

Artificial Intelligence in Digital Pathology from a Computer Scientist's Perspective

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M U N I

Who We Are

RationAI

- a research group at Masaryk University (in Brno)
- 8 senior members, approx. 20 students
- Research in AI in digital pathology, especially explainable methods and clinically relevant problems
- Collaboration with MMCI, IKEM, Med Uni Graz, FN Brno, etc.

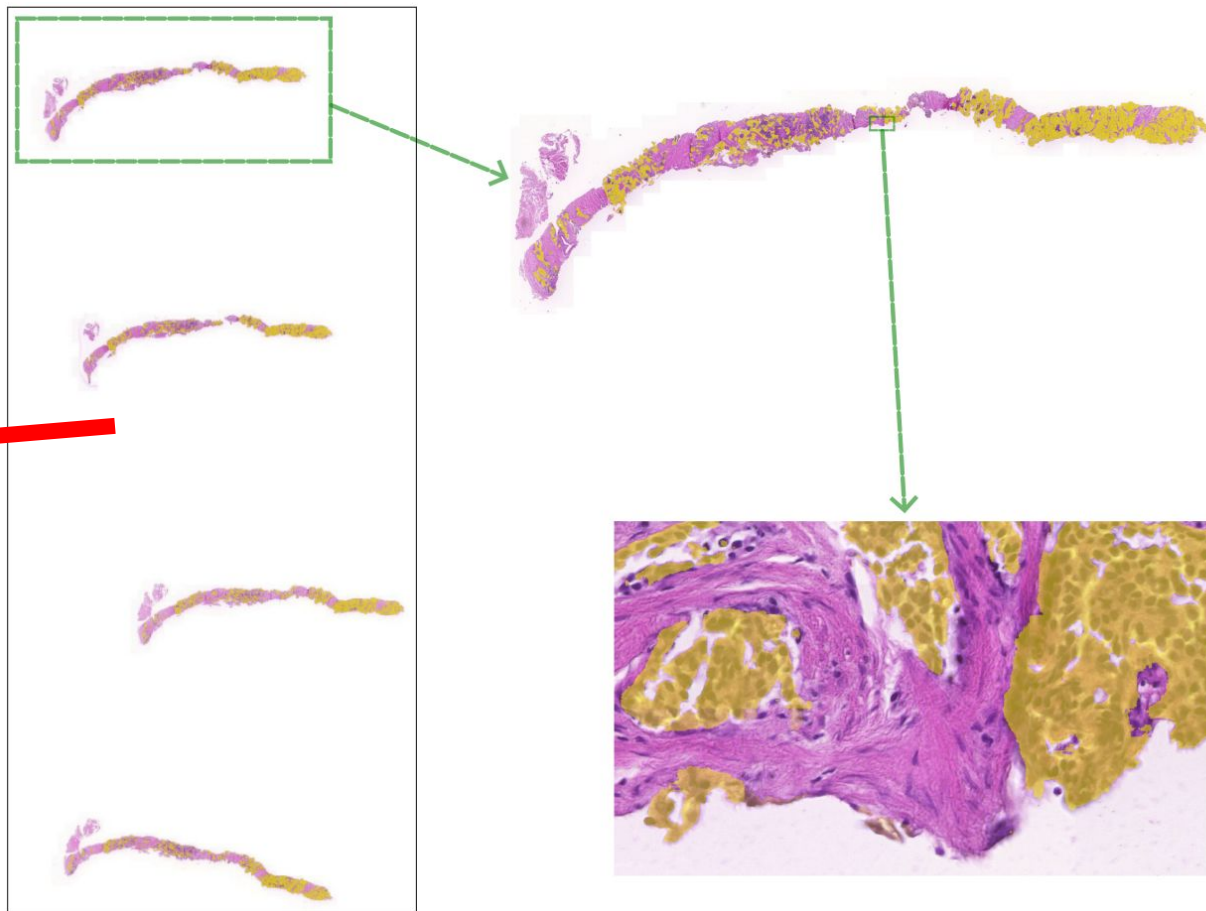
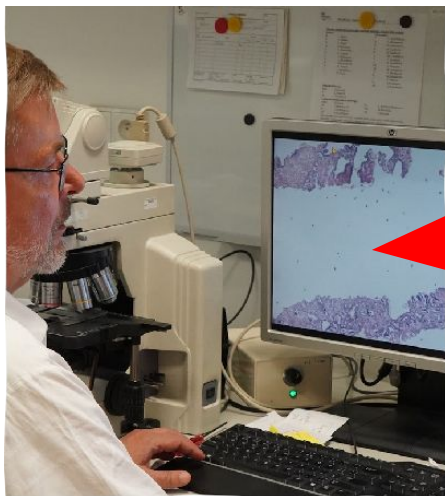


AigoPath

- a startup closely related to RationAI
- *Products*: xOpat viewer enhanced with AI tools



AI in DP

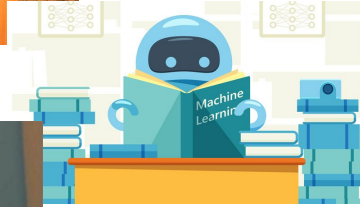


AI learns from data: Here microscopic images (WSI) of tissue labeled with cancer

AI in Digital Pathology

Needs to be

- Developed
- Trained
- Deployed
- Monetized



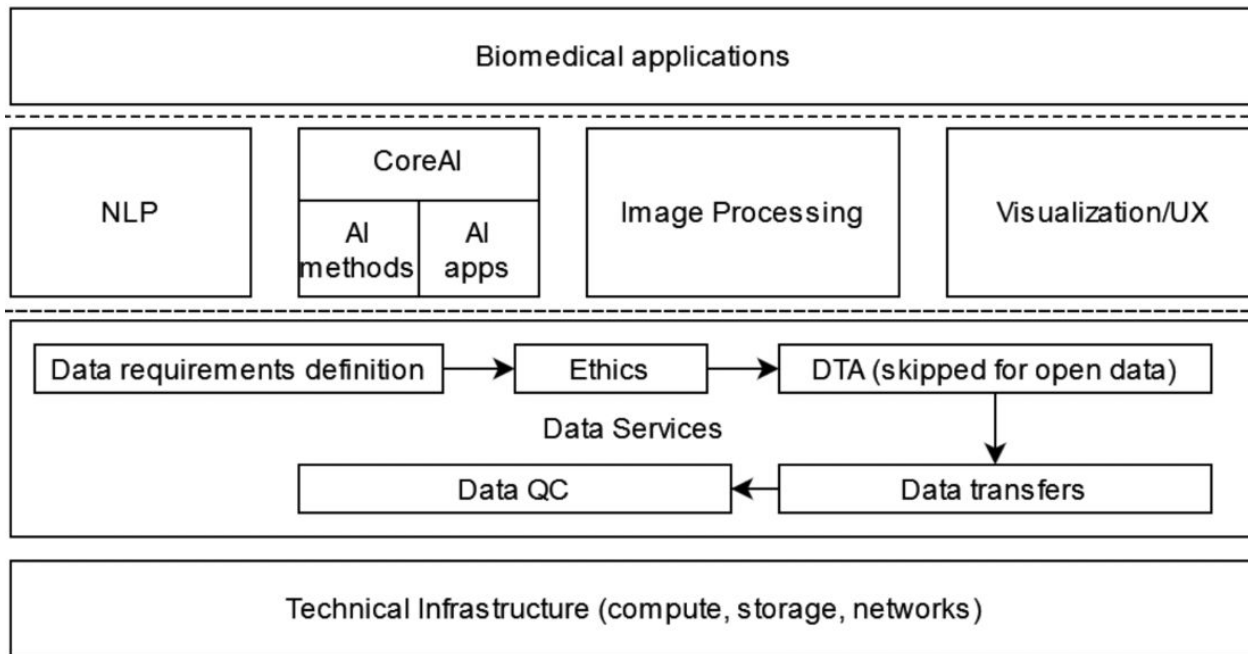
How to do the above in a university research group and a small startup?

Typical AI training workflow - the research group

- The **pathologist formulates** the **medical problem**
E.g. detect prostate cancer in whole-slide images (WSI) of prostate needle biopsies
- The **pathologist selects** the appropriate **cases/slides** for AI training
Cooperation of IT experts and pathologists ...
IT experts do not understand pathology the pathologists are short of time
- The **slides** are **scanned**
 - No standard of WSI format (DICOM, MIRAX, SVS, NDPI, VSI, new proprietary formats)
small group/company does not have official access to the formats
- The resulting data transferred to safe storage
 - smaller projects = thousands of WSI (**terabytes**)
 - large projects = hundreds of thousands/millions of WSI (**petabytes**)
- AI *trained, evaluated, visualized* -> feedback from the **pathologist**

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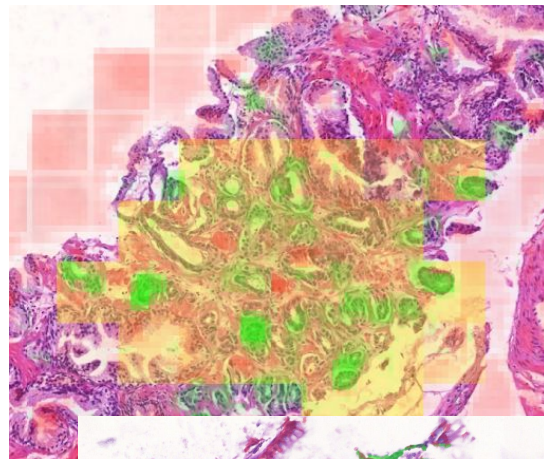
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Some results of RationAI

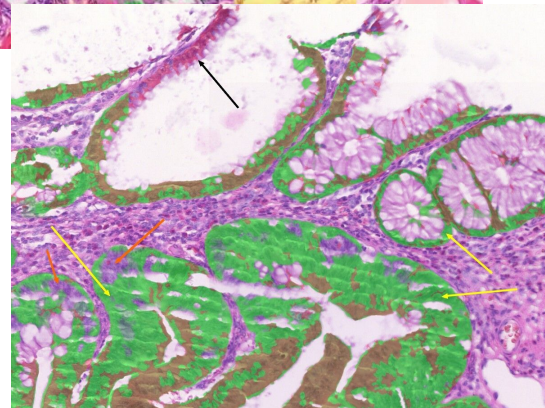
Cancer segmentation

- *Prostate, breast, colon* tumor segmentation
- Explainable
- Using foundation models
(Prov-GigaPath, etc.)
- Tested on multi-centric data



Epithelium segmentation

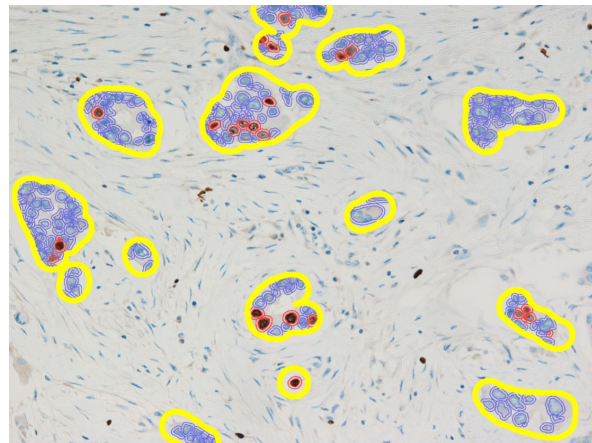
- Trained on double stained WSI
(cytokerating, H&E)
- Various types of tissue, reliable, precise



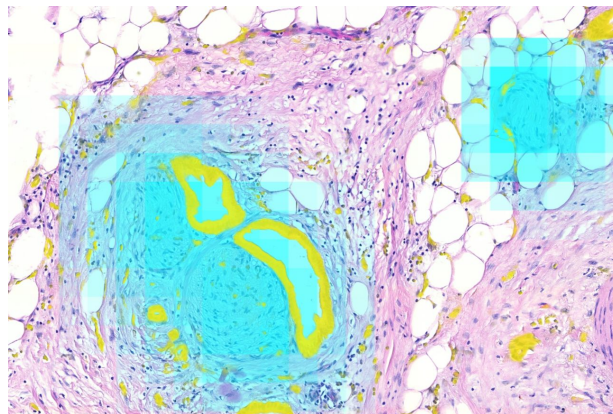
Some results of RationAI

Ki-67 proliferation index estimator in ROI

- Trained on clinical data
no manual annotation
- More precise than usual
“working estimates”



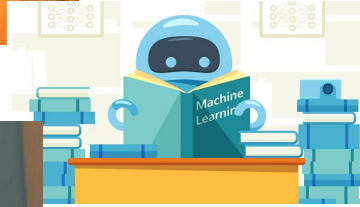
Perineural invasions detection in pancreatic resections



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How to do the above in a university research group and a small startup?

Solved by AI?

Pathologist's view: It does not bother me and sometimes it's helpful


IT expert's view: In 2018

cancer patients with clinicopathological and outcome data available. The results show that deep learning-based outcome prediction with only small tissue areas as input outperforms (hazard ratio 2.3; CI 95% 1.79–3.03; AUC 0.69) visual histological assessment performed by human experts on both TMA spot (HR 1.67; CI 95% 1.28–2.19; AUC 0.58) and whole-slide level

Bychkov et al (2018). Deep learning based tissue analysis predicts outcome in colorectal cancer. *Scientific Reports*, 8

Solved by AI?

IT expert's view: In 2024 (and still in 2025)

Researchers have published many promising algorithmic solutions.^{11,12} However, the path to wide clinical adoption is difficult. A core problem is a lack of standardization and interoperability for the seamless integration of image analysis methods into diverse image management and laboratory information systems. Commercialization and clinical implementation of pathology AI must overcome additional hurdles,^{13,14} namely the transformation of an idea into an AI prototype (which requires data acquisition), a validation process towards market readiness, and certification as a medical product. Finally, reimbursement and billing issues must be solved to generate revenue. 

Prototype deployment:

What is expected from a clinical solution?

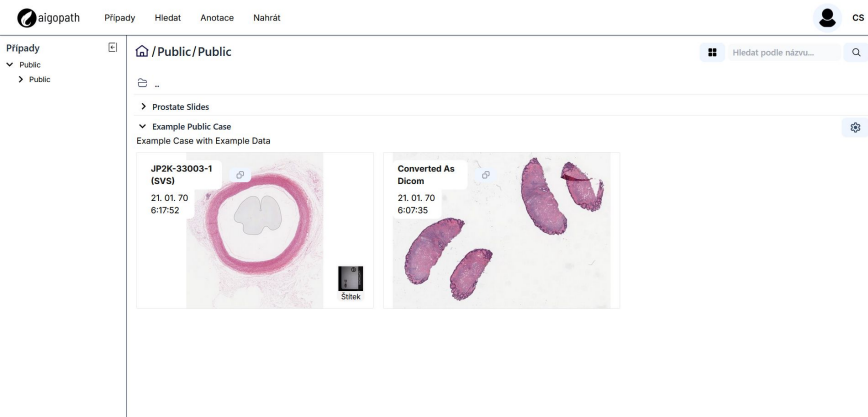
End users (pathologists, lab technicians) usually need

- “Invisible” connection between scanner and the user interface

Quickly

- get images from the scanner to storage
- process them
- visualize results
- **Ergonomy, clarity, reliability of the user interface (UX classics)**
 - Single system for everything - LIS integration
 - Images rendered quickly
- **Necessary tools at hand**
 - Annotation, morphometry, area measurement tools
 - Reporting system (possibly pre filled by the AI system)





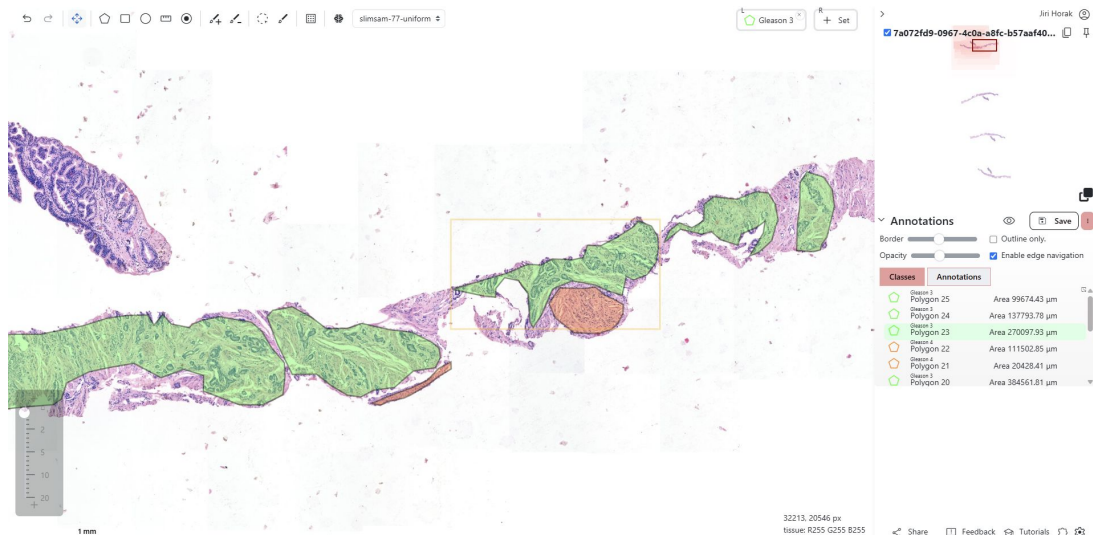
Our Deployment at MMCI

Case browser

- Execution of AI models
- Metadata examination
- Can be substituted by LIS

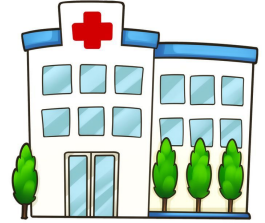
xOpat viewer

- Fast and reliable
- Advance visualization of AI outputs, annotations
- Measurements, morphometry



Prototype deployment:

What is expected from a clinical solution?



Institutional needs (laboratory, hospital)

- Security of processed data and access control

Dependent on the solution: Cloud vs on-premise (next slide)

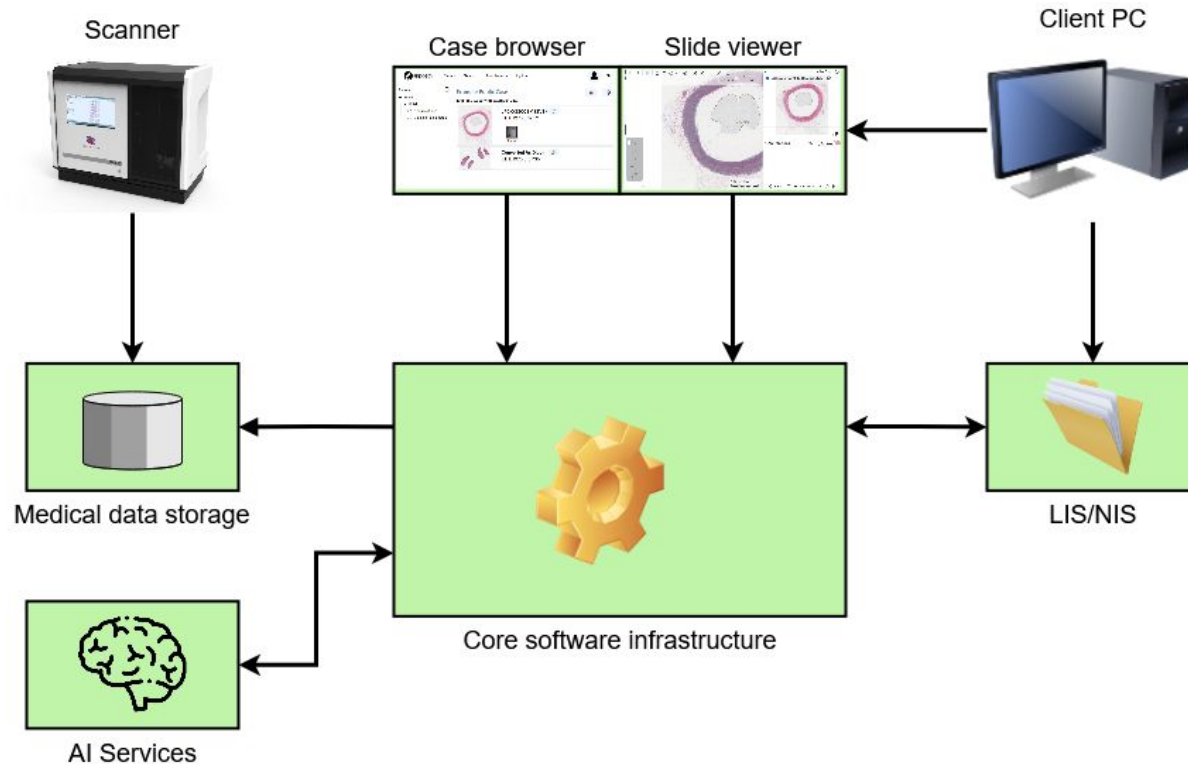
- Certification (ISO, MDR, IVDR)

For small startup: Expensive, complicated, opaque

- Integration with existing processes

Typically the most complicated part: User education, integration with other software solutions, cooperation of IT departments, etc.

Prototype deployment: Infrastructure setup



Question: Have all this on-premise or in cloud?

Prototype deployment: Self-managed vs cloud-based infrastructure

Self-managed

- Complete control over the data
- Easier security measures
- Fast access to the data (short network travel time)
- Possible to optimize costs
- **Requires dedicated personnel**



Cloud-based

- Hardware managed for you
- Dedicated technicians not needed
- Complex services such as backups, disaster recovery, etc.
- **Secure clouds are expensive**



Smooth integration

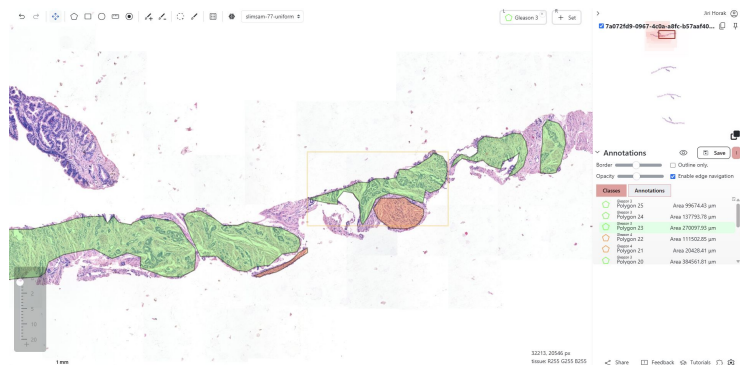


Multiple separate applications: Cumbersome, annoying, and potentially dangerous

Few providers manage to deliver smooth integration with LIS/NIS

xOpat viewer is *lightweight and web-based*

-> ready for integration with larger systems

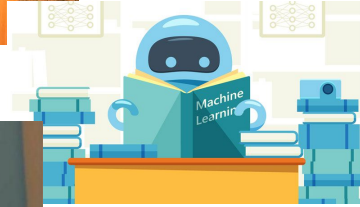


First integrated solution with DS Soft Olomouc was developed in **two weeks!**

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How to do the above in a university research group and a small startup?

The market - AI company view

- The usual **clients**

- Hospitals
- Private labs



- The usual **providers**

- Scanner manufacturers (Roche, Leica)
- Companies from related disciplines (Sectra, Dedalus, Fujifilm)
- Startups and dedicated companies (Ibex, Proscia, Smart In Media, Pathozoom)
- Majority of LIS/NIS providers did NOT jump in yet!

- The **market** is relatively **small** - compared, e.g., with radiology
- *Specialized* - only larger institutions have pathology departments
- Tender based acquisition in state-owned hospitals

Institutional issues when buying DP (AI) solution

Which scanner to get?

- Expensive, long-term purchase
- Can be borrowed for a trial
- Who will operate it

Where to keep the digital images?

- Self-managed in-house servers
- Cloud storage (or even compute)

From whom to get the software?

- Large differences in service and products
- Scanner manufacturers vs sw only companies
- Large companies often less amenable to customizations
- Localization, customer support



Institutional issues when buying DP (AI) solution

How much data will we work with digitally?

- Slides per day, gigabytes
- Scanner and storage space

How users will access the digital slides?

- Access from home or purely on-site?
- On-site network capacity must be sufficient
- Hospital security allowing remote access

Is LIS/NIS integration desirable?

- Willingness of vendors to accommodate other vendors
- Smaller providers usually more flexible and cheaper
- Larger providers offer more complex services



Conclusions

- We have considered **challenges** and **opportunities** encountered by a small research group and a small company in AI in digital pathology
- The **research** in the area is much faster than **testing, deployment, certification, and market release**
- AI will assist digital pathology only when sensibly integrated within processes of pathology departments and labs

Now comes LLM based
Agentic AI !

