



Classification of LEGO Bricks Based on Images

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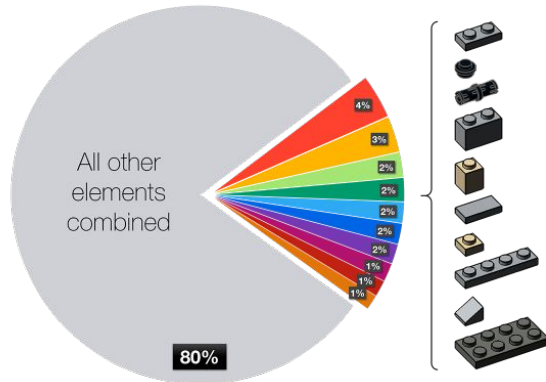
Introduction

- 3700+ unique LEGO bricks
- 10 most common bricks \cong 20% of all elements



<https://absatzwirtschaft-1cf93.kxcdn.com/wp-content/uploads/2022/08/lego-1024x676.jpg>

Most common LEGO elements



<https://brickarchitect.com/bricks/>

- over 600 billion parts in the world
- 74 different brands of building blocks systems compatible with LEGO
- most bricks made from ABS (Acrylonitrile Butadiene Styrene)
 - cannot be recycled

Data

- dataset:

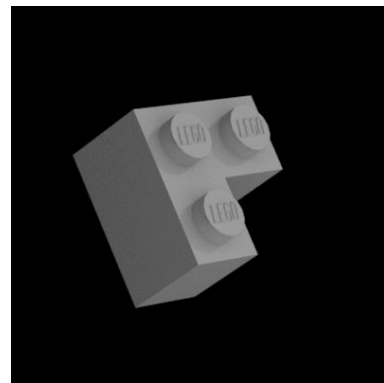
Images of LEGO Bricks

40,000 images of 50 different LEGO bricks

<https://www.kaggle.com/datasets/joosthazelzet/lego-brick-images>



- 50 bricks computer rendered by 800 different angles
- 2 camera setup
- 400 x 400 pixels, 3 RGB channels



Used Methods/Models - Preprocessing

- load pictures into pandas dataframe, get labels from file names

```
filenames = os.listdir(path)
filenames.remove('.gitkeep')
df = pd.DataFrame(filenames, columns=['FileNames'])

df['Label'] = df['FileNames'].apply(lambda x: get_label(x))
df = df.sample(frac=1, random_state=1).reset_index()
```

- get train and validation generator using flow_from_dataframe

```
train_generator = datagen.flow_from_dataframe(df,
                                             directory=path,
                                             x_col='FileNames',
                                             y_col='Label',
                                             target_size=image_size,
                                             batch_size=batch_size,
                                             color_mode='grayscale',
                                             class_mode='categorical',
                                             subset='training',
                                             )
```

Used Methods/Models - Tuner

- Setting up a Keras Tuner

```
tuner = keras_tuner.RandomSearch(  
    hypermodel=build_model,  
    objective="val_accuracy",  
    max_trials=10,  
    executions_per_trial=1,  
    overwrite=True,  
    directory=" ../experiments",  
    project_name="Hyperparametertuning",  
)
```

- Hyperparameter Grid based on Baseline Model
- Retraining of best model found

```
# define search space  
convolution_layers_total_min = 2  
convolution_layers_total_max = 6  
convolution_layers_total_step = 1  
  
filter_count_min = 32  
filter_count_max = 64  
filter_count_step = 32  
  
kernel_size_min = 2  
kernel_size_max = 4  
kernel_size_step = 2  
  
dense_layers_total_min = 9  
dense_layers_total_max = 13  
dense_layers_total_step = 1  
  
units_count_min = 150  
units_count_max = 250  
units_count_step = 50
```

Baseline Model

```
model = tf.keras.  
# Flatten to  
tf.keras.laye  
tf.keras.laye  
  
tf.keras.laye  
tf.keras.laye  
l)
```

```
model.compile(loss=  
optimizer=met
```

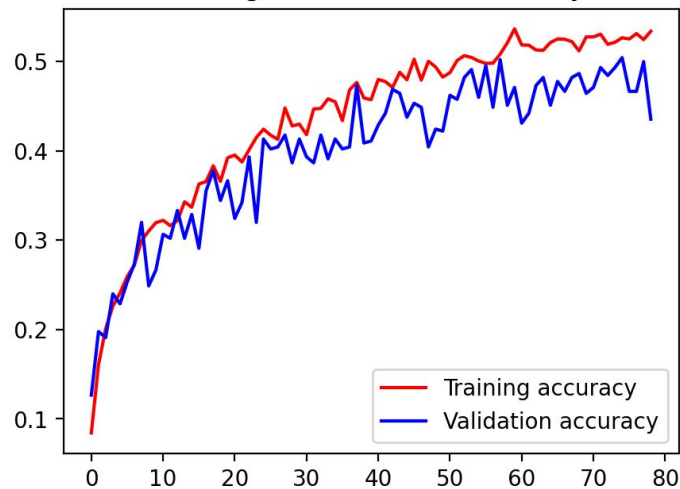
Model: "sequential_6"

Layer (type)	Output Shape	Param #
=====		
flatten_6 (Flatten)	(None, 22500)	0
dropout_6 (Dropout)	(None, 22500)	0
dense_12 (Dense)	(None, 128)	2880128
dense_13 (Dense)	(None, 50)	6450

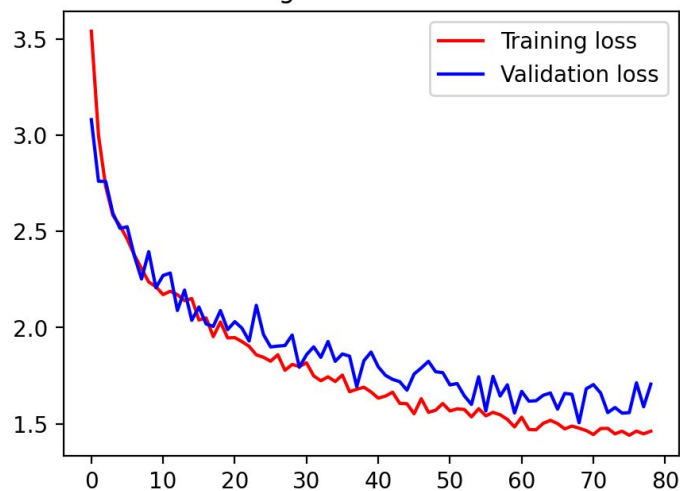
=====

Total params: 2,886,578
Trainable params: 2,886,578
Non-trainable params: 0

Training and validation accuracy



Training and validation loss



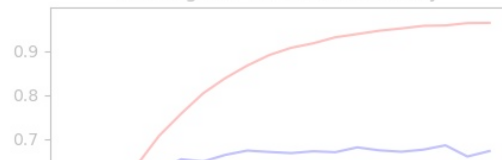
Results

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 149, 149, 64)	320
max_pooling2d (MaxPooling2D)	(None, 74, 74, 64)	0
conv2d_1 (Conv2D)		
conv2d_2 (Conv2D)		
max_pooling2d_1 (MaxPooling2D)		
flatten (Flatten)		
dense (Dense)	(None, 80)	6635600
dense_1 (Dense)	(None, 90)	7290
dropout (Dropout)	(None, 90)	0
dense_2 (Dense)	(None, 50)	4550

=====
Total params: 6,676,552
Trainable params: 6,676,552
Non-trainable params: 0

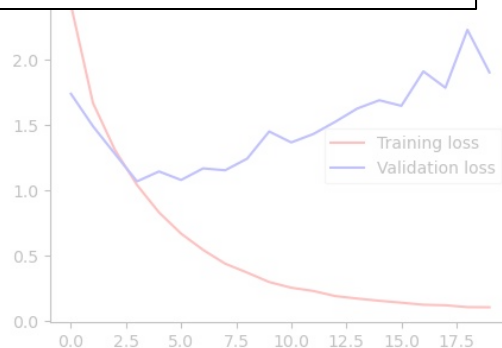
Training and validation accuracy



```
1 model.evaluate(test_generator, verbose=0)
```

[79] ✓ 9.3s

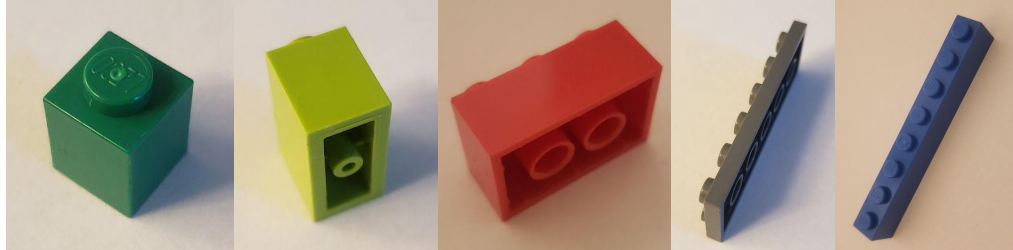
... [1.984946846961975, 0.6685000061988831]



Problems

- Unexplainable Overfitting
 - Reason: Train- and Testgenerator contain different classes
 - Solution: shuffle data
- Can not load previously saved model (RuntimeWarning: Unexpected end-group tag: Not all data was converted)
 - solution: <3
- Defining a reasonable grid of Hyperparameters to prevent running out of memory
 - Solution: Trial-and-Error

Testing the model



```
1 # prepare image to fit model input
2 image = np.array(image)
3 image = image.reshape((150, 150, 1))
4 image = image/255
5 image = np.expand_dims(image, axis=0)
6
7 # predict image
8 prediction = model.predict(image, verbose=0)
9 index_class_predicted = np.argmax(prediction)
10 keys=list(validation_generator.class_indices.keys())
11 class_predicted = keys[index_class_predicted]
12 class_predicted
```

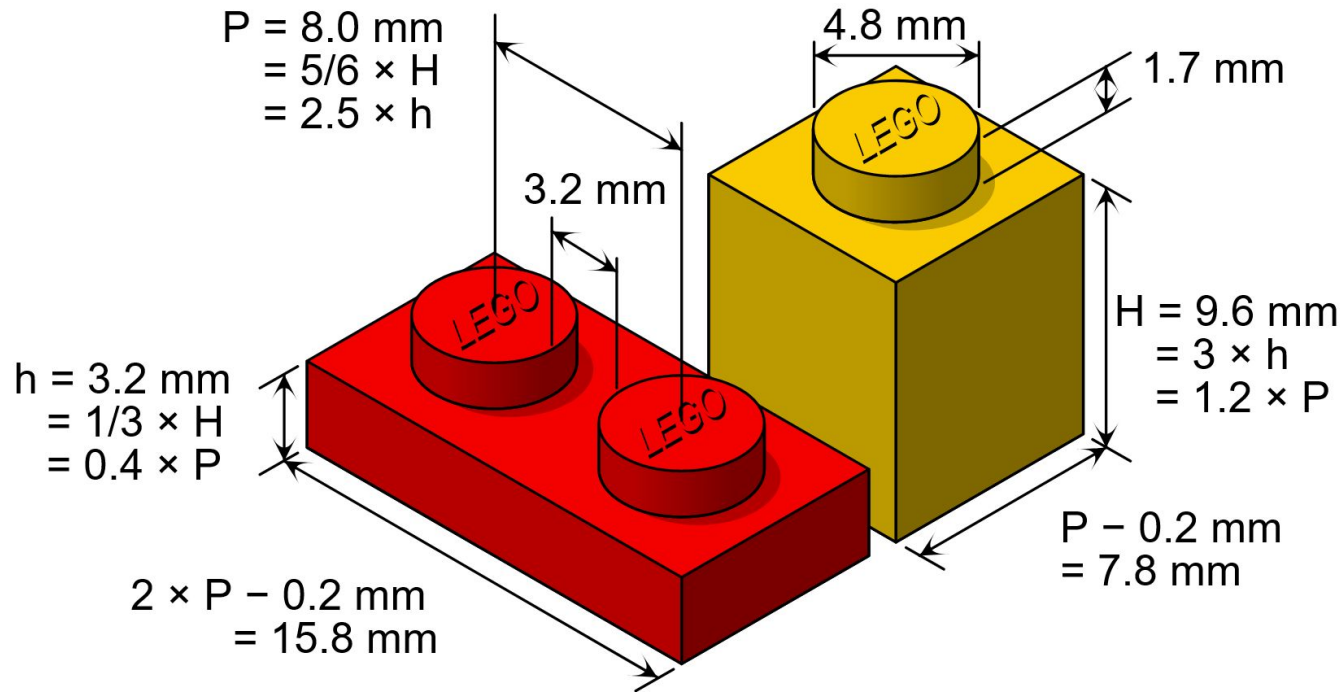
✓ 0.4s

'27925 flat tile round 2x2'

```
1 # prepare image to fit model input
2 image = np.array(image)
3 image = image.reshape((150, 150, 1))
4 image = image/255
5 image = np.expand_dims(image, axis=0)
6
7 # predict image
8 prediction = model.predict(image, verbose=0)
9 index_class_predicted = np.argmax(prediction)
10 keys=list(validation_generator.class_indices.keys())
11 class_predicted = keys[index_class_predicted]
12 class_predicted
```

✓ 0.4s

'3001 brick 2x4'



Thank you for your attention!