# An update on the genomic landscape of breast cancer: new opportunity for personalized therapy?

## Anneleen Daemen

Department of Bioinformatics & Computational Biology, Genentech Inc., 1 DNA Way, South San Francisco, California 94080, USA *Corresponding to:* Anneleen Daemen. Department of Bioinformatics & Computational Biology, Genentech Inc., 1 DNA Way, South San Francisco, California 94080, USA. Email: daemena@gene.com.



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Breast cancer is a clinically and genomically heterogeneous disease, and may therefore benefit from personalized therapeutic strategies. Incorporating genomic information obtained by molecular analysis of an individual patient's tumor into treatment decisions is a promising area of focus. This approach requires knowledge of the genomic landscape and its possible therapeutic implications, but will ultimately allow employing investigational therapeutic compounds in a more intelligent way. Many institutes worldwide have contributed to the genomic landscaping of breast cancer, of which the results of six studies were recently published in Nature. These studies, summarized in Table 1, differ in breast cancer spectrum covered, primary focus, and technology platforms used. Integrating the genomic data from these studies, which is the focus of this review, provides an unprecedented opportunity to better understand the substantial variation in drug response and clinical outcome and to rethink treatment strategy.

Mutational profiling for the identification of driver mutations was assessed in five studies, with the number of somatic point mutations summarized in *Table 1*. Somatic putative driver substitutions and small insertions and deletions were identified in cancer genes previously implicated in breast cancer and in cancer genes involved in other cancer types (*Figure 1*). Of particular interest for cancer treatment are new genes that may be causally implicated in oncogenesis. Cancer genes not previously associated with breast cancer but confirmed through mutation recurrence screening are potential tumor suppressors ARID1B, CASP8, CBFB, CDKN1B, MAP3K13, NCOR1, RUNX1, SMARCD1, and TBX3. These genes with their mutation frequency per clinical and transcriptional subtype are shown in *Figure 1*. This heatmap confirms subtype-specific mutation patterns for genes such as *AKT1*, *BRCA1*, *GATA3*, *PIK3CA* and *TP53*, with the majority of basal-like and HER2-enriched samples carrying a *TP53* mutation, whilst luminal samples are more likely to carry a *PIK3CA* mutation. It also emerges from *Figure 1* that the mutation spectrum in luminal breast cancers is heterogeneous with mutations in the majority of highly recurrent genes. HER2-enriched and particularly basal-like breast cancers harbor less recurrently mutated genes. Notably, both Banerji *et al.* and the TCGA group observed mutation rates to be lowest in luminal A and highest in the basal-like and HER2-enriched subtypes (2,4).

A new recurrent observation across the studies is the inactivation of the RUNX1/CBFB complex reported to be essential for normal hematopoietic cell differentiation (7). CBFB encodes the non-DNA-binding component of the transcription factor complex, whilst RUNX1 is a transcription factor encoding the DNA-binding RUNX protein. Isolated cases of breast cancer with a somatic mutation in CBFB had been noted before, but not as frequently as in these recent studies with 15 tumor samples carrying a CBFB somatic mutation (of which 87% luminal/ER+). Interestingly, RUNX1 was mutated in a mutually exclusive fashion in an additional 21 samples (90% luminal/ER+), and showed homozygous deletion in 2 basal-like samples. With RUNX1 a tethering interaction partner of estrogen receptor  $\alpha$  (8), loss of transcriptional regulation by this complex is suggested to promote breast cancer progression.

Structural variation was focused on in three studies by sequencing at the whole-genome level (*Table 1*). Consistent with mutation rates, luminal A samples showed on average 30 rearrangements, basal-like samples 237 and HER2enriched samples 246 rearrangements (2). Within luminal

studies with sequencing data						
Study	Patient characteristics	Number of patients	Collected data	Subtype information	# somatic substitutions	# somatic indels
Sanger (1)	Primary carcinoma	100	WES [100]	54% ER+/HER2- 30% HER2+ 16% TN	6,964	277
MIT (2)	Primary, treatment- naïve carcinoma	108	WES [103] WGS [22]	35% luminal A 20% luminal B 20% Her2-enriched 12% basal-like 13% normal-like or unknown	4,668	317
WashU (3)	Pretreatment biop- sies from 2 neoadj. aromatase inhibitor trials	77	WES [31] WGS [46]	100% ER+	3,221	133
TCGA (4)	Invasive carcinoma	825	WES [507] Expr [547] Meth [802] SNP [773] miRNA Seq [697] RPPA [403]	44% luminal A 25% luminal B 11% HER2-enr 18% basal-like 2% normal-like	28,319	2,302
UBC (5)	Primary, treatment- naïve carcinoma	104	WES [54] WGS [15] RNA-seq [80] SNP [104]	100% TN	2,386	107
Cambridge (6)	Primary carcinoma	1,992	Expr [1,992] SNP [1,992]	36% luminal A 25% luminal B 12% HER2-enr 17% basal-like 10% normal-like	N/A	N/A

**Table 1** Overview of reviewed studies with patient and omics information. The number of somatic point mutations are shown for 5 out of 6 studies with sequencing data

Collected data: WES = whole exome sequencing; WGS = whole genome sequencing; Expr = expression microarray; Meth = methylation; SNP = single nucleotide polymorphism; RPPA = reverse-phase protein. Subtype information: ER = estrogen receptor; PR = progesterone receptor; TN = triple negative; N/A = not applicable

samples, the number of rearrangements was associated with treatment response, with fewer rearrangements observed in ER+ samples sensitive to aromatase inhibition (3). Although no new structural rearrangements were discovered, in-frame fusion events were revalidated, including the MAGI3-AKT3 fusion (2). It is anticipated that the latter fusion results in both loss of function of *PTEN* and activation of oncogene *AKT3*, making it targetable with an ATP-competitive AKT inhibitor. Screening for this fusion event in an additional 235 breast cancers revealed a rearrangement rate of 3.4%, among which 62% were triple negative (2). This reveals a new therapeutic option to investigate for fusion-positive triple negative breast cancer.

In the TCGA study, significant clusters obtained from

a variety of data types such as methylation, copy number and miRNA expression, showed moderate concordance with the transcriptional subtypes (4). Curtis *et al.* also revealed a steady increase in number of distinct subtypes, with substantial heterogeneity within the transcriptional subtypes (6). Ten clusters were obtained by joint clustering of copy number and gene expression data. These clusters showed distinct disease-specific survival, and might influence treatment strategies. For example, the HER2enriched cluster contained luminal cases that might benefit from HER2-specific targeted treatment. These associations with patient outcome, however, are not yet strong enough for clinical utility.

Finally, many infrequently mutated genes about which



Figure 1 Heatmap of mutation prevalence per subtype for the top 30 non-silent, recurrently mutated genes. Genes were selected based on a percentage per subtype  $\geq$ 5% for at least 1 study, and sum of subtype-specific percentages across all subtypes and studies  $\geq$ 15%

little is known together may substantially contribute to cancer, which significantly challenges therapy development. In the 100 sample cohort of Stephens et al., a small portion (17%) of cancer genes (TP53, PIK3CA, ERBB2, MYC, FGFR1/ZNF703, GATA3, and CCND1, each frequently mutated in over 10% of cases) contributed over half (58%) of all observed driver mutations. The remaining 42% of driver mutations occurred in 83% of cancer genes, each contributing infrequently (1). Similar observations were made in two of the other studies, with somatic mutations in 21% and 12% of cases restricted to genes mutated at an insignificant frequency, many of which were part of key regulatory pathways characterized by highly recurrent cancer genes (3,5). In order for patients without recurrent abnormalities to benefit from targeted treatment, a pathwaydriven approach that takes rare and tumor-specific changes into account will become a necessity. Shah et al. analyzed sparse mutation patterns in functionally connected genes (5). This approach revealed significantly mutated pathways for triple negative breast cancer such as p53-related pathways, ERBB signaling, and WNT/cadherin signaling. Targeting dysregulated pathways regardless of the mechanism by which

they are dysregulated will potentially yield more and better treatment options for patients without somatic aberrations in the frequent drivers. The TCGA group and Ellis *et al.* made suggestions of drug targets based on mutation and genomic aberration spectra, with a target defined as a gene or protein with an approved drug or investigational drug in late stage development targeting the molecular pathway. These suggestions include PIK3CA, AKT1 and KIT inhibitors for luminal/ER+ cancers, PARP and DDR2 inhibitors for basallike cancers, and combined inhibitors of *HER2*, *HER3* and/ or *EGFR* for HER2-enriched cancers (3,4). It is therefore plausible that patients with chromosomal aberrations or carrying mutations in the targeted pathways might benefit from those respective inhibitors.

To move this information into routine clinical practice, it will be important to further exploit the clinical heterogeneity and new drug targets for improved treatment of breast cancer patients. Genomics can direct towards dysregulated signaling pathways or molecular functions instead of individual mutations, and therapeutic combinations can potentially reduce acquired resistance. To accelerate clinical investigation, *in vitro* cell line systems on one hand will be key to associate mutation patterns and pathway activity with response to investigational compounds (Daemen A *et al.*, submitted). On the other hand, a model for rapid assessment of phase II drugs is required. The neoadjuvant I-SPY2 trial for locally advanced breast cancer was set up according to an adaptive trial scheme to allow for concurrent drug testing using fewer patients and resources: compounds with a high probability of being more effective than standard therapy in a certain subpopulation graduate from the trial towards smaller, population-targeted phase III trials, whilst compounds with a low probability of improved efficacy are dropped (9). This design allows the simultaneous selection of biomarkers to guide patient selection, avoiding drug failure in an all-comers trial due to low aberrant pathway prevalence.

As revealed by the studies in *Table 1*, breast cancer heterogeneity cannot be fully addressed with current knowledge and standard technologies. Additional means of profiling such as epigenetics and metabolomics could aid in better defining key pathways and drivers, and add to future therapeutic selection (10). Sequencing, expression, copy number and other new technologies in combination with a pathway-targeted therapeutic approach poise the field to identify new prognostic and predictive biomarkers, and discover additional new targets and pathways for therapeutic intervention.

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