

Creating a Cancer-free World. One Person, One Discovery at a Time.

The James



THE OHIO STATE UNIVERSITY WEXNER MEDICAL CENTER

Implementing Digital Pathology For Clinical, **Educational and Research Applications – The Nuts** and Bolts Anil V Parwani, MD, PhD, MBA

The Ohio State University Comprehensive Cancer Center - Arthur G. James Cancer Hospital and Richard I. Solove Research Institute

The author has no relevant conflicts to disclose

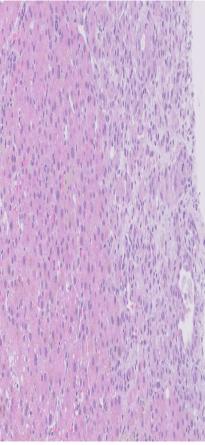






Objectives of my talks:

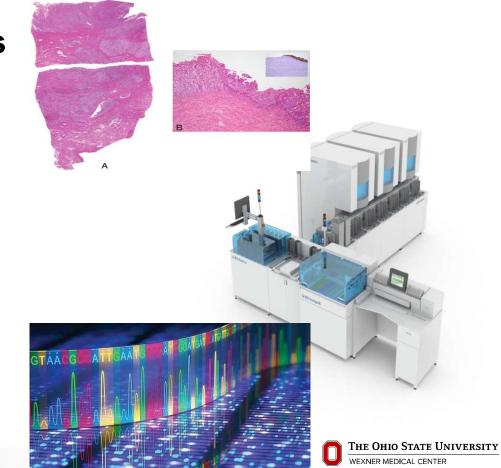
- Provide an overview of the evolution of digital pathology and AI and its current state for clinical diagnostics, research and education
- Provide an understanding of the challenges and opportunities for implementing digital pathology and AI in the clinical areas using OSU as an example
- Explore the future directions of Digital Pathology /Artificial Intelligence technology's role in advanced diagnostics

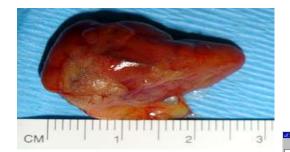




NEW TOOLS FOR HIGH VALUE CLINICAL DIAGNOSTICS

- Rapid innovations.
- More variety and complexities of available tests and services.
- New molecular tests, advanced equipment and testing techniques.
- More automation
- Digital and Computational pathology
- Artificial Intelligence





PATHOLOGY REPORTS ARE BECOMING MORE COMPLEX

montic Di ignosis Worksheet for P<u>HS05-1</u> R Worksheet#1 of 1 Part 2: Rib Cartilage Page 1 of 4 Non-Hodgkins Lymphoma Biopsy/Resection Synoptic Template <u>B-cell Lymphoma</u> B-cell lymphoma, subtype cannot be determined B-cell lymphoma with high grade features Precursor B-lymphoblastic leukemiatlymphoma C3 - MACROSCOPIC SPECIMEN TYPE** C4 C5 C8 Ale Scalibraphoran with high grade features Processor 54 physiolates it Leavanianmal/physiolate Comparing the second secon AL Lymphadenectomy (specify sites): ______ A2 Other (specify): ______ A3 Not specified Splenectomy Softer extranodal (specify): <mark>Mediastinal soft lissue </code></mark> TUMOR SITE (check all that apply)** B1 Lymph node(s), site unknown B2 Lymph node(s) - Specify site(s), B3 Other tissue(s) - Specify site(s): Mediastinum B4 B4 Not specified B5 Only one site biopsied, see above - MICROSCOPIC ---MICROSCOPIC HISTOLOGIC TYPE (WHO CLASSIFICATION)** Note: This is NOT the final diagnosis. Use final diagnos /comment for therapeutic decisions. Histologic type cannot be assessed Non-Hodgkin lymphoma vs Hodgkin lymphoma OK Cancel Comment A Menu Help



Acute and chronic inflammation, most consistent with recent biopsy procedure. 3. Scalene lymph node, left, no pathologic diagnosis. Mike O'Seen,MD

Path. No.: \$91-210 Name: Sally Sacqueski

Surgeon: So Long, MD

GROSS DESCRIPTION

MICROSCOPIC

DIAGNOSIS:

Address:____

Reg. No

Age: 66 Sex: Female Race: White Location: ____ Date: 02/11/91

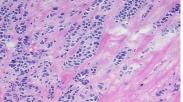
The specimen is received in two parts. They are labeled #1, "biopsy bladder tumor", and #2, "scalene node, left". Part #1 consists of multiple are submitted in their entirety for processing. Part #2 consists of multiple fragments of fatty vellow tissue which range in size from 0.2 to 1.0 cm in diameter. They are submitted in their entirety for processing.

Section of bladder contains areas of transitional cell carcinoma. Na area of invasion can be identified. A marked acute and chronic inflammatory reaction with eosinophils is noted together with some necrosis. Sections are examined at six levels. Section of lymph node contains normal node with reactive neurinal centers.

1. Papillary transitional cell carcinoma, grade II, bladder, biopsy

Occupation HISTORY OF CASE: Multiple TURB for Grade II TCC with microinvasior multiple tumors CLINICAL DIAGNOSIS: Carcinoma of bladder; R/O scalene POST-OPERATIVE DIAGNOSIS: Carcinoma of bladder

Operation: Bx of bladder & L scalene nod



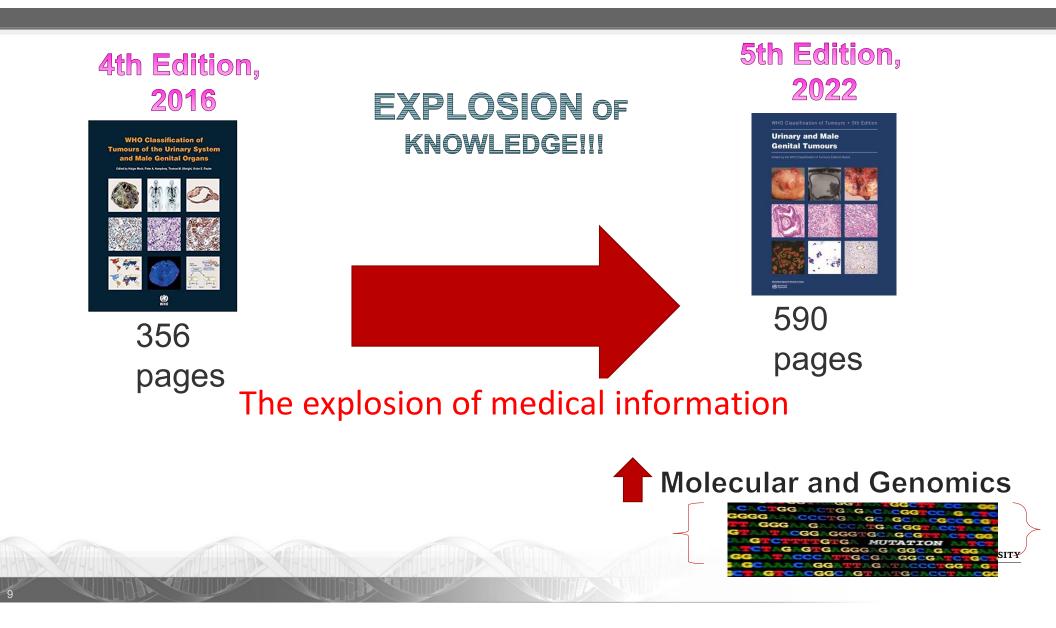


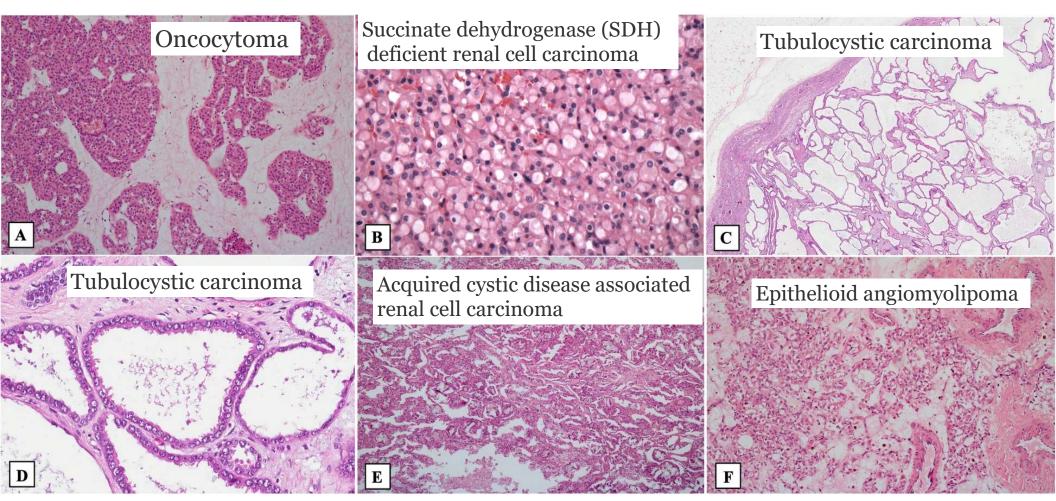


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Oncocytic renal tumors and some of their differential diagnoses

Athanazio, D.A., Amorim, L.S., da Cunha, I.W. *et al.* Classification of renal cell tumors – current concepts and use of ancillary tests: recommendations of the Brazilian Society of Pathology. *Surg Exp Pathol* **4**, 4 (2021).

VISUAL INFORMATION IS CENTRAL TO PATHOLOGY

James

STATE UNIVERSITY

CAL CENTER

H&E SLIDE = THE GOLD STANDARD

THE H&E SLIDE FOR THE PATHOLOGIST IS EQUIVALENT TO A GOOD H&P



Accurate interpretation of the hematoxylin and eosin (H&E) slide has remained the foundation of pathological analysis and diagnostic medicine for over a century

Powerful microscope = Whole Slide Imaging (WSI)

COMPUTATIONAL PATHOLOGY/AI

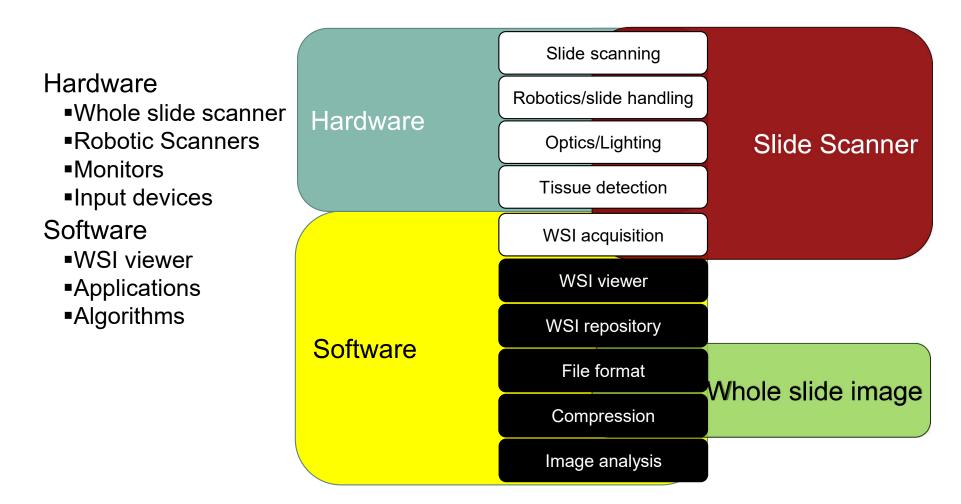
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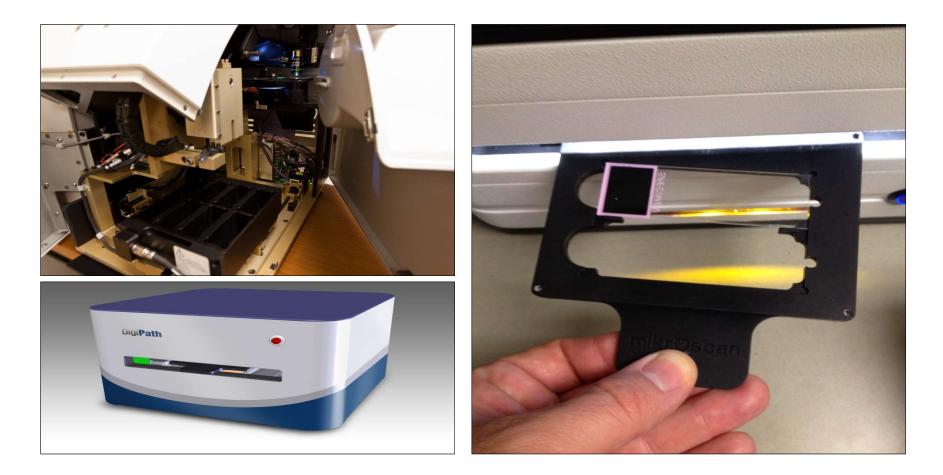


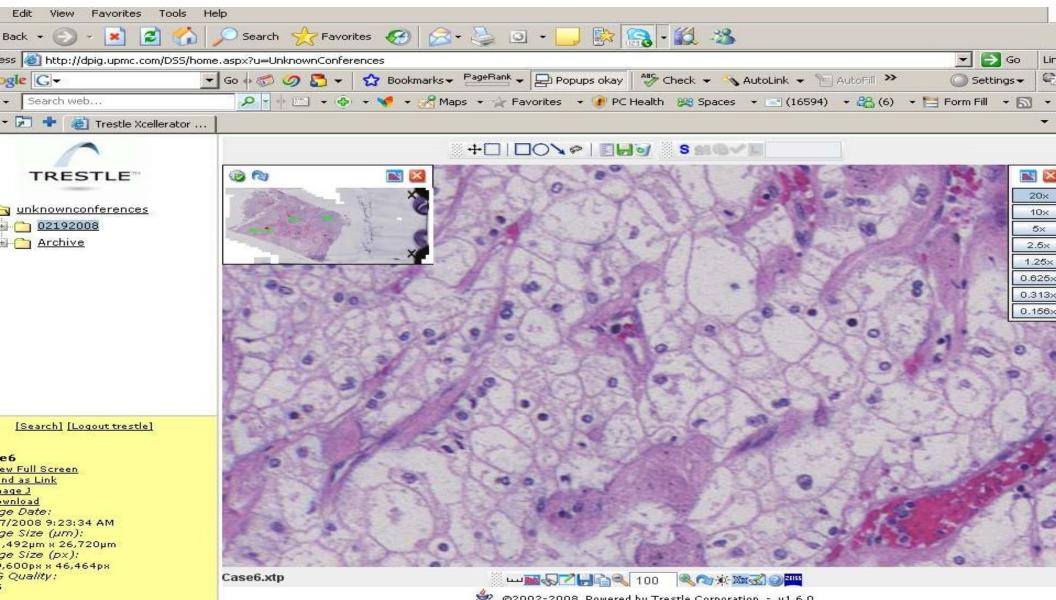


Digital Pathology Subsystem



WSI scanners





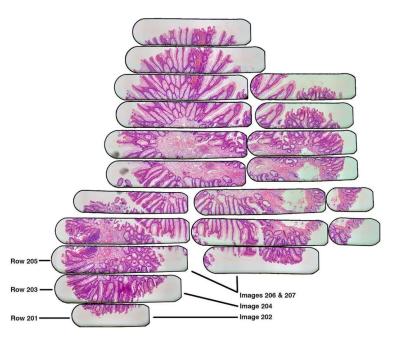
🖆 @2002-2008 Powered by Trestle Corporation - v1.6.0

STATE OF THE WSI INDUSTRY

•Automated, high-speed, high resolution whole slide imaging systems

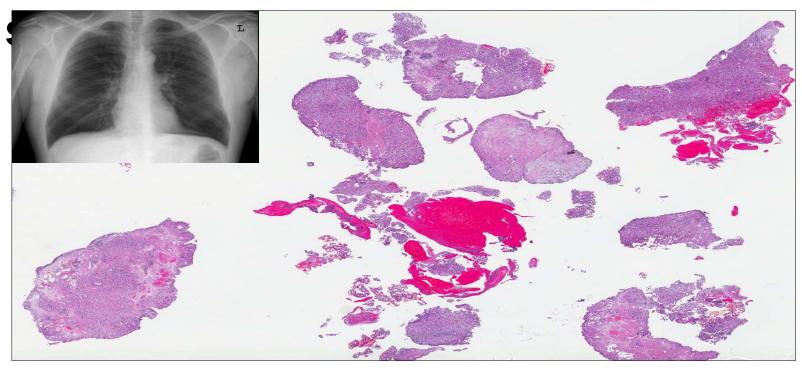
•Scan 1 to 1000 slides, even reading barcodes on slides.

•1.5 x1.5 cm tissue section in approximately 1-4 minutes with spatial sampling periods of between 0.25 to 0.5 microns/pixel.



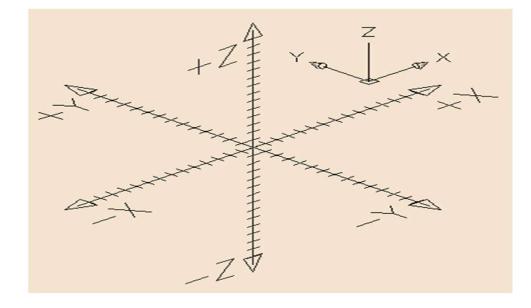
Up to 100 slides per hour, 2000 slides per day (at 40x resolution, 0.25 μ m/pixel, single layer).

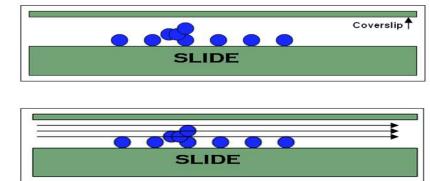
Magnitude of whole slide image dataset

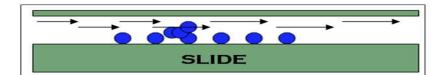


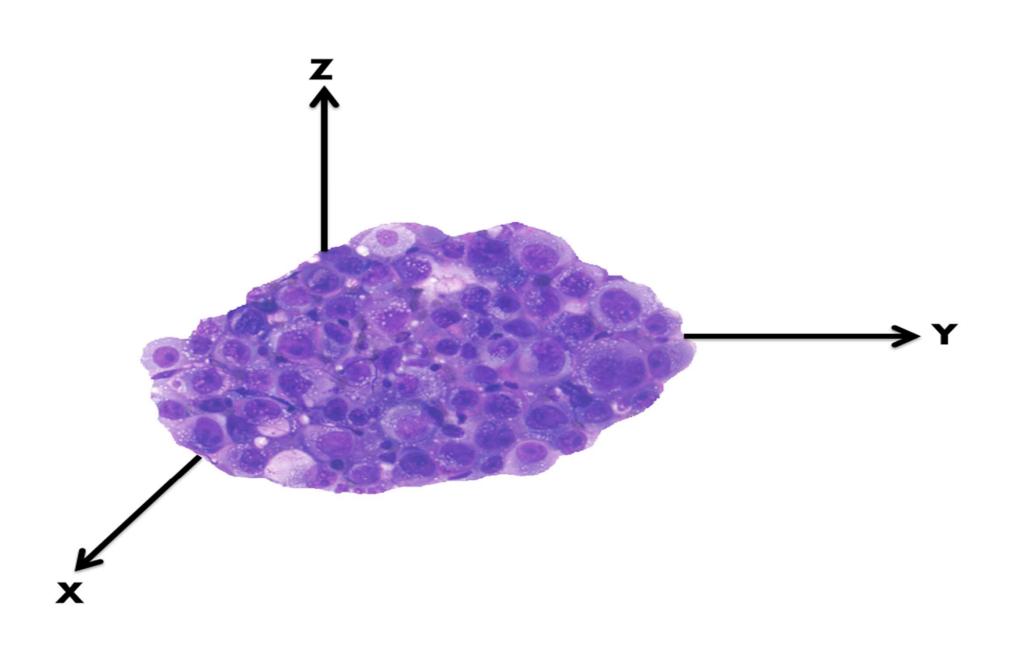
In terms of total pixel-normalized display in the same field of view, a 2k x 2k pixel digital radiographic chest X-ray image (A) is dwarfed when compared to a 40x scan of a typical 2.5 x 2.0 cm biopsy (654Mb with 20:1 loss compression).

Z stacking









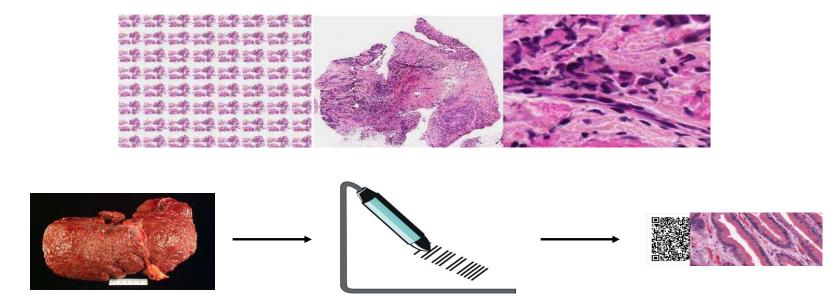
spatial sampling periods of between 0.25 to 0.5

WebPathology.com

COMBINING MORPHOLOGY WITH PATIENT INFORMATION: AP-LIS INTEGRATION IS KEY



MULTIPLE AREAS OF THE PATHOLOGY LAB HAVE MANUAL STEPS, AND LACK STANDARDIZATION



AUTOMATION AND STANDARDIZATION OF WORFLOW IS CRITICAL

FACTORS TO CONSIDER WHEN SELECTING AN WSI

- Type of lab and application
 - Reference/research/public health
 - Clinical
 - Hybrid
- Volume of specimens
- Types and number of slides
- Size of staff/users
- Existing system
 - Determine which areas will be affected
- Requirements and expectations
 - Be realistic
 - Adoption issues

Some WSI Imaging Devices



NEWER TECHNOLOGIES: COST DECREASING



WHOLE SLIDE IMAGING: Decade of Evolution COST ADOPTION













Favorable Regulatory Environment for AI/Digital Pathology

The Food and Drug Administration (FDA) has approved a number of digital pathology solutions for use in the United States



April 13, 2017

Philips receives FDA clearance to market Philips Intellisite Pathology Solution for primary diagnostic use in the US





May 29, 2019

Leica Biosystems receives FDA clearance for Aperio AT2 DX digital pathology system for primary diagnosis

Sept 21, 2021

(FDA) has granted de novo marketing authorization for Paige Prostate, a clinicalgrade AI solution for prostate cancer detection

Favorable Regulatory Environment for Al/Digital Pathology

The Food and Drug Administration (FDA) has approved a number of digital pathology solutions for use in the United States

Hamamatsu Photonics Announces U.S. FDA clearance for the NanoZoomer S360MD Slide scanner system for Surgical Pathology Diagnostics

2022/10/04



10/04/2022

Digital slides will introduce new costs!

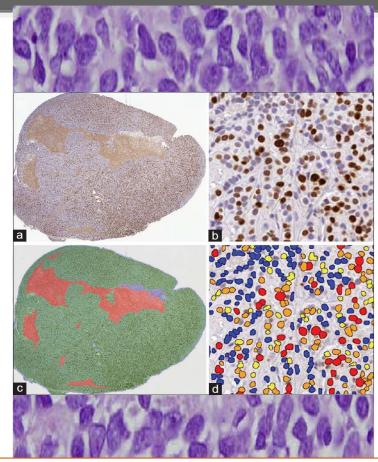
- New capital equipment
 - Scanners
 - Servers
 - Storage (terabytes per week! But storage is cheap)
 - BARCODES are a MUST



The Ohio State University

Challenges in Pathology Today

- Lack of standardization
- Subjective diagnosis
- Manual process
- Shortage of pathologists worldwide
- Over-worked
- Explosion of medical knowledge
- Time-cosuming tasks such as counting cells



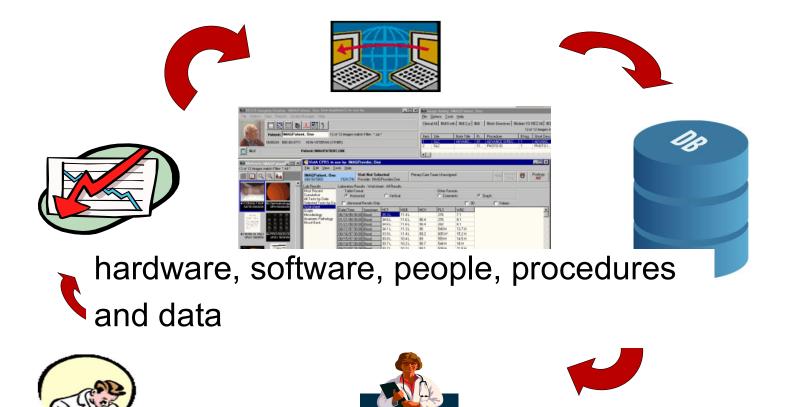
How can DP/AI help pathology?

- Standardization
- Objective Dx
- Automated process
- Accurate Dx
- Faster Dx
- More time to do other tasks
- Keep up with the knowledge

AI-POWERED PATHOLOGY



DIGITIZING PATHOLOGY: OPPORTUNITY TO IMPROVE ALL ASPECTS OF PATHOLOGY WORKFLOW



PATHOLOGY INFORMATICS DRIVEN WORKFLOWS INTEGRATION OF EMR, LIS, WSI

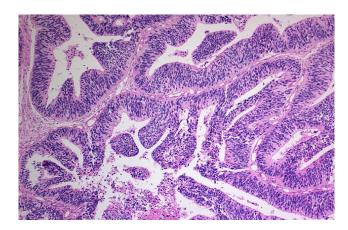


ENTERPRISE IMAGING INTEGRATED DATA AND IMAGES MORE DIAGNOSTIC AND PROGNOSTIC TOOLS

THE OHIO STATE UNIVERSITY

Integration: LIS, Scanner, Workstations interfaces

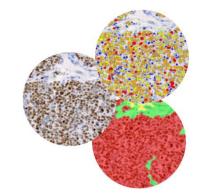
between WSI systems, laboratory information systems and the electronic medical record



The Solution for Enterprise Digital Transformation







Scanners

Any Scanner. Any LIS. Any Image Analysis.

Image Analysis and AI Tools

Creating an OPEN platform for Digital Pathology and AI tools Slide courtesy of Inspirata



TOP 5 REASONS FOR IMPLEMENTING DIGITAL PATHOLOGY and AI FOR DIAGNOSTICS

- **1. INCREASED PRODUCTIVITY**
 - IMPROVED INFORMATION MANAGEMENT, WORKFLOW DISTRIBUTION, INTEGRATION OF DATA
- **2. IMPROVED QUALITY/BETTER MEDICINE**
 - QUALITY ASSURANCE, RAPID SECOND REVIEWS, EASIER ACCESS TO SUB-SPECIALIST
- **3. INCREASE REVENUES**
 - INSOURICING (Digital Consults), PULL-THROUGH REVENUES, BRAND RECOGNITION
- 4. COST SAVINGS
 - CONSOLIDATION, REDUCED COSTS WITH MOVING SLIDES AROUND
- **5.** ARTIFICIAL INTELLIGENCE
 - IMAGE ANALYSIS/WORK FLOW ALGORITHMS, COMPUTER AIDED DIAGNOSIS

OSU Strategic Pathology Imaging Roadmap



Static Images – Patient reports/Gross/Microscopic

Robotic Microscopes - Telepathology

WSI-Education/QA

WSI-LIS Integration

WSI-Primary Diagnosis

WSI- Pathology PACS

DEEP LEARNING TOOLS/ARTIFICIAL INTELLIGENCE

Department of Pathology OSUWMC

- 85 faculty
 - AP, CP, MP
 - Outreach
 - Experimental
 - Nationwide Children's Hospital (NCH)
- Sites
 - OSUWMC/James CCC
 - Molecular lab
 - OSU East
 - 3 outreach hospitals
 - NCH

40





Clinical Volumes Increasing

Surgical Pathology cases – 90,000

Approximately 600,000 slides including H&E, IHCs and special stains

Approximately 3,000 slides/day





OSU Digital Pathology Timeline



January 2017 Digital pathology lab space secured on the 18th floor of the James



May 2017 Digital pathology scanners installed



July 2017 Full scale digital pathology WIS services are launched



EPIC Beaker goes live. This allows for a list of scanned slides, easily accessible link to click through slides, and the ability to sort slides by Part ID, Block ID, Slide ID, Image creation time.



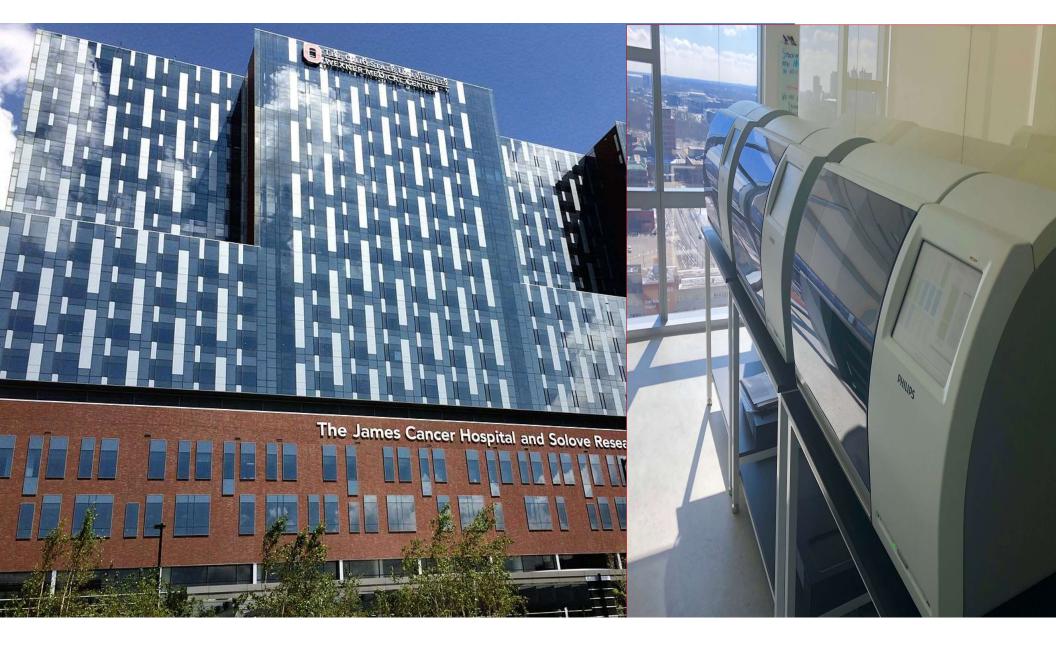
December 2019

A large portion of digital pathology services are moved to a larger centralized lab space

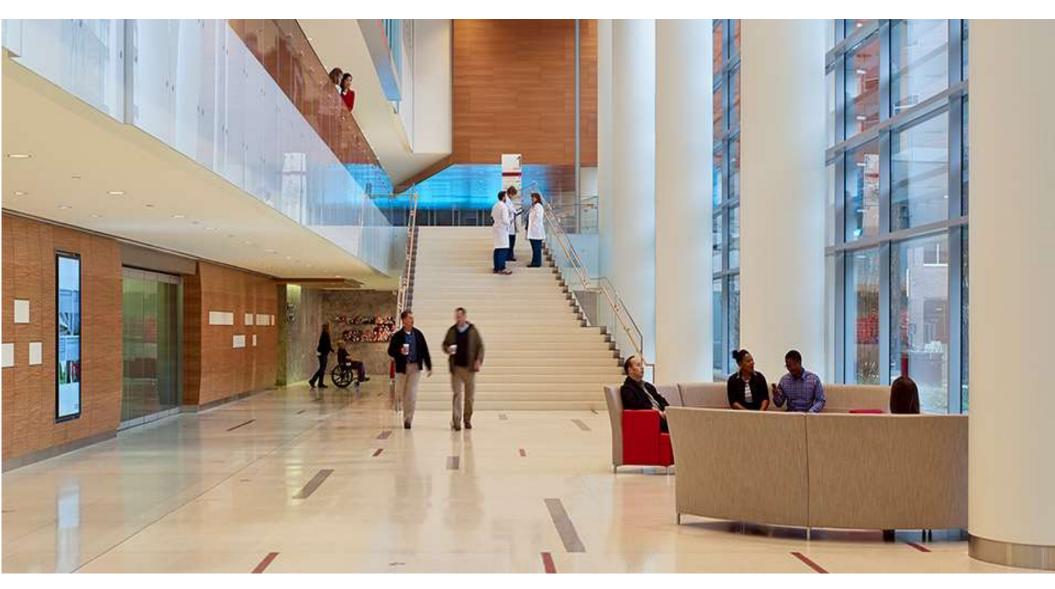


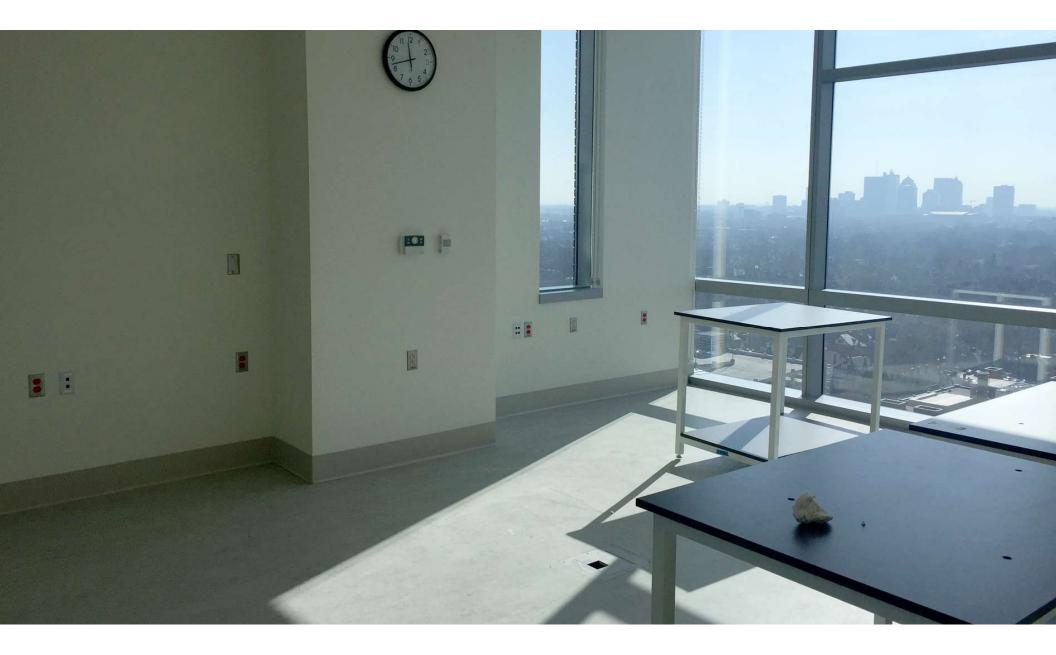
November 2019

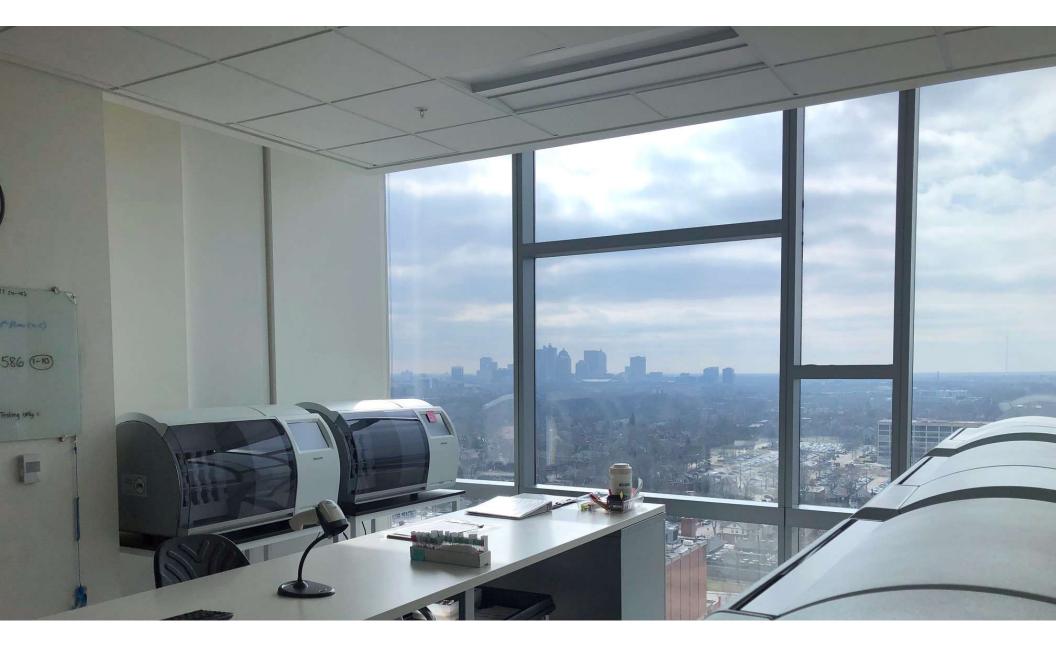
Telepathology scanner installed at Adena Regional Medical Center





















Thousands of Patient Slides in Archives

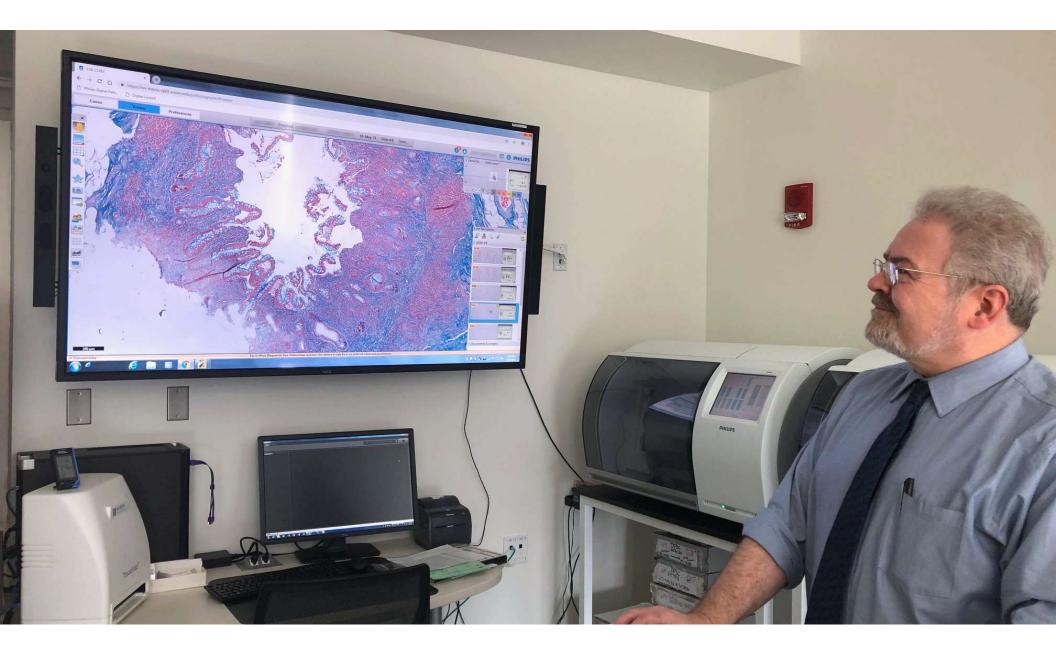


GLASS SLIDES IN RACKS









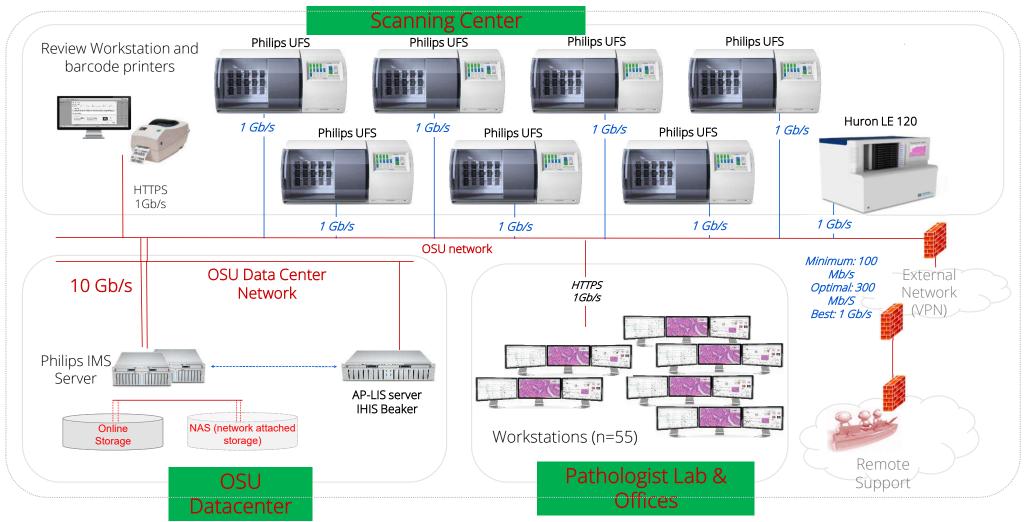


TRAINING THE NEXT GENERATION OF HISTOTECHS

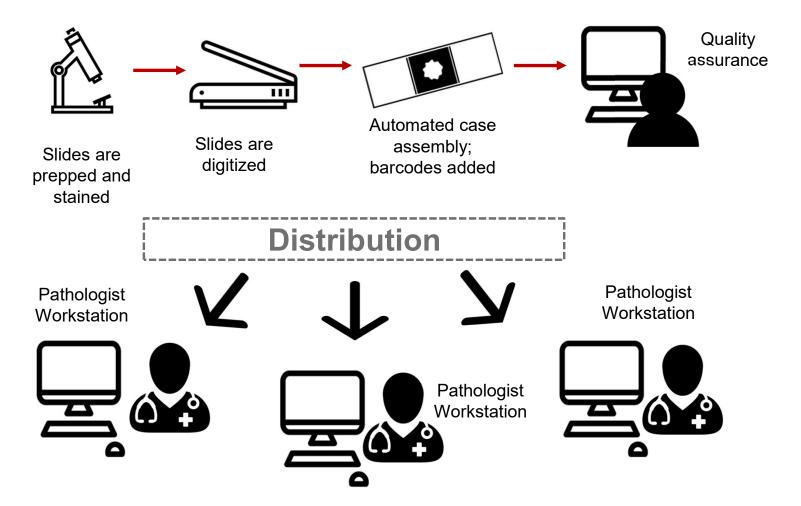




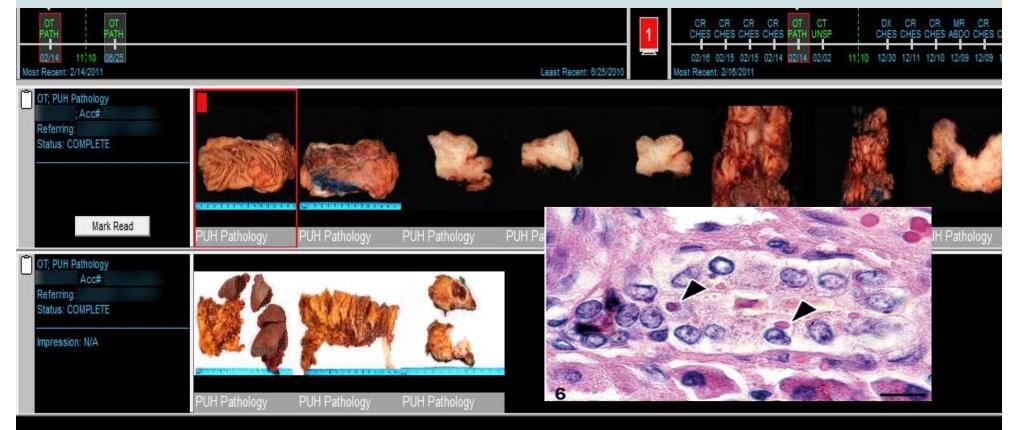
OSU Digital Pathology Clinical Workflow System

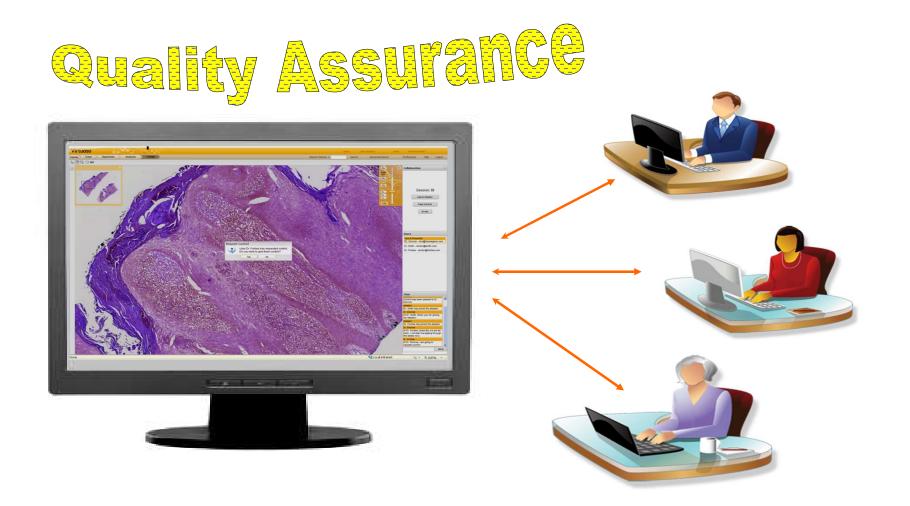


Whole Slide Imaging (WSI) Workflow



BETTER QUALITY: DIGITAL TUMOR BOARDS- REVIEWING PRIORS





Impact on pathology workflows

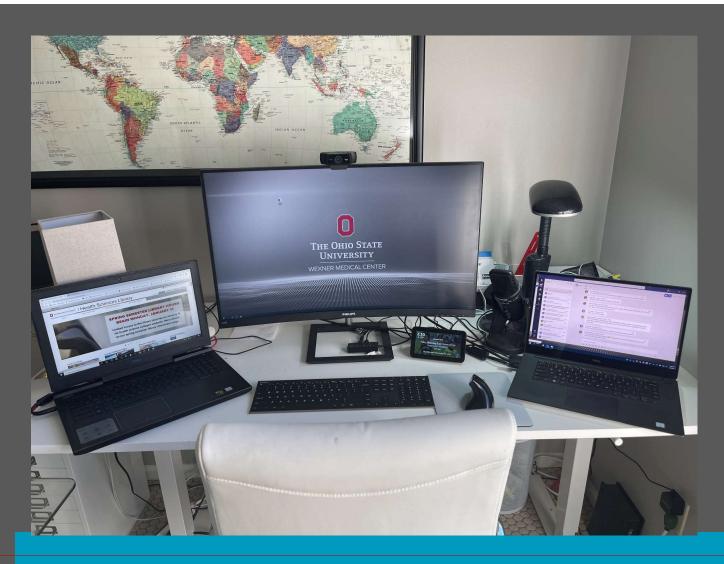


CREATING A WORKPLACE THE OHIO STATE UNIVERSITY

WEXNER MEDICAL CENTER



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Outstanding List - CLINICAL LAB UH - AP Pathologist Workqueue - 1 out of 10

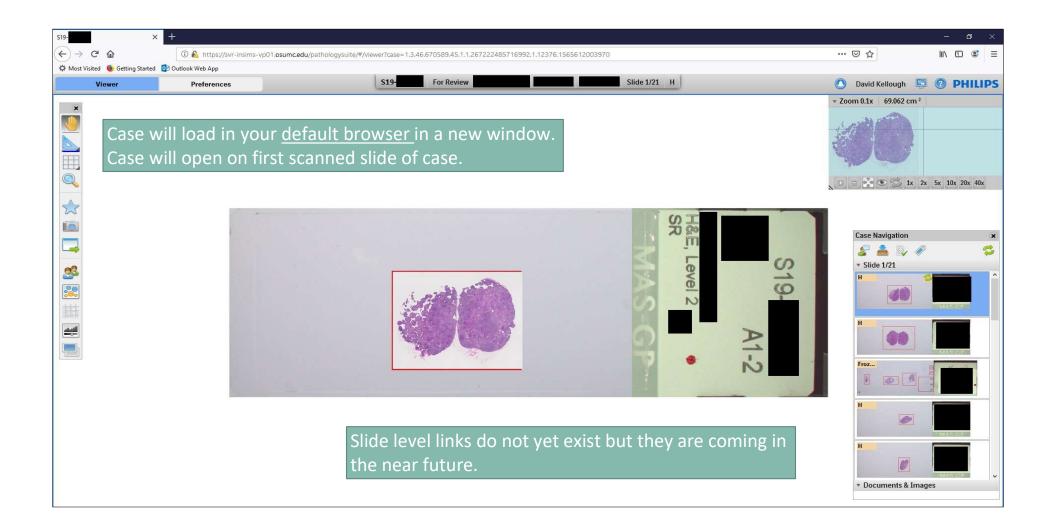
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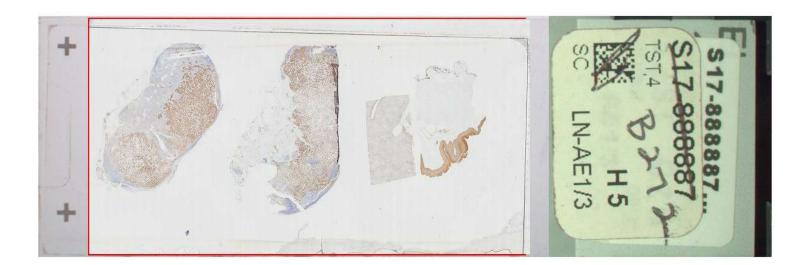
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Slide-level Links in Laboratory Information System Provides a More Efficient Sign-out Workflows to the Pathologists

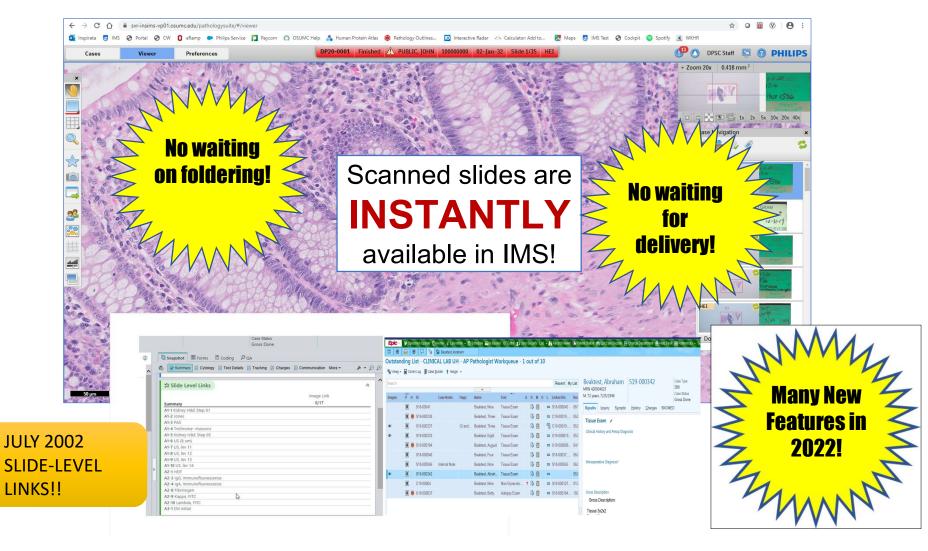
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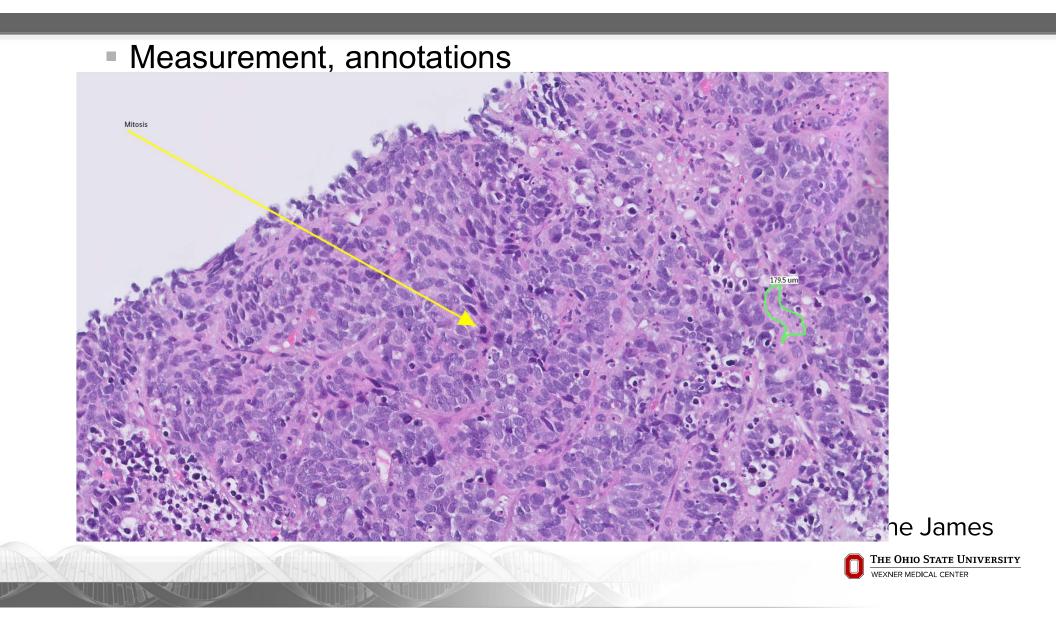


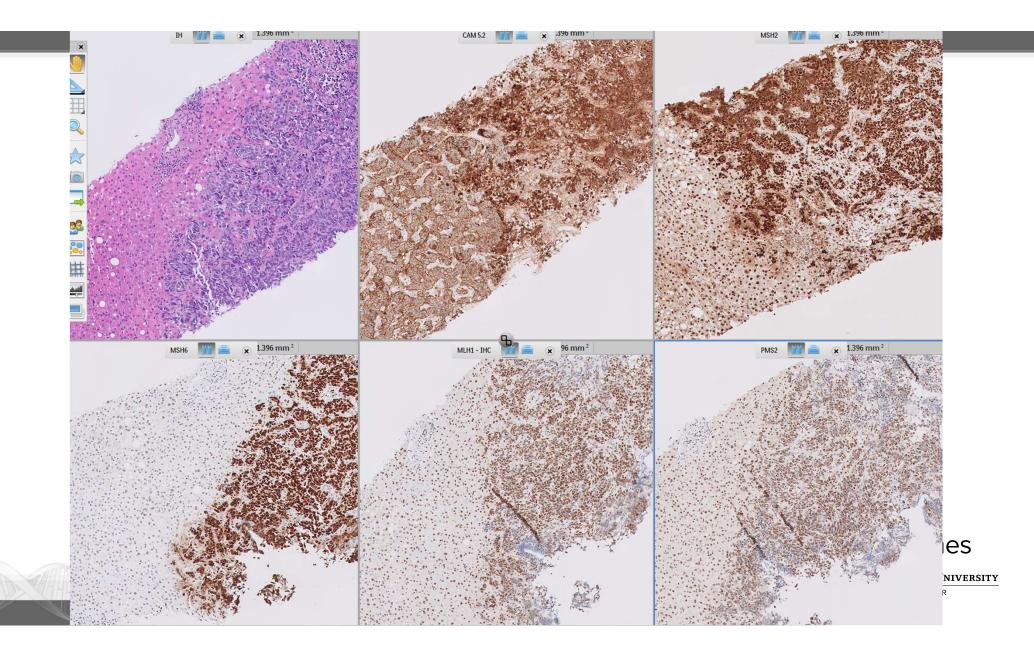
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Fully Integrated Digital Pathology Workflow For Improved Patient Care: Digital Pathology for Primary Diagnosis





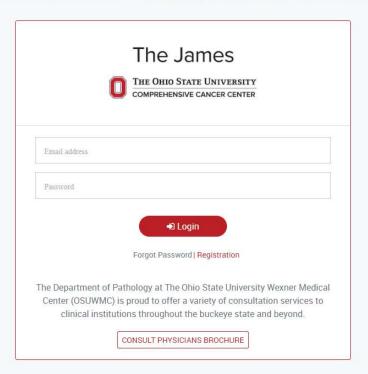


Consults with colleagues



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The OSUWMC Pathology Consult Services

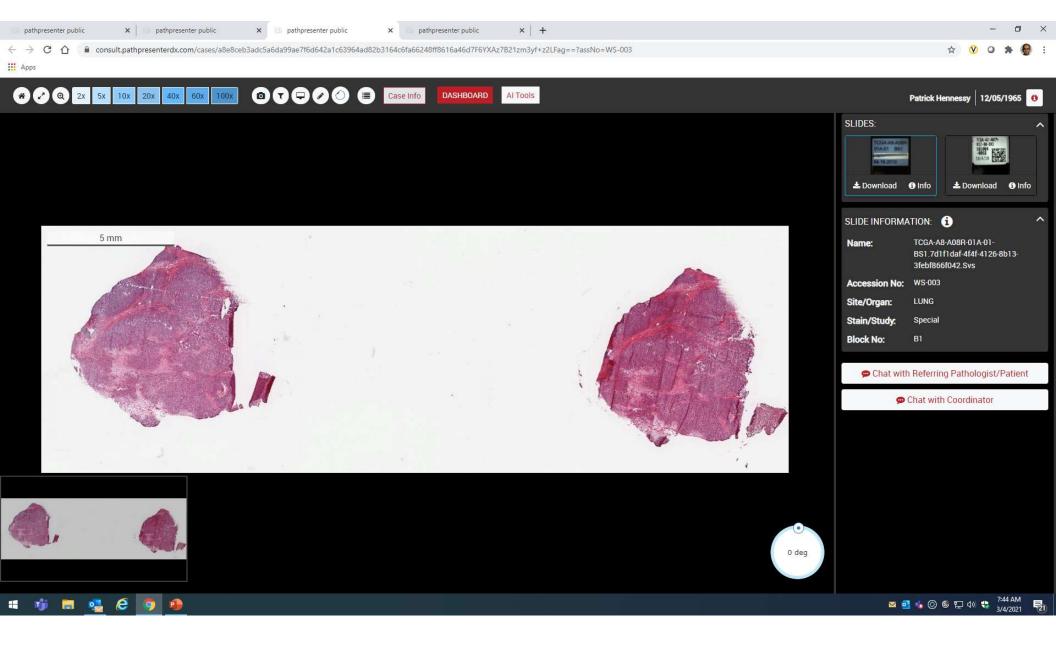


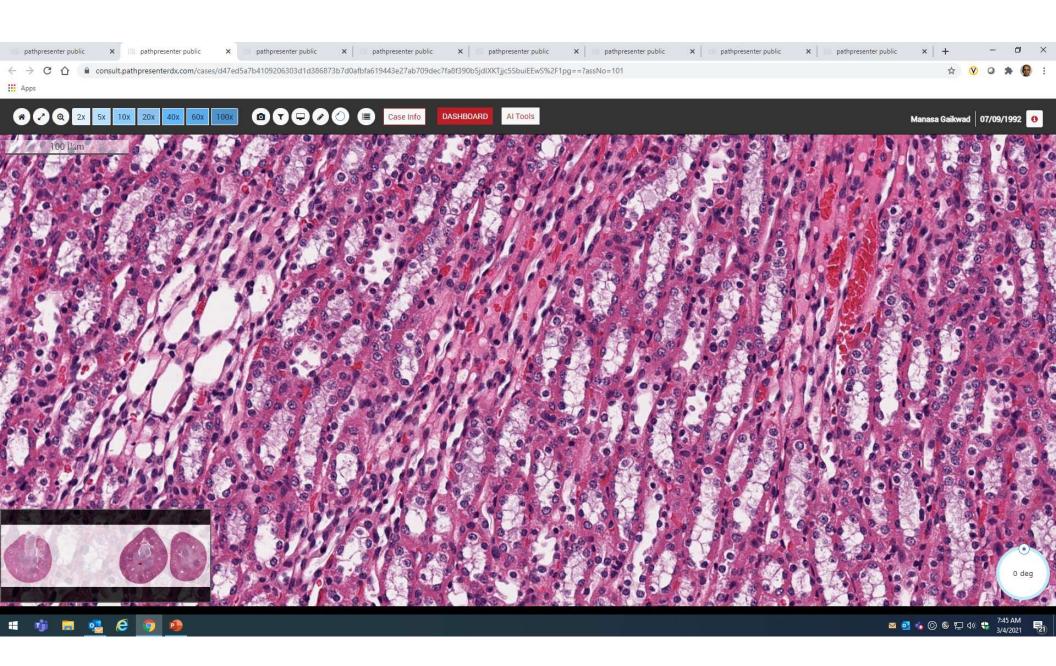


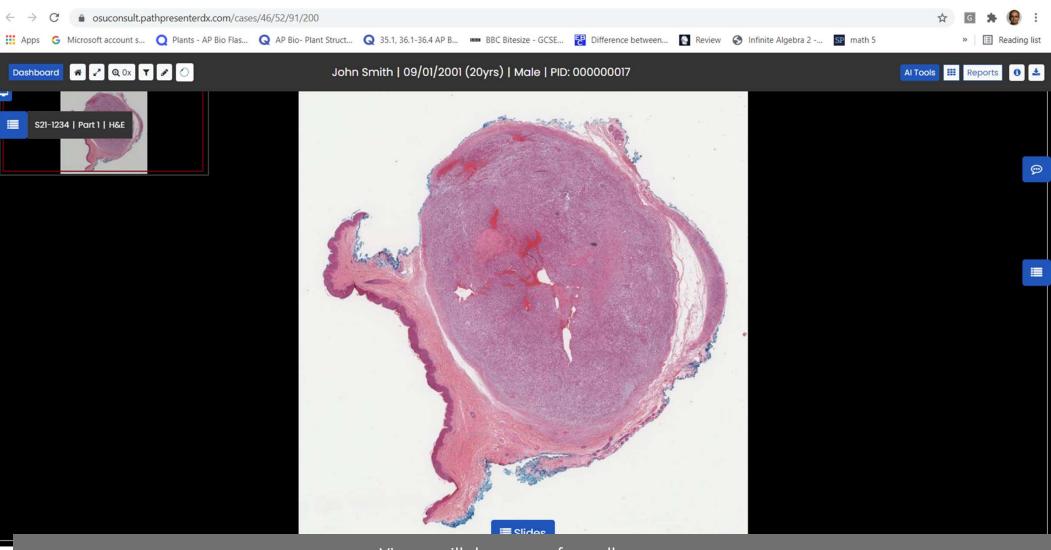
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case-59	abc	def		RAJENDRA SINGH	.	MALE	Draft		Assign Edit	
case-58	WS-005			Tara jacob Tyre	06/06/1970	FEMALE	Assigned	Lung	WS-005-B , WS-005-A 🛛 🖉 Un Assign	
case-57	WS-004			Amy sd Hardy	05/04/1985	FEMALE	Assigned	Lung	WS-004 WS-004	
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case-55	WS-002			Jason Arndt	04/05/1978	MALE	Assigned	Lung	WS-002 WS-002	
case-54	WS-001			Olivia st Whisman	04/19/1991	FEMALE	Assigned	Skin	WS-001 WS-001	
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Viewer will show cases from all scanners



FINE NEEDLE ASPIRATION SERVICE The James

Solution for FNA





WEXNER MEDICAL CENTER

Hybrid robotic microscope/WSI scanners



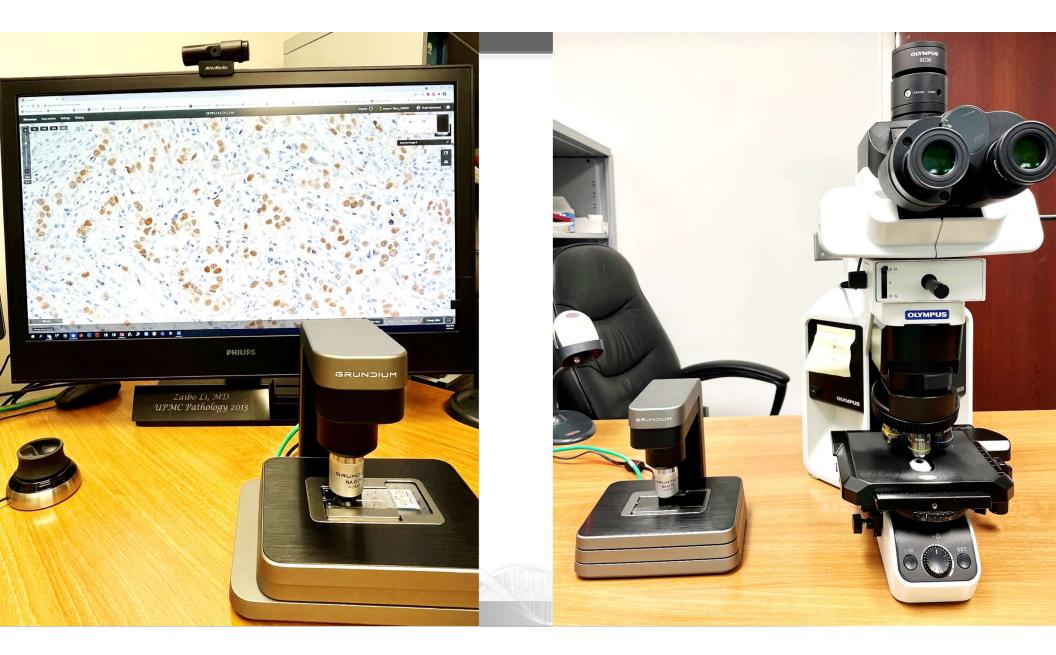


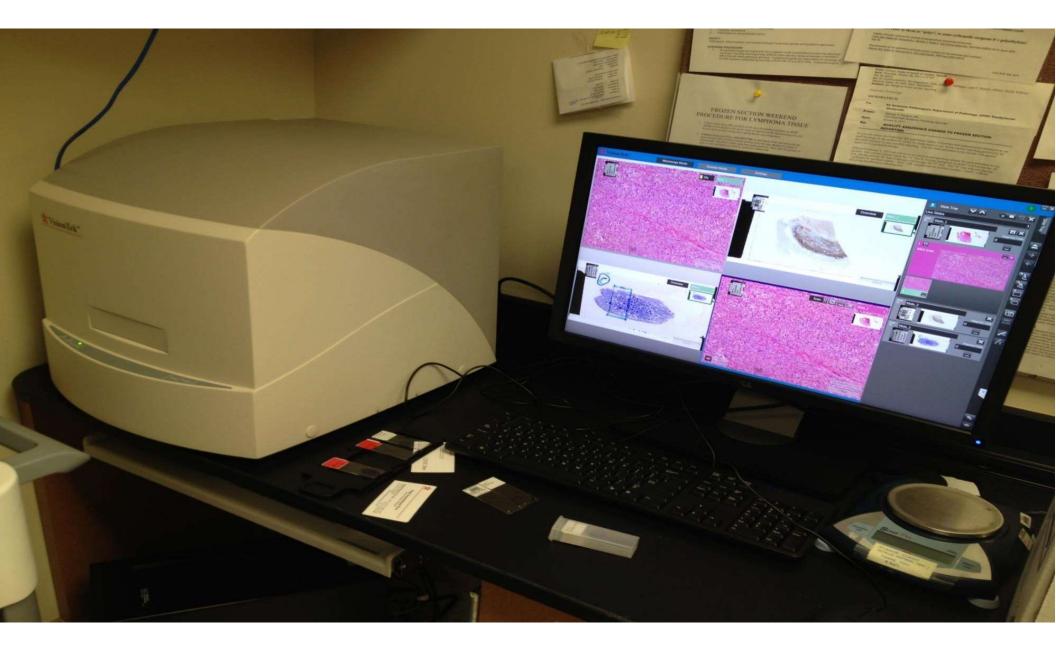
MikroScan SL5

Grandium OCUS40









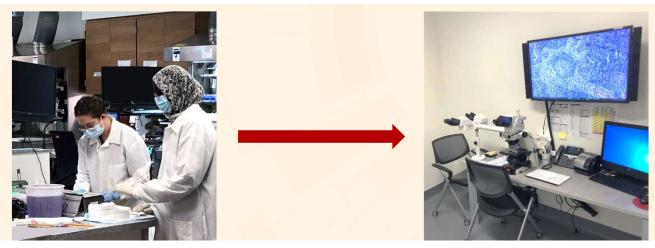












FROZEN SECTION LABORATORY – LEARNING IN A DIGITAL WORLD





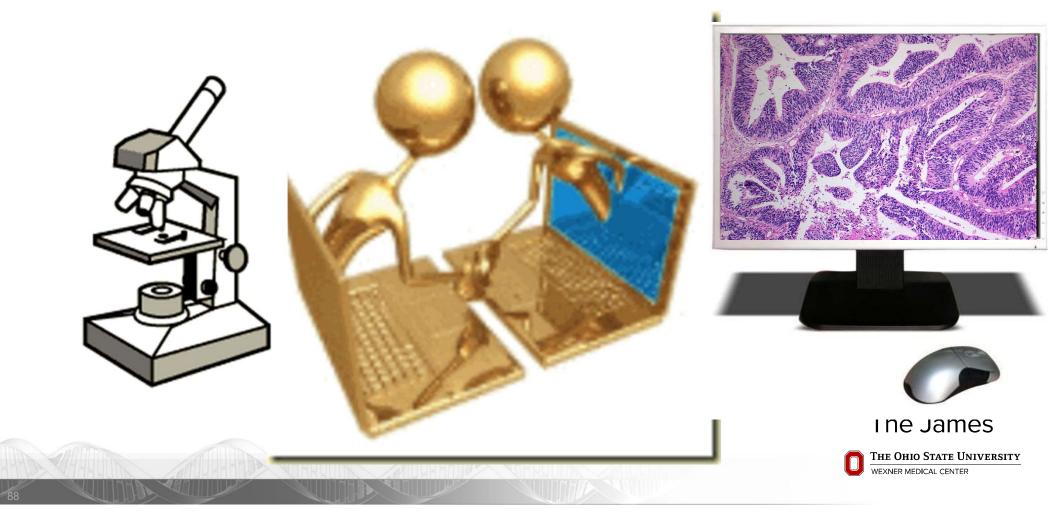




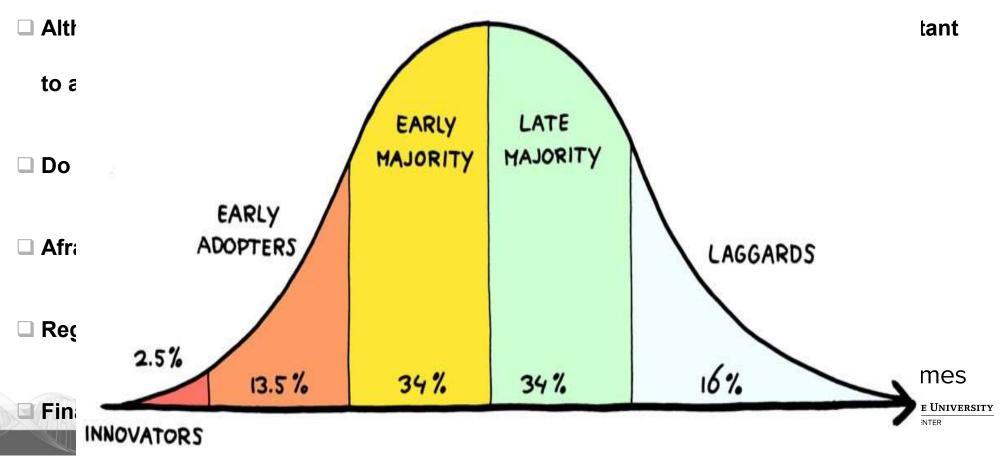
Pathologist fully controls the computer which connects to the instrument remotely, can switch between slides, change magnification, move the slide and adjust the fine focus.

2,900,000 SLIDES SCANNED SINCE DIGITAL PATHOLOGY GO-LIVE, 2017-2022

PATHOLOGIST ADOPTION ISSUES







Digital slides has <u>IMPROVED my workflow by...</u>

- Direct interface to LIS
- Sharing cases with consultants, clinicians
 - Tumor conferences
- Flagging cases

90

- Controls instantly available
- Digital slides has <u>IMPROVED my TAT by...</u> automatically distributing workload
- Glass slides go directly from cover slipper to scanner to proper pathologist
- Cases are instantly available upon scanning
- Previous cases are instantly available

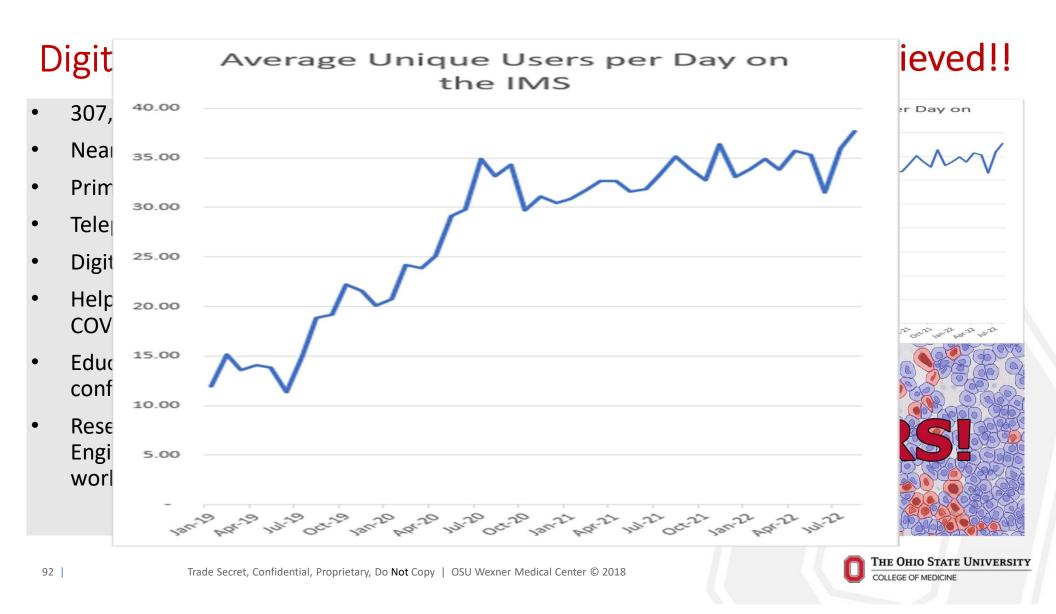




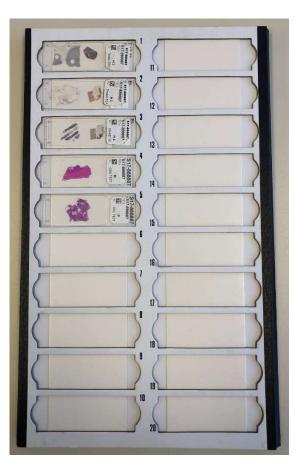
THINGS THAT STILL ARE AN ISSUE

- We still can't use digital slides to perform some necessary tasks!
- Polarization to identify foreign material
- New ways to teach residents?
- 3D imaging for cytopathology - Z stacking-Expensive
- Oil for the hemepath guys Messy and not a good solution out there

THE JAILES



Those 2,900,000 slides and 307,000 cases include:



Clinical cases:

Sep 2011 to the present day.

All consults:

2017 to the present day

Tumor boards

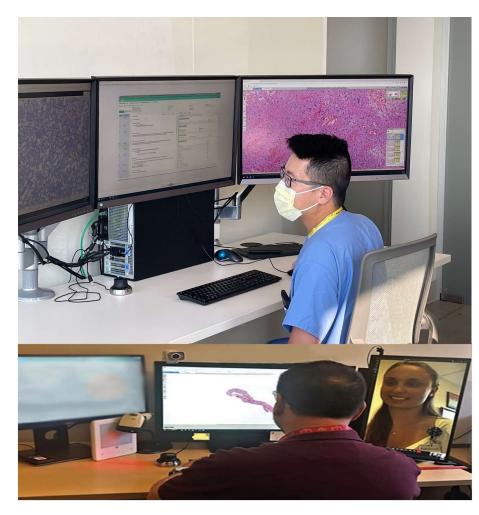
Residents' Slide Set

Teaching sets

Various research collections

Special requests.

Digital Pathology Enhanced Resident Training



Over 200 primary diagnosis sign-outs a day via digital pathology.

Nearly 80 consult sign-outs a day.

. . .

Supports numerous tumor boards.

Digital images used for dozens of research endeavors.

System used regularly by 144 faculty, residents, fellows, and staff.

Pathology Department Benefits: The Quality Advantage

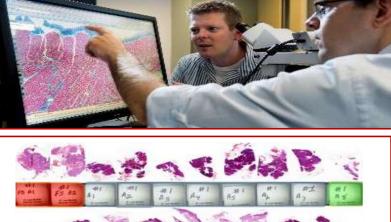
1 Decrease the time to diagnosis

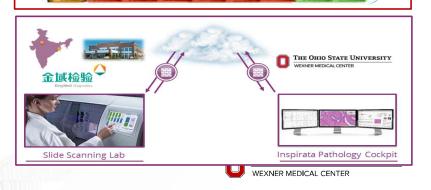
- The digital pathology cockpit focuses pathologists time on what they do best diagnose disease
- Diagnostic algorithm support make pathologists even faster

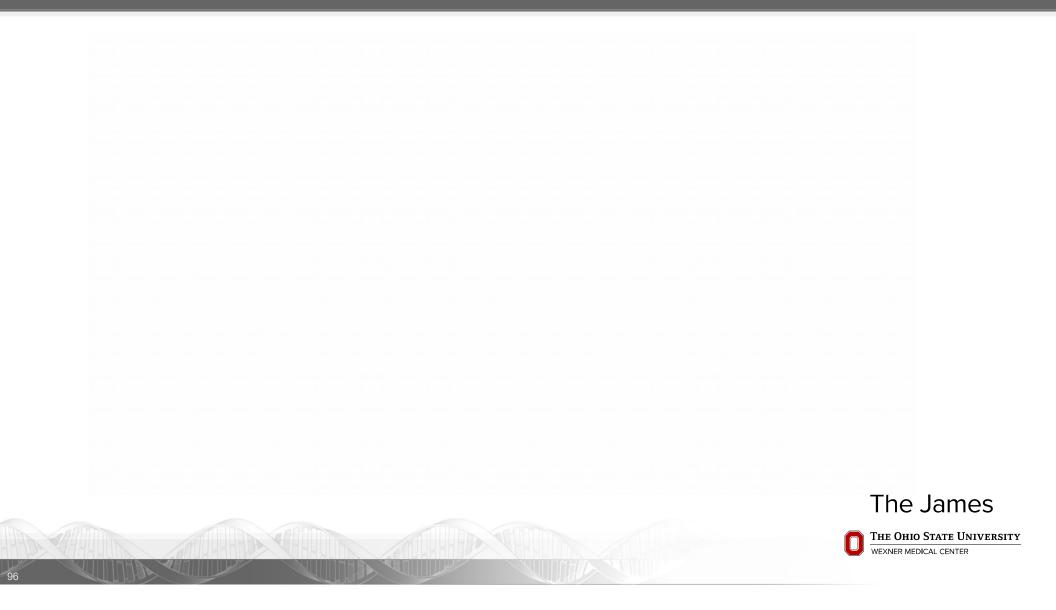
Consult Instantly

- **2** ~ Telepathology to load balance in real time
 - ~ Consult locally, regionally and globally
 - Gain access to complex cases
 - ~ Rapid second reviews
 - Improved and streamlined quality reviews

Al-Janabi, Shaimaa, André Huisman, and Paul J. Van Diest. "Digital pathology: current status and future perspectives." *Histopathology* 61.1 (2012): 1-9.







QUESTIONS???





Creating a Cancer-free World. One Person, One Discovery at a Time.

The James



Current and Future Applications of Artificial Intelligence in Pathology: Are we ready for prime time? Anil V Parwani, MD, PhD, MBA

The Ohio State University Comprehensive Cancer Center - Arthur G. James Cancer Hospital and Richard J. Solova Pasaarch Institute



CREATING THE PIXEL PIPELINE

COMPUTATIONAL PATHOLOGY/AI APPLICATIONS

WHAT CAN WE DO WITH THE PIXELS??

1000

Pathologists today perform many complex manual tasks

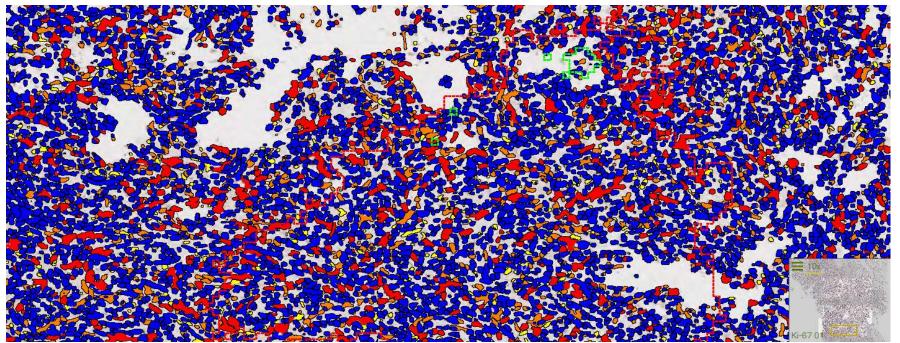




Al can automate many of the complex manual tasks – such as object quantification, tissue classification based on morphology and rare target identification – that used to take hours or days to perform.

Some of those tasks now only take minutes using image analysis and AI

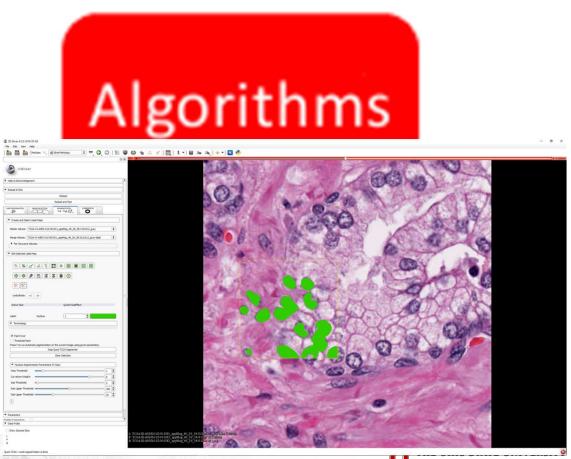
Quantification of Biomarkers: MORE OBJECTIVE, ACCURATE AND FASTER



	Measuremen	Count	Count	Count	Count	Total Cell	Total Positive Ki-	% Positive Ki-	% Low Ki-	% Medium Ki-	% High Ki-		
Study Unit	t	Low	Medium	High	Negative	Count	67	67	67	67	67		
Demo									15.043499		8.6470403		The lames
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GOING BEYOND THE GLASS SLIDE

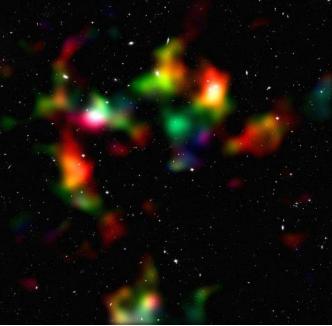
- Digital pathology/Al opens new doors in algorithms
 Identification
 - Quantification
 - Synthesis



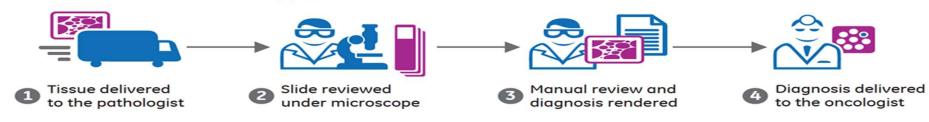
WEXNER MEDICAL CENTER

What ELSE can we do now?

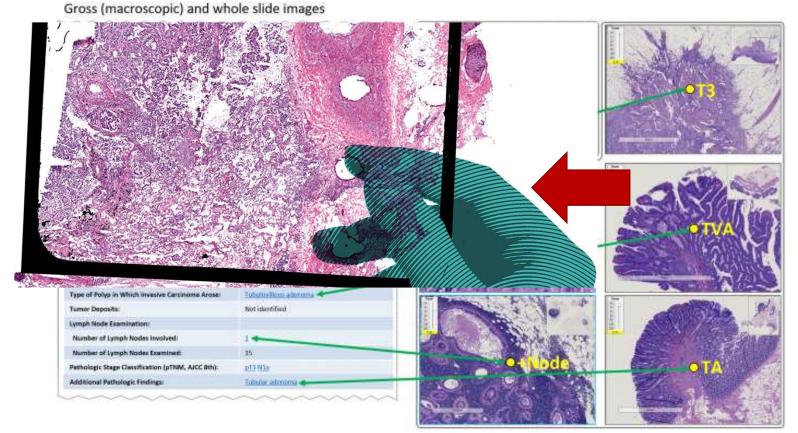




Traditional Pathology

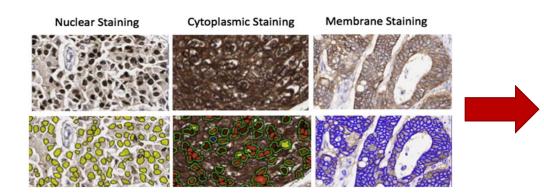


Interactive Multimedia Pathology Reporting



Roth, Christopher J et al. "Multispecialty Enterprise Imaging Workgroup Consensus on Interactive Multimedia Reporting Current State and Road to the Future: HIMSS-SIIM Collaborative White Paper." *Journal of digital imaging* vol. 34,3 (2021): 495-522. doi:10.1007/s10278-021-00450-5

Future State: Quantitative Immunohistochemistry Objective Data Directly to Patient Chart and Integrated Reporting



INTEGRATED AI-DRIVEN SMART REPORTING SYSTEMS..MULTIMODALITY DRIVEN WITH DATA ANALYTICS AND VISUALIZATION

Age: 66	Sex: Female	Race: White	Location:	Da	te: 02/11/91
Address:		Nuclear Staining	Cytoplasmic Si	aining	Membrane Staining
Address		DITA SCHOOL AND			LINESSAN MEN SARREN
HISTORY (OF CASE: Multi	p · · · ·		S KAR	CALLY S
inditiple to	amors	and some	and the second	3	1911 0
CLINICAL	DIAGNOSIS: Ca	ar is set			
POST-OPE	RATIVE DIAGN	IC S S S			
Surgeon: S	So Long, MD		8		
GROSS DE	SCRIPTION:	1098 577 86° 80	and Managements	Die Carlo	States and the second sec
The spe	cimen is recei	ved in two par	ts. They are la	beled	#1. "biopsy
	mor", and #2.				
					orrhagic. They
					ists of multiple
	of fatty yellov They are subn				.2 to 1.0 cm in
diameter.	They are subn	hitted in their	entirety for pr	ocessin	ig.
MICROSCO	OPIC:				
Section	of bladder cor	ntains areas of	f transitional o	ell card	inoma No
	asion can be i				
	ory reaction w				
	Sections are ex				nph node
contains n	ormal node w	ith reactive ge	rminal centers	s.	
DIAGNOSI	S:				
1. Papil	lary transition	al cell carcinor	ma, grade II, b	ladder,	biopsy.
2. Acut	e and chronic	inflammation.	most consiste	ent with	recent biopsy
procedure					, , ,
3. Scale	ne lymph nod	e, left, no path	nologic diagno	sis.	
Mike O'Se					
Pathologis					

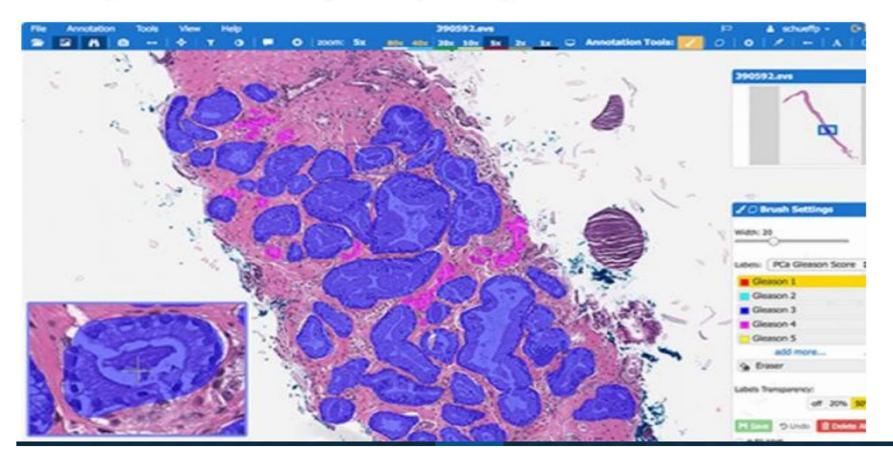
Hamilton PW, et al. Digital pathology and image analysis in tissue biomarker research. Methods. 2014 Nov;70(1):59-73.



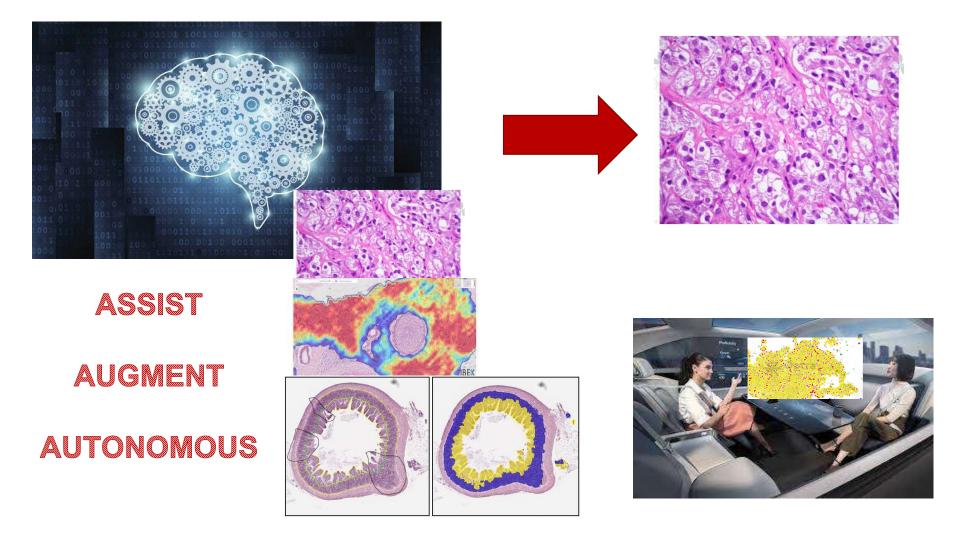


FDA Grants Breakthrough Designation to Paige.Al

Designation expedites product development and provides priority regulatory review for clinical-grade artificial intelligence in pathology



HOW CAN AI TOOLS HELP A PATHOLOGIST?



Deep Learning in Breast Pathology

intensive with poor reproducibility and

quality concerns. New approaches use fundamental AI research to build tools to make pathologic analysis more efficient, accurate, and predictive. In the 2016

Camelyon Grand Challenge for metastatic

cancer detection,⁶⁹ the top-performing

entry in the competition was an Al-based

computational system that achieved an error rate of 7.5%.⁷⁰ A pathologist

reviewing the same set of evaluation images achieved an error rate of 3.5%.

Combining the predictions of the AI system

with the pathologist lowered the error rate

to down to 0.5%, representing an 85%

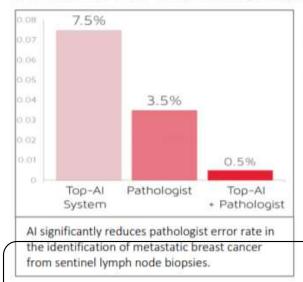
reduction in error (see image)." This

example illustrates how fundamental research in AI can drive the development

National Institutes of Health (NIH) grants-supported research

ARTIFICIAL INTELLIGENCE FOR COMPUTATIONAL PATHOLOGY

Image interpretation plays a central role in the pathologic diagnosis of cancer. Since the late 19th century, the primary tool used by pathologists to make definitive cancer diagnoses is the microscope. Pathologists diagnose cancer by manually examining stained sections of cancer tissues to determine the cancer subtype. Pathologic diagnosis using conventional methods is labor-



of high performing computational systems that offer great potential for making pathological diagnoses more efficient and more accurate.

THE NATIONAL ARTIFICIAL INTELLIGENCE RESEARCH AND DEVELOPMENT STRATEGIC PLAN

National Science and Technology Council

Networking and Information Technology Research and Development Subcommittee

October 2016



Automated Biomarker Analysis for Clinical Diagnostics Regions of Interest Scoring Methods

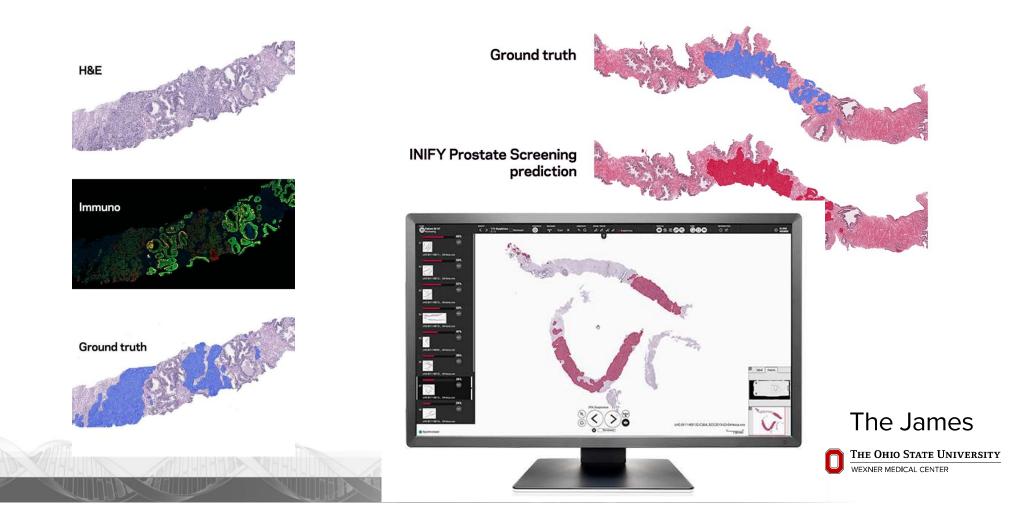
Slide Score: Multiple Regions

Highlighted Region Score(s)

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Annotations	Algorithms	🗹 Sna	pshots	se	earch		So	rt By: Date 👻		2.5		1	00	
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1+ Nuclei (#)	171	74	4	2	9	7	58	17	- (B)			-		
2+ Nuclei (#)	317	110	43	35	40	36	47	6	1 1		- CO.			10 :
3+ Nuclei (#)	36	4	4	8	11	8	1	0	11-11			198	1 m //	
Positive Percentage	83.4	87.9	83.6	86.5	84,5	92.7	71.6	85.2	3-3-4	01 0	Se des	1.4		
H-Score	145	143	167	185	172	187	105	107						0.0

Future State: DP/QIA Standardized Diagnostic Tests

Al-powered tool that provides pathologists with valuable decision support when identifying cancer in prostate biopsies



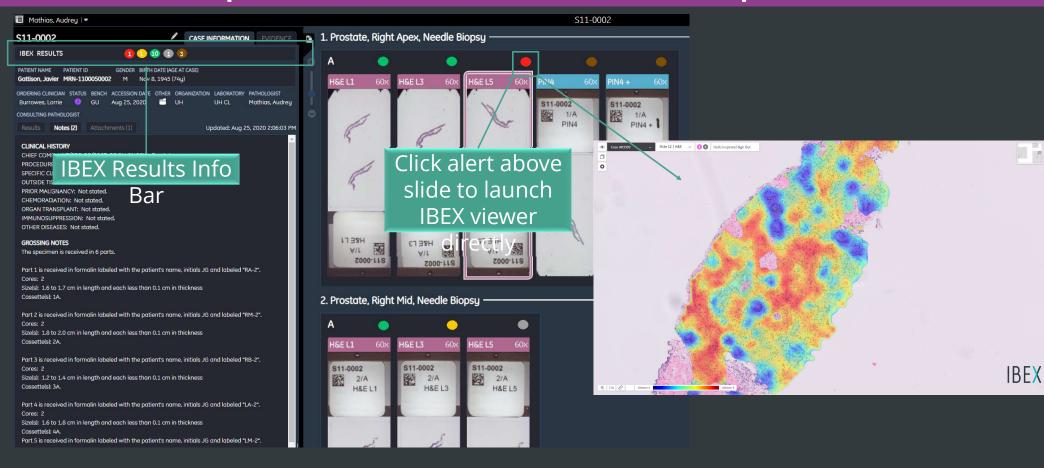
IBEX: Suspicious Lesions in Prostate Biopsies

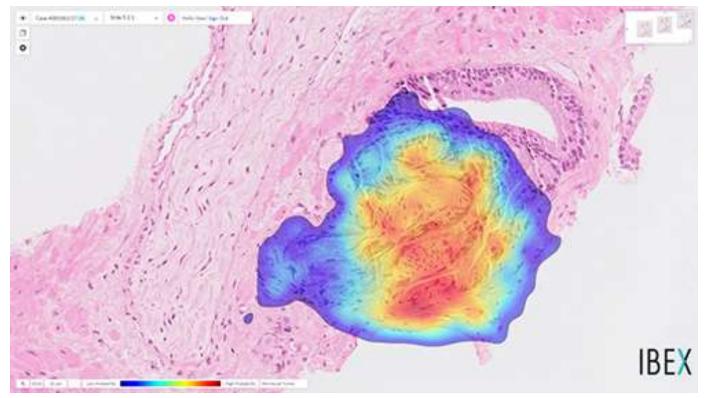
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O My Folders	o \$11-0016	Aug 25, 2020	Koll, Guy	0	Wilgus, Allan	2	GU	= 1	1	Testis and Proximal Spermatic Cord	1 10		1
Kidney	o \$11-0029	Aug 25, 2020	Dyess, Erik		Santini, Julianne	0	GU	-	1	Left Testicle, Radical Orchiectomy	11	titore enteriment	1
Prostate	o 511-0030	Aug 25, 2020	Burgo, Neva		Cogswell, Tia	÷	GU	=	1	Retroperitoneal Mass, Excision	27		1
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Notifications in expanded case row

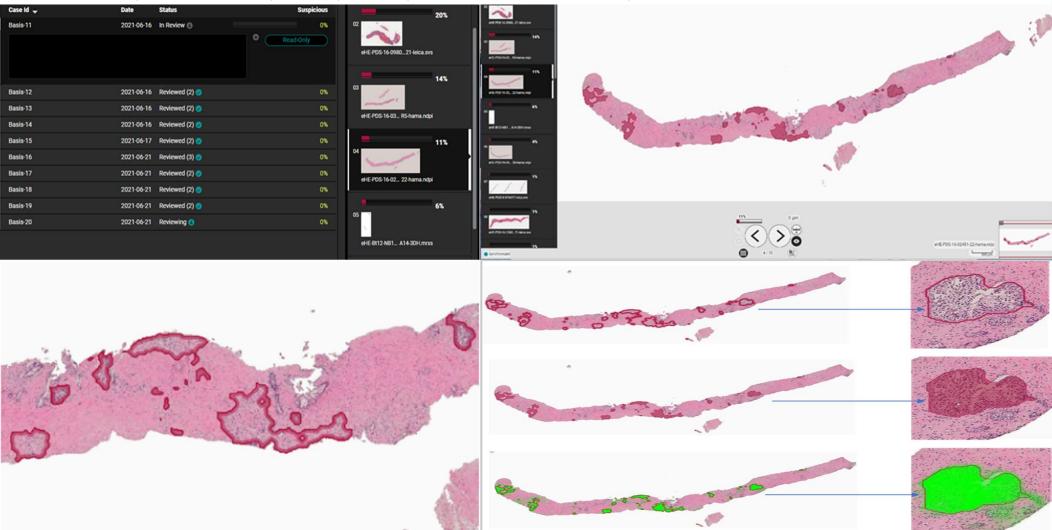
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S11-0004	Aug 25, 2020	Bodiford, Marylo	u 4	Maddix, Allan	0	ТΧ	≓ ∞	1	Kidney Allo	ograft, Needle Biopsy	6	1	romosiny in neuropy rearies.
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MRN-1100050002	GENDER BIRTH DATE (AG M Nov 8, 194		ORGANIZATION University Hosp	LABORATORY Dit UH Core L	PATHOLOGIST Mathias, A		2. F 3. F	Prostate, Right Prostate, Right	Mid, Needle Bio Base, Needle Bi) Likely benign
							5. F	Prostate, Left M	bex, Needle Bio id, Needle Biops ose, Needle Bio) Out of Focus) IHC slide

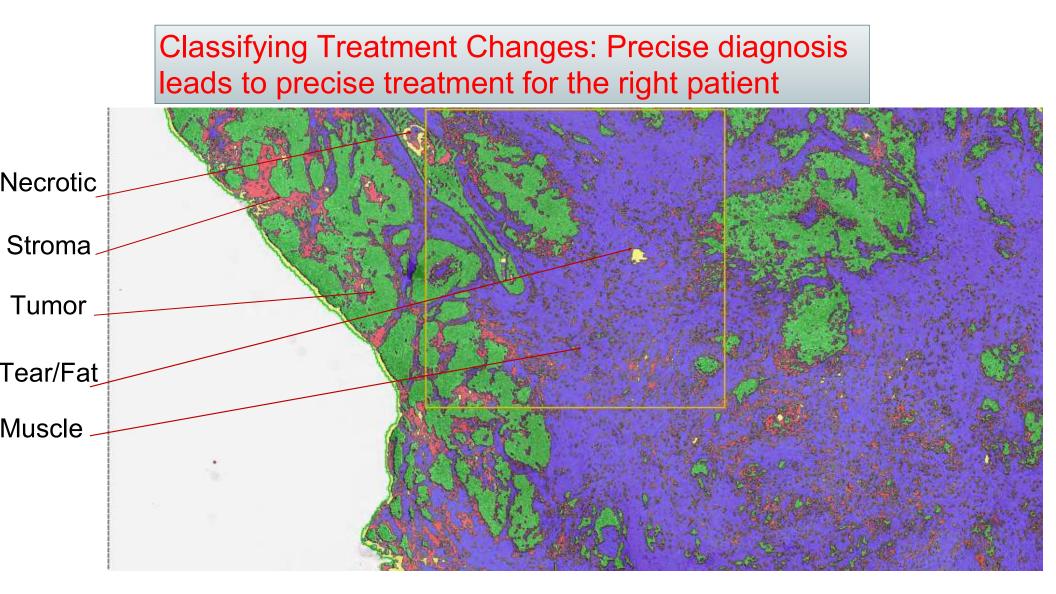
IBEX: Suspicious Lesions in Prostate Biopsies





The algorithm achieved an AUC of 0.997 (95% CI 0.995 to 0.998) for cancer detection in the internal test set and 0.991 (0.979 to 1.00) in the external validation set. The AUC for distinguishing between a low-grade (Gleason score 6 or ASAP) and high-grade (Gleason score 7–10) cancer diagnosis was 0.941 (0.905 to 0.977) and the AUC for detecting Gleason pattern 5 was 0.971 (0.943 to 0.998) in the external validation set. Cancer percentage calculated by pathologists and the algorithm showed good agreement (r=0.882, 95% CI 0.834 to 0.915; p<0.0001) with a mean bias of -4.14% (-6.36 to -1.91) Evaluation of 3 different scanners' performance to create prostate biopsy images with suitable quality for accurate INIFY Prostate predictions. Impact of preanalytical factors and user experience.





Renal Cell Carcinoma Whole-Slide Image Classification and Search Using Deep Learning

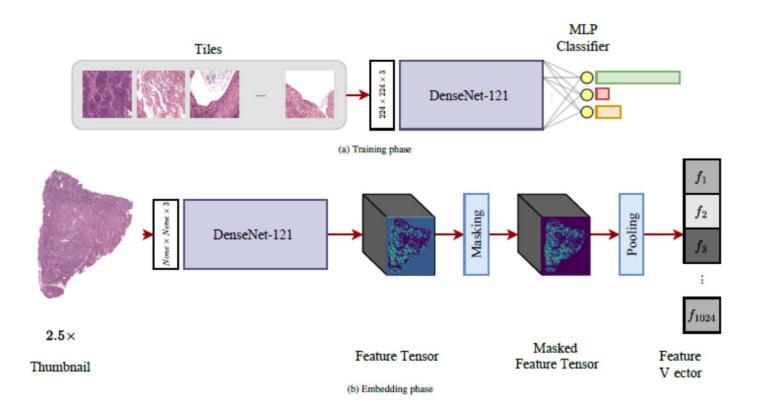
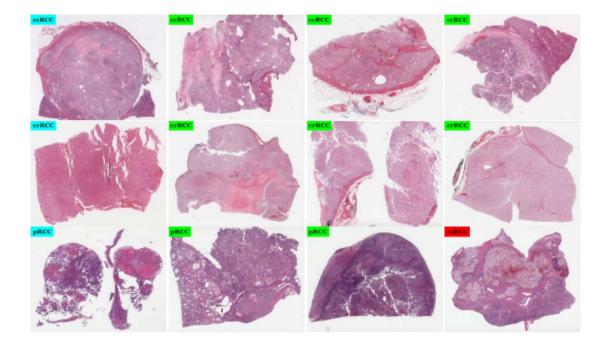
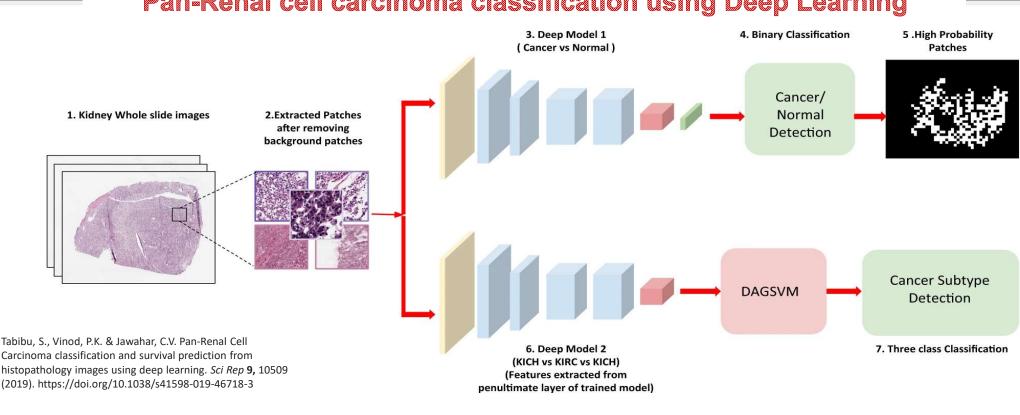


Figure 1: The outline of the proposed algorithm for encoding WSIs. (a) shows the training step in the algorithm. this In step, а DenseNet-121 [27] is trained using 224 × 224×3 tiles extracted from $2.5 \times$ WSIs in the training subsets to classify different RCC subtypes

Renal Cell Carcinoma Whole-Slide Image Classification and Search Using Deep Learning



The top three search results for queries related to different RCC subtypes from the TCGA search dataset. The images with a light blue tag are the query WSIs, while the correct and the wrong retrievals are shown in green and red tags, respectively.



Pan-Renal cell carcinoma classification using Deep Learning

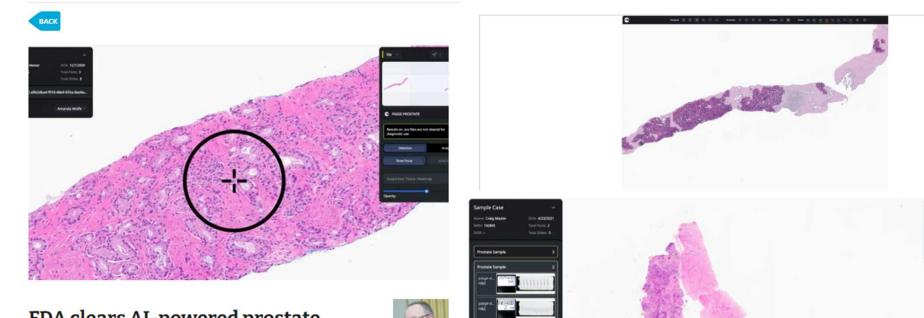
Classification Pipeline

Cancer Classification Pipeline. (1) Kidney Whole slide images. (2) 512*512 patches extracted from images with 50% overlap and background removed using pixel thresholding. (3) Patches from normal and cancerous slides fed to the deep network. (4) Patches classified as cancerous or non-cancerous. (5) High probability patches identified by the trained network and binary mask is applied. (6) The patches from three subtypes used to train a similar deep architecture for a three-way classification. (7) Features extracted from the penultimate layer of the network and fed to DAG-SVM and a three-way classification is performed by it.

119		

AI-BASED PATHOLOGY DIAGNOSTIC TOOLS ARE NOW AVAILABLE!! ge, a

commercial distribution agreement for Paige's comprehensive portfolio of diagnostic software solutions.



FDA clears AI-powered prostate cancer detection software

Phil Taylor

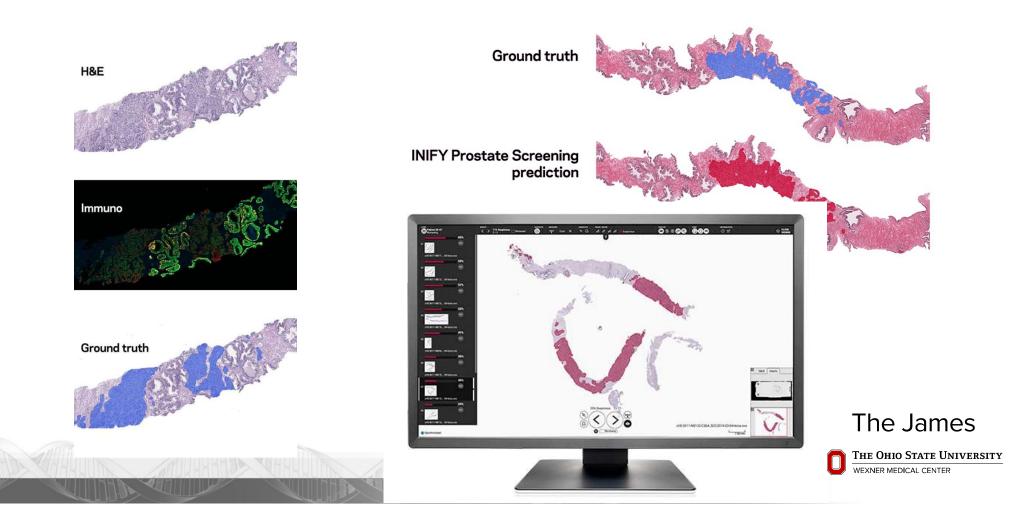
September 22, 2021

Software that can help pathologists detect prostate cancer from slides of biopsies more effectively has been approved by the FDA.

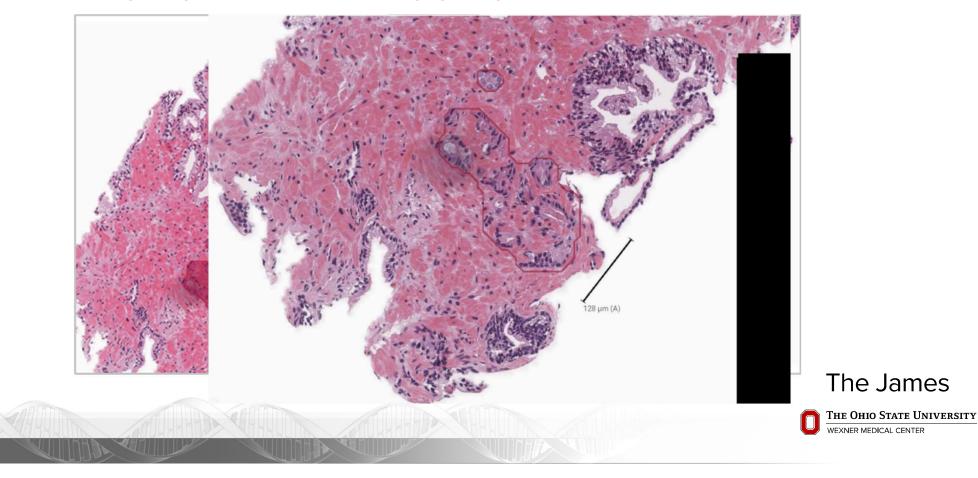
Boehringer backs AI-powered anti-infective startup ArrePath

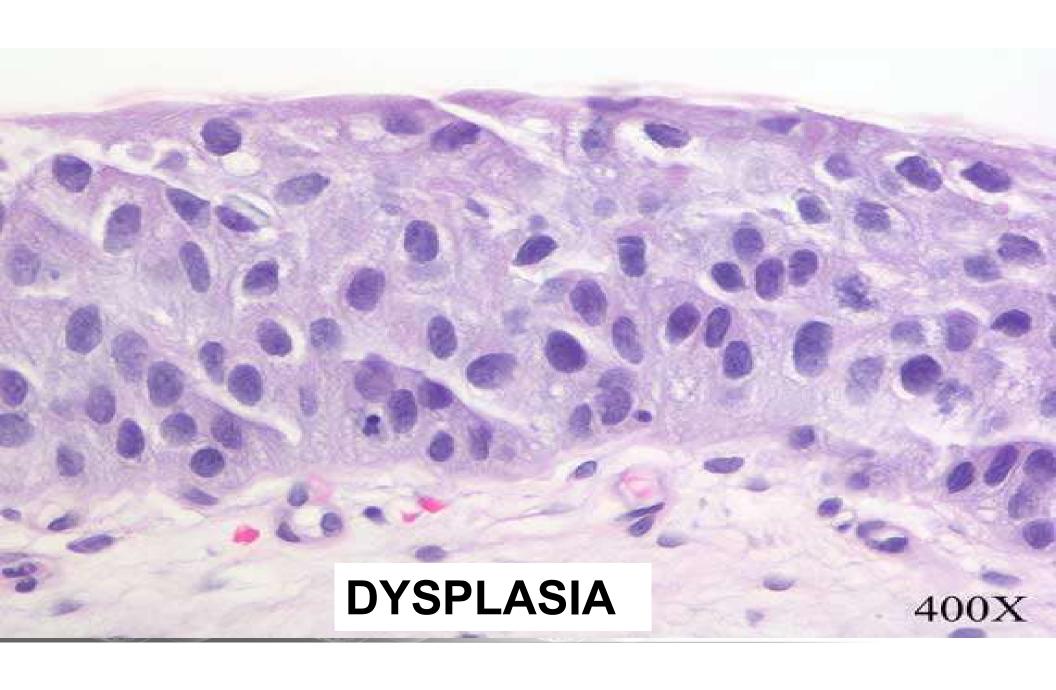


Al-powered tool that provides pathologists with valuable decision support when identifying cancer in prostate biopsies



Artificial intelligence (AI) algorithms can be used as a tool to augment diagnosis, grading and volume quantification of prostate cancer (PCa) in prostate biopsy (PBx)



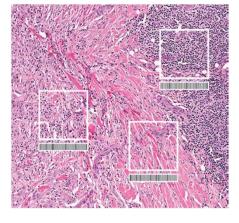


End-to-end system example: Bladder cancer diagnosis

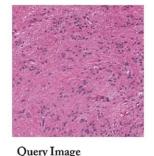


Searching WSIs: Image retrieval: Another Clinical Tool

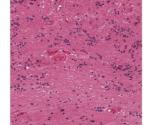








Brain - Subependymoma



Best Match Brain – Subependymoma

THE OHIO STATE UNIVERSITY

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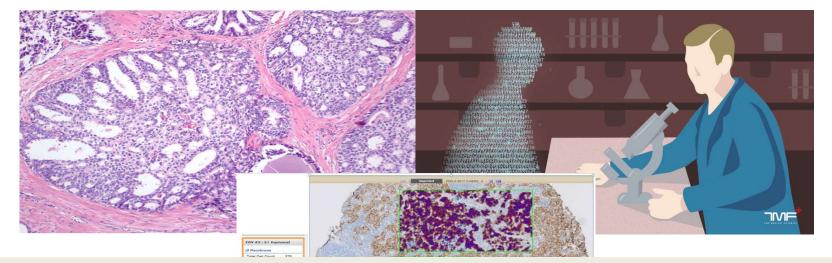
Scan Index/label

Search



Searching Is Intelligence. Image retrieval – the next revolution in pathology. Hamid Tizhoosh | 10/11/2018. The Pathologist The James

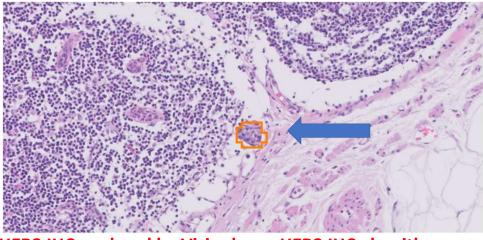
Digital Pathology and AI is here...NOW!



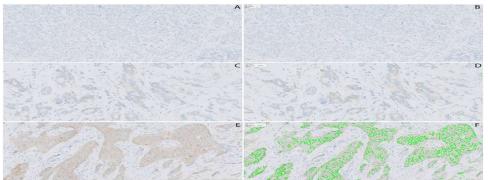
IMPACTING CLINICAL DIAGNOSTICS AND RESEARCH Research, Clinical, Education, Innovations

Examples of Quantitative Image Analysis and AI Algorithms Validated at OSU in the last 12 months

FINDING METASTATIC CANCER IN LYMPH NODES



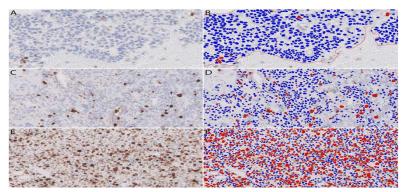
HER2 IHC analyzed by Visiopharm HER2 IHC algorithm



DETECTING AND GRADING PROSTATE CANCER

				Total 20
Cases			Enable live update 🛛 🍞 💠	Images *
Case Id 🖕	Date	Status	Suspicious	20%
Basis-11	2021-06-16	In Review 🕲	Read-Only	02 eHE-POS-16-0980_21-leica.svs
Basis-12 Basis-13		Reviewed (2) 🥏 Reviewed (2) 🥏		14%
Basis-14 Basis-15	2021-06-17	Reviewed (2) 🥥 Reviewed (2) 🥥		11%
Basis-16 Basis-17 Basis-18	2021-06-21 2021-06-21 2021-06-21	Reviewed (3) 🥥 Reviewed (2) 🥥 Reviewed (2) 🥥	0% 0% 0%	eHE-PDS-16-02 22-hama.ndpi
Basis-19 Basis-20	2021-06-21 2021-06-21	Reviewed (2) 🥏	0% 0%	6%

Digital imaging analysis of Ki-67 in cytology





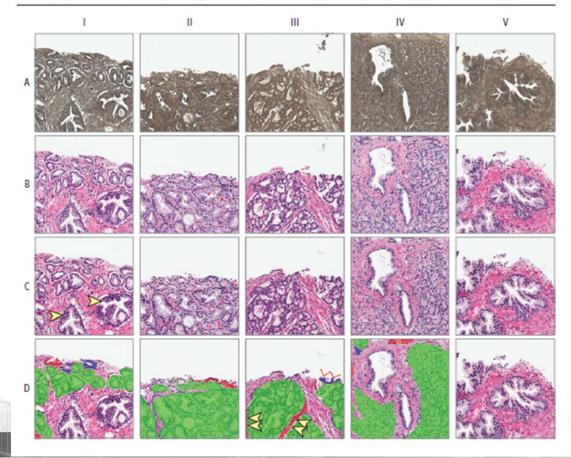
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Original Investigation | Health Informatics

Use of Deep Learning to Develop and Analyze Computational Hematoxylin and Eosin Staining of Prostate Core Biopsy Images for Tumor Diagnosis

Aman Rana, MS; Alarice Lowe, MD; Marie Lithgow, MD; Katharine Horback, MD; Tyler Janovitz, MD; Annacarolina Da Silva, MD; Harrison Tsai, MD; Vignesh Shanmugam, MD; Akram Bayat, PhD; Pratik Shah, PhD

> Figure 2. Representative Image Patches Generated by the Computational Staining Neural Network and Their Comparison With Corresponding Ground Truth Hematoxylin and Eosin (H&E) Dye-Stained Images



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Row A, Deparaffinized native nonstained image patches entered into the neural network. Row B, Ground truth H&E dye–stained patches. Row C, computationally H&E stained patches generated by the neural network. Arrows in C-I indicate the 2 benign glands, all other glands represent tumors. Row D, shows computationally H&E stained patches overlaid with colors indicating agreements and disagreements between physician annotations on these images compared with ground truth H&E dye–stained images. Variation in labeling detail by annotators (arrows) are shown in D-III. Green indicates true positive; blue, falsenegative; and red, false positive.



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CONVERGENCE AND TECHNOLOGIES | JANUARY 18 2022

Prostate Cancer Risk Stratification via Nondestructive 3D Pathology with Deep Learning–Assisted Gland Analysis 👌

Weisi Xie 💿 ; Nicholas P. Reder; Can Koyuncu 💿 ; Patrick Leo; Sarah Hawley; Hongyi Huang;

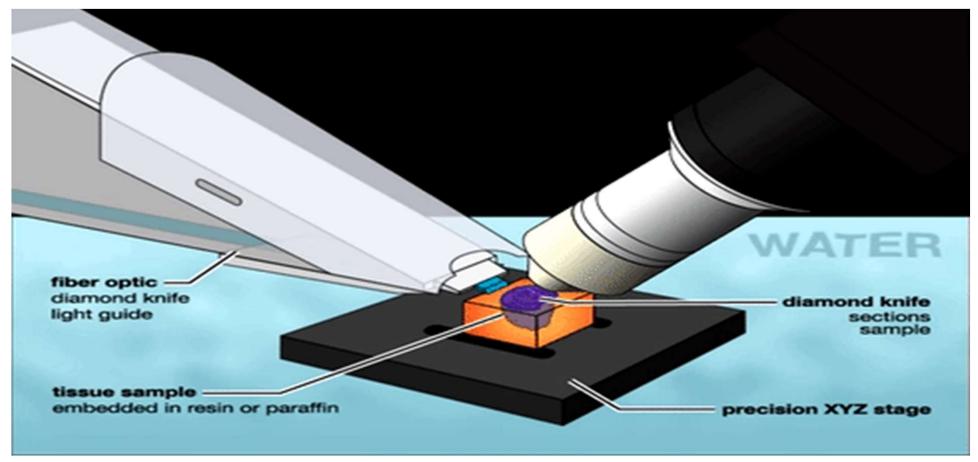
Chenyi Mao; Nadia Postupna 💿 ; Soyoung Kang; Robert Serafin; Gan Gao; Qinghua Han;



BEYOND WHOLE SLIDE IMAGING

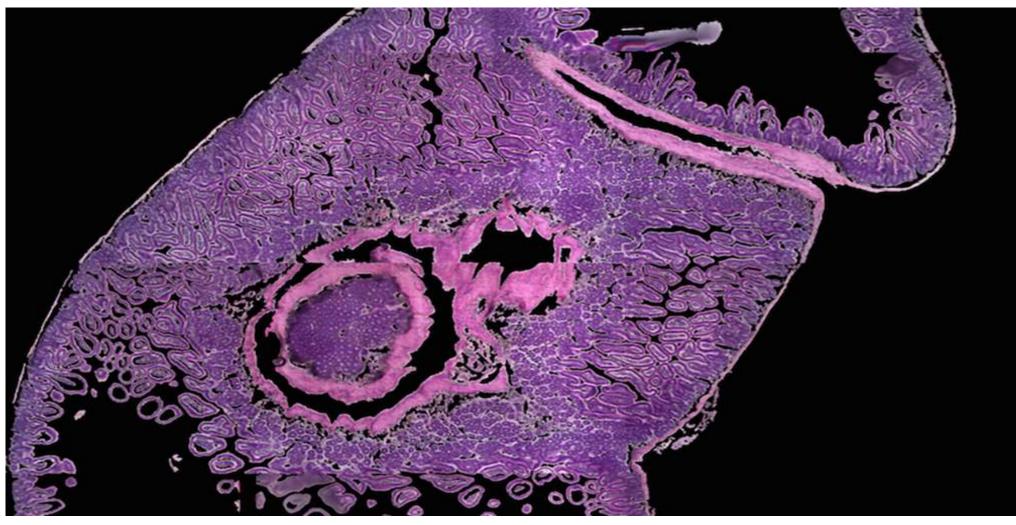


Knife Edge Scanning Microscope



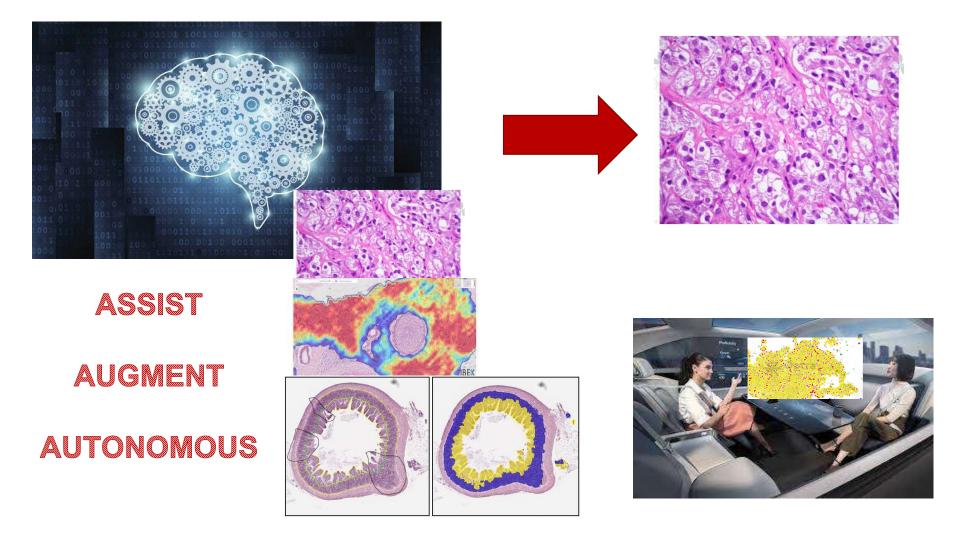
http://www.3scan.com/

Knife Edge Scanning Microscope



PUTTING IT ALL TOGETHER

HOW CAN AI TOOLS HELP A PATHOLOGIST?



DIGITAL WORKFLOW: WHAT WORKS WELL AND WHAT NEEDS TO BE FIXED? AND WHAT NEEDS TO BE FIXED?



- IMAGE QUALITY
- ANNOTATIONS
- MEASURING THINGS
- **IMAGE ACCESSIBILITY**
- SHARING IMAGES
- REVIEWING CASES
- CO-REGISTRATION
- COLLABORATION
- MANY NEWS TOOLS ON THE HORIZON THE OHIO STATE UNIVERSITY WEXNER MEDICAL CENTER

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WHAT WORKS OK!

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- ERGONOMICS
- INTEGATION WITH LIS NEEDS TO BE IMPROVED
- ANNOTATIONS TRANSFER TO REPORT
- MEASUREMENTS NO LINK TO LIS
- IMAGE ACCOUNTABILITY
- AUDIT SYSTEMS

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DIGITAL **WORKFLOW**: WHAT WORKS WELL AND WHAT NEEDS TO BE FIXED? WITH A MARKED AND WHAT WEEDS TO BE FIXED AND A MARKED AND

WHAT WE NEED TO FIX

- MANY IMAGE FORMATS
- CLOSED SYSTEMS BY VENDORS
- AI TOOLS STILL NOT WELL-INTEGRATED
- IMAGE WORKFLOW AND STORAGE –
 NOT STANDARDIZED
- COMPRESSION ISSUES
- MONITORS IT IS A WILD WILD WEST

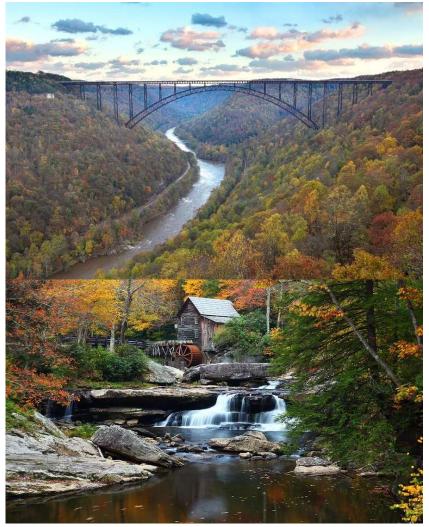
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Conclusions

- New tools such as digital pathology, image analysis and deep learning are here and transforming pathology and integration with LIS/EMR is vital to this.
- The pandemic served as a catalyst to pathologists adopting a digital workflow.
- Our journey into digital pathology/AI has allowed us to improve our pathology practice, workflows, quality of our diagnosis.
- The use of AI in pathology continues to evolve and will likely be use in mainstream pathology within this decade



Future Directions

- Cyto, Renal Pathology, Hematopathology digital signout
- Look at Compression algorithms to improve storage
- Continue to deploy more AI solutions for pathologists
- Work on integrating AI solutions within the LIS to perform analysis and improve result reporting
- Create easier de-identification of Images and data for helping with education and translational research
- Create additional revenue streams from digital pathology program
- CPT codes for image analysis- CLIA tests- implement 2022
- CPT codes for WSI. Implement 2023

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 Launch an online Masters Program in Digital and Computational Pathology in 2023









Thank You to Our **Amazing Digital Pathology** Team

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