

Individualized treatment risk stratification using histopathology-based genomics prediction in multiple myeloma: A multicenter study in 1429 participants

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INTRODUCTION

- Multiple myeloma (MM) genomic profiling by NGS/FISH is time-consuming and resource-intensive, limiting universal access to molecular-guided treatment decisions.
- ISS/R-ISS^{1,2,3} enables accessible but population-level risk stratification, while individualized models like IRMMa⁴ provide superior patient-specific accuracy through comprehensive genomic profiling but remain inaccessible in resource-limited settings.
- Histopathology could democratize precision risk assessment by extracting molecular information from routine H&E bone marrow biopsies.

AIMS

To develop **CORAL-histomorph**, a novel, next-generation machine learning deep whole slide attention-based⁵ model that predicts genomic abnormalities directly from routine H&E histopathology bone marrow biopsy images and integrates clinical/treatment data to provide individualized treatment guidance and predict clinical outcomes in MM patients.

METHODS

- Cohort:** 1,309 MM patients and 120 controls from University of Miami (UM; n=519), HealthTree Foundation (HT; n=805), and Ministry of Health Malaysia (MOHM; n=105).
- Processing:** H&E bone marrow slides scanned at 40x (Aperio AT2), deep learning trained to predict seven genomic subgroups: t(11;14), t(4;14), 1q gain, del 17p/13q/1p, hyperdiploidy.
- Validation:** **CORAL-histomorph** trained on UM, validated against FISH (HT, AUC), tested for OS prognostic performance (MOHM cohort, c-index, >3-year follow-up).
- Analysis:** Unsupervised clustering identified histomolecular and treatment-responsive clusters (TRC) from histopathology features.

RESULTS

Figure 1: CORAL-histomorph establishes histopathology-based molecular classification bridging tissue architecture with genomic complexity.

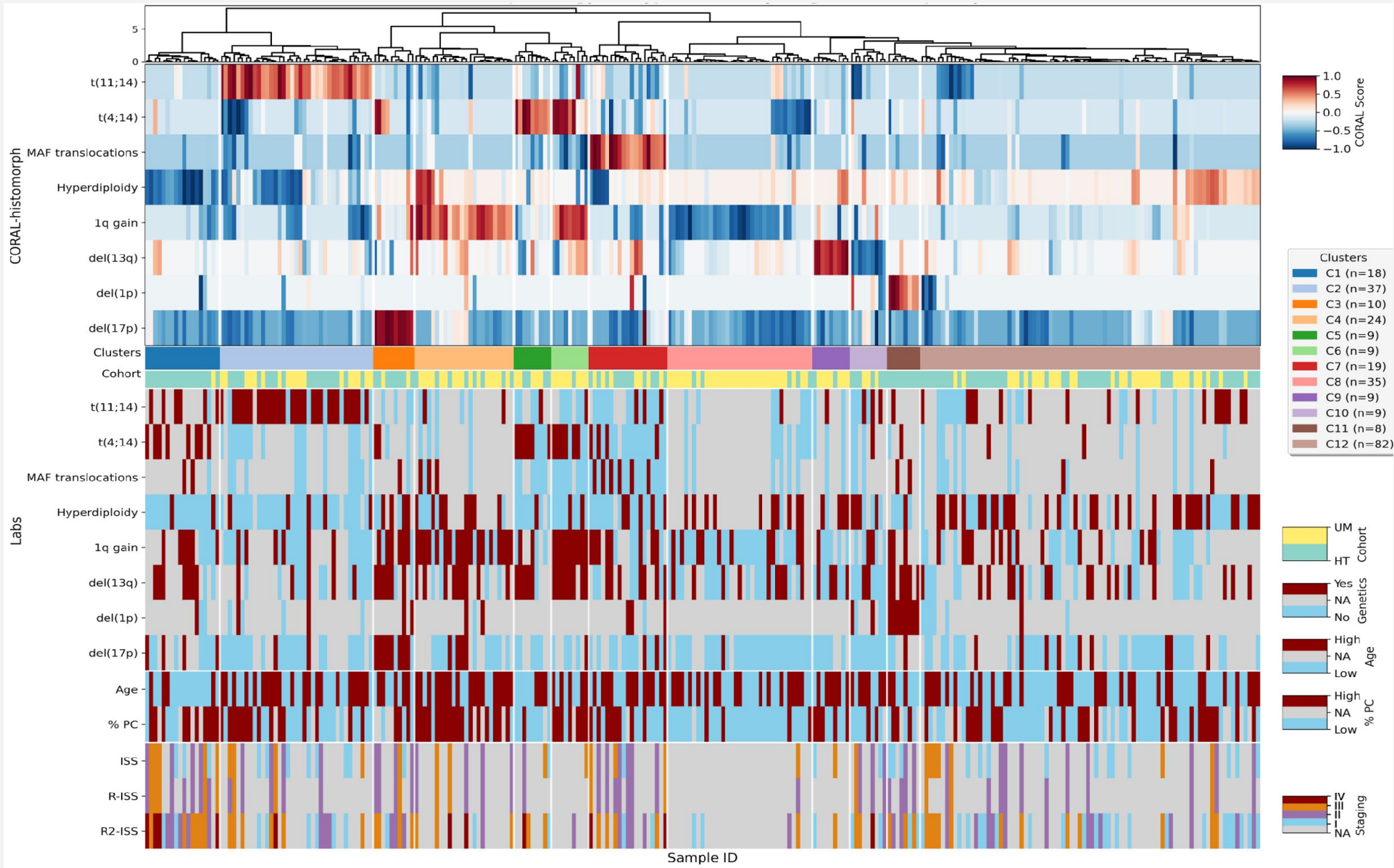


Figure 2: The histopathology image features provide risk stratification with significant p-values based on double hits and IMWG³ criteria

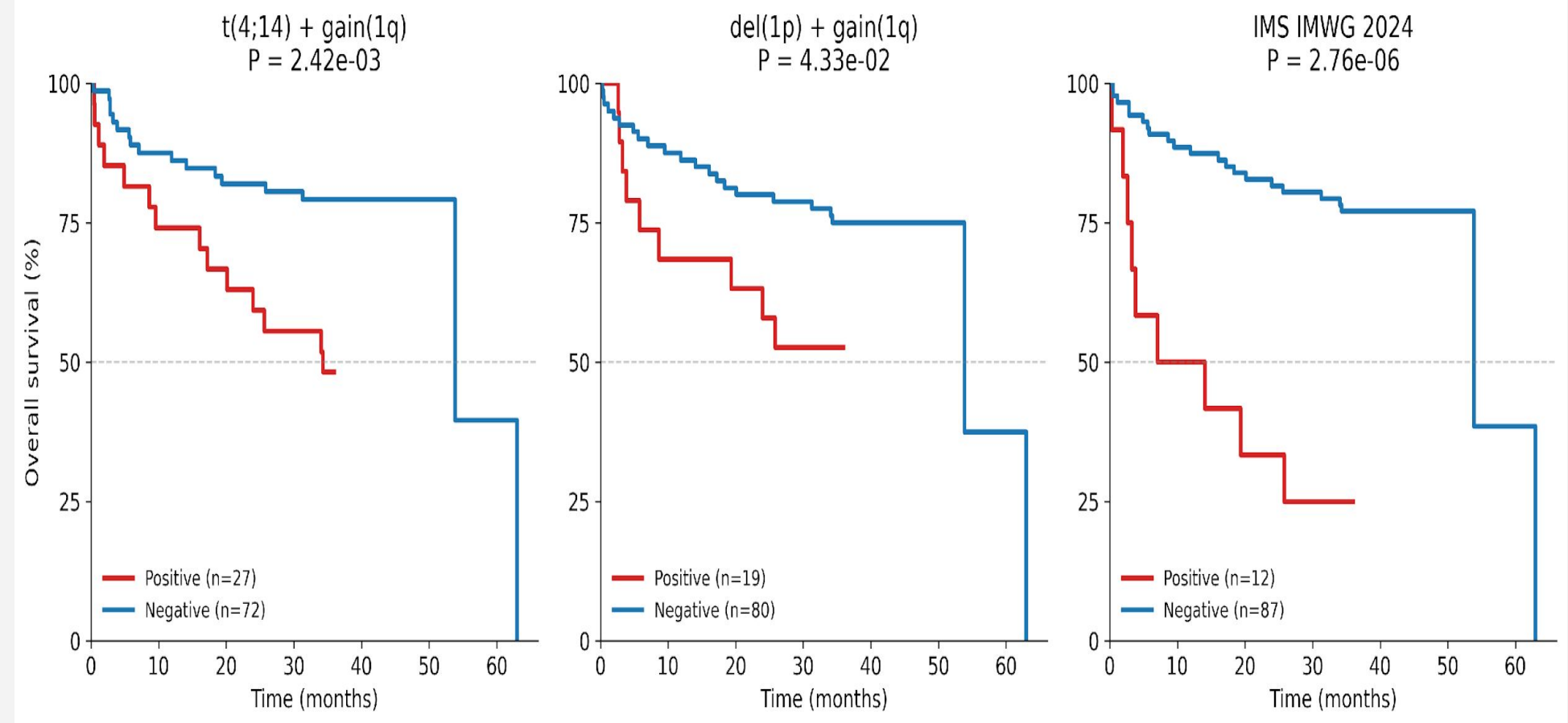


Figure 3: TRCs showed significant survival differences (OS p<0.01, PFS p<0.05). Non-ASCT patients (TRC1) demonstrated superior OS benefit (47.5 mo vs 32.2 mo) despite ASCT patients (TRC2) showing better PFS outcomes (28.9 mo vs 26 mo). **High-risk clusters** (TRC4/TRC5) showed poor responses to intensification with rapid progression.

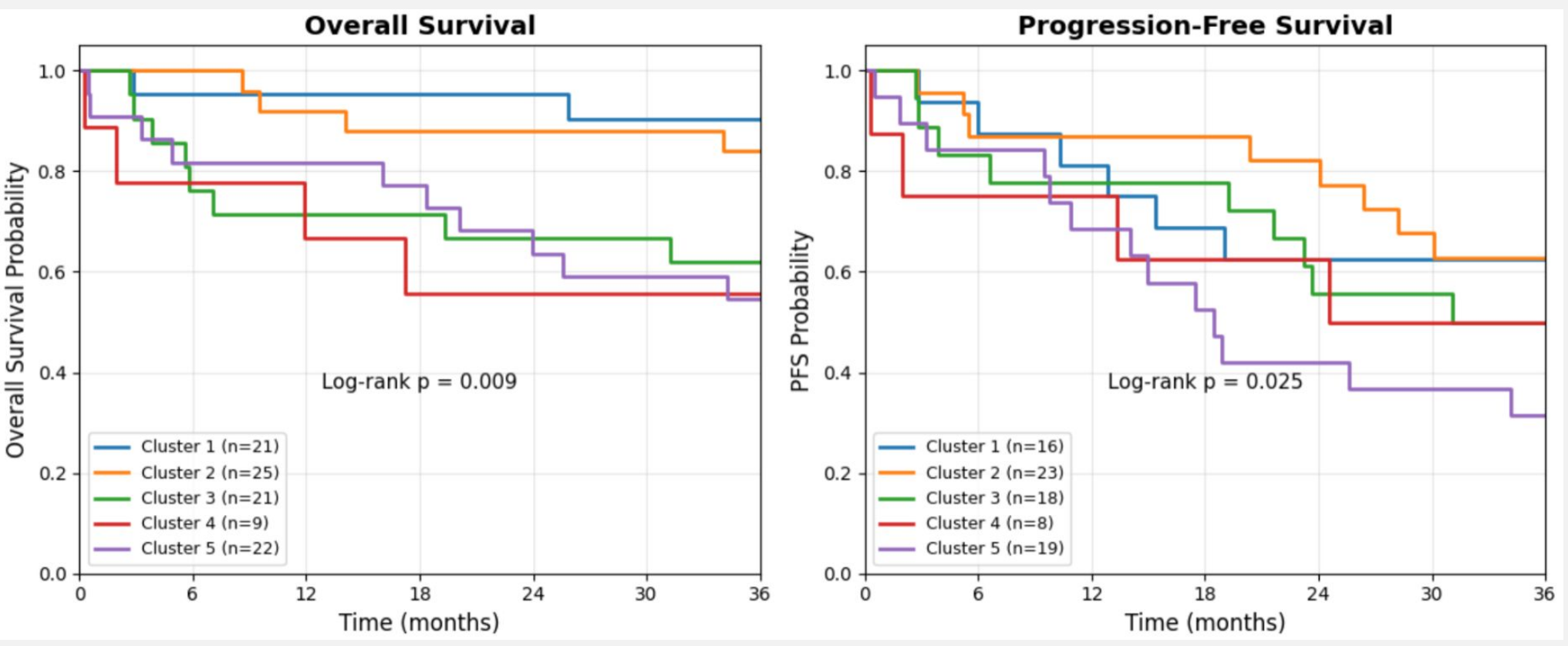


Figure 4: CORAL-histomorph deep attention maps of predicted and FISH test as t(11;14) positive.

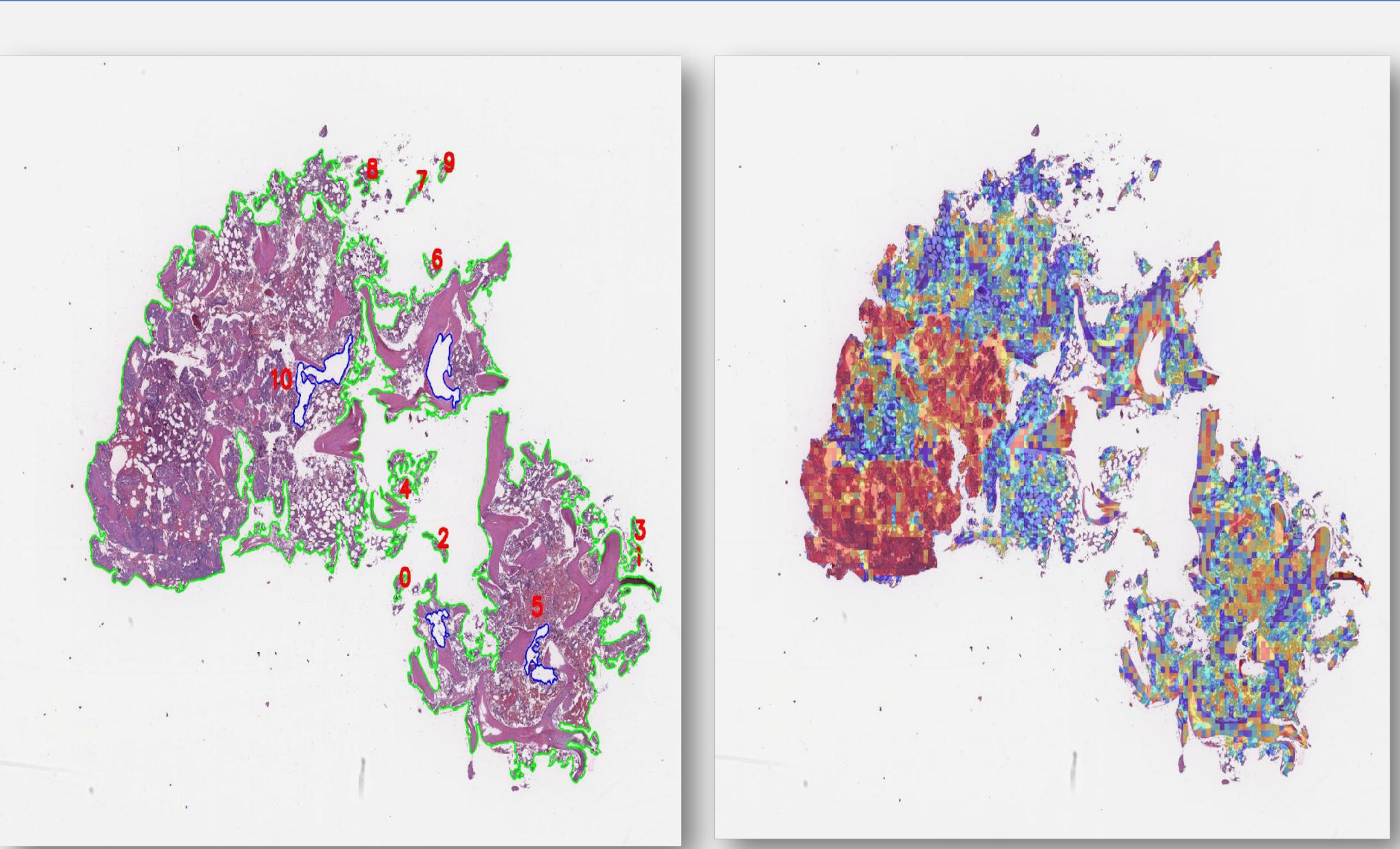
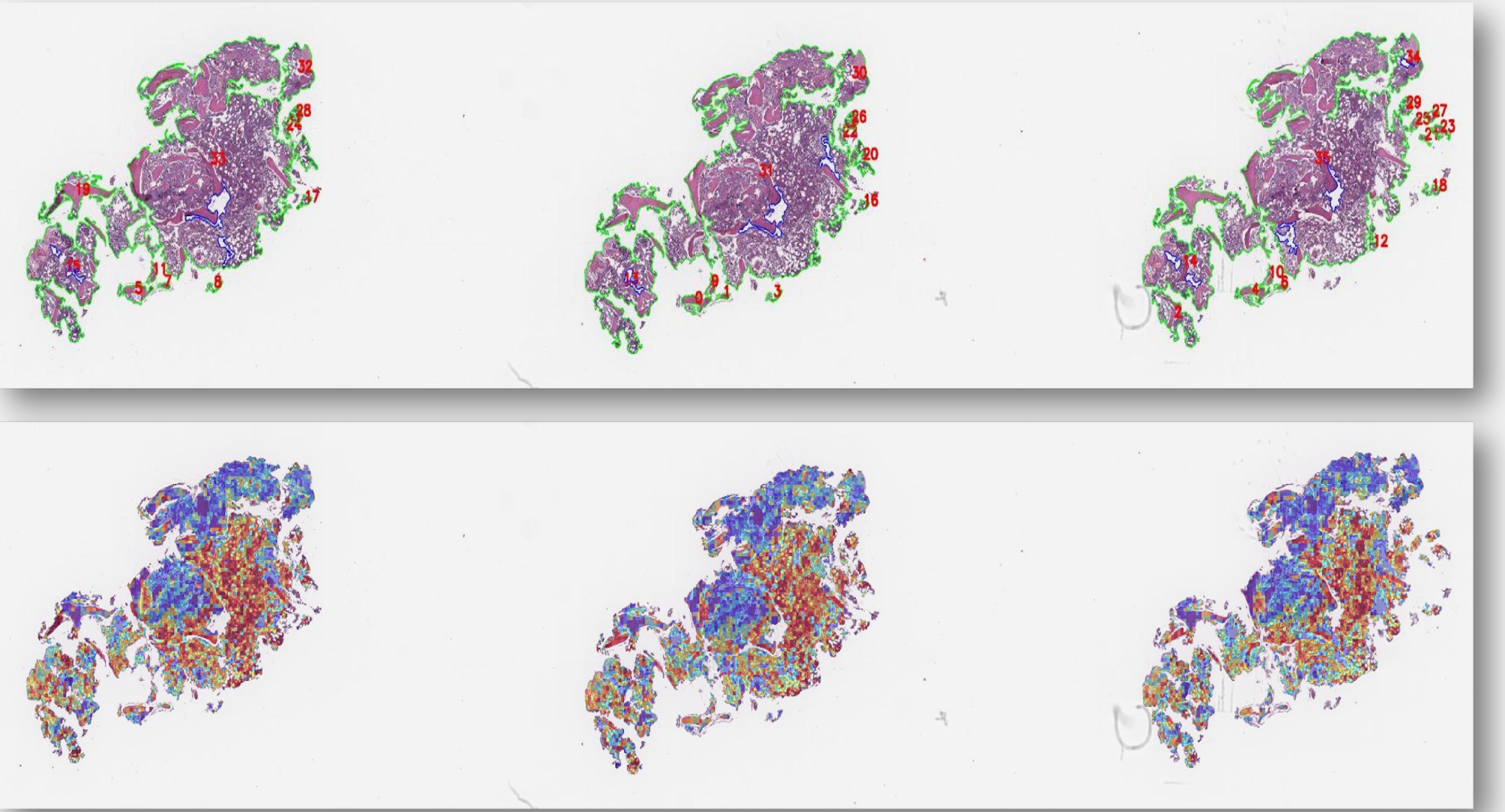
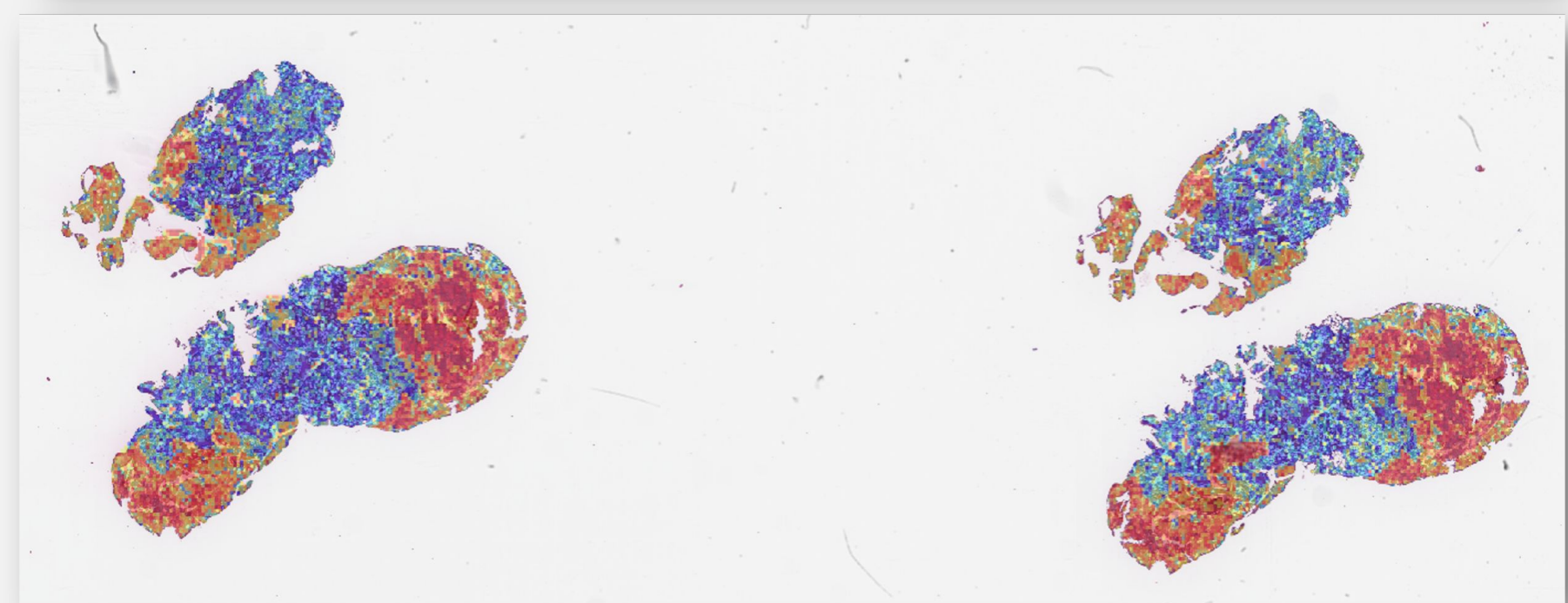
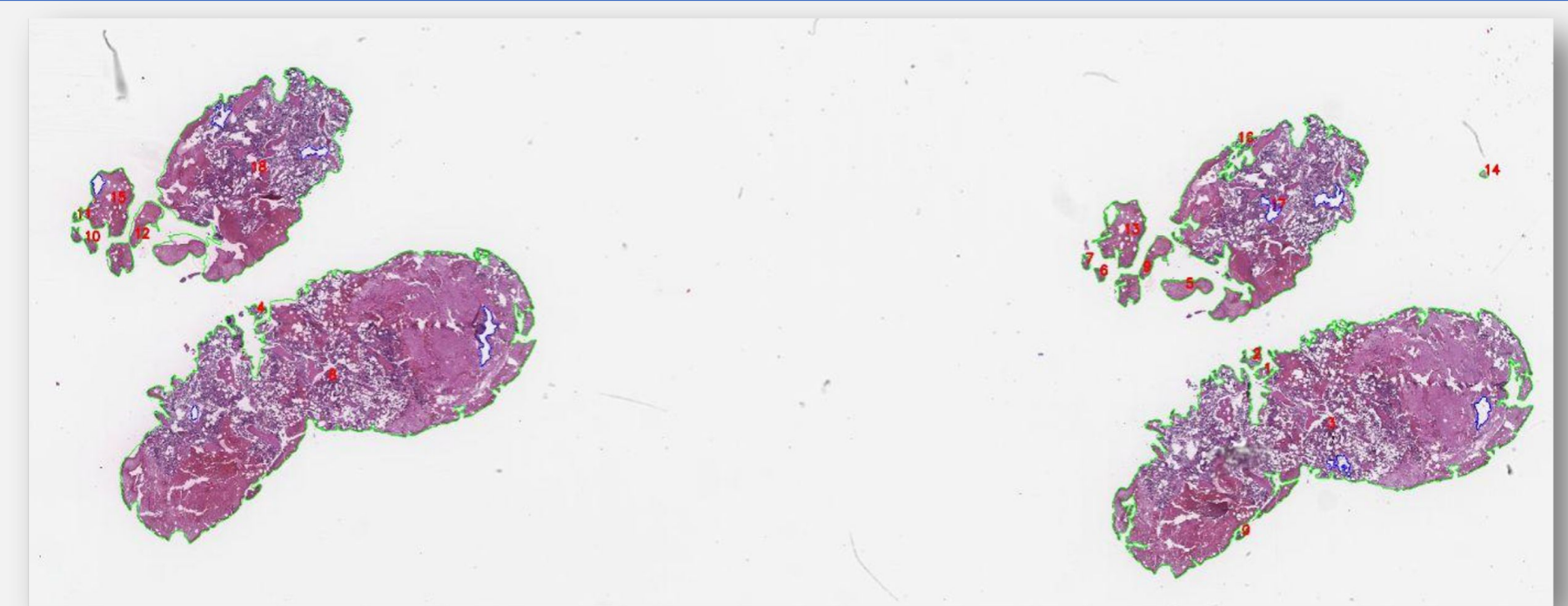
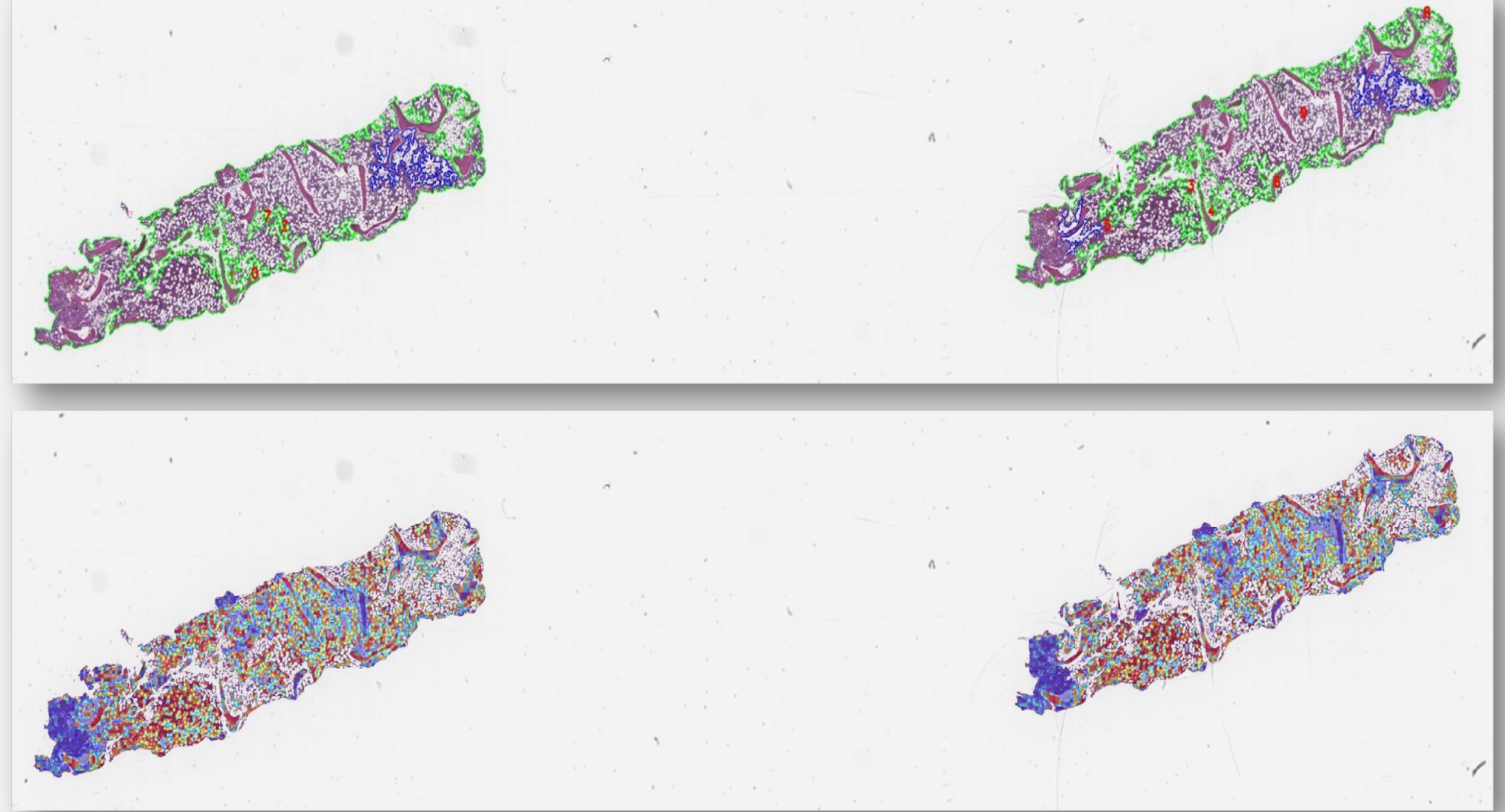


Figure 6: Our model's deep attention maps of predicted and FISH test as t(11;14) negative.



CONCLUSIONS

- Genomic prediction:** AUCs: amp 1q (0.836), hyperdiploidy (0.800), t(4;14) (0.779), del 13q (0.767), del 17p (0.759), t(11;14) (0.754), 1q gain (0.747), del 1p (0.723); OS c-index (0.7).
- Novel classification:** 12 histomolecular clusters; 7 genomic subgroups, 3 hyperdiploidy variants, 2 genomically silent.
- Treatment insights:** 5 TRCs identified; TRC1 superior OS without ASCT (47.5 mo, CR 33.3%); TRC2 better PFS with ASCT (28.9 mo, CR 50%) but shorter OS (32.2 mo) than TRC1; TRC4/5 high-risk with poor responses (CR 14.3%, 25%).
- Clinical impact:** Accessible genomic classification from routine H&E with scalable multicenter architecture for expansion to other malignancies.

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ACKNOWLEDGEMENTS



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