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The CIN4 Chromosomal Instability qPCR Classifier Defines Tumor Aneuploidy and Stratifies Outcome in Grade 2 Breast Cancer

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Abstract

Purpose: Quantifying chromosomal instability (CIN) has both prognostic and predictive clinical utility in breast cancer. In order to establish a robust and clinically applicable gene expression-based measure of CIN, we assessed the ability of four qPCR quantified genes selected from the 70-gene Chromosomal Instability (CIN70) expression signature to stratify outcome in patients with grade 2 breast cancer.

Methods: AURKA, FOXM1, TOP2A and TPX2 (CIN4), were selected from the CIN70 signature due to their high level of correlation with histological grade and mean CIN70 signature expression in silico. We assessed the ability of CIN4 to stratify outcome in an independent cohort of patients diagnosed between 1999 and 2002. 185 formalin-fixed, paraffin-embedded (FFPE) samples were included in the qPCR measurement of CIN4 expression. In parallel, ploidy status of tumors was assessed by flow cytometry. We investigated whether the categorical CIN4 score derived from the CIN4 signature was correlated with recurrence-free survival (RFS) and ploidy status in this cohort.

Results: We observed a significant association of tumor proliferation, defined by Ki67 and mitotic index (MI), with both CIN4 expression and aneuploidy. The CIN4 score stratified grade 2 carcinomas into good and poor prognostic cohorts (mean RFS: 83.8±4.9 and 69.4±8.2 months, respectively, p = 0.016) and its predictive power was confirmed by multivariate analysis outperforming MI and Ki67 expression.

Conclusions: The first clinically applicable qPCR derived measure of tumor aneuploidy from FFPE tissue, stratifies grade 2 tumors into good and poor prognosis groups.

Introduction

Chromosomal instability (CIN) is a key determinant of biological behavior of breast cancer [1], yet remains challenging to determine using high throughput methodologies [2]. We and others have shown that CIN may play a role in determining response to taxane treatment in ovarian cancer [3] and intrinsic multidrug resistance in colon cancer [4] and it also appears to be an important determinant of breast cancer prognosis [5,6]. While direct determination of CIN by counting centromeres in a sufficient number of breast cancer cells or measuring DNA index is possible [7], it is technically challenging and time-consuming [8,9]. Consequently, a simple measure of CIN, based on for example qPCR measurement of a handful of genes, would greatly facilitate its introduction into general oncological practice. Therefore, we decided to investigate whether a clinically implementable a qPCR-based gene expression based CIN measure could be created. The ability of such signatures to reflect CGH based genomic instability has been previously demonstrated [5].
The potential utility of a gene expression based measure of CIN is further emphasized by its complex relationship with histological grade [10]. It has been known for many years that grade 2 tumors classified by the Nottingham grading system display heterogeneous characteristics in terms of clinical outcome [11,12]. It has also been shown that relatively simple gene expression based methods were able to stratify grade 2 cases into low risk and high risk patients in a robust fashion [13]. Genes intimately involved with CIN prominently featured in such gene expression signatures, but it has become clear recently that quantifying ploidy has improved prognostic potential than stratifying intermediate histological grade in terms of clinical outcome. The vast majority of hormone receptor negative cases (ER-/PgR-/HER2-) fall into histological grade 3 category [14], but even within those, the level of CIN may define subgroups of patients with distinct clinical outcomes [5]. Therefore in this study, we determined a minimal gene set that seems to capture the information contained in the previously published CIN70 signature, tested its correlation with directly quantified CIN and verified its ability to stratify grade 2 breast carcinomas according to good and poor clinical outcome in routine formalin-fixed, paraffin-embedded (FFPE) pathological samples. We compared the predictive power of the resulting signature to mitotic index and Ki67 expression.

**Materials and Methods**

**In silico selection of CIN genes for expression analysis**

All microarray data sets used in this analysis were normalized by robust multi-array average (RMA) [15]. Probe sets were selected from each gene in the various signatures (CIN70 [1], NKI70 [16], 21-gene Recurrence Score (DX21) [17], intrinsic subtype [18], Ishihana’s molecular grade [11], Mas’s 5-gene HOXB13:IL17BR ratio (M5) [19], Sotiropoulos’ Genomic Grade Index (GGI) [20]) analyzed were selected by Jetset as described previously [21]. If no reliable probe set was found for a given gene that was excluded from further analysis. The data sets are displayed in Table S1.

**Tissue samples**

Focusing on histological grade, we evaluated 185 invasive breast carcinomas consisting of 63 grade 1, 62 grade 2 and 60 grade 3 FFPE tissue samples regardless of other pathological features from the Buda MAV Hospital (1999–2002), diagnosed and graded by a single pathologist (J.K.). Recurrence-free survival (RFS) time was defined either by loco-regional relapse or the appearance of a distant metastasis, and whichever shorter if both applicable. The study was a retrospective analysis. General written consent was obtained from all patients at time of surgery. The samples were anonymised for the study. The study was approved by the Ethical Committee of the Semmelweis University (#7/2008 and #7-1/2008).

**Patient characteristics**

In line with bioinformatics, the clinicopathological properties of the selected 185 breast cancer patients were analyzed. The mean age of patients was 58.8±12.8 years (59.9±11.8 years, 59.2±13.0 years, 56.5±13.7 years in grade 1, 2 and 3 tumor groups, respectively). Among the histological types, invasive ductal carcinoma was the most common overall (63.9%), but less frequent types of cancer were also included in the analysis (8.1%): 1 papillary, 1 tubular, 1 micropapillary and 3 mucinous carcinomas in the grade 1 group; 1 micropapillary and 1 mucinous in grade 2; and 4 with medullary features, 2 metaplastic and 1 micropapillary carcinomas in grade 3 cancers. Tumor size, frequency of vascular invasion, presence of necrosis, Nottingham Prognostic Index (NPI) and number of relapses showed an increase, while ER, PgR expression and RFS decreased with higher grade (Table 1). When characterizing tumors according to immunophenotype, expression of ER, PgR and Her2 were evaluated and fluorescence in situ hybridisation (FISH) was performed to assess Her2 amplification. ER and PgR expressing tumors with lower than 20% Ki67 expression were considered as Luminal A (LumA). For ER or PgR and Her2 positive tumors and ER and PgR expressing tumors with more than 20% Ki67 index, Luminal B subtype (LumB) was assigned [22]. Her2 expressing and/or Her2 amplified and ER and PgR negative tumors were considered as the Her2 subgroup (none were treated with trastuzumab at that time). Triple negative breast cancers (TNBC) were ER, PgR, Her2 negative tumors with no Her2 amplification. The disease free survival of patients in the different grade- and immunophenotype groups were plotted on Kaplan-Meier graphs (Figure S1).

**RNA purification and qPCR**

Five to ten 5 μm thick sections were used from each case for RNA purification after assessment of cellularity on HE stained slides (minimum of 70% tumor cell content was required). RNA was extracted with Qiagen FFPE RNeasy kit according to the manufacturer’s protocol (Qiagen, Venlo, The Netherlands). High Capacity RNA-to-cDNA kit was used to reverse transcribe 1000 ng of RNA (Applied Biosystems, Foster City, CA, USA). The Eppendorf epMotion 5070 robotic system was used to transfer samples and reagents to 384-well full-skirted white plates (Eppendorf, Hamburg, Germany). The qPCR was performed in duplicates with Taqman® Assays (Table S2) in Gene Expression Master Mix (all from Applied Biosystems) according to the manufacturer’s protocol. The reactions were run in Roche LightCycler 480 real-time PCR system (Roche Diagnostics, Mannheim, Germany). The CIN4 signature resulting from the qPCR measurement was assessed based on the average expression of the four genes (AURKA, FOXM1, TOP2A, TPX2) normalized to the average expression of the three control genes. In order to find robust internal reference control genes for clinical evaluation of CIN4 by qPCR in formalin-fixed tumor tissue, we chose three known housekeeping genes which bare consistently low variance in the microarray profiles of all breast cancer datasets as normalizing genes: B4GALT3, SLCA93R2 [23] and PUM1 [24].

**Flow cytometry**

Flow cytometry was performed for the analysis of ploidy. Briefly, a 50 μm section was cut from all the FFPE blocks. A scroll of tissue was placed in a microcentrifuge tube and xylene was added to remove the paraffin wax. The tissue was then serially rehydrated through 100%, 95%, 70% and 50% ethanol for 5 minutes respectively at room temperature. The tissue was washed twice with distilled water. A suspension of nuclei was made by incubating the tissue in a 0.5% pepsin solution (Sigma-Aldrich, Dorset, UK) prepared in 0.9% saline at pH 1.5. Incubation was carried out at 37°C for 30 minutes. The nuclei were washed once with PBS, stained with propidium-iodine and analyzed using the Calibur 1 FACS machine and CellQuest software (Beckman Coulter, Buckinghamshire, UK). DNA index was assigned as follows: diploids were ‘1.0’, a tumor with a DNA index greater than 1.10 was classified as aneuploid [25].

**Statistics**

The assignment of each patient into two cohorts using the CIN4 expression signature was performed in the R statistical environment (R, version 2.10.1) using the package Prediction Analysis for
Table 1. Clinicopathological data of the 185 breast cancer patients included in the analysis.

<table>
<thead>
<tr>
<th>Groups according to grade</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>All</th>
</tr>
</thead>
<tbody>
<tr>
<td>Patients (n)</td>
<td>63</td>
<td>62</td>
<td>60</td>
<td>185</td>
</tr>
<tr>
<td>Age – years/mean, (range)/</td>
<td>59.9 (35–95)</td>
<td>59.2 (23–87)</td>
<td>56.5 (29–87)</td>
<td>58.8 (23–95)</td>
</tr>
<tr>
<td>Histology (n, %)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IDC</td>
<td>45 (71.4%)</td>
<td>37 (59.6%)</td>
<td>40 (66.6%)</td>
<td>122 (65.9%)</td>
</tr>
<tr>
<td>ILC</td>
<td>4 (6.3%)</td>
<td>3 (4.8%)</td>
<td>1 (1.8%)</td>
<td>8 (4.3%)</td>
</tr>
<tr>
<td>Mixed</td>
<td>8 (12.6%)</td>
<td>20 (32.2%)</td>
<td>12 (20.0%)</td>
<td>40 (21.6%)</td>
</tr>
<tr>
<td>Other</td>
<td>6 (9.5%)</td>
<td>2 (3.2%)</td>
<td>7 (11.6%)</td>
<td>15 (8.1%)</td>
</tr>
<tr>
<td>Tumour size - mm (mean ± SE)</td>
<td>20.49±1.10</td>
<td>26.81±1.54</td>
<td>28.55±1.92</td>
<td>25.13±1.87</td>
</tr>
<tr>
<td>Vascular Invasion/n (%)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Necrosis/n (%)/</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>NPI (mean ± SE)</td>
<td>3.00±0.11</td>
<td>4.39±0.13</td>
<td>5.58±0.15</td>
<td>4.39±1.39</td>
</tr>
<tr>
<td>Immunophenotype (n, %)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lum A</td>
<td>55 (87.3%)</td>
<td>40 (64.5%)</td>
<td>22 (36.6%)</td>
<td>117 (63.2%)</td>
</tr>
<tr>
<td>Lum B</td>
<td>4 (6.3%)</td>
<td>8 (12.9%)</td>
<td>15 (25.0%)</td>
<td>27 (14.5%)</td>
</tr>
<tr>
<td>HER2</td>
<td>0</td>
<td>3 (4.8%)</td>
<td>7 (11.6%)</td>
<td>10 (5.4%)</td>
</tr>
<tr>
<td>TNBC</td>
<td>4 (6.3%)</td>
<td>11 (17.7%)</td>
<td>16 (26.6%)</td>
<td>31 (16.7%)</td>
</tr>
<tr>
<td>Recurrence (n, %)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Local relapse</td>
<td>8 (12.6%)</td>
<td>5 (8.1%)</td>
<td>7 (11.6%)</td>
<td>20 (10.8%)</td>
</tr>
<tr>
<td>Distant metastasis</td>
<td>8 (12.6%)</td>
<td>21 (33.8%)</td>
<td>21 (35.0%)</td>
<td>50 (27.0%)</td>
</tr>
<tr>
<td>RFS – months/mean (range)</td>
<td>85.9 (12–122)</td>
<td>75.9 (0–123)</td>
<td>74.1 (0–119)</td>
<td>78.8 (0–123)</td>
</tr>
</tbody>
</table>

Microarrays (PAM, version 2.19) as described previously [26]. By performing a soft thresholding to produce a shrunken centroid, which allows the weighting and ranking of genes with high predictive potential, PAM is capable to classify the samples into two cohorts. After this assignment, Kaplan-Meier curves were used to plot the efficiency of the prediction on recurrence-free survival. Chi-squared test was performed to test relation of qPCR expression of CIN genes, those were selected purely based on their highest average correlation coefficients to tumor CIN score [1]. Correlation with grade was not used in selecting the PCR quantified CIN genes, which were selected purely based on their highest average correlation coefficients to molecular grade [11], Ma5 [19], GGI [20], DX21 [17] and NKI70 [16] in 3 datasets (GIS [11], JBI [20], JBI1 [20,27]). CIN4 performed almost as well as CIN25. The univariate model based on which the meta analysis performed clearly suggested the HR of CIN4 is slightly better than the other predictors (Figure S2a), however, we wanted to know if CIN4 classification adds any predictive power to existing classifiers (predictorX). In another analysis, testing the additive power from CIN4 to a given predictor we determined the following model: survival = CIN4+predictorX. If CIN4 adds any predictive power, the benefit is represented by a significant interaction (CIN4+predictorX). However, we did find that even though CIN4 has a higher HR in the univariate model it does not add significant predictive power to other signatures (Figure S2b).

Results

CIN4 signature derived from in silico ranking of CIN70 genes also reflects histological grade

In order to select a more limited set of genes that optimally reflect the CIN70 signature we retrieved expression profiles for the CIN70 signature genes from 10 publicly available breast cancer datasets [1]. The CIN70 genes were then ranked by Pearson’s correlation coefficients to the CIN score (mean CIN70 expression) within these breast cancer microarray cohorts [supplementary references]. Next, we assessed the association individually on data sets between the mean-expression level of CIN4 and clinical outcome across the same 10 cohorts containing expression data from 1928 breast cancers. We observed significant discriminative power (p<0.05, for all) by CIN4 for the stratification of good from poor clinical outcome in all of the breast cancer cohorts [supplementary references]. Therefore, the expression of CIN4 appears almost as efficient at predicting cancer outcome as the extended CIN25 and CIN70 signatures. We were able to compare the performance in silico of the CIN4 signature to a number of previously published predictors of outcome such as CIN25, CIN70 [1], Ki67, Ishihara’s molecular grade [11], Ma5 [19], GGI [20], DX21 [17] and NKI70 [16] in 3 datasets (GIS [11], JBI [20], JBI1 [20,27]). CIN4 performed almost as well as CIN25. The univariate model based on which the meta analysis performed clearly suggested the HR of CIN4 is slightly better than the other predictors (Figure S2a), however, we wanted to know if CIN4 classification adds any predictive power to existing classifiers (predictorX). In another analysis, testing the additive power from CIN4 to a given predictor we determined the following model: survival = CIN4+predictorX. If CIN4 adds any predictive power, the benefit is represented by a significant interaction (CIN4+predictorX). However, we did find that even though CIN4 has a higher HR in the univariate model it does not add significant predictive power to other signatures (Figure S2b).

Identification of control genes in datasets and their variability in the tissue samples

B4GALT3, SLC9A3R2 and PUM1 were previously chosen based on their low variability in gene expression datasets described previously [23,24]. The three genes showed low variation between the investigated samples as well (Table S3).
CIN4 stratifies outcome in grade 2 tumors in a cohort of 185 breast cancers

Expression of CIN4 was assessed in a retrospective cohort of 185 patients for which we had FFPE primary breast cancer samples available from Buda MAV hospital, treated between 1999 and 2002 (Table 1). Kaplan-Meier curves of the pathological evaluation of Ki67 expression and mitotic index performances on prognosis are shown in Figure S3.

In order to define a threshold for CIN4 expression for distinct outcome groups, we trained the PAM algorithm using the continuous CIN4 gene expression signature to discriminate clinical outcome of 63 patients with grade 1 breast cancer compared to the poorer outcome associated with 60 grade 3 breast cancers from within this cohort of 185 patients. Using this CIN4 expression threshold that best distinguished grade 1 compared to grade 3 breast cancers, the PAM defined CIN4 score was established, and we assessed the ability of this CIN4 score to stratify cancer outcome in the remaining 62 patients with grade 2 breast cancer from this 185 patient cohort.

Using this threshold, the CIN4 score was able to stratify patients with grade 2 breast cancers into good (44 patients) and poor (18 patients) prognostic groups (mean, relapse-free survival: 83.8 months [95%CI: 73.6–94.2] and 69.4 months [95%CI: 53.1–90.6], respectively [p = 0.016], HR = 2.155[1.007–4.612]). (Figure 1A).

Performance of the components of the CIN4 signature/score regardless of immunophenotype

For the identified genes that have been previously described to be of prognostic value, we have evaluated the prognostic power of AURKA, FOXM1, TOP2A, and TPX2 separately. Although, in public breast cancer datasets (grade 2, follow-up: 10 years, n = 497, split at median) all the genes showed strong predictive potential [AURKA: p = 0.047, HR = 1.33[1.00–1.78]; FOXM1: p<0.001, HR = 1.69[1.26–2.27]; TOP2A: p<0.001, HR = 1.68[1.26–2.25]; TPX2: p<0.001, HR = 1.64[1.23–2.18] for relapse-free survival (Figure S4), the data generated by qPCR was unable to distinguish prognostic subgroups for AURKA, FOXM1, and TPX2 at 10, 25, 75, 90 percentiles, median or average expression in our limited number of patients (Table S4). While TOP2A was already able to split prognostic subgroups at 25 percentile and median expression by itself (Table S4). The gain in performance is presented in the same table when considering all the four (CIN4) genes simultaneously.

Performance of CIN4 in ER positive breast tumors

We assessed the performance of the CIN4 score in ER+/HER2– tumors classified by immunohistochemistry. This cohort was also stratified into good and poor prognostic groups with reasonable confidence (p = 0.009, HR = 2.269[1.117–5.486], Figure 1B), suggesting that the CIN4 score-defined group with poor outcome is not driven by the worse outcome of the HER2+ subgroup diagnosed and treated before the introduction of trastuzumab. Other subtypes were under-represented and comparison could not be made in the herein investigated cohort.

CIN4 expression signature reflects aneuploidy directly quantified by FACS analysis

We assessed tumor DNA index through flow cytometry, classifying cancers into aneuploid and diploid categories according to standard threshold (threshold = 1.10 [25]). We evaluated the relationship of DNA index and the CIN4 expression by regression analysis. The variables showed a significant relationship between CIN4 expression and increasing DNA index (p = 0.036, Figure 2A).
(p = 0.001, Figure 3A) and Ki67 expression (p = 0.005, Figure 3C), suggesting that proliferation is correlated with tumor aneuploidy (p = 0.014, p = 0.023, respectively; Figure 3B and D). The Nottingham Prognostic Index (Figure 3E and F) or tumor size (Figure 3G and H) showed a trend only towards correlation with CIN4 signature or with DNA index.

Estrogen and progesterone receptor expression as measured by IHC inversely correlated with CIN4 expression (p = 0.012, p = 0.017, respectively, Figure S5A and S5B). HER2 expression and HER2/C17 FISH counts correlated with CIN4 expression (p = 0.001, p = 0.013, respectively, Figure S5C and S5D), indicating that ER negativity and HER2 positivity are associated with higher CIN4 expression. The data derived from a qPCR assessment of CIN4 mRNA expression from FFPE breast cancers, confirms that the expression of this signature appears to be relatively enriched within ER negative and HER2 positive breast cancers (Figure S6).

CIN4 outperforms known clinical clinicopathological parameters in multivariate outcome models

CIN4 score and clinicopathological variables were tested for their correlation in grade 2 cancers (Table S5). Cox multivariate regression analysis was then performed in grade 2 breast carcinomas: in multiple models CIN4 outperformed clinicopathological variables including hormone receptor status, tumor size,
Table 2. Multivariate regression analysis considering grade 2 breast cancers regarding the CIN4 score and clinicopathological variables.

<table>
<thead>
<tr>
<th></th>
<th>(1a)</th>
<th></th>
<th>(1b)</th>
<th></th>
<th>(2a)</th>
<th></th>
<th>(2b)</th>
<th></th>
<th>(3a)</th>
<th></th>
<th>(3b)</th>
<th></th>
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</thead>
<tbody>
<tr>
<td></td>
<td>P-value</td>
<td>HR</td>
<td>CI</td>
<td>P-value</td>
<td>HR</td>
<td>CI</td>
<td>P-value</td>
<td>HR</td>
<td>CI</td>
<td>P-value</td>
<td>HR</td>
<td>CI</td>
</tr>
<tr>
<td>HR STATUS</td>
<td>0.820</td>
<td>0.87</td>
<td>0.36–3.64</td>
<td>0.099</td>
<td>0.10</td>
<td>0.32–3.13</td>
<td>0.339</td>
<td>0.98</td>
<td>0.92–1.02</td>
<td>0.493</td>
<td>1.02</td>
<td>1.28–2.18</td>
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<tr>
<td>KI67</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0.714</td>
<td>1.20</td>
<td>0.31–2.18</td>
<td>0.799</td>
<td>1.13</td>
<td>0.34–2.27</td>
</tr>
<tr>
<td>LYMPH NODE STATUS</td>
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<td></td>
<td></td>
<td></td>
<td>0.067</td>
<td>0.43</td>
<td>0.94–5.88</td>
<td>0.183</td>
<td>0.55</td>
<td>0.75–4.30</td>
</tr>
<tr>
<td>AGE</td>
<td>0.114</td>
<td>1.03</td>
<td>0.94–1.00</td>
<td>0.168</td>
<td>1.02</td>
<td>0.94–1.01</td>
<td>0.816</td>
<td>0.96</td>
<td>0.65–1.39</td>
<td>0.914</td>
<td>0.98</td>
<td>0.70–1.47</td>
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<tr>
<td>VASCULAR INVASION</td>
<td>0.332</td>
<td>1.04</td>
<td>0.90–1.03</td>
<td>0.658</td>
<td>1.02</td>
<td>0.92–1.06</td>
<td>0.714</td>
<td>1.20</td>
<td>0.31–2.18</td>
<td>0.799</td>
<td>1.13</td>
<td>0.34–2.27</td>
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<tr>
<td>NECROSIS</td>
<td>0.067</td>
<td>0.43</td>
<td>0.94–5.88</td>
<td>0.183</td>
<td>0.55</td>
<td>0.75–4.30</td>
<td>0.067</td>
<td>0.43</td>
<td>0.94–5.88</td>
<td>0.183</td>
<td>0.55</td>
<td>0.75–4.30</td>
</tr>
<tr>
<td>NPI</td>
<td>0.450</td>
<td>1.07</td>
<td>0.78–1.10</td>
<td>0.625</td>
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<td>0.067</td>
<td>0.43</td>
<td>0.94–5.88</td>
<td>0.183</td>
<td>0.55</td>
<td>0.75–4.30</td>
</tr>
<tr>
<td>MITOTIC INDEX</td>
<td>0.332</td>
<td>1.04</td>
<td>0.90–1.03</td>
<td>0.658</td>
<td>1.02</td>
<td>0.92–1.06</td>
<td>0.714</td>
<td>1.20</td>
<td>0.31–2.18</td>
<td>0.799</td>
<td>1.13</td>
<td>0.34–2.27</td>
</tr>
<tr>
<td>TUMOR SIZE</td>
<td>0.047</td>
<td>1.00</td>
<td>0.98–1.03</td>
<td>0.903</td>
<td>1.00</td>
<td>0.98–1.03</td>
<td>0.690</td>
<td>1.01</td>
<td>0.97–1.03</td>
<td>0.919</td>
<td>1.00</td>
<td>0.97–1.02</td>
</tr>
<tr>
<td>CIN4</td>
<td>0.028</td>
<td>0.38</td>
<td>1.11–6.21</td>
<td>0.011</td>
<td>0.33</td>
<td>1.28–6.99</td>
<td>0.041</td>
<td>2.41</td>
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<td>Score (Logrank) Test</td>
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<td>0.9437</td>
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Multiple variables tested in separate runs when CIN4 was included (1–3a) or excluded (1–3b) from the comparison.

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Discussion

In this study, we developed a 4 gene-based measure of CIN applicable to FFPE material demonstrating clinical utility as a fairly robust marker of breast cancer prognosis. Aneuploidy determined by FACs-based DNA index correlated with the CIN4 signature, indicating that the CIN4 signature may in fact serve as a surrogate method to determine tumor aneuploidy status.

These data derived from a qPCR assessment of CIN4 mRNA expression from FFPE breast cancers, seems to confirm our previous analysis from microarray expression datasets [5], that CIN signature expression appears to be relatively enriched within ER negative breast cancers as well.

Directly or indirectly quantifying CIN in human tumor biopsies may hold significant potential for clinical practice [2,7,10,28,29]. The aim of the current work has been to establish such a gene expression based indirect measure and validate its utility in an already existing retrospective FFPE cohort. In multivariate regression analysis the CIN4 signature confers prognostic power in comparisons across grade 2 cancers specifically. ER status and vascular invasion showed predictive power, while the CIN4 score seems to outperform mitotic index and Ki67 expression in multivariate analysis.

The CIN4 signature performed comparably to previously published gene signatures in similar cohorts. For example, the previously published Genomic Grade Index was further simplified and converted into a qPCR-based test from formalin-fixed, paraffin-embedded tissues [13,20]. It examines the expression of 4 genes, CDC2, CDC20, KPNB2 and MYBL2 and its utility has been demonstrated to predict outcome following tamoxifen treatment. A 5-gene assay was also developed quantifying the expression of CTHDH, HOXB13, IL17BR, MIB1 and MKI67 to identify a subgroup of early stage estrogen receptor–positive breast cancer patients with very poor outcome despite endocrine therapy [19]. Considering the previously published correlation between Genomic Grade Index and CIN [1,20], it is reassuring to see that both CIN4 and the 4-gene Genomic Grade Index signature yield similar results [13,30], indicating that replacing histological grade 2 with gene expression based low and high risk cases could be considered in prospective studies.

Interestingly, in the study we identified a group of samples with a diploid DNA index which have a wide range of CIN scores. There might be several explanations behind this phenomenon including tumor heterogeneity or the presence of yet unknown other compensatory mechanisms [31]. Clearly, a substantial proportion of tumors with high CIN4 scores have a normal DNA ploidy, indicating that the CIN4 score likely reflects factors in addition to DNA copy number. In these cases, it is possible that highly proliferative cells, although with relatively stable genomes, may have high levels of the genes comprising the signature

While establishing a qPCR-based simple test to improve histological classification is an important clinical goal, – following further validation – the CIN4 signature as a quantifier of CIN is expected to serve as a potential marker of other clinical characteristics as well. Most prominently, the putative role of CIN in determining taxane response would suggest that the predictive role of CIN4 should be tested in the neoadjuvant setting when comparing response to therapy with or without taxanes in the treatment of hormone receptor negative breast cancer.

Supporting Information

Figure S1 Kaplan-Meier plot showing the disease-free survival of the different grade and immunophenotype groups in the validation cohort, respectively. (TIF)

Figure S2 In silico comparison of the performance of CIN4 vs. CIN25, CIN70, Ki67, genetic grade, HOXB13:IL17BR index, Genomic Grade Index, the 21-gene recurrence score and NKI70 in the GIS, JB1 and JB11 datasets [11,20,27] (A), and assessing the additive power of CIN4 (B). (PDF)

Figure S3 Kaplan-Meier curves of Ki67 and mitotic index performances in the tissue samples of grade 2 breast tumors. (TIF)

Figure S4 Individual prognostic performance of AURKA, FOXM1, TOP2A and TPX2 in breast cancer datasets as compared to CIN4 signature (groups split at median expression). (PDF)

Figure S5 Correlation of CIN4 and markers used in the determination of immunophenotype, CIN4 and A) ER, B) PR, and C) HER2 expression, and D) HER2/chromosome 17 score determined by FISH. (TIF)

Figure S6 CIN4 expression in ER-negative and ER-positive tumors. (TIFF)

Table S1 The public datasets analysed in the study and their supplementary references. (XLS)

Table S2 The probes used for the qPCR evaluation in the study. (XLS)

Table S3 Expression of control genes in the breast cancer tissue samples according to grade. (XLS)

Table S4 Performance of the individual components of CIN4 in tissue samples of grade 2 breast tumors. (XLS)

Table S5 Correlation of the clinicopathological/prognostic variables with each other in grade 2 breast tumours only. Pearson correlation coefficients displayed, significant correlation: gray highlight. (XLS)

Acknowledgments

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Author Contributions
Conceived and designed the experiments: AMS JK CS ZsS. Performed the experiments: AMS AE QL ZsS AT AR BS. Analyzed the data: AMS ZsS

References