

TRAINING: Introduction to Deep Learning

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SACHSEN Diese Maßnahme wird gefördert durch die Bundesregierung aufgrund eines Beschlusses des Deutschen Bundestages. Diese Maßnahme wird mitfinanziert durch Steuermittel auf der Grundlage des von den Abgeordneten des Sächsischen Landtags beschlossenen Haushaltes.













































Deep Learning in in Medicine

Data Structuring

Medical Imaging, e.g. Radiology, Pathology, Dermatology

Diagnosis and Triage

Robotics and Surgery

Application Areas

Drug Discovery



Disease Management, e.g. Chronic Disease, Mental Disease

Decision Support /
Personalized Medicine

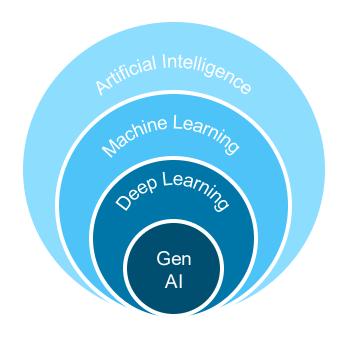
Virtual Health Assistants







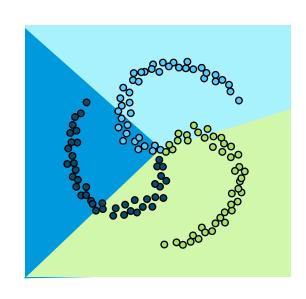
Why Machine Learning is not enough sometimes...



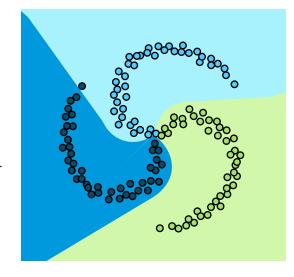
	Feature 1	Feature 2	Feature 3
Pat. 1	21	7	8
Pat. 2	5	35	9
Pat. 3	87	58	3



- Unknown features
- Manual labour
- Complex interactions

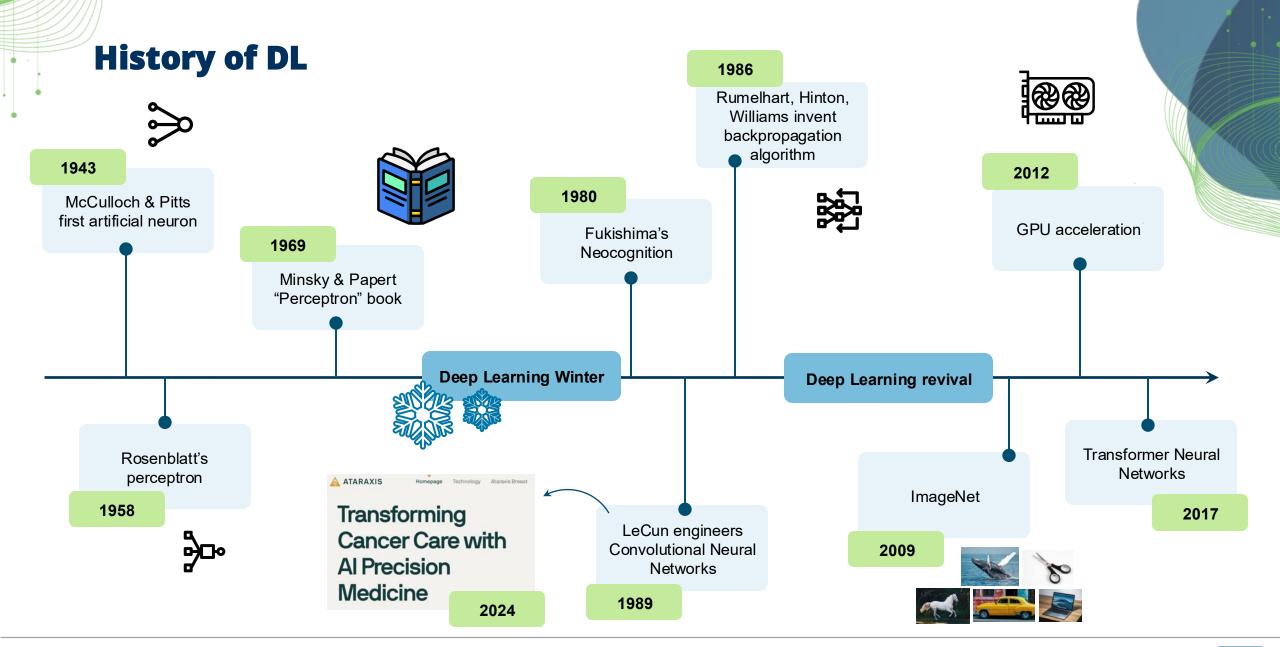


Deep learning can be used to model non-linear relations





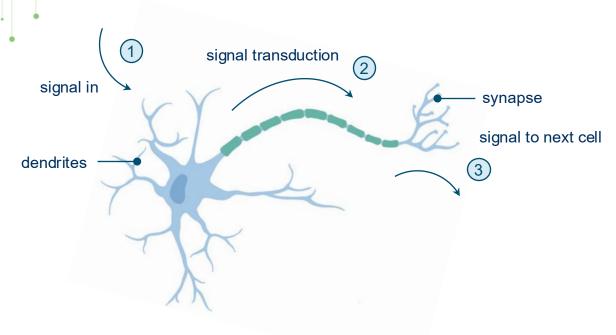


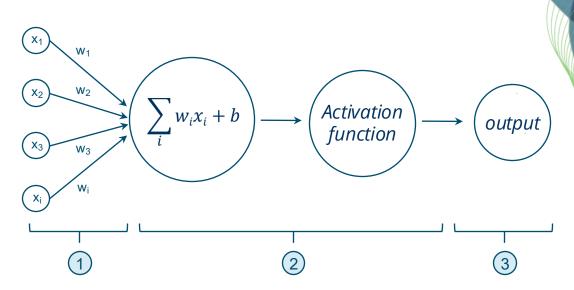




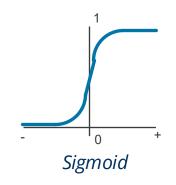


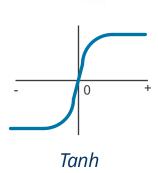
Basics of Neural Networks

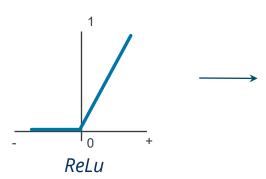


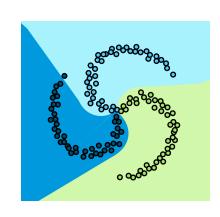


Activation functions:









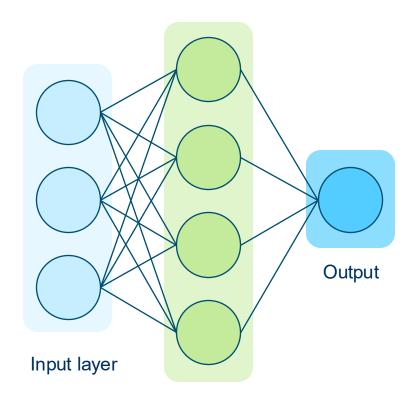
Activation functions introduce non-linearity to neural networks



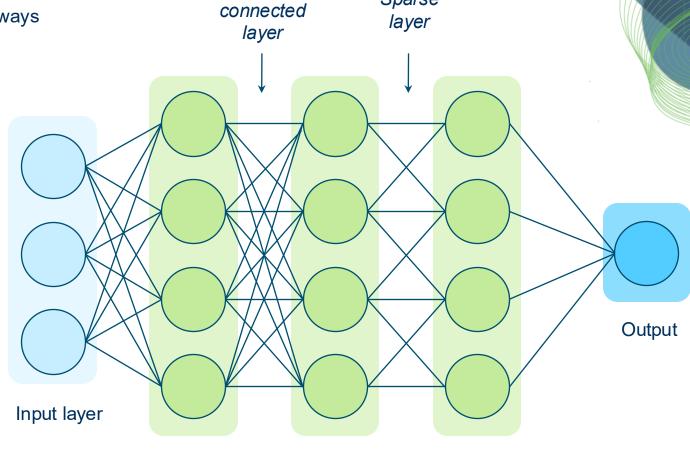


Deep Neural Networks

Artificial neurons can be stacked together in various ways







Hidden layer

Sparse

Dense / fully

Hidden layer

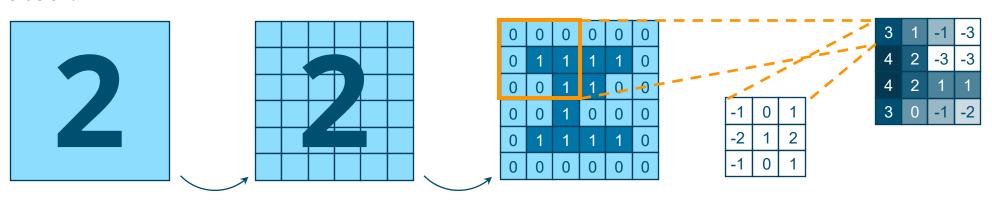




Hidden layer

Convolutional Neural Networks

Convolution:



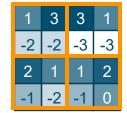
Filters detect structures within the image matrix

Pooling:







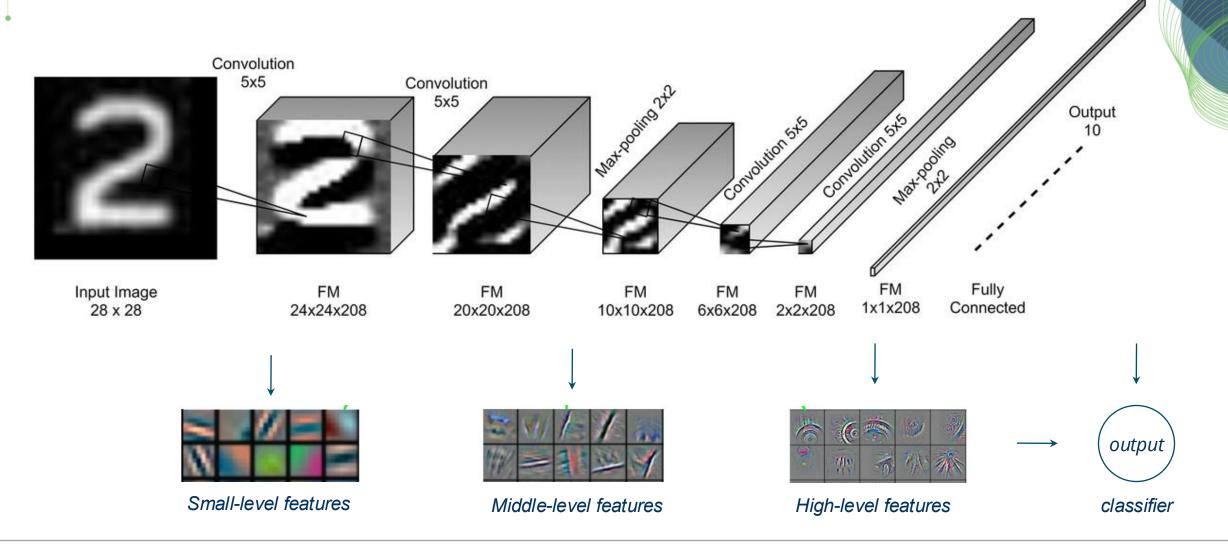


- Pooling reduces spatial size
 Makes features more robust.
 - Makes features more robust to variation





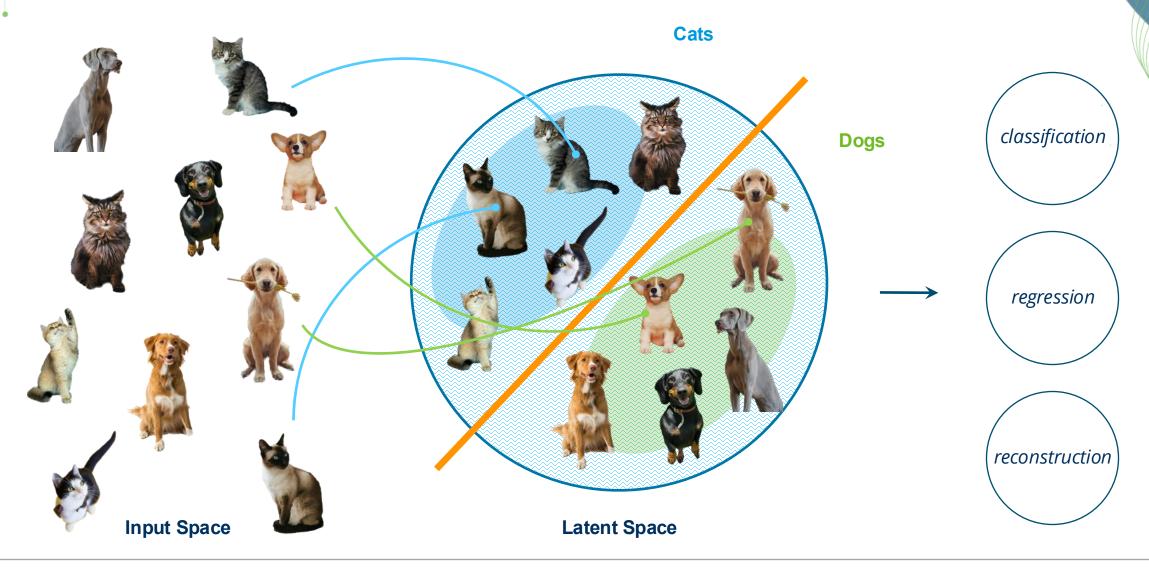
Convolutional Neural Networks







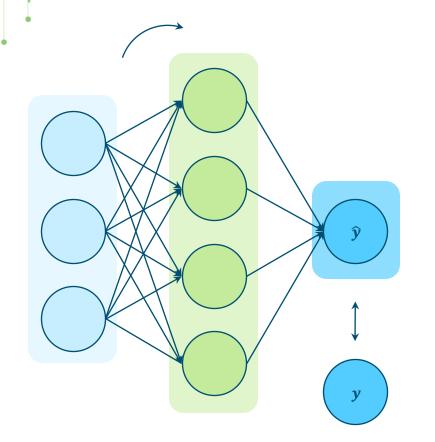
The Latent Space

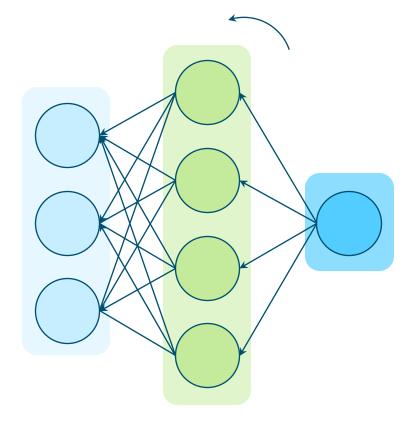


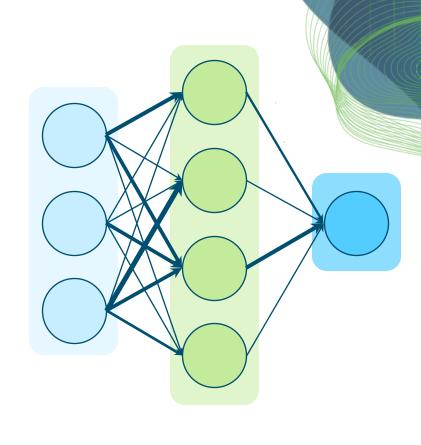




How do Neural Networks learn?





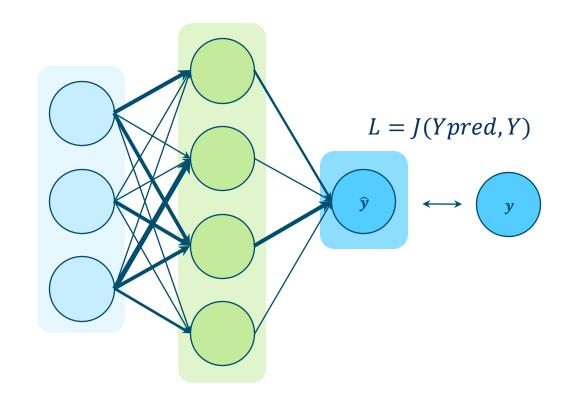


- 1 **Error**: difference between prediction and output
- 2 Error is sent back 3 through all neurons
- Gradient of error is calculated for each weight
- 4 Weights are updated in regard to error
- 5 New error is sent back ...



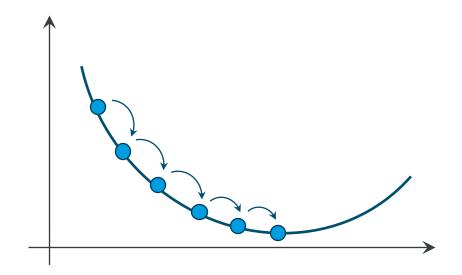


How do Neural Networks learn?



Loss Function

We want to minimize loss function and with this decrease the error between prediction and label.



Loss Function for Regression

$$L = \frac{1}{N} \sum_{i=1}^{N} (yi - \hat{y}_i)^2$$

Loss Binary Classification

$$L = \frac{1}{N} \sum_{i=1}^{N} (yi - \hat{y}_i)^2 \qquad \qquad L = -\frac{1}{N} \sum_{i=1}^{N} [yi \log(\hat{y}_i) + (1 - yi)\log(1 - \hat{y}_i)] \qquad \qquad L = -\frac{1}{N} \sum_{i=1}^{N} \sum_{c=1}^{C} y_{ic} log(\hat{y}_{ic})$$

Loss Multi-Class Classification

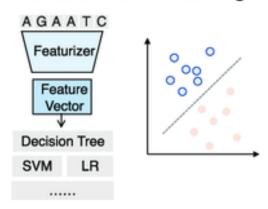
$$L = -\frac{1}{N} \sum_{i=1}^{N} \sum_{c=1}^{C} y_{ic} log(\hat{y}_{ic})$$



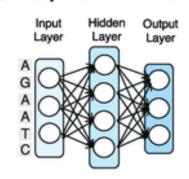


Todays Diversity

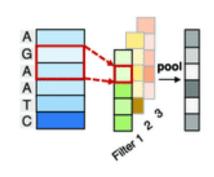
A Classic Machine Learning



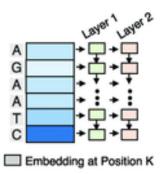
B Deep Neural Network



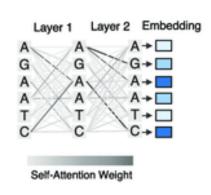
c Convolutional Neural Network



D Recurrent Neural Network

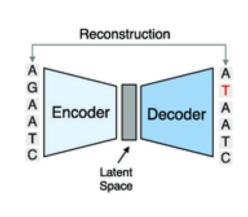


E Transformer



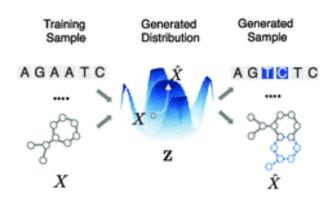
Come2Data

F Graph Neural Network



G Autoencoder

H Generative Model

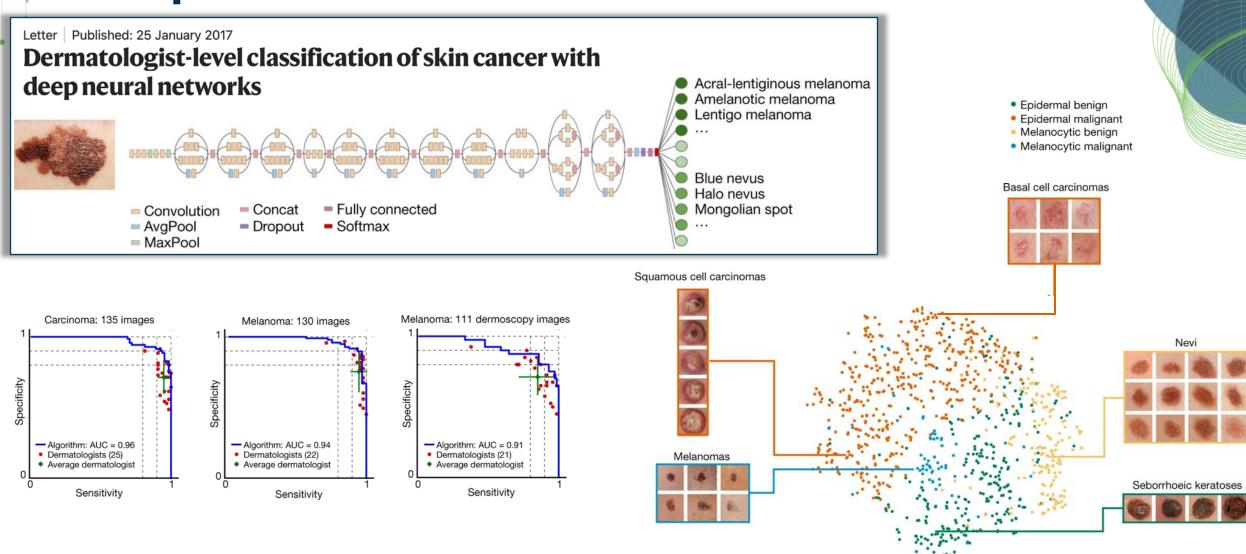






K-th Hop Neighbor

Example - Melanomas



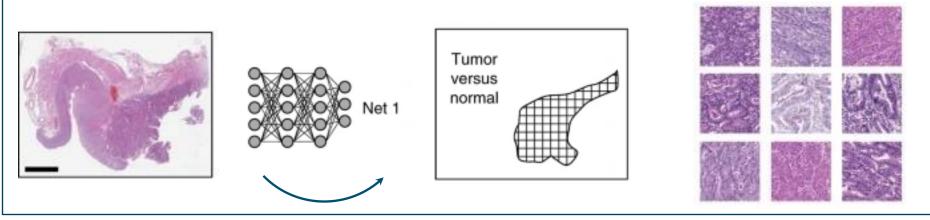


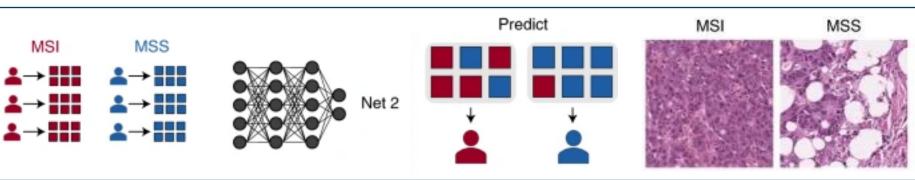


Example - Pathology

Brief Communication | Published: 03 June 2019

Deep learning can predict microsatellite instability directly from histology in gastrointestinal cancer





Data Science and Al for Medicine Training School

Training: Introduction to Deep Learning









Step

Step





Challenges and Limitations

Data Challenges

- Bias in training data
- Imbalance of training data
- Data availability / openaccess
- Artefacts

Model Challenges

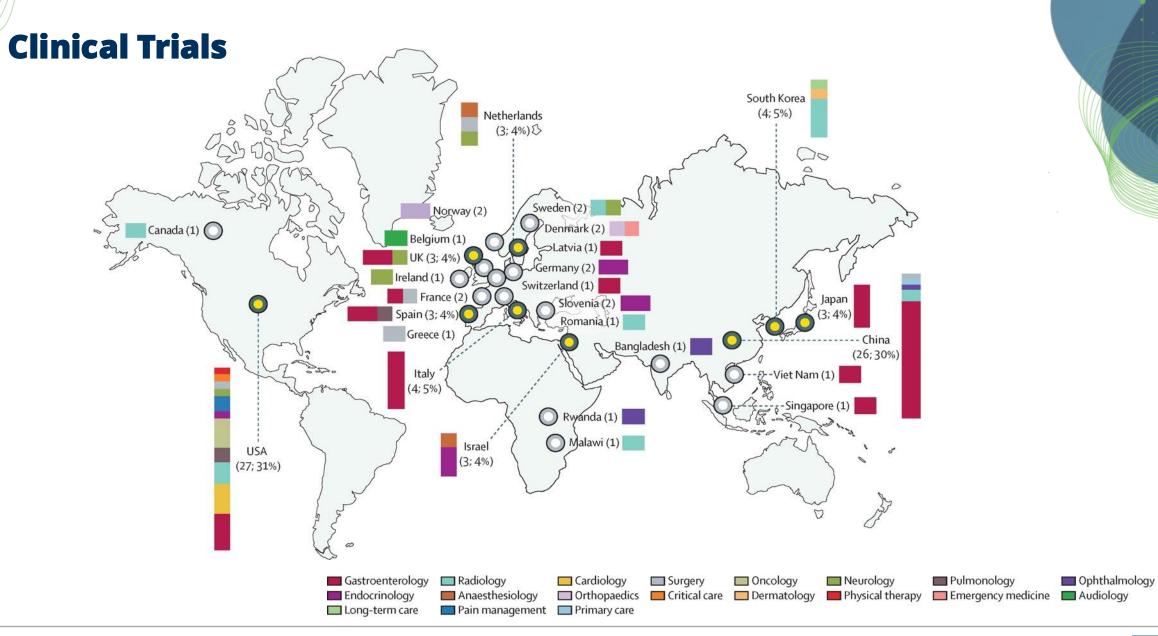
- Digitization
- Explainability
- Decisions aligned with newest clinical guidelines
- Compute
- Size

Clinical Challenges

- Safe use in best interest of patients
- Performance degradation over time / domain shift
- Compliance with regulatory standards
- Easy and accessible use













Any questions or remarks?

Let's practice – Deep Learning Basics on Jupyter Hub



