

Use of a Deep Learning Artificial Intelligence Model for Differentiating Alveolar/Bronchiolar Adenomas and Carcinomas of the Lung in B6C3F1 Mice

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Background

- Toxicologic pathology is rapidly transitioning to digitalization. The use of digital pathology (DP) combined with novel technologies such as artificial intelligence (AI) platforms, particularly deep learning (DL) (a subclass of AI), to assist pathologists in diagnosing lesions on pathology slides is increasing.
- Standard diagnostic toxicologic pathology is an expensive, time-consuming, labor-intensive process and hence there has been a growing need to develop faster, automated processes that would reduce costs and improve diagnostic accuracy and consistency to facilitate more efficient workflows.
- Development and use of automated DL/AI algorithms designed to objectively detect and classify proliferative lesions may be useful in assisting with decisional support in performing diagnostic toxicologic pathology.

Objective

- Explore the use of AI/DL for routine diagnostic purposes in toxicologic pathology.
- Develop and train a computer-assisted, automated AI model/algorithm using convoluted neural network (CNN) to assist with diagnosis of mouse lung tumors in digital whole-slide image (WSI) scans of lung tissue sections.
- Apply the trained AI algorithm retrospectively to pathologist peer reviewed alveolar/bronchiolar (A/B) adenoma and carcinoma diagnoses to test the neural network's concordance with the peer reviewed diagnoses.

Methods

- Hematoxylin & Eosin-stained sections of B6C3F1 mouse lung tumors mounted on glass slides from US National Toxicology Program studies were scanned at 40X using a NanoZoomer S60 Digital WSI scanner.
- The digitized WSI (resolution = 0.2473 μ m per pixel) were uploaded to the Aiforia™ Cloud Version 5.1.1 image processing and management platform (Aiforia Inc., Cambridge, United States) for processing with DL/CNNs and supervised learning.
- Using hand-drawn annotations on the digitized images, a supervised, multi-layered, CNN algorithm was created and trained to detect and recognize normal and tumor tissue, and the histological features/patterns that distinguish A/B adenoma from A/B carcinoma.
- The AI algorithm was trained on a diverse and representative data set of 27 WSI of mouse lung tumors to create a generalizable deep learning model.
- Known tissue artifacts were identified and trained into the AI model to exclude them from the analysis as background.
- Finally, the trained AI model was applied to WSI data set of 70 lung tumors (17 A/B adenomas; 53 A/B carcinomas) with known, NTP peer-reviewed diagnoses uploaded to the Aiforia™ Cloud platform.



Scan the code with your smart phone camera to view the results

- Click on a photo
- Select "% Analyze" on the right side of the screen
- Press "LungTumor_Tissue1Ki_FOV500_noaug"
- View the results

https://cloud.aiforia.com/Public/NIH_NIEHS_Herbert/uM3y-F5ftP_YhpUrSag3rz-QNWEHrgoSWuQBbPIUI2g

Legend

- Tissue
- Normal
- Tumor
- Adenoma
- Carcinoma

Methods

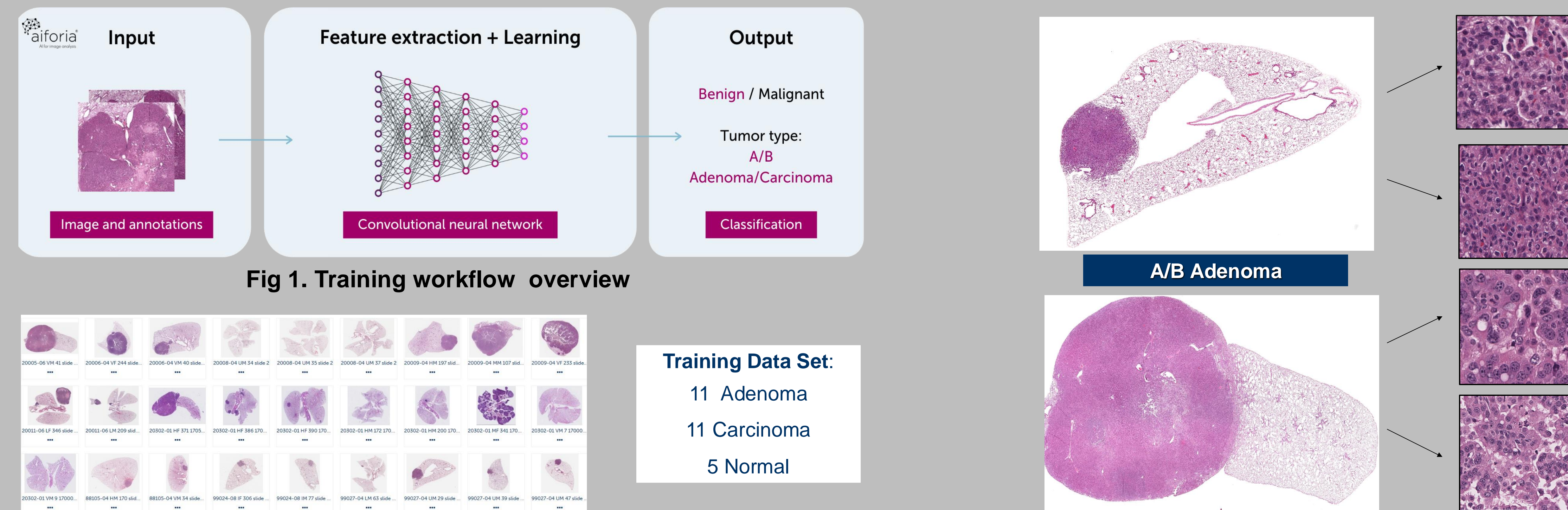


Fig 2. Variability in WSI used in the AI model training image set

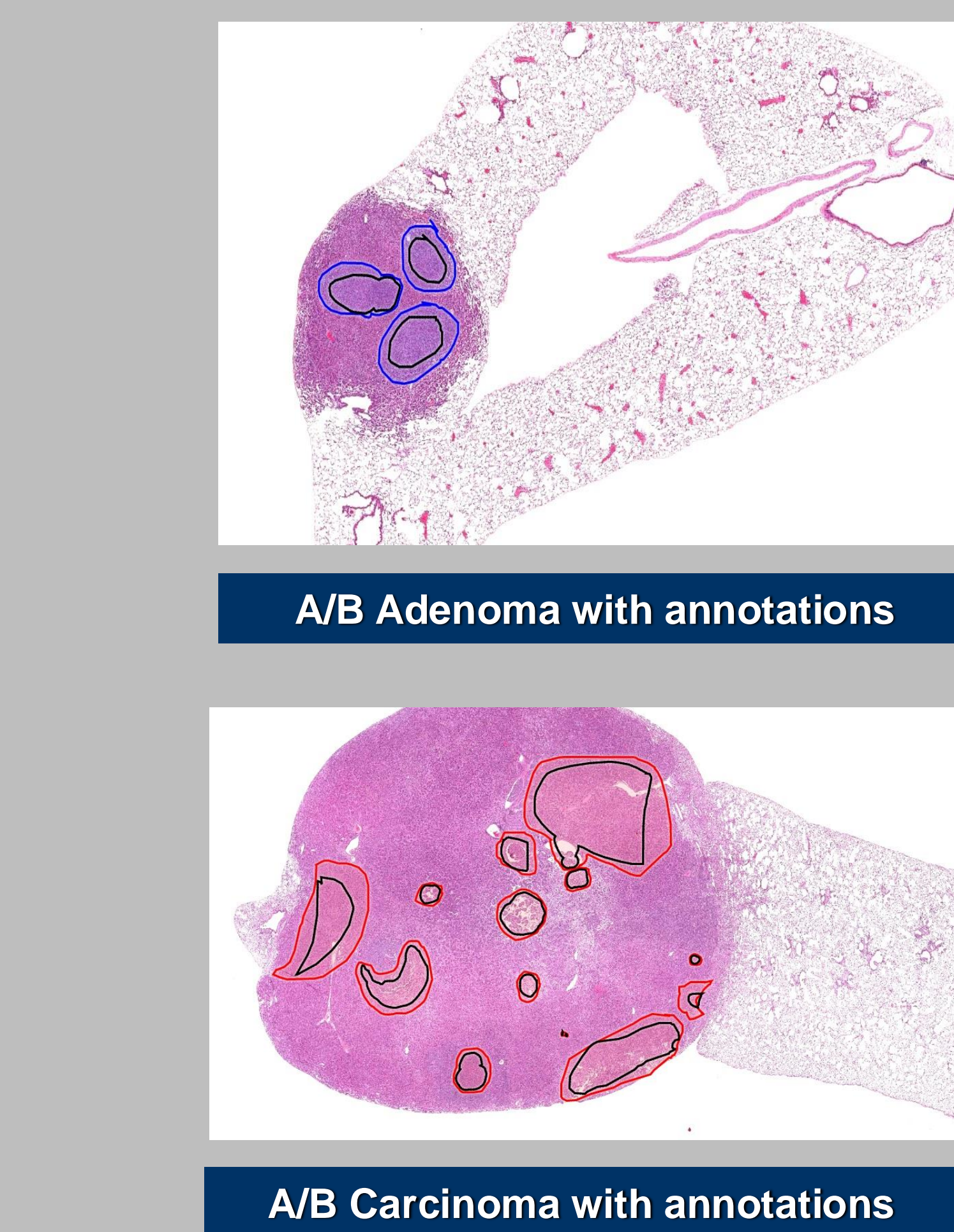
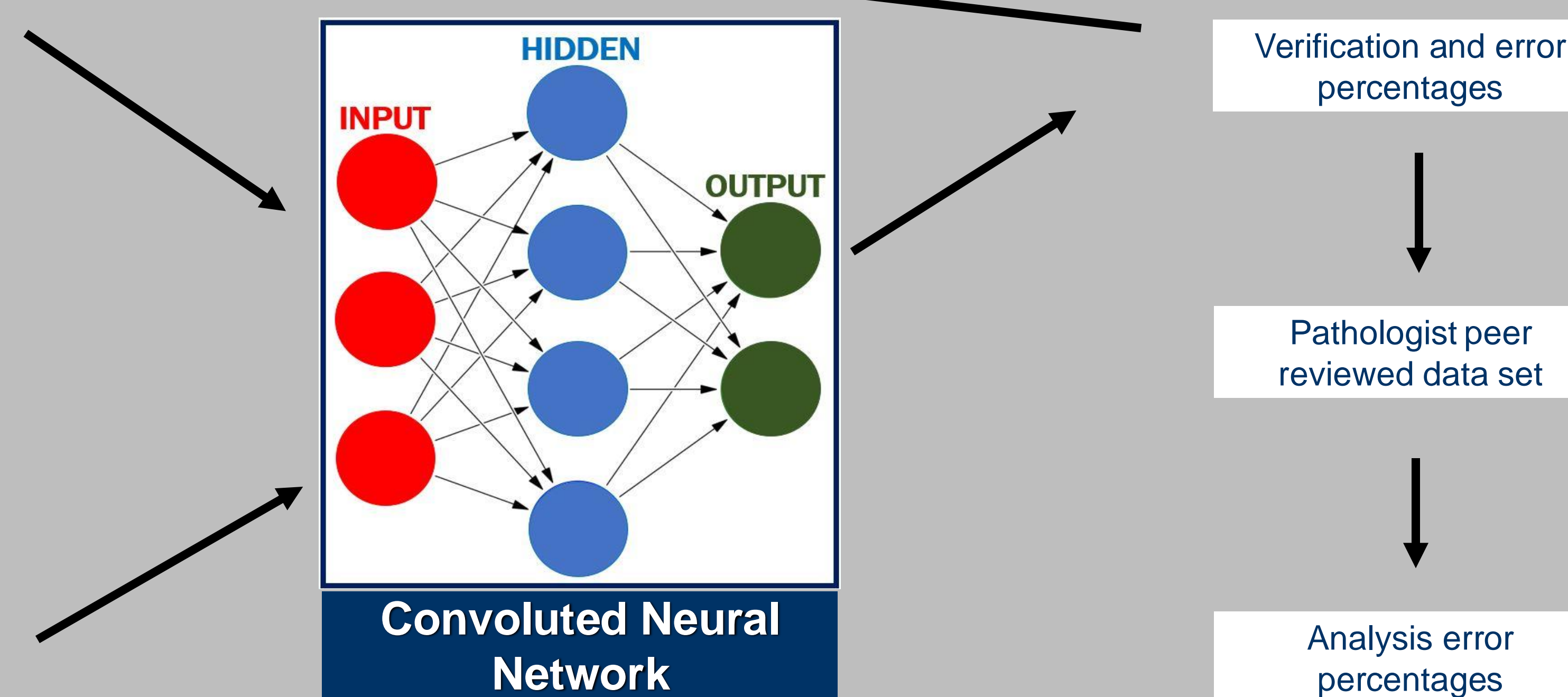


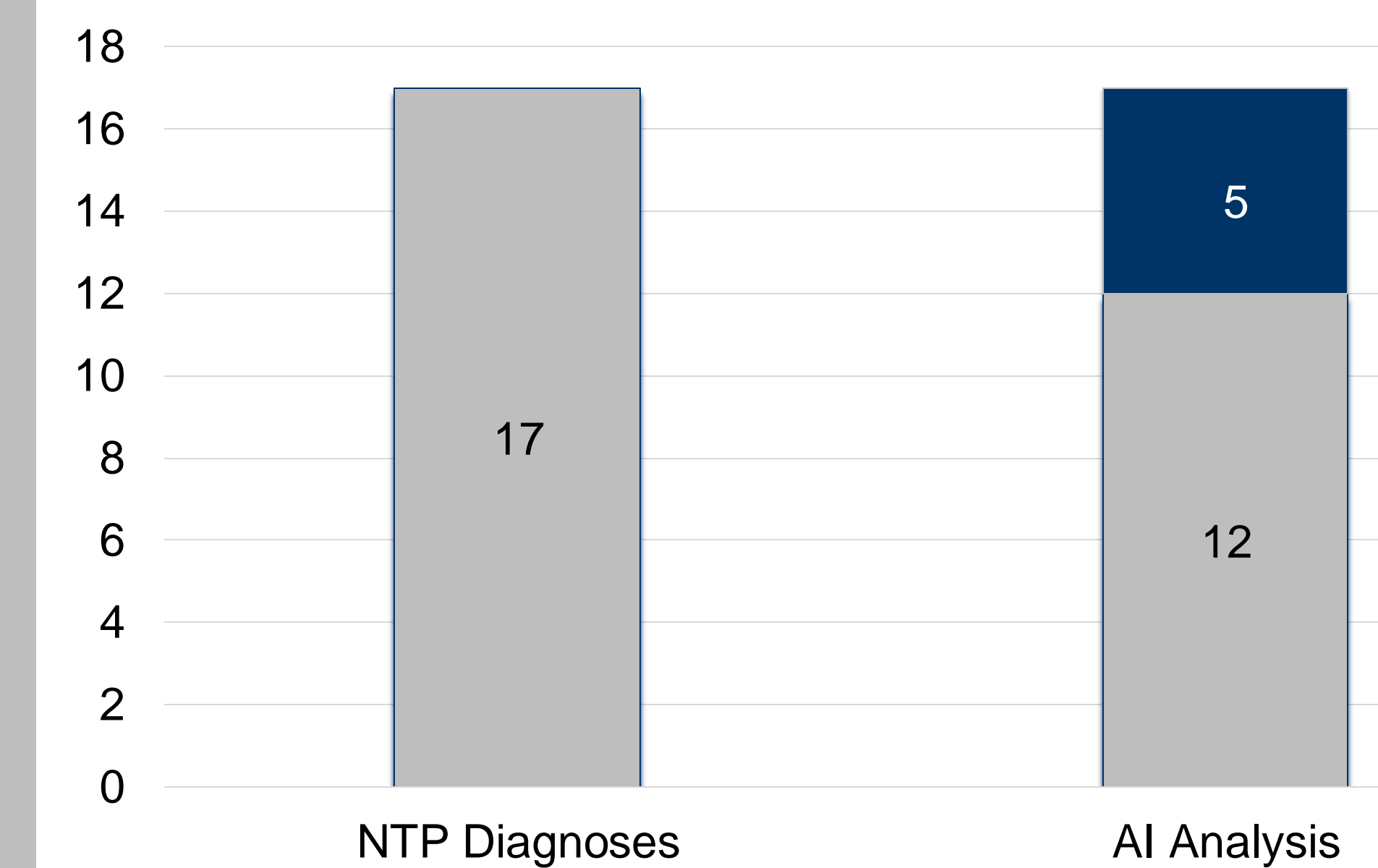
Fig 3. Examples of adenoma and carcinoma patterns



Results

(7a.) Verification Sensitivity Rates:
Tissue layer 99.9%
Tumor/Normal layer: 98.4%
Adenoma/Carcinoma layer: 77.1%

(7b.) Adenoma: 70.6% Sensitivity



Carcinoma: 43.4% Sensitivity

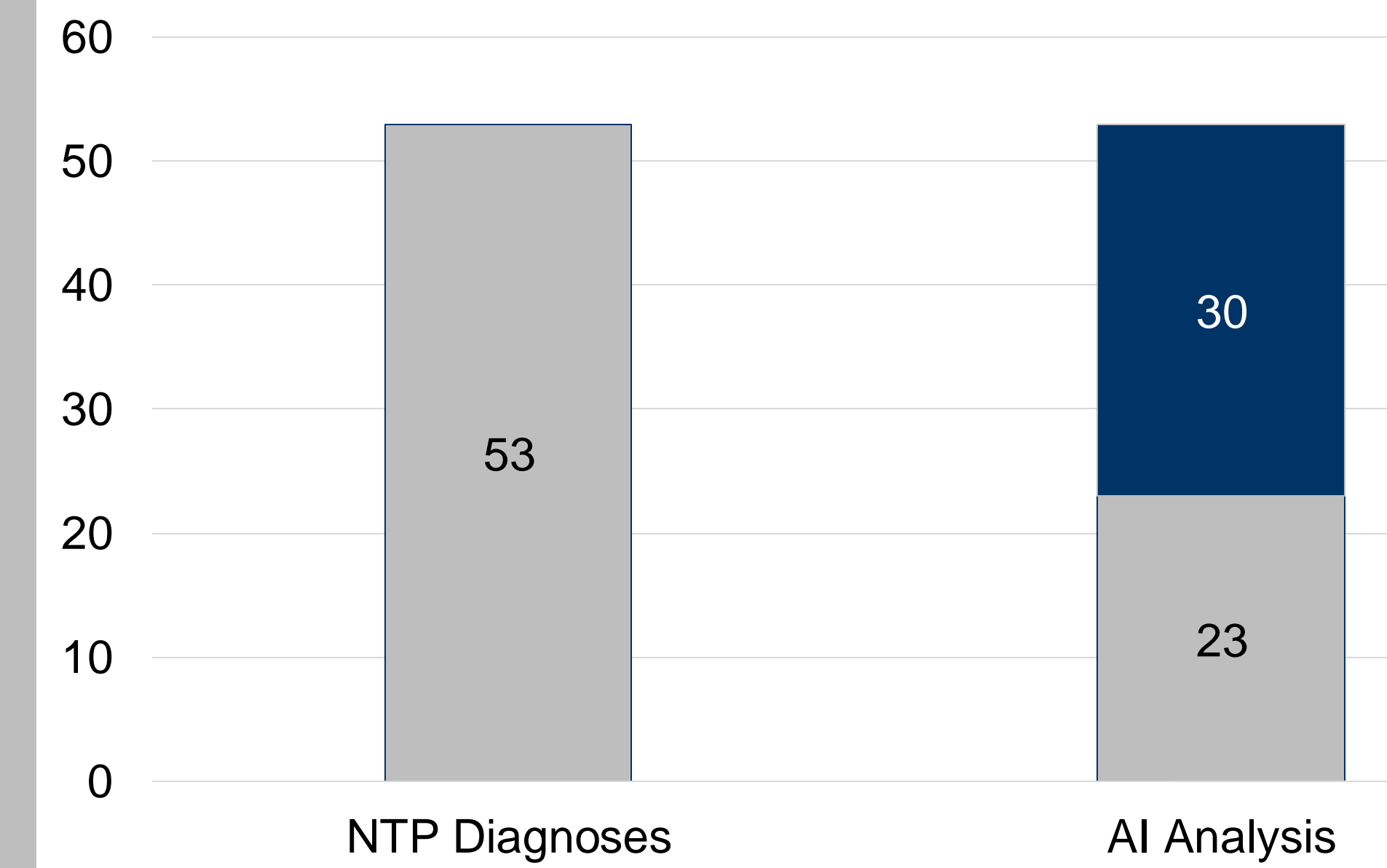


Fig 7b. Comparison of results from NTP diagnoses and AI analysis.

Conclusion

- Results suggest that the CNN algorithm could increase sensitivity for classifying carcinomas, and an automated model to screen, detect, and diagnose proliferative lesions in lung tissue sections from B6C3F1 mice.
- Increasing annotations and images to the data set are likely to improve the CNN to decrease error percentages.

References

- Tokarz DA, *et al* (2021). Using Artificial Intelligence to Detect, Classify, and Objectively Score Severity of Rodent Cardiomyopathy. *Toxicol Pathol.* Vol. 49(4):888-896.
- Rudmann D, *et al* (2021). Using Deep Learning Artificial Intelligence Algorithms to Verify N-Nitroso-N-Methylurea and Urethane Positive Control Proliferative Changes in Tg-RasH2 Mouse Carcinogenicity Studies. *Toxicol Pathol.* 49(4):938-949.
- Turner O, *et al* (2020). Society of Toxicologic Pathology Digital Pathology and Image Analysis Special Interest Group Article*: Opinion on the Application of Artificial Intelligence and Machine Learning to Digital Toxicologic Pathology. *Toxicol Pathol.* 48(2):277-294.
- Gertych, A., *et al.* (2019) Convolutional neural networks can accurately distinguish four histologic growth patterns of lung adenocarcinoma in digital slides. *Sci Rep* 9, 1483.

Acknowledgements

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Results

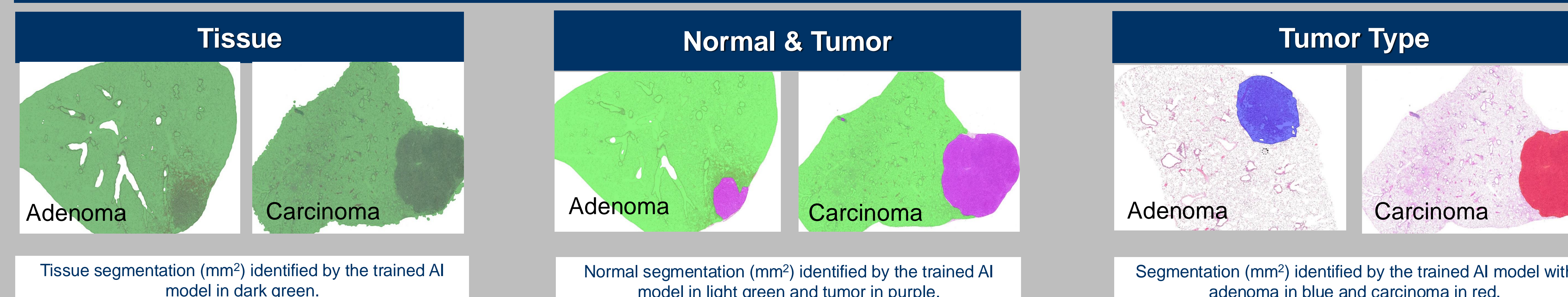


Fig 5. WSI AI analysis results

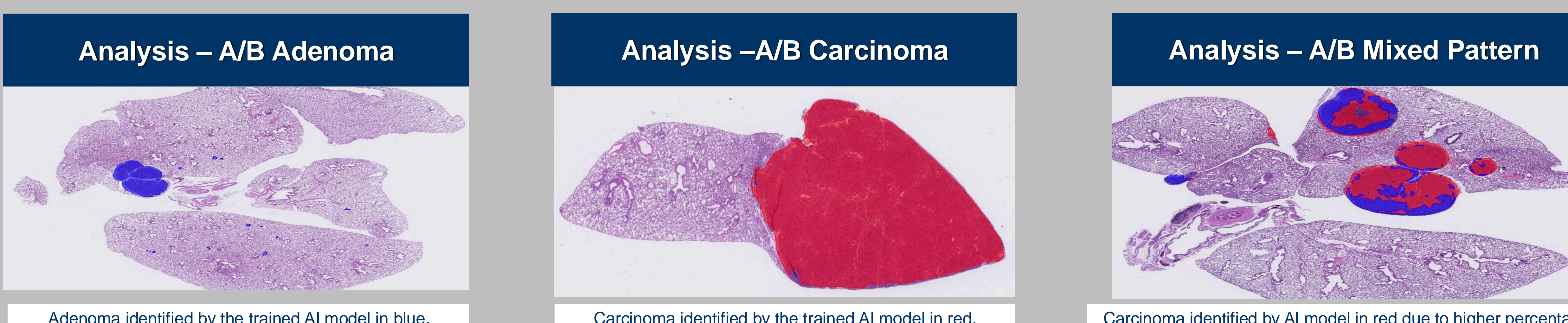


Fig 6. Analysis of NTP study set