Association of *ESR1* germline variants with *TP53* somatic variants in breast tumors in a genome-wide study

Running title: Germline associations with somatic mutations

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ABSTRACT

Background: In breast tumors, somatic mutation frequencies in *TP53* and *PIK3CA* vary by tumor subtype and ancestry. HER2 positive and triple negative breast cancers (TNBC) have a higher frequency of *TP53* somatic mutations than other subtypes. *PIK3CA* mutations are more frequently observed in hormone receptor positive tumors. Emerging data suggest tumor mutation status is associated with germline variants and genetic ancestry. We aimed to identify germline variants that are associated with somatic *TP53* or *PIK3CA* mutation status in breast tumors.

Methods: A genome-wide association study was conducted using breast cancer mutation status of *TP53* and *PIK3CA* and functional mutation categories including *TP53* gain of function (GOF) and loss of function mutations and *PIK3CA* activating/hotspot mutations. The discovery analysis consisted of 2850 European ancestry women from three datasets. Germline variants showing evidence of association with somatic mutations were selected for validation analyses based on predicted function, allele frequency, and proximity to known cancer genes or risk loci. Candidate variants were assessed for association with mutation status in a multi-ancestry validation study, a Malaysian study, and a study of African American/Black women with TNBC.

Results: The discovery Germline x Mutation (GxM) association study found five variants associated with one or more TP53 phenotypes with P values $<1x10^{-6}$, 33 variants associated with one or more TP53 phenotypes with P values $<1x10^{-5}$, and 44 variants associated with one or more PIK3CA phenotypes with P values $<1x10^{-5}$. In the multi-ancestry and Malaysian validation studies, germline ESR1 locus variant, rs9383938, was associated with the presence of TP53 mutations overall (P values $6.8x10^{-5}$ and $9.8x10^{-8}$, respectively) and TP53 GOF mutations (P value $8.4x10^{-6}$). Multiple variants showed

suggestive evidence of association with *PIK3CA* mutation status in the validation studies, but none were significant after correction for multiple comparisons.

Conclusions: We found evidence that germline variants were associated with *TP53* and *PIK3CA* mutation status in breast cancers. Variants near the estrogen receptor alpha gene, *ESR1*, were significantly associated with overall *TP53* mutations and GOF mutations. Larger multi-ancestry studies are needed to confirm these findings and determine if these variants contribute to ancestry-specific differences in mutation frequency.

KEY WORDS

TP53, PIK3CA, breast cancer, GWAS, germline, somatic, germline by somatic interaction, ESR1

BACKGROUND

TP53 and PIK3CA are among the most frequently mutated genes in breast tumors [1]. The frequency of somatic mutations in these genes varies by tumor subtype as well as ancestry [1-4]. Pan-cancer and breast cancer specific studies have found that tumors arising in individuals of African ancestry (AFA), particularly West African ancestry, are more likely to have somatic TP53 mutations and less likely to have somatic *PIK3CA* mutations than tumors arising in individuals of European ancestry (EUR) [2-9]. TP53 somatic mutations are more common in triple negative [estrogen receptor (ER) negative, progesterone receptor (PR) negative, human epidermal growth factor receptor 2 (HER2) negative] breast cancers (TNBC), while PIK3CA mutations are more common in hormone receptor (HR) positive HER2tumors. However, even after adjusting for breast cancer subtype, ancestral differences in TP53 and PIK3CA somatic mutation frequencies persist for some subtypes [2-4,10]. For example, one study found that 39% of hormone receptor (HR) positive HER2- tumors from individuals of AFA had TP53 alterations compared to 24% of those of EUR [8]. Similarly, in HR+ HER2- tumors PIK3CA somatic mutations are less frequent in individuals of AFA (26%) versus EUR (42%) [8]. The biological mechanisms leading to the observed differences in TP53 and PIK3CA somatic mutation frequency across populations and breast tumor subtypes are not understood.

TP53 encodes transcription factor TP53 and is mutated in a high proportion of breast and other cancers, resulting in altered expression of genes important for response to cellular stress and apoptosis. Unlike many genes involved in tumorigenesis, TP53 can have either loss of function (LOF) mutations, which lead to total loss of the ability of the protein to transactivate, or gain of function (GOF) mutations, which result in TP53 binding to new promoters to activate genes not typically associated with TP53 [11,12]. TP53 is a tetramer but can also bind to related proteins TP63 and TP73 [13]. Some TP53 tumor-

associated mutations act in a dominant negative manner where the mutant version of the protein interferes with the function of wildtype proteins in the tetramer. In previous studies, we found that in breast tumors with *TP53* mutations, those from AFA women were less likely to have GOF mutations than those from EUR women [14]. Mutations without dominant negative activity were associated with TNBC and ER negative status. These data suggest that types of *TP53* mutations in breast tumors differ by self-reported race and tumor subtypes which may be due to different functional consequences of these mutations within cells.

Whilst most somatic events in tumors are likely due to exogenous or endogenous mutators, recent evidence suggests that germline variants may influence the type and burden of somatic changes. Tumor mutational burden, caused in part by somatic mutations in DNA repair genes, is a polygenic trait with an estimated 13% of the variation explained by common germline variants [15]. Some tumor mutational signatures are associated with common inherited variants in genes such the apolipoprotein B mRNA editing enzyme catalytic polypeptide (APOBEC) mutation signature and variants in *GNB5* [16]. Pathogenic variants (PV) in high-risk cancer susceptibility genes also associate with the presence of somatic mutations and specific mutational signatures. Breast tumors arising in individuals with a germline *BRCA1* PV have more frequent occurrence of somatic *TP53* mutations compared to those without a *BRCA1* PV [17-20]. Breast and ovarian tumors arising in individuals with *BRCA1* and *BRCA2* PVs typically show homology directed repair deficiency signatures [20,21].

Based on these studies, we hypothesized that the germline genetic background of an individual can influence specific mutational processes, tumor promotion, and/or mutations in specific cancer-related

genes during tumorigenesis, any of which could lead to the observed differences in the frequency of key cancer driver mutations by ancestry [22]. The goal of this study was to identify inherited common germline variants (G) that are associated with *TP53* or *PIK3CA* somatic mutation status (M) in breast tumors using a Germline Variant by Mutation (GxM) genome-wide association study (GWAS) design to assess the influence of genetic background on mutation frequency of these genes.

METHODS

Discovery Breast Cancer Datasets

Existing datasets of women with breast cancer from the Cancer Genome Atlas (TCGA) [1], Molecular Taxonomy of Breast Cancer International Consortium (METABRIC) [23], and the Welcome Trust Sanger Institute [24] were used for the discovery GxM GWAS. Each study had existing genome-level single nucleotide variant (SNV) genotyping data, somatic mutation data for *TP53* and *PIK3CA*, and associated clinical and tumor details such as self-reported race/ethnicity, age at diagnosis (reported as decade of diagnosis), and ER, PR and HER2 status (Supplemental Information Tables 2-4).

PIK3CA and TP53 Somatic Mutation Classification

For discovery and validation analyses, *PIK3CA* mutation status (yes or no) was defined for the following phenotypes: any non-loss of function somatic variant in *PIK3CA*, any activating/hot-spot mutation [25], or specific activating mutations (e.g. p.E542K, p.E545K, p.H1047R/L). *TP53* mutation status was classified as the presence of any somatic variant in a codon exon or splice-site (yes/no), and variants resulting in *TP53* LOF or GOF as described (Supplemental Information Table S5) [14,26]. Somatic

variants that resulted in a synonymous change and were not predicted to affect splicing were not considered to be a mutation. GOF mutations displayed one or more of the following phenotypes in functional studies: interference with TP63 or TP73 activity, transactivation of genes repressed by wildtype TP53, or cooperation with oncogenes in rat or mouse embryonic fibroblasts [27-61]. TP53 LOF variants were those that abolished transactivation activity and/or resulted in altered splicing, frameshift, or nonsense changes. TP53 somatic missense variants with insufficient data to functionally score as LOF or GOF were called unknown and were not included in LOF or GOF specific analyses (Supplemental Information Table S5). Larger copy number loss of TP53 was not included as a mutation type in the analyses due to lack of annotated data for multiple datasets. Controls for each analysis were individuals with breast cancer with no somatic mutation in the gene being assessed.

Ancestry and GWAS analyses

PLINK was used to merge data sets, filter, and analyze data. Ancestry SNVs for principal component analyses (PCA) were determined using the Affymetrix annotation accomplished by subtracting the minor allele frequency (MAF) from each of four populations [Han Chinese in Beijing (CHB), Yoruba in Ibadan (YRI), Northern Europeans from Utah (CEU), Japanese in Tokyo (JPT)] in the annotation in a pairwise manner and taking the top 1000 SNVs from each comparison. This resulted in the use of 4486 unique "ancestry" SNVs; 4212 of those had a MAF of greater than 1%. PCA were performed on these 4212 SNVs to identify individuals of non-European ancestry (e.g. those who did not cluster with the European ancestry group); these individuals were removed from the discovery analyses and were included in the validation studies. There was a high concordance of ancestry assignment with self-reported race. Filtering also included removal of SNVs with MAF less than 0.01, male participants, and samples and SNVs with greater than 10% missing values. Imputed SNVs were not included. SNVs

showing Hardy Weinberg equilibrium P values less than $1x10^{-50}$ were also removed. Association analyses were run on a final set of 2850 females of European ancestry and 739,537 SNVs with PLINK using a logistic model with a covariate for study. An additive model was assumed. P values were false discovery rate (FDR) corrected and visualized using R [62].

Selection of variants for validation analyses.

Variants were prioritized for validation studies through multiple qualitative and quantitative filtering steps (Supplemental Information Tables S6 and S7). Information used to rank variants included P values <1x10⁻⁴, odds ratios (OR), MAF greater than 10% in at least one of three populations (European, African or East Asian), allele frequency differences by ancestry, proximity to a variant identified in GWAS for breast cancer risk or other relevant phenotypes (e.g. other cancers, age at menarche, obesity), proximity to a gene showing a role in tumorigenesis, and mapping to a functionally active region (e.g. transcription start site, active chromatin markers, estimated or actual transcription factor binding site, disruption of a transcription factor binding motif, ChIP-seq region for breast cancer cell line, characterized gene enhancer, characterized promoter region, or expression quantitative trait locus). Online resources used for in silico screening of candidate SNVs included UCSC Genome Browser (https://genome.ucsc.edu/), GTEx Portal (https://gtexportal.org/home/), RegulomeDB v.2.0.3 (https://www.regulomedb.org/regulome-search/), HaploReg v4.0 (https://pubs.broadinstitute.org/mammals/haploreg/haploreg.php), and dbSNP (https://www.ncbi.nlm.nih.gov/snp/). In addition to variants chosen from the discovery GxM GWAS findings, additional variants were analyzed including two SNVs mapping near SETD9/MAP3K1 previously shown to be associated with PIK3CA somatic mutations in breast cancer [63], an XPC variant rs2228001 previously shown to be associated with TP53 mutation status [64], and a variant in AURKA

(rs2273535) shown in our unpublished studies to be associated with somatic GOF *TP53* mutations in human papilloma virus negative head and neck squamous cell carcinomas from TCGA. When a genotyping assay for a variant could not be designed for technological reasons, another variant from that locus or a variant in high linkage disequilibrium (LD) (r²>0.8) with the original variant was included as a replacement.

Validation Genotyping

Validation genotyping for 188 SNVs of interest (95 for *TP53* and 93 for *PIK3CA*) was completed for cohorts without existing genome-wide genotyping data including individuals from the Stefanie Spielman Breast Cancer Cohort (n=144), OSU Total Cancer Care (TCC) (n=351) and the City of Hope (COH) Latina Breast Cancer Study (n=120) using a Fluidigm HD Biomark in a 96x96 format in the Ohio State University (OSU) Comprehensive Cancer Center (CCC) Genomics Shared Resource (GSR) (Supplemental Tables S8, S9). Each genotyping plate contained two duplicate DNA samples, three notemplate controls (water), and one control DNA sample genotyped on all plates. DNAs that failed for more than 10% of SNVs from a plate were repeated and if failed again were removed from analysis. SNVs that failed for more than 10% of samples or failed to consistently form three clear genotyping groups were removed from analyses. For genetic ancestry, 96 SNVs were chosen for genotyping from existing ancestry informative marker (AIM) panels [65-67] (supplemental table S10-S12). Of the 96 AIM SNVs, two were removed for poor genotyping performance.

Somatic Mutational Analyses

For the validation studies, *TP53* and *PIK3CA* mutational status from the clinical testing reports or targeted or exome sequencing of tumor DNA was available for breast cancer cases from the COH Latina Breast Cancer Study, TCGA, and a subset of the TCC cases. For cases in which mutation status was not known, tumor tissue or DNA was available from the Spielman Breast Cancer Cohort and the TCC program.

Next-generation sequencing mutational analysis. An Agilent custom-targeted sequencing panel that included *TP53* and *PIK3CA* was designed for an unrelated project. Tumor DNA samples isolated from frozen tumors from TCC cases (n=12) were assessed for somatic mutations in *TP53* and *PIK3CA* using this panel. Libraries were generated according to Agilent protocols and sequenced on a HiSeq4000 by the OSU CCC GSR. Sequencing run data was received in the form of paired FASTQ files (forward and reverse reads). FASTQ file pairs were aligned to the hg19 reference genome using Genomic short-read nucleotide alignment program (GSNAP) (version 2019-03-04)[68]. Variants were called with Pisces (version 5.2.10.49) using the hg19 reference genome [69]. At this stage, low depth (<100 read count) variants were removed. Finally, both SnpSift (version 4.3t) and Annovar (version 20190316) were used to annotate called variants [70,71].

Sanger sequencing mutational analysis: The remainder of tumor samples lacking existing somatic mutation data (n=126 for *TP53*, n=184 for *PIK3CA*) were screened for somatic mutations in *TP53* coding exons (exons 2-10) and *PIK3CA* exons 4, 9 and 20 using Sanger sequencing. Tumor DNA (10-20ng) was PCR amplified and products were confirmed for size by gel electrophoresis (Supplemental Table S13). PCR products were Exo/SAP-IT treated and Sanger sequenced in both forward and reverse directions by the GSR. Sequence chromatograms were evaluated for mutations using DNASTAR Lasergene v.17 by two different laboratory members.

GxM Validation Analyses

Data used for validation of key findings included genotype and tumor mutational data from individuals of non-European ancestry from the three discovery datasets as well as samples (germline and/or tumor DNA) or existing data from 1285 individuals of multiple ancestries from the METABRIC (n=166), Stefanie Spielman Breast Cancer Cohort (n=144), OSU TCC (n=352), a Nigerian breast cancer study (n=100), the COH Latina Breast Cancer Study (n=120), TCGA (n=302), and a TCGA study ("Banerji study") of women from Mexico and Vietnam (n=101) (Supplemental Tables S2-S4, S14-S18)[3,4,72,73]. Genetic ancestry by PCA classified 341 women as AFA (26.5%), 572 women as EUR (44.5%), and 133 women as East Asian ancestry (EAS) (10.4%). The remainder of women (18.6%) were admixed (falling between PC clusters), most of whom self-identified as Hispanic/Latino. Due to some missing genotypes, not every variant had data for all 1285 individuals (Supplemental Table S26-S27).

For association analyses, logistic models were employed with an additive effect for the SNV. Study and ancestry were included as covariates in the models. For the study and ancestry-specific analyses, the study analysis omitted the effect of study, and the ancestry analyses omitted the ancestry principal component (PC) from the models. Because two different panels were used for ancestry determination, individuals of known ancestry (HapMap, TCGA) were used as anchors for each panel. The PC1/PC2 were rotated so that the known ancestry groups overlapped and the distance from the anchor group was calculated as the PC covariables. For individuals with available genome-wide genotyping data, imputation of validation SNVs not present on the GWAS genotyping panels was performed. Imputation was carried out after removing genotypes with no calls or Y chromosome calls. Eagle was used to phase SNV, and imputation was done using Minimac3. The maximum expected error rate across imputed

validation SNPs was 0.086. Formats were converted to PLINK format, and variants with greater than 2 alleles were removed.

Independent Validation Studies

SNVs of interest also were assessed independently in two cohorts: 859 women with breast cancer from the Malaysian Breast Cancer Study (MyBrCa) [74,75] and 393 AFA women with TNBC from the Breast Cancer in African Americans: Understanding Somatic Mutations and Etiology (B-CAUSE) study (Supplemental Tables S19 and S20) [76]. Validation SNVs for the MyBrCa study were excluded from analyses if they mapped to the X chromosome or if they had a MAF less than 1% in Malaysian individuals. SNVs were excluded from analyses for the B-CAUSE study if they mapped to the X chromosome. For the MyBrCa study association tests were conducted using SNPtest adjusted to information for ancestry [4 PCs], age of diagnosis, and ER status. B-CAUSE data came from women who self-identified as Black and were diagnosed with TNBC. The African-ancestry Breast Cancer Genetic (AABCG) is a large breast cancer consortium which provided genome-wide genotyping data for the B-CAUSE study. AFA ancestry was confirmed by estimating global African ancestry using ADMIXTURE [77] (Supplemental Tables S20). As the frequency of somatic TP53 mutations in the B-CAUSE TNBC cases was high, analyses were run for TP53 GOF-associated germline variants using individuals with LOF TP53 mutations and those with no mutations as controls; conversely for TP53 LOF-associated variants, analyses were run using individuals with GOF plus those with no mutations as controls.

RESULTS

To identify germline variants associated with *TP53* or *PIK3CA* somatic mutations in tumors, we identified existing datasets with GWAS-level germline variant information, somatic mutation information for *TP53* and *PIK3CA*, and demographic and clinical information such as age of diagnosis, tumor subtype defined by hormonal (ER and PR) status and HER2 amplification. Three datasets were identified that fit these criteria (Supplemental Tables S2-S4). After filtering for SNVs with MAF less than 1%, individuals with 10% or higher SNV genotypes missing, SNVs out of Hardy-Weinberg equilibrium (p<1x10⁻⁵⁰) and individuals of non-European ancestry, 2850 females of EUR with breast cancer and 739,537 SNVs were included in the discovery GWAS for variants associated with *TP53* and *PIK3CA* mutation status.

Discovery GxM for TP53 mutation status

Somatic mutations in TP53 (237 GOF, 536 LOF, 106 unknown) were identified in 879 of the 2850 women (30.8%) (Supplemental Tables S2-S5). Analyses for association with any TP53 mutation, GOF TP53 mutation, and LOF TP53 mutation were performed. Following analysis, no SNV met the genomewide statistical significance threshold of a P value $<1.0\times10^{-8}$; four variants were identified with P values $\le 1.0\times10^{-6}$ and 34 variants had P values less than $\le 1.0\times10^{-5}$ across 22 loci (Figure 1A, 1B and 1C; Supplemental Table S21 and S22). Two variants showed P values of $<1.0\times10^{-5}$ for more than one TP53 phenotypic comparison: rs1561072 for any TP53 mutation and GOF TP53 mutations and rs2886631 for any TP53 mutation and LOF TP53 mutations.

Discovery GxM for PIK3CA mutation status

Thirty-eight percent (n=1095) of breast tumors from the 2850 women included in the discovery analyses had a somatic PIK3CA mutation. Of these, 858 tumors had a mutation known to functionally activate PIK3CA encoded protein PIK, including 112 with p.E542K, 193 with p.E545K, and 387 with p.H1047R/L alterations. Following association analyses for PIK3CA mutation status, no SNV met genome-wide significance of P value of <1.0x10⁻⁸ (Figure 1B). Forty-four SNVs were associated with one or more PIK3CA mutation phenotypes with P value <1x10⁻⁵ (Figure 1B; Supplemental Table S23 and S24). Of these, rs2026801 showed evidence of association (P value <1x10⁻⁵) for any PIK3CA mutation and activating PIK3CA mutations, and rs1712829 showed evidence of association with both p.H1047R and any PIK3CA mutation.

Selection of variants for validation studies

Using *in silico* filtering approaches, all variants with *P* values < 1x10⁻⁴ for any phenotype were evaluated for potential inclusion in validation studies. Variants were prioritized for further evaluation by allele frequency (MAF>10%), potential function using *in silico* prediction models, location near a known GWAS hit for breast cancer or related phenotype (e.g. age of menarche, obesity), location near a gene involved in tumor development, or known relationship to PI3K or TP53 pathways (Supplemental Tables S6 and S17). Of these, 188 variants from *TP53* (n=95) and *PIK3CA* (n=93) GxM analyses were chosen for validation studies and successfully genotyped in multi-ancestral populations (Supplemental Table S25). For individuals with GWAS-level genotyping data, 119 variants for *TP53* and 106 variants for *PIK3CA* were tested (Supplemental Tables S26-S31).

Mutation Status and Ancestry in Validation Populations

In the multi-ancestry validation data set, ancestry classifications by PCA yielded 340 AFA individuals, 602 EUR individuals and 134 EAS individuals. The remainder of study individuals (n=209) were considered admixed and not assigned to a specific group; these included individuals of Hispanic/Latino background who demonstrated a high degree of admixture. In the multi-ancestry validation cohort, 419 of 1036 individuals (40%) had a *TP53* somatic mutation of which 110 (10.6%) were GOF and 277 (27%) were LOF. The overall frequency of *PIK3CA* mutations in the multi-ancestry validation cohort was 28% (290 of 1036) with 235 of those being activating (81% of all mutations and 23% of the total). For specific *PIK3CA* activating mutations, 40/1035 (3.9%) of tumors had a p.E542K mutation, 58/1035 (5.6%) had a p.E545K mutation, and 133/1035 (13%) had a p.H1047R/L mutation.

The MyBrCa study included 859 women from Malaysia with breast cancer, of whom 369 carried a *TP53* mutation (43%). Of the MyBrCa women with a *TP53* mutation, 114 had a GOF mutation (13%), 241 (28%) had LOF mutations, and the remainder (n=14) were of unknown GOF/LOF status. For *PIK3CA* mutation status, 247 of the MyBrCa women (29%) had *PIK3CA* mutations in their breast tumor; of these, 217 (25% of total; 88% of all with a mutation), were activating. For the specific *PIK3CA* mutations, 20 had a p.E542K mutation (2% of total), 55 had a p.E545K mutation (6% of total) and 115 (13% of total) had a p.H047R/L mutation. Of the 393 women of African ancestry with TNBC in the B-CAUSE study, 365 (93%) had a TP53 somatic mutation including 260 (66%) with LOF mutations and 85 (22%) with GOF mutations. Only nine (2.3%) of tumors from women in the B-CAUSE study had any *PIK3CA* somatic mutations, of which four were activating/hotspot mutations.

Association of variants at the ESR1 locus and TP53 mutation status

In the multi-ancestry validation study, rs9383938, which maps to chromosome 6 at the *ESR1* locus, was significantly associated with having any *TP53* mutation (OR 1.46, *P* value 6.8x10⁻⁵) after correction for multiple comparisons of 119 variants/*TP53* phenotypes (<0.00042). In the MyBrCa study, rs9383938 also showed association with having a *TP53* mutation (OR=1.81; *P* value 9.8x10⁻⁸) (Table 1) (Supplemental Tables S27 and S32) and *TP53* GOF mutation status (*P* value 8.4x10⁻⁶). Another *ESR1* locus variant, rs9479090, was associated with *TP53* mutations (*P* value 2.8x10⁻⁷) in the MyBrCa Study.

After correcting for multiple comparisons, no other variants were significantly associated with any *TP53* phenotype in the validation analyses, MyBrCA, or B-CAUSE studies. However, multiple variants showed nominal *P* values (<0.05), including five in the multi-ancestry validation cohort (rs10931697, rs4767726, rs12470238, rs9911226, rs17103093), three in the MyBrCa Study (rs6890674, rs6709393, rs4254949), and four in the B-CAUSE study (rs7633912, rs9926714, rs28547342, rs12445424) (Table 1; Supplemental Tables S26-S28, S32-S33). Variants showing a trend for association in more than one dataset included rs10931697 in the multi-ancestry validation and MyBrCa studies (*P* values 0.001 and 0.09), and rs6709393 (*P* values 0.003 and 0.16) in the MyBrCa and B-CAUSE studies.

Association of germline variants and PIK3CA mutation status

After correction for multiple comparisons for 106 variants/phenotype (*P* value <0.00047), no SNVs were significantly associated with any *PIK3CA* phenotype in the validation set, MyBrCA study (*P* value <0.00048) or B-CAUSE study (*P* values <0.0021) (Table 2) (Supplemental Tables S29-S31, S34-S35). Four variants showed nominal association (*P* value <0.05) for a *PIK3CA* mutation phenotype in the validation data set (rs6955337, rs10975835, rs13230836 and rs1886675) and nine variants in the

MyBrCa study (rs2026801, rs1051894, rs8084310, rs2332431, rs479995, rs194537, rs859074, rs9791638, rs2127064) (Table 2). Three variants with *P* values for association with *PIK3CA* mutation status of less than 0.05 were observed in the B-CAUSE study (rs893344, rs252913, rs331499), but these data should be interpreted with caution due to the very small numbers of *PIK3CA* mutation carriers. Variants showing a trend towards association in the multi-ancestry validation study and the MyBrCa study included rs2026801 with activating mutations (OR 0.88, *P* value 0.16 and OR 0.74, *P* value 0.009, respectively) and rs24420246 with p.E542K mutations (OR=1.64, *P* value 0.07 and OR=1.92, *P* value 0.14).

Association with African Ancestry

As *TP53* mutation frequency is generally higher in breast tumors arising in individuals of AFA compared to other populations, we tested if any of the variants in our validation analyses showed stronger evidence of association in this population. Variants rs16951139, rs10514489 and rs10931697 showed nominal evidence of association with *TP53* GOF mutation status in AFA breast cancer cases from the multi-ancestry study (*P* values 0.018, 0.022 and 0.022, respectively) (Supplemental Table S28). rs10514489 also showed evidence of association with *TP53* GOF mutation status in individuals of EUR (*P* value 0.008) and rs10931697 showed a non-significant trend for association with GOF *TP53* status in the MyBrCa study (*P* value 0.09). In the B-CAUSE study, none of these variants showed evidence of association. However, four variants (rs7633912, rs9926714, rs28547342, and rs12445424) in the B-CAUSE study showed nominal evidence of association with one or more *TP53* phenotypes (*P* value < 0.05) (Supplemental Table S33).

PIK3CA mutation status and genotype data were available for 127 to 338 females of AFA in the multi-ancestry study depending on the variant. Three variants showing evidence of association with a PIK3CA phenotype in the discovery analysis showed nominal evidence of association in AFA women in the multi-ancestry study. These included rs2586532 (p.E542K, P value 0.005), rs4322362 (p.H1047R/L, P value 0.022) and rs3812471 (any PIK3CA mutation; P value 0.026) (Supplemental Table S31). None of these variants showed evidence of association in other populations or in the combined analysis. In the B-CAUSE study, rs893344 showed evidence of association for any PIK3CA somatic mutation (P value 0.044) and activating PIK3CA mutations (P value 0.09). While this variant did not show evidence of association in the multi-ancestry or MyBrCa studies, there was a trend towards association when only the AFA women in the multi-ancestry study were considered (PIK3CA any mutation P value 0.09 and activating mutation P value 0.13) (Supplemental Table S35).

DISCUSSION

To our knowledge, this is the first genome-wide breast cancer specific study to identify germline variants that are associated with TP53 or PIK3CA somatic mutation status. As different types of mutations may have differential effects on cancer-related phenotypes, we also tested for association of specific subcategories of TP53 (any, LOF, GOF) and PIK3CA (any, activating, specific site) mutations with common SNVs. Five variants from the discovery analyses of women of EUR showed suggestive evidence (P value $< 1 \times 10^{-6}$) for association with TP53 mutation status. Analyses of candidate variants in a multi-ancestry cohort and a Malaysian study, MyBrCa, confirmed that variants at the ESR1 locus were associated with multiple TP53 mutation classifications (any, GOF) and remained significant after corrections of multiple comparisons.

ESR1 locus variants and association with TP53 mutation status

We found evidence that multiple ESR1 locus variants were associated with TP53 mutation status. In our discovery study, ten variants showed a trend towards association (P value <1x10⁻⁴) for one or more of the three TP53 mutation phenotypes. From breast cancer GWAS, multiple variants near ESR1 have been associated with breast cancer of all subtypes as well as ER- tumors [76,78-81]. Some variants at the ESR1 locus have been reported to exhibit ancestry-specific association with breast cancer risk [81-83]. For example, ESR1 variant rs140068132 which is thought to have originated in Indigenous Americans, is protective for breast cancer risk [83]. In ER- breast tumors, TP53 and ESR1 mutations tend to be mutually exclusive [84]. This may be due in part to the regulatory relationship between TP53 and ESR1. Mutant TP53 is correlated with lower ESR1 gene expression which is thought to be due in part to TP53 binding to the ESR1 promoter to activate expression [85]. Mutant TP53 tumors have lower estrogen response signatures compared to TP53 wildtype tumors which may be caused by both decreased transcriptional activation of ESR1 by mutant TP53 and increased levels of ESR1-targeting miRNAs [84]. These studies suggest the possibility that mutation of TP53 may be an early event that promotes lineage towards ER- breast tumors; it is possible that variants at the ESR1 locus may enhance or reverse this association. Further functional studies are warranted to understand the connection between ESR1 variants, TP53 mutational status, and breast cancer subtypes.

Variants associated with TP53 mutation status

Other variants of interest that showed evidence of association with *TP53* included rs17103093 which was associated with any *TP53* mutation phenotype (discovery OR 1.54, *P* value 3.3x10⁻⁵ and multi-ancestry validation OR 1.4, *P* value 0.03) and maps to an intron of *TACC2*. This variant did not show evidence of association with *TP53* mutations in the MyBrCa study. *TACC2* encodes one of three homologous coiled-coiled proteins (TACC1, TACC2 and TACC3); it shows increased expression in higher grade breast tumors and is associated with local recurrence and reduced survival [86,87]. Variants at the *TACC2* locus are associated with risk of low-grade breast cancer, overall breast cancer, and epithelial ovarian cancer [81,88,89]. Two variants, rs6703393 and rs6890674, showed consistent direction of association for *TP53* GOF mutations in the discovery analyses (OR 0.79, *P* value 7.5x10-5) and the MyBrCa study (OR 0.28, *P* value 0.003) but had no evidence of association in the multi-ancestry validation study (*P* values 0.99 and 0.83 respectively). rs6709393 maps near the *RAB17* gene which encodes for a small GTPase associated with invasion [90]. rs6890674 is located in the 3'UTR of *CD180*, an orphan toll-like receptor that is expressed on B cells and is involved in inflammatory and autoimmune diseases [91].

rs10931697, located near *SLC39A10*, showed opposite effects by study, as it demonstrated increased risk for GOF *TP53* status for the C allele in the discovery study (OR 1.82, *P* value 2.79x10⁻⁵) but protective effects in the multi-ancestry validation (OR 0.53, *P* value 0.001) and MyBrCa studies (OR 0.72, *P* value 0.09). Further studies are needed to determine if findings represent a false association or if there are ancestry-related differences driving the findings. SLC39A10 is a zinc transporter whose expression is associated with breast cancer metastasis and increased migration *in vitro* [92]. Another variant showing inconsistent directionality of association with a *TP53* mutation phenotype included rs4767726 that maps within a long non-coding RNA (*LNC-RNA 02439*) of which little is known. The C allele of rs4767726

was associated with reduced likelihood of having a *TP53* GOF mutation in the discovery dataset (OR 0.67, *P* value 0.0002) but increased likelihood for a GOF mutation in the multi-ancestry datasets (OR 1.49, *P* value 0.009).

Germline variants associated with PIK3CA mutations and impact on the PI3K/AKT pathway

A variant near *LPAR1*, rs2026801, showed evidence of association in the discovery dataset (*P* value <1x10⁻⁵) for both any *PIK3CA* mutation and activating *PIK3CA* mutation phenotypes and was the top variant associated with activating *PIK3CA* mutations in the MyBrCA study (G allele OR 0.5, 95% CI 0.38-0.65; *P* value 0.009). However, it showed only a trend towards association in the multi-ancestry analysis (OR 0.88, *P* value 0.16). This variant maps near multiple GWAS hits for hip-to-waist ratio [93] and birth weight [94-96]; both of which are phenotypes associated with increased breast cancer risk [97-99] but has not been associated with risk in women of European, African, or East Asian descent [100]. The nearest gene to rs2026801, *LPAR1*, encodes the lysophosphatidic acid receptor which activates the PI3K/AKT pathway and whose expression is elevated in ER+HER- and lower stage breast cancers [101-103]. In the GTEx database, rs2026801 maps to an expression quantitative trait locus (eQTL) for *LPAR1* in venous blood with the G allele showing lower expression (*P* value 3.93x10⁻¹¹). These data suggest that lower expression of *LPAR1* may provide a cellular environment less likely to require activating *PIK3CA* somatic mutations during tumor development. Further studies are needed to evaluate this hypothesis.

An intronic variant in KDM4C, rs10975835, showed evidence of association with PIK3CA H1047R/L mutations in the discovery set (OR 1.54, P value 1.1x10⁻⁵) and in the multi-ancestry validation study (OR 1.5, P value 0.011), but not in the MyBrCa study (OR 0.92, P value 0.66). KDM4C is a lysine

demethylase that removes a methylation mark from histones, specifically converting H3K9Me3, a repressive histone mark, to H3K9Me2, a more active mark [104]. Interestingly, a coding variant in *KDM4C*, rs2296067, renders the protein resistant to caspase-3 cleavage and is associated with worse outcomes in breast cancer [105]. *KDM4C* expression is increased in TNBC in which it is thought to be important in chromosomal stability and proliferation [105,106]. In prostate cancer cells, knock-down of *KDM4C* leads to decreased activation of the AKT pathway [107]. KDM4C acts as a lysine demethylase for the TP53 protein leading to decreased TP53 activity and apoptosis [108]. It is unknown if rs10975835 impacts *KDM4C* expression.

Two variants on chromosome 7 (rs194537 and rs9791638) showed suggestive evidence of association with *PIK3CA* p.E545K mutations in the MyBrCa study (unadjusted *P* values 0.042 and 0.047) but not in the multi-ancestry study (*P* values 0.53 and 0.78, respectively). These variants maps near *STEAP2*, a gene in which low expression is associated with poorer outcomes in breast cancer and activation of the PI3K/AKT/mTOR pathway. Conversely, up-regulation of STEAP2 suppresses this pathway [109].

The validation studies included the *MAP3K1/SETD9* variants rs252913 and rs331499 reported in the literature to be associated with *PIK3CA* mutation status in ER+ breast tumors, [63]. These variants and others at the *MAP3K1/SETD9* locus have been associated with breast cancer risk in multiple GWAS as well as body size [79,93,110,111]. Neither variant showed evidence of association with *PIK3CA* mutation status in our multi-ancestry validation studies (*P* values 0.54 and 0.37), but both showed suggestive evidence in MyBrCa Study (*P* values 0.09 and 0.14) and the B-CAUSE study (*P* values

0.005 and 0.008) for association with overall *PIK3CA* mutation status. These results should be taken with caution given the very small number of *PIK3CA* mutation carriers in the B-CAUSE study.

Other variants of potential interest included a MGAT5 intronic variant, rs2442046, and two SNX13/HDAC9 locus variants, rs13230836 and rs6955337. The MGAT5 variant showed a similar but non-significant effect size for association with p.E542K mutations in the discovery, multi-ancestry validation and MyBrCa studies (ORs 1.86, 1.64, and 1.92). MGAT5 encodes Nacetylgluocosaminyltransferase V which is a glycosyltransferase important in the synthesis of branched N-glycans, many of which are found on key cancer signaling proteins including EGFR and TGFBR [112,113]. MGAT5 expression is increased in multiple tumor types; in breast tumors overexpression is associated with metastasis, epithelial to mesenchymal transitions, and cell motility [114]. In mice, MGAT5 and PTEN act antagonistically to regulate PI3K/AKT signaling [115]. rs13230836 and rs6955337 map to chromosome 7 between SNX13 and HDAC and are in LD. They show evidence of association with PIK3CA mutations in the discovery (P values 7.4x10⁻⁶ and 3.2x10⁻⁶) and multi-ancestry validation (P values 0.023 and 0.007, respectively) but show non-significance findings in the MyBrCa study. rs6955337 maps within 10kb of rs11560253, a variant associated with gastric cancer risk [116]. However, rs6955337 and rs11560253 are only in LD in European and Japanese population (r²>0.8) but not in African or Han Chinese populations ($r^2 < 0.6$). HDAC9 is a class II histone deacetylase whose gene is hypermethylated in HR+ and/or TP53 wildtype breast cancers [117]; conversely, increased HDAC9 expression is seen in more aggressive breast cancers and those with poorer prognosis [118,119]. Increased HDAC9 expression in cell line models is associated with enhanced proliferation, decreased apoptosis and decreased ESR1 mRNA [118,120].

Ancestral differences in TP53 and PIK3CA mutation frequencies across cancer types

Associations with genetic ancestry and specific somatic driver mutations have been observed in other cancer types [23,121]. Genetic ancestry is associated with specific somatic driver mutations in EGFR, KRAS and STK11 in lung cancer in individuals of Indigenous American ancestry relative to those of EUR or EAS ancestry [122,123]. TP53 mutations are found at a higher frequency in individuals of AFA relative to individuals of EUR tumors in multiple tumor types (lung, colon, gastric, human papilloma virus negative head and neck), suggesting that genetic background and/or differences in exposures/socio-determinants of health may influence selection of TP53 somatic mutations [124-127]. PIK3CA somatic mutations also show differences by ancestry in different tumor types. For example, PIK3CA mutations have been observed at lower frequencies in bladder tumors arising in EAS individuals and in head and neck squamous cell carcinomas from AFA individuals [128,129]. Conversely, *PIK3CA* mutations are more often observed in colorectal tumors from AFA individuals [130]. Variants identified in this study may have utility in explaining TP53 and PIK3CA somatic mutation frequencies arising in different tissues that differ by genetic ancestry. We did not observe any significant ancestry-specific associations after corrections for multiple comparisons, but we were not adequately powered to assess this.

Study Limitations

There are some limitations to this study. Our discovery analyses were performed in non-Hispanic individuals of EUR, which means that variants enriched in or specific to non-European populations may not have been identified. We were underpowered to determine if our GxM findings were responsible for the observed differences in breast cancer *TP53* and *PIK3CA* mutation frequency for individuals of non-

European populations and for variants associated with specific PIK3CA mutations (e.g. p.E542K, p.E545K and p.H1047R/L). In our validation study, we did not genotype all variants/loci with P values of less than 1x10⁻⁴ observed in our discovery set, some of which were not included due to low MAF in one or more populations. As such, we may have missed key variants/loci associated with TP53 or PIK3CA mutation status. The source of somatic mutation information used as our phenotype varied widely with some information coming from clinical reports, some from whole genome/whole exome sequencing of tumors, some from targeted sequencing studies, and some from in-house Sanger sequencing studies. Next-generation sequencing is more sensitive than Sanger sequencing for somatic mutations that are present in fewer than 20% of cells or for tumors with a high degree of immune or stromal infiltrate. Our study was based on the premise that TP53 and PIK3CA mutations would be early driver events in tumor development and mutations in these genes should be present in a high proportion of tumor cells. In a previous study, in which we evaluated types of TP53 mutation by self-reported race and ethnicity, we found no differences in TP53 mutation frequency across studies by modality of somatic variation detection suggesting that Sanger sequencing is reasonable for mutation detection of early driver events present in a large proportion of cells [14]. Copy number information was not available for a large proportion of tumors; thus, TP53 mutations due to larger deletions (e.g. chromosome 17p loss) were not included. We expect that a subset of tumors defined as not having a mutation in TP53 may have had large copy number losses at that locus resulting in the missing of individuals with LOF mutations due to larger deletions.

Across populations, somatic mutations in *TP53* are more common in TNBC and HER2+ tumors; conversely, somatic mutations in *PIK3CA* are much more frequent in ER+ tumors and Luminal breast cancers [2,4]. Even with adjustment based on tumor subtype, it is difficult to sort out the association of the SNV with somatic mutation versus association of the SNV with tumor subtype. Previous studies

stratifying by ER- and ER+ tumor status have found ancestry differences in mutation frequency for these genes, but this was not the case for all studies stratifying by tumor subtype [4,8]. Future mechanistic studies are needed to determine if germline variants help drive tumor subtypes that are characterized by certain gene mutations and/or if germline variants impact a cellular context in which a particular mutation is more likely to be selected and the mutation is important for determining tumor subtype.

CONCLUSIONS

This study provides evidence that germline variants may shape somatic mutation processes or mutation selection of *TP53* and *PIK3CA* in breast tumors. In the future, polygenic risk scores based on findings from this study could identify individuals who are at increased risk of *TP53* or *PIK3CA* mutations should they develop breast cancer which may ultimately inform prevention strategies, such as potential vaccination-based prevention for high-risk individuals more likely to carry a specific somatic mutation. Larger multi-ancestry studies are warranted to confirm the study findings and determine if germline variants explain some of the differences in *TP53* and *PIK3CA* breast cancer mutation frequencies by genetic ancestry. Functional and mechanistic studies are needed to understand the target genes and pathways for variants associated with these mutations in breast tumors.

DECLARATIONS

AUTHOR CONTRIBUTIONS

Study conception and design (McElroy, Toland), Experimental Design (Ramroop, Tjader, Toland, Teo, Yao), cohort and sample collection, participant ascertainment and data extraction (Aeilts, Ambrosone,

Carpten, Hampel, Huo, Neuhausen, Palmer, Ping, Teo, Toland, Yao, Zhang, Ziv), mutational analysis (Beer, Dauch, Gandhi, Ghanekar, Gillespie, Hu, Liu, Meadows, Pan, Paredes, Ramroop, Sotelo, Tjader, Wei,), Genotyping (Fadda, Pan, Ramroop, Tjader, Toland), and data analysis (Ho, Hu, Liu, Jia, McElroy, Ping, Stevens, Tai, Teo, Wei). All authors provided edits and approved the final draft of manuscript.

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CONFLICTS OF INTEREST

HH is on the scientific advisory board for Invitae Genetics, Promega, and Genome Medical and has stock/stock options in Genome Medical and GI OnDemand. None of these are direct conflicts with this study.

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ETHICS APPROVAL AND CONSENT TO PARTICIPATE

This study was approved by the Ohio State University Cancer Institutional Review Board (IRB) (protocol number 2005C0082). All data and samples were from de-identified individuals who had undergone informed consent for participation in research studies. Patients and sample ids reported in the supplementary information are de-identified (cannot reveal study participant identify) or are not able to

be linked to participant identity outside of the researchers on those studies. Nigerian Study: The VOH IRB and the University of Chicago IRB approved study for participants enrolled at their respective sites.

COH Latina Study: One hundred and twenty Latina breast-cancer patients seen at City of Hope (COH) in Duarte, California were included in this study. All participants signed a written informed consent approved by the COH Institutional Review Board. MyBrCa was approved by the Independent Ethics Committee, Ramsay Sime Darby Health Care (reference no: 201208.1), and the Medical Ethics Committee, University Malaya Medical Centre (reference no: 842.9).

DATA AVAILABILITY STATEMENT

The majority of data generated or analyzed during this study are included in this published article (Supplementary Tables S2-S35), in TCGA, dbGAP and/or the following data repositories as listed below. TCGA tumor mutation data and SNV genotyping data are available in dbGAP under accession numbers phs001687.v1.p1, phs000178.v11.p8, and phs002387.v1.p1. METABRIC sequencing data of tumors and SNV genotyping data are available on the European Genome-Phenome archive using accession numbers EGAD0001000164, EGAS000000000083, EGAD00010000158, EGAD00010000266, EGAS00001004518 and EGAD0001006399. The Welcome Trust Sanger Institute data are available in the European Genome-Phenome archive using accessing number EGAS00001001178 and EGAD0010000915. Sequencing data and processed genomic data from the Nigerian breast cancer cases are in dbGAP under study accession number phs001687.v1.p1. Tumor/normal WES and RNAseq data and accompanying phenotypic and clinical/histologic data for the COH Latina Breast Cancer Study are deposited in dbGAP (dbGaP Study Accession: phs003218)[71]. MyBrCa WES and sWGS files are available on the European Genome-phenome Archive under the study accession number

EGAS00001004518. Access to controlled patient data will require the approval of the MyBrCa Tumour Genomics Data Access Committee upon request to genetics@cancerresearch.my. Sequence and genotyping data for the Banerji et al. study [72] are available in dbGAP under accession number phs000369.v1.p1. Summary-level statistics genotyping data for the AABCG study are available at GWAS Catalog (accession number: GCST90296719, GCST90296720, GCST90296721, and GCST90296722). B-CAUSE TNBC sequencing data is in the process being deposited into dbGaP with accession number pending.

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Tables

Table 1: Select variants showing evidence of association with TP53 mutation status

SNV rs# Ref allele	Chr:GRCh38 Nearest gene	Phenotype	Disc OR	Disc P	Valid OR	Valid P	MyBr Ca OR	MyBr Ca P	AABC OR	AABC P
rs10931697 C	2:195440726 SLC39A10	GOF	1.82	2.79E-05	0.53 (0.36-0.77)	0.001	0.72	0.09	1.00	0.91
rs6709393 A	2: 237596982 RAB17	GOF	0.79	0.0003	1.002 (0.65- 1.54)	0.99	0.28	0.003	1.04	0.16
rs6890674 T	5: 67180023 CD180	Any	0.61	7.48E-05	1.02 (0.83-1.2	0.88	0.63	0.0006	1.01	0.71
rs4254949 A	5: 110258567 TMEM232	LOF	0.69	0.0001	0.98 (0.8-1.2)	0.83	0.75	0.03	1.03	0.36
rs9397436 T	6:151630867 ESR1	Any LOF GOF	1.53 1.48 1.79	1.39E-05 0.0007 0.0001	1.21 (0.95-1.58) 1.20 (0.89-1.6) 1.09 (0.7- 1.64)	0.11 0.699 0.213	1.43 1.35 1.51	0.001 0.02 0.01	1.01	0.87
rs9383938 G	6:151666222 ESR1	Any GOF	ND	ND	1.46	7E-05	1.81 2.07	9.8E- 08 8.4E- 06	1.03 1.00	0.34 0.97
rs9479090 A	6:151668165 ESR1	Any	1.43	5.35E-05	1.10 (0.9-1.36)	0.36	1.76	2.8E- 07	1.03	0.22
rs17103093 C	10:122128410 TACC2	Any	1.54	3.32E-05	1.4 (1.03-1.92)	0.03	1.02	0.87	1.01	0.58
rs4767726 C	12:118831314 LINC02439	GOF	0.67	0.0002	1.49 (1.14-2.04)	0.009	1.06	0.7	0.95	0.17

Rs#, rs number; Ref allele, reference allele; Disc, Discovery Study; Valid, Multi-ancestry Validation Study; OR, odds ratio; P, P values.

Table 2: Variants associated with PIK3CA mutation status

SNV Rs#	Chr:GRCh38	Phenotype	Disc	Disc P	Validation OR	Validation	MyBrCa	MyBrCa
Ref Allele	Nearest Gene		OR		(95% CI)	P	OR	P
rs859074	1:94883881	E542K	0.54	0.0002	1.02 (0.68-1.53)	0.90	0.45	0.04
G	SLC44A3							
rs13230836	7:17974272	Any	0.78	7.41E-06	0.81 (0.68-0.97)	0.023	1.2	0.12
Α	SNX13/HDCA9							
rs6955337	7:17974951	Any	1.3	3.22E-06	1.41 (1.1-1.82)	0.007	0.83	0.13
A	SNX13/HDCA9							
rs9791638	7:90158086	E545K	1.68	1.54E-05	0.93 (0.59-1.47)	0.78	1.67	0.047
T	STEAP1/STEAP2							
rs194537	7:90245730	E545K	1.73	4.6E-06	0.81 (0.54-1.19)	0.53	1.67	0.042
T	STEAP1/STEAP2							
rs10975835	9:6797498	H1047R/L	1.54	1.12E-05	1.5 (1.1-2.04)	0.011	0.92	0.66
С	KDM4C				·			
rs2026801	9:111143171	Activating	0.75	2.52E-06	0.88 (0.73-1.05)	0.16	0.74	0.009
A	LPAR1							
rs2332431	14:70609134	E542K	1.80	0.0002	1.26 (0.80-1.93)	0.31	0.46	0.03
A	MED6							
rs10518943	15:57863151	E542K	0.51	8.11E-05	0.99 (0.65-1.49)	0.96	0.32	0.016
C	POLR2M							
	ALDH1A2							
rs8084310	18:8950506	E542K	1.80	0.0002	1.09 (0.63-1.82)	0.75	3.39	0.017
T	SOGA2							
	CCDC165							
rs2442046	2:135014078	E542K	1.86	1.94E-05	1.64 (0.95-2.77)	0.07	1.92	0.14
G	MGAT5							
rs2738348	X:348699	E542K	0.7	0.0003	0.78 (0.59-1.03)	0.08	ND	ND
C								

Ref allele, reference allele; CI, confidence interval; ND: No data; Disc, Discovery Study; Validation, Multi-ancestry Validation Study; OR, odds ratio; p, *P* values.

FIGURE LEGENDS:

Figure 1: Manhattan Plots for TP53 Discovery GxM Analyses

Discovery GWAS data for (A) 879 *TP53* mutation carriers and 1965 breast cancer cases without *TP53* mutations, (B) 237 cases with *TP53* GOF mutations and 1965 breast cancer cases without *TP53* mutations, and (C) 536 cases with *TP53* LOF mutations and 1965 controls are plotted by -log₁₀(*P* values). Blue lines represent *P* values of less than 1x10⁻⁵. Chromosome numbers are indicated. GXM, Germline Variant by Mutation; GWAS, genome-wide association study; GOF, gain of function; LOF,

loss of function.

Figure 2: Manhattan Plots for PIK3CA Discovery GxM Analyses

Discovery GWAS data for (A) 1095 *PIK3CA* mutation carriers and 1642 breast cancer cases without *PIK3CA* mutations, (B) 858 cases with PIK3CA activating/hotspot mutations and 1642 breast cancer cases without *PIK3CA* mutations, (C) 112 cases with *PIK3CA* p.E542K mutations and 1642 breast cancer cases without *PIK3CA* mutations, (D) 193 cases with *PIK3CA* p.E545K mutations and 1642 breast cancer cases without *PIK3CA* mutations, and (E) 387 cases with *PIK3CA* p.H1047R mutations and 1642 breast cancer cases without *PIK3CA* mutations are plotted by -log₁₀(*P* value). Blue lines represent *P* values of less than 1x10⁻⁵. Chromosome numbers are indicated. GXM, Germline Variant by Mutation; GWAS, genome-wide association study.

SUPPORTING INFORMATION

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