

Review

Advances in Digital Histopathology: A Comparative Review of Automated Tissue Analysis Tools and the Innovation Behind the Histopathology Image Analysis Platform

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Abstract:

The integration of computational methods into histopathology represents a paradigm shift in diagnostic accuracy and laboratory efficiency. Despite the proliferation of digital pathology tools, the field lacks integrated platforms that seamlessly combine morphometric analysis, H&E stain quantification, and standardized biomarker scoring in an accessible format. This review examines the capabilities and limitations of existing digital pathology solutions including QuPath, HALO, ImageJ, and CellProfiler and introduces the Histopathology Image Analysis Tool, a web-based platform engineered to address critical gaps in the current market. The tool integrates automated cell counting, stain deconvolution algorithms, nuclear-to-cytoplasmic ratio calculation, and standardized biomarker scoring (ER, PR, HER2, Ki-67) with built-in diagnostic interpretation modules. By synthesizing quantitative analysis with clinical rule-based interpretation, the platform accelerates tissue evaluation, reduces inter-observer variability, and enhances reporting consistency without requiring whole-slide imaging infrastructure. This review demonstrates that unified, accessible digital pathology platforms bridge the gap between high-cost commercial solutions and basic open-source tools, positioning them as essential assets for clinical diagnostics, translational research, and medical education.

Keywords: Digital pathology, histopathology image analysis, biomarker scoring, H&E stain deconvolution, automated diagnosis, computational pathology, immunohistochemistry quantification, diagnostic interpretation, precision medicine, healthcare informatics

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1. Introduction

Histopathology remains the gold standard for tissue diagnosis, yet the field faces mounting pressures that demand technological innovation. The exponential growth in cancer incidence with an estimated 20 million new cancer cases annually worldwide places extraordinary demands on

pathology services already stretched by increasing caseloads and workforce shortages (Aljehani et al., 2023). Traditional microscopic examination, while fundamental to diagnostics, suffers from inherent limitations: inter-observer variability in biomarker scoring, subjective interpretation of staining intensity patterns, and time-intensive

manual assessment of nuclear morphology (Mezei et al., 2024).

The diagnostic landscape has been further complicated by the expanding repertoire of biomarkers essential for therapeutic decision-making. Hormone receptor (ER/PR) status, human epidermal growth factor receptor 2 (HER2) expression, and proliferation indices (Ki-67) now constitute mandatory assessments in breast cancer pathology (Thakur et al., 2018). Similarly, emerging markers such as p53, e-cadherin, and tissue-specific keratins demand quantitative precision that manual scoring cannot reliably provide.

Digital pathology the application of computational and imaging technologies to pathological analysis has emerged as a transformative solution. However, the current ecosystem reveals a fragmented landscape. High-end artificial intelligence platforms like PathAI and Aiforia demand significant capital investment and technical infrastructure. Conversely, open-source tools such as QuPath and ImageJ, while free and flexible, require substantial programming expertise and lack integrated interpretation modules. The result is a critical market gap: a shortage of accessible, user-friendly platforms that unify image analysis with clinical decision support (Fatima et al., 2024).

This review evaluates the existing digital pathology ecosystem and introduces the Histopathology Image Analysis Tool a purpose-built platform designed to democratize quantitative histopathology. By comparing it against established competitors, we demonstrate how integrated design, standardized workflows, and accessible interfaces create measurable advantages in accuracy, reproducibility, and clinical utility.

2. The Digital Pathology Landscape: Growth and Challenges

The digital pathology market has experienced substantial growth over the past decade. Advances in slide scanning technology, computational processing power, and cloud infrastructure have reduced technical barriers to implementation. Yet paradoxically, adoption remains inconsistent

(Jahn et al., 2020). Several factors explain this paradox:

Market Barriers: Whole-slide imaging (WSI) systems cost \$150,000–\$300,000 per instrument, creating prohibitive entry costs for many laboratories, particularly in resource-limited settings. Additionally, WSI scanners generate enormous data files (500 MB–4 GB per slide), demanding robust infrastructure for storage and network transmission (Farahani et al., 2015).

Technical Standardization: Lack of interoperability between proprietary platforms has fragmented the ecosystem. Slides digitized on one vendor's scanner often cannot be seamlessly analyzed using another vendor's software, creating "vendor lock-in" and limiting institutional flexibility (Jahn et al., 2020).

Regulatory Uncertainty: While the FDA has cleared certain digital pathology systems for diagnostic use, most institutions remain cautious about relying exclusively on digital images for critical clinical decisions, particularly in challenging cases. This hesitancy reflects both technical concerns and medico-legal conservatism (Farahani et al., 2015).

Accessibility Gap: The most sophisticated AI-powered platforms (PathAI, Aiforia, Visiopharm) are expensive, require technical support, and often demand WSI infrastructure. Meanwhile, open-source tools (QuPath, CellProfiler, ImageJ) are free but demand steep learning curves, with limited built-in clinical interpretation frameworks. Within this landscape, a critical need has emerged: accessible, integrated platforms that require minimal infrastructure while providing clinically relevant outputs. The Histopathology Image Analysis Tool addresses this need directly.

3. Comparative Overview of Existing Digital Pathology Tools

3.1 QuPath

QuPath is an open-source digital pathology platform released in 2016 that has become the de facto standard for biomedical image analysis research (Gonzalez et al., 2025). **Strengths:** QuPath excels in customizability, supporting complex scripting workflows and integration with external tools like CytoMap for spatial analysis.

Its open-source nature eliminates licensing costs. Comparative studies demonstrate strong correlations with manual quantification ($r > 0.89$) for immune cell phenotyping and tau pathology assessment (Gonzalez et al., 2025). **Limitations:** QuPath's power comes at a usability cost. It requires scripting knowledge (Groovy language) for most advanced analyses, making it inaccessible to non-technical clinicians. Built-in biomarker scoring modules are limited, and interpretation guidance is minimal. Many users report lengthy setup times before analyses can commence (Gonzalez et al., 2025).

3.2 HALO (Indica Labs)

HALO is a proprietary commercial platform optimized for high-throughput automated analysis. **Strengths:** HALO offers user-friendly interfaces and strong performance in specific applications. Studies comparing tau pathology quantification found HALO showed strong concordance with manual object density measurements ($\rho \geq 0.70$) in some tissue regions (Gonzalez et al., 2025). It supports multiplex immunofluorescence analysis and provides reasonable integrations with standard pathology workflows. **Limitations:** HALO's high cost (often >\$50,000 annually) limits accessibility. Software stability issues have been reported, with users noting frequent crashes, particularly when using AI-driven tissue detection modules. Optical density measurements show inconsistent performance across different tissue types (Gonzalez et al., 2025). Furthermore, HALO offers limited flexibility for customization and integration with external analytical tools.

3.3 ImageJ/Fiji

ImageJ remains ubiquitous in biomedical imaging laboratories, with the Fiji distribution providing pre-configured plugins. **Strengths:** ImageJ is freely available, highly modular, and supported by an enormous user community. Countless plugins enable diverse analytical functions. **Limitations:** Biomarker scoring is limited and requires manual

plugin configuration. H&E analysis capabilities are basic and lack standardized interpretation frameworks. Most critically, ImageJ demands significant user expertise; automated workflows must be constructed through macro programming. It is poorly suited to non-expert users or clinical laboratory environments.

3.4 CellProfiler

CellProfiler is an open-source software designed specifically for high-throughput cellular image analysis. **Strengths:** It provides powerful, modular pipelines for quantifying cellular morphology, particularly nuclear segmentation and object tracking. **Limitations:** CellProfiler requires pipeline design and parameter optimization before use. It offers no integrated biomarker interpretation modules and minimal clinical guidance. Its steep learning curve and lack of graphical workflow design make it impractical for routine clinical deployment.

3.5 Aiforia and PathAI

These commercial AI platforms represent the frontier of computational pathology. **Strengths:** Both employ deep learning models for tissue classification, tumor detection, and prognosis prediction. They integrate WSI support and cloud-based workflows, enabling enterprise-scale deployment. **Limitations:** These platforms cost \$100,000+ annually. They require substantial technical infrastructure and vendor relationships. Most critically, they are black-box systems; pathologists and researchers cannot easily understand or modify underlying algorithms, raising reproducibility and validation concerns (Xu et al., 2024). These high-end solutions are poorly suited to resource-constrained settings or institutions seeking transparent, interpretable analyses.

4. Comparative Evaluation Framework

To fairly evaluate digital pathology tools, a structured framework is essential. The following criteria are proposed:

| Criterion | Description | Clinical Relevance |
|----------------------------|--|--|
| Image Preprocessing | Automated stain normalization, artifact removal, tissue segmentation | Essential for handling batch effects and variable staining |

| | | |
|---------------------------------------|--|--|
| Nuclear Segmentation Accuracy | Reliability of nucleus detection and quantification | Critical for cell counting, density analysis, and proliferation assessment |
| H&E Stain Deconvolution | Separation of hematoxylin (nuclear) and eosin (cytoplasmic) channels | Fundamental to tissue morphology interpretation and quality control |
| Biomarker Scoring Automation | Support for ER, PR, HER2, Ki-67, p53, and other clinically relevant markers | Directly impacts diagnostic decision-making and therapy planning |
| Standardized Output Metrics | Allred score, H-score, N/C ratio, percentage positivity | Enables comparison across laboratories and clinical implementation |
| Built-in Interpretation Module | Diagnostic rules and clinical guidance integration | Bridges gap between quantitative data and clinical actionability |
| User Interface Accessibility | Ease of use for non-experts; minimal training requirements | Determines adoption likelihood in clinical and educational settings |
| Analysis Speed | Time from image upload to final report | Critical for high-throughput laboratory workflows |
| Export Capabilities | Support for standard formats (PDF, Excel, CSV); customizable reports | Essential for documentation and regulatory compliance |
| Infrastructure Requirements | Does platform require WSI scanners, cloud services, or advanced IT infrastructure? | Determines affordability and accessibility in diverse settings |
| Cost | Licensing, infrastructure, training, ongoing support | Fundamental barrier to adoption, particularly in resource-limited regions |

This framework ensures fair, multidimensional comparison.

5. Comparative Analysis: Existing Tools vs. the Histopathology Image Analysis Tool

Comparative Feature Matrix

| Feature | QuPath | HALO | ImageJ | CellProfiler | Aiforia | Our Tool |
|------------------------------------|---------------------|------|---------|--------------|---------|-----------------------|
| H&E Stain Deconvolution | ✓ (manual setup) | ✓ | ✗ | ✗ | ✓ | ✓ Automated |
| Nuclear Segmentation | ✓ | ✓ | Limited | ✓ | ✓ | ✓ Optimized |

| | | | | | | |
|--|-----------------------|----------|-----------|-----------|-----------|--|
| Biomarker Scoring (ER/PR/HER2 /Ki-67) | ✓ (manual) | ✓ | ✗ | ✗ | ✓ | ✓ Automated + Interpretation |
| Allred Score Calculation | ✗ | ✓ | ✗ | ✗ | Limited | ✓ Built-in |
| H-Score Calculation | ✗ | ✓ | ✗ | ✗ | Limited | ✓ Built-in |
| N/C Ratio Quantification | ✗ | Limited | ✗ | ✗ | ✗ | ✓ Automated |
| Diagnostic Interpretation Module | ✗ | Limited | ✗ | ✗ | ✓ | ✓ Comprehensive |
| Distance Measurement | ✓ | ✓ | ✓ | Limited | ✗ | ✓ User-friendly |
| Ease of Use (non-experts) | Poor | Medium | Poor | Very Poor | Good | Excellent |
| User Interface Quality | Moderate | Good | Basic | Poor | Excellent | Excellent |
| Training Requirement | Extensive (scripting) | Moderate | Extensive | Extensive | Minimal | Minimal |
| WSI Requirement | Optional | No | No | No | Preferred | No |
| Native WSI Support | Limited | Yes | Yes | No | Yes | Planned |
| Customizability | Excellent | Moderate | Excellent | Excellent | Low | Medium |
| Integration with External Tools | Excellent | Poor | Excellent | Excellent | Poor | Good |
| Automated Report Generation | ✗ | ✓ | ✗ | ✗ | ✓ | ✓ |

| | | | | | | |
|---|----------|-------------------|---------|----------|-----------------------|-------------------|
| Downloadable Reports (PDF/Excel) | X | ✓ | X | X | ✓ | ✓ |
| Cost | Free | >\$50,000/year | Free | Free | >\$100,000/year | Affordable |
| Infrastructure Needed | Moderate | Substantial | Minimal | Moderate | Substantial (Cloud) | Minimal |
| Regulatory Validation | Limited | Yes (CAP-cleared) | Limited | Limited | FDA-cleared (limited) | In development |

6. Unique Differentiators of the Histopathology Image Analysis Tool

The tool's design philosophy prioritizes **integrated functionality, clinical relevance, and accessibility** three dimensions where existing platforms fall short.

6.1 Unified H&E and Biomarker Analysis

Most digital pathology tools specialize in either morphometric analysis OR biomarker scoring, but not both in a single workflow. The Histopathology Image Analysis Tool integrates both dimensions:

- A single image upload initiates parallel analysis pathways for tissue morphology and biomarker expression
- Results display side-by-side, enabling pathologists to correlate morphologic features (nuclear density, tissue composition) with biomarker intensity
- This unified approach mirrors clinical decision-making, where morphology and biomarker status are inseparable

6.2 Automated Scoring with Clinical Interpretation

Unlike tools that output raw data, the platform includes integrated diagnostic interpretation:

- **Allred Score calculation** (0–8 scale) for ER/PR scoring, critical in breast cancer therapy selection
- **H-Score generation** for quantitative IHC assessment
- **Built-in clinical rules** (e.g., ER/PR positivity thresholds per ASCO guidelines) translate raw intensities into actionable clinical categories

- **Ki-67 categorization** (low/intermediate/high proliferation) based on established cutoffs

This bridge between data and diagnosis is rare among competitors and highly valued in clinical settings.

6.3 Lightweight, Accessible Design

The tool requires only standard JPEG/PNG image uploads no WSI scanners, no institutional IT infrastructure, no cloud service mandates:

- Works on standard microscope image captures or cropped regions of larger slides
- Operates within web browsers, eliminating software installation barriers
- Requires <5 minutes of training for basic functionality
- Suitable for clinical laboratories, research institutions, and educational settings in diverse resource contexts

6.4 User-Centric Interface

Designed for **clinicians and researchers without programming expertise**:

- Drag-and-drop image upload with visual feedback
- Modular tool selection (Count Cells, Analyze H&E, Analyze Biomarkers) accessible via simple buttons
- Real-time visualization of detected nuclei, stain distribution, and biomarker intensity regions
- Output panels display results in standardized formats immediately usable for reporting

6.5 Structured, Downloadable Outputs

Clinical and research use demands documentation:

- **H&E Analysis Reports** include staining percentages, N/C ratio, tissue architecture summary
- **Biomarker Scoring Reports** provide intensity distributions, H-score, Allred score, and clinical interpretation
- **Cell Count Data** with nuclear density metrics exportable for research analysis
- **Annotated Images** highlight detected nuclei or staining intensity regions for verification
- All outputs download as PDF or Excel for integration into laboratory information systems (LIS)

6.6 Modular Expansion

The platform's architecture supports incremental feature addition:

- Current modules: Cell Count, H&E Analysis, Biomarker Scoring (ER/PR/HER2/Ki-67/p53/E-cadherin)
- Planned: Distance measurement, multi-tissue comparison, immune cell phenotyping, malignancy prediction scoring
- This modularity contrasts with monolithic competitors and enables tailored deployments

7. Strengths and Gaps in Current Market Solutions

7.1 Where Existing Tools Excel

QuPath: Unmatched customizability and research flexibility; strong community support; validated performance in diverse tissue types and staining modalities (Gonzalez et al., 2025).

HALO: Robust performance for established analyses; strong WSI support; good integration with pathology workflows in well-resourced institutions; CAP validation (Gonzalez et al., 2025).

Aiforia/PathAI: Sophisticated AI models for tissue classification and prognosis prediction; enterprise-scale deployment capabilities; cutting-edge research performance (Xu et al., 2024).

7.2 Persistent Gaps

Universal Gap: No existing tool seamlessly bridges affordability, accessibility, diagnostic interpretation, and integrated functionality. QuPath demands expertise; HALO and commercial platforms demand capital; open-source tools lack clinical guidance.

Interpretation Gap: High-end AI platforms (Aiforia, PathAI) employ black-box deep learning, limiting interpretability and reproducibility critical concerns in regulated clinical environments (Xu et al., 2024).

Accessibility Gap: Educational institutions and resource-limited laboratories lack practical tools; QuPath and CellProfiler remain too complex, while commercial platforms remain prohibitively expensive (Madoori et al., 2023).

Standardization Gap: Biomarker scoring remains subjective across platforms. While Allred and H-scores are established standards (Thakur et al., 2018), most tools implement these inconsistently or require manual parameter tuning, reintroducing variability (Chaurasia et al., 2025). The Histopathology Image Analysis Tool directly addresses each gap.

8. Clinical and Research Applications: Practical Validation

8.1 Illustrative Case: Automated Breast Tissue Assessment

To demonstrate practical value, a representative H&E-stained breast tissue image was analyzed using the platform. The tissue exhibited dense cellularity typical of proliferative lesions. Platform analysis identified 107 nuclei with a computed nuclear density of 2,780.67 cells/mm², consistent with high-proliferation zones. H&E deconvolution quantified hematoxylin coverage (27.80%) and eosin distribution (23.81%), yielding a nuclear-to-cytoplasmic ratio of 0.27 suggestive of increased nuclear prominence associated with proliferative disease.

Subsequent biomarker analysis of an IHC-stained slide from the same tissue determined 99.5% total positive staining, H-score of 169.1, and Allred score of 7, consistent with hormone receptor positivity in breast cancer. These metrics directly informed therapeutic recommendations.

Time efficiency: Analysis and report generation required 3 minutes substantially less than manual scoring requiring 15–30 minutes per case in routine pathology practice (Madoori et al., 2023).

8.2 Clinical Value Proposition

For pathology laboratories, the platform offers:

- **Accuracy:** Reproducible quantification minimizes inter-observer variability, particularly critical for ER/PR scoring where small differences influence therapy selection
- **Efficiency:** Rapid analysis accelerates turnaround times, reducing diagnostic delays
- **Documentation:** Automatically generated reports create audit trails essential for quality assurance and regulatory compliance
- **Research Support:** Exportable quantitative data enables biomarker-driven research and retrospective cohort studies

8.3 Educational Utility

Medical and pathology trainees benefit from:

- **Structured Learning:** Standardized workflows teach fundamental principles of stain interpretation and biomarker assessment
- **Feedback Mechanism:** Quantitative outputs provide objective benchmarks for trainee performance
- **Case Libraries:** Accumulated cases with standardized outputs create learning repositories

9. Discussion: Market Positioning and Clinical Translation

The digital pathology market continues bifurcating into high-end enterprise solutions and basic open-source tools. The Histopathology Image Analysis Tool occupies a strategically important middle ground combining clinical utility, diagnostic interpretation, accessibility, and affordability.

9.1 Why Unified Platforms Matter

Integration of H&E analysis with biomarker scoring represents more than mere convenience. Clinically, morphology and biomarker expression are inseparable: ER/PR status must be interpreted

in context of tissue architecture and proliferation rate. Technologically, unified platforms reduce redundancy and potential error accumulation across sequential tools. From workflow efficiency, integration dramatically accelerates analysis. The market gap exists precisely because achieving this integration requires balancing complexity (to support sophisticated analysis) with accessibility (to serve non-expert users). Most existing tools optimize for one at the expense of the other.

9.2 Accessibility as Innovation

In global health contexts, digital pathology typically means high-cost WSI infrastructure requiring capital investment, IT support, and personnel training barriers unsurmountable in most low- and middle-income countries. Yet computational pathology need not mandate WSI. Standard microscope image capture, already ubiquitous in pathology laboratories globally, suffices for most quantitative analyses. By designing for standard imagery rather than WSI, the platform democratizes computational pathology, enabling resource-limited laboratories to adopt quantitative analysis without infrastructure transformation.

9.3 Standardization as Clinical Requirement

Biomarker scoring variability remains a significant source of diagnostic discordance (Thakur et al., 2018). Allred and H-scores are well-established standards, yet implementation varies across laboratories and platforms, reintroducing the very variability standardization aims to eliminate. The platform addresses this by embedding standardized scoring algorithms with validated cutoffs directly into the user interface. When every pathology laboratory uses identical scoring logic, inter-laboratory concordance improves, benefiting patients through consistent, reproducible diagnosis.

9.4 Clinical Adoption Pathway

Successful medical technology adoption depends on multiple factors: clinical validation, regulatory approval, workflow integration, and cost-benefit demonstration. The platform is well-positioned:

- **Clinical validation** is underway through prospective comparative studies against manual pathology scoring

- **Regulatory pathway** involves CAP/CLIA alignment and potential FDA clearance for specific applications
- **Workflow integration** occurs through LIS connectivity and standardized report formats
- **Cost-benefit** is compelling: minimal infrastructure investment, rapid analysis, reduced labor, improved consistency

10. Future Directions in Computational Histopathology

10.1 Planned Platform Enhancements

Whole-slide imaging support: Integration of gigapixel WSI analysis while maintaining the platform's lightweight accessibility. This requires advances in image tiling strategies and memory-efficient computation.

Predictive modeling: Machine learning integration to predict clinical outcomes (recurrence risk, therapy response) from histopathologic features, advancing toward precision medicine applications (Xu et al., 2024).

Multiplex immunofluorescence: Extension to simultaneously quantify multiple biomarkers (5–7 concurrent stains) with spatial analysis of immune infiltration.

Automated malignancy scoring: Deep learning models to assess tissue-level malignancy likelihood, providing a quantitative measure to supplement pathologist interpretation.

Integration with clinical data systems: Direct connectivity to laboratory information systems (LIS) and electronic health records (EHR) for seamless result reporting and quality tracking.

10.2 Broader Ecosystem Evolution

The field will likely converge toward **transparent, interpretable AI systems** that provide explainable outputs rather than black-box predictions. This reflects growing regulatory and clinical demands for algorithmic transparency (FDA guidance on AI/ML medical devices).

Cloud-edge hybrid models will likely emerge, enabling privacy-preserving analysis (on-premise computation) while leveraging cloud resources for model updates and collaborative review.

Multi-modal integration will expand, combining histopathology with genomic data, clinical

outcomes, and imaging, enabling integrative precision medicine workflows.

11. Conclusion

Digital pathology has transitioned from experimental novelty to essential clinical tool. Yet fragmentation persists: high-end AI platforms serve well-resourced institutions; open-source tools serve sophisticated researchers; most practicing pathologists and pathology trainees lack accessible, integrated solutions that combine quantitative analysis with clinical interpretation.

The Histopathology Image Analysis Tool addresses this critical market need by integrating H&E morphometry and biomarker quantification within an intuitive, clinically guided interface requiring minimal infrastructure and training. Comparative analysis demonstrates that existing solutions excel in specific dimensions (QuPath's customizability, HALO's validation, Aiforia's AI capabilities) yet fall short in providing integrated, accessible, clinically relevant workflows.

By strategically positioning between high-cost commercial platforms and complex open-source tools, the platform enables:

- **Clinical laboratories** to accelerate diagnostics and reduce reporting variability
- **Research institutions** to conduct large-scale biomarker studies with standardized quantification
- **Educational programs** to teach quantitative histopathology with structured workflows
- **Resource-limited settings** to adopt computational pathology without prohibitive infrastructure investment

The convergence of increasing diagnostic complexity, workforce constraints, and technological maturity creates unique opportunity. Platforms that successfully balance clinical utility, interpretability, accessibility, and affordability will define the next generation of digital pathology. The Histopathology Image Analysis Tool exemplifies this convergence, positioning computational histopathology not as specialized research tool but as essential infrastructure for modern diagnostic medicine.

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