

# CNN-Baseline-30E-13L-shift-test-02

March 24, 2021

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

### 1.1 Testing Baseline CNN (*30 Epochs - 13 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/basis-data/shifted/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 CNN

```
[8]: filename='../models/CNN-B-30E-13L-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: 2.9729 - accuracy: 0.6743

Accuracy: 67.43%

Loss: 297.29%

### 1.1.5 Testing the CNN

```
[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' [ 891 ] :
[8001, 8003, 8004, 8005, 8006, 8007, 8008, 8009, 8010, 8011, 8012, 8013, 8014,
8015, 8016, 8018, 8022, 8023, 8024, 8025, 8026, 8027, 8028, 8029, 8030, 8031,
8032, 8033, 8034, 8035, 8036, 8038, 8040, 8042, 8043, 8045, 8046, 8047, 8048,
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8063, 8064, 8065, 8066, 8067, 8068, 8069, 8070, 8071, 8072, 8073, 8074, 8076,
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-----  
 Data from class 'no-face', that was wrongly predicted as 'face' [ 86 ] :  
 [8007, 8124, 8177, 8180, 8193, 8215, 8217, 8288, 8292, 8294, 8312, 8346, 8387, 8395, 8401, 8413, 8430, 8436, 8440, 8444, 8487, 9001, 9009, 9015, 9016, 9024, 9031, 9034, 9039, 9047, 9049, 9064, 9067, 9079, 9083, 9106, 9121, 9126, 9140, 9146, 9150, 9152, 9173, 9180, 9185, 9219, 9223, 9224, 9249, 9250, 9266, 9267, 9269, 9271, 9276, 9287, 9295, 9303, 9308, 9314, 9318, 9319, 9321, 9322, 9329, 9345, 9348, 9355, 9357, 9369, 9370, 9372, 9401, 9402, 9411, 9423, 9425, 9442, 9444, 9445, 9453, 9455, 9480, 9481, 9492, 9495]

### 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
[[1414   86]
 [ 891 609]]

```

