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Automated prediction of response to neoadjuvant chemotherapy from digitized H&E slides of pretreatment biopsies of germline BRCA carriers with HER2-negative breast cancer (TBCRC 031)



Abstract # 605

Nadine M. Tung¹, Stuart J. Schnitt^{2,3}, Judy Ellen Garber³, Satabhisa Mukhopadhyay⁴, Tathagata Dasgupta⁴, Craig A. Bunnell³

1. Beth Israel Deaconess Medical Center, Boston, MA; 2. Brigham and Women's Hospital, Boston, MA; 3. Dana-Farber Cancer Institute, Boston, MA; 4. 4D Path Inc., Newton, MA

BACKGROUND

- The TBCRC031 (INFORM), a randomized phase II trial of neoadjuvant cisplatin (C) vs doxorubicin-cyclophosphamide (AC) in germline BRCA mutation carriers with HER2negative breast cancer, showed no difference in response rates between the two arms ¹.
- A systematic and direct computational measurement of cell cycle deregulation (CCD) and its interaction with tumor immune microenvironment (TIME) from the pretreatment core needle biopsies (CNB) could potentially be useful in automated prediction of the neoadjuvant chemotherapy (NAC) response 2,3, achieving residual cancer burden 0/1 (RCB 0,1) at surgery.
- The current study examines if quantification of the immune, proliferative, and key cell cycle deregulation signatures from digital images of H&E-stained pre-treatment CNB could distinguish patients with a good response to NAC (RCB 0,1) from those with a poor response (RCB 2,3).

METHODS

- CNBs (n = 92 with RCB scores available) scanned at 40x on a Hamamatsu Nanozoomer scanner were evaluated using the 4D Q-Plasia OncoReader (QPOR™) generated image-based signatures (Figure 1). 96% (88/92) of the CNBs could be analyzed successfully.
- **QPOR™** is a statistical physics and tumor biology-based hard-coded algorithm that quantifies CCD and TIME dynamics as continuous biomarkers from digitized H&E CNB whole slide images (WSI) alone.
- **QPOR™** biomarkers used in the study:
 - IHI: complex immune response signature based on the intra-slide spatial heterogeneity of immune infiltrates.
 - Cmbl: combined IHI, proliferative (PI), and cell cycle G1S deregulation (G1SI) signatures based on the hypotheses and prediction-logic described in Figure-2. **CmbI-Tree** and **CmbI** are categorical and continuous signatures respectively.
- All the biomarkers were validated out of sample both as i) continuous scales, and ii) <u>dichotomized</u> at median value to predict RCB 0,1 vs. RCB 2,3 status.
- Receiver operator characteristic (ROC) curves with area under the curve (AUC) measures and odds ratios (OR) per unit increase with 95% confidence intervals (CI) were used to assess the predictive performance of each continuous biomarker.
- Accuracy, specificity, sensitivity, positive and negative predictive values (PPV, NPV), and OR with 95% CI were used to assess the predictive performance of the dichotomized biomarkers.
- The analysis is repeated for each therapy arms (C and AC), the TNBC and the BRCA-1 sub-cohorts, using two different image formats (ndpi 40x and tiff 40x).
- Other **QPOR™** indices were analyzed using AUC-ROC to test further reduction of false positive (FP) and false negative (FN) rates. Pathology review was done on FP and FN cases to determine if they were associated with any specific features.

IHI: Complex immune response signature from spatial heterogeneity of immune **CmbI**: Combined **IHI**, proliferative (**PI**) and cell cycle G1S deregulation (G1SI) signatures Pre-treatment core Automated Prediction of RCB needle biopsies (n = 88) by **QPOR**" Core Computation Biological Signature Map Image Preprocess and Prediction 1. Computer vision-based hard-. Tumor biology-based hard-L. Statistical physics-based hardcoded module coded module coded module 2. Auto-selection of IAA. 2. Computes the CCD map and 2. Aligns the biomarkers with the TIME dynamics. underlying scientific 3. Converting the WSI to FOVs. 3. Extracts digital analyte values 4. Uses grayscale images and 3. Maps them into various per FOV and its derivatives. nuclear segmentation. categorical predictive outputs 5. Auto-screening of FOVs to be 4. Aggregates them into a set of 4. Reports both continuous biomarker values per slide, passed to the core capturing local and non-local scales and categorical computation module predictions. 5. Back-projects select biomarkers on the original H&E WSI for ROI annotation.

Figure 1: Study design and QPOR workflow following image upload; IAA = Initial Analyzable Area, FOV = Field of View, ROI = Region of Interest, CCD = Cell Cycle Deregulation, TIME = Tumor Immune Micro-Environment

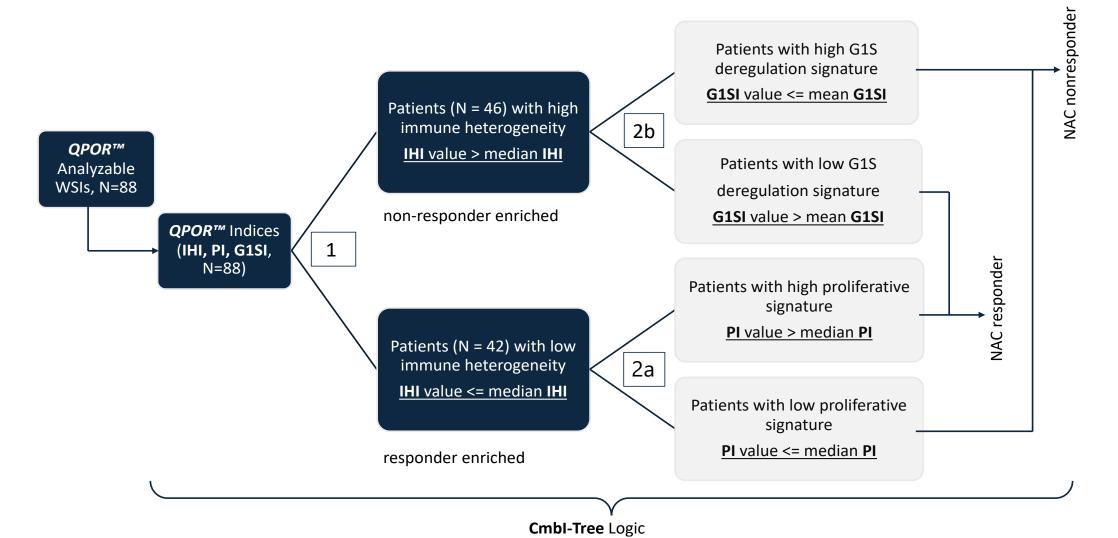


Figure 2: Hypotheses and prediction logic behind CmbI-Tree and CmbI construction: 1) higher immune heterogeneity is likely to associate with poor chemotherapy response; 2a) lower proliferative status is likely to associate with poor chemotherapy response, particularly for patients with low immune heterogeneity; 2b) higher G1S deregulation is likely to associate with poor chemotherapy response, which is likely more significant in patients with high immune heterogeneity. Assigning projected %confidence weights for each arm and then aggregating produces the continuous CmbI value. CmbI >= 0 predicts responders (median cut-off =0).

• Low IHI was predictive of NAC response in the overall population (Table 1) and showed better prediction of response compared to pathologist identified % sTILs (the latter had nonsignificant Wilcoxon test p-values). • High Cmbl was significantly predictive of NAC response in the full cohort with higher accuracy, as well as in the AC, C, and TNBC sub-cohorts and predicted response better than IHI alone (Table 1, 2). • Cmbl was predictive of NAC response across different image filetypes (ndpi, tiff, svs) and in BRCA proficient TNBC cohort. • Most of the FP and FN cases (> 90%) were found to be associated with specific features (e.g., prominent necrosis).

RESULTS

Table 1: Combined signature (CmbI*) is superior to immune heterogeneity signature (IHI*) to predict neoadjuvant chemotherapy response (RCB 0,1)

Subgroup	Biomarker	AUC (95% CI)	OR (95% CI)	P value (Wilcoxon test)
AII (N = 88)	IHI	0.65 (0.53, 0.77)	0.58 (0.33, 0.95)	0.016
AII (N = 88)	CmbI	0.80 (0.71, 0.90)	1.23 (1.13, 1.36)	3.3e-07
AC (N = 43)	IHI	0.62 (0.45, 0.80)	0.68 (0.31,1.36)	0.18
AC (N = 43)	CmbI	0.83 (0.71, 0.96)	1.26 (1.11, 1.47)	<u>1e-04</u>
C (N = 45)	IHI	0.68 (0.51, 0.84)	0.48 (0.2, 0.99)	0.047
C (N = 45)	CmbI	0.77 (0.63, 0.91)	1.21 (1.07, 1.39)	<u>0.0015</u>
TNBC (N = 54)	IHI	0.64 (0.48, 0.80)	0.60 (0.31, 1.08)	0.086
TNBC (N = 54)	CmbI	0.85 (0.75, 0.95)	1.32 (1.16, 1.57)	<u>4.3e-06</u>

*All biomarkers were evaluated as continuous scales. High CmbI, and low IHI predicted NAC response (RCB 0,1) as continuous biomarkers. The unit increase/decrease of IHI, and CmbI is in a scale of 100, and 0.1, respectively.

Table 2: Performance of <u>dichotomized</u> CmbI** to predict response (RCB 0,1) to neoadjuvant chemotherapy

Biomarker	TP: RCB 0, 1	TN: RCB 2, 3	FP	FN	Accuracy	Sensitivity	Specificity	PPV	NPV	OR (95% CI)
CmbI	29	41	6	12	80%	71%	87%	83%	77%	15.56 (5.52, 50.96)

**The dichotomized CmbI predicted NAC response (RCB 0,1) according to the rule explained Figure 2.

CONCLUSIONS

- Low immune heterogeneity (IHI), high proliferative (PI), and high cell cycle deregulation measured by low G1S length (G1SI) signature indices generated by the computational algorithm ($QPOR^{TM}$) individually, and much more strongly in combination (CmbI), are predictive of NAC response in germline BRCA mutation carriers with HER2-negative breast cancers.
- The combined index (CmbI) demonstrated superior predictive performance in the overall population, in each therapy arm and in the TNBC sub-cohort.

References: 1. Tung, N. et al. J Clin Oncol 2020;

2. Williams, G. H. J Pathol. 2012;

3. Tisled, C. et al. Frontiers in Oncol 2022.

Corresponding author: Nadine M. Tung, MD Contact: ntung@bidmc.harvard.edu



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