

Configuration Manual

MSc Research Project Data Analytics

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MSc Project Submission Sheet



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Configuration Manual

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1 Introduction

This configuration handbook provides information on the system configuration, software, and hardware requirements, as well as the steps necessary to complete the Research Project: Application of Data Analytics to Suggest Gifts to Retail Customers Based on their Emotions.

Software and hardware specifications are covered in Section 2 of this document. The environment setup, data collection, data preparation, and the libraries to be imported are covered in Section 3. The different steps taken for the image processing dataset are explained in Section 4. In Section 5, the model design for both the image dataset and the gift dataset is discussed. In the last section (Section 6) all the steps taken for the final Testing and Deployment stage have been shown.

2 System Configuration

The hardware and Software specifications required for this project have mentioned below in this section.

2.1 Hardware Requirements

Operating System	Windows 10
RAM	8 GB
Hard Disc Space	250 GB
CPU	64 bit x 86 multi-core

2.2 Soft Requirements

Table 2: So	ftware Req	uirements
-------------	------------	-----------

Programming Language Tools	Jupyter Notebook
Web Browser	Google Chrome or Mozilla Firefox
Other Software	Microsoft Excel, Microsoft Word, and Overleaf

3 Project Development

In this section, information related to the environment setup, data collection, data preparation, and the libraries to be imported are covered.

3.1 Environmental Setup

Jupyter Notebook has been used to carry out the implementation part of the project.

💭 jupyter	Quit Log
Files Running Clusters	
elect items to perform actions on them.	Upload New -
0 - IRESEARCH PROJECT FER	Name Last Modified File s
C	seconds ago
C C FER	2 months ago
Ca models	6 days ago
the process	7 days ago
DT_gifts (1).ipynb	7 days ago 2.83
🗌 🖉 DT_gifts.ipynb	6 days ago 5.14
ER_uning_Normal_ML.ipynb	a month ago 49.
Finla preds.ipynb	3 days ago 10.
B Res_Net_trial_1 (1).ipynb	7 days ago 38
B Res_Net_trial_1.ipynb	25 days ago 39
B RES_NET_Trial_2.ipynb	6 days ago 24
Z Tryout_CNN.ipynb	7 days ago 28
MOCK_DATA_1.csv	20 days ago 43.
MOCK_DATA_2.csv	20 days ago 43.
MOCK_DATA_3.csv	20 days ago 43.
MOCK_DATA_4.csv	20 days ago 43.
MOCK_DATA_5.csv	20 days ago 45.
model_weights.h5	a month ago 53.9
	4E dava and 77

Figure 1: Jupyter Notebook

3.2 Data Collection

The Image dataset used in this project is FER2013. The dataset has been taken from Kaggle1. It is a reliable data source that is also open to the public. In the form of ".jpg" files, image data is collected.



Figure 2: Data Source- Kaggle

¹ https://www.kaggle.com/datasets/msambare/fer2013

3.3 Data Preparation

The method used to prepare the data is described in this section. Training, Validation, and Testing are the three folders into which the data is separated and saved in the directory.

```
    This PC > Windows (C:) > Users > SAYAN GHOSH > RESEARCH PROJECT FER > FER > images > images
    Name
    Date modified
    Type
    Size
    Test
    09-12-2022 16:25
    File folder
    train
    14-10-2022 16:27
    File folder
    validation
    14-10-2022 16:27
    File folder
```

Figure 3: Data in Directory

3.4 Importing Libraries

The main libraries required for this project are NumPy, Seaborn, Matplotlib, Tensorflow, Keras, sklearn, and Pandas.

```
import numpy as np
import seaborn as sns
# from keras.preprocessing.image import img_to_array
from tensorflow.keras.utils import load_img,img_to_array
import matplotlib.pyplot as plt
import os
%matplotlib inline
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import os
import pprint
import pandas as pd
from sklearn.preprocessing import LabelEncoder
# pp = pprint.PrettyPrinter(indent=4)
```

Figure 4: Imported Libraries

4 Image Processing

All the steps taken for pre-processing and organizing the image data are shown in this section.

4.1 Pre-processing and augmentation for ResNet50 Model

```
# building data generator
from keras.preprocessing.image import ImageDataGenerator
batch_size = 128
base_path = r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\FER\images\images'
train_datagen = ImageDataGenerator(rescale = 1.0/255.0,
                                  width_shift_range = 0.1,
                                   height_shift_range = 0.1,
                                   rotation_range = 20,
                                   horizontal_flip = True)
validation_datagen = ImageDataGenerator(rescale= 1.0/255)
train_generator = train_datagen.flow_from_directory(base_path + "/train",
                                                    target_size=(56,56),
                                                    color_mode="rgb",
                                                    batch_size=batch_size,
                                                    class_mode='categorical',
                                                    shuffle=True)
validation_generator = validation_datagen.flow_from_directory(base_path + "/validation",
                                                    target_size=(56,56),
                                                    color_mode="rgb",
                                                    batch size=batch size,
                                                    class_mode='categorical',
                                                    shuffle=False)
Found 28821 images belonging to 7 classes.
Found 7066 images belonging to 7 classes.
from collections import Counter
counter = Counter(train_generator.classes)
max_val = float(max(counter.values()))
class_weights = {class_id : max_val/num_images for class_id, num_images in counter.items()}
```

Figure 5: Pre-processing for ResNet50 Model

4.2 Pre-processing and image augmentation for Convolutional Neural Network (CNN) Model

Image augmentation using keras ImageDataGenerator	
# building data generator	
<pre>from keras.preprocessing.image import ImageDataGenerator</pre>	
<pre>batch_size = 128 base_path = r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\FER\images\images'</pre>	
<pre>train_datagen = ImageDataGenerator(rescale = 1.0/255.0,</pre>	
validation_datagen = ImageDataGenerator(rescale= 1.0/255)	
<pre>train_generator = train_datagen.flow_from_directory(base_path + "/train",</pre>	
<pre>validation_generator = validation_datagen.flow_from_directory(base_path + "/val target_size=(56,56), color_mode="grayscale", batch_size=batch_size, class_mode='categorical', shuffle=False)</pre>	idation",

Figure 6: Pre-processing for CNN Model

4.3 Pre-processing and image augmentation for Random Forest Model (Used for Image Processing)

```
import joblib
from skimage.io import imread
from skimage.transform import resize
def resize_all(src, include, width=150, height=None):
   load images from path, resize them and write them as arrays to a dictionary, together with labels and metadata. The dictionary is written to a pickle file named '{pklname}_{width}x{height}px.pkl'.
    Parameter
    src: str
         path to data
    pklname: str
        path to output file
    width: int
        target width of the image in pixels
    include: set[str]
    set containing str
   height = height if height is not None else width
    data = dict()
    data['description'] = 'resized ({0}x{1})FER images in grayscale'.format(int(width), int(height))
    data['label'] = []
data['filename'] = []
    data['data'] = []
      pklname = f"{pklname}_{width}x{height}px.pkl"
    # read all images in PATH, resize and write to DESTINATION_PATH
    for subdir in os.listdir(src):
        if subdir in include:
            print(subdir)
             current_path = os.path.join(src, subdir)
             for file in os.listdir(current_path):
                  if file[-3:] in {'jpg', 'png'}:
im = imread(os.path.join(current_path, file))
im = resize(im, (width, height)) #[:,:,::-1]
#
                  data['label'].append(subdir)
data['filename'].append(file)
                 data['data'].append(im.flatten())
   return data
#
           joblib.dump(data, pklname)
data_path = r'C:\\Users\\SAYAN GHOSH\\RESEARCH PROJECT FER\\FER\\images\\images\\train'
os.listdir(data_path)
['angry', 'disgust', 'fear', 'happy', 'neutral', 'sad', 'surprise']
# base_name = 'emots'
width = 80
include = {'angry', 'disgust', 'fear', 'happy', 'neutral', 'sad', 'surprise'}
data = resize_all(src=data_path, width=width, include=include)
angry
disgust
fear
happy
neutral
sad
surprise
# base name = 'emotions
val_path = r'C:\\Users\\SAYAN GHOSH\\RESEARCH PROJECT FER\\FER\\images\\images\\validation'
width = 80
include = {'angry', 'disgust', 'fear', 'happy', 'neutral', 'sad', 'surprise'}
data_val = resize_all(src=val_path, width=width, include=include)
```

```
from collections import Counter
# data = joblib.load(f'{base_name}_{width}x{width}px.pkl')
print('number of samples: ', len(data['data']))
print('keys: ', list(data.keys()))
print('description: ', data['description'])
print('image shape: ', data['data'][0].shape)
print('labels:', np.unique(data['label']))
Counter(data['label'])
number of samples: 28821
keys: ['description', 'label', 'filename', 'data']
description: resized (80x80)FER images in grayscale
image shape: (6400,)
labels: ['angry' 'disgust' 'fear' 'happy' 'neutral' 'sad' 'surprise']
Counter({'angry': 3993,
'disgust': 436,
'fear': 4103,
'happy': 7164,
'neutral': 4982,
            'sad': 4938,
'surprise': 3205})
X = np.array(data['data'])
y = np.array(data['label'])
# data['label']
X_val = np.array(data_val['data'])
y_val = np.array(data_val['label'])
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(
     Х,
     y,
test_size=0.3,
     shuffle=True,
     random_state=69,
from sklearn.base import BaseEstimator, TransformerMixin
class RGB2GrayTransformer(BaseEstimator, TransformerMixin):
     Convert an array of RGB images to grayscale
     def __init__(self):
           pass
     def fit(self, X, y=None):
            """returns itself""
           return self
     def transform(self, X, y=None):
    """perform the transformation and return an array"""
           return np.array([skimage.color.rgb2gray(img) for img in X])
```

Figure 7: Pre-processing for Random Forest Model

4.4 Pre-processing for the Gift dataset

from sklearn.preprocessing import OneHotEncoder		
<pre>y_train = df['Gift Categories'] y_test = df_test['Gift Categories']</pre>		
<pre>OHE = OneHotEncoder(categories='auto', # Categories per feature drop=None, # Whether to drop one of the features sparse=True, # Will return sparse matrix if set True handle_unknown='error' # Whether to raise an error) OHE.fit(df[['Gender']]) transformed = OHE.transform(df[['Gender']]) transformed_tst = OHE.transform(df_test[['Gender']]) df[OHE.categories_[0]] = transformed.toarray() df_test[OHE.categories_[0]] = transformed_tst.toarray()</pre>		
<pre>import pickle with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\OHE_Gend', "wb") as f: pickle.dump(OHE,f)</pre>		
<pre>OHE = OneHotEncoder(categories='auto', # Categories per feature drop=None, # Whether to drop one of the features sparse=True, # Will return sparse matrix if set True handle_unknown='error' # Whether to raise an error) OHE.fit(df[['Occasion']]) transformed = OHE.transform(df[['Occasion']]) transformed_tst = OHE.transform(df_test[['Occasion']]) df[OHE.categories_[0]] = transformed.toarray() df tost/OHE categories_[0]] = transformed.toarray()</pre>		
<pre>import pickle with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\OHE_OCC', "wb") as f: pickle.dump(OHE,f)</pre>		
<pre>le = LabelEncoder() le.fit(df['Emotions']) # le.classes_ df['Emotions'] = le.transform(df['Emotions']) df_test['Emotions'] = df.Emotions.astype('category') df['Emotions'] = df_test.Emotions.astype('category')</pre>		
<pre>import pickle with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\le_Emo', "wb") as f: pickle.dump(le,f)</pre>		
<pre>le_y = LabelEncoder() le_y.fit(y_train) y_train = le_y.transform(y_train) y_test = le_y.transform(y_test)</pre>		
<pre>import pickle with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\le_op', "wb") as f: pickle.dump(le_y,f)</pre>		
<pre>df.drop(columns=['Gender','Occasion','Age','Gift Categories'],inplace=True) df_test.drop(columns=['Gender','Occasion','Age','Gift Categories'],inplace=True)</pre>		
df.dtypes		
Emotions category Price (Euros) int64 Female float64 Male float64 Casual float64 Festivals float64 Special Events float64 dtype: object		

Figure 8: Pre-processing for the Gift Dataset

5 Model Design

5.1 Modelling for Image Dataset

5.1.1 ResNet50 Model

```
# Create and compile model
from tensorflow.keras.applications import ResNet50
from keras.models import Model, Sequential
from keras.layers import Dense, Input, Dropout, GlobalMaxPooling2D, Flatten, Conv2D, BatchNormalization, Activation, MaxPooling2[
from keras.optimizers import Adam
model = Sequential()
model.add(ResNet50(input_shape=(56, 56, 3), weights='imagenet', =False))
model.add(GlobalMaxPooling2D())
model.add(Dropout(0.5))
model.add(Dense(7,activation='softmax'))
from tensorflow.keras.utils import plot_model
model.compile(optimizer=Adam(learning_rate=1e-4), loss='binary_crossentropy', metrics=['accuracy'])
model.summary()
history = model.fit(train_generator,
validation_data = validation_generator,
                      class_weight=class_weights,
                      epochs = 35,
                     steps_per_epoch=train_generator.n//train_generator.batch_size,
validation_steps = validation_generator.n//validation_generator.batch_size)
```

Figure 9: Model Building for ResNet50

5.1.2 CNN Model

```
from keras.layers import Dense, Input, Dropout, GlobalAveragePooling2D, Flatten, Conv2D, BatchNormalization, Activation, MaxPool
from keras.models import Model, Sequential
from keras.optimizers import Adam
# number of possible label values
nb_classes = 7
# Initialising the CNN
model = Sequential()
# 1 - Convolution
model.add(Conv2D(64,(3,3), padding='same', input_shape=(56, 56,1)))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
# 2nd Convolution layer
model.add(Conv2D(128,(5,5), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
# 3rd Convolution layer
model.add(Conv2D(512,(3,3), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
# 4th Convolution layer
model.add(Conv2D(512,(3,3), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.25))
```



Figure 10: Model Building for CNN

5.1.3 Random Forest Model

```
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import cross_val_predict
from sklearn.preprocessing import StandardScaler, Normalizer
import skimage
# create an instance of each transformer
grayify = RGB2GrayTransformer()
rfc = RandomForestClassifier(n_estimators = 300)
# call fit_transform on each transform converting X_train step by step
# X_train_gray = grayify.fit_transform(X_train_1)
scalify = StandardScaler()
X_train_prepared = scalify.fit_transform(X_train)
rfc.fit(X_train_prepared, y_train)
# print(X_train_prepared, shape)
```

Figure 11: Model Building for Random Forest (Image processing)

5.2 Modelling for Gift Dataset

5.2.1 Random forest Model

```
from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier(n_estimators = 500,random_state = 69)
rfc.fit(df, y_train)
RandomForestClassifier(n_estimators=500, random_state=69)
import pickle
with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\models\RandF', "wb") as f:
    pickle.dump(rfc,f)

y_pred = rfc.predict(df)
print('Train Acc: ', 100*np.sum(y_pred == y_train)/len(y_train))
y_pred = rfc.predict(df_test)
print('Validation Acc: ', 100*np.sum(y_pred == y_test)/len(y_test))
```

Figure 12: Model Building for Random Forest (Gift Dataset)

5.2.2 Decision Tree Model

Trying Decission Tree clf = DecisionTreeClassifier(random_state = 70) max depth =10 clf.fit(df, y_train) DecisionTreeClassifier(random_state=70) import pickle with open(r^C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\models\DTree', "wb") as f: pickle.dump(clf,f) from sklearn import tree import matplotlib.pyplot as plt plt.figure(figsize=(50,60)) a = tree.plot_tree(clf, feature_names = df.columns, rounded = True, filled = True, fontsize=14) plt.show() y_pred = clf.predict(df) y_h(i' Train Acc: ', 100*np.sum(y_pred == y_train)/len(y_train)) y_pred = clf.predict(df_test) print('Validation Acc: ', 100*np.sum(y_pred == y_test)/len(y_test))

Figure 13: Model Building for Decision Tree

5.2.3 Logistic Regression Model

Trying logistic regression

```
from sklearn.tree import DecisionTreeClassifier
from sklearn.linear_model import LogisticRegression
LR = LogisticRegression(multi_class= 'multinomial', solver = 'newton-cg')
LR.fit(df, y_train)
LogisticRegression(multi_class='multinomial', solver='newton-cg')
import pickle
with open(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\models\Logit', "wb") as f:
    pickle.dump(LR,f)
y_pred = LR.predict(df)
print('Train Acc: ', 100*np.sum(y_pred == y_train)/len(y_train))
y_pred = LR.predict(df_test)
print('Validation Acc: ', 100*np.sum(y_pred == y_test)/len(y_test))
```

Figure 14: Model Building for Logistic Regression

6 Testing & Deployment

def	<pre>DL_Pred(batch_size = 128, path = r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\FER\images\images\\',model_path = r'C:\Users\SA\ bat_size = batch_size base path = path</pre>
	test datagen = ImageDataGenerator(rescale= 1.0/255)
	test generator = test datagen.flow from directory(base path + "Test",
	color_mode="rgb",
	batch_size=batch_size,
	<pre>class_mode='categorical',</pre>
	shuffle=False)
	# Label_map
	<pre>label_map = {0: 'Angry',1: 'Disgust',2: 'Fear',3: 'Happy',4: 'Neutral',5: 'Sad',6: 'Surprise'}</pre>
	RNN = DL_model(model_path)
	preds = RNN.predict(test_generator)
	<pre>pp = preds.argmax(axis=-1)</pre>
	cls = []
	for 1 in pp:
	cis.append(label_map[1])
def	Pickle file loader(file name):
	file = open(file name, "rb")
	<pre>pkl = pickle.load(file)</pre>
	return pkl
def	gift_data_prep(Price, Gender, Occasion):
	df = pd.DataFrame({'Price (Euros)':Price.split(','), 'Gender':Gender.split(','), 'Occasion':Occasion.split(','), 'Emotions':[
	OHE_Gend = Pickle_file_loader(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\OHE_Gend')
	<pre>transformed = OHE_Gend.transform(df[['Gender']])</pre>
	df[OHE_Gend.categories_[0]] = transformed.toarray()
	OHE_OCC = Pickle_file_loader(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\OHE_OCC')
	transformed = OHE_OCC.transform(df[['Occasion']])
	df[OHE_OCC.categories_[0]] = transformed.toarray()
	<pre>le_Emo = Pickle_tile_Loader(r'C:\Users\SAYAN GHOSH\RESEARCH PROJECT FER\pre_process\le_Emo')</pre>
	dt['Emotions'] = le_Emo.transform(dt[('Emotions']])
	dt['Emotions'] = dt.Emotions.astype('category')
	at price (curos) j = at price (curos) .astype(int)
	al.arop(columns=[dender', occasion],inplace=rrue)
gift	: data prep(Price = input('Enter Price:'), Gender = input("Enter Gender (Male\Female):"), Occasion = input("Enter Occasion:"))

Figure 15: Testing & Deployment