

Colorful Image Colorization with Tensorflow

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Implementing Artificial Neural Networks with Tensorflow

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1 Introduction/Motivation

Based on the paper *Colorful Image Colorization* (Zhang et al., 2016) this project aims to reimplement a similar artificial neuronal net that transforms grayscale images into colorful pictures.

This involves first creating a dataset based on pictures that are converted into the CIELAB colorspace (Lab), such that the first channel “L” can be considered as input as it is grayscale whereas the “a” and “b” channel are the target labels to be predicted. Thus, the problem can be handled as classification task. In the second step, the aim was to closely rebuilt the layers of the original model (which used “caffe” (Jia et al., 2014)) using tensorflow 2.0 (richtige version?).

Other project have trained convolutional neural networks (CNNs) on the color prediction problem before (e.g. Cheng et al. (2015), Dahl). The training data is easily available which enables training on large datasets. Problem about previous approaches is that they try to predict the ground truth rather than a possible truth. A conservative loss function tries to minimize Euclidean error between estimate and ground truth. As objects can have various plausible colors, these predictions are multimodal. Thus, the approach of Zhang et al. (2016) innovates a loss function that predicts plausible colors for pixels, rather than the original color (Zhang et al., 2016).

we want to build a network that can colorize images

- this project aims to produce colorful images, given a greyscale picture.
- transforming greyscale into plausible colors is an easy task for humans
- We see a greyscale picture showing a woman playing volleyball at the beach. As we can recog-

nize the scene and the form and relate to it. The sand is yellow, the sea is blue and the ball is white.

- But coloring it in life would be a much more difficult task. As we also need to consider different textures, shades and so on. Seeing and imagining things does not make people a proper painter.
- Surface structure and the semantics of the scene are necessary to validly color images.
- this project aims does not aim to generate the true color for pictures but at least a good and prediction.
- aus dem paper: model enough of the statistical dependencies between semantics and the textures of greyscale images and their color versions in order to produce visually compelling results.

2 Important background knowledge (including reference to most relevant publications)

```
#packages needed
import numpy as np
import tensorflow as tf
import cv2
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Activation,
    BatchNormalization
    for loop
'ich bin ein string'
if
return
MyClass,__init__
```

3 “In-text” listing highlighting

- We started looking at this paper:
<https://arxiv.org/abs/1603.08511> ,
- at their best solution

4 The model and the experiment (MAIN PART). This part should feature code.

4.1 Dataset

- loading large amount of data

<https://machinelearningmastery.com/how-to-load-large-datasets-from-directories-for-deep>

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4.2 Preprocessing

- image preprocessing documentation

<https://keras.io/preprocessing/image/#imagedatagenerator-class>

- preprocessing via ImageDataGenerator() from keras.preprocessing.image
- takes in traindata, validation data and test data
- featurewise-center and featurewise std
- classmode: none -> is for predictions

4.3 Layer

4.4 Loss-Function

5 Visualization and discussion of your results.

oder lieber bilder als screenshot einfügen?

```
1 import numpy as np
2 import tensorflow as tf
3 #from skimage import color
4 import cv2
5
6 from tensorflow.keras.preprocessing.image import ImageDataGenerator
7
8 from tensorflow.keras.models import Sequential
9
10 from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Activation, BatchNormalization
11 #from keras.layers import Dense
```

Figure 1: Hier kann man dann auch noch etwas dazu schreiben

5.1 Training

5.2 Testing

6 Literaturverzeichnis

Cheng, Z., Yang, Q., and Sheng, B. (2015). Deep colorization. *2015 IEEE International Conference on Computer Vision (ICCV)*.

Dahl, R.

Jia, Y., Shelhamer, E., Donahue, J., Karayev, S., Long, J., Girshick, R., Guadarrama, S., and Darrell, T. (2014). Caffe: Convolutional architecture for fast feature embedding. *arXiv preprint arXiv:1408.5093*.

Zhang, R., Isola, P., and Efros, A. A. (2016). Colorful image colorization. *CoRR*, abs/1603.08511.