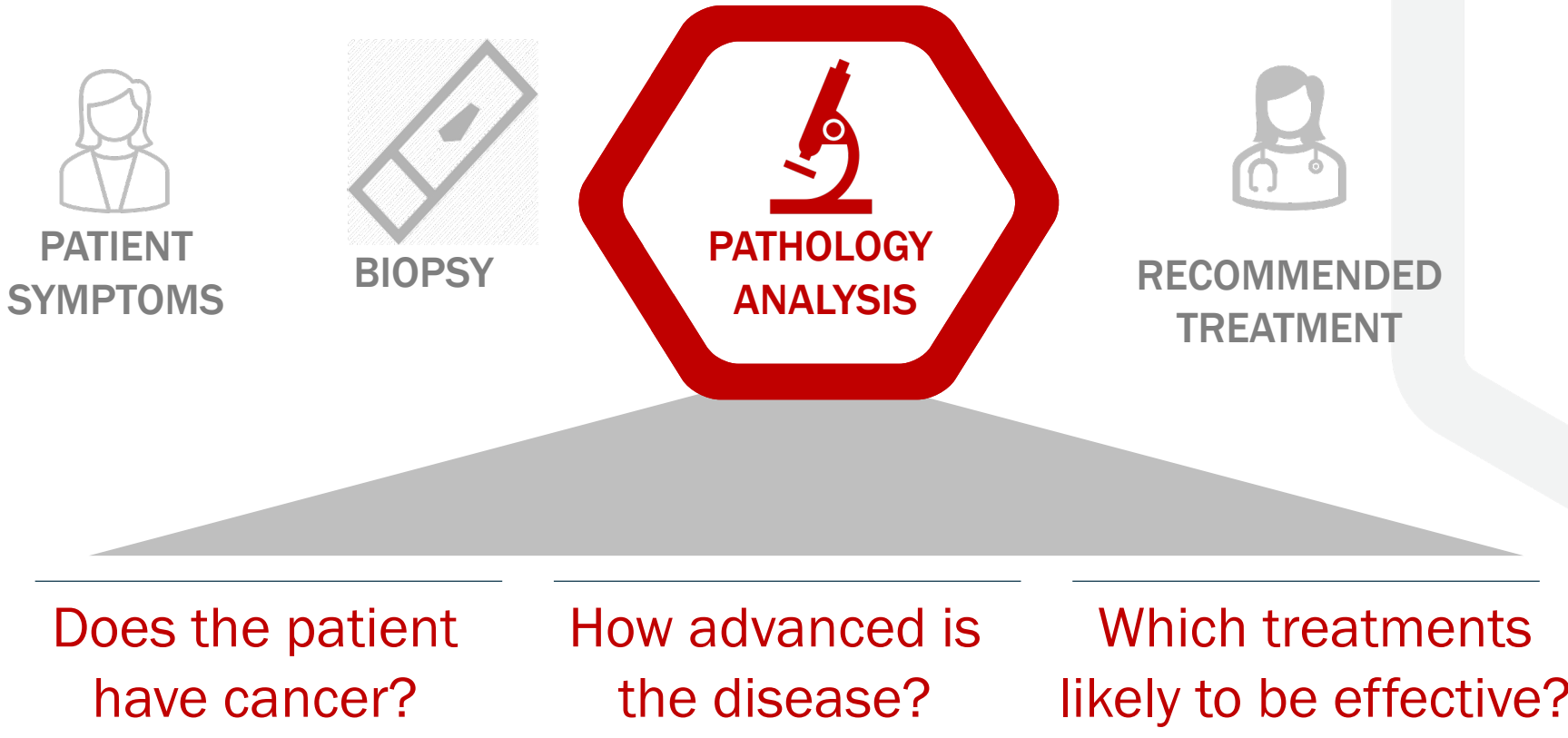


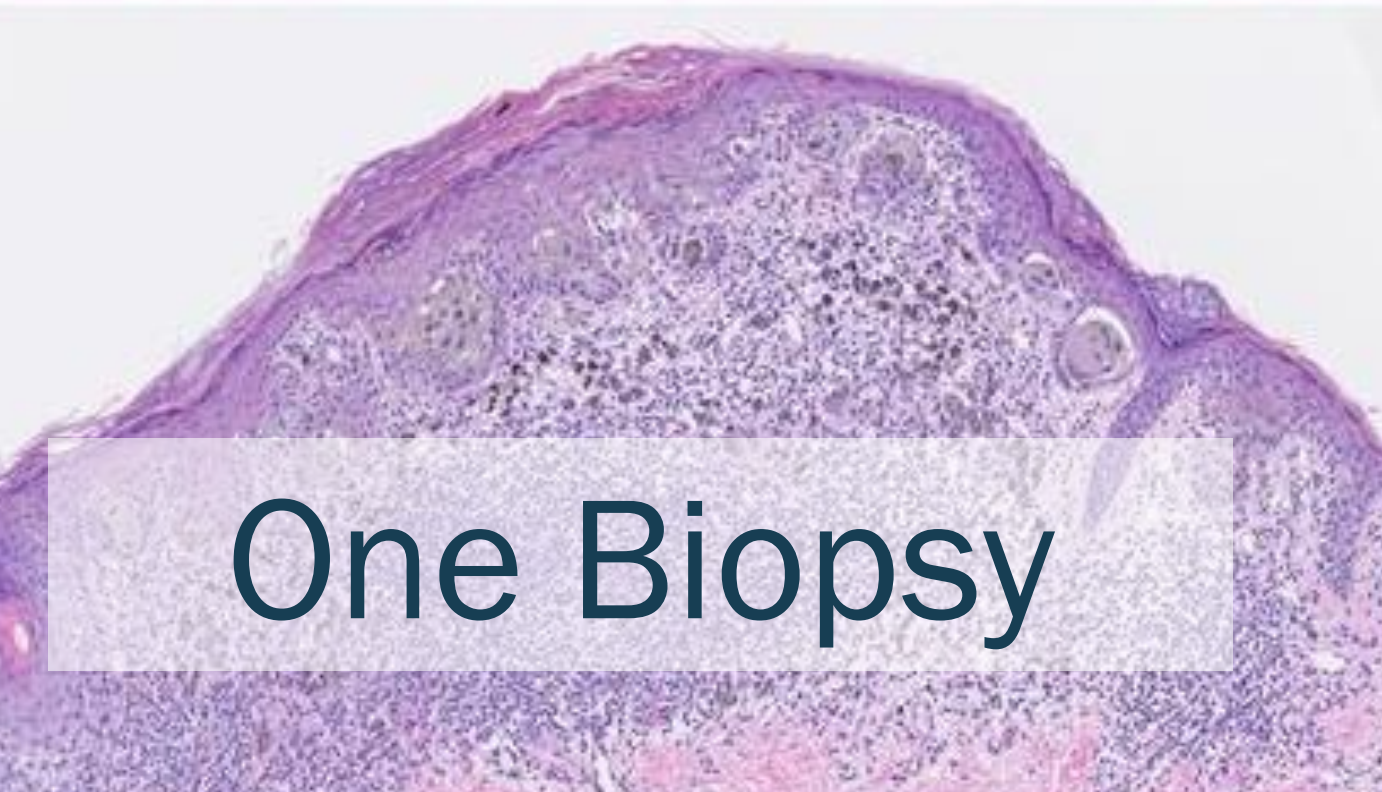
Novel Technologies and Strategies that Aim to Optimize Diagnosis via Pathology

Andrew H Beck MD, PhD
CEO at PathAI

Workshop on Achieving Excellence in Cancer Diagnosis

Pathology plays a fundamental role in the patient journey





One Biopsy



36 Pathologists

Joann Elmore et al. Pathologists' diagnosis of invasive melanoma and melanocytic proliferations: observer accuracy and reproducibility study. BMJ 2017;357:j2813

Diagnostic terms given

MPATH-Dx class I

Common nevus, junctional

Dysplastic nevus - mild

Halo nevus (1)

Atypical melanocytic neoplasm, junctional (suggested treatment)

MPATH-Dx class II

Spitz nevus (conventional), (junctional, compound, or intradermal)

Dysplastic nevus - moderate

Pigmented spindle cell nevus (junctional or compound)

Atypical nevus not otherwise specified, including atypical nevus

Atypical intraepithelial melanocytic proliferation (AIMP) (suggested treatment)

<5 mm margins (narrow but complete))

Atypical melanocytic neoplasm, junctional (suggested treatment)

(narrow but complete))

MPATH-Dx class III

Atypical/dysplastic Spitz lesion, (junctional, compound, or dermal)

Melanoma in situ, common/pagetoid/superficial spreading

Dysplastic nevus - severe

Atypical nevus not otherwise specified, including atypical nevus

Melanoma in situ, not otherwise specified

Atypical melanocytic neoplasm, junctional (suggested treatment)

5mm (but <1 cm) margins)

MPATH-Dx class IV

Invasive melanoma, superficial spreading melanoma

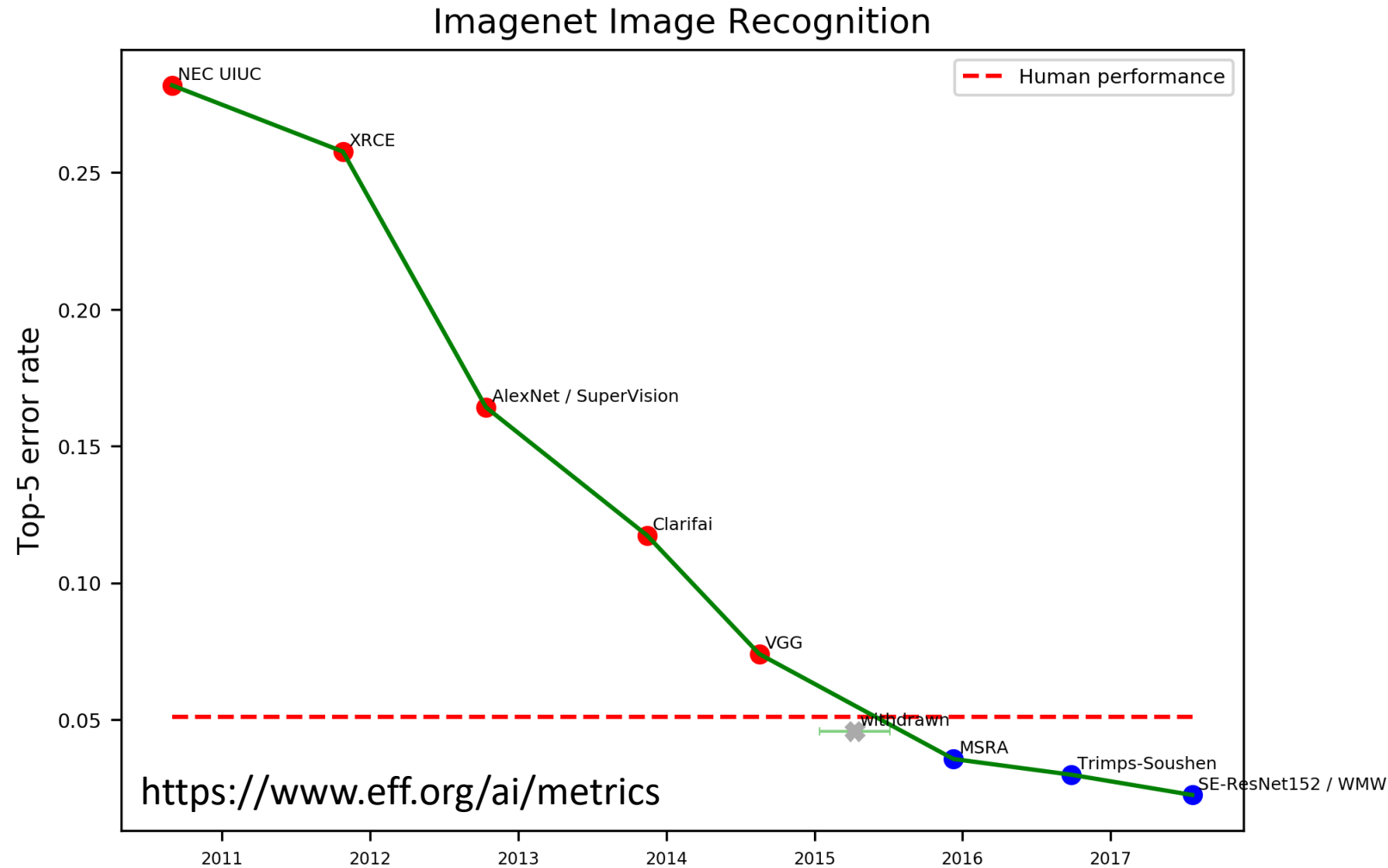
MPATH-Dx class V

Invasive melanoma, heavily pigmented melanoma

18

DIFFERENT
DIAGNOSES

Deep Learning for Computer Vision has shown Remarkable Progress in Recent Years



Alex Krizhevsky, Ilya Sutskever, Geoffrey E. Hinton. ImageNet classification with deep convolutional neural networks
<https://dl.acm.org/doi/10.1145/3065386>

The landscape of precision medicine testing is becoming increasingly complex

J.T. Jørgensen

Translational Oncology 14 (2021) 101063

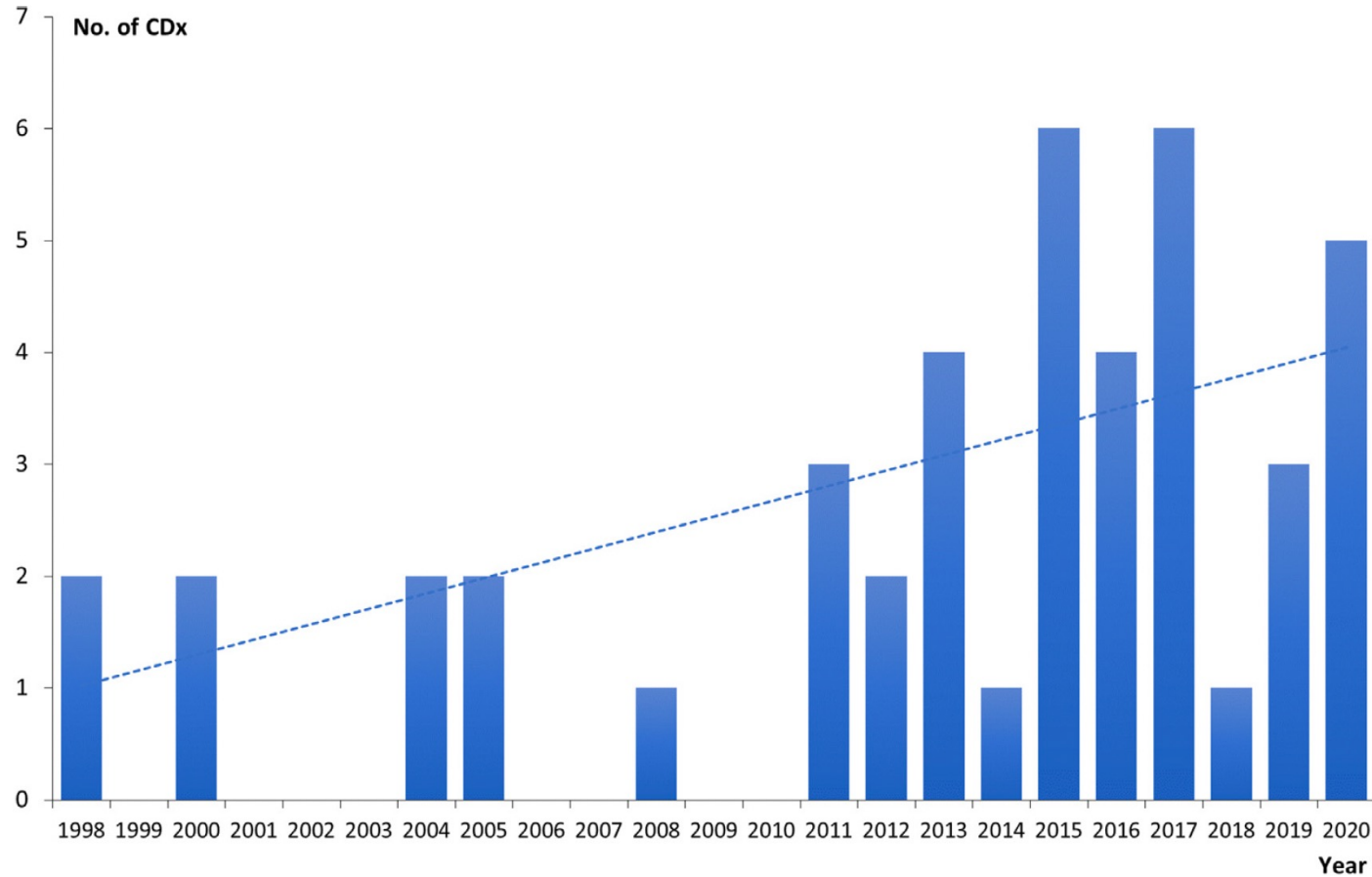
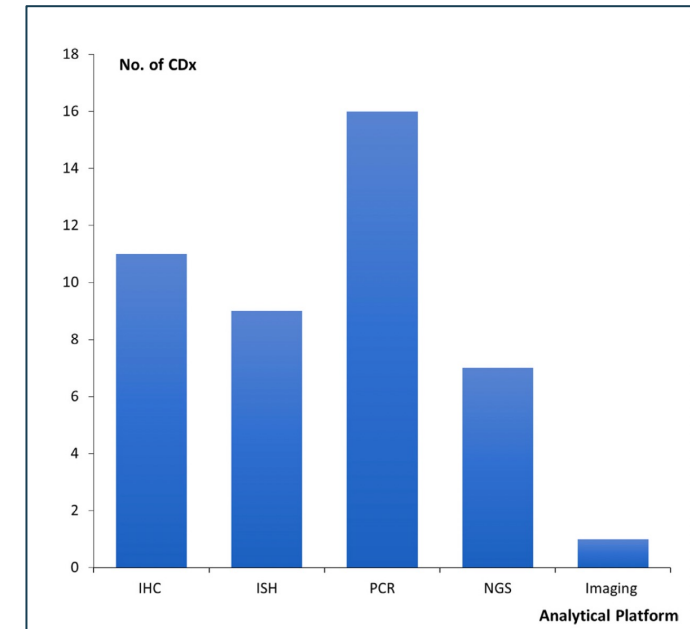


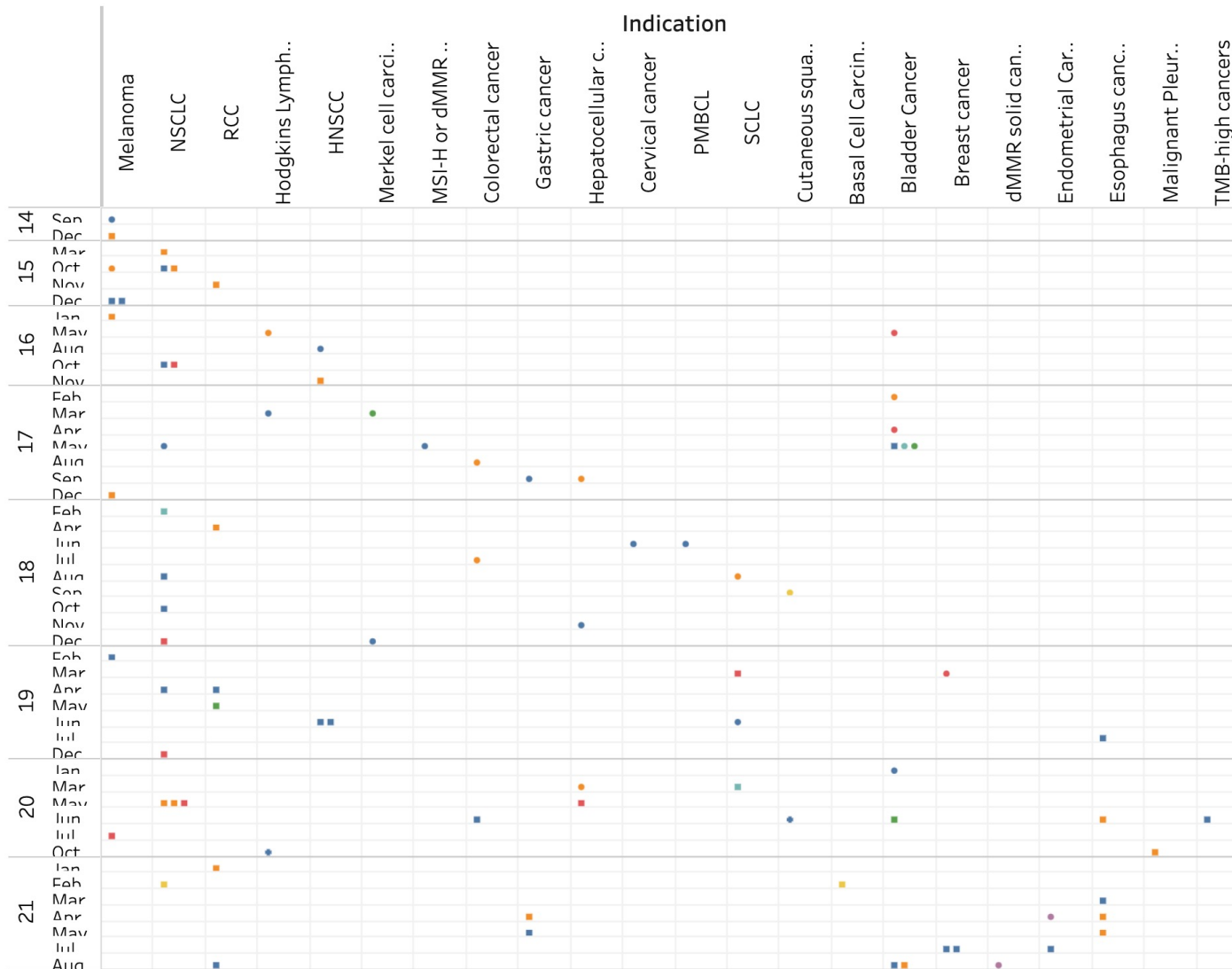
Fig. 1. FDA approval of companion diagnostic (CDx) assays by year. The total number of approvals by the end of 2020 are 44.



Timeline of Anti-PD-1/L1 Antibody Approvals by the FDA

Updated August 31, 2021

Sources: CRI, CRI Analytics, and FDA



Drug & Company

- Pembrolizumab, Merck & Co
- Nivolumab, Bristol Myers Squibb
- Atezolizumab, Roche
- Durvalumab, AstraZeneca
- Avelumab, EMD Serono
- Cemiplimab, Regeneron
- Dostarlimab, GlaxoSmithKline

<https://www.cancerresearch.org/en-us/scientists/immunology-landscape/pd-1-pd-l1-landscape#timelines>

Complex, Quantitative Criteria Required for Patient Selection

Quantitative PD-L1 IHC CDx includes:

- NSCLC - TPS 1%
- Bladder - IC 5%
- TNBC - IC 1%
- Bladder - CPS 10%
- Gastric - CPS 1%
- Cervical - CPS 1%
- HNSCC - CPS 1%
- Esoph. - CPS 10%

CDx includes immune cell PDL1+ quantification from IHC

Traditional manual pathology review of PDL1-IHC performs very poorly on immune cells

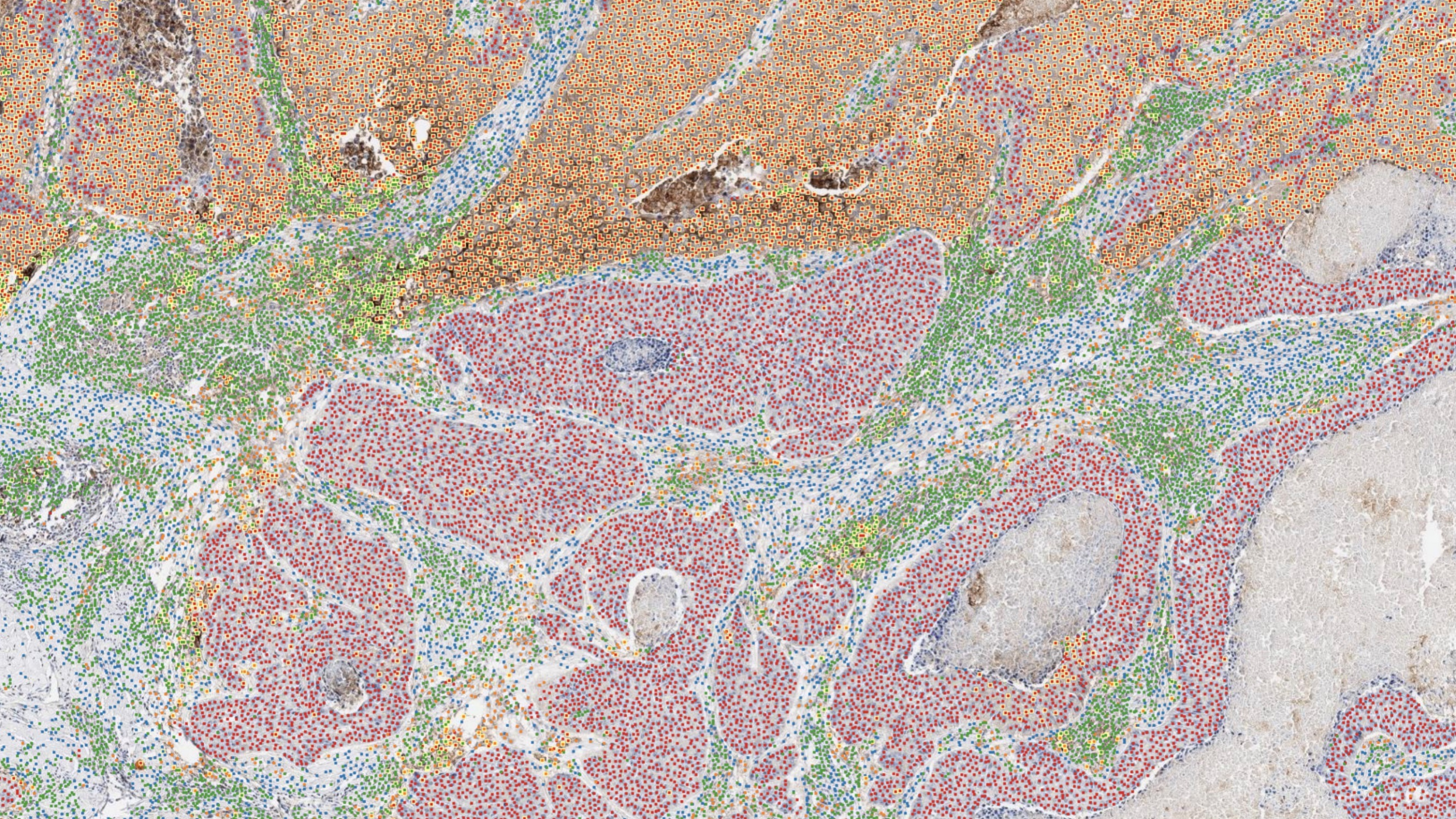
PDL1 manual IHC scores on immune cells are unreliable

Table 2. ICC for the Pathologist Scores and Concordance Statistics

Cells ^a	Antibody, ICC (95% CI)				Summary, Mean (SD)
	22c3	28-8	SP142	E1L3N	
Tumor cells	0.882 (0.873-0.891)	0.832 (0.820-0.844)	0.869 (0.859-0.879)	0.859 (0.849-0.869)	0.86 (0.02)
Immune cells	0.207 (0.190-0.226)	0.172 (0.156-0.189)	0.185 (0.169-0.203)	0.229 (0.211-0.248)	0.19 (0.03)

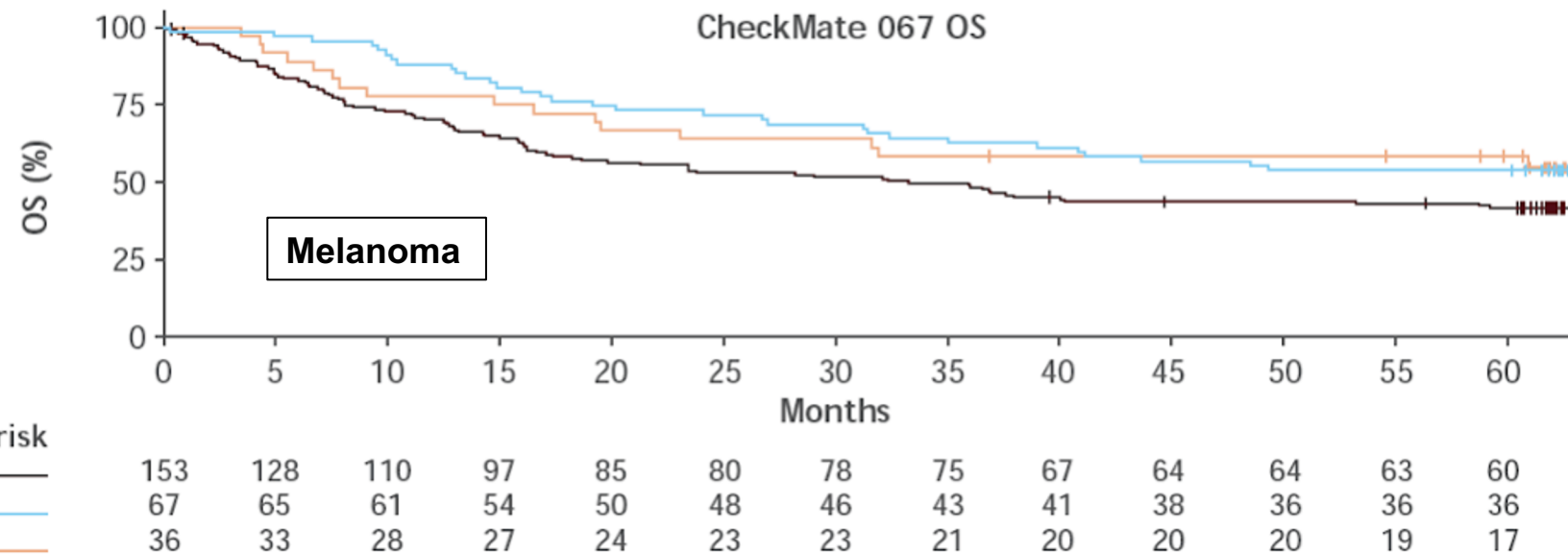
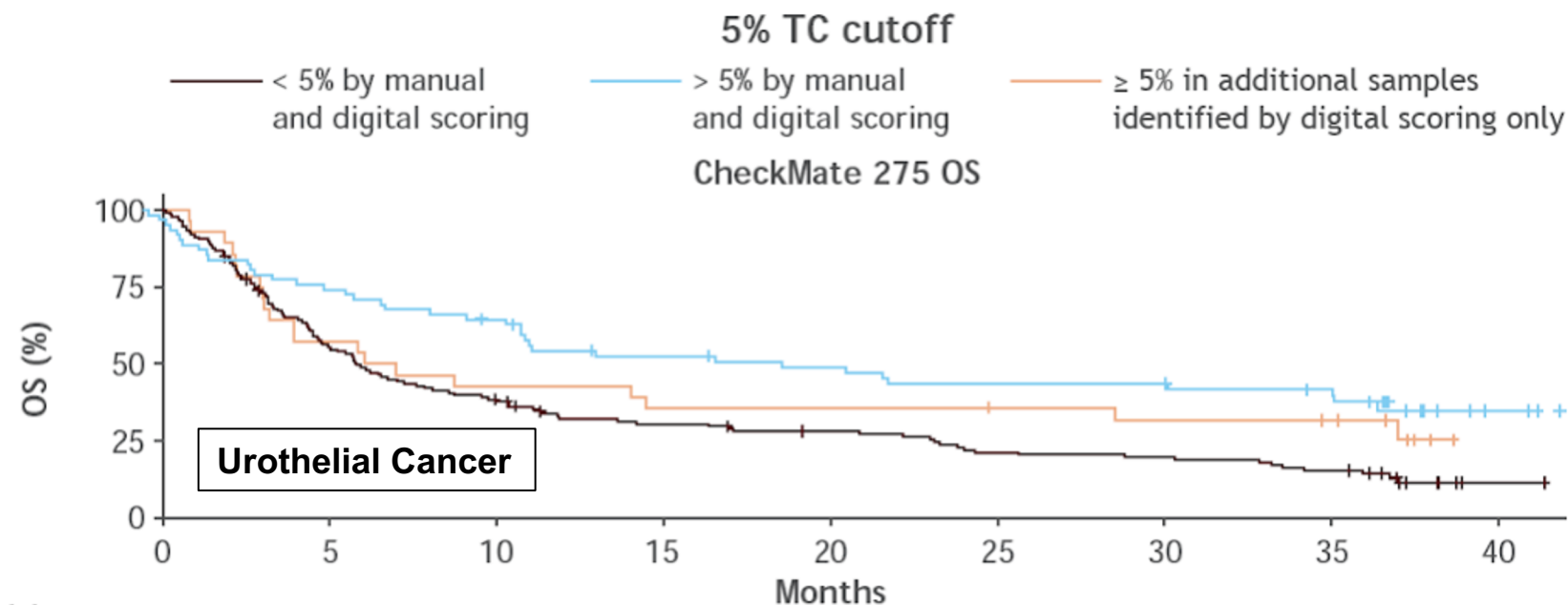
Abbreviation: ICC, intraclass correlation coefficient.

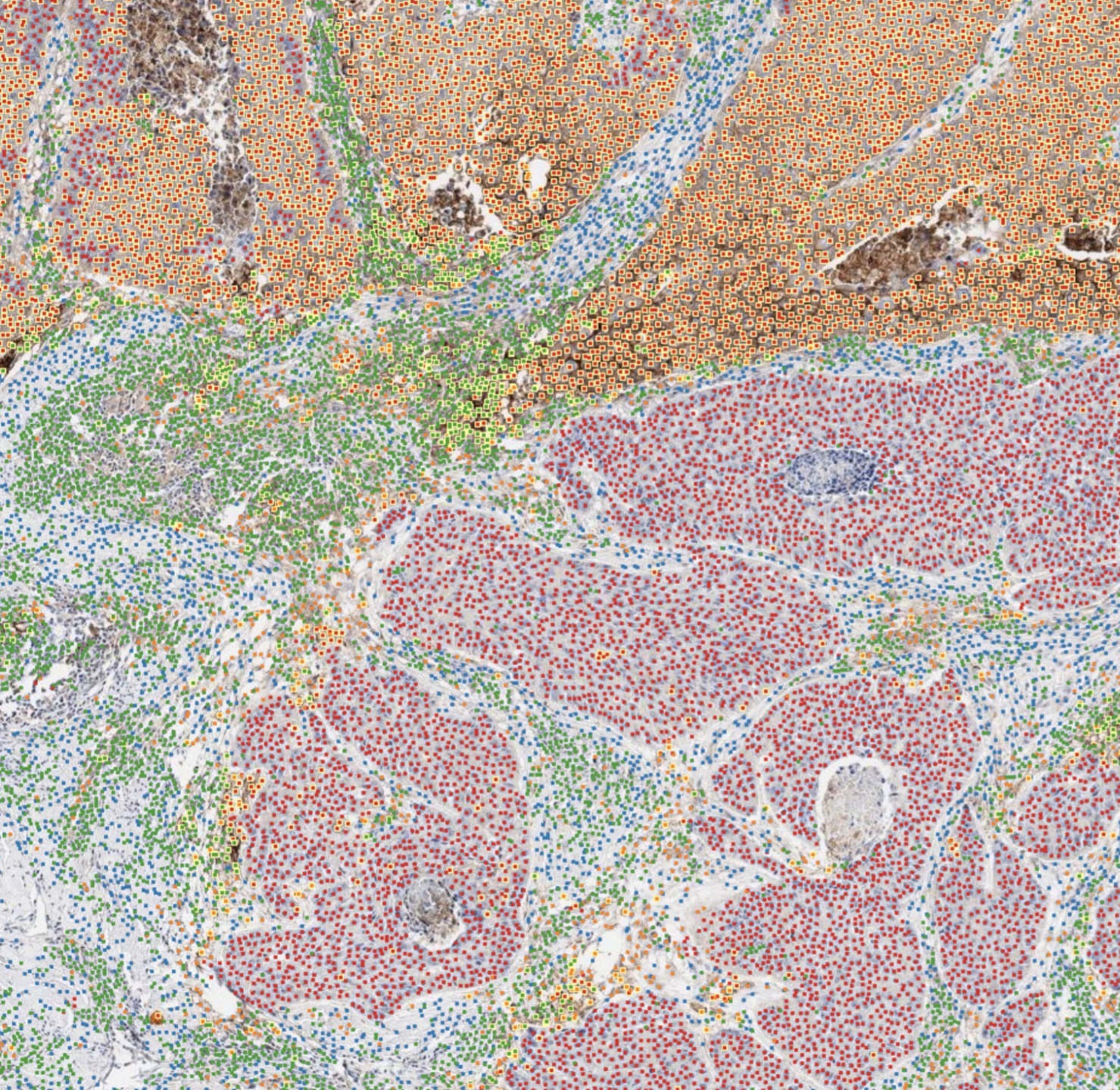
^a N = 90.



AI-powered PD-L1 scoring identifies manual-negative patients with long term survival

Duan et al. Presented at the American Association for Cancer Research (AACR), June 2020





Potential of AI-Driven Pathology:

More accurate diagnosis

More lifesaving drug approvals

Matching patients with the most effective treatments

Thank you

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