Al Transforming Healthcare: The Future of Al in Pathology



By Saeed Pirmoradi (Al PhD) Winter 2025

Presenter Profile

- Founder of Khayyam Al Innovation Lab
- b **KHAYYAM** AI Lab
- **10** years of experience in AI developing
- More than 20 projects in AI application in medicine
- Data Scientist in Tabriz Valiasr Hospital (Clinical Research Development Unit)



Dr Saeed Primoradi (Al PhD) Al Developer The **First** (TVAI) Skyroom International Virtual Congress on the practical Application of Artificial Intelligence in **Medical Sciences**

Date & Time: 1-5 February 2025 (09:00 Am . 12:00)



تاریخورمان بیگزاری ۲۰ تا ۲۷ بیمین ۲۰۰۳ (۱۰۰۰ میچ) **اولین** کنگره بین المللی مجازی <mark>کاربرد هوش مصنوعی</mark> در علوم پزشکی











Previous AI in Medicine Applications workshops

Americans errors a ye	experience ear.	e 12	million	diagnostic
(Reference	<u>e link from C</u>	<u>BS ne</u>	ws)	
	agnostic mist death or pe			
(Reference	link from NO	CPA)		
Breast cano	er misdiagn	osis c	osts <mark>\$4</mark> E	<mark>8</mark> /year.
Dicast can				

Facts

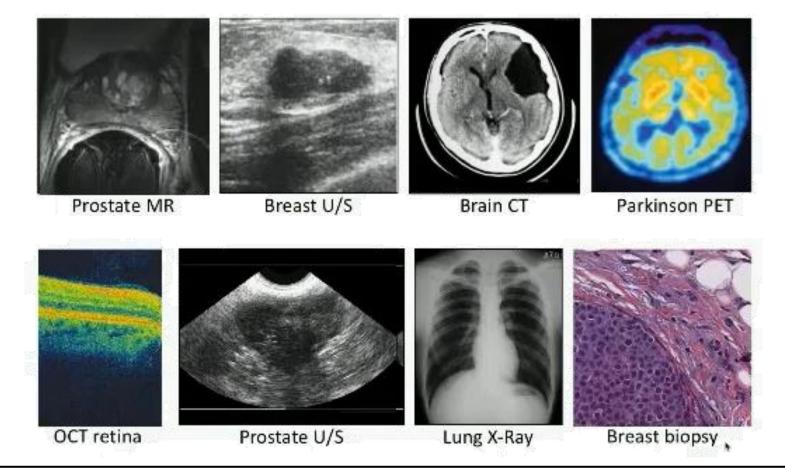
General Types of error:

NGUIGA

- Scanning error (we fail to fixate in specific areas)
- Recognition error (we fail to detect abnormality)
 Decision-making error
 - (almost 50% of error by incorrect interpretation of a malignant/benign)

Medical imaging

Many modalities/case: 700 Billion images/year in the US alone



Proportion answering in the affirmative (%)

Anatomic Laboratory pathologists medical directors

Errors and misdiagnosis

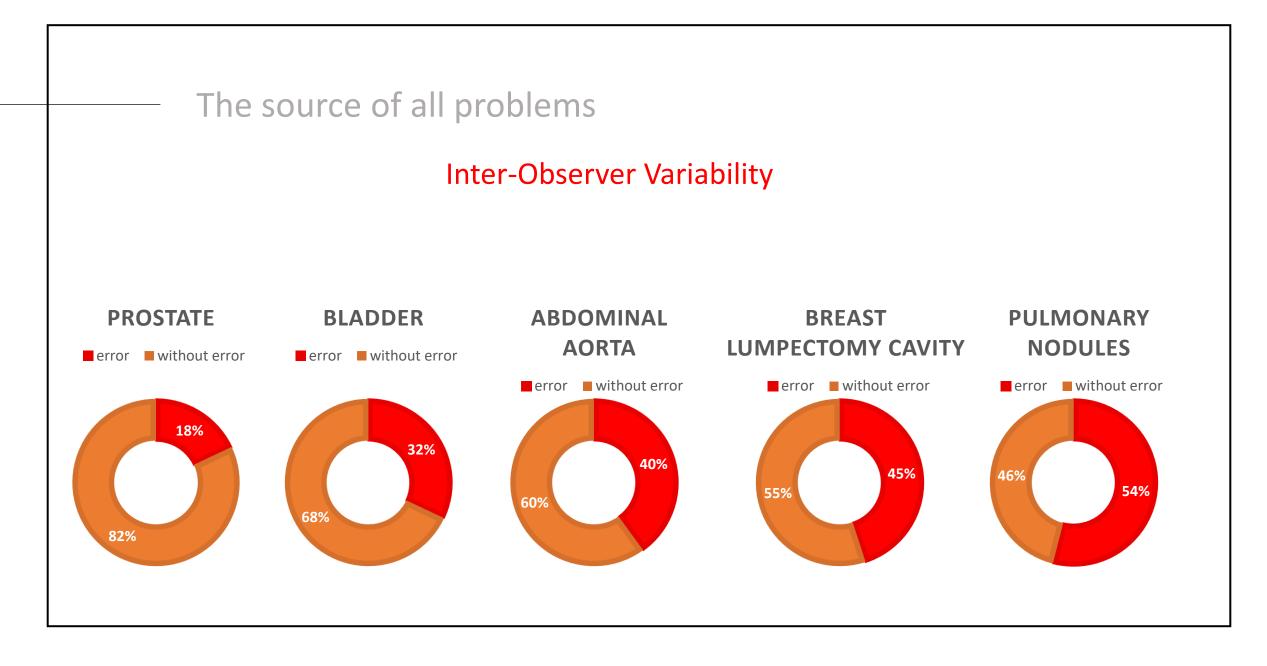
We need to talk: Pathologists, Patients, and Diagnostic errors.

Survey of 260 pathologists and 81 laboratory medical directors with response rate 51%

The pathologist 06/23/2016

(Reference link)

	pathologists	medical directo
Near misses should be disclosed to patients	22	18
Minor errors should be disclosed to patients	75	57
Serious errors should be disclosed to patients	96	99
Have you been personally involved with a near miss?	79	n
Have you been personally involved with a minor error?	71	63
Have you been personally involved with a serious error?	47	34
Have you disclosed a minor error to a patient?	5	7
Have you disclosed a serious error to a patient?	17	14
Where you disclosed a minor error to a patient, were you satisfied with the results? (Did you experience relief after disclosure?)	92	(70)
Where you disclosed a serious error to a patient, were you satisfied with the results? (Did you experience relief after disclosure?)	88	(80)

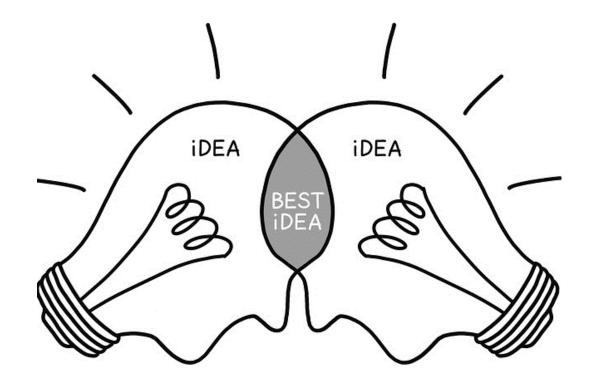


Inter-Observer Variability is more scary



Important question

What can Artificial Intelligence do?



A.I., especially Machine Learning, allows for a computer model to learn and extract meaningful patterns from data in a semi-automatic manner.

Introduction to Digital Pathology

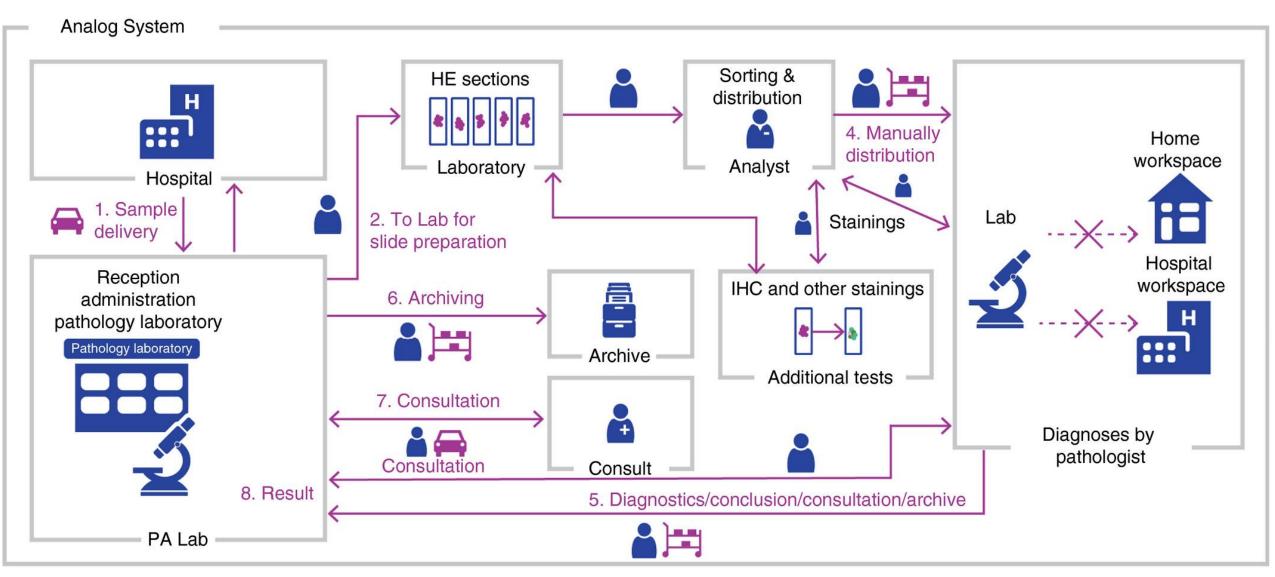
Pathology Process





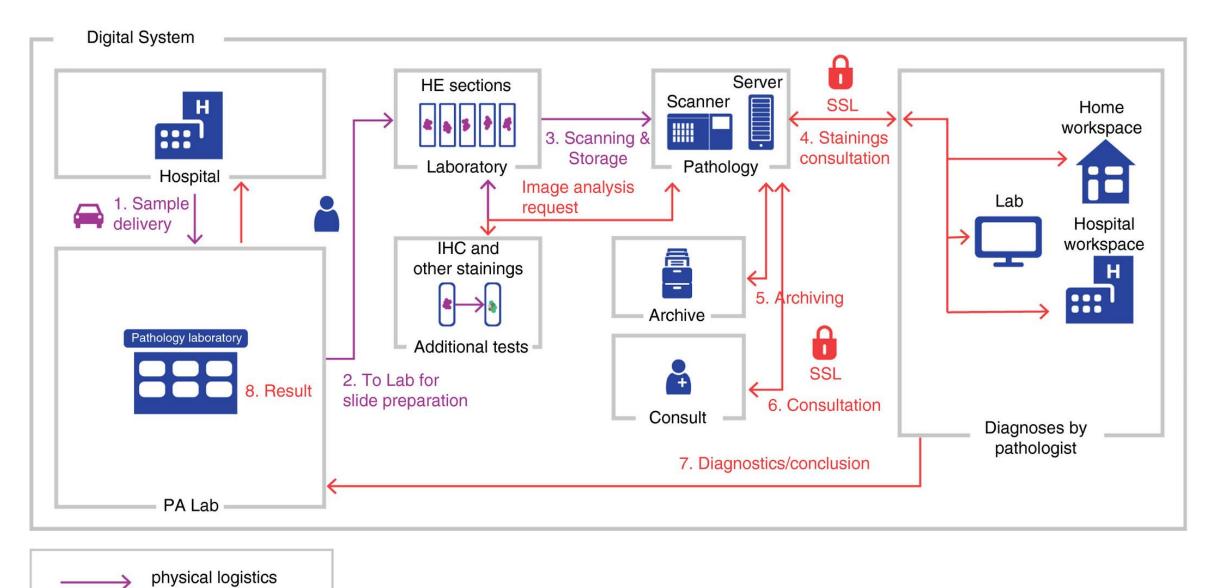


Pathology: Conventional Microscopy



Pathology: Virtual Microscopy

digital transmission



the Pathologist Going 100 % Digital YEARS

Making the move to 100 Percent Digital: The LabPon is the first laboratory in the world to digitise its histopathology service fully



DP has four key benefits:

- Efficient workflows
- Connected teams
- Increased safety
- New insights from analysing large datasets



The potential operational cost savings for 5 years following the implementation of a digital pathology solution were estimated at around \$18 million.

The main contributing factors were gains in pathologist time by higher productivity and better workload distribution (\$12.4 million), and reduced costs of incorrect treatment.

The over- and under-treatment costs in oncology were estimated at \$5.4 million.

Digital Pathology: Scanners

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HURON Digital Pathology



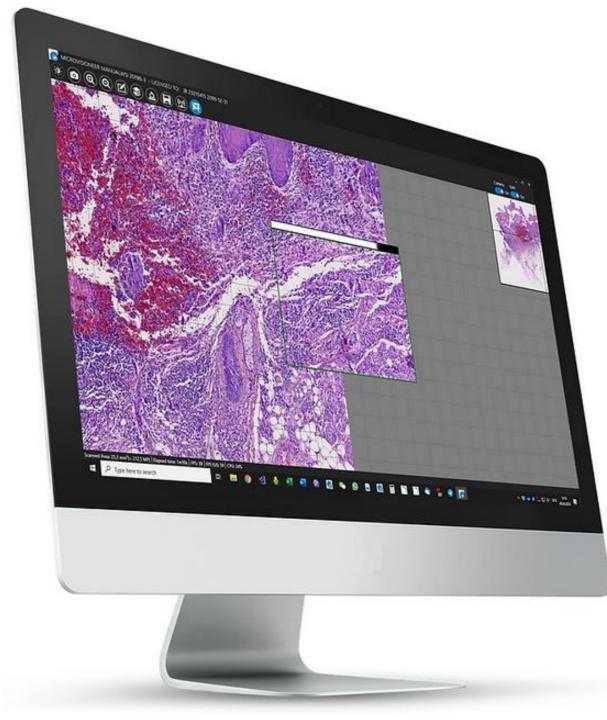


Grundium



Whole Slide Images (WSI): Setup

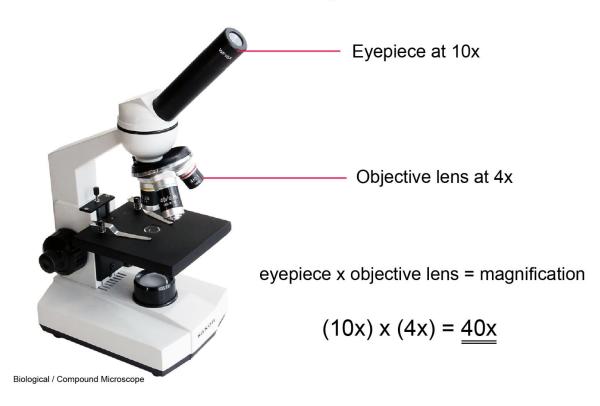




Whole Slide Images (WSI): Image View

Magnification

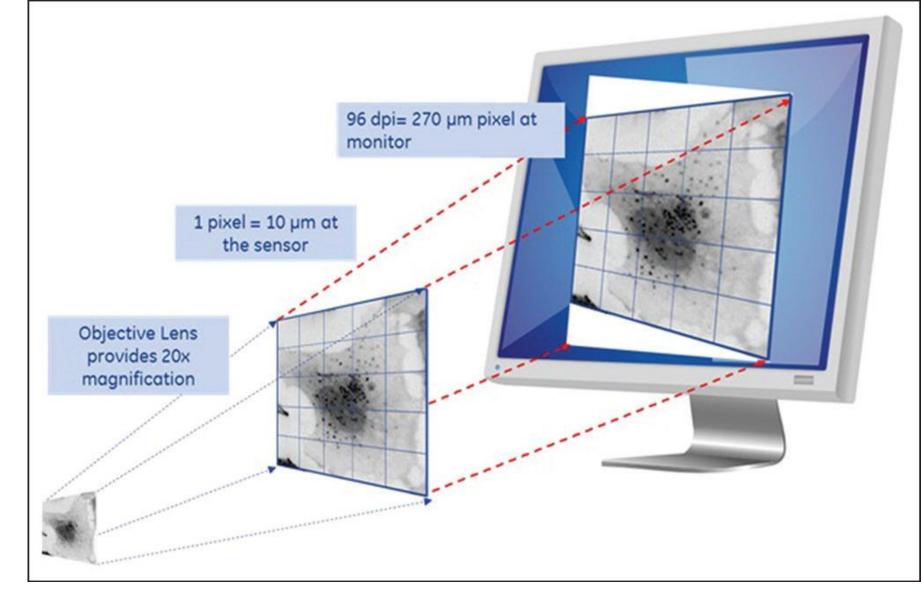
To calculate magnification



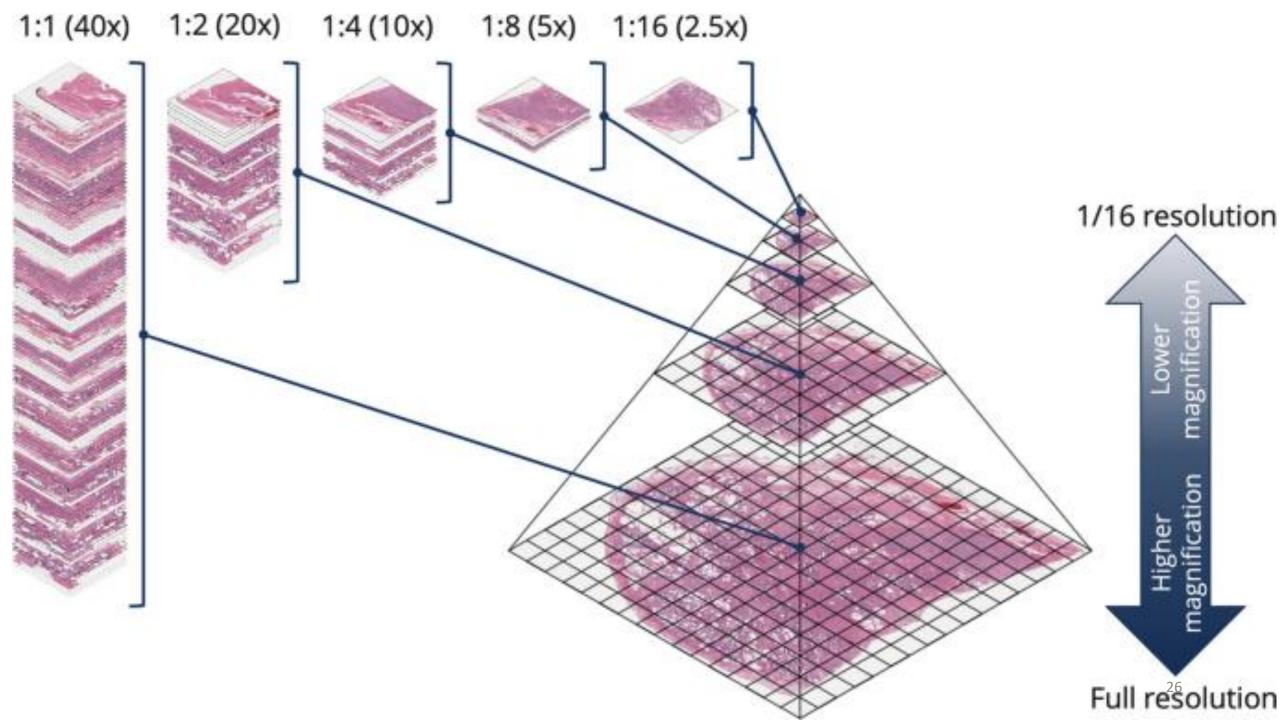
Magnification \neq Resolution

Optical Resolution = f(Objective Lenses)

The added boost at magnification comes from the monitor's workstation.



Digital Resolution = f(Objective Lens, Digital Camera Sensor, Monitor)



Concordance between Digital Pathology and Light Microscopy

2014

Concordance between digital pathology and light microscopy in general surgical pathology: a pilot study of 100 cases

Joseph P Houghton,¹ Aaron J Ervine,² Sarah L Kenny,² Paul J Kelly,² Seamus S Napier,² W Glenn McCluggage,² Maureen Y Walsh,² Peter W Hamilton³

There was concordance between the original light microscopy diagnosis and digital pathology-based diagnosis in 95 of the 100 cases while the remaining 5 cases showed only slight discordance (with no clinical consequence). None of the cases were categorised as discordant. Participants had mixed experiences using digital pathology technology.

Journal of Clinical Pathology 2017



of Pathology & Laboratory Medicine

The Diagnostic Concordance of Whole Slide Imaging and Light Microscopy

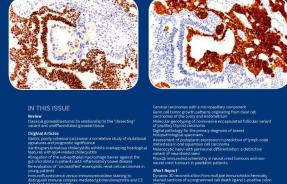
A Systematic Review

Edward Goacher, BSc; Rebecca Randell, PhD; Bethany Williams, MBBS; Darren Treanor, MB, BSc, PhD, FRCPath

Thirty-eight studies were included in the review. The mean diagnostic concordance of WSI and LM, weighted by the number of cases per study, was 92.4%. The weighted mean κ coefficient between WSI and LM was 0.75, signifying substantial agreement. Of the 30 studies quoting percentage concordance, 18 (60%) showed a concordance of 90% or greater, of which 10 (33%) showed a concordance of 95% or greater. This review found evidence to support a high level of diagnostic concordance. However, there were few studies, many were small, and they varied in quality, suggesting that further validation studies are still needed.

2018







Digital pathology for the primary diagnosis of breast histopathological specimens: an innovative validation and concordance study on digital pathology validation and training

Bethany Jill Williams 🔀 Andrew Hanby, Rebecca Millican-Slater, Anju Nijhawan, Eldo Verghese, Darren Treanor

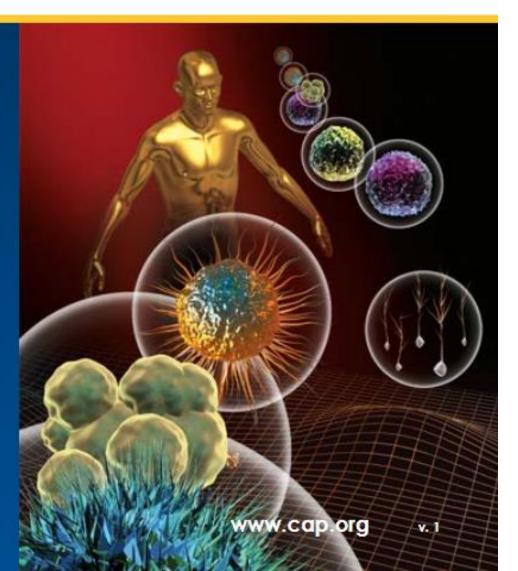
Three specialty breast pathologists completed training in using a digital microscopy system. They were exposed to a training set of 20 challenging cases, designed to help them identify personal digital diagnostic pitfalls. Following this, the three pathologists viewed a total of 694 live, entire breast cases. All primary diagnoses were made on digital slides, with immediate glass slide review and reconciliation before the final case sign-out. There was complete clinical concordance between the glass and digital impression of the case in 98.8% of cases. 30



Pathology and Laboratory Quality Center

Recommendations for Validating Whole Slide Imaging Systems for Diagnostic Purposes in Pathology

Anil V Parwani, MD, PhD On behalf of the CAP WSI Validation Expert Panel



The validation study should encompass the entire WSI system



-The validation process should include at least 60 cases for one application... reflecting the spectrum and complexity of specimen types and diagnoses likely to be encountered during routine practice.

-The validation study should establish a diagnostic concordance between digital and glass slides for the same observer (i.e., intraobserver variability).

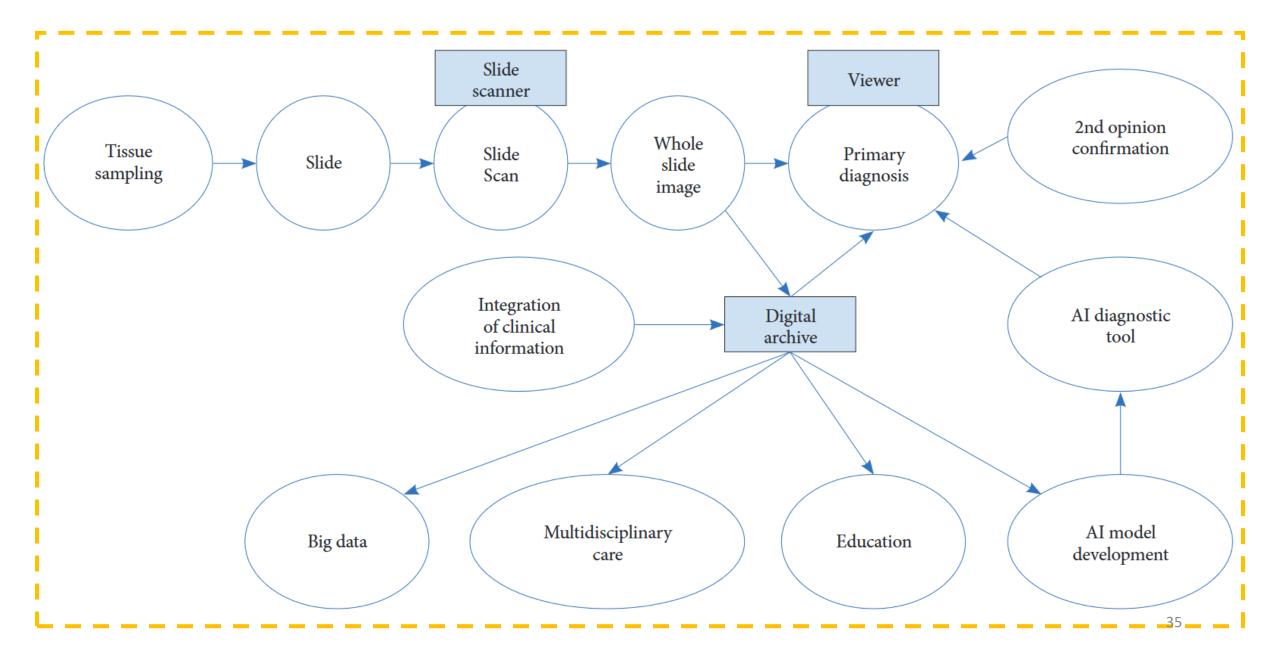
-The washout period of at least 2 weeks should occur between viewing digital and glass slides.

Computer Vision and AI in Digital Pathology

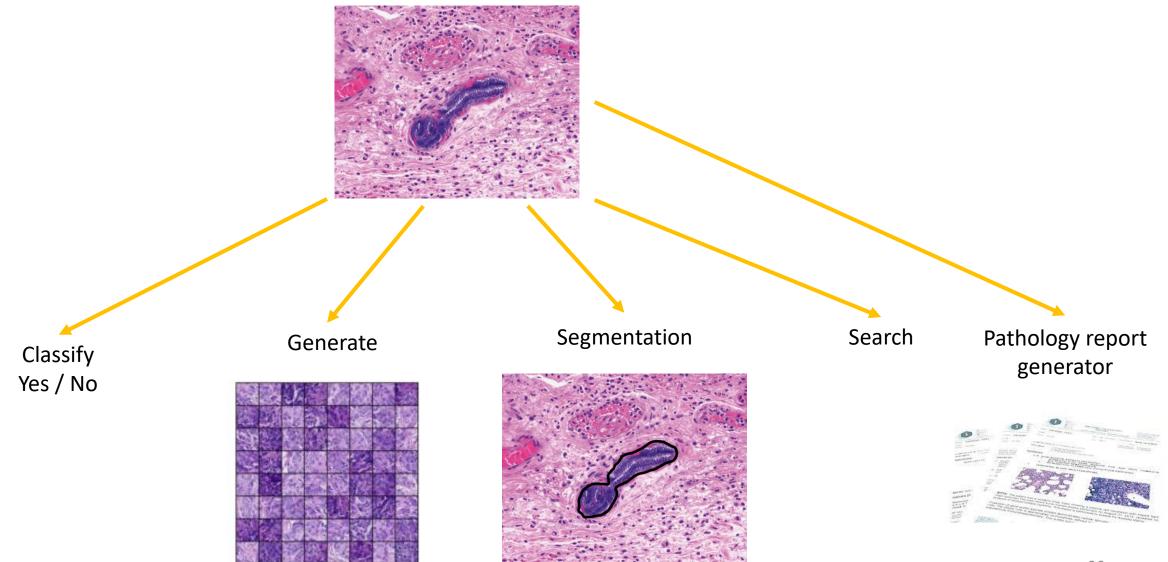
Clinical Pathology Challenges:

- Inter-observer variance
- Intra-observer variance
- Lack of quantitative, objective, and reproducible measures to assess patient biopsies
- The human brain can keep track of only a few variables

Workflow of digital pathology



What can AI do with images?



NIH NATIONAL CANCER INSTITUTE Center for Cancer Genomics



Genomic Data Commons Data Portal

Harmonized Cancer Datasets

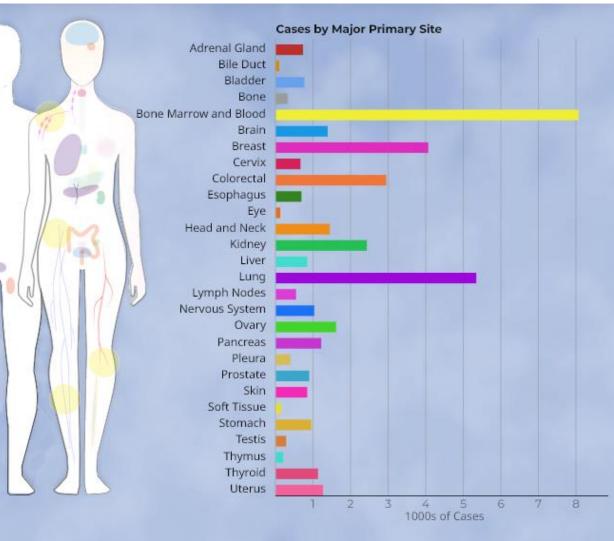
A repository and computational platform for cancer researchers who need to understand cancer, its clinical progression, and response to therapy.

Explore Our Cancer Datasets

Data Portal Summary

🖸 Data Release 41.0 - August 28, 2024







The data in this challenge contains whole-slide images (WSI) of hematoxylin and eosin (H&E) stained lymph node sections.

Depending on the particular data set (see below), ground truth is provided:

- •On a lesion-level: with detailed annotations of metastases in WSI.
- •On a **patient-level:** with a pN-stage label per patient.

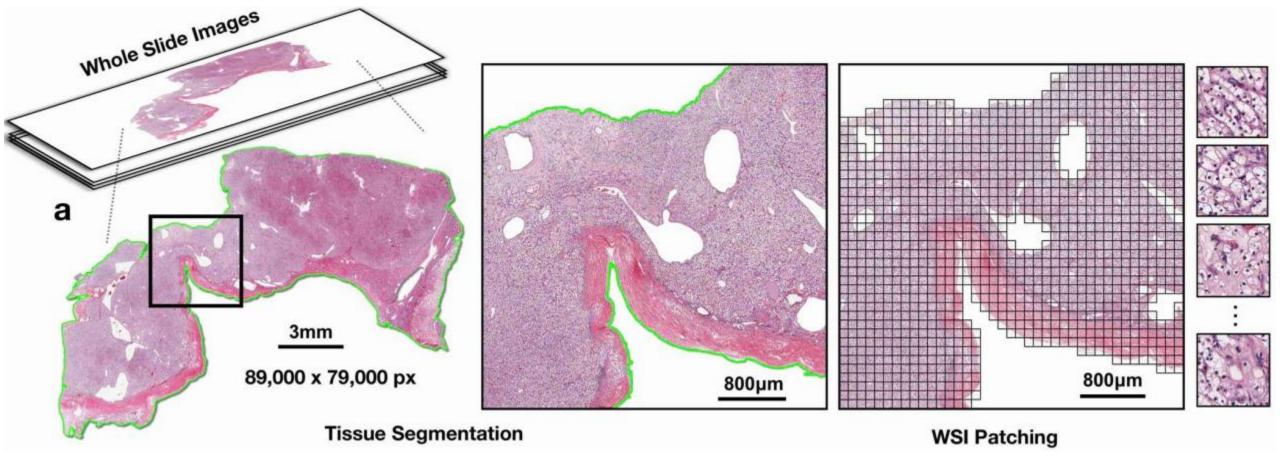


Datasets

The TCGA has many cancers, with over 1.2 petabytes of data, including pathology slides.

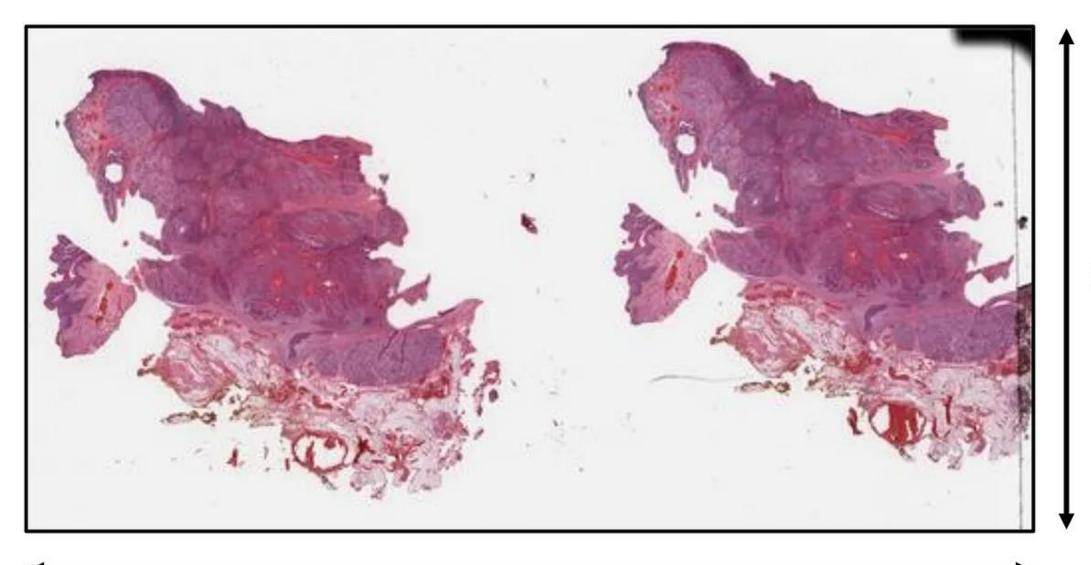
The Camelyon has 1399 WSIs with tumor marking in metastasis, without subtyping, not configuring a diagnosis.

However, most public datasets are not in WSIs or do not have pathologists' pixel-level annotations.



Images that can have more than 10 billion pixels, cannot be used fully as an input of a neural network without overflowing memory.

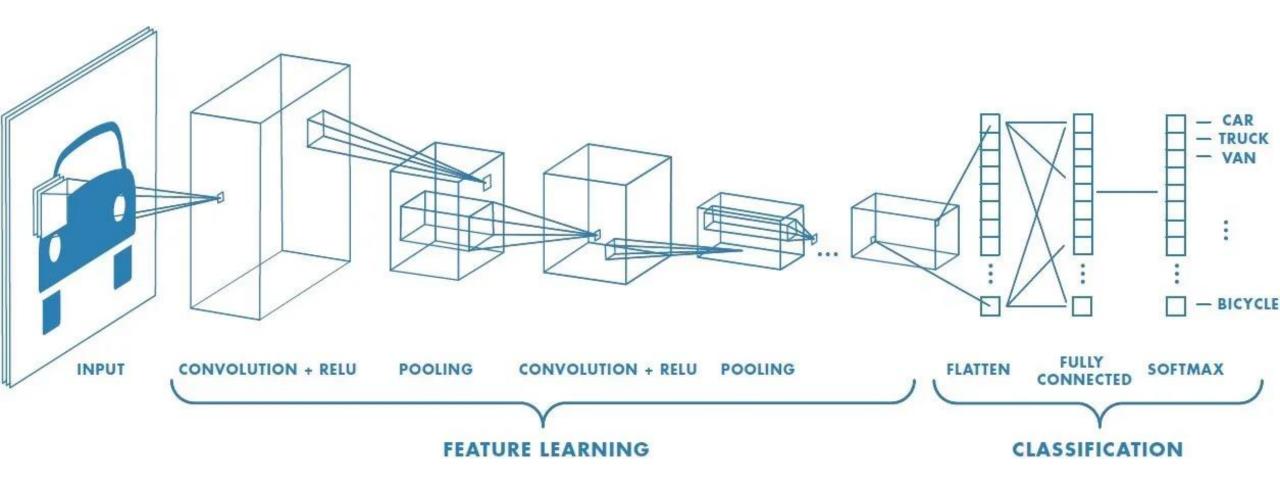
Therefore, all studies divided the WSIs into smaller patches (such as 256x256 pixels), using overlap or not as data augmentation, to feed their models.



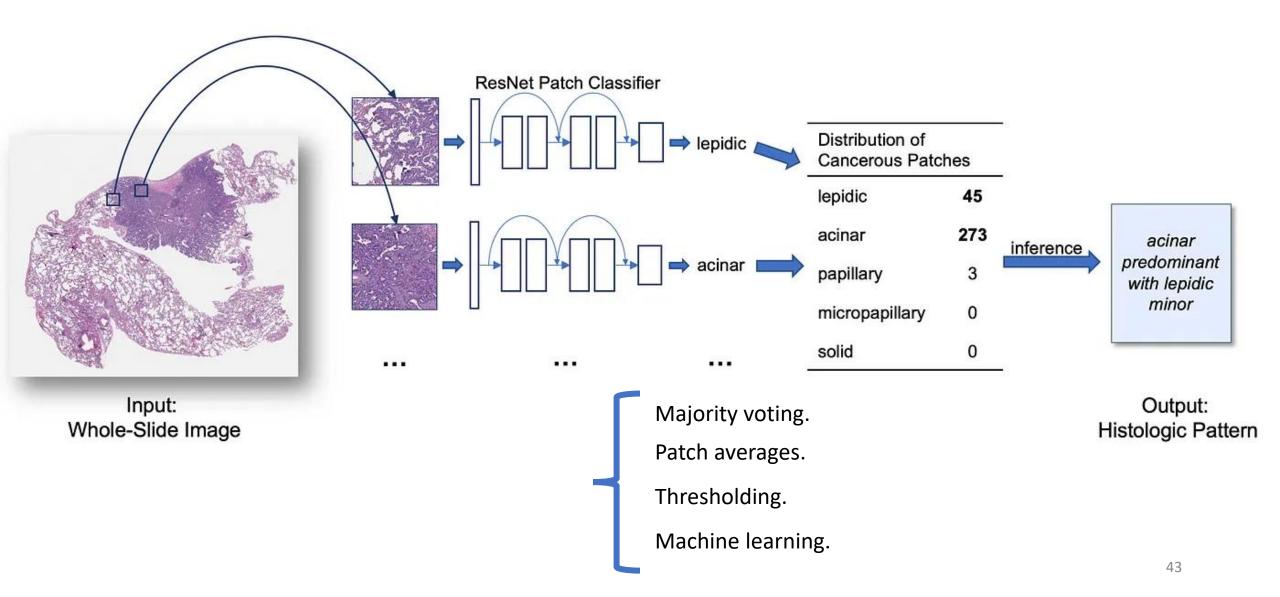
77781 pixels

158002 pixels

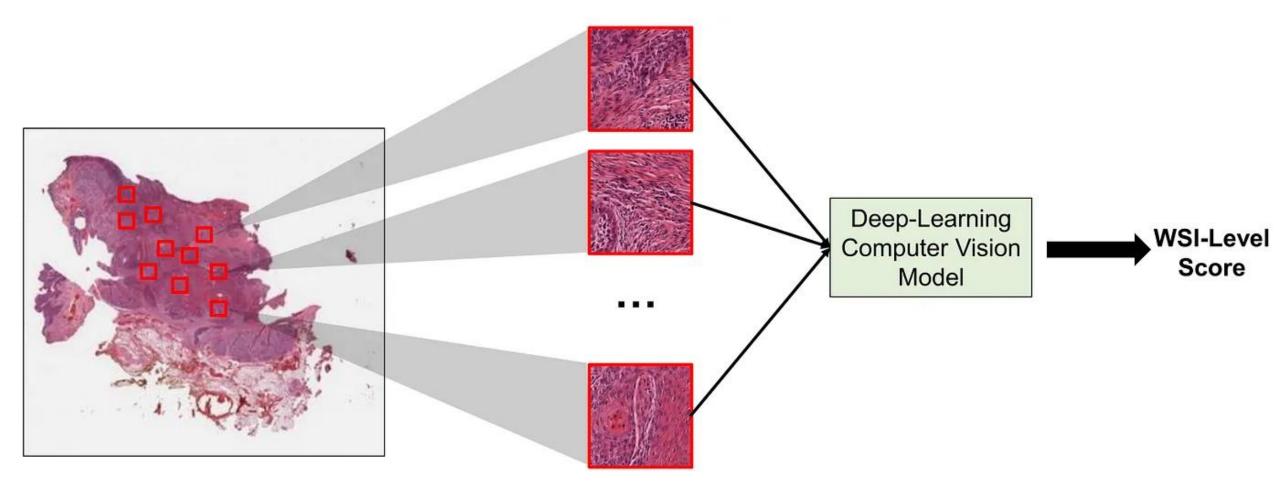
Deep Learning & Convolutional Neural Nets



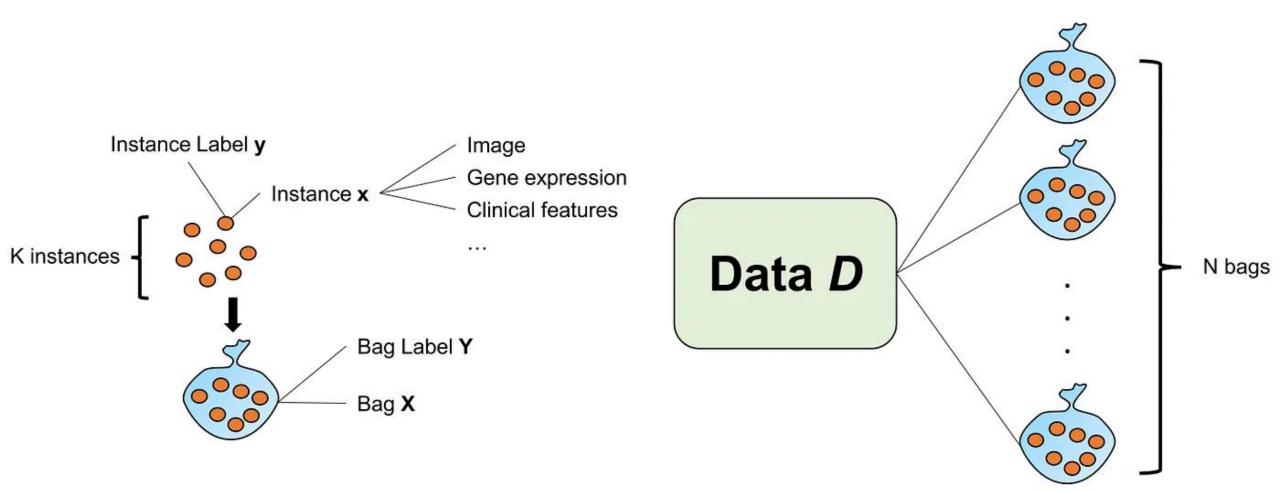
Whole-Slide Inference

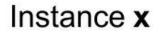


The idea of Multiple Instance Learning (MIL) in WSI analyses



Problem Statement of general MIL (MIL-G)









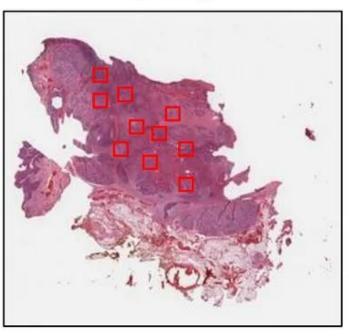


Refined Problem Statement of MIL-WSI

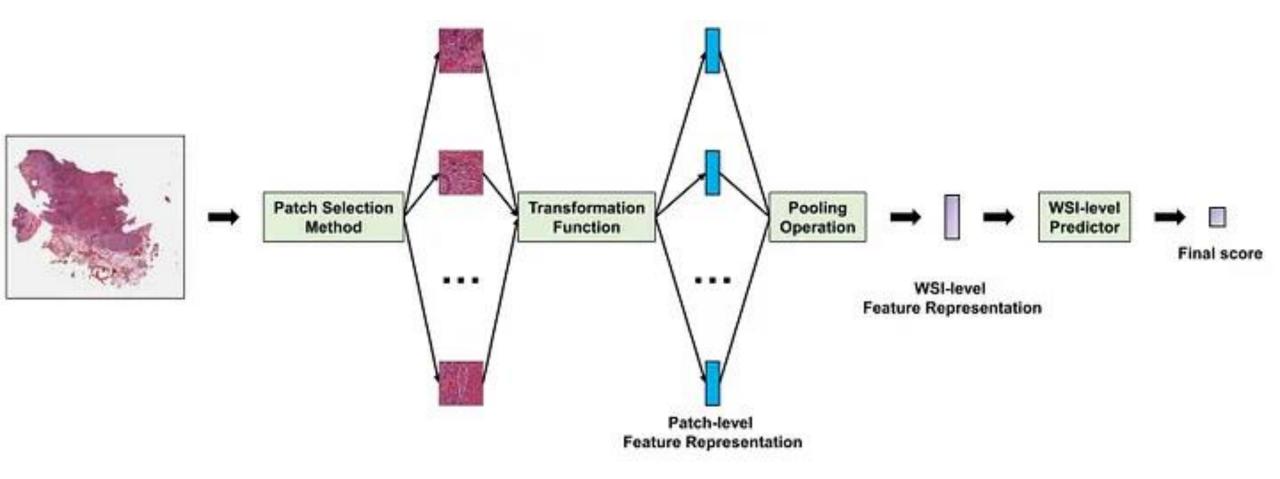
WSI X







Main Components of MIL-WSI Framework



Categories of Patch Selection Methods

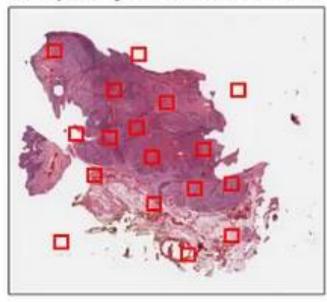
(i) Completely random selection

(ii) Random selection with the region of interest (ROI).

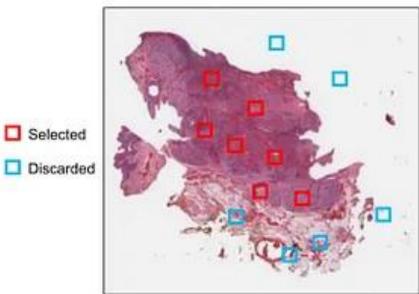
(iii) Feature-based selection.

(iv) Hierarchical selection.

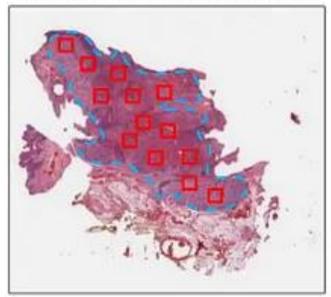
Completely Random Selection



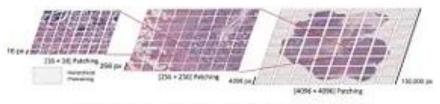
Feature-based Selection



Random Selection with ROI



Hierarchical Selection



Source: https://github.com/mahmoodlab/HIPT

Categories of Pooling Operations

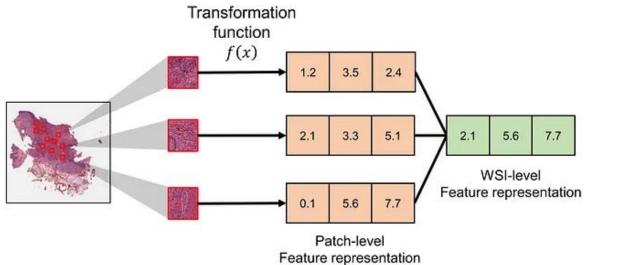
1- Static Pooling

2- Adaptive Pooling

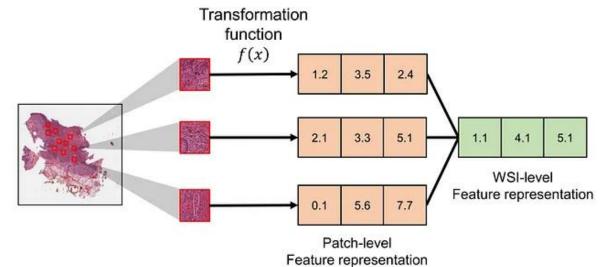
These methods adaptively update the patch contribution to its WSI. This category contains trainable pooling, dynamic pooling, differential evolutionary pooling, and so on. Generally, the adaptive pooling operation is changed iteration by iteration as we train the MIL-WSI framework. Interesting examples of this category are Attention-based pooling and Hopfield pooling, both of which belong to trainable pooling.

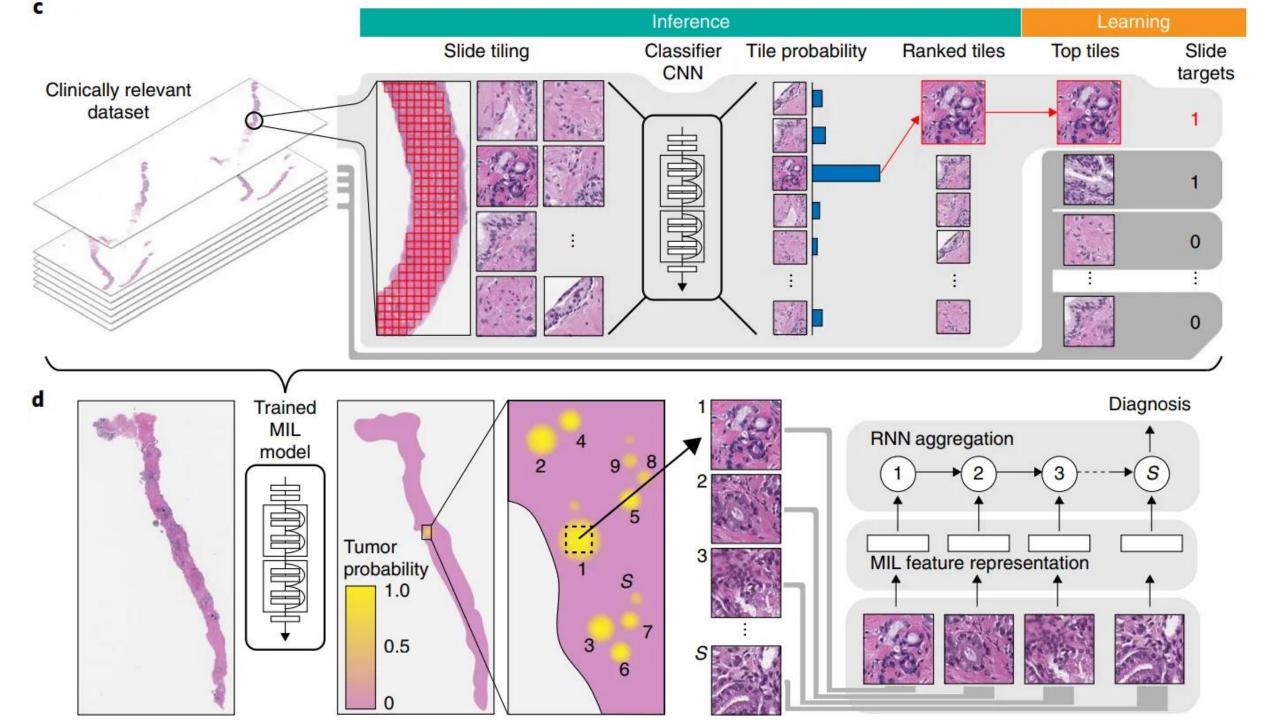
Static Pooling

Max Pooling



Average Pooling







Education

Digital Pathology



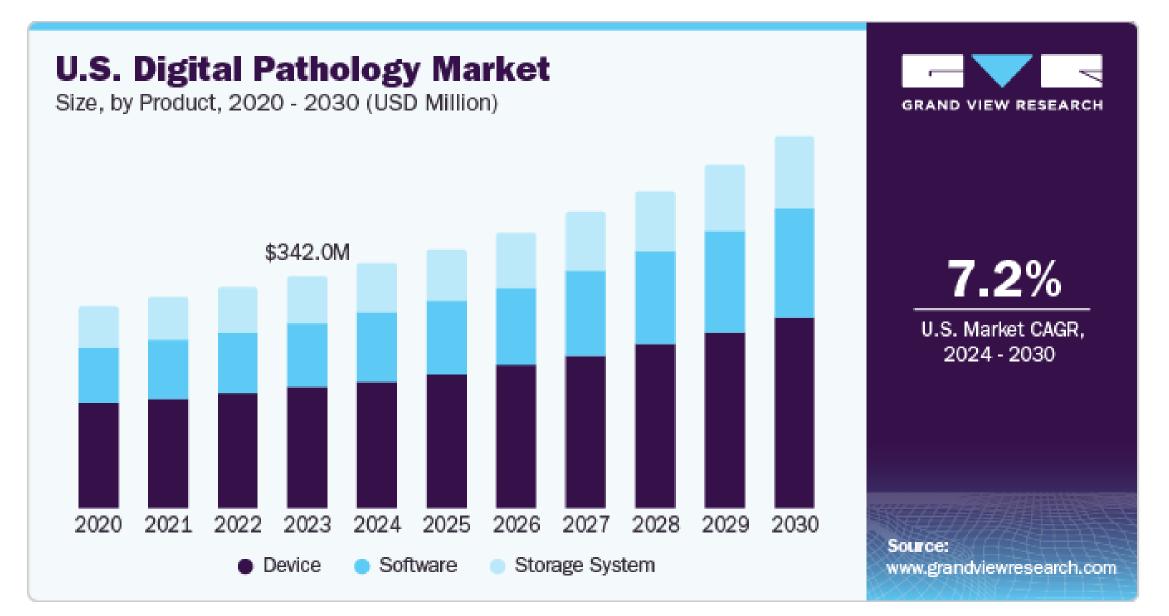
Traditional Pathology

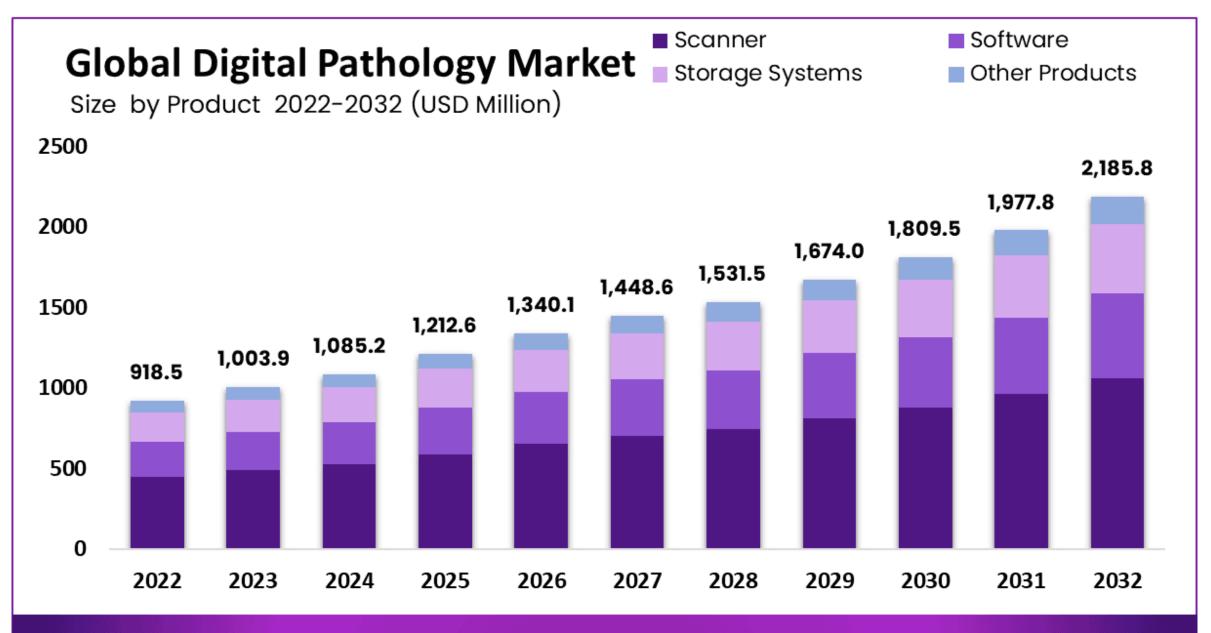
Global DIGITAL PATHOLOGY Market

Opportunities and Forecasts, 2021-2030 Global Digital Pathology Market is expected to reach **\$1791.3 Million** by 2030

Allied

Growing at a CAGR of 9.3% (2021-2030)

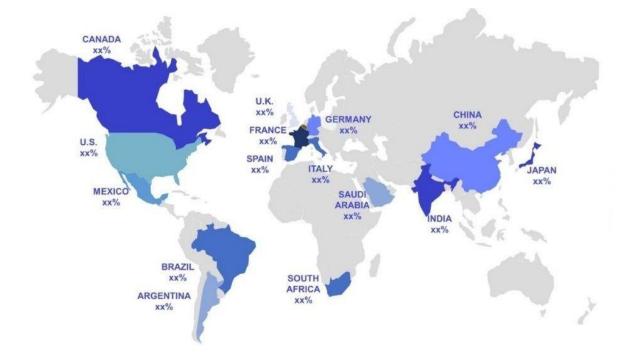


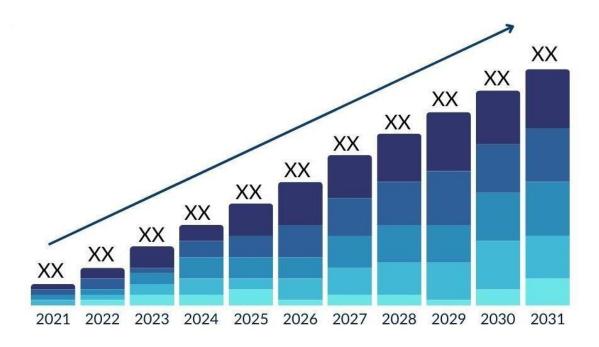


The Market will Grow 9.3% The forecasted market \$2,185.8M I market.us At the CAGR of 9.3% size for 2032 in USD

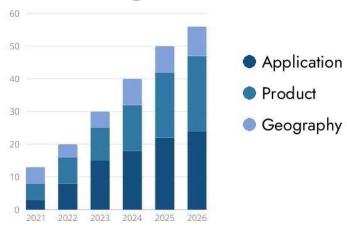
Global Digital Pathology Market Size and Scope



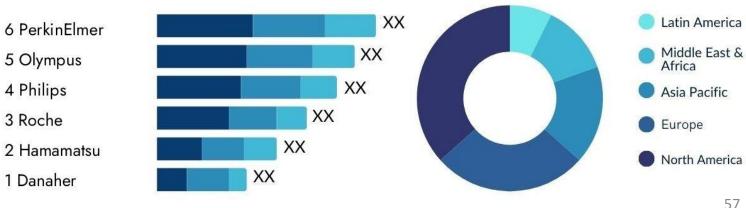




Market Segmentation



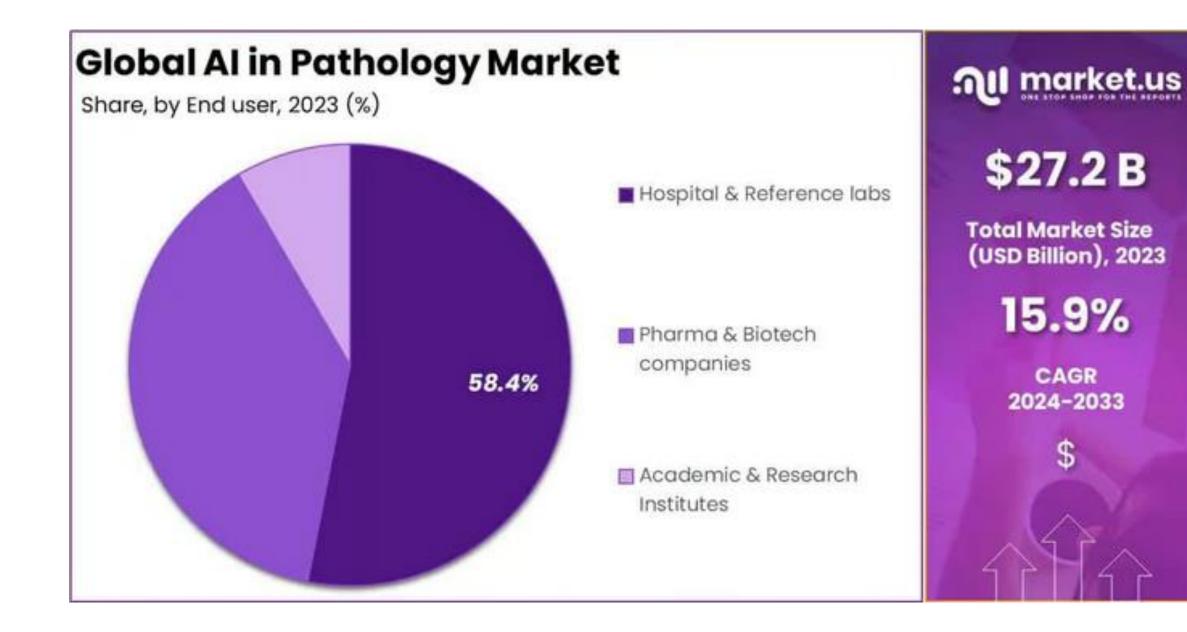
Top Key Players



57 Source : www.marketresearchintellect.com

Regional Analysis



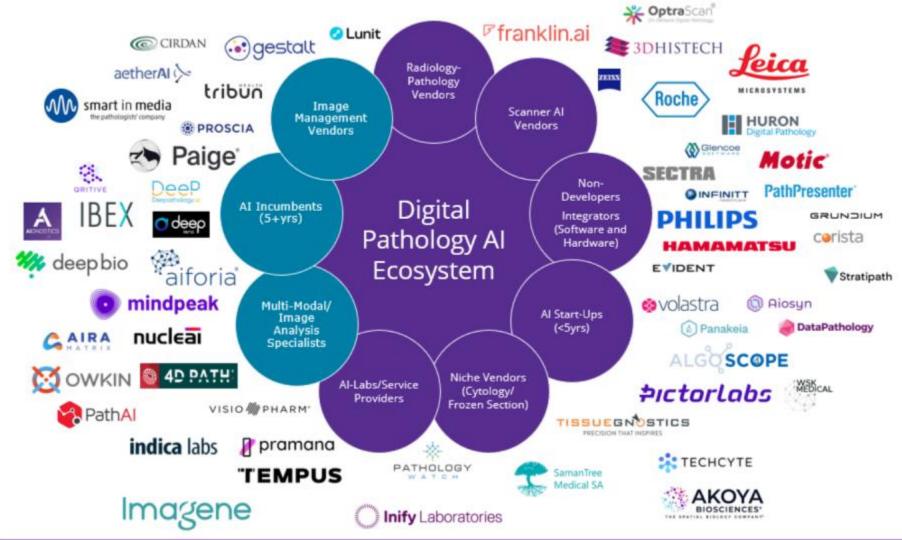


AI Startups and Companies

Companies



Signify RESEARCH Digital Pathology Al Ecosystem









Thank you

Phone Number: 09143116396

Email: said.Pirmoradi@gmail.com

Website: Khayyamlab.ir