

Classification of a Melanoma with Deep Learning

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1 Introduction – Classes

Goal: Classify three types of skin-lesions.

<u>Class</u>	Nevus	Seborrheic keratosis	Melanoma
Appears at	Birth (congenital)	Increased age	Increased age
<u>Risk</u>	Benign (harmless)	Benign (harmless)	malign (harmfull cancer)
<u>Characet</u> <u>ristics</u>	 hyperpigmented mostly single colored mostly arised skin Can turn into a Melanoma! 	 Flat/raised papule /plaque mostly one single smooth, waxy or warty surface Can look similar to a Melanoma 	ABCDE Rule (vs. Nevus) - Asymmetry - Border irregular in shape. - Color multicolored - Diameter > 6mm in diameter, - Evolution change characteristics over time.

2 Data

Melanoma Detection Dataset¹ from Kaggle.

- Unbalanced data

	# total images
test	600
train	2000
valid	150

3 Models and Methods

- 1. Baseline Model (with Random Forest [RF])
- 2. Pretrained Neuronal Network [NN]
- 3. Train a ConvolutionalNN from scratch

RF and pretrained NN used the weights of the CNN VGG:

```
## Load base model weights ----
base_model <- application_vgg16(weights = "imagenet",</pre>
                              include_top = FALSE,
                              input_shape = c(image_size, image_size, 3L))
summary(base_model)
# i.e. freeze all convolutional VGG16 lavers
freeze_weights(base_model)
                                       transfer_model %>% compile(
All models were complied
                                          optimizer = "adam",
                                          loss = "categorical crossentropy",
the same:
                                          metrics = c("accuracy")
```

4 Results and Discussion

Model	Loss	Accuracy
Baseline model with RF	NA	0.70
Pretrained VGG	1.52	0.63
CNN from scratch	0.80	0.68

1. Baseline model with RF

The Baseline model using the Random Forest as classification lead to the highest accuracy with 70 %.

RF uses no distribution assumption and can handle well imbalanced data sets (weighted random forest, balanced random

2. Pretrained NN with VGG

The pretrained VGG has the highest loss function and the lowest accuracy of all the trained models. This is surprising as the model provides 4096 features for the final densely connected classifier.

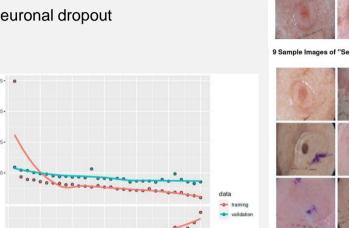
[...] 4 Results and Discussion

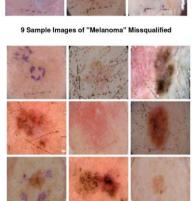
3. CNN from scratch:

Visible overfitting (accuracy of train data > test data)

To avoid overfitting:

- data augmentation4,
- neuronal dropout





5 Conclusion

Data difficulties:

- Different pixel size/coloring of images
- Imbalanced data
- Cumbersome data load -> writing of new data import function

Language/Versioning difficulties:

- Tensorflow/Keras better documented for Python than for R
- Image data augmentation seems to depend on the version of keras and tensorflow. -> data augmentation skipped

Findings:

- Baseline model with RF -> best classification accuracy
- From scratch trained CNN: minimization of loss and the accuracy vary among the batch_sizes.
 - Hypothesis⁵: model optimization algorithm ADAM aims for solution with larger weights -> lower test accuracy

6 Base script

Script: https://github.com/hosts1/ethz-deepl-melanoma

References

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- 3. https://tensorflow.rstudio.com/.
- 4. https://tensorflow.rstudio.com/guides/keras/preprocessing_layers.
- 5. https://medium.com/mini-distill/effect-of-batch-size-on-training-dynamics-21c14f7a716e