

The background is a dark blue gradient with a subtle pattern of white dots. On the left side, there are several concentric circles and a large circular scale with degree markings from 140 to 260. Some of the circles have arrows indicating a clockwise direction. The text is positioned on the right side of the image.

# ADVANCED MACHINE LEARNING SEMINAR

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# PIPELINE FOR TRAINING & EVALUATION

- Supports logging with Weights & Biases
- Easy configuration through CLI arguments
- Example:
  - `python train.py capsnet_2d --epochs 5 --slice_width=28 --dataset=mnist --batch_size=32`  

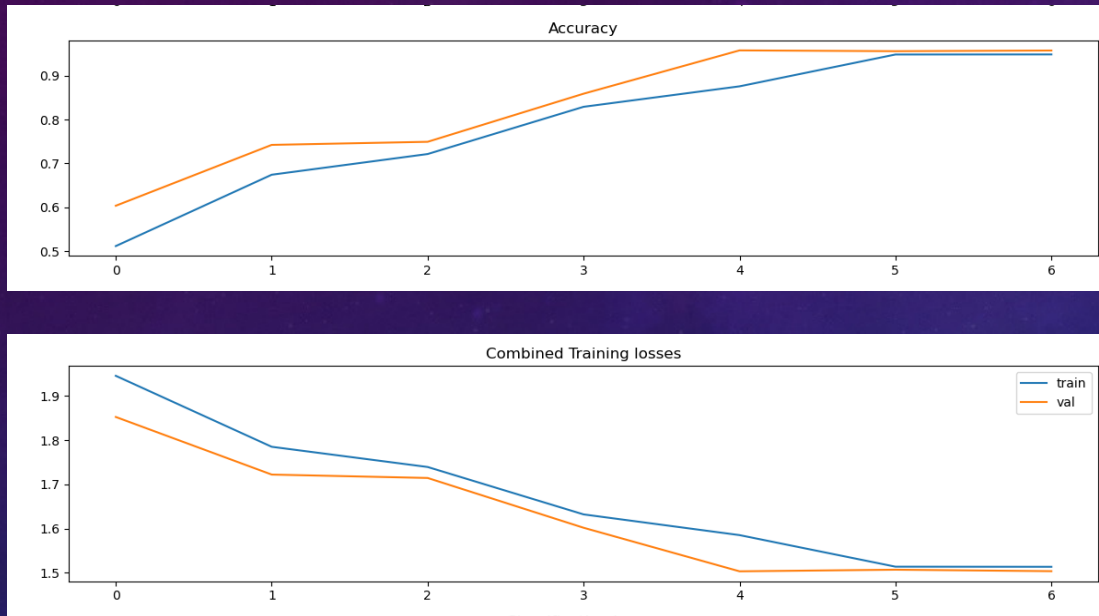
model (capsnet / resnet)      target image size      dataset (mnist, lungpetctdx)
- Evaluation with confusion matrix and explainability through SHAP



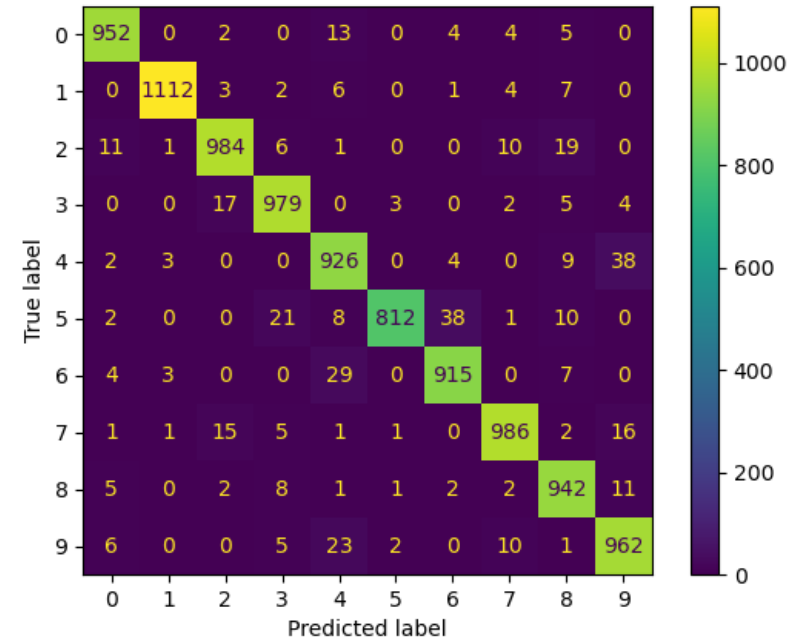
# PIPELINE FOR TRAINING & EVALUATION

- CLI arguments:
  - `--epochs EPOCHS`
  - `--max_loaded_samples MAX_LOADED_SAMPLES`
  - `--learning_rate LEARNING_RATE`
  - `--slice_width SLICE_WIDTH`
  - `--batch_size BATCH_SIZE`
  - `--wandb`
  - `--early_stopping`
  - `--class_imbalance {class_weights,undersample,none}`
  - `--dataset {lungpetctx,mnist}`
- + model specific arguments

# EVALUATION ON MNIST – RESNET 152



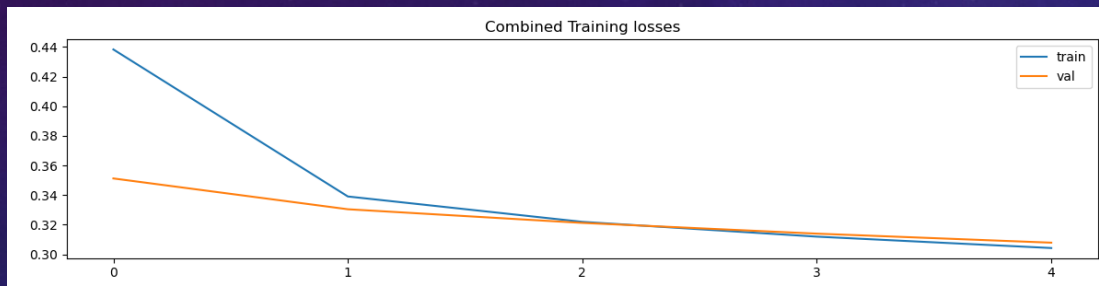
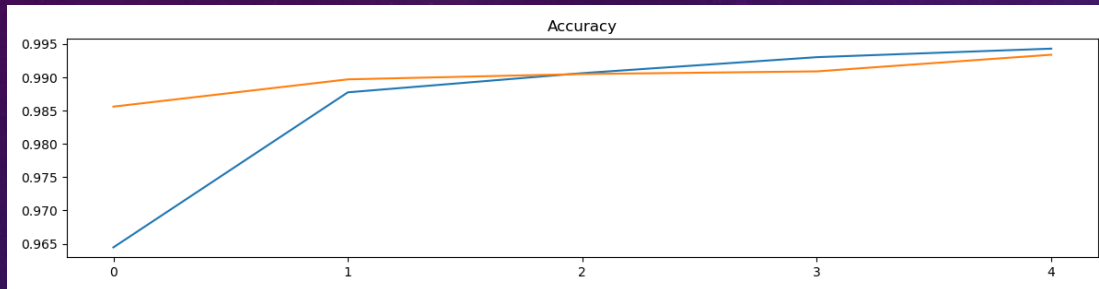
✓ Looks good  
Best validation accuracy at ~0.97



```
python train.py resnet_2d --epochs 20 --slice_width=28 --dataset=mnist --batch_size=32
```

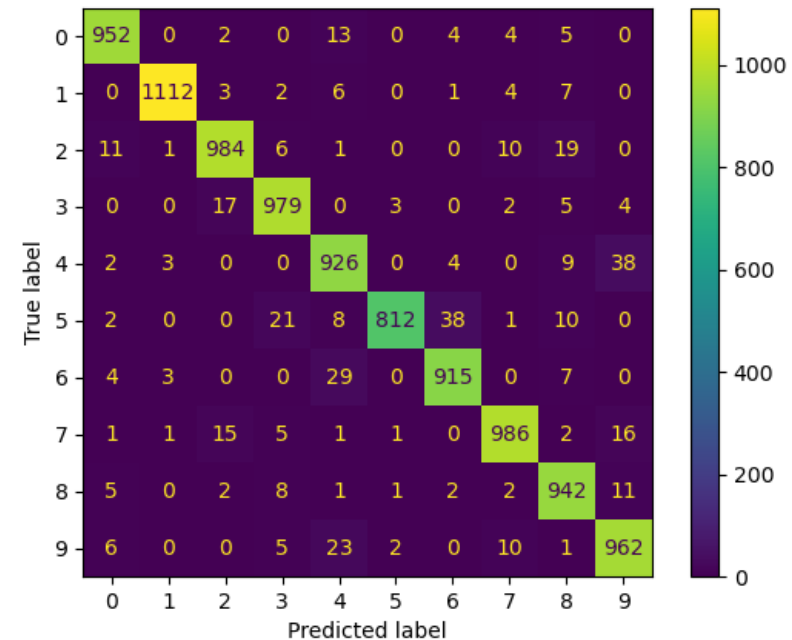


# EVALUATION ON MNIST – CAPSNET 2D



Even better

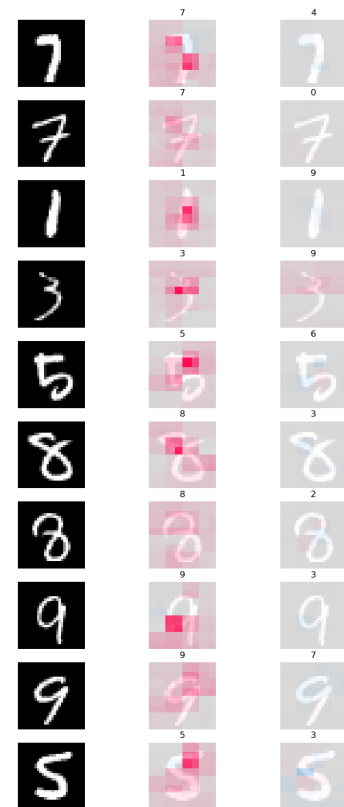
Best validation accuracy at >0.99, very low loss



```
python train.py capsnet_2d --epochs 20 --slice_width=28 --dataset=mnist --batch_size=32
```

# OBSERVATIONS

- Capsnet has broader attention, while Resnet focusses smaller spots



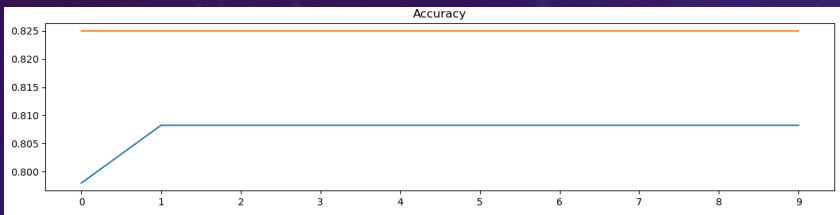
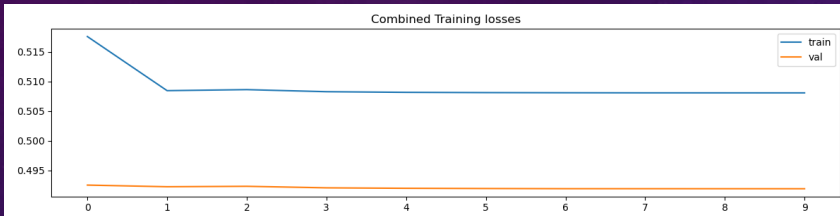
Capsnet



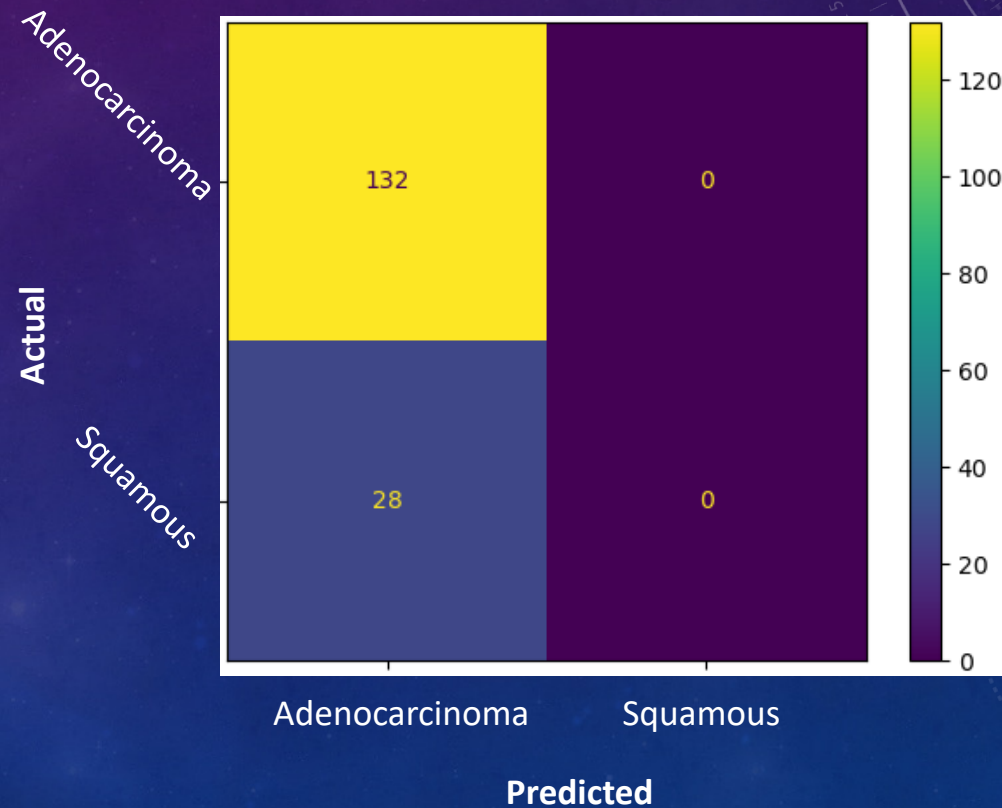
Resnet



# EVALUATION ON LUNGPETCTDX – 2D CAPSNET

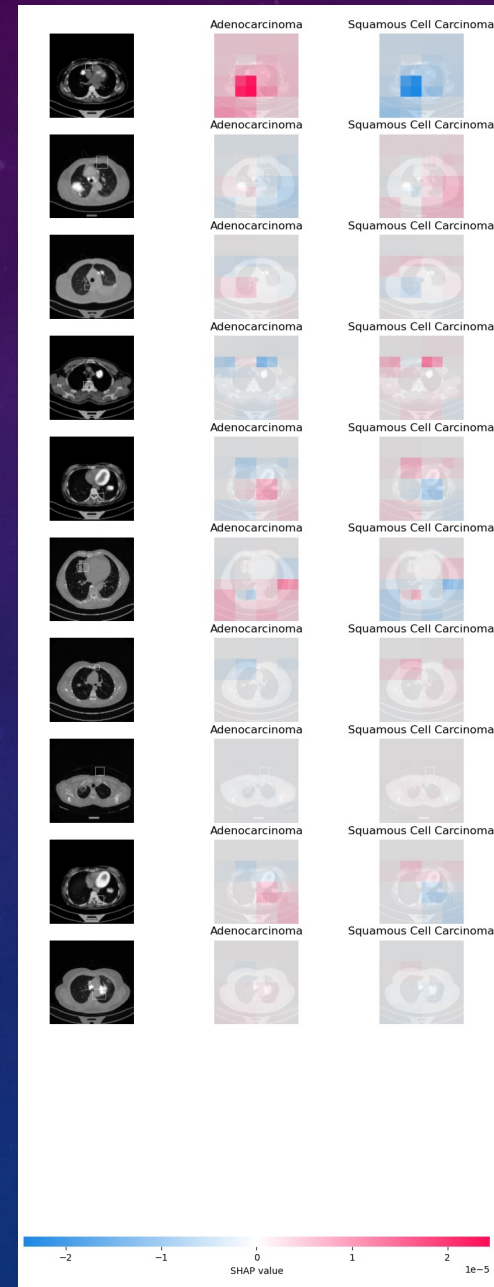


😓 Does not train at all  
- Similar results for Capsnet 3D and Resnet



# EVALUATION ON LUNGPETCTDX

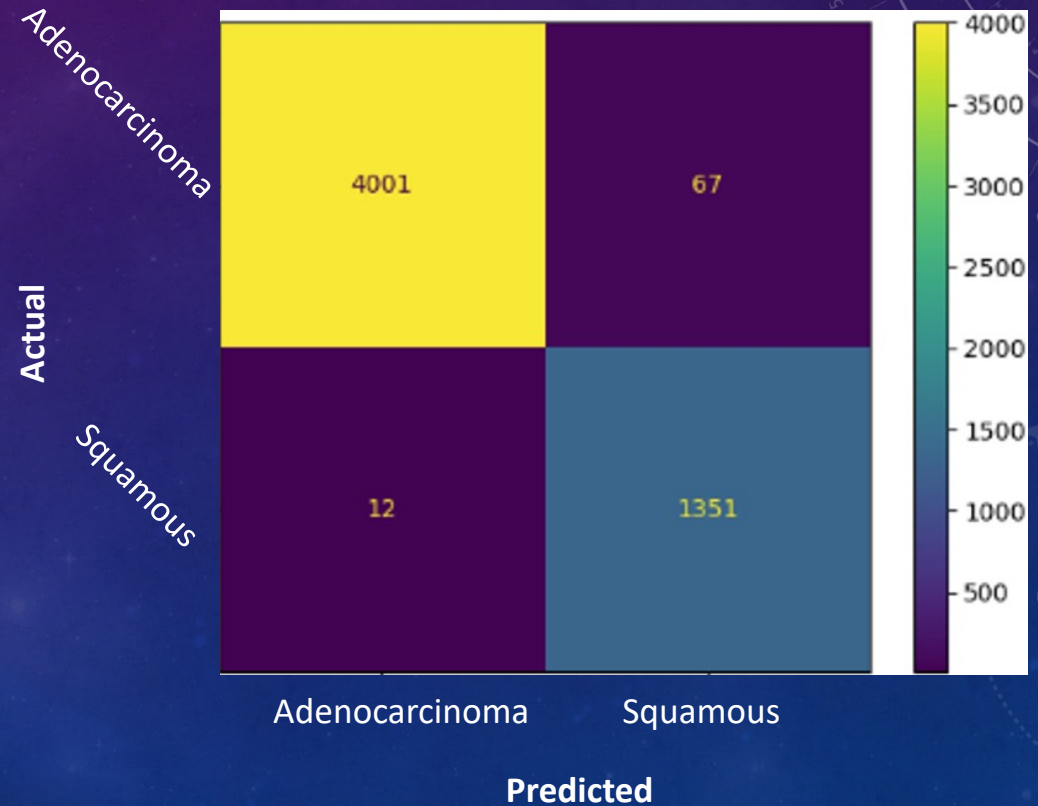
- SHAP value very low for all samples
- „random“ behaviour → not focussing on the tumor
- Changes tried:
  - Different Capsulenet implementation, increasing model complexity
  - Preprocessing (linear & standard normalization, per image or over all)
  - Resizing to higher or lower resolution
  - Different optimizers (Adam, RMSProp, SGD)
  - Hyperparameter optimization
  - Using 3D CT-scans
  - Padding / Cropping to the tumor
  - Various ways to deal with imbalance (Undersampling, class weights, ...)
  - Validate on train set
- No significant improvement ☹️





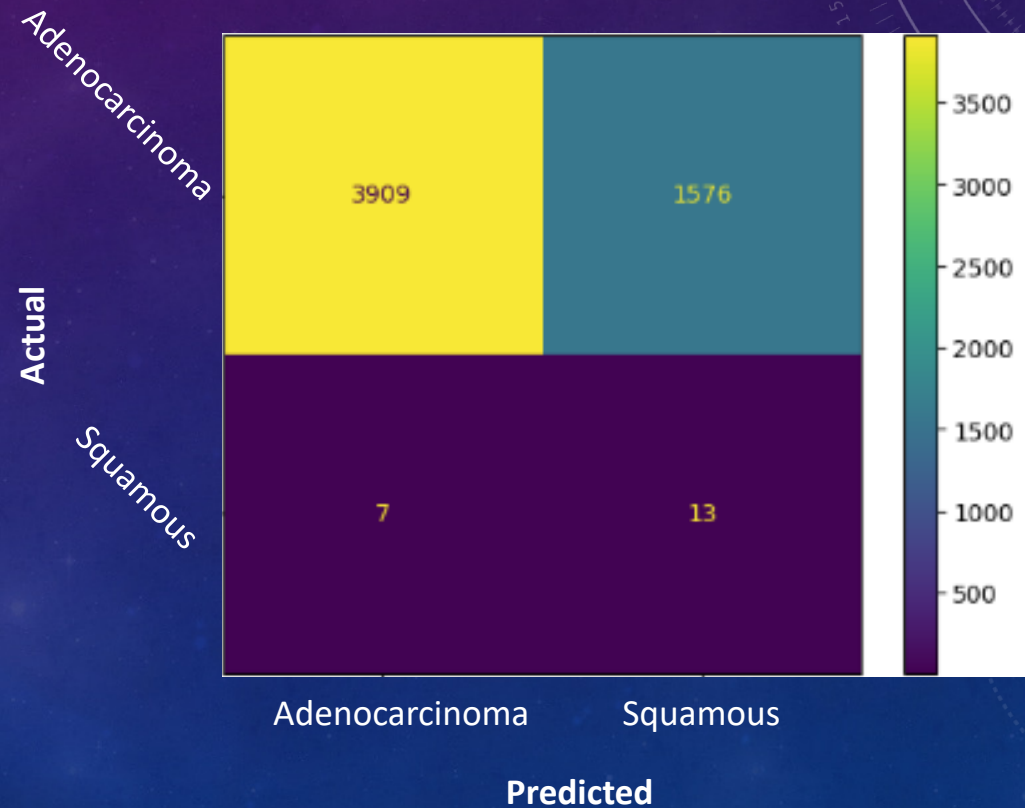
# DEEP DIVE: RANDOM SPLIT

- Model can learn case specific images / features  
→ It performs well
- Model complexity seems to be high enough to remember images
- Training loop is working



# DEEP DIVE: CROPPING TO TUMOR

- The tumor only affects a small part of the whole lung
  - It might perform better if cropped to only the tumor
  - Many papers combine segmentation & classification, we “simulate” the segmentation
- However, does not affect the results
  - This raises the question if classification on the tumor is possible for this dataset





# CONCLUSION

- Our Resnet & Capsule Net implementations work
- The dataset might have some issues or need different preprocessing
  - We did not find any paper with repeatability in mind that uses Lung PET-CT-Dx dataset
  - question remains: can the tumor be classified based only on CT images?
- Whats left:
  - With the new argument parsing, it is very easy to execute experiments with the pipeline
  - It might be a good idea to look for a dataset with repeatable implementation

Thanks for the weekly support and for listening! :)



# REFERENCES

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- Patrick, Mensah Kwabena, et al. "Capsule networks—a survey." Journal of King Saud University-computer and information sciences 34.1 (2022): 1295-1310.
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