sna-project1

June 16, 2023

1 Image labeling on social network metadata

Package import

[ ]:
```
import tensorflow as tf
from tensorflow.keras import Sequential
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.preprocessing import image
from tensorflow.keras.layers import Flatten, Dense, Dropout,␣
↪BatchNormalization, Conv2D, MaxPool2D
print(tf.__version__)
```

```
2.12.0
```

[ ]:
```
import numpy as np
import pandas as pd
from tqdm import tqdm
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
```

Dataset import

[ ]:
```
data = pd.read_csv('/content/drive/MyDrive/Movies-Poster_Dataset-master/train.
↪csv')
data.shape
```

[ ]:
```
(99, 27)
```

[ ]:
```
from google.colab import drive
drive.mount('/content/drive')
```

```
Drive already mounted at /content/drive; to attempt to forcibly remount, call
drive.mount("/content/drive", force_remount=True).
```

Tags overview

[ ]:
```
data.head()
```

[ ]:

	Id	Genre	Action	Adventure	\
0	tt0086425	['Comedy', 'Drama']	0	0	
1	tt0085549	['Drama', 'Romance', 'Music']	0	0	
2	tt0086465	['Comedy']	0	0	
3	tt0086567	['Sci-Fi', 'Thriller']	0	0	
4	tt0086034	['Action', 'Adventure', 'Thriller']	1	1	

	Animation	Biography	Comedy	Crime	Documentary	Drama	…	N/A	News	\
0	0	0	1	0	0	1	…	0	0	
1	0	0	0	0	0	1	…	0	0	
2	0	0	1	0	0	0	…	0	0	
3	0	0	0	0	0	0	…	0	0	
4	0	0	0	0	0	0	…	0	0	

	Reality-TV	Romance	Sci-Fi	Short	Sport	Thriller	War	Western
0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	0	0	0
2	0	0	0	0	0	0	0	0
3	0	0	1	0	0	1	0	0
4	0	0	0	0	0	1	0	0

[5 rows x 27 columns]

Processing images in batch

[ ]:
```
img_width = 350
img_height = 350

X = []

for i in tqdm(range(data.shape[0])):
  path = '/content/drive/MyDrive/Movies-Poster_Dataset-master/Images/' +␣
 ↪data['Id'][i] + '.jpg'
  img = image.load_img(path, target_size=(img_width, img_height, 3))
  img = image.img_to_array(img)
  img = img/255.0
  X.append(img)

X = np.array(X)
```

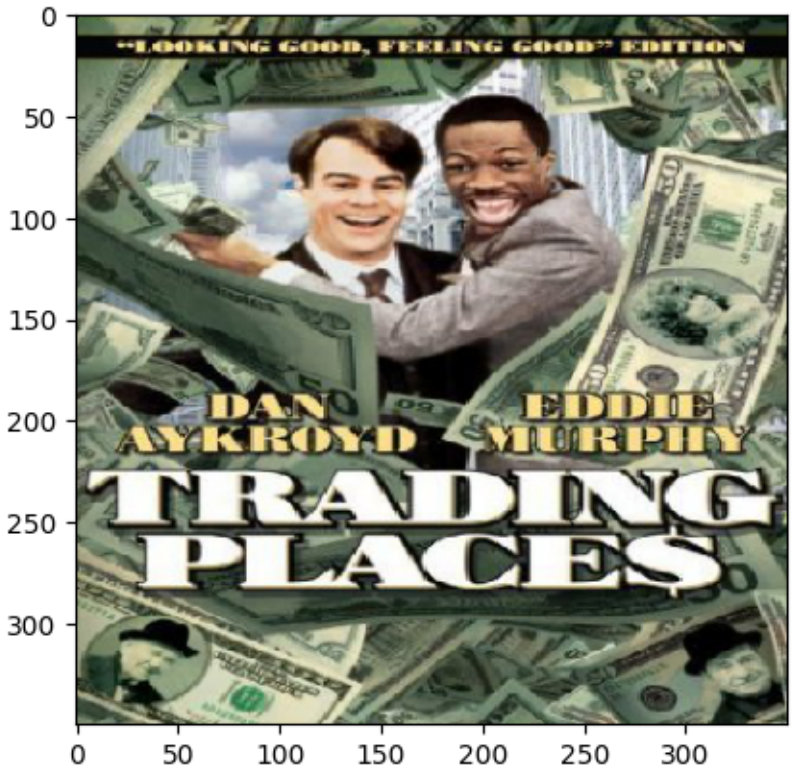
100%| | 99/99 [00:00<00:00, 163.72it/s]

[ ]:
```
X.shape
```

[ ]: (99, 350, 350, 3)

[ ]:
```
plt.imshow(X[2])
```

[ ]: <matplotlib.image.AxesImage at 0x7f8b3aff8460>



Labelling image

[ ]:
```
data['Genre'][2]
```

[ ]: "['Comedy']"

[ ]:
```
y = data.drop(['Id', 'Genre'], axis = 1)
y = y.to_numpy()
y.shape
```

[ ]: (99, 25)

Splitting dataset

[ ]:
```
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state = 0,␣
↪test_size = 0.15)
```

[ ]:
```
X_train[0].shape
```

[ ]: (350, 350, 3)

Building CNN network

[ ]:
```
model = Sequential()
model.add(Conv2D(16, (3,3), activation='relu', input_shape = X_train[0].shape))
model.add(BatchNormalization())
model.add(MaxPool2D(2,2))
model.add(Dropout(0.3))

model.add(Conv2D(32, (3,3), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPool2D(2,2))
model.add(Dropout(0.3))

model.add(Conv2D(64, (3,3), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPool2D(2,2))
model.add(Dropout(0.4))

model.add(Conv2D(128, (3,3), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPool2D(2,2))
model.add(Dropout(0.5))

model.add(Flatten())

model.add(Dense(128, activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.5))


model.add(Dense(128, activation='relu'))
model.add(BatchNormalization())
model.add(Dropout(0.5))


model.add(Dense(25, activation='sigmoid'))
```

Parameter summary

[ ]:
```
model.summary()
```

```
Model: "sequential_1"
_________________________________________________________________
 Layer (type)                Output Shape              Param #
=================================================================
 conv2d_4 (Conv2D)           (None, 348, 348, 16)      448

 batch_normalization_6 (Batc  (None, 348, 348, 16)     64
```

```
hNormalization)

max_pooling2d_4 (MaxPooling  (None, 174, 174, 16)     0
2D)

dropout_6 (Dropout)          (None, 174, 174, 16)     0

conv2d_5 (Conv2D)            (None, 172, 172, 32)     4640

batch_normalization_7 (Batc  (None, 172, 172, 32)     128
hNormalization)

max_pooling2d_5 (MaxPooling  (None, 86, 86, 32)       0
2D)

dropout_7 (Dropout)          (None, 86, 86, 32)       0

conv2d_6 (Conv2D)            (None, 84, 84, 64)       18496

batch_normalization_8 (Batc  (None, 84, 84, 64)       256
hNormalization)

max_pooling2d_6 (MaxPooling  (None, 42, 42, 64)       0
2D)

dropout_8 (Dropout)          (None, 42, 42, 64)       0

conv2d_7 (Conv2D)            (None, 40, 40, 128)      73856

batch_normalization_9 (Batc  (None, 40, 40, 128)      512
hNormalization)

max_pooling2d_7 (MaxPooling  (None, 20, 20, 128)      0
2D)

dropout_9 (Dropout)          (None, 20, 20, 128)      0

flatten_1 (Flatten)          (None, 51200)            0

dense_3 (Dense)              (None, 128)              6553728

batch_normalization_10 (Bat  (None, 128)              512
chNormalization)

dropout_10 (Dropout)         (None, 128)              0

dense_4 (Dense)              (None, 128)              16512
```

```
 batch_normalization_11 (Bat  (None, 128)              512
 chNormalization)

 dropout_11 (Dropout)        (None, 128)              0

 dense_5 (Dense)             (None, 25)               3225

=================================================================
Total params: 6,672,889
Trainable params: 6,671,897
Non-trainable params: 992
_________________________________________________________________
```

Training dataset

[ ]:
```
model.compile(optimizer='adam', loss = 'binary_crossentropy',␣
 ↪metrics=['accuracy'])
```

[ ]:
```
history = model.fit(X_train, y_train, epochs=5, validation_data=(X_test,␣
 ↪y_test))
```

```
Epoch 1/5
3/3 [==============================] - 25s 7s/step - loss: 0.9768 - accuracy:
0.0833 - val_loss: 0.6755 - val_accuracy: 0.4667
Epoch 2/5
3/3 [==============================] - 20s 6s/step - loss: 0.9406 - accuracy:
0.0595 - val_loss: 0.6901 - val_accuracy: 0.0000e+00
Epoch 3/5
3/3 [==============================] - 20s 6s/step - loss: 0.9265 - accuracy:
0.0833 - val_loss: 0.7903 - val_accuracy: 0.0000e+00
Epoch 4/5
3/3 [==============================] - 23s 7s/step - loss: 0.9340 - accuracy:
0.0357 - val_loss: 0.9271 - val_accuracy: 0.0000e+00
Epoch 5/5
3/3 [==============================] - 19s 7s/step - loss: 0.8965 - accuracy:
0.0952 - val_loss: 1.0717 - val_accuracy: 0.0000e+00
```

Plot for training and validation loss

[ ]:
```
def plot_learningCurve(history, epoch):
  epoch_range = range(1, epoch+1)

  plt.plot(epoch_range, history.history['loss'])
  plt.plot(epoch_range, history.history['val_loss'])
  plt.title('Model loss')
  plt.ylabel('Loss')
  plt.xlabel('Epoch')
  plt.legend(['Train', 'Val'], loc='upper left')
  plt.show()
```

```
plot_learningCurve(history, 5)
```

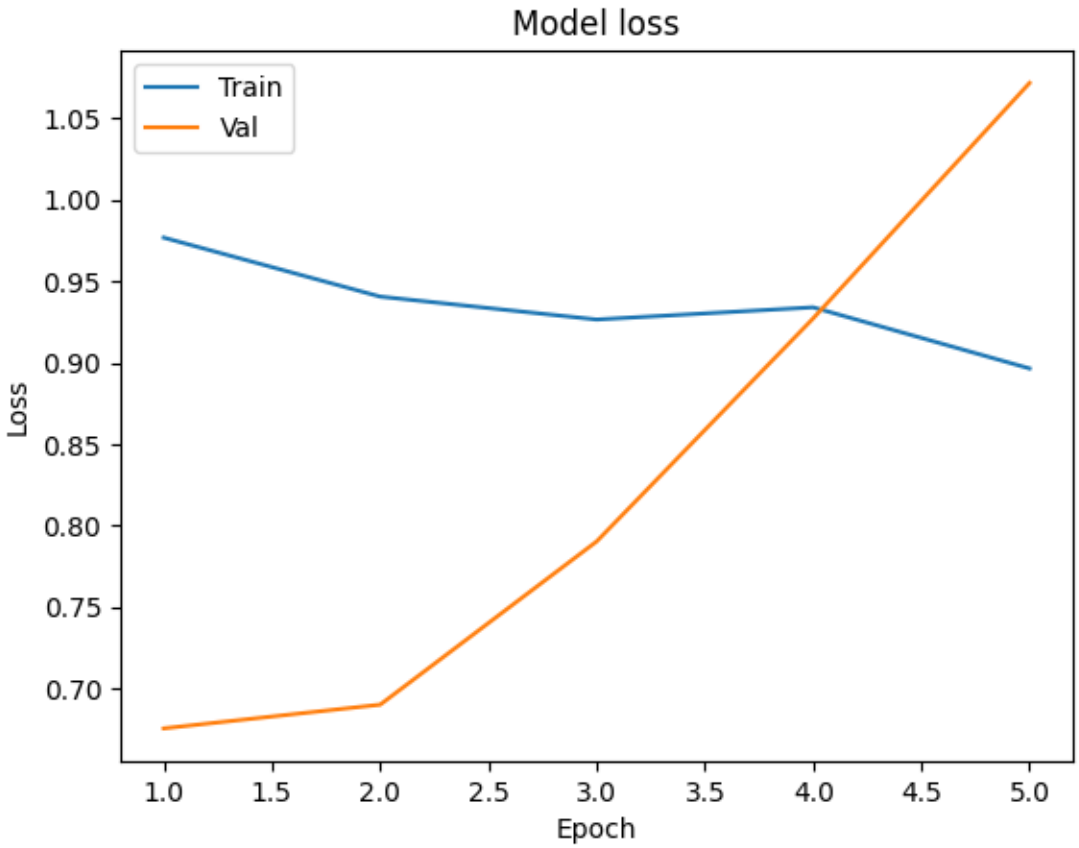


Image labelling on sample movie poster

[ ]:
```
img = image.load_img('free.jpg', target_size=(img_width, img_height, 3))
plt.imshow(img)
img = image.img_to_array(img)
img = img/255.0

img = img.reshape(1, img_width, img_height, 3)

classes = data.columns[2:]
print(classes)
y_prob = model.predict(img)
top3 = np.argsort(y_prob[0])[:-4:-1]

for i in range(3):
  print(classes[top3[i]])
```

```
Index(['Action', 'Adventure', 'Animation', 'Biography', 'Comedy', 'Crime',
       'Documentary', 'Drama', 'Family', 'Fantasy', 'History', 'Horror',
       'Music', 'Musical', 'Mystery', 'N/A', 'News', 'Reality-TV', 'Romance',
       'Sci-Fi', 'Short', 'Sport', 'Thriller', 'War', 'Western'],
      dtype='object')
1/1 [==============================] - 1s 929ms/step
N/A
Western
Crime
```