['class']]
      df_aux = df_aux.reset_index(drop = True)
       # Plots the probability distribution for the class and the rest
      ax = plt.subplot(2, 6, i+1)
      sns.histplot(x = "prob", data = df_aux, hue = 'class', color = 'b', ax_{\sqcup}
\Rightarrow= ax, bins = bins)
      ax.set title(title)
      ax.legend([f"Class 1: \{c1\}", f"Class 0: \{c2\}"])
      ax.set_xlabel(f"P(x = {c1})")
       # Calculates the ROC Coordinates and plots the ROC Curves
      ax_bottom = plt.subplot(2, 6, i+7)
      tpr, fpr = get_all_roc_coordinates(df_aux['class'], df_aux['prob'])
      plot_roc_curve(tpr, fpr, scatter = False, ax = ax_bottom)
      ax_bottom.set_title("ROC Curve OvO")
       # Calculates the ROC AUC OvO
      roc_auc_ovo[title] = roc_auc_score(df_aux['class'], df_aux['prob'])
  plt.tight_layout()
```

```
[256]: plot_roc_auc(test, y_pred)
```

```
[['Var_1_Cat_6', 'Var_1_Cat_4'], ['Var_1_Cat_4', 'Var_1_Cat_6'], ['Var_1_Cat_6',
'Var_1_Other'], ['Var_1_Other', 'Var_1_Cat_6'], ['Var_1_Cat_4', 'Var_1_Other'],
['Var_1_Other', 'Var_1_Cat_4']]
```



ROC curves are not very promising in this mode. You can see that for most of the classes, ROC curve is below the 0.5 probability which means that model is performing worse than a coin toss. This is most probably because of a small number of data that we have. For neural network, minimum of a few million data-sets are required. However, if we look at 4th ROC graph, we might say that a False Positive of around 0.5 and True Positive of around 0.6 are good thresholds since it maximises the area under the curve.

0.0.5 After Applying SMOTE for Over Sampling

```
[257]: def smote_train_pipeline():
           performs all the necessary functions for the training
           dataset
           111
           df = load_csv('archive/train.csv')
           df = pre processing(df, train=True, only label=True)
           df smote = apply smote(df)
           print("Var_1_Cat_4 after SMOTE", len(df_smote.loc[df_smote["Var_1"] ==_

¬"Cat 4"]))
           print("Var_1_Cat_6 after SMOTE", len(df_smote.loc[df_smote["Var_1"] ==__

¬"Cat_6"]))
           print("Other after SMOTE", len(df_smote.loc[df_smote["Var_1"] == "Other"]))
           target = df_smote[['Var_1']]
           target = pd.get dummies(target)
           df_smote.drop(columns=['Var_1'], inplace=True)
           df smote = pd.concat([df smote, target], axis=1)
           X_train, X_valid, y_train, y_valid = train_validation_split(0.10, df_smote)
           # Removing normalization: Because most of the columns
           # Are categorical and if we normalize the effect of
           # Numerical data will be very low
           # min_max_scaler(X_train)
           # min_max_scaler(X_valid)
           history_smote, model smote = fit model(X_train, y_train, X_valid, y_valid)
```

return history, model_smote

We applied SMOTE to increase the number minority classes. As you can see below all three classes have the same number of data points.

[258]: history_smote, model_smote = smote_train_pipeline()

dropping columns...

dropping ['ID', 'Segmentation', 'Age']...

replacing numerical nans with mode...

replacing categorical nans with None string...

changing anything otehr than cat_6 and cat_4 to other...

label encoding categorical data...

SUCCESSFULLY PERFORMED PREPROCESSING

Var_1_Cat_4 after SMOTE 5238

Var_1_Cat_6 after SMOTE 5238

Other after SMOTE 5238 Model: "sequential_26"

Layer (type)	Output Shape	Param #
dense_182 (Dense)	(None, 512)	4096
dense_183 (Dense)	(None, 256)	131328
dense_184 (Dense)	(None, 128)	32896
dropout_78 (Dropout)	(None, 128)	0
dense_185 (Dense)	(None, 64)	8256
dropout_79 (Dropout)	(None, 64)	0
dense_186 (Dense)	(None, 16)	1040
<pre>dropout_80 (Dropout)</pre>	(None, 16)	0
dense_187 (Dense)	(None, 8)	136
dense_188 (Dense)	(None, 3)	27

Total params: 177,779 Trainable params: 177,779 Non-trainable params: 0

None

Epoch 1/500

```
f1_score: 0.3837 - accuracy: 0.3837 - val_loss: 1.0688 - val_f1_score: 0.4078 -
val_accuracy: 0.4078
Epoch 2/500
f1_score: 0.4190 - accuracy: 0.4190 - val_loss: 1.0668 - val_f1_score: 0.4281 -
val accuracy: 0.4281
Epoch 3/500
f1_score: 0.4137 - accuracy: 0.4137 - val_loss: 1.0596 - val_f1_score: 0.4256 -
val_accuracy: 0.4256
Epoch 4/500
f1 score: 0.4314 - accuracy: 0.4314 - val_loss: 1.0549 - val_f1_score: 0.4567 -
val_accuracy: 0.4567
Epoch 5/500
f1 score: 0.4330 - accuracy: 0.4330 - val_loss: 1.0518 - val_f1_score: 0.4230 -
val_accuracy: 0.4230
Epoch 6/500
f1_score: 0.4314 - accuracy: 0.4314 - val_loss: 1.0571 - val_f1_score: 0.4358 -
val_accuracy: 0.4358
Epoch 7/500
f1 score: 0.4332 - accuracy: 0.4332 - val_loss: 1.0554 - val_f1_score: 0.4440 -
val_accuracy: 0.4440
Epoch 8/500
f1_score: 0.4447 - accuracy: 0.4447 - val_loss: 1.0579 - val_f1_score: 0.4307 -
val_accuracy: 0.4307
Epoch 9/500
f1_score: 0.4371 - accuracy: 0.4371 - val_loss: 1.0565 - val_f1_score: 0.4345 -
val accuracy: 0.4345
Epoch 10/500
f1_score: 0.4380 - accuracy: 0.4380 - val_loss: 1.0544 - val_f1_score: 0.4548 -
val_accuracy: 0.4548
Epoch 11/500
f1 score: 0.4439 - accuracy: 0.4439 - val_loss: 1.0505 - val_f1_score: 0.4459 -
val_accuracy: 0.4459
Epoch 12/500
f1_score: 0.4439 - accuracy: 0.4439 - val_loss: 1.0500 - val_f1_score: 0.4580 -
val_accuracy: 0.4580
Epoch 13/500
```

```
f1_score: 0.4446 - accuracy: 0.4446 - val_loss: 1.0518 - val_f1_score: 0.4523 -
val_accuracy: 0.4523
Epoch 14/500
f1_score: 0.4448 - accuracy: 0.4448 - val_loss: 1.0512 - val_f1_score: 0.4504 -
val accuracy: 0.4504
Epoch 15/500
f1_score: 0.4510 - accuracy: 0.4510 - val_loss: 1.0443 - val_f1_score: 0.4377 -
val_accuracy: 0.4377
Epoch 16/500
f1 score: 0.4474 - accuracy: 0.4474 - val_loss: 1.0324 - val_f1_score: 0.4555 -
val_accuracy: 0.4555
Epoch 17/500
f1 score: 0.4464 - accuracy: 0.4464 - val_loss: 1.0382 - val_f1_score: 0.4739 -
val_accuracy: 0.4739
Epoch 18/500
f1_score: 0.4564 - accuracy: 0.4564 - val_loss: 1.0299 - val_f1_score: 0.4447 -
val_accuracy: 0.4447
Epoch 19/500
f1 score: 0.4579 - accuracy: 0.4579 - val_loss: 1.0451 - val_f1_score: 0.4587 -
val_accuracy: 0.4587
Epoch 20/500
f1_score: 0.4642 - accuracy: 0.4642 - val_loss: 1.0313 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 21/500
442/442 [=========== ] - 1s 2ms/step - loss: 1.0309 -
f1_score: 0.4570 - accuracy: 0.4570 - val_loss: 1.0377 - val_f1_score: 0.4733 -
val accuracy: 0.4733
Epoch 22/500
f1_score: 0.4598 - accuracy: 0.4598 - val_loss: 1.0318 - val_f1_score: 0.4542 -
val_accuracy: 0.4542
Epoch 23/500
f1 score: 0.4623 - accuracy: 0.4623 - val_loss: 1.0361 - val_f1_score: 0.4497 -
val_accuracy: 0.4497
Epoch 24/500
f1_score: 0.4588 - accuracy: 0.4588 - val_loss: 1.0291 - val_f1_score: 0.4618 -
val_accuracy: 0.4618
Epoch 25/500
```

```
f1_score: 0.4633 - accuracy: 0.4633 - val_loss: 1.0391 - val_f1_score: 0.4389 -
val_accuracy: 0.4389
Epoch 26/500
f1_score: 0.4598 - accuracy: 0.4598 - val_loss: 1.0263 - val_f1_score: 0.4644 -
val accuracy: 0.4644
Epoch 27/500
f1_score: 0.4670 - accuracy: 0.4670 - val_loss: 1.0448 - val_f1_score: 0.4758 -
val_accuracy: 0.4758
Epoch 28/500
f1 score: 0.4658 - accuracy: 0.4658 - val_loss: 1.0354 - val_f1_score: 0.4587 -
val_accuracy: 0.4587
Epoch 29/500
f1 score: 0.4663 - accuracy: 0.4663 - val_loss: 1.0401 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 30/500
f1_score: 0.4682 - accuracy: 0.4682 - val_loss: 1.0472 - val_f1_score: 0.4612 -
val_accuracy: 0.4612
Epoch 31/500
f1 score: 0.4700 - accuracy: 0.4700 - val_loss: 1.0282 - val_f1_score: 0.4663 -
val_accuracy: 0.4663
Epoch 32/500
f1_score: 0.4702 - accuracy: 0.4702 - val_loss: 1.0237 - val_f1_score: 0.4714 -
val_accuracy: 0.4714
Epoch 33/500
f1_score: 0.4617 - accuracy: 0.4617 - val_loss: 1.0244 - val_f1_score: 0.4555 -
val accuracy: 0.4555
Epoch 34/500
f1_score: 0.4715 - accuracy: 0.4715 - val_loss: 1.0370 - val_f1_score: 0.4383 -
val_accuracy: 0.4383
Epoch 35/500
f1 score: 0.4691 - accuracy: 0.4691 - val_loss: 1.0203 - val_f1_score: 0.4637 -
val_accuracy: 0.4637
Epoch 36/500
f1_score: 0.4669 - accuracy: 0.4669 - val_loss: 1.0275 - val_f1_score: 0.4612 -
val_accuracy: 0.4612
Epoch 37/500
```

```
f1_score: 0.4685 - accuracy: 0.4685 - val_loss: 1.0271 - val_f1_score: 0.4555 -
val_accuracy: 0.4555
Epoch 38/500
f1_score: 0.4674 - accuracy: 0.4674 - val_loss: 1.0248 - val_f1_score: 0.4542 -
val accuracy: 0.4542
Epoch 39/500
f1_score: 0.4629 - accuracy: 0.4629 - val_loss: 1.0293 - val_f1_score: 0.4682 -
val_accuracy: 0.4682
Epoch 40/500
f1 score: 0.4679 - accuracy: 0.4679 - val_loss: 1.0253 - val_f1_score: 0.4402 -
val_accuracy: 0.4402
Epoch 41/500
f1 score: 0.4726 - accuracy: 0.4726 - val_loss: 1.0270 - val_f1_score: 0.4555 -
val_accuracy: 0.4555
Epoch 42/500
f1_score: 0.4661 - accuracy: 0.4661 - val_loss: 1.0402 - val_f1_score: 0.4548 -
val_accuracy: 0.4548
Epoch 43/500
f1 score: 0.4692 - accuracy: 0.4692 - val_loss: 1.0261 - val_f1_score: 0.4618 -
val_accuracy: 0.4618
Epoch 44/500
f1_score: 0.4680 - accuracy: 0.4680 - val_loss: 1.0216 - val_f1_score: 0.4733 -
val_accuracy: 0.4733
Epoch 45/500
f1_score: 0.4688 - accuracy: 0.4688 - val_loss: 1.0259 - val_f1_score: 0.4669 -
val accuracy: 0.4669
Epoch 46/500
f1_score: 0.4699 - accuracy: 0.4699 - val_loss: 1.0316 - val_f1_score: 0.4618 -
val_accuracy: 0.4618
Epoch 47/500
f1 score: 0.4605 - accuracy: 0.4605 - val_loss: 1.0266 - val_f1_score: 0.4828 -
val_accuracy: 0.4828
Epoch 48/500
f1_score: 0.4700 - accuracy: 0.4700 - val_loss: 1.0204 - val_f1_score: 0.4574 -
val_accuracy: 0.4574
Epoch 49/500
```

```
f1_score: 0.4711 - accuracy: 0.4711 - val_loss: 1.0209 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 50/500
f1_score: 0.4738 - accuracy: 0.4738 - val_loss: 1.0226 - val_f1_score: 0.4517 -
val accuracy: 0.4517
Epoch 51/500
f1_score: 0.4722 - accuracy: 0.4722 - val_loss: 1.0105 - val_f1_score: 0.4669 -
val_accuracy: 0.4669
Epoch 52/500
f1 score: 0.4728 - accuracy: 0.4728 - val_loss: 1.0351 - val_f1_score: 0.4561 -
val_accuracy: 0.4561
Epoch 53/500
f1 score: 0.4728 - accuracy: 0.4728 - val_loss: 1.0268 - val_f1_score: 0.4695 -
val_accuracy: 0.4695
Epoch 54/500
f1_score: 0.4714 - accuracy: 0.4714 - val_loss: 1.0073 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 55/500
f1 score: 0.4782 - accuracy: 0.4782 - val_loss: 1.0242 - val_f1_score: 0.4504 -
val_accuracy: 0.4504
Epoch 56/500
f1_score: 0.4720 - accuracy: 0.4720 - val_loss: 1.0277 - val_f1_score: 0.4777 -
val_accuracy: 0.4777
Epoch 57/500
f1_score: 0.4808 - accuracy: 0.4808 - val_loss: 1.0221 - val_f1_score: 0.4739 -
val accuracy: 0.4739
Epoch 58/500
f1_score: 0.4711 - accuracy: 0.4711 - val_loss: 1.0246 - val_f1_score: 0.4574 -
val_accuracy: 0.4574
Epoch 59/500
f1 score: 0.4695 - accuracy: 0.4695 - val_loss: 1.0219 - val_f1_score: 0.4720 -
val_accuracy: 0.4720
Epoch 60/500
f1_score: 0.4694 - accuracy: 0.4694 - val_loss: 1.0168 - val_f1_score: 0.4529 -
val_accuracy: 0.4529
Epoch 61/500
```

```
f1_score: 0.4782 - accuracy: 0.4782 - val_loss: 1.0202 - val_f1_score: 0.4695 -
val_accuracy: 0.4695
Epoch 62/500
f1_score: 0.4774 - accuracy: 0.4774 - val_loss: 1.0284 - val_f1_score: 0.4650 -
val accuracy: 0.4650
Epoch 63/500
f1_score: 0.4766 - accuracy: 0.4766 - val_loss: 1.0395 - val_f1_score: 0.4510 -
val_accuracy: 0.4510
Epoch 64/500
f1 score: 0.4762 - accuracy: 0.4762 - val_loss: 1.0159 - val_f1_score: 0.4650 -
val_accuracy: 0.4650
Epoch 65/500
f1 score: 0.4760 - accuracy: 0.4760 - val_loss: 1.0060 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 66/500
f1_score: 0.4806 - accuracy: 0.4806 - val_loss: 1.0206 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 67/500
f1 score: 0.4734 - accuracy: 0.4734 - val_loss: 1.0124 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 68/500
f1_score: 0.4784 - accuracy: 0.4784 - val_loss: 1.0120 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 69/500
f1_score: 0.4803 - accuracy: 0.4803 - val_loss: 1.0145 - val_f1_score: 0.4796 -
val accuracy: 0.4796
Epoch 70/500
f1_score: 0.4782 - accuracy: 0.4782 - val_loss: 1.0351 - val_f1_score: 0.4656 -
val_accuracy: 0.4656
Epoch 71/500
f1 score: 0.4787 - accuracy: 0.4787 - val_loss: 1.0309 - val_f1_score: 0.4676 -
val_accuracy: 0.4676
Epoch 72/500
f1_score: 0.4756 - accuracy: 0.4756 - val_loss: 1.0268 - val_f1_score: 0.4631 -
val_accuracy: 0.4631
Epoch 73/500
```

```
f1_score: 0.4721 - accuracy: 0.4721 - val_loss: 1.0316 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 74/500
f1_score: 0.4755 - accuracy: 0.4755 - val_loss: 1.0121 - val_f1_score: 0.4701 -
val accuracy: 0.4701
Epoch 75/500
f1_score: 0.4770 - accuracy: 0.4770 - val_loss: 1.0174 - val_f1_score: 0.4625 -
val_accuracy: 0.4625
Epoch 76/500
f1 score: 0.4753 - accuracy: 0.4753 - val_loss: 1.0054 - val_f1_score: 0.4828 -
val_accuracy: 0.4828
Epoch 77/500
f1 score: 0.4859 - accuracy: 0.4859 - val_loss: 1.0071 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 78/500
f1_score: 0.4793 - accuracy: 0.4793 - val_loss: 1.0183 - val_f1_score: 0.4777 -
val_accuracy: 0.4777
Epoch 79/500
f1 score: 0.4850 - accuracy: 0.4850 - val_loss: 1.0130 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 80/500
f1_score: 0.4825 - accuracy: 0.4825 - val_loss: 1.0085 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 81/500
442/442 [=========== ] - 1s 1ms/step - loss: 1.0038 -
f1_score: 0.4777 - accuracy: 0.4777 - val_loss: 1.0146 - val_f1_score: 0.4714 -
val accuracy: 0.4714
Epoch 82/500
f1_score: 0.4782 - accuracy: 0.4782 - val_loss: 1.0163 - val_f1_score: 0.4669 -
val_accuracy: 0.4669
Epoch 83/500
f1 score: 0.4836 - accuracy: 0.4836 - val_loss: 1.0166 - val_f1_score: 0.4695 -
val_accuracy: 0.4695
Epoch 84/500
f1_score: 0.4761 - accuracy: 0.4761 - val_loss: 1.0183 - val_f1_score: 0.4733 -
val_accuracy: 0.4733
Epoch 85/500
```

```
f1_score: 0.4796 - accuracy: 0.4796 - val_loss: 1.0095 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 86/500
f1_score: 0.4851 - accuracy: 0.4851 - val_loss: 0.9958 - val_f1_score: 0.4733 -
val accuracy: 0.4733
Epoch 87/500
f1_score: 0.4774 - accuracy: 0.4774 - val_loss: 1.0087 - val_f1_score: 0.4618 -
val_accuracy: 0.4618
Epoch 88/500
f1 score: 0.4799 - accuracy: 0.4799 - val_loss: 1.0154 - val_f1_score: 0.4567 -
val_accuracy: 0.4567
Epoch 89/500
f1 score: 0.4799 - accuracy: 0.4799 - val_loss: 1.0077 - val_f1_score: 0.4644 -
val_accuracy: 0.4644
Epoch 90/500
f1_score: 0.4739 - accuracy: 0.4739 - val_loss: 1.0024 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 91/500
f1 score: 0.4839 - accuracy: 0.4839 - val_loss: 1.0132 - val_f1_score: 0.4555 -
val_accuracy: 0.4555
Epoch 92/500
f1_score: 0.4768 - accuracy: 0.4768 - val_loss: 1.0065 - val_f1_score: 0.4587 -
val_accuracy: 0.4587
Epoch 93/500
f1_score: 0.4767 - accuracy: 0.4767 - val_loss: 1.0342 - val_f1_score: 0.4485 -
val accuracy: 0.4485
Epoch 94/500
f1_score: 0.4772 - accuracy: 0.4772 - val_loss: 1.0423 - val_f1_score: 0.4230 -
val_accuracy: 0.4230
Epoch 95/500
f1 score: 0.4808 - accuracy: 0.4808 - val_loss: 1.0054 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 96/500
f1_score: 0.4731 - accuracy: 0.4731 - val_loss: 1.0021 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 97/500
```

```
f1_score: 0.4878 - accuracy: 0.4878 - val_loss: 1.0080 - val_f1_score: 0.4650 -
val_accuracy: 0.4650
Epoch 98/500
f1_score: 0.4819 - accuracy: 0.4819 - val_loss: 1.0052 - val_f1_score: 0.4726 -
val accuracy: 0.4726
Epoch 99/500
f1_score: 0.4804 - accuracy: 0.4804 - val_loss: 1.0046 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 100/500
f1 score: 0.4810 - accuracy: 0.4810 - val_loss: 1.0220 - val_f1_score: 0.4733 -
val_accuracy: 0.4733
Epoch 101/500
f1 score: 0.4868 - accuracy: 0.4868 - val_loss: 1.0184 - val_f1_score: 0.4587 -
val_accuracy: 0.4587
Epoch 102/500
f1_score: 0.4827 - accuracy: 0.4827 - val_loss: 0.9976 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 103/500
f1_score: 0.4883 - accuracy: 0.4883 - val_loss: 1.0116 - val_f1_score: 0.4784 -
val_accuracy: 0.4784
Epoch 104/500
f1_score: 0.4821 - accuracy: 0.4821 - val_loss: 1.0042 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 105/500
f1_score: 0.4856 - accuracy: 0.4856 - val_loss: 1.0030 - val_f1_score: 0.4828 -
val accuracy: 0.4828
Epoch 106/500
f1_score: 0.4796 - accuracy: 0.4796 - val_loss: 0.9920 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 107/500
f1 score: 0.4825 - accuracy: 0.4825 - val_loss: 1.0085 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 108/500
f1_score: 0.4794 - accuracy: 0.4794 - val_loss: 1.0246 - val_f1_score: 0.4561 -
val_accuracy: 0.4561
Epoch 109/500
```

```
f1_score: 0.4866 - accuracy: 0.4866 - val_loss: 1.0100 - val_f1_score: 0.4714 -
val_accuracy: 0.4714
Epoch 110/500
f1_score: 0.4813 - accuracy: 0.4813 - val_loss: 1.0116 - val_f1_score: 0.4701 -
val accuracy: 0.4701
Epoch 111/500
f1_score: 0.4838 - accuracy: 0.4838 - val_loss: 0.9988 - val_f1_score: 0.4790 -
val_accuracy: 0.4790
Epoch 112/500
f1 score: 0.4856 - accuracy: 0.4856 - val_loss: 1.0008 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 113/500
f1 score: 0.4835 - accuracy: 0.4835 - val_loss: 1.0015 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 114/500
f1_score: 0.4876 - accuracy: 0.4876 - val_loss: 0.9919 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 115/500
f1 score: 0.4851 - accuracy: 0.4851 - val_loss: 0.9966 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 116/500
f1_score: 0.4880 - accuracy: 0.4880 - val_loss: 0.9899 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 117/500
f1_score: 0.4830 - accuracy: 0.4830 - val_loss: 1.0014 - val_f1_score: 0.4714 -
val accuracy: 0.4714
Epoch 118/500
f1_score: 0.4810 - accuracy: 0.4810 - val_loss: 1.0057 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 119/500
f1 score: 0.4873 - accuracy: 0.4873 - val_loss: 0.9981 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 120/500
f1_score: 0.4786 - accuracy: 0.4786 - val_loss: 1.0038 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 121/500
```

```
f1_score: 0.4881 - accuracy: 0.4881 - val_loss: 1.0003 - val_f1_score: 0.4746 -
val_accuracy: 0.4746
Epoch 122/500
f1_score: 0.4860 - accuracy: 0.4860 - val_loss: 1.0121 - val_f1_score: 0.4593 -
val accuracy: 0.4593
Epoch 123/500
f1_score: 0.4833 - accuracy: 0.4833 - val_loss: 1.0164 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 124/500
f1 score: 0.4841 - accuracy: 0.4841 - val_loss: 0.9976 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 125/500
f1 score: 0.4856 - accuracy: 0.4856 - val_loss: 1.0129 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 126/500
f1_score: 0.4870 - accuracy: 0.4870 - val_loss: 1.0083 - val_f1_score: 0.4656 -
val_accuracy: 0.4656
Epoch 127/500
f1 score: 0.4783 - accuracy: 0.4783 - val_loss: 1.0001 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 128/500
f1_score: 0.4843 - accuracy: 0.4843 - val_loss: 0.9873 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 129/500
442/442 [=========== ] - 1s 1ms/step - loss: 0.9909 -
f1_score: 0.4799 - accuracy: 0.4799 - val_loss: 0.9979 - val_f1_score: 0.4822 -
val accuracy: 0.4822
Epoch 130/500
f1_score: 0.4878 - accuracy: 0.4878 - val_loss: 1.0087 - val_f1_score: 0.4739 -
val_accuracy: 0.4739
Epoch 131/500
f1 score: 0.4898 - accuracy: 0.4898 - val_loss: 1.0107 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 132/500
442/442 [=========== ] - 1s 1ms/step - loss: 0.9928 -
f1_score: 0.4837 - accuracy: 0.4837 - val_loss: 0.9954 - val_f1_score: 0.4720 -
val_accuracy: 0.4720
Epoch 133/500
```

```
f1_score: 0.4809 - accuracy: 0.4809 - val_loss: 1.0103 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 134/500
f1_score: 0.4834 - accuracy: 0.4834 - val_loss: 1.0068 - val_f1_score: 0.4612 -
val accuracy: 0.4612
Epoch 135/500
f1_score: 0.4867 - accuracy: 0.4867 - val_loss: 1.0248 - val_f1_score: 0.4682 -
val_accuracy: 0.4682
Epoch 136/500
f1 score: 0.4830 - accuracy: 0.4830 - val_loss: 1.0012 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 137/500
f1 score: 0.4854 - accuracy: 0.4854 - val_loss: 0.9965 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 138/500
f1_score: 0.4827 - accuracy: 0.4827 - val_loss: 1.0020 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 139/500
f1 score: 0.4852 - accuracy: 0.4852 - val_loss: 1.0016 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 140/500
f1_score: 0.4846 - accuracy: 0.4846 - val_loss: 0.9948 - val_f1_score: 0.4841 -
val_accuracy: 0.4841
Epoch 141/500
f1_score: 0.4855 - accuracy: 0.4855 - val_loss: 1.0020 - val_f1_score: 0.4892 -
val accuracy: 0.4892
Epoch 142/500
f1_score: 0.4887 - accuracy: 0.4887 - val_loss: 1.0100 - val_f1_score: 0.4606 -
val_accuracy: 0.4606
Epoch 143/500
f1 score: 0.4958 - accuracy: 0.4958 - val_loss: 1.0028 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 144/500
442/442 [============ ] - 1s 2ms/step - loss: 0.9892 -
f1_score: 0.4861 - accuracy: 0.4861 - val_loss: 1.0019 - val_f1_score: 0.4695 -
val_accuracy: 0.4695
Epoch 145/500
```

```
f1_score: 0.4914 - accuracy: 0.4914 - val_loss: 1.0126 - val_f1_score: 0.4529 -
val_accuracy: 0.4529
Epoch 146/500
f1_score: 0.4911 - accuracy: 0.4911 - val_loss: 1.0026 - val_f1_score: 0.4866 -
val accuracy: 0.4866
Epoch 147/500
f1_score: 0.4960 - accuracy: 0.4960 - val_loss: 0.9941 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 148/500
f1 score: 0.4871 - accuracy: 0.4871 - val_loss: 0.9992 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 149/500
f1 score: 0.4829 - accuracy: 0.4829 - val_loss: 1.0023 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 150/500
f1_score: 0.4896 - accuracy: 0.4896 - val_loss: 0.9957 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 151/500
f1 score: 0.4944 - accuracy: 0.4944 - val_loss: 1.0087 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 152/500
f1_score: 0.4868 - accuracy: 0.4868 - val_loss: 1.0050 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 153/500
f1_score: 0.4868 - accuracy: 0.4868 - val_loss: 1.0025 - val_f1_score: 0.4822 -
val accuracy: 0.4822
Epoch 154/500
f1_score: 0.4929 - accuracy: 0.4929 - val_loss: 1.0120 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 155/500
f1 score: 0.4963 - accuracy: 0.4963 - val_loss: 1.0099 - val_f1_score: 0.4682 -
val_accuracy: 0.4682
Epoch 156/500
442/442 [=========== ] - 1s 2ms/step - loss: 0.9831 -
f1_score: 0.4922 - accuracy: 0.4922 - val_loss: 1.0022 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 157/500
```

```
f1_score: 0.4873 - accuracy: 0.4873 - val_loss: 1.0029 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 158/500
f1_score: 0.4932 - accuracy: 0.4932 - val_loss: 1.0038 - val_f1_score: 0.4771 -
val accuracy: 0.4771
Epoch 159/500
f1_score: 0.4948 - accuracy: 0.4948 - val_loss: 1.0213 - val_f1_score: 0.4669 -
val_accuracy: 0.4669
Epoch 160/500
f1 score: 0.4950 - accuracy: 0.4950 - val_loss: 0.9865 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 161/500
f1 score: 0.4929 - accuracy: 0.4929 - val_loss: 1.0005 - val_f1_score: 0.4688 -
val_accuracy: 0.4688
Epoch 162/500
f1_score: 0.4895 - accuracy: 0.4895 - val_loss: 1.0008 - val_f1_score: 0.4606 -
val_accuracy: 0.4606
Epoch 163/500
f1 score: 0.4874 - accuracy: 0.4874 - val_loss: 0.9916 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 164/500
f1_score: 0.4953 - accuracy: 0.4953 - val_loss: 0.9957 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 165/500
f1_score: 0.4914 - accuracy: 0.4914 - val_loss: 1.0012 - val_f1_score: 0.4911 -
val accuracy: 0.4911
Epoch 166/500
f1_score: 0.4931 - accuracy: 0.4931 - val_loss: 1.0107 - val_f1_score: 0.4784 -
val_accuracy: 0.4784
Epoch 167/500
442/442 [============== ] - 1s 1ms/step - loss: 0.9859 -
f1 score: 0.4946 - accuracy: 0.4946 - val_loss: 1.0073 - val_f1_score: 0.4695 -
val_accuracy: 0.4695
Epoch 168/500
f1_score: 0.4849 - accuracy: 0.4849 - val_loss: 0.9993 - val_f1_score: 0.4739 -
val_accuracy: 0.4739
Epoch 169/500
```

```
f1_score: 0.4858 - accuracy: 0.4858 - val_loss: 1.0129 - val_f1_score: 0.4701 -
val_accuracy: 0.4701
Epoch 170/500
f1_score: 0.4815 - accuracy: 0.4815 - val_loss: 1.0037 - val_f1_score: 0.4828 -
val accuracy: 0.4828
Epoch 171/500
f1_score: 0.4765 - accuracy: 0.4765 - val_loss: 1.0070 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 172/500
f1 score: 0.4842 - accuracy: 0.4842 - val_loss: 1.0017 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 173/500
f1_score: 0.4811 - accuracy: 0.4811 - val_loss: 0.9961 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 174/500
f1_score: 0.4866 - accuracy: 0.4866 - val_loss: 1.0017 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 175/500
f1 score: 0.4926 - accuracy: 0.4926 - val_loss: 0.9999 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 176/500
f1_score: 0.4928 - accuracy: 0.4928 - val_loss: 0.9971 - val_f1_score: 0.4784 -
val_accuracy: 0.4784
Epoch 177/500
f1_score: 0.4917 - accuracy: 0.4917 - val_loss: 1.0075 - val_f1_score: 0.4752 -
val accuracy: 0.4752
Epoch 178/500
f1_score: 0.4905 - accuracy: 0.4905 - val_loss: 1.0011 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 179/500
f1 score: 0.4942 - accuracy: 0.4942 - val_loss: 1.0109 - val_f1_score: 0.4637 -
val_accuracy: 0.4637
Epoch 180/500
f1_score: 0.4895 - accuracy: 0.4895 - val_loss: 0.9991 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 181/500
```

```
f1_score: 0.4883 - accuracy: 0.4883 - val_loss: 1.0108 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 182/500
f1_score: 0.4922 - accuracy: 0.4922 - val_loss: 1.0130 - val_f1_score: 0.5083 -
val accuracy: 0.5083
Epoch 183/500
f1_score: 0.4902 - accuracy: 0.4902 - val_loss: 0.9955 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 184/500
f1 score: 0.4906 - accuracy: 0.4906 - val_loss: 0.9913 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 185/500
f1 score: 0.4922 - accuracy: 0.4922 - val_loss: 1.0031 - val_f1_score: 0.4714 -
val_accuracy: 0.4714
Epoch 186/500
f1_score: 0.4897 - accuracy: 0.4897 - val_loss: 1.0032 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 187/500
f1 score: 0.4972 - accuracy: 0.4972 - val_loss: 0.9950 - val_f1_score: 0.4777 -
val_accuracy: 0.4777
Epoch 188/500
f1_score: 0.4869 - accuracy: 0.4869 - val_loss: 1.0210 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 189/500
442/442 [============ ] - 1s 1ms/step - loss: 0.9875 -
f1_score: 0.4832 - accuracy: 0.4832 - val_loss: 1.0028 - val_f1_score: 0.4765 -
val accuracy: 0.4765
Epoch 190/500
f1_score: 0.4799 - accuracy: 0.4799 - val_loss: 1.0051 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 191/500
f1 score: 0.4931 - accuracy: 0.4931 - val_loss: 0.9982 - val_f1_score: 0.4720 -
val_accuracy: 0.4720
Epoch 192/500
f1_score: 0.4869 - accuracy: 0.4869 - val_loss: 0.9979 - val_f1_score: 0.4739 -
val_accuracy: 0.4739
Epoch 193/500
```

```
f1_score: 0.4967 - accuracy: 0.4967 - val_loss: 1.0153 - val_f1_score: 0.4676 -
val_accuracy: 0.4676
Epoch 194/500
f1_score: 0.4934 - accuracy: 0.4934 - val_loss: 0.9983 - val_f1_score: 0.4765 -
val accuracy: 0.4765
Epoch 195/500
f1_score: 0.4874 - accuracy: 0.4874 - val_loss: 0.9994 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 196/500
f1 score: 0.4765 - accuracy: 0.4765 - val_loss: 1.0270 - val_f1_score: 0.4555 -
val_accuracy: 0.4555
Epoch 197/500
f1 score: 0.4926 - accuracy: 0.4926 - val_loss: 0.9924 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 198/500
f1_score: 0.4877 - accuracy: 0.4877 - val_loss: 1.0020 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 199/500
f1 score: 0.4902 - accuracy: 0.4902 - val_loss: 0.9927 - val_f1_score: 0.4847 -
val_accuracy: 0.4847
Epoch 200/500
f1_score: 0.4960 - accuracy: 0.4960 - val_loss: 0.9890 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 201/500
f1_score: 0.4934 - accuracy: 0.4934 - val_loss: 0.9872 - val_f1_score: 0.4822 -
val accuracy: 0.4822
Epoch 202/500
f1_score: 0.4934 - accuracy: 0.4934 - val_loss: 0.9935 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 203/500
f1 score: 0.4941 - accuracy: 0.4941 - val_loss: 1.0120 - val_f1_score: 0.4854 -
val_accuracy: 0.4854
Epoch 204/500
f1_score: 0.4961 - accuracy: 0.4961 - val_loss: 0.9930 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 205/500
```

```
f1_score: 0.4878 - accuracy: 0.4878 - val_loss: 1.0056 - val_f1_score: 0.4637 -
val_accuracy: 0.4637
Epoch 206/500
f1_score: 0.4948 - accuracy: 0.4948 - val_loss: 0.9985 - val_f1_score: 0.4924 -
val accuracy: 0.4924
Epoch 207/500
f1_score: 0.4979 - accuracy: 0.4979 - val_loss: 1.0049 - val_f1_score: 0.4790 -
val_accuracy: 0.4790
Epoch 208/500
f1 score: 0.4941 - accuracy: 0.4941 - val_loss: 1.0055 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 209/500
f1 score: 0.4857 - accuracy: 0.4857 - val_loss: 1.0018 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 210/500
f1_score: 0.4821 - accuracy: 0.4821 - val_loss: 0.9960 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 211/500
f1 score: 0.4886 - accuracy: 0.4886 - val_loss: 0.9920 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 212/500
f1_score: 0.4835 - accuracy: 0.4835 - val_loss: 0.9910 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 213/500
f1_score: 0.4852 - accuracy: 0.4852 - val_loss: 0.9942 - val_f1_score: 0.4828 -
val accuracy: 0.4828
Epoch 214/500
f1_score: 0.4881 - accuracy: 0.4881 - val_loss: 0.9895 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 215/500
f1 score: 0.4925 - accuracy: 0.4925 - val_loss: 0.9917 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 216/500
f1_score: 0.4951 - accuracy: 0.4951 - val_loss: 0.9927 - val_f1_score: 0.4987 -
val_accuracy: 0.4987
Epoch 217/500
```

```
f1_score: 0.4897 - accuracy: 0.4897 - val_loss: 0.9959 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 218/500
f1_score: 0.4911 - accuracy: 0.4911 - val_loss: 0.9933 - val_f1_score: 0.4917 -
val accuracy: 0.4917
Epoch 219/500
f1_score: 0.4994 - accuracy: 0.4994 - val_loss: 1.0028 - val_f1_score: 0.4682 -
val_accuracy: 0.4682
Epoch 220/500
f1 score: 0.4944 - accuracy: 0.4944 - val_loss: 0.9994 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 221/500
f1_score: 0.4931 - accuracy: 0.4931 - val_loss: 0.9891 - val_f1_score: 0.4752 -
val_accuracy: 0.4752
Epoch 222/500
f1_score: 0.4898 - accuracy: 0.4898 - val_loss: 1.0048 - val_f1_score: 0.4676 -
val_accuracy: 0.4676
Epoch 223/500
f1 score: 0.4960 - accuracy: 0.4960 - val_loss: 0.9983 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 224/500
f1_score: 0.4946 - accuracy: 0.4946 - val_loss: 0.9897 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 225/500
f1_score: 0.5020 - accuracy: 0.5020 - val_loss: 1.0013 - val_f1_score: 0.4726 -
val accuracy: 0.4726
Epoch 226/500
f1_score: 0.4965 - accuracy: 0.4965 - val_loss: 0.9955 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 227/500
f1 score: 0.4935 - accuracy: 0.4935 - val_loss: 1.0005 - val_f1_score: 0.4847 -
val_accuracy: 0.4847
Epoch 228/500
f1_score: 0.4946 - accuracy: 0.4946 - val_loss: 0.9949 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 229/500
```

```
f1_score: 0.4955 - accuracy: 0.4955 - val_loss: 0.9978 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 230/500
f1_score: 0.4968 - accuracy: 0.4968 - val_loss: 0.9995 - val_f1_score: 0.4949 -
val accuracy: 0.4949
Epoch 231/500
f1_score: 0.5018 - accuracy: 0.5018 - val_loss: 1.0017 - val_f1_score: 0.4777 -
val_accuracy: 0.4777
Epoch 232/500
f1 score: 0.4943 - accuracy: 0.4943 - val_loss: 0.9947 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 233/500
f1 score: 0.5025 - accuracy: 0.5025 - val_loss: 1.0020 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 234/500
f1_score: 0.4950 - accuracy: 0.4950 - val_loss: 1.0056 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 235/500
f1 score: 0.4943 - accuracy: 0.4943 - val_loss: 1.0034 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 236/500
f1_score: 0.4975 - accuracy: 0.4975 - val_loss: 1.0017 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 237/500
f1_score: 0.4992 - accuracy: 0.4992 - val_loss: 0.9936 - val_f1_score: 0.4733 -
val accuracy: 0.4733
Epoch 238/500
f1_score: 0.4919 - accuracy: 0.4919 - val_loss: 1.0014 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 239/500
f1 score: 0.5006 - accuracy: 0.5006 - val_loss: 0.9909 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 240/500
f1_score: 0.4965 - accuracy: 0.4965 - val_loss: 0.9918 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 241/500
```

```
f1_score: 0.4960 - accuracy: 0.4960 - val_loss: 0.9897 - val_f1_score: 0.4854 -
val_accuracy: 0.4854
Epoch 242/500
f1_score: 0.4937 - accuracy: 0.4937 - val_loss: 1.0007 - val_f1_score: 0.4847 -
val accuracy: 0.4847
Epoch 243/500
f1_score: 0.5025 - accuracy: 0.5025 - val_loss: 0.9940 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 244/500
f1 score: 0.4984 - accuracy: 0.4984 - val_loss: 1.0143 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 245/500
f1 score: 0.4931 - accuracy: 0.4931 - val_loss: 1.0047 - val_f1_score: 0.4790 -
val_accuracy: 0.4790
Epoch 246/500
f1_score: 0.4955 - accuracy: 0.4955 - val_loss: 1.0034 - val_f1_score: 0.4841 -
val_accuracy: 0.4841
Epoch 247/500
f1 score: 0.4996 - accuracy: 0.4996 - val_loss: 0.9973 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 248/500
f1_score: 0.4902 - accuracy: 0.4902 - val_loss: 0.9951 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 249/500
f1_score: 0.4948 - accuracy: 0.4948 - val_loss: 0.9998 - val_f1_score: 0.4777 -
val accuracy: 0.4777
Epoch 250/500
f1_score: 0.4958 - accuracy: 0.4958 - val_loss: 0.9929 - val_f1_score: 0.4777 -
val_accuracy: 0.4777
Epoch 251/500
f1 score: 0.4956 - accuracy: 0.4956 - val_loss: 1.0026 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 252/500
f1_score: 0.4991 - accuracy: 0.4991 - val_loss: 0.9932 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 253/500
```

```
f1_score: 0.5041 - accuracy: 0.5041 - val_loss: 0.9841 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 254/500
f1_score: 0.4972 - accuracy: 0.4972 - val_loss: 0.9940 - val_f1_score: 0.4981 -
val accuracy: 0.4981
Epoch 255/500
f1_score: 0.4955 - accuracy: 0.4955 - val_loss: 0.9936 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 256/500
f1 score: 0.4999 - accuracy: 0.4999 - val_loss: 0.9841 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 257/500
f1 score: 0.5004 - accuracy: 0.5004 - val_loss: 0.9887 - val_f1_score: 0.4987 -
val_accuracy: 0.4987
Epoch 258/500
f1_score: 0.5040 - accuracy: 0.5040 - val_loss: 0.9974 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 259/500
f1 score: 0.5046 - accuracy: 0.5046 - val_loss: 0.9874 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 260/500
f1_score: 0.4969 - accuracy: 0.4969 - val_loss: 1.0176 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 261/500
f1_score: 0.4958 - accuracy: 0.4958 - val_loss: 0.9888 - val_f1_score: 0.4955 -
val accuracy: 0.4955
Epoch 262/500
f1_score: 0.5013 - accuracy: 0.5013 - val_loss: 0.9854 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 263/500
f1 score: 0.5003 - accuracy: 0.5003 - val_loss: 0.9843 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 264/500
f1_score: 0.4956 - accuracy: 0.4956 - val_loss: 0.9938 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 265/500
```

```
f1_score: 0.5020 - accuracy: 0.5020 - val_loss: 0.9861 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 266/500
f1_score: 0.5037 - accuracy: 0.5037 - val_loss: 0.9960 - val_f1_score: 0.4885 -
val accuracy: 0.4885
Epoch 267/500
f1_score: 0.5013 - accuracy: 0.5013 - val_loss: 0.9867 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 268/500
f1 score: 0.5030 - accuracy: 0.5030 - val_loss: 0.9938 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 269/500
f1 score: 0.5076 - accuracy: 0.5076 - val_loss: 0.9988 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 270/500
f1_score: 0.5018 - accuracy: 0.5018 - val_loss: 0.9872 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 271/500
f1 score: 0.5013 - accuracy: 0.5013 - val_loss: 0.9912 - val_f1_score: 0.4847 -
val_accuracy: 0.4847
Epoch 272/500
f1_score: 0.5047 - accuracy: 0.5047 - val_loss: 0.9871 - val_f1_score: 0.4962 -
val_accuracy: 0.4962
Epoch 273/500
f1_score: 0.4935 - accuracy: 0.4935 - val_loss: 0.9976 - val_f1_score: 0.4892 -
val accuracy: 0.4892
Epoch 274/500
f1_score: 0.4936 - accuracy: 0.4936 - val_loss: 0.9871 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 275/500
f1 score: 0.5029 - accuracy: 0.5029 - val_loss: 0.9942 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 276/500
f1_score: 0.4999 - accuracy: 0.4999 - val_loss: 0.9940 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 277/500
```

```
f1_score: 0.5069 - accuracy: 0.5069 - val_loss: 0.9972 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 278/500
f1_score: 0.5057 - accuracy: 0.5057 - val_loss: 0.9918 - val_f1_score: 0.5006 -
val accuracy: 0.5006
Epoch 279/500
f1_score: 0.5028 - accuracy: 0.5028 - val_loss: 0.9959 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 280/500
f1 score: 0.4992 - accuracy: 0.4992 - val_loss: 0.9903 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 281/500
f1 score: 0.5069 - accuracy: 0.5069 - val_loss: 0.9850 - val_f1_score: 0.4854 -
val_accuracy: 0.4854
Epoch 282/500
f1_score: 0.4984 - accuracy: 0.4984 - val_loss: 0.9821 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 283/500
f1 score: 0.5054 - accuracy: 0.5054 - val_loss: 0.9865 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 284/500
f1_score: 0.5100 - accuracy: 0.5100 - val_loss: 1.0051 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 285/500
f1_score: 0.5034 - accuracy: 0.5034 - val_loss: 0.9919 - val_f1_score: 0.5019 -
val accuracy: 0.5019
Epoch 286/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 1.0045 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 287/500
f1 score: 0.4906 - accuracy: 0.4906 - val_loss: 0.9914 - val_f1_score: 0.4854 -
val_accuracy: 0.4854
Epoch 288/500
f1_score: 0.5067 - accuracy: 0.5067 - val_loss: 0.9887 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 289/500
```

```
f1_score: 0.5098 - accuracy: 0.5098 - val_loss: 0.9925 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 290/500
f1_score: 0.4992 - accuracy: 0.4992 - val_loss: 0.9855 - val_f1_score: 0.4917 -
val accuracy: 0.4917
Epoch 291/500
f1_score: 0.5014 - accuracy: 0.5014 - val_loss: 0.9939 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 292/500
f1 score: 0.5002 - accuracy: 0.5002 - val_loss: 1.0072 - val_f1_score: 0.4784 -
val_accuracy: 0.4784
Epoch 293/500
f1 score: 0.4979 - accuracy: 0.4979 - val_loss: 0.9962 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 294/500
f1_score: 0.5012 - accuracy: 0.5012 - val_loss: 0.9933 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 295/500
f1 score: 0.5054 - accuracy: 0.5054 - val_loss: 0.9965 - val_f1_score: 0.4758 -
val_accuracy: 0.4758
Epoch 296/500
f1_score: 0.5045 - accuracy: 0.5045 - val_loss: 0.9854 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 297/500
f1_score: 0.5074 - accuracy: 0.5074 - val_loss: 0.9834 - val_f1_score: 0.4962 -
val accuracy: 0.4962
Epoch 298/500
f1_score: 0.5076 - accuracy: 0.5076 - val_loss: 0.9905 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 299/500
f1 score: 0.5061 - accuracy: 0.5061 - val_loss: 0.9954 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 300/500
f1_score: 0.5069 - accuracy: 0.5069 - val_loss: 1.0002 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 301/500
```

```
f1_score: 0.5004 - accuracy: 0.5004 - val_loss: 0.9884 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 302/500
f1_score: 0.5044 - accuracy: 0.5044 - val_loss: 0.9936 - val_f1_score: 0.4924 -
val accuracy: 0.4924
Epoch 303/500
f1_score: 0.5060 - accuracy: 0.5060 - val_loss: 1.0092 - val_f1_score: 0.4765 -
val_accuracy: 0.4765
Epoch 304/500
f1 score: 0.5042 - accuracy: 0.5042 - val_loss: 0.9971 - val_f1_score: 0.4714 -
val_accuracy: 0.4714
Epoch 305/500
f1 score: 0.5013 - accuracy: 0.5013 - val_loss: 0.9894 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 306/500
f1_score: 0.5047 - accuracy: 0.5047 - val_loss: 0.9821 - val_f1_score: 0.4975 -
val_accuracy: 0.4975
Epoch 307/500
f1 score: 0.5045 - accuracy: 0.5045 - val_loss: 0.9980 - val_f1_score: 0.4707 -
val_accuracy: 0.4707
Epoch 308/500
f1_score: 0.5025 - accuracy: 0.5025 - val_loss: 0.9934 - val_f1_score: 0.4847 -
val_accuracy: 0.4847
Epoch 309/500
f1_score: 0.5045 - accuracy: 0.5045 - val_loss: 1.0039 - val_f1_score: 0.5025 -
val accuracy: 0.5025
Epoch 310/500
f1_score: 0.4996 - accuracy: 0.4996 - val_loss: 1.0004 - val_f1_score: 0.4841 -
val_accuracy: 0.4841
Epoch 311/500
f1 score: 0.4992 - accuracy: 0.4992 - val_loss: 0.9855 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 312/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 0.9791 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 313/500
```

```
f1_score: 0.5074 - accuracy: 0.5074 - val_loss: 0.9838 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 314/500
f1_score: 0.5076 - accuracy: 0.5076 - val_loss: 0.9742 - val_f1_score: 0.5032 -
val accuracy: 0.5032
Epoch 315/500
f1_score: 0.5011 - accuracy: 0.5011 - val_loss: 0.9992 - val_f1_score: 0.4968 -
val_accuracy: 0.4968
Epoch 316/500
f1 score: 0.5028 - accuracy: 0.5028 - val_loss: 1.0089 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 317/500
f1 score: 0.5059 - accuracy: 0.5059 - val_loss: 0.9933 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 318/500
f1_score: 0.5067 - accuracy: 0.5067 - val_loss: 0.9840 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 319/500
f1 score: 0.5035 - accuracy: 0.5035 - val_loss: 0.9827 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 320/500
f1_score: 0.5012 - accuracy: 0.5012 - val_loss: 0.9865 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 321/500
f1_score: 0.5001 - accuracy: 0.5001 - val_loss: 0.9819 - val_f1_score: 0.4765 -
val accuracy: 0.4765
Epoch 322/500
f1_score: 0.5043 - accuracy: 0.5043 - val_loss: 0.9976 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 323/500
f1 score: 0.5109 - accuracy: 0.5109 - val_loss: 0.9950 - val_f1_score: 0.5000 -
val_accuracy: 0.5000
Epoch 324/500
f1_score: 0.5078 - accuracy: 0.5078 - val_loss: 0.9861 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 325/500
```

```
f1_score: 0.5046 - accuracy: 0.5046 - val_loss: 0.9875 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 326/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 0.9961 - val_f1_score: 0.4854 -
val accuracy: 0.4854
Epoch 327/500
f1_score: 0.5037 - accuracy: 0.5037 - val_loss: 1.0058 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 328/500
f1 score: 0.4993 - accuracy: 0.4993 - val_loss: 0.9908 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 329/500
f1 score: 0.5073 - accuracy: 0.5073 - val_loss: 1.0043 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 330/500
f1_score: 0.5054 - accuracy: 0.5054 - val_loss: 0.9943 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 331/500
f1 score: 0.5105 - accuracy: 0.5105 - val_loss: 0.9919 - val_f1_score: 0.4981 -
val_accuracy: 0.4981
Epoch 332/500
f1_score: 0.5033 - accuracy: 0.5033 - val_loss: 0.9854 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 333/500
f1_score: 0.5031 - accuracy: 0.5031 - val_loss: 0.9861 - val_f1_score: 0.4898 -
val accuracy: 0.4898
Epoch 334/500
f1_score: 0.5047 - accuracy: 0.5047 - val_loss: 0.9913 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 335/500
f1 score: 0.5082 - accuracy: 0.5082 - val_loss: 0.9958 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 336/500
f1_score: 0.5045 - accuracy: 0.5045 - val_loss: 1.0112 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 337/500
```

```
f1_score: 0.5012 - accuracy: 0.5012 - val_loss: 0.9932 - val_f1_score: 0.4975 -
val_accuracy: 0.4975
Epoch 338/500
f1_score: 0.5086 - accuracy: 0.5086 - val_loss: 0.9913 - val_f1_score: 0.5089 -
val accuracy: 0.5089
Epoch 339/500
f1_score: 0.5063 - accuracy: 0.5063 - val_loss: 0.9896 - val_f1_score: 0.4975 -
val_accuracy: 0.4975
Epoch 340/500
f1 score: 0.5035 - accuracy: 0.5035 - val_loss: 0.9958 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 341/500
f1 score: 0.5064 - accuracy: 0.5064 - val_loss: 0.9835 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 342/500
f1_score: 0.5063 - accuracy: 0.5063 - val_loss: 0.9954 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 343/500
f1 score: 0.5076 - accuracy: 0.5076 - val_loss: 0.9905 - val_f1_score: 0.4835 -
val_accuracy: 0.4835
Epoch 344/500
f1_score: 0.5025 - accuracy: 0.5025 - val_loss: 0.9964 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 345/500
f1_score: 0.5077 - accuracy: 0.5077 - val_loss: 0.9780 - val_f1_score: 0.5070 -
val accuracy: 0.5070
Epoch 346/500
f1_score: 0.5088 - accuracy: 0.5088 - val_loss: 0.9832 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 347/500
f1 score: 0.5066 - accuracy: 0.5066 - val_loss: 0.9926 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 348/500
f1_score: 0.5055 - accuracy: 0.5055 - val_loss: 1.0012 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 349/500
```

```
f1_score: 0.5085 - accuracy: 0.5085 - val_loss: 0.9758 - val_f1_score: 0.5159 -
val_accuracy: 0.5159
Epoch 350/500
f1_score: 0.5100 - accuracy: 0.5100 - val_loss: 0.9861 - val_f1_score: 0.5019 -
val accuracy: 0.5019
Epoch 351/500
f1_score: 0.5023 - accuracy: 0.5023 - val_loss: 0.9840 - val_f1_score: 0.5102 -
val_accuracy: 0.5102
Epoch 352/500
f1 score: 0.5052 - accuracy: 0.5052 - val_loss: 0.9855 - val_f1_score: 0.5006 -
val_accuracy: 0.5006
Epoch 353/500
f1 score: 0.5035 - accuracy: 0.5035 - val_loss: 0.9898 - val_f1_score: 0.5013 -
val_accuracy: 0.5013
Epoch 354/500
f1_score: 0.5085 - accuracy: 0.5085 - val_loss: 0.9913 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 355/500
f1 score: 0.5074 - accuracy: 0.5074 - val_loss: 0.9907 - val_f1_score: 0.5000 -
val_accuracy: 0.5000
Epoch 356/500
f1_score: 0.5106 - accuracy: 0.5106 - val_loss: 0.9832 - val_f1_score: 0.4968 -
val_accuracy: 0.4968
Epoch 357/500
442/442 [=========== ] - 1s 1ms/step - loss: 0.9584 -
f1_score: 0.5133 - accuracy: 0.5133 - val_loss: 0.9902 - val_f1_score: 0.4898 -
val accuracy: 0.4898
Epoch 358/500
f1_score: 0.5141 - accuracy: 0.5141 - val_loss: 0.9965 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 359/500
f1 score: 0.5077 - accuracy: 0.5077 - val_loss: 0.9899 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 360/500
f1_score: 0.4923 - accuracy: 0.4923 - val_loss: 1.0004 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 361/500
```

```
f1_score: 0.5041 - accuracy: 0.5041 - val_loss: 0.9864 - val_f1_score: 0.5025 -
val_accuracy: 0.5025
Epoch 362/500
f1_score: 0.5028 - accuracy: 0.5028 - val_loss: 0.9838 - val_f1_score: 0.4936 -
val accuracy: 0.4936
Epoch 363/500
f1_score: 0.4994 - accuracy: 0.4994 - val_loss: 0.9957 - val_f1_score: 0.4790 -
val_accuracy: 0.4790
Epoch 364/500
f1 score: 0.5069 - accuracy: 0.5069 - val_loss: 0.9987 - val_f1_score: 0.4905 -
val_accuracy: 0.4905
Epoch 365/500
f1 score: 0.5132 - accuracy: 0.5132 - val_loss: 0.9799 - val_f1_score: 0.5102 -
val_accuracy: 0.5102
Epoch 366/500
f1_score: 0.5057 - accuracy: 0.5057 - val_loss: 0.9908 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 367/500
f1 score: 0.5044 - accuracy: 0.5044 - val_loss: 0.9869 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 368/500
f1_score: 0.5015 - accuracy: 0.5015 - val_loss: 0.9844 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 369/500
f1_score: 0.5083 - accuracy: 0.5083 - val_loss: 0.9802 - val_f1_score: 0.4981 -
val accuracy: 0.4981
Epoch 370/500
f1_score: 0.5157 - accuracy: 0.5157 - val_loss: 0.9921 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 371/500
f1 score: 0.5038 - accuracy: 0.5038 - val_loss: 0.9817 - val_f1_score: 0.4987 -
val_accuracy: 0.4987
Epoch 372/500
f1_score: 0.5100 - accuracy: 0.5100 - val_loss: 0.9950 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 373/500
```

```
f1_score: 0.5053 - accuracy: 0.5053 - val_loss: 0.9920 - val_f1_score: 0.5025 -
val_accuracy: 0.5025
Epoch 374/500
f1_score: 0.5082 - accuracy: 0.5082 - val_loss: 0.9784 - val_f1_score: 0.5070 -
val accuracy: 0.5070
Epoch 375/500
f1_score: 0.5085 - accuracy: 0.5085 - val_loss: 0.9834 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 376/500
f1 score: 0.5101 - accuracy: 0.5101 - val_loss: 0.9765 - val_f1_score: 0.5019 -
val_accuracy: 0.5019
Epoch 377/500
f1 score: 0.5075 - accuracy: 0.5075 - val_loss: 0.9796 - val_f1_score: 0.5083 -
val_accuracy: 0.5083
Epoch 378/500
f1_score: 0.5040 - accuracy: 0.5040 - val_loss: 0.9768 - val_f1_score: 0.5038 -
val_accuracy: 0.5038
Epoch 379/500
f1 score: 0.5099 - accuracy: 0.5099 - val_loss: 0.9726 - val_f1_score: 0.5064 -
val_accuracy: 0.5064
Epoch 380/500
f1_score: 0.5135 - accuracy: 0.5135 - val_loss: 0.9980 - val_f1_score: 0.5000 -
val_accuracy: 0.5000
Epoch 381/500
f1_score: 0.5112 - accuracy: 0.5112 - val_loss: 0.9930 - val_f1_score: 0.5019 -
val accuracy: 0.5019
Epoch 382/500
f1_score: 0.5114 - accuracy: 0.5114 - val_loss: 0.9821 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 383/500
f1 score: 0.4996 - accuracy: 0.4996 - val_loss: 0.9892 - val_f1_score: 0.5064 -
val_accuracy: 0.5064
Epoch 384/500
f1_score: 0.5075 - accuracy: 0.5075 - val_loss: 0.9980 - val_f1_score: 0.4968 -
val_accuracy: 0.4968
Epoch 385/500
```

```
f1_score: 0.5037 - accuracy: 0.5037 - val_loss: 0.9886 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 386/500
f1_score: 0.4968 - accuracy: 0.4968 - val_loss: 0.9847 - val_f1_score: 0.5083 -
val accuracy: 0.5083
Epoch 387/500
f1_score: 0.5045 - accuracy: 0.5045 - val_loss: 0.9832 - val_f1_score: 0.5019 -
val_accuracy: 0.5019
Epoch 388/500
f1 score: 0.5125 - accuracy: 0.5125 - val_loss: 0.9797 - val_f1_score: 0.5038 -
val_accuracy: 0.5038
Epoch 389/500
f1 score: 0.5117 - accuracy: 0.5117 - val_loss: 0.9829 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 390/500
f1_score: 0.5119 - accuracy: 0.5119 - val_loss: 0.9794 - val_f1_score: 0.5019 -
val_accuracy: 0.5019
Epoch 391/500
f1 score: 0.5061 - accuracy: 0.5061 - val_loss: 0.9968 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 392/500
f1_score: 0.5029 - accuracy: 0.5029 - val_loss: 0.9863 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 393/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 1.0031 - val_f1_score: 0.4936 -
val accuracy: 0.4936
Epoch 394/500
f1_score: 0.5035 - accuracy: 0.5035 - val_loss: 0.9920 - val_f1_score: 0.5070 -
val_accuracy: 0.5070
Epoch 395/500
f1 score: 0.5087 - accuracy: 0.5087 - val_loss: 0.9872 - val_f1_score: 0.5038 -
val_accuracy: 0.5038
Epoch 396/500
f1_score: 0.5049 - accuracy: 0.5049 - val_loss: 0.9944 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 397/500
```

```
f1_score: 0.5004 - accuracy: 0.5004 - val_loss: 0.9853 - val_f1_score: 0.5089 -
val_accuracy: 0.5089
Epoch 398/500
f1_score: 0.5057 - accuracy: 0.5057 - val_loss: 0.9834 - val_f1_score: 0.4860 -
val accuracy: 0.4860
Epoch 399/500
f1_score: 0.5038 - accuracy: 0.5038 - val_loss: 0.9878 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 400/500
f1 score: 0.5062 - accuracy: 0.5062 - val_loss: 0.9834 - val_f1_score: 0.5013 -
val_accuracy: 0.5013
Epoch 401/500
f1 score: 0.5108 - accuracy: 0.5108 - val_loss: 0.9859 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 402/500
f1_score: 0.4999 - accuracy: 0.4999 - val_loss: 0.9998 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 403/500
f1 score: 0.5078 - accuracy: 0.5078 - val_loss: 0.9815 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 404/500
f1_score: 0.5142 - accuracy: 0.5142 - val_loss: 0.9757 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 405/500
f1_score: 0.5074 - accuracy: 0.5074 - val_loss: 0.9924 - val_f1_score: 0.5051 -
val accuracy: 0.5051
Epoch 406/500
f1_score: 0.5031 - accuracy: 0.5031 - val_loss: 0.9732 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 407/500
f1 score: 0.4977 - accuracy: 0.4977 - val_loss: 0.9873 - val_f1_score: 0.5025 -
val_accuracy: 0.5025
Epoch 408/500
f1_score: 0.5029 - accuracy: 0.5029 - val_loss: 0.9973 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 409/500
```

```
f1_score: 0.5059 - accuracy: 0.5059 - val_loss: 0.9867 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 410/500
f1_score: 0.5109 - accuracy: 0.5109 - val_loss: 1.0321 - val_f1_score: 0.4606 -
val accuracy: 0.4606
Epoch 411/500
f1_score: 0.5005 - accuracy: 0.5005 - val_loss: 0.9907 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 412/500
f1 score: 0.5091 - accuracy: 0.5091 - val_loss: 0.9910 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 413/500
f1 score: 0.5075 - accuracy: 0.5075 - val_loss: 0.9870 - val_f1_score: 0.4885 -
val_accuracy: 0.4885
Epoch 414/500
f1_score: 0.5063 - accuracy: 0.5063 - val_loss: 0.9873 - val_f1_score: 0.4930 -
val_accuracy: 0.4930
Epoch 415/500
f1 score: 0.5047 - accuracy: 0.5047 - val_loss: 0.9946 - val_f1_score: 0.4981 -
val_accuracy: 0.4981
Epoch 416/500
f1_score: 0.5068 - accuracy: 0.5068 - val_loss: 0.9839 - val_f1_score: 0.4873 -
val_accuracy: 0.4873
Epoch 417/500
f1_score: 0.5065 - accuracy: 0.5065 - val_loss: 0.9859 - val_f1_score: 0.4930 -
val accuracy: 0.4930
Epoch 418/500
f1_score: 0.5088 - accuracy: 0.5088 - val_loss: 0.9861 - val_f1_score: 0.4975 -
val_accuracy: 0.4975
Epoch 419/500
f1 score: 0.5136 - accuracy: 0.5136 - val_loss: 0.9865 - val_f1_score: 0.4784 -
val_accuracy: 0.4784
Epoch 420/500
f1_score: 0.5081 - accuracy: 0.5081 - val_loss: 0.9870 - val_f1_score: 0.5083 -
val_accuracy: 0.5083
Epoch 421/500
```

```
f1_score: 0.5111 - accuracy: 0.5111 - val_loss: 0.9939 - val_f1_score: 0.4955 -
val_accuracy: 0.4955
Epoch 422/500
f1_score: 0.5125 - accuracy: 0.5125 - val_loss: 0.9876 - val_f1_score: 0.4955 -
val accuracy: 0.4955
Epoch 423/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 0.9878 - val_f1_score: 0.4911 -
val_accuracy: 0.4911
Epoch 424/500
f1 score: 0.5028 - accuracy: 0.5028 - val_loss: 0.9841 - val_f1_score: 0.4917 -
val_accuracy: 0.4917
Epoch 425/500
f1 score: 0.5089 - accuracy: 0.5089 - val_loss: 0.9755 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 426/500
f1_score: 0.5114 - accuracy: 0.5114 - val_loss: 0.9863 - val_f1_score: 0.4746 -
val_accuracy: 0.4746
Epoch 427/500
f1 score: 0.5153 - accuracy: 0.5153 - val_loss: 0.9913 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 428/500
f1_score: 0.5067 - accuracy: 0.5067 - val_loss: 0.9896 - val_f1_score: 0.4828 -
val_accuracy: 0.4828
Epoch 429/500
f1_score: 0.5064 - accuracy: 0.5064 - val_loss: 0.9905 - val_f1_score: 0.4809 -
val accuracy: 0.4809
Epoch 430/500
f1_score: 0.5099 - accuracy: 0.5099 - val_loss: 1.0193 - val_f1_score: 0.4809 -
val_accuracy: 0.4809
Epoch 431/500
f1 score: 0.5030 - accuracy: 0.5030 - val_loss: 0.9776 - val_f1_score: 0.4924 -
val_accuracy: 0.4924
Epoch 432/500
f1_score: 0.5033 - accuracy: 0.5033 - val_loss: 0.9858 - val_f1_score: 0.4898 -
val_accuracy: 0.4898
Epoch 433/500
```

```
f1_score: 0.5025 - accuracy: 0.5025 - val_loss: 0.9951 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 434/500
f1_score: 0.4992 - accuracy: 0.4992 - val_loss: 0.9896 - val_f1_score: 0.4981 -
val accuracy: 0.4981
Epoch 435/500
f1_score: 0.4953 - accuracy: 0.4953 - val_loss: 0.9952 - val_f1_score: 0.4726 -
val_accuracy: 0.4726
Epoch 436/500
f1 score: 0.4924 - accuracy: 0.4924 - val_loss: 0.9908 - val_f1_score: 0.4879 -
val_accuracy: 0.4879
Epoch 437/500
f1 score: 0.4835 - accuracy: 0.4835 - val_loss: 1.0090 - val_f1_score: 0.4612 -
val_accuracy: 0.4612
Epoch 438/500
f1_score: 0.4987 - accuracy: 0.4987 - val_loss: 1.0002 - val_f1_score: 0.4739 -
val_accuracy: 0.4739
Epoch 439/500
f1 score: 0.5006 - accuracy: 0.5006 - val_loss: 1.0114 - val_f1_score: 0.4803 -
val_accuracy: 0.4803
Epoch 440/500
f1_score: 0.4976 - accuracy: 0.4976 - val_loss: 1.0031 - val_f1_score: 0.4720 -
val_accuracy: 0.4720
Epoch 441/500
f1_score: 0.4952 - accuracy: 0.4952 - val_loss: 1.0031 - val_f1_score: 0.4758 -
val accuracy: 0.4758
Epoch 442/500
f1_score: 0.4973 - accuracy: 0.4973 - val_loss: 1.0077 - val_f1_score: 0.4854 -
val_accuracy: 0.4854
Epoch 443/500
f1 score: 0.5057 - accuracy: 0.5057 - val_loss: 0.9908 - val_f1_score: 0.4866 -
val_accuracy: 0.4866
Epoch 444/500
f1_score: 0.5109 - accuracy: 0.5109 - val_loss: 0.9924 - val_f1_score: 0.4860 -
val_accuracy: 0.4860
Epoch 445/500
```

```
f1_score: 0.5079 - accuracy: 0.5079 - val_loss: 0.9941 - val_f1_score: 0.5019 -
val_accuracy: 0.5019
Epoch 446/500
f1_score: 0.5086 - accuracy: 0.5086 - val_loss: 1.0018 - val_f1_score: 0.4873 -
val accuracy: 0.4873
Epoch 447/500
f1_score: 0.5007 - accuracy: 0.5007 - val_loss: 0.9914 - val_f1_score: 0.4892 -
val_accuracy: 0.4892
Epoch 448/500
f1 score: 0.5062 - accuracy: 0.5062 - val_loss: 1.0011 - val_f1_score: 0.4968 -
val_accuracy: 0.4968
Epoch 449/500
f1 score: 0.5107 - accuracy: 0.5107 - val_loss: 0.9917 - val_f1_score: 0.5019 -
val_accuracy: 0.5019
Epoch 450/500
f1_score: 0.5134 - accuracy: 0.5134 - val_loss: 0.9830 - val_f1_score: 0.5000 -
val_accuracy: 0.5000
Epoch 451/500
f1 score: 0.5113 - accuracy: 0.5113 - val_loss: 0.9878 - val_f1_score: 0.4987 -
val_accuracy: 0.4987
Epoch 452/500
f1_score: 0.5104 - accuracy: 0.5104 - val_loss: 0.9894 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 453/500
f1_score: 0.5095 - accuracy: 0.5095 - val_loss: 0.9925 - val_f1_score: 0.4962 -
val accuracy: 0.4962
Epoch 454/500
f1_score: 0.5133 - accuracy: 0.5133 - val_loss: 0.9792 - val_f1_score: 0.5095 -
val_accuracy: 0.5095
Epoch 455/500
f1 score: 0.5140 - accuracy: 0.5140 - val_loss: 0.9894 - val_f1_score: 0.5089 -
val_accuracy: 0.5089
Epoch 456/500
f1_score: 0.5057 - accuracy: 0.5057 - val_loss: 0.9876 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 457/500
```

```
f1_score: 0.5066 - accuracy: 0.5066 - val_loss: 0.9975 - val_f1_score: 0.4822 -
val_accuracy: 0.4822
Epoch 458/500
f1_score: 0.5100 - accuracy: 0.5100 - val_loss: 0.9804 - val_f1_score: 0.5153 -
val accuracy: 0.5153
Epoch 459/500
f1_score: 0.5052 - accuracy: 0.5052 - val_loss: 0.9969 - val_f1_score: 0.4936 -
val_accuracy: 0.4936
Epoch 460/500
f1 score: 0.5062 - accuracy: 0.5062 - val_loss: 0.9691 - val_f1_score: 0.5115 -
val_accuracy: 0.5115
Epoch 461/500
f1_score: 0.5021 - accuracy: 0.5021 - val_loss: 0.9912 - val_f1_score: 0.5013 -
val_accuracy: 0.5013
Epoch 462/500
f1_score: 0.5095 - accuracy: 0.5095 - val_loss: 0.9882 - val_f1_score: 0.4994 -
val_accuracy: 0.4994
Epoch 463/500
f1 score: 0.5095 - accuracy: 0.5095 - val_loss: 0.9878 - val_f1_score: 0.5038 -
val_accuracy: 0.5038
Epoch 464/500
f1_score: 0.5085 - accuracy: 0.5085 - val_loss: 0.9868 - val_f1_score: 0.5083 -
val_accuracy: 0.5083
Epoch 465/500
f1_score: 0.5124 - accuracy: 0.5124 - val_loss: 0.9854 - val_f1_score: 0.5076 -
val accuracy: 0.5076
Epoch 466/500
f1_score: 0.5146 - accuracy: 0.5146 - val_loss: 0.9800 - val_f1_score: 0.5172 -
val_accuracy: 0.5172
Epoch 467/500
f1 score: 0.5089 - accuracy: 0.5089 - val_loss: 0.9868 - val_f1_score: 0.4968 -
val_accuracy: 0.4968
Epoch 468/500
f1_score: 0.5093 - accuracy: 0.5093 - val_loss: 0.9828 - val_f1_score: 0.5134 -
val_accuracy: 0.5134
Epoch 469/500
```

```
f1_score: 0.5090 - accuracy: 0.5090 - val_loss: 0.9848 - val_f1_score: 0.5070 -
val_accuracy: 0.5070
Epoch 470/500
f1_score: 0.5128 - accuracy: 0.5128 - val_loss: 0.9820 - val_f1_score: 0.5172 -
val accuracy: 0.5172
Epoch 471/500
f1_score: 0.5152 - accuracy: 0.5152 - val_loss: 0.9892 - val_f1_score: 0.5134 -
val_accuracy: 0.5134
Epoch 472/500
f1 score: 0.5145 - accuracy: 0.5145 - val_loss: 0.9989 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 473/500
f1 score: 0.5054 - accuracy: 0.5054 - val_loss: 1.0037 - val_f1_score: 0.5013 -
val_accuracy: 0.5013
Epoch 474/500
f1_score: 0.5110 - accuracy: 0.5110 - val_loss: 0.9921 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 475/500
f1 score: 0.5092 - accuracy: 0.5092 - val_loss: 0.9902 - val_f1_score: 0.5064 -
val_accuracy: 0.5064
Epoch 476/500
f1_score: 0.5140 - accuracy: 0.5140 - val_loss: 0.9849 - val_f1_score: 0.5013 -
val_accuracy: 0.5013
Epoch 477/500
f1_score: 0.5149 - accuracy: 0.5149 - val_loss: 1.0084 - val_f1_score: 0.5045 -
val accuracy: 0.5045
Epoch 478/500
f1_score: 0.5216 - accuracy: 0.5216 - val_loss: 0.9815 - val_f1_score: 0.5089 -
val_accuracy: 0.5089
Epoch 479/500
f1 score: 0.5135 - accuracy: 0.5135 - val_loss: 0.9874 - val_f1_score: 0.5038 -
val_accuracy: 0.5038
Epoch 480/500
442/442 [=========== ] - 1s 2ms/step - loss: 0.9621 -
f1_score: 0.5135 - accuracy: 0.5135 - val_loss: 0.9824 - val_f1_score: 0.4816 -
val_accuracy: 0.4816
Epoch 481/500
```

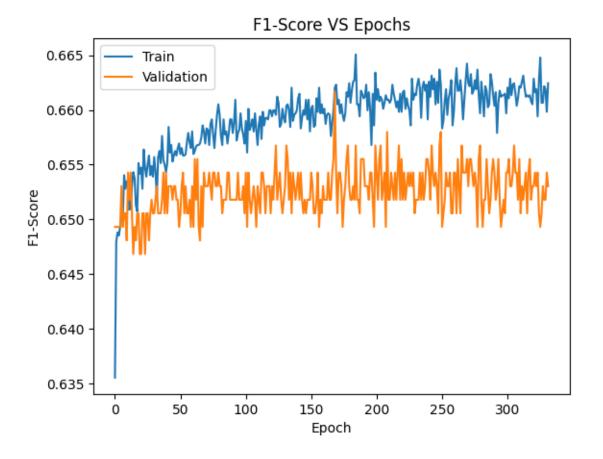
```
f1_score: 0.5093 - accuracy: 0.5093 - val_loss: 0.9840 - val_f1_score: 0.4796 -
val_accuracy: 0.4796
Epoch 482/500
f1_score: 0.5144 - accuracy: 0.5144 - val_loss: 0.9963 - val_f1_score: 0.4930 -
val accuracy: 0.4930
Epoch 483/500
f1_score: 0.5148 - accuracy: 0.5148 - val_loss: 0.9881 - val_f1_score: 0.4962 -
val_accuracy: 0.4962
Epoch 484/500
f1 score: 0.5055 - accuracy: 0.5055 - val_loss: 1.0077 - val_f1_score: 0.5045 -
val_accuracy: 0.5045
Epoch 485/500
f1 score: 0.5077 - accuracy: 0.5077 - val_loss: 0.9974 - val_f1_score: 0.4987 -
val_accuracy: 0.4987
Epoch 486/500
f1_score: 0.5083 - accuracy: 0.5083 - val_loss: 0.9819 - val_f1_score: 0.5095 -
val_accuracy: 0.5095
Epoch 487/500
f1 score: 0.5183 - accuracy: 0.5183 - val_loss: 0.9812 - val_f1_score: 0.5057 -
val_accuracy: 0.5057
Epoch 488/500
f1_score: 0.5109 - accuracy: 0.5109 - val_loss: 0.9820 - val_f1_score: 0.5051 -
val_accuracy: 0.5051
Epoch 489/500
f1_score: 0.5135 - accuracy: 0.5135 - val_loss: 0.9828 - val_f1_score: 0.4987 -
val accuracy: 0.4987
Epoch 490/500
f1_score: 0.5071 - accuracy: 0.5071 - val_loss: 0.9875 - val_f1_score: 0.4949 -
val_accuracy: 0.4949
Epoch 491/500
f1 score: 0.5130 - accuracy: 0.5130 - val_loss: 0.9800 - val_f1_score: 0.5089 -
val_accuracy: 0.5089
Epoch 492/500
f1_score: 0.5138 - accuracy: 0.5138 - val_loss: 1.0027 - val_f1_score: 0.4943 -
val_accuracy: 0.4943
Epoch 493/500
```

```
f1_score: 0.5073 - accuracy: 0.5073 - val_loss: 0.9896 - val_f1_score: 0.5178 -
    val_accuracy: 0.5178
    Epoch 494/500
    f1_score: 0.5153 - accuracy: 0.5153 - val_loss: 1.0072 - val_f1_score: 0.4968 -
    val accuracy: 0.4968
    Epoch 495/500
    f1_score: 0.4999 - accuracy: 0.4999 - val_loss: 0.9941 - val_f1_score: 0.4987 -
    val_accuracy: 0.4987
    Epoch 496/500
    f1 score: 0.5091 - accuracy: 0.5091 - val_loss: 0.9907 - val_f1_score: 0.4981 -
    val_accuracy: 0.4981
    Epoch 497/500
    f1 score: 0.5015 - accuracy: 0.5015 - val_loss: 0.9933 - val_f1_score: 0.5083 -
    val_accuracy: 0.5083
    Epoch 498/500
    f1_score: 0.5127 - accuracy: 0.5127 - val_loss: 0.9874 - val_f1_score: 0.5312 -
    val_accuracy: 0.5312
    Epoch 499/500
    f1 score: 0.5081 - accuracy: 0.5081 - val_loss: 0.9777 - val_f1_score: 0.5159 -
    val_accuracy: 0.5159
    Epoch 500/500
    f1_score: 0.5117 - accuracy: 0.5117 - val_loss: 0.9949 - val_f1_score: 0.5000 -
    val_accuracy: 0.5000
[259]: pred_smote, test_smote, y_pred_smote, y_test_smote, X_test_smote = ___
     →test_pipeline(model_smote)
    dropping ID column
    dropping ['ID', 'Var_1', 'Age']...
    replacing numerical nans with mode...
    replacing categorical nans with None string...
    label encoding categorical data...
    SUCCESSFULLY PERFORMED PREPROCESSING
    83/83 [=======] - Os 477us/step
[272]: # Classification metrics for Test
     print(f'f1_score for the test data is:', f1_score(pred_smote, test_smote, ⊔
     ⇔average='macro'))
```

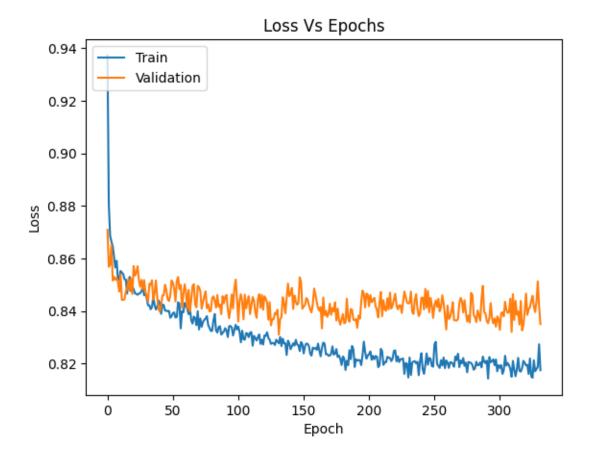
f1_score for the test data is: 0.42865639531183425 recall for the test data is: 0.4319648903554185 precision for the test data is: 0.45932621682535535 accuracy for the test data is: 0.49105443471640653

f1_score, and precision increased compare to previous data which means that number of true positive incrased and number of false positives decreased. The model is no longer identifying what belongs to class B and C as Class A.This makes sense as we increased number of minority classes and made a balanced dataset. Now that recall and precision is closer to each other and greater than 0.4 the Harmonic mean (f1_score) has also increased significantly from 0.28 to 0.42. If we had larger dataset to better trian our model, we would have good metrics.

```
[261]: plt.plot(history_smote.history['f1_score'])
   plt.plot(history_smote.history['val_f1_score'])
   plt.title('F1-Score VS Epochs')
   plt.ylabel('F1-Score')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Validation'], loc='upper left')
   plt.show()
```



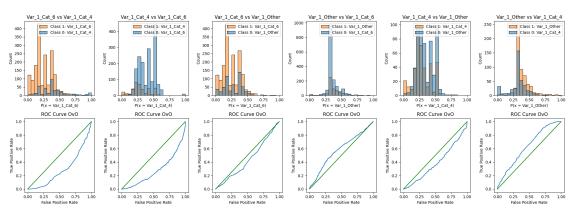
```
[262]: plt.plot(history_smote.history['loss'])
   plt.plot(history_smote.history['val_loss'])
   plt.title('Loss Vs Epochs')
   plt.ylabel('Loss')
   plt.xlabel('Epoch')
   plt.legend(['Train', 'Validation'], loc='upper left')
   plt.show()
```



Similiar to previous situation, model is learning and loss is decreasing while f1_score is increasing. We would have achieved a better result if we had more data and trained with a more complicated (more layers and neurons) model.

[263]: plot_roc_auc(test_smote, y_pred_smote)

[['Var_1_Cat_6', 'Var_1_Cat_4'], ['Var_1_Cat_4', 'Var_1_Cat_6'], ['Var_1_Cat_6',
'Var_1_Other'], ['Var_1_Other', 'Var_1_Cat_6'], ['Var_1_Cat_4', 'Var_1_Other'],
['Var_1_Other', 'Var_1_Cat_4']]



While overall our model is not a good model but after SMOTE you can see that because we increased number of Cat_4 and Other classes we are getting slightly better ROC curve (more area under the curve) for Other vs Cat_6, Cat_6 vs Other, Other vs Cat_4 and Cat_4 vs Other.