

Transcriptomic insights into lobular breast cancer biology: a retrospective analysis of the MINDACT clinical trial

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BACKGROUND & OBJECTIVES

- Invasive lobular breast cancer (ILC) represents the 2nd most common type of breast cancer and differs from invasive breast cancer of no special type (NST) at different levels, recently reviewed in [1],
- Most ILC are expressing the estrogen receptor (ER) and lacking HER2 amplification (ER+/HER2-).
- It is crucial to better characterize biological features characterizing ILC from NST, those differing between classic and non-classic ILC as well as features associated with prognosis.

Here, using the transcriptomic data from the MINDACT trial, we aimed at identifying/refining the transcriptomic differences between:

Objective 1: ER+/HER2- NST versus ER+/HER2- ILC

Objective 2: non-classic and classic ER+/HER2-ILC

Objective 3: recurring and non-recurring ER+/HER2- ILC in the subgroup of patients with a low clinical and low genomic (cL/gL) risk

PATIENTS & METHODS

- Central pathology review was performed for histological subtype, grade and Ki67 (G.V.) for 5929/6693 (88.6%) of the patients included in the MINDACT trial (NCT00433589)
- **Transcriptomic analyses:** Analysis of bulk transcriptomic data [1] was performed using the R/Bioconductor package 'limma' (version 3.46.0) to identify differentially expressed genes according to histologic subtype (NST vs ILC), subclassification of ILC (non-classic vs classic), genomic risk (high vs low), and relapse incidence (relapse vs no relapse). Linear models were adjusted for age (> 50 vs \leq 50) and tumor grade (G3, G2 vs G1). False discovery rate (FDR) was controlled by p-value adjustment using the Benjamin-Hochberg method. Differentially expressed genes (DEGs) were determined as those having absolute log-fold change (logFC) \geq 0.2 and FDRadjusted p-value (q-value) < 0.05.
- In a corresponding manner, we performed aene set enrichment analyses using two independent approaches: the supervised population-based Gene Set Enrichment Analysis (GSEA – version 4.1.0)[2], and the unsupervised single sample-based method Gene Set Variation Analysis (R package 'GSVA' – version 1.40.1)[3]. The former method was executed using the complete list of genes preranked by the logFC of the prior differential gene expression analysis. Hallmark gene sets available in the H collection of MSigDB (version 7.5.1) were used as references[4]. Hallmarks having an absolute normalized enrichment score (NES) \geq 1 and q-value < 0.05 were considered differentially enriched
- Statistical analyses: Statistical analyses were performed using R version 4.1.1. All statistical tests were two-sided and were considered statistically significant when p-value < 0.05. Fisher's exact test was used to assess the association of clinicopathological variables with histologic subtype (NST vs ILC), subclassification of ILC (non-classic vs classic) and genomic risk (high vs low). The evaluated clinicopathological variables include age (>50 vs \leq 50), menopausal status (post- vs pre-menopausal), tumor grade (G3, G2 vs G1), tumor size ($\geq 2 \text{ cm vs} < 2 \text{ cm}$), nodal status (positive vs negative), receptor status (positive vs negative), Ki67 level (20% - 100%, 14% - < 20% vs < 14%), genomic risk (if appropriate, high vs low) and clinical risk. Univariable and multivariable Cox regression models were used to evaluate the association between GSVA scores of molecular hallmarks with disease-free survival (DFS).
- The clinical risk and genomic risk was defined by a modified version of Adjuvant Online! and the 70-gene signature, respectively.



RESULTS **Clinico-pathological comparison ILC vs NST**

After central pathological review, 464 patients with ER+/HER2-ILC and 3798 patients with ER+/HER2- NST were identified. Patients with ILC were significantly older at diagnosis, had larger tumors, less axillary nodal involvement, more grade 2 tumors than patients with NST (Table 1).

Table	1: Clinical	and	pathological	ch
and N				

		ILC
		(
		n
Δσe	≤ 50	117
	> 50	347
	Pre-menopausal	139
Menopausal	Post-	305
status	menopausal	
	Unknown	20
	G1	81
	G2	361
	G3	22
	Unknown	0
Tumor size	< 2cm	258
	≥ 2cm	206
Nodal status	Negative	378
Noual Status	Positive	86
	Negative	37
PR status	Positive	424
	Missing	3
	< 14%	255
Ki67	14% - < 20%	136
	20% - 100%	70
	Missing	3
Clinical risk	Low (cL)	249
Cillicat Lisk	High (cH)	215
Genomic risk	Low (gL)	394
	High (gH)	70
	cL/gL	221
Corrected rick	cL/gH	28
confected fisk	cH/gL	173
	cH/gH	42

FLOWCHART

OBJECTIVE 1 Transcriptomic comparison NST vs ILC

Figure 1. Transcriptomic differences ER+/HER2- NST and ILC. (A) Volcano plots of Figure 2: Transcriptomic differences between the non-classic and classic differential gene expression analysis by 'limma'. The x-axis represents the log2 fold subclassifications of ILC ER+/HER2- tumours. (A) Volcano plots of differential gene change (logFC) of gene expression in NST tumours relative to ILC tumours. The y- expression analysis. Genes with absolute logFC > 0.2 and q-value < 0.05 are axis represents the -log10 of FDR-adjusted p-value (q-value), the higher -log10(q- highlighted and labelled (red: upregulated in non-classic ILC, green: upregulated value) the smaller q-value. Genes with absolute logFC > 0.2 and q-value < 0.001 in classic ILC). (B) Lollipop plots displaying differentially enriched molecular are highlighted and labelled (red: upregulated in NST, green: upregulated in ILC). hallmarks according to ILC subclassification detected by GSEA (q-value < 0.1) (B) Lollipop plots displaying differentially enriched molecular hallmarks according and heatmap showing their corresponding average enrichment scores to histological subtype detected by GSEA (q-value < 0.1) and heatmap showing computed by GSVA (positive NES: enriched in variant ILC (non-classic, NC), their corresponding average enrichment scores computed by GSVA. The length negative NES: enriched in classic ILC (C)). of the lollipops represents the absolute value of the normalized enrichment score (NES) of a hallmark in NST tumours compared to in ILC tumours. The sign of the NES indicates the orientation of the differential enrichment (positive: enriched in NST, negative: enriched in ILC)

CDH1

NST ER+/HER2-+/HER2-Fisher's exact p-value 25.2 1221 32.1 0.0025 2577 67.9 1372 31.3 37.5 2291 62.5 0.0123 68.7 135 17.5 953 25.2 2310 61.0 77.8 0.0005 4.7 526 13.9 2614 55.6 68.8 < 0.0001 44.4 1184 31.2 2938 81.5 77.4 0.0443 22.6 18.5 860 8.0 306 8.1 92.0 3488 91.9 1.0000 55.3 1386 36.6 29.5 1090 28.8 0.0005 15.2 1315 34.7 2170 57.1 53.7 0.1646 46.3 1628 42.9 2758 72.6 84.9 < 0.0001 15.1 1040 27.4 1813 47.7 47.6 357 6.0 9.4 0.0005 37.3 945 24.9

683

9.1

18.0

naracteristics of patients with ER+/HER2- ILC



OBJECTIVE 2

Transcriptomic comparison non-classic vs classic ILC



C NC

Recurring vs not-recurring cL/gL ILC

Figure 3. Association of enrichment of molecular hallmarks with DFS in the subgroup of ER+/HER2- patients with cL/gL. Forest plots presenting hazard ratios (HRs) of GSVA scores of molecular hallmarks with DFS estimated by univariable and multivariable Cox regression models. Hallmarks associated with DFS (p-value < 0.1) are shown. (HR > 0: higher enrichment is associated with worse DFS; HR < 1: higher enrichment is associated with better DFS).

Samples Event HALLMARK_APOPTOSIS Model 2 HALLMARK COMPLEMEN Model 1 Model 2 HALLMARK_DNA_REPAI Model 2 HALLMARK E2F TARGET Model 2 HALLMARK HYPOXI Model 2 HALLMARK IL2 STAT5 SIGNALIN Model Model 2 HALLMARK_IL6_JAK_STAT3_SIGNALIN Model 2 HALLMARK INFLAMMATORY RESPONS HALLMARK_INTERFERON_ALPHA_RESPONS Model 2 HALLMARK INTERFERON GAMMA RESPONSI Model Model 2 HALLMARK_KRAS_SIGNALING_DM Model 2 HALLMARK MTORC1 SIGNALIN Model 2 HALLMARK_MYC_TARGETS_V Model 2 HALLMARK_MYC_TARGETS_V Model 2 HALLMARK_PI3K_AKT_MTOR_SIGNALI HALLMARK_TNFA_SIGNALING_VIA_NFK Model 2 HALLMARK UNFOLDED PROTEIN RESPONSE Model 1 Model 2 HALLMARK_UV_RESPONSE_U Model 2

CONCLUSIONS

- signaling and increased PI3K/Akt signaling.
- signaling in classic ILC.
- 3. Enrichment of hallmarks related to apoptosis, inflammatory associated with worse survival in patients with cL/gL ILC.

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OBJECTIVE 3

		HR (95% CI)	P Value	
		13.705 (1.157 - 162.299) 11.605 (0.841 - 160.221)	0.038 0.067	
		5.41 (0.978 - 29.943) 5.642 (0.896 - 35.552)	0.053 0.065	
ł		4.965 (0.355 - 69.499) 12.864 (0.689 - 240.084)	0.234 0.087	
		2.971 (0.615 - 14.343) 5.528 (0.982 - 31.12)	0.175 0.052	
		11.149 (0.863 - 144.093) 10.303 (0.695 - 152.624)	0.065 0.090	
		7.67 (0.954 - 61.666) 6.707 (0.761 - 59.142)	0.055 0.087	
		3.706 (0.824 - 16.664) 3.822 (0.785 - 18.618)	0.088 0.097	
		4.012 (0.911 - 17.68) 3.466 (0.722 - 16.65)	0.066 0.121	 Model 1: Univariable Model 2: Adjusted for: Age,
	⊨ ∎ -1 ⊨ ∎ -1	2.484 (0.881 - 7.005) 2.756 (0.905 - 8.394)	0.086 0.074	Tumor grade, Tumor size, Nodal status,
	┝╾ Ш →┥ ┝╼ Ш →┥	2.85 (0.851 - 9.549) 3.089 (0.821 - 11.623)	0.090 0.095	Endocrinetherapy, Radiotherapy
		0.013 (0.001 - 0.288) 0.012 (0 - 0.316)	0.006 0.008	
		7.129 (0.807 - 62.984) 9.645 (1.045 - 89.006)	0.077 0.046	
		3.694 (0.619 - 22.057) 6.005 (0.877 - 41.117)	0.152 0.068	
	⊨- ■ -1 ⊨- ■ -1	3.556 (0.918 - 13.777) 5.596 (1.317 - 23.769)	0.066 0.020	
		11.817 (1.24 - 112.613) 13.56 (1.337 - 137.498)	0.032 0.027	
		3.881 (0.952 - 15.822) 3.518 (0.816 - 15.16)	0.059 0.091	
		6.204 (0.623 - 61.796) 8.225 (0.705 - 95.907)	0.120 0.093	
		30.916 (2.321 - 411.753) 54.236 (3.537 - 831.633)	0.009 0.004	

0.01 0.1 1 10 100 Hazard ratio

Marked transcriptomic differences were identified between ER+/HER2- NST and ILC, with ILC presenting differences in lipid metabolism and in the extracellular matrix, a decreased ER-

2. Differences between classic and non-classic ER+/HER2- ILC were more subtle with enrichment of the hallmarks related to cell cycle in the non-classic ILC, and of the hallmarks related to epithelial-tomesenchymal transition, hypoxia, adipogenesis and IL6/JAK/STAT3

response, hypoxia and oncogenic signaling (PI3K/Akt, c-Myc) is

REFERENCES