

sna-project

May 23, 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
```

(10, 27)

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)
```

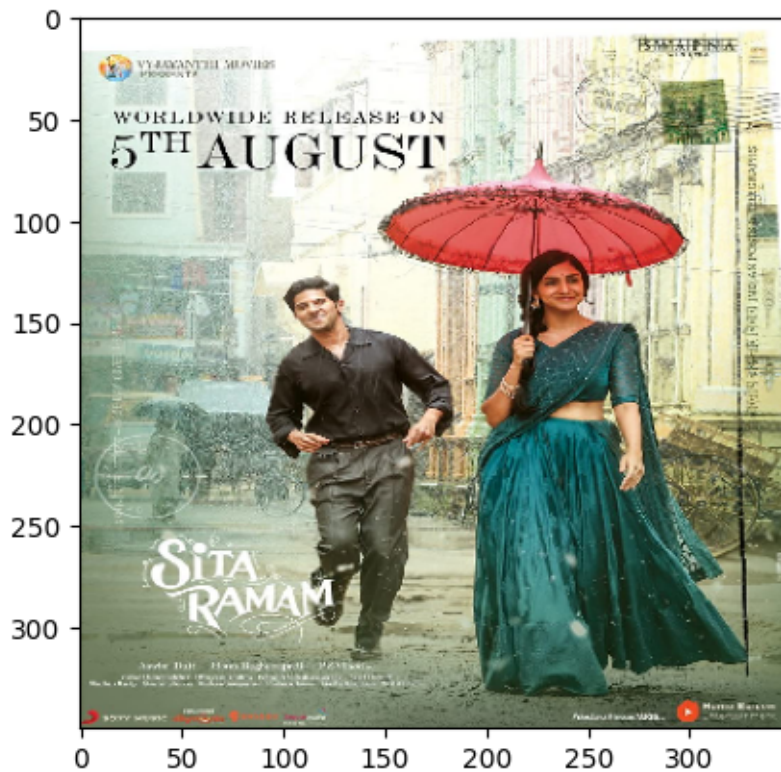
100%| | 10/10 [00:04<00:00, 2.38it/s]

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

```
[ ]: (10, 350, 350, 3)
```

```
[ ]: plt.imshow(X[1])
```

```
[ ]: <matplotlib.image.AxesImage at 0x7f2f56b3fdf0>
```



Labelling image

```
[ ]: data['Genre'][1]
```

```
[ ]: "['Drama', 'Romance', 'Music']"
```

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

```
[ ]: (10, 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"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 348, 348, 16)	448
batch_normalization (Batch Normalization)	(None, 348, 348, 16)	64

max_pooling2d (MaxPooling2D)	(None, 174, 174, 16)	0
dropout (Dropout)	(None, 174, 174, 16)	0
conv2d_1 (Conv2D)	(None, 172, 172, 32)	4640
batch_normalization_1 (Batch Normalization)	(None, 172, 172, 32)	128
max_pooling2d_1 (MaxPooling 2D)	(None, 86, 86, 32)	0
dropout_1 (Dropout)	(None, 86, 86, 32)	0
conv2d_2 (Conv2D)	(None, 84, 84, 64)	18496
batch_normalization_2 (Batch Normalization)	(None, 84, 84, 64)	256
max_pooling2d_2 (MaxPooling 2D)	(None, 42, 42, 64)	0
dropout_2 (Dropout)	(None, 42, 42, 64)	0
conv2d_3 (Conv2D)	(None, 40, 40, 128)	73856
batch_normalization_3 (Batch Normalization)	(None, 40, 40, 128)	512
max_pooling2d_3 (MaxPooling 2D)	(None, 20, 20, 128)	0
dropout_3 (Dropout)	(None, 20, 20, 128)	0
flatten (Flatten)	(None, 51200)	0
dense (Dense)	(None, 128)	6553728
batch_normalization_4 (Batch Normalization)	(None, 128)	512
dropout_4 (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 128)	16512
batch_normalization_5 (Batch Normalization)	(None, 128)	512

dropout_5 (Dropout)	(None, 128)	0
dense_2 (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
1/1 [=====] - 6s 6s/step - loss: 0.8828 - accuracy:
0.0000e+00 - val_loss: 0.6933 - val_accuracy: 0.0000e+00
Epoch 2/5
1/1 [=====] - 3s 3s/step - loss: 1.0501 - accuracy:
0.0000e+00 - val_loss: 0.6877 - val_accuracy: 0.0000e+00
Epoch 3/5
1/1 [=====] - 3s 3s/step - loss: 0.9229 - accuracy:
0.0000e+00 - val_loss: 0.6806 - val_accuracy: 0.0000e+00
Epoch 4/5
1/1 [=====] - 2s 2s/step - loss: 0.9467 - accuracy:
0.0000e+00 - val_loss: 0.6739 - val_accuracy: 0.0000e+00
Epoch 5/5
1/1 [=====] - 2s 2s/step - loss: 0.8569 - accuracy:
0.1250 - val_loss: 0.6709 - val_accuracy: 0.0000e+00
```

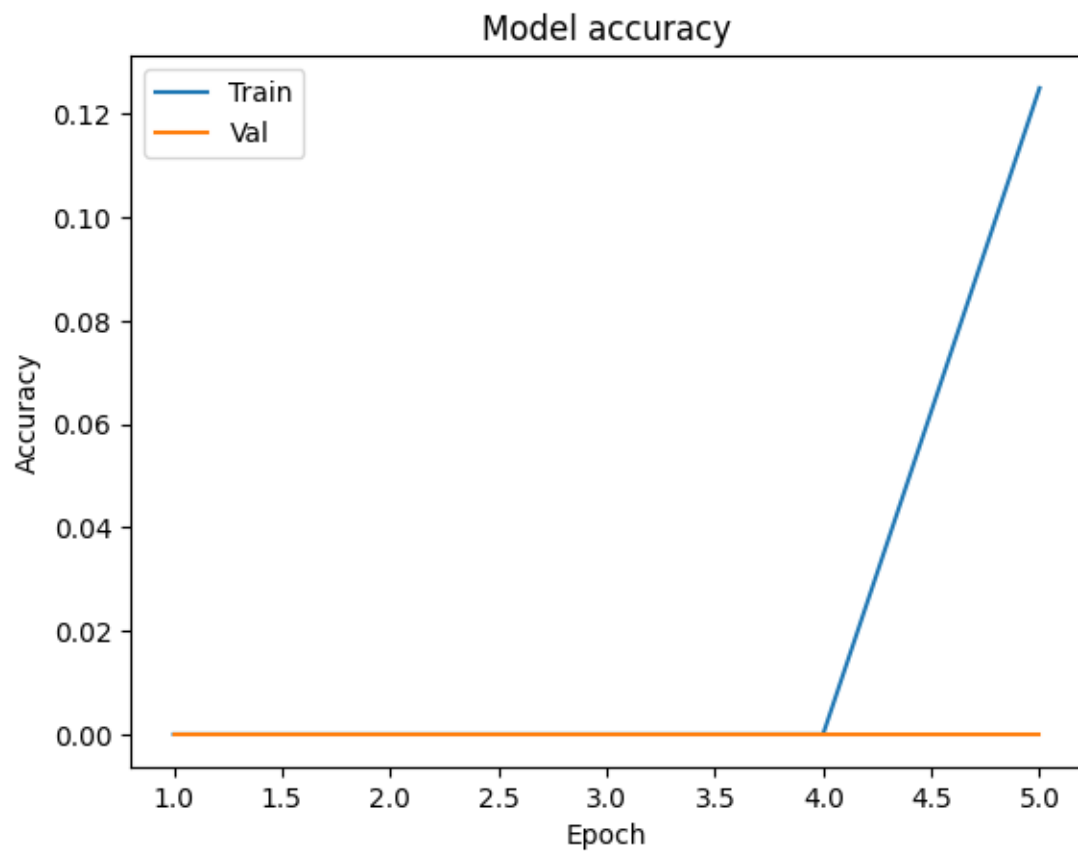
Plot for training and validation loss

```
[ ]: def plot_learningCurve(history, epoch):
    epoch_range = range(1, epoch+1)
    plt.plot(epoch_range, history.history['accuracy'])
    plt.plot(epoch_range, history.history['val_accuracy'])
    plt.title('Model accuracy')
    plt.ylabel('Accuracy')
    plt.xlabel('Epoch')
    plt.legend(['Train', 'Val'], loc='upper left')
    plt.show()

    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)
```



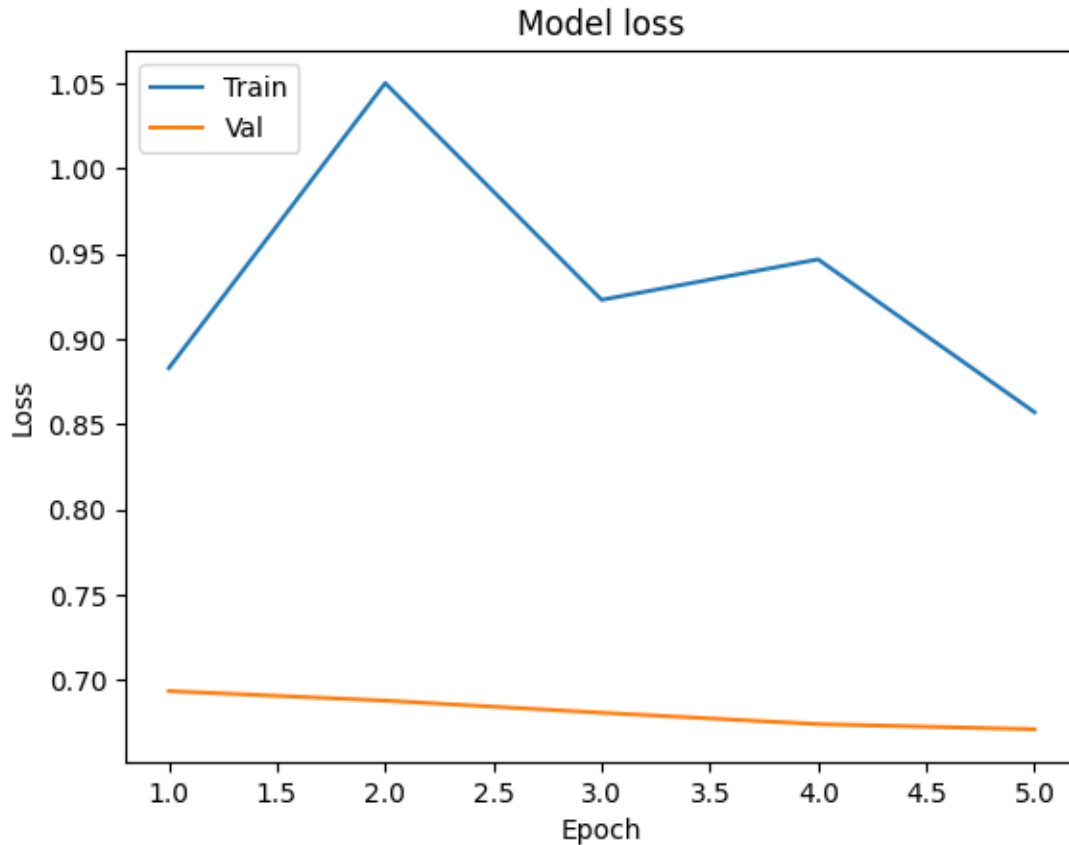


Image labelling on sample movie poster

```
[ ]: img = image.load_img('IB71.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 [=====] - 0s 72ms/step  
Horror  
Musical  
Sci-Fi
```

