



Neural networks detect cutaneous basal cell carcinomas in histological sections

Susanne Kimeswenger

Department of Dermatology and Venerology, KUK

Department of Soft Matter Physics, JKU









Digital pathology



Processing of histological information via digital slides

Goals:

- Simplification of data management
- Image quality stays the same for years
- Spacial independence of pathologists
- Possibility for discussion with specialists
- Improvement of diagnosis

Next step:

Improve digital pathology with artificial intelligence

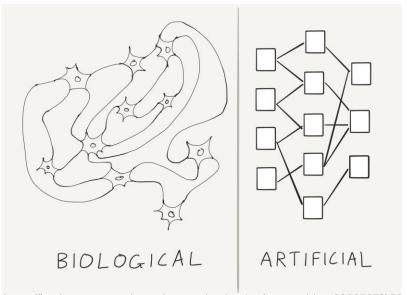
Artificial Neural Networks



Deep neural networks are the #1 method in Machine Learning

Convolutional neural networks (CNNs) are used to detect patterns in images, state-of-the-art in

- Image recognition tasks
- Face recognition
- Self-driving cars



https://hackernoon.com/neural-networks-the-1-minute-guide-a2909507f350



UNIVERSITÄT LINZ

LETTER

doi:10.1038/nature21056

Dermatologist-level classification of skin cancer with deep neural networks

Andre Esteva¹*, Brett Kuprel¹*, Roberto A. Novoa^{2,3}, Justin Ko², Susan M. Swetter^{2,4}, Helen M. Blau⁵ & Sebastian Thrun⁶

CNN is as good at evaluation of skin lesions (photos and dermatoscopic photos) as experienced dermatologists

→ Is that also possible for histopathological images?

A. Esteva, B. Kuprel, R. A. Novoa, J. Ko, S. M. Swetter, H. M. Blau and S. Thrun, Dermatologist-level classification of skin cancer with deep neural networks, *Nature*, 2017, **542**, 115–118.

Machine learning in histopathology





Goals:

- → Accurate predictions
- → Interpretable methods

Proof-of-concept

- → Easy task: Basal cell carcinomas (BCCs) vs. normal skin
- → Classification
- Interpretation











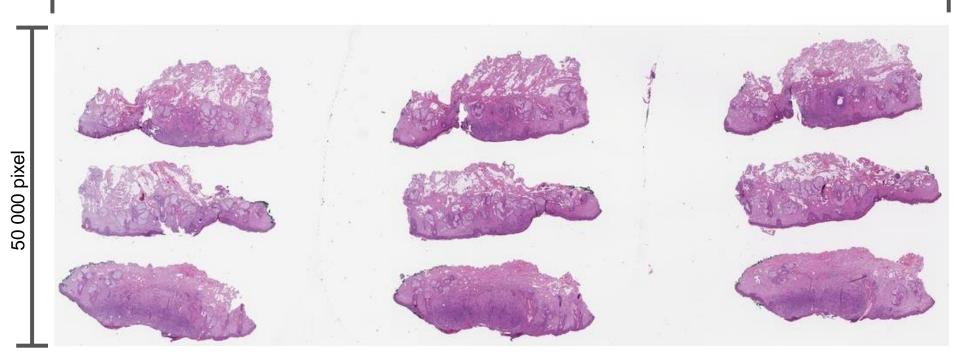


Challenges



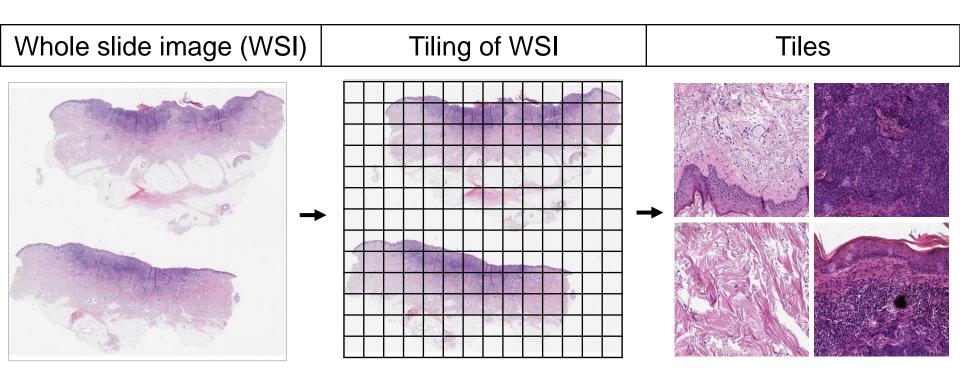


100 000 pixel



Approach

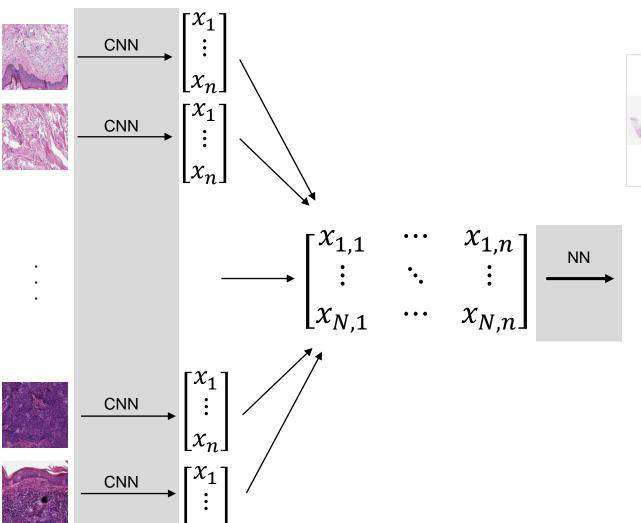




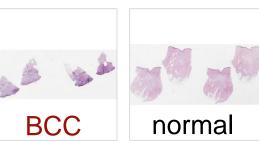


Approach



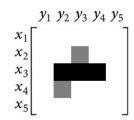


 $\lfloor x_n \rfloor$



Prediction

Attention weight matrix



Results: classification

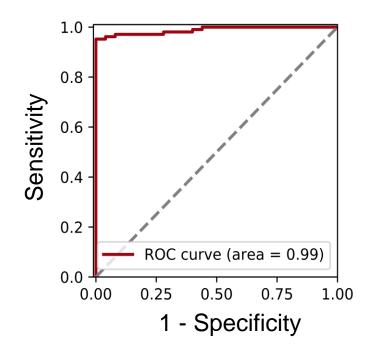


Anonymized images: **709** trainings images, **129** test images

Divided to tiles of size 244x244x3

Pretrained CNN (VGG11)

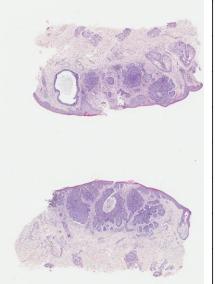
Metric	<u>Value</u>				
Accuracy (threshold = 0.5)	0.96 (0.95-0.96)				
Sensitivity (threshold = 0.5)	0.97 (0.96-0.97)				
Specificity (threshold = 0.5)	0.91 (0.89-0.94)				
F1 score (threshold = 0.5)	0.97 (0.97-0.98)				
Area under the curve (AUC of ROC curve)	0.99 (0.99-0.99)				



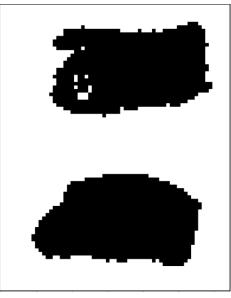
Results: interpretation of CNN



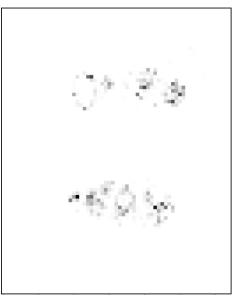
Original image

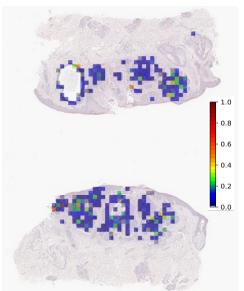


Pre-selected tiles



Attention weight matrix





Results: important tiles for the prediction





diagnosis

Nodular +, cystic BCC					
Superficial BCC					
Nodular + adenoid BCC					
Normal skin					
Normal skin					

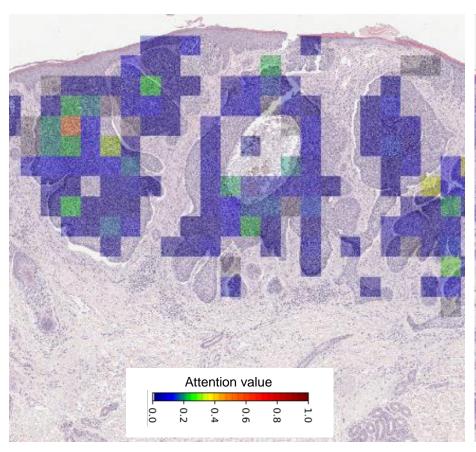
Results: CNN vs. pathologist

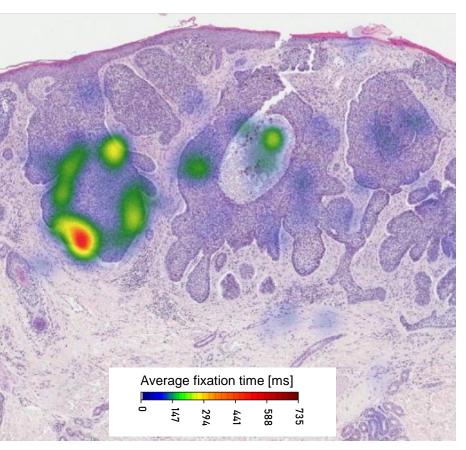
Kepler Universitäts Klinikum

JOHANNES KEPLER UNIVERSITÄT LINZ

12

BCC example





CNN

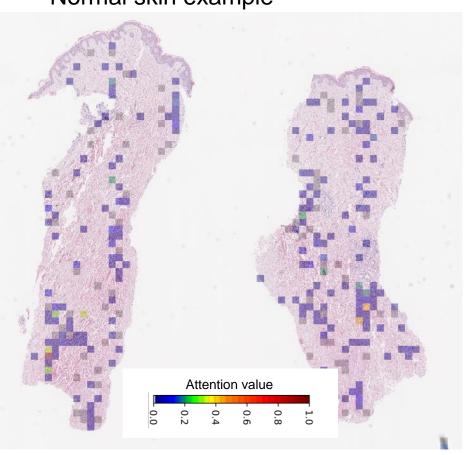
4 board-certified pathologists

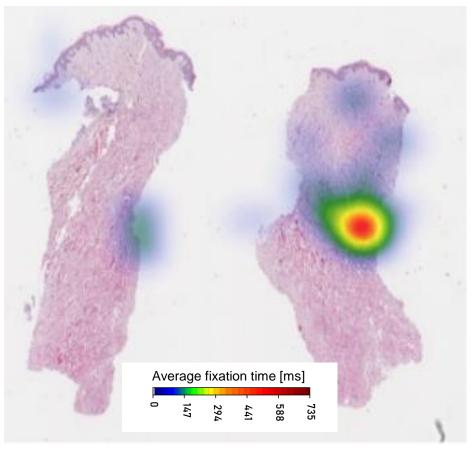
Results: CNN vs. pathologist





Normal skin example





CNN

4 board-certified pathologists

Differences between CNN and pathologists

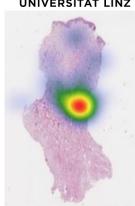
Kepler Universitäts Klinikum



UNIVERSITÄT LINZ

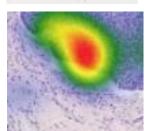
Pathologists focus mainly at individual areas in the sections - for the NN important areas are distributed over the entire section/lesion



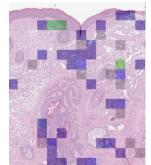


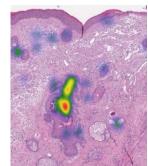
Pathologists consider the artificial retraction gap as important





Pathologists focus on the epidermis, glands and hair follicles - NN considers connective tissue





Conclusion



- CNN/attention architecture is able to distinguish between BCC and non-BCC images with a high accuracy
- CNN/attention architecture focuses on partly different patterns and structures than pathologists for diagnosis

Outlook

- Quantify differences between CNN and pathologists
- Apply similiar methods to
 - Several diseases on one slide
 - More applications (other diseases, other cancer types)
 - More images
- More targets (mutation prediction, protein expression prediction, ...)

Thank you!

Kepler Universitäts Klinikum



Wolfram Hötzenecker

Department of Dermatology



Günter Klambauer Markus Hofmarcher Elisabeth Rumetshofer Sepp Hochreiter Institute for Machine Learning



Rene Silye Petar Noack Gudrun Lang Department of Pathology and Microbiology



Phillip Tschandl Christoph Sinz Giuliana Petronio Harald Kittler Department of Dermatology

