

Application of deep learning in diagnostic pathology: own experience and perspectives

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University Hospital Cologne

Institute of Pathology

What do we do:

- Diagnostics
- Molecular pathology
- Oncology / Precision medicine
 - Research

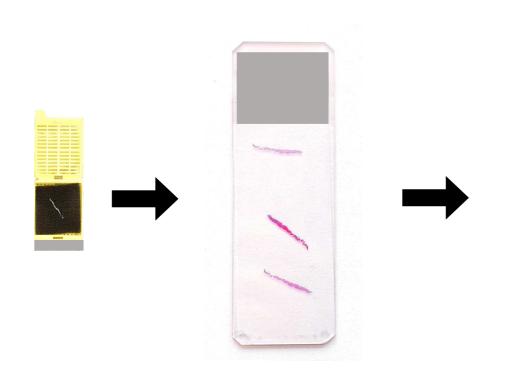




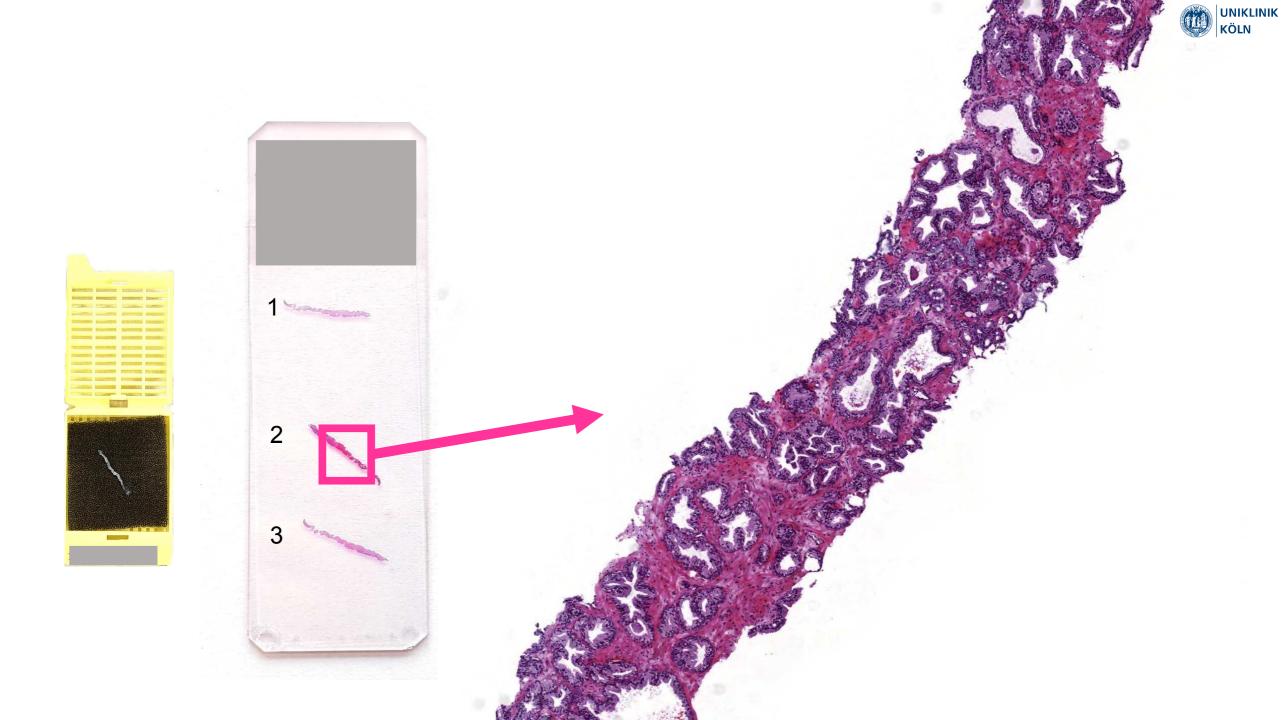


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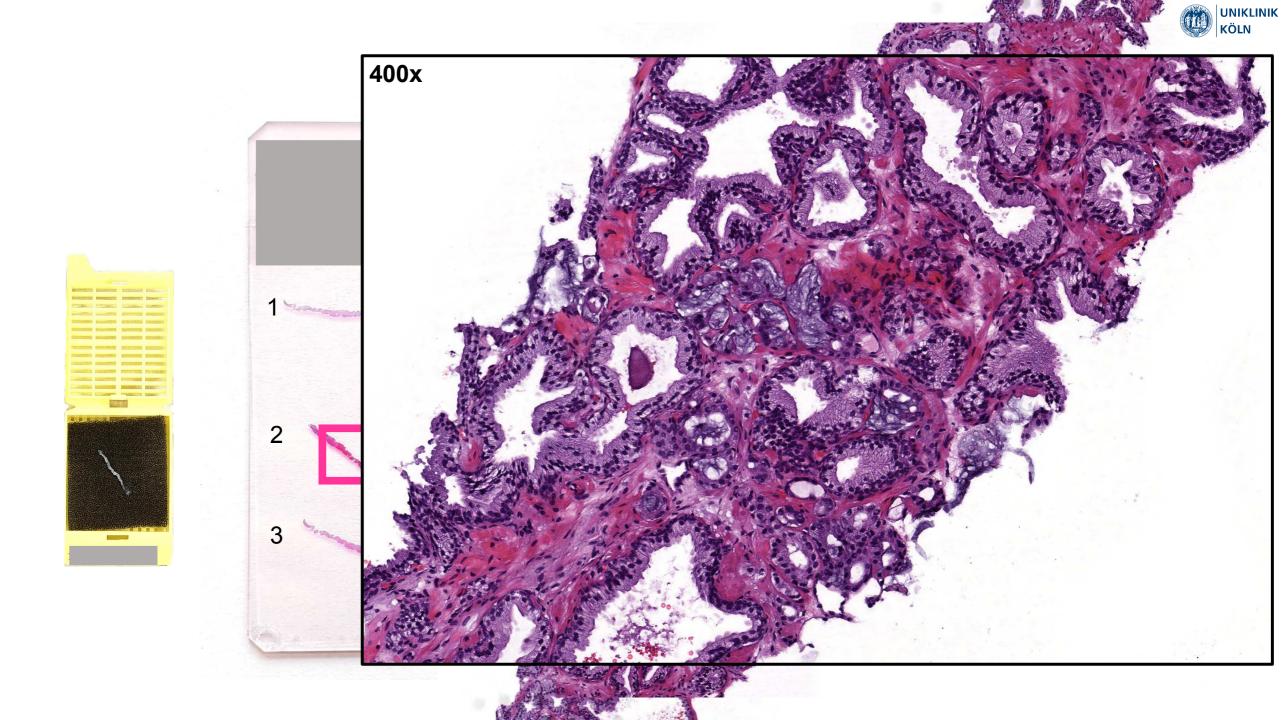


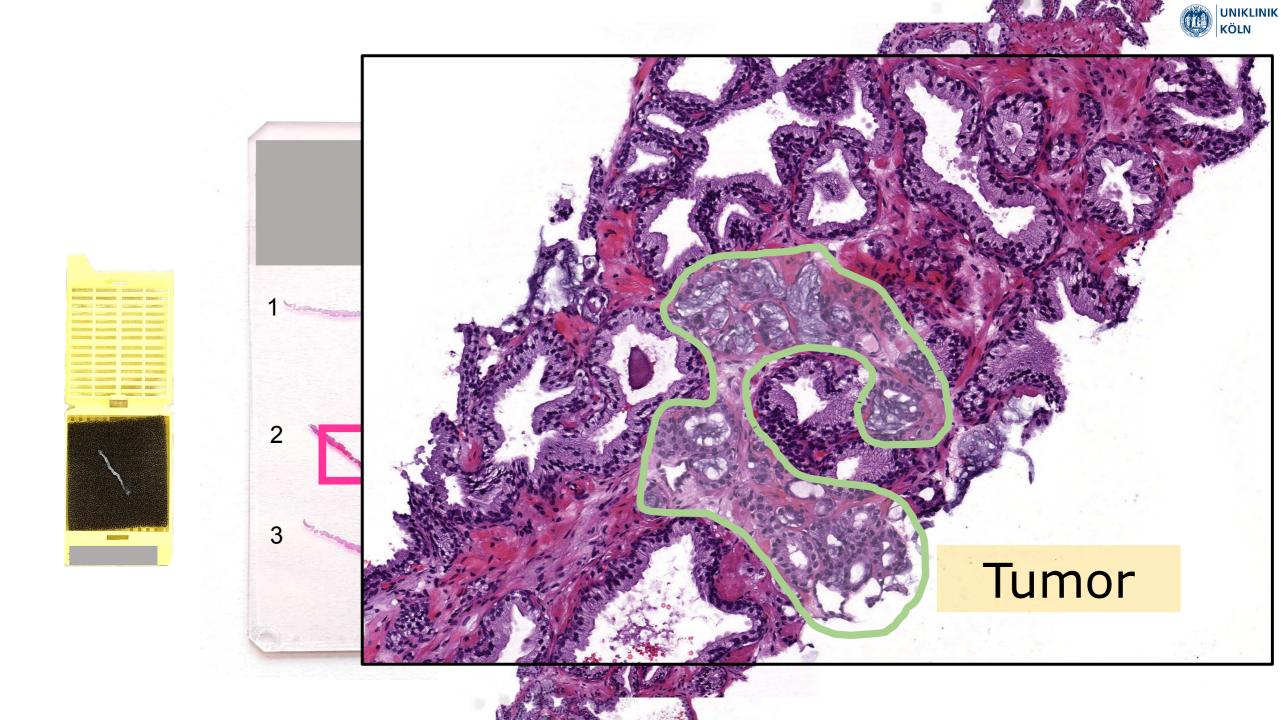








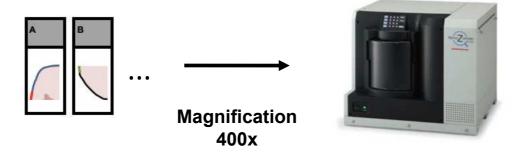




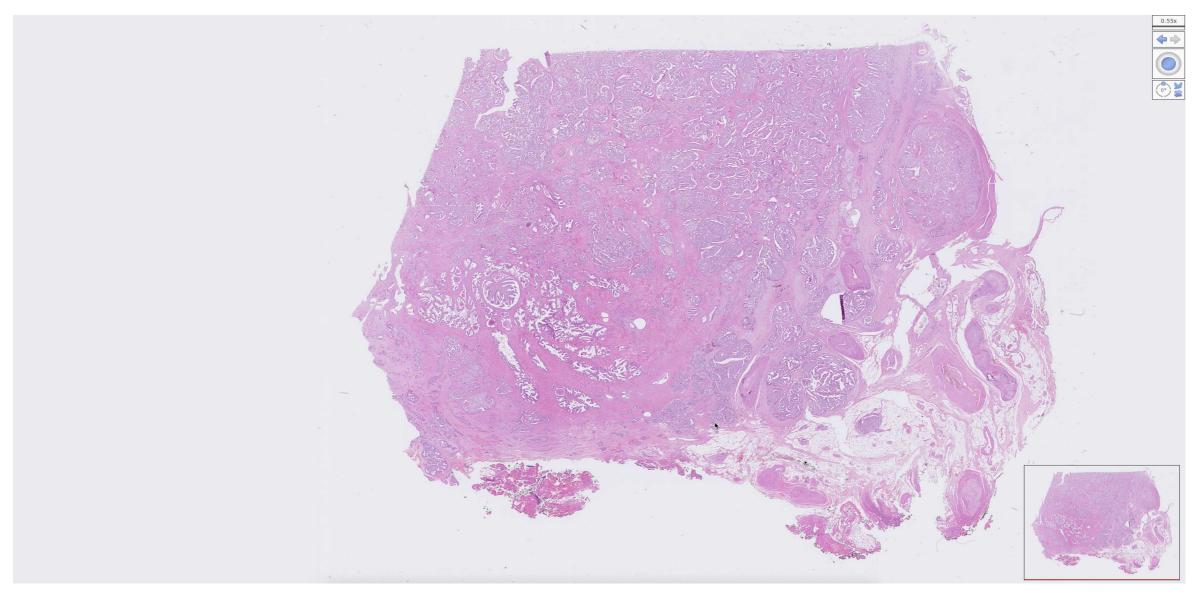


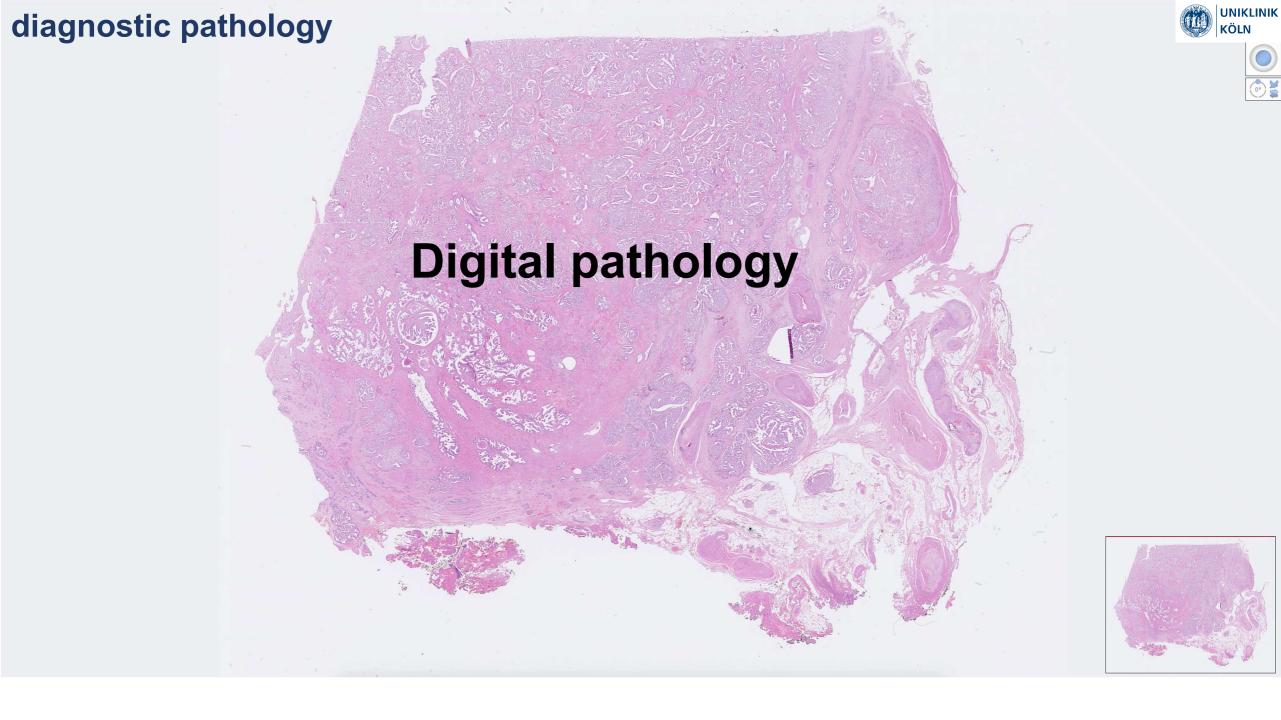




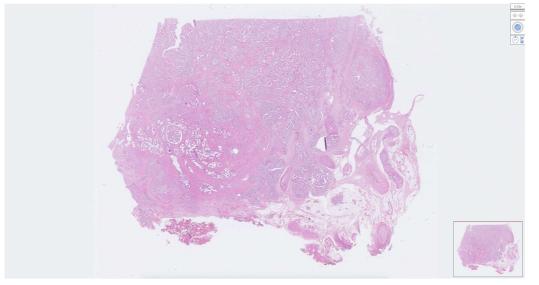






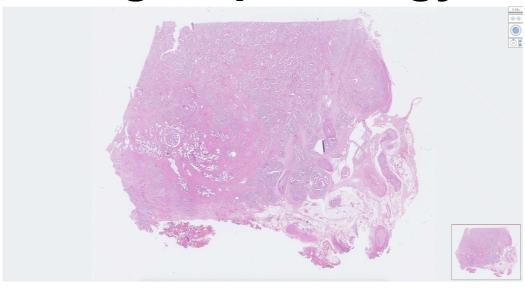






Digital pathology

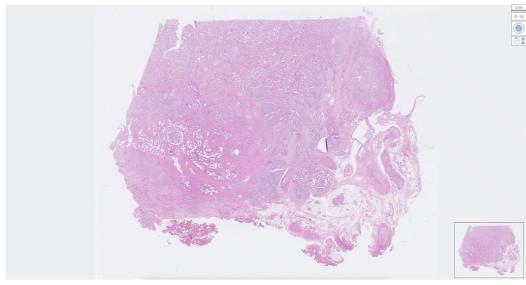




Timing Dimensions Sense

Digital pathology





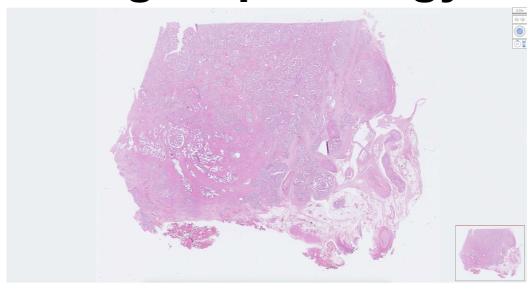
Timing

Transformation next 5-10 years

Dimensions

Digital pathology





Timing

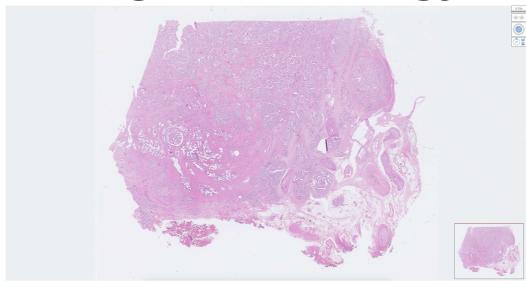
Transformation next 5-10 years

Dimensions

1 Slide = 1-4 Gb

Digital pathology





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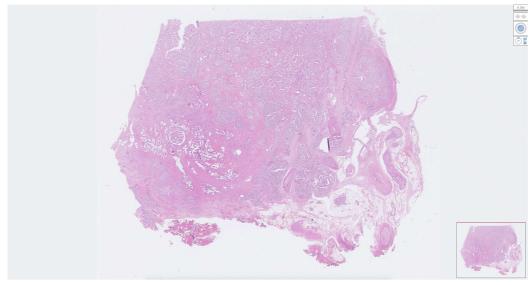
UK Cologne:

>1000 Slides / day

up to 1 Tb / day

Digital pathology





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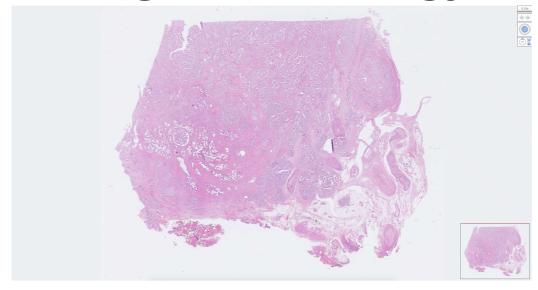
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Very Big Data

Digital pathology







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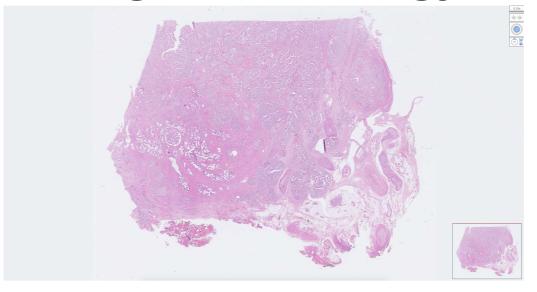
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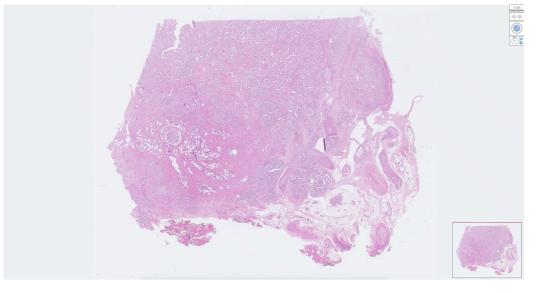
Sense

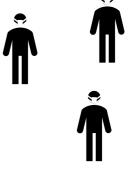
Diagnostic

- Staging (size, local etc.)
- Aggressive?
- Metastasis?
- Resection complete?

Digital pathology







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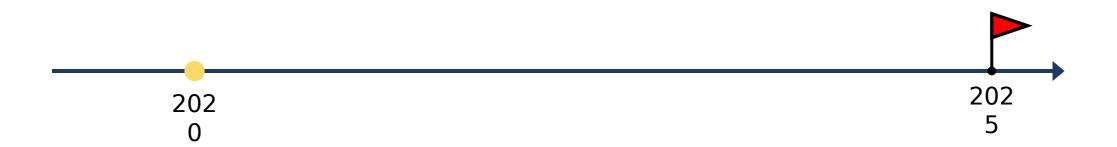
Advanced

- Prognosis (recurrence, death)?
- Metastasis?
- Genetic changes?
- Individual therapy options?
- Response to therapy?

Bulk of evidence

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- 1) "low-hanging fruit" effect
- 2) Wow! effect
- 3) Overestimation of modest results





Technology

Networks:

Classification
Segmentation
GANs
Architectures
Principles of training
AutoML



Bulk of evidence



Technology

Networks:

Classification Segmentation

GANs

Architectures

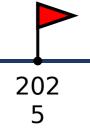
Principles of training

AutoML

Data

202

- Lack of ground truth



Bulk of evidence



Technology

Networks:

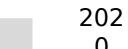
Classification Segmentation **GANs**

Architectures

Principles of training

AutoML

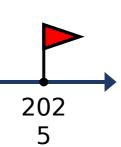
Data



- Lack of ground truth

Performance

- Perspective directions
- Market formation





Diagnostic applications

Tumor recognition Aggressivity grading



Advanced applications

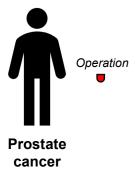














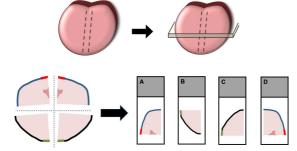


Pathology department



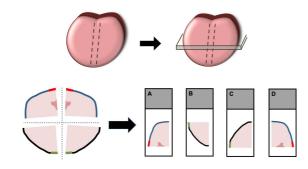


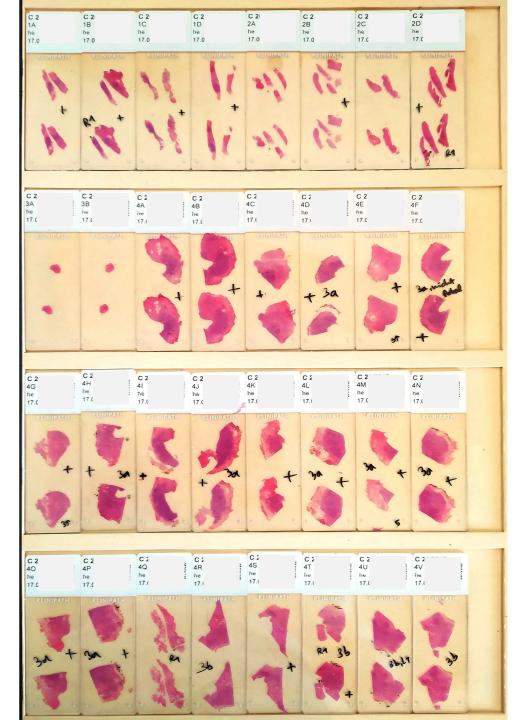






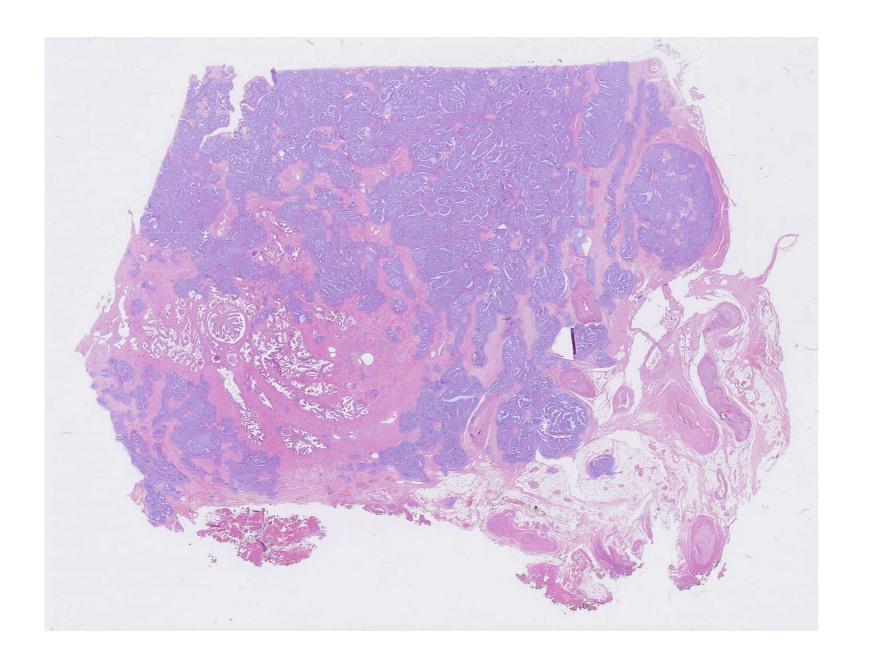






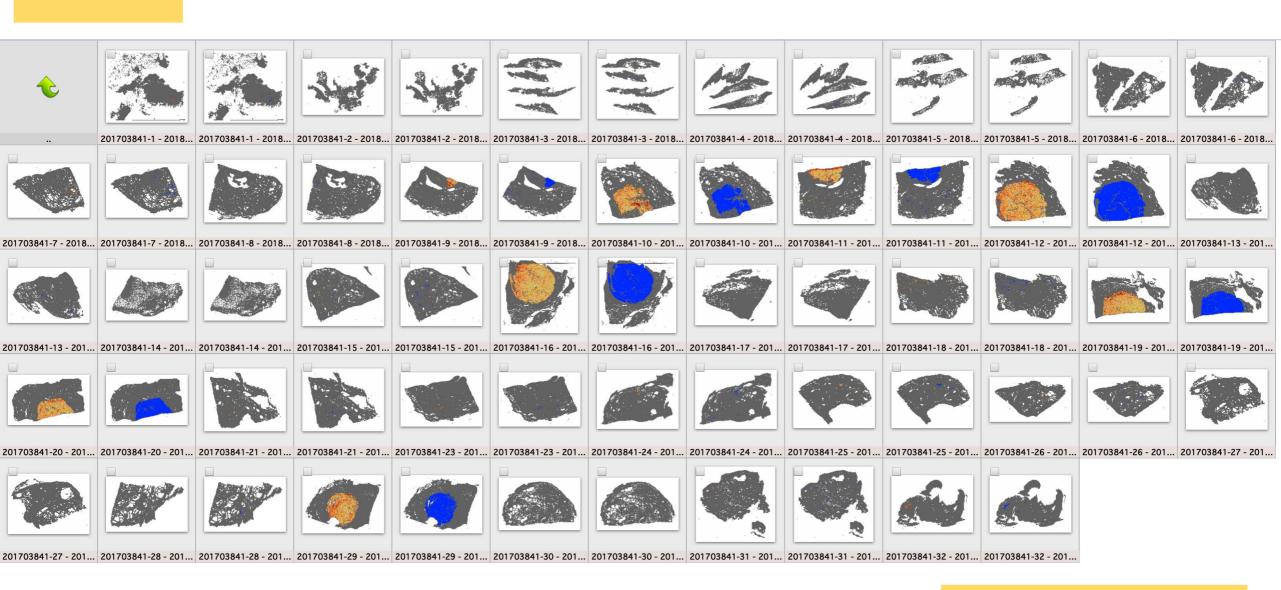






Al Tool







Pathology report

Mi 20:02 UNIKLINIK KÖLN QuPath (0.1.2) - TCGA-G9-6498 File Edit Tools View Objects TMA Measure Automate Analyze Classify Extensions Help ① 0.51x Project Image Annotations Hierarchy Workflow None ■ GP3 ■ GP4 ■ GP5 ■3+3 3+4 **■** 4+3 ■ 4+4 ■ N 5+4 4+5 HGPIN ■ N_STR TU_IDC ■ SB ■ TU_MUC ■ N_NONGL Set class Auto set Delete Delete all Value Key Name

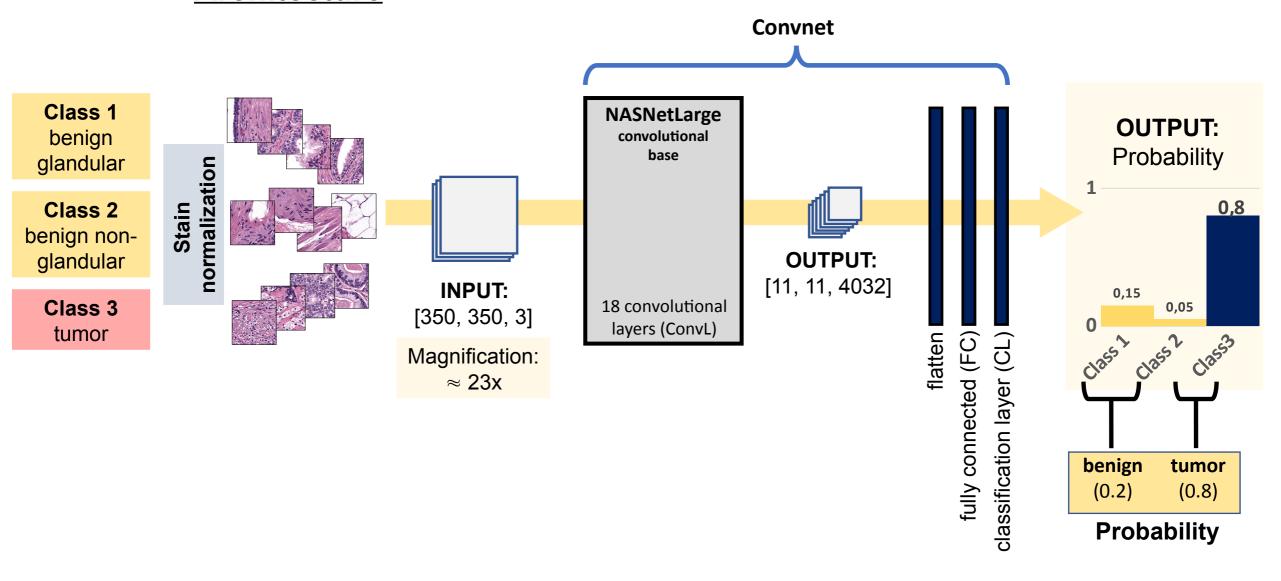
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Mi 20:03 UNIKLINIK KÖLN QuPath (0.1.2) - TCGA-G9-6498 File Edit Tools View Objects TMA Measure Automate Analyze Classify Extensions Help 0 0 N 0.51x Project Image Annotations Hierarchy Workflow None Area (3+4) C Area (GP4) ■ GP3 ■ GP4 O Area (4+3) ■ GP5 Area (3+4) ■3+3 Area (3+4) 3+4 Area (3+4) ■ 4+3 Area (3+4) 4+4 Area (3+4) ■ N Area (3+4) 5+4 (Area (GP4) 4+5 Polygon (GP4) HGPIN O Polygon (GP4) ■ N_STR Polygon (GP4) TU_IDC Area (GP4) ■ SB & Area (GP4) TU_MUC Area (4+5) ■ N_NONGL O Polygon (GP4) Auto set Delete Delete all Set class Value Key Name 11591.76, 6597.56 μm 201, 118, 157

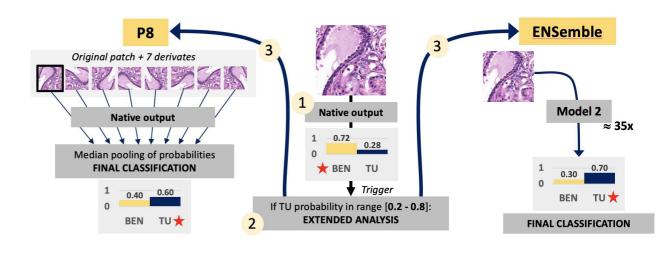
Project Image Annotations Hierarchy Workflow None Area (3+4) G Area (GP4) ■ GP3 ■ GP4 O Area (4+3) ■ GP5 Area (3+4) ■3+3 Area (3+4) 3+4 Area (3+4) ■ 4+3 Area (3+4) 4+4 Area (3+4) ■ N Area (3+4) 5+4 (Area (GP4) 4+5 Polygon (GP4) HGPIN O Polygon (GP4) ■ N_STR O Polygon (GP4) TU_IDC Area (GP4) ■ SB & Area (GP4) ■ TU_MUC Area (4+5) ■ N_NONGL O Polygon (GP4) Delete Auto set Delete all Set class Key Value Name 11668.90, 10558.18 µm 134, 85, 131

Convolutional neural network Architecture

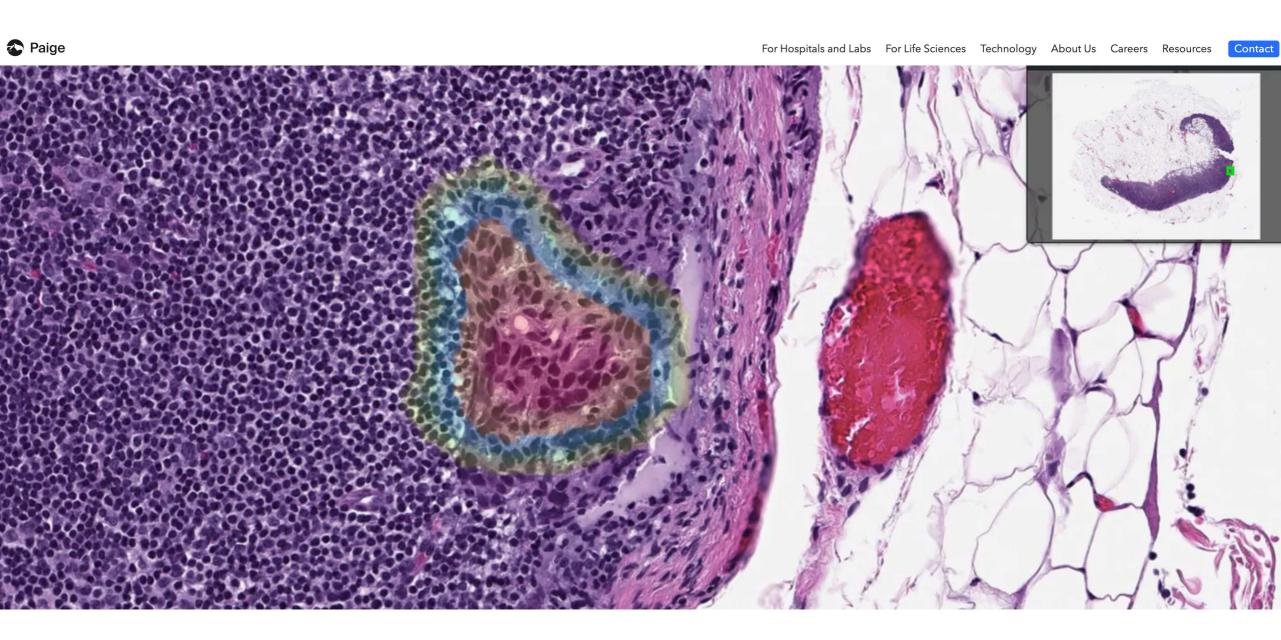




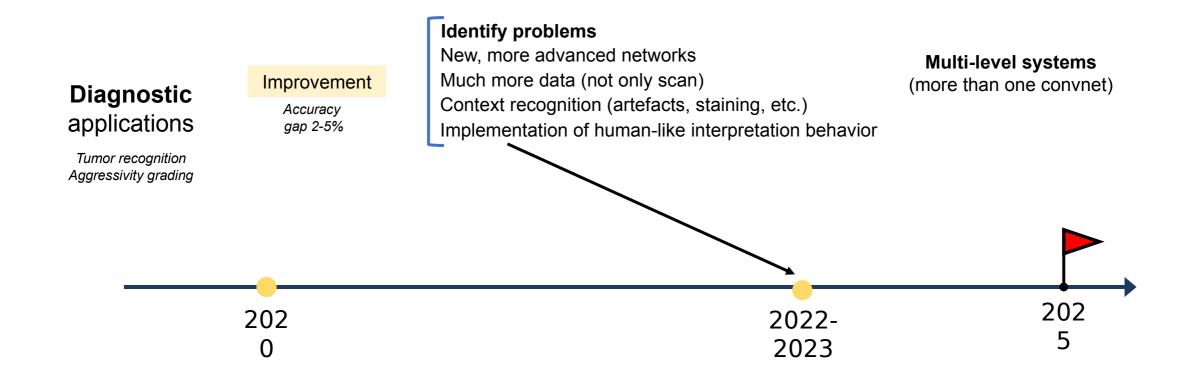
Accuracy Tumor detection



Cohort	Precision	Recall	F1 Score	Overall accuracy
VAL 1 native	0.937	0.961	0.948	96.2%
VAL 2 native	0.927	0.957	0.942	96.7%
VAL 2 <i>p. VAL 1</i>	0.951	0.955	0.953	97.4%
VAL 1 <i>P8</i>	0.950	0.967	0.959	97.0%
VAL 2 <i>P8</i>	0.944	0.962	0.953	97.3%
VAL 1 ENS	0.936	0.967	0.951	96.4%
VAL 2 ENS	0.932	0.962	0.947	97.0%



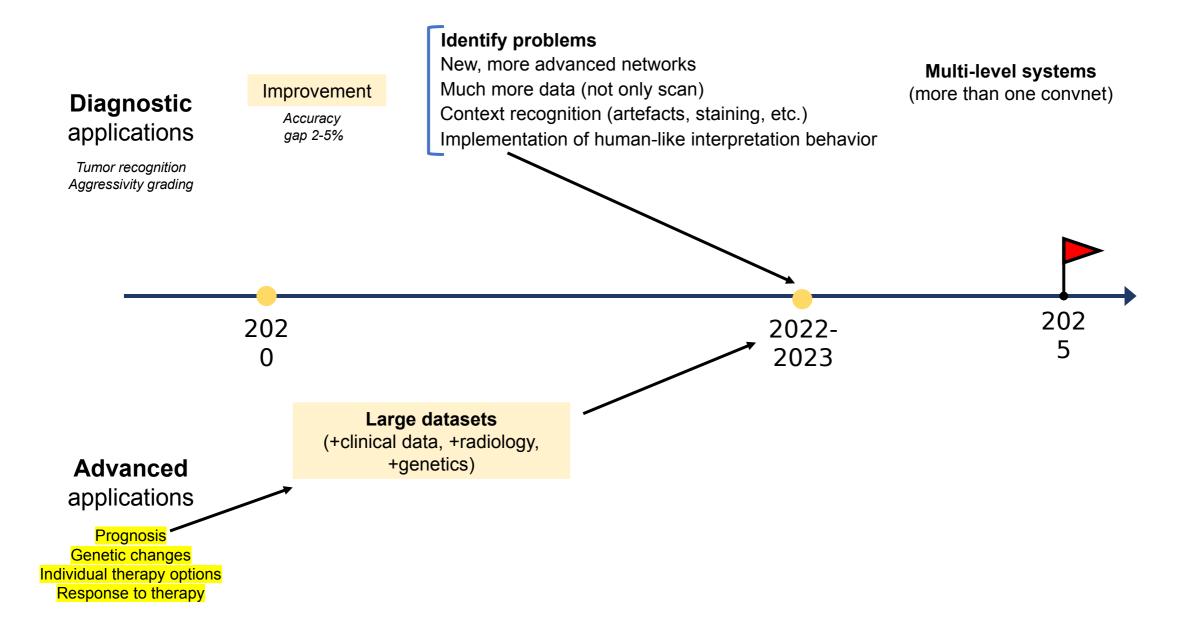




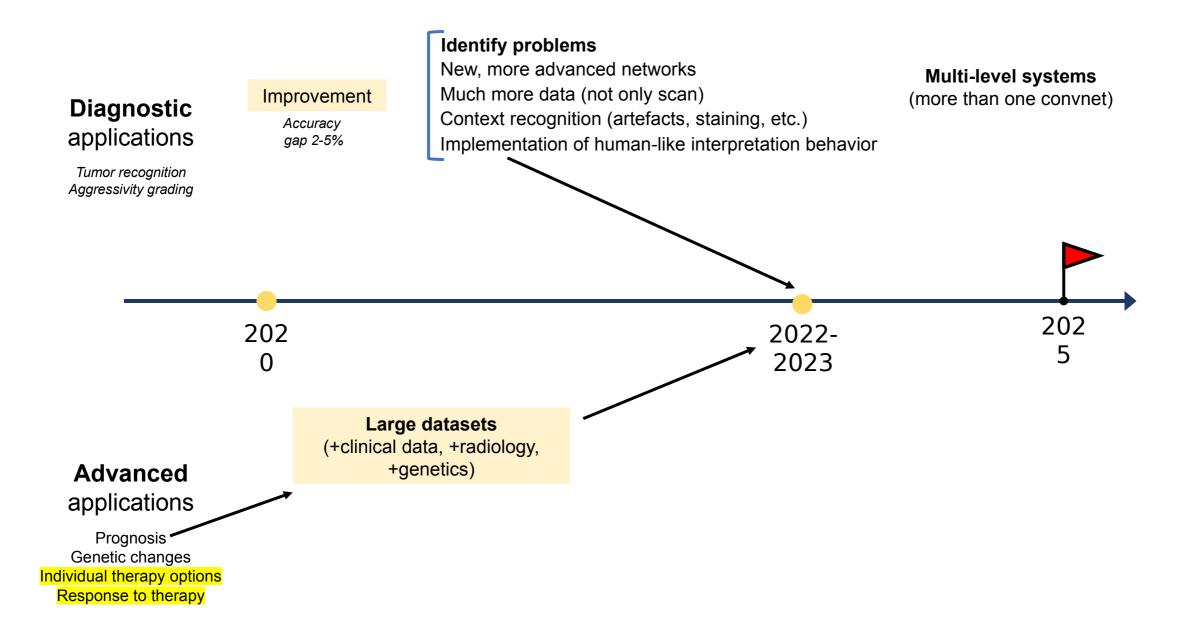
Advanced applications

Prognosis
Genetic changes
Individual therapy options
Response to therapy



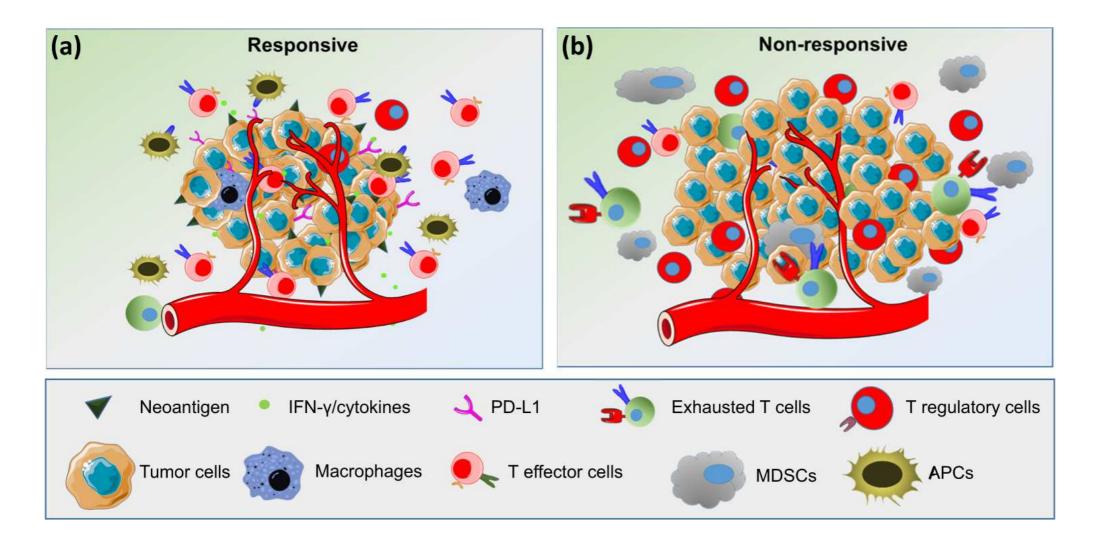






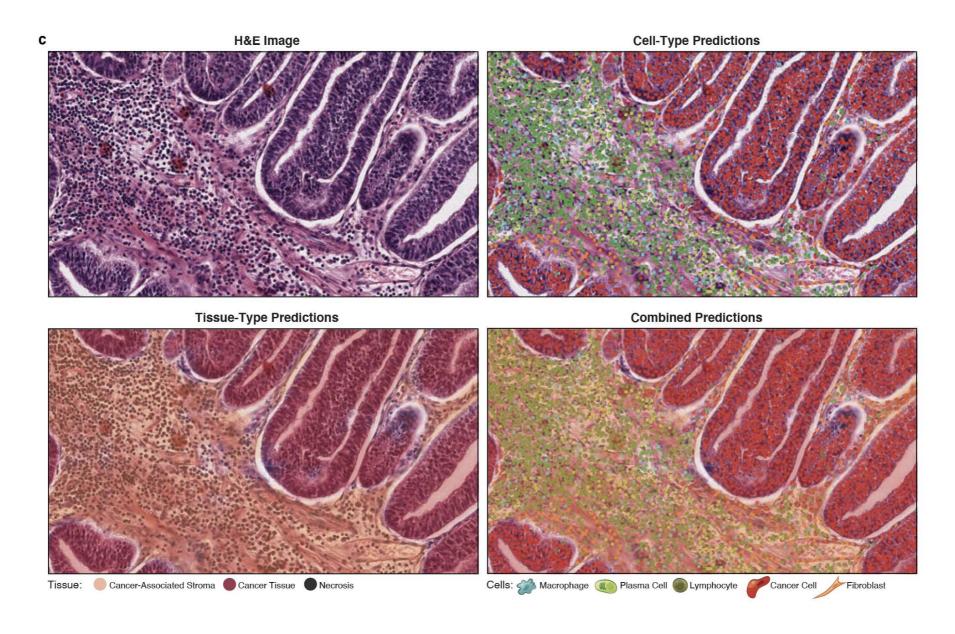


Immunooncology

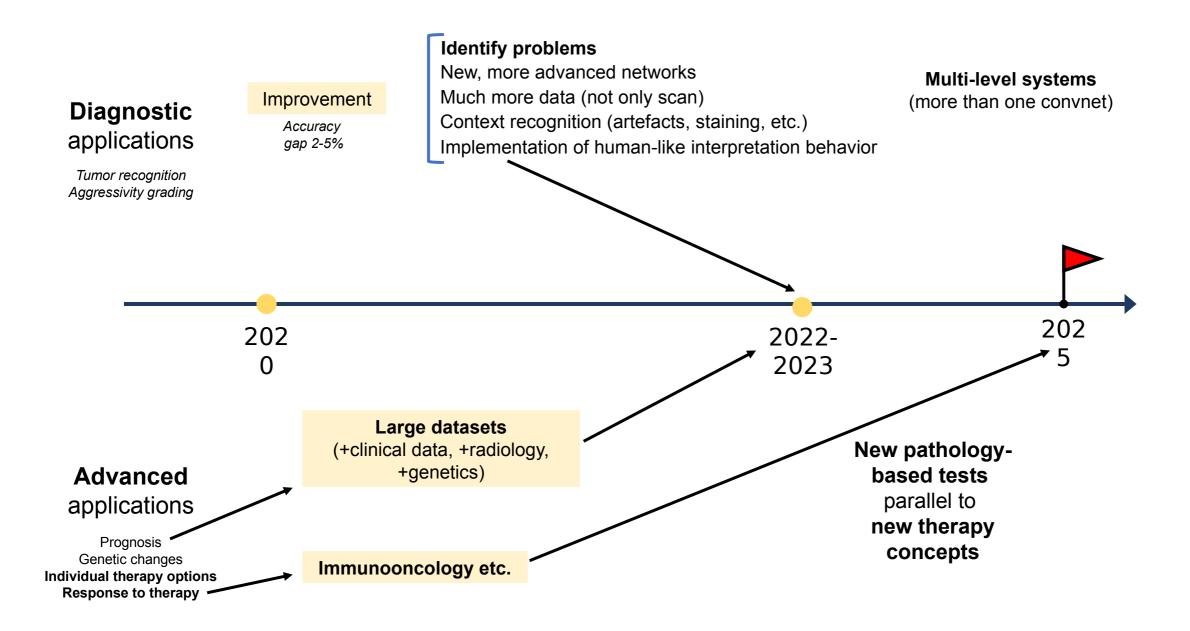


Immunooncology











Take home messages

- Digital transformation of pathology has just begun
- Impressive results from studies to diagnostical applications
 - Commercialized field to 2025
- New therapeutic agents (immunooncology) developed parallel to companion digital pathology tests
- High-quality data is everything
 - Pathological archives
- Big opportunities for research
- Cooperation of medical and technical specialists needed