

FCN-Baseline-20E-16L-FFA-test-02

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1 Are Relations Relevant in CNNs? *A Study Based on a Facial Dataset*

1.1 Testing Baseline FCN (*20 Epochs - 15 Layers*)

1.1.1 Imports, Seed, GPU integration

```
[1]: import numpy as np
import random
import tensorflow as tf
```

```
[2]: # Seeds for better reproducibility
seed = 42
np.random.seed(seed)
random.seed(seed)
tf.random.set_seed(seed)
```

```
[3]: from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.metrics import confusion_matrix
import itertools
import matplotlib.pyplot as plt
import warnings
warnings.simplefilter(action='ignore', category=FutureWarning)
%matplotlib inline
```

```
[4]: physical_devices = tf.config.experimental.list_physical_devices('GPU')
print("Num GPUs Available: ", len(physical_devices))
tf.config.experimental.set_memory_growth(physical_devices[0], True)
```

Num GPUs Available: 1

1.1.2 Data preparation

```
[5]: test_path = '../..../picasso_dataset/FFA-data/middle/test'
```

```
[6]: test_batches = ImageDataGenerator(preprocessing_function=tf.keras.applications.
    ↪vgg16.preprocess_input) \
```

```
.flow_from_directory(directory=test_path, target_size=(224,224),  
→classes=['no_face', 'face'], batch_size=10, shuffle=False)
```

Found 3000 images belonging to 2 classes.

```
[7]: assert test_batches.n == 3000  
assert test_batches.num_classes == 2
```

1.1.3 Loading the trained FCN

```
[8]: filename='../models/FCN-B-20E-16L-02.h5'  
loaded_model = load_model(filename)
```

1.1.4 Accuracy and loss of the trained model

```
[9]: scores = loaded_model.evaluate(test_batches, verbose=2)  
print("Accuracy: %.2f%%" % (scores[1]*100))  
print("Loss: %.2f%%" % (scores[0]*100))
```

```
300/300 - 7s - loss: 0.0316 - accuracy: 0.9913  
Accuracy: 99.13%  
Loss: 3.16%
```

1.1.5 Testing the FCN

```
[10]: predictions = loaded_model.predict(x=test_batches, steps=len(test_batches),  
→verbose=0)
```

1.1.6 Index of wrongly predicted pictures

```
[11]: y_true=test_batches.classes  
y_pred=np.argmax(predictions, axis=-1)  
cm = confusion_matrix(y_true = y_true, y_pred = y_pred)
```

```
[12]: face_but_predicted_no_face=[]  
no_face_but_predicted_face=[]  
  
for i in range(len(predictions)):  
    if y_true[i] != y_pred[i]:  
        if y_true[i] == 1:  
            face_but_predicted_no_face.append(i+8001-1500) #Index of file  
→on disk  
        else:  
            no_face_but_predicted_face.append(i+8001) #Index of file on disk
```

```

print("Data from class 'face', that was wrongly predicted as 'no-face' [",
      ↪len(face_but_predicted_no_face), "] :")
print(face_but_predicted_no_face)
print("-----")
print("Data from class 'no-face', that was wrongly predicted as 'face' [",
      ↪len(no_face_but_predicted_face), "] :")
print(no_face_but_predicted_face)

```

Data from class 'face', that was wrongly predicted as 'no-face' [21] :
 [8207, 8313, 8358, 8381, 8487, 8501, 8534, 8556, 8610, 8858, 8862, 9015, 9108,
 9113, 9122, 9207, 9272, 9323, 9330, 9401, 9491]

 Data from class 'no-face', that was wrongly predicted as 'face' [5] :
 [8063, 8082, 8125, 9137, 9193]

1.1.7 Confusion matrix

```

[13]: def plot_confusion_matrix(cm, classes,
                                normalize=False,
                                title='Confusion matrix',
                                cmap=plt.cm.Blues):

    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation=45)
    plt.yticks(tick_marks, classes)

    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')

    print(cm)

    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, cm[i, j],
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")

    plt.tight_layout()
    plt.ylabel('True label')
    plt.xlabel('Predicted label')

```

```
[14]: test_batches.class_indices
```

```
[14]: {'no_face': 0, 'face': 1}
```

```
[15]: cm_plot_labels = ['no_face', 'face']  
      plot_confusion_matrix(cm=cm, classes=cm_plot_labels, title='Confusion Matrix')
```

Confusion matrix, without normalization

```
[[1495    5]  
 [  21 1479]]
```

