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The path to individualised breast cancer screening

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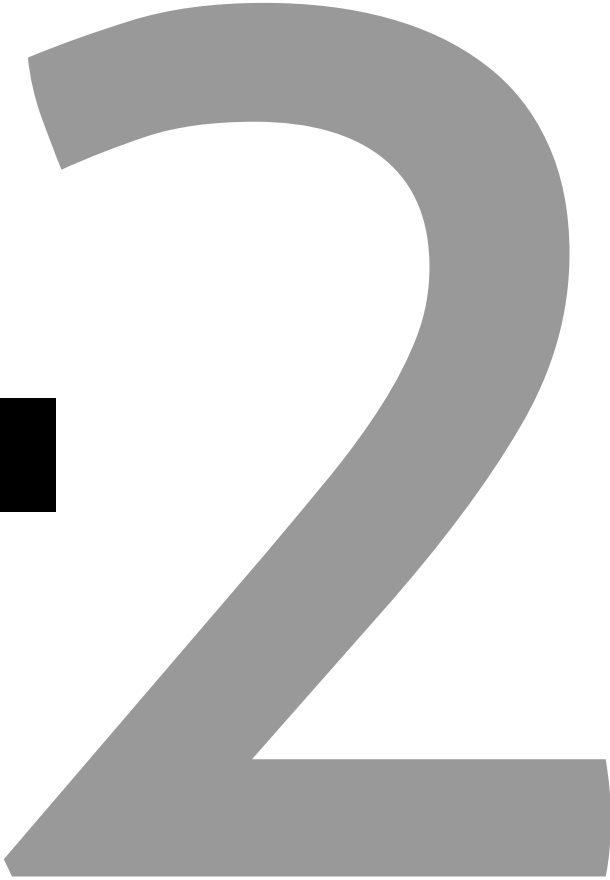
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CHAPTER 2



Addition of a 161-SNP Polygenic Risk Score to family history-based risk prediction: impact on clinical management in non-*BRCA1/2* breast cancer families

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Abstract

Background

The currently known breast cancer associated Single Nucleotide Polymorphisms (SNPs) are presently not used to guide clinical management. We explored whether a genetic test that incorporates a SNP-based Polygenic Risk Score (PRS) is clinically meaningful in non-*BRCA1/2* high-risk breast cancer families.

Methods

101 non-*BRCA1/2* high-risk breast cancer families were included; 323 cases and 262 unaffected female relatives were genotyped. The 161-SNP PRS was calculated and standardised to 327 population controls (sPRS). Association analysis was performed using a Cox-type random effect regression model adjusted by family history. Updated individualised breast cancer lifetime risk scores were derived by combining the Breast and Ovarian Analysis of Disease Incidence and Carrier Estimation Algorithm (BOADICEA) breast cancer lifetime risk with the effect of the sPRS.

Results

The mean sPRS for cases and their unaffected relatives was 0.70 (SD=0.9) and 0.53 (SD=0.9), respectively. A significant association was found between sPRS and breast cancer, HR=1.16, 95%CI=1.03-1.28, p=0.026. Addition of the sPRS to risk prediction based on family history alone changed screening recommendations in 11.5%, 14.7%, and 19.8% of the women according to breast screening guidelines from the USA (National Comprehensive Cancer Network), UK (National Institute for Health and Care Excellence) and the Netherlands (Netherlands Comprehensive Cancer Organisation), respectively.

Conclusion

Our results support the application of the PRS in risk prediction and clinical management of women from genetically unexplained breast cancer families.

Introduction

Breast cancer is the most common cancer in women in the Western world. For women with a first-degree relative with breast cancer, the risk for developing breast cancer is twofold in comparison with women without such a family history¹. Approximately 20% of this familial relative risk is explained by pathogenic variants in the high-risk genes *BRCA1* and *BRCA2*, 2-5% by variants in other breast cancer genes (e.g. *CHEK2*, *PALB2*, and *ATM*), and 18% by the currently known common low risk variants, mostly single nucleotide polymorphisms (SNPs)²⁻⁵.

Individually these SNPs confer a very small increase in breast cancer risk but jointly they may confer a substantial increase of the risk². This combined risk of all SNPs associated with breast cancer can be summarised in a Polygenic Risk Score (PRS). The PRS can stratify women into different risk categories^{2,6-8}, which for 8% of women from the general population might be high enough to be clinically relevant, regardless of family history².

The PRS may also be combined with other risk factors, such as *BRCA1/2* status or breast cancer family history, to further refine and individualise risk estimation. The large majority of breast cancer families seen in Family Cancer Clinics today cannot be linked to pathogenic variants in *BRCA1* or *BRCA2*. Risk management for women from these families is based mainly on family history, which can be used as a variable to calculate individual breast cancer risk in various risk prediction algorithms⁹, such as the Breast and Ovarian Analysis of Disease Incidence and Carrier Estimation Algorithm (BOADICEA)¹⁰.

Until now, the PRS is not included in clinical genetic practice to guide clinical management. Several studies have shown an improved discriminative power between breast cancer cases and controls by combining the PRS with a breast cancer risk prediction tool¹¹⁻¹⁴. However, little is currently known of the discriminative power of the PRS between family members, with respect to who will develop breast cancer. A recent study genotyped cases and controls in 52 Finnish non-*BRCA1/2* breast cancer families to calculate a 75-SNP PRS. The PRS for healthy women from breast cancer families was lower in comparison to affected family members¹⁵. This suggests that the PRS can help to individualise risk stratification and advice for surveillance for women in breast cancer families.

Here, we explore the clinical applicability of the 161-SNP PRS for risk prediction in a cohort of 101 high-risk breast cancer families not explained by pathogenic variants in the *BRCA1* and *BRCA2* genes. The clinical impact of the PRS on breast cancer risk prediction based on family history alone was investigated by determining the potential change in clinical management, as stipulated by three currently used guidelines (the National Comprehensive Cancer Network guideline (NCCN)¹⁶, the National Institute for Health

and Care Excellence guideline (NICE)¹⁷, and the Netherlands Comprehensive Cancer Organisation guideline (IKNL)¹⁸).

Materials and Methods

Study cohorts

Two cohorts were included, a hospital-based case-control (Oorsprong van borstkanker integraal onderzocht (ORIGO)) and a family-based case-control cohort. Informed consent was obtained for all individuals. Population controls were irreversibly anonymised. Only women were included in this study.

The ORIGO cohort consists of incident breast cancer cases, not selected for breast cancer family history enrolled between 1996 and 2006 in the context of the ORIGO study, as described elsewhere¹⁹. For the present study, 357 ORIGO cases were selected for which genotyping had been performed on the iCOGS array. Likewise, 327 healthy genotyped bloodbank donors were included in the ORIGO cohort as controls. Age of last follow up was determined as the age at diagnosis for cases and the age at inclusion for controls.

The families from the family-based cohort were selected between 1990 and 2012 through five Clinical Genetic Services (Rotterdam, Groningen, Nijmegen, Leiden, the Netherlands and Budapest, Hungary) and the Foundation for the Detection of Hereditary Tumours in the Netherlands, as previously described²⁰. At least one family member affected with breast cancer was tested for *BRCA1* and *BRCA2*. We did not have informed consent for testing other specific genes besides *BRCA1* and *BRCA2*. The selection criteria for families included: breast cancer (invasive/in situ) before the age of 60 in at least three women, or in two women if at least one of them had bilateral breast cancer before the age of 60. In total, 102 families without a pathogenic variant in *BRCA1* or *BRCA2* were included of which a blood DNA sample was available for 612 women. Of these women, 340 were affected with breast cancer and 272 were unaffected relatives. The unaffected relatives were censored regarding breast cancer, irrespective of other types of cancer. Most cancers were verified with a pathology report. Date of last follow up was determined as the date of last contact with the family.

Genotyping

DNA samples of all included individuals were genotyped with the iCOGS SNP array, designed for association analysis in breast, ovarian and prostate cancer, containing 211,155 SNPs³. Genotyping and quality control of the ORIGO cohort was performed as part of association studies conducted by the Breast Cancer Association Consortium (BCAC)³. For the family-based cohort, quality control led to the exclusion of 27 individuals

(see supplementary material and methods). Therefore further analysis was done with 323 breast cancer cases and 262 unaffected relatives from 101 families for this cohort.

Imputation

Some of the 182 currently known SNPs are associated primarily with Estrogen Receptor (ER)-negative or ER-positive breast cancer. We constructed a PRS for overall breast cancer with 161 SNPs, selecting all SNPs significantly associated ($p < 5 \cdot 10^{-8}$) with overall breast cancer in case-control studies performed by BCAC⁴ (Table S1). ER-status was not known for all cases in our study, and substrata would become too small to reach sufficient statistical power for ER-specific PRSs. The 85 SNPs that were not directly genotyped by the iCOGS array were imputed by pre-phasing with SHAPEIT and IMPUTE2^{21,22}. To improve imputation quality both the reference panels 1000 genomes phase 3 and GoNL were used^{23,24}.

Polygenic risk score

The following formula was used to calculate the PRS based on 161 SNPs:

$$PRS_j = \sum_{i=1}^{161} n_{ij} \ln(OR_i)$$

where n_{ij} is the number of risk alleles (0, 1 or 2) for SNP i carried by individual j and OR_i is the per-allele log odds ratio (OR) for breast cancer associated with SNP i . The ORs were the most recent estimates from analysis of the OncoArray data⁴ (Table S1). The majority of studies used for this analysis were population-based case-control studies⁴.

The PRS was calculated for all included individuals. For the descriptive analysis, the PRS was standardised to the mean and standard deviation (SD) in healthy population controls. The mean standardised PRS (sPRS) in population controls is therefore 0 with an SD of 1. Standardisation facilitates the comparison between different groups. For further analysis in the family-based cohort, the PRS was standardised to the mean and SD in the family-based cohort including both cases and unaffected relatives.

Total BOADICEA score and polygenic load (BOADICEA_{FH})

The pedigrees were collected and drawn for all families, including all known first-degree and second-degree relatives of the genotyped individuals. For 25 of the 561 family members affected with breast cancer, the age of breast cancer diagnosis was not known. For these affected family members, the age at diagnosis was assumed to equal the average age of developing breast cancer in the Netherlands (61 years), or the age at last follow up if this was earlier.

Two different scores were calculated for all individuals in the family cohort by the online risk prediction tool BOADICEA¹⁰, the total BOADICEA score and the polygenic load. The total BOADICEA score (hereafter termed $BOADICEA_{LTR}$) is a measure for lifetime breast cancer risk, and incorporates *BRCA1* and *BRCA2* status, age, birth cohort and a polygenic load. The polygenic load in the BOADICEA model is an estimated polygenetic component representing a large number of loci of small effect to capture the residual familial aggregation of breast cancer and is therefore a measure of the breast cancer family history¹⁵. Calculation of the polygenic load is described previously by Muranen et al.¹⁵. To avoid confusion between the variables polygenic load and the PRS, the polygenic load is hereafter termed $BOADICEA_{FH}$. The $BOADICEA_{LTR}$ and $BOADICEA_{FH}$ were calculated by simulating an individual to be at an age of 1 year and unaffected (for cases), that is, lifetime risk at birth, given the family history.

Statistical analysis

To define the degree of correlation between the sPRS and the $BOADICEA_{FH}$, the Pearson correlation coefficient was calculated. A Cox-type random effect regression model was used to estimate the association between the sPRS and breast cancer, adjusting by family history, using the $BOADICEA_{FH}$ (FH) as covariate:

$$\lambda(t_{ij}) = u_i \lambda_0(t_{ij}) \exp(\beta_1 sPRS_{ij} + \beta_2 FH_{ij}) \quad (1)$$

where t_{ij} is the age at first diagnosis of breast cancer or the age at censoring for member j in family i . Censoring was done at age of last contact with the family or death. Censoring at the age of diagnosis for other tumours, if present, did not affect the result. $\lambda_0(t_{ij})$ refers to the baseline hazard, which is left completely unspecified (Cox-type model), β_1 is the main effect of interest, the regression coefficient of the sPRS and β_2 is the effect of the $BOADICEA_{FH}$. In comparing affected to unaffected relatives, it is important to adjust for different numbers of affected versus unaffected relatives per family. We therefore added a family specific random effect $u > 0$ in our model, shared by the members of the same family. This unobserved heterogeneity shared within families was assumed to follow a gamma distribution.

To evaluate the potential of the sPRS on the reclassification of breast cancer risk, we constructed a new individual breast cancer risk score based on both the $BOADICEA_{LTR}$ and the estimated effect of the sPRS with the model defined by expression 1. Namely, since $BOADICEA_{LTR}$ is defined as the probability of experiencing breast cancer before age 80 years, the new score is calculated as the distribution function at 80 of a Cox proportional hazard model using $BOADICEA_{LTR}$ as baseline (average risk in the sample) and the sPRS as covariate:

$$\text{BOADICEA}_{\text{sPRS}} = 1 - (1 - \text{BOADICEA}_{\text{LTR}})^{e(\beta_1 * \text{sPRS})} \quad (2)$$

The sPRS is expected to individualise cancer risk estimates, but not to alter the overall average risk level computed by BOADICEA in the joint sample, that is, the higher risks given to some individuals are expected to be compensated by lower risks in others. For this reason we centred the sPRS at the mean of the whole family cohort.

The risk calculation based on BOADICEA alone ($\text{BOADICEA}_{\text{LTR}}$) and the new individual breast cancer risk score ($\text{BOADICEA}_{\text{sPRS}}$) were compared for all individuals in the family-based cohort to define the change in risk category and thus advice for breast cancer surveillance according to three different guidelines, NICE¹⁷, NCCN¹⁶ and IKNL¹⁸ (Table S2).

Statistical significance was established at 5%, analysis was performed using R version 3.4.1²⁵.

Results

The analysis of the ORIGO cohort included 357 breast cancer cases and 327 population controls. The analysis of the family-based cohort included 323 breast cancer cases and 262 unaffected relatives from 101 families. Unaffected relatives derived from 49 of these 101 families.

Descriptive analysis

Virtually all breast cancers were invasive in both cohorts, and second breast cancers were more prevalent in familial cases (Table 1). In both the ORIGO and family-based cohort, the sPRS was on average higher in cases than in controls (Table 2). The unaffected relatives in the family-based cohort had on average a higher sPRS in comparison with ORIGO cases and controls. The mean sPRS for sporadic cases was 0.35 (SD=0.92), and in the family-based cohort, the mean sPRS was 0.70 (SD=0.90) and 0.53 (SD=0.95) for the affected and unaffected relatives respectively. In the family-based cohort, the sPRS was higher for cases with two invasive breast tumours in comparison with cases with one breast tumour (invasive/in situ), with a mean sPRS of 0.66 (SD=0.89) and 0.89 (SD=0.93) respectively. The distributions of the sPRS in both cohorts are shown in Figure 1. Information about the 95% Confidence Interval (CI) and Standard Error (SE) in different groups are shown in Table 2.

Table 1: Characteristics of all included individuals

		ORIGO cohort		Family-based cohort	
		cases	controls	cases	unaffected relatives
Number		357	327	323	262
Age	Mean (SD)	56 (10)	46 (14)	51 (11)	62 (17)
	Range	23-84	18-90	26-90	17-94
Country of origin	The Netherlands	357	327	317	249
	Hungary	-	-	6	14
First breast tumour	Invasive (%)	313 (88)	-	317 (98)	-
	DCIS (%)	32 (9)	-	4 (1)	-
	Unknown (%)	12 (3)	-	2 (1)	-
Second breast tumour	Invasive (%)	19 (5)	-	51 (16)	-
	DCIS (%)	2 (1)	-	4 (1)	-
	Unknown (%)	0 (0)	-	5 (2)	-
Family score	BOADICEA_{FH} (SD)	-	-	1.03 (0.40)	1.05 (0.39)
	BOADICEA_{LTR} (SD)	-	-	0.23 (0.07)	0.23 (0.06)

Abbreviations: BOADICEA_{FH}, Breast cancer family history score; BOADICEA_{LTR}, Breast cancer lifetime risk at age 80; DCIS, ductal carcinoma in situ.

Table 2: Descriptive analysis 161-SNP PRS

Group	Mean sPRS	SD sPRS	SE sPRS	n	95% CI	
					lower limit	upper limit
Family breast cancer cases	0.70	0.90	0.05	323	0.60	0.80
1 breast tumour	0.66	0.89	0.05	267	0.55	0.76
2 breast tumours	0.89	0.93	0.12	56	0.65	1.13
Unaffected relatives	0.53	0.95	0.06	262	0.41	0.64
ORIGO cases	0.35	0.92	0.05	357	0.26	0.45
Population controls	0.00	1.00	0.06	327	-0.11	0.11

Abbreviations: PRS, polygenic risk score; SNP, single nucleotide polymorphism; sPRS, standardised PRS.

Correlation

Further analyses were performed only for the family-based cohort. A weak but statistically significant positive correlation was detected between the BOADICEA_{FH} (measure of the family history) and the sPRS. The Pearson correlation coefficient was 0.103, 95% confidence interval (CI)=0.022-0.183, P=0.013, which means that 1.1% of the variance in the sPRS is explained by the BOADICEA_{FH}. Larger correlation was found in the unaffected relatives

(correlation coefficient 0.153, 95%CI=0.032-0.269, P=0.013). No evidence of correlation was found in family cases only (correlation coefficient 0.057, 95%CI=-0.052-0.165, P=0.306).

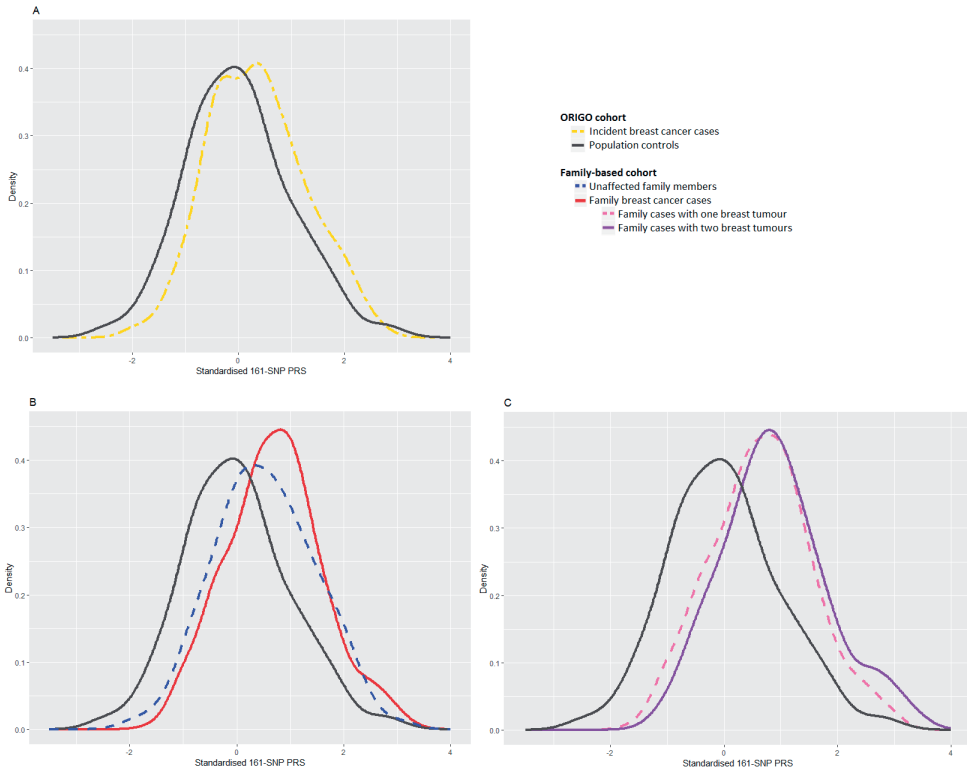


Figure 1: Distribution of the standardised 161-SNP PRS

The standardised 161-SNP PRS was plotted against the density in the different cohorts. (A) incident breast cancer cases and population controls from the ORIGO cohort; (B) population controls from the ORIGO cohort, breast cancer cases and unaffected relatives from the family-based cohort; (C) population controls from the ORIGO cohort, breast cancer cases with one and two primary breast tumours from the family-based cohort.

Abbreviations: PRS, polygenic risk score; SNP, single nucleotide polymorphism

Cox-type random effects modelling

The sPRS should not be directly combined with the $BOADICEA_{LTR}$ because the PRS is a part of the familial relative risk, captured by $BOADICEA$ by its polygenic component, the $BOADICEA_{FH}$. For this reason, adjustment was made by the $BOADICEA_{FH}$ in the association analysis, using model (1). Furthermore, adjusting for the $BOADICEA_{FH}$ helps to correct for ascertainment bias. The $BOADICEA_{FH}$ was calculated for cases assuming they were at age 1 year and unaffected. Consequently controls have, in our sample, a larger $BOADICEA_{FH}$ than cases. Hence, adding the $BOADICEA_{FH}$ as a covariate in the model indirectly corrects the oversampling of cases of our design. Within the family-based cohort, the

sPRS was significantly associated with breast cancer, conferring a hazard ratio (HR) of 1.16 (95%CI=1.03-1.28; P=0.026) per SD. No statistical significant association was found without adjustment, HR 1.10, 95%CI= 0.98-1.23, P=0.122.

PRS-based individualised risk score

To calculate a PRS-based breast cancer risk score ($BOADICEA_{sPRS}$), the individual sPRS was combined with the $BOADICEA_{LTR}$. Both risk scores for each individual in the family-based cohort are plotted against each other in Figure 2. This resulted in a change in breast cancer lifetime risk for all individuals. We evaluated the proportions of individuals that would fall in another risk management category, given risk cut-off levels from three different clinical guidelines. Risk management changed for 19.8%, 14.7%, and 11.5% of women under the IKNL¹⁸, NICE¹⁷, and NCCN¹⁶ guidelines, respectively (Table 3). The percentage of family cases and unaffected relatives who changed to a lower or higher risk category based on these guidelines are shown in Table S3. Examples of the change in breast cancer risk category are shown for individuals in three pedigrees in Figure 3 and Table S4.

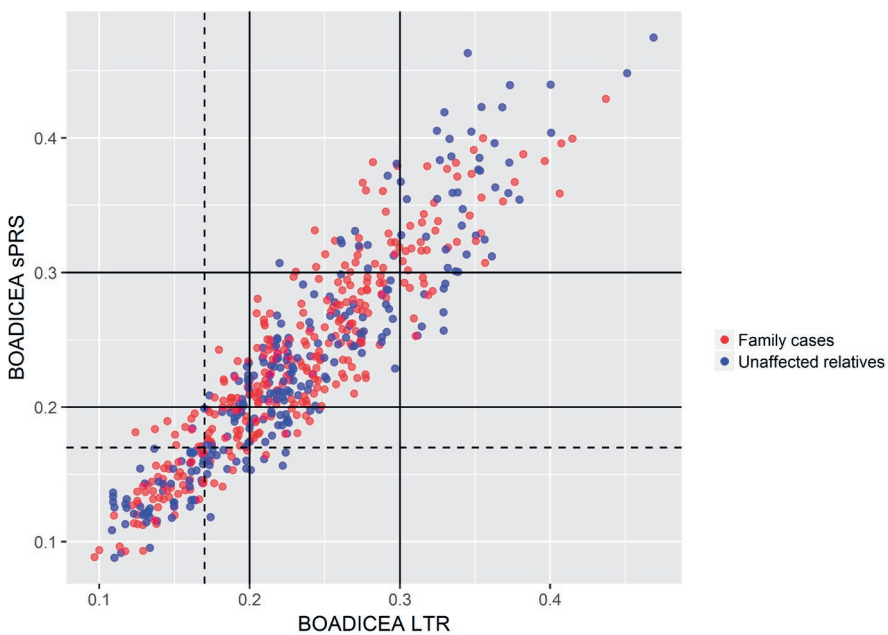


Figure 2: Change in breast cancer lifetime risk score

For every individual, $BOADICEA_{sPRS}$ was plotted against $BOADICEA_{LTR}$. The dotted lines represent the 17% breast cancer lifetime risk cut-off level. The solid lines represent the 20% and 30% breast cancer lifetime risk cut-off levels.

Abbreviations: $BOADICEA_{sPRS}$ 161-SNP PRS based breast cancer lifetime risk score; $BOADICEA_{LTR}$ breast cancer lifetime risk at age 80, based on $BOADICEA$ alone.

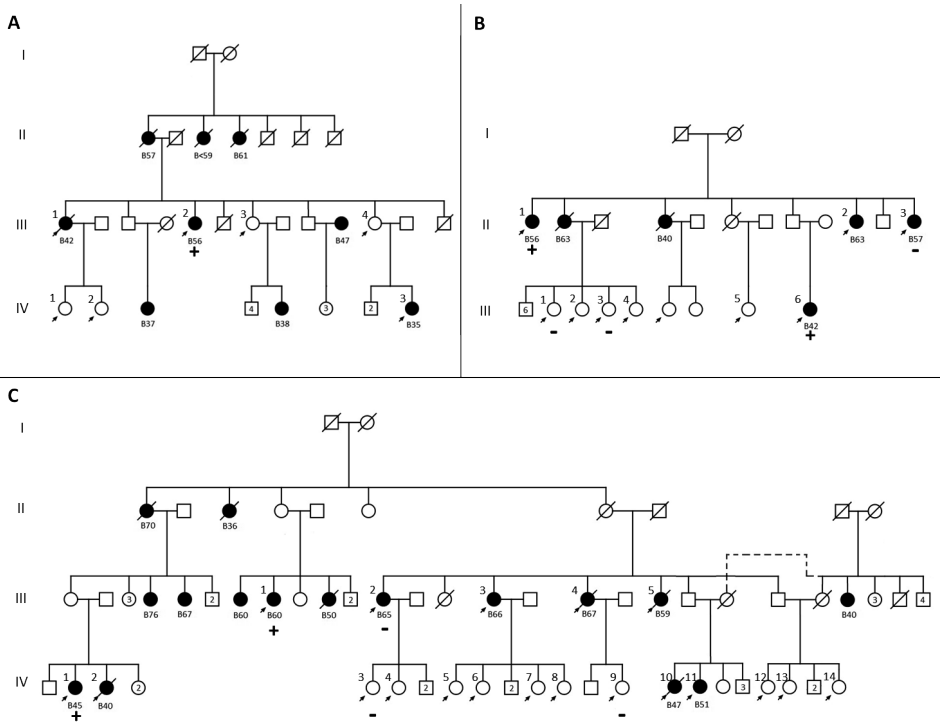


Figure 3: Risk management change for 11 women from three pedigrees

Risk changes are based on the Dutch IKNL screening guideline¹⁸ (Table S2). An arrow indicates that a woman has been genotyped. Generations in the pedigree are numbered with I, II, III and IV.

Based on the individual $BOADICEA_{sPRS}$ score, 11 individuals will change to a higher (+) or lower (-) risk category compared to the $BOADICEA_{LTR}$ score and will receive other breast screening surveillance.

Abbreviations: B, breast cancer; $BOADICEA_{sPRS}$, 161-SNP PRS based breast cancer lifetime risk score; $BOADICEA_{LTR}$, breast cancer lifetime risk at age 80, based on $BOADICEA$ alone; IKNL, Netherlands Comprehensive Cancer Organisation.

Table 3: Breast cancer risk category change in the family-based cohort

Lifetime risk		IKNL ¹⁸		NICE ¹⁷		NCCN ¹⁶	
BOADICEA _{LTR}	BOADICEA _{sPRS}	N	% change	N	% change	N	% change
<17%	<17%			108	10.7%		
<17%	>17%			13			
17-30%	17-30%			317	15.5%		
17-30%	<17%			24			
17-30%	>30%			34			
<20%	<20%	175	14.2%			175	14.2%
<20%	>20%	29				29	
>20%	>20%					343	10.0%
>20%	<20%					38	
20-30%	20-30%	220	24.7%				
20-30%	<20%	38					
20-30%	>30%	34					
>30%	>30%	74	16.9%	74	16.9%		
>30%	<30%	15		15			
overall change			19.8%		14.7%		11.5%

Following the Dutch IKNL guideline, cut off levels of 20% and 30% represent low, moderate and high risk categories. Following the NICE guideline, 17% and 30% represent low, moderate and high risk categories. Following the NCCN guideline, 20% represent a cut off level for the high risk category. Abbreviations: BOADICEA_{LTR}, breast cancer lifetime risk at age 80, based on BOADICEA alone; BOADICEA_{sPRS}, 161-SNP PRS based breast cancer lifetime risk score; IKNL, Netherlands Comprehensive Cancer Organisation; NCCN, National Comprehensive Cancer Network; NICE, National Institute for Health and Care Excellence

Discussion

Polygenic risk scores, derived from a combination of disease-associated SNPs, are gaining importance as predictive factor for a range of disease phenotypes, including breast cancer²⁶. All discovered breast cancer SNPs to date explain 18% of the familial relative risk⁴. Here, we use a PRS based on these SNPs, to show the potential clinical utility within high-risk breast cancer families. While most studies use population controls as a reference group^{2, 8, 12, 13}, we used the healthy relatives of breast cancer cases as a reference to make it more compatible with clinical practice in Family Cancer Clinics. Similar to population-based case-control studies^{2, 12, 13}, we found that the PRS was significantly associated with breast cancer within high-risk breast cancer families. In addition, the PRS may change breast screening recommendations in a substantial proportion of women from these families, according to currently used screening guidelines¹⁶⁻¹⁸. For incompleteness of data on ER-status, we did not calculate PRSs predictive for ER-positive or ER-negative disease^{5, 27}. While breast cancer screening guidelines are mainly based on overall breast cancer risk,

some guidelines suggest discussing the use of chemoprevention with women at high risk of breast cancer^{16,17}. We expect these ER-specific PRSs, similar to the overall PRS, to individualise these discussions within these families.

Some studies have described an association between the PRS and contralateral breast cancer^{8,28}. In agreement with this, we found the average sPRS in women diagnosed with two primary breast cancers in our family cohort to be higher in comparison with women with one breast cancer (similarly in ORIGO cases, Figure S1 and Table S5). Thus, the PRS may be helpful managing contralateral breast cancer risk and guide the choice for treatment or risk reducing mastectomy.

The family-based cohort used in our study was not part of the cohort used to discover the breast cancer associated SNPs by GWAS, while the ORIGO cohort was^{3,4}. A notable finding in our family-based cohort was that unaffected relatives of familial breast cancer cases had on average a higher sPRS than ORIGO incident breast cancer cases, not selected by family history. This may be due to our selection of families with multiple cases of breast cancer, since SNPs of this PRS are expected to cluster in breast cancer families. Moreover, the mean sPRS we calculated for ORIGO cases was lower than found in a large population-based study². Since we found no evidence for substructures in the ORIGO cohort (Figure S1 and Table S5), this effect is probably due to the relatively small number of ORIGO cases included in this study.

Three previous studies have also genotyped breast cancer cases and their unaffected relatives^{7,15,29}. These studies found an association with breast cancer as well, but effect-sizes are difficult to compare because of differences in methodology and cohort selection criteria. Furthermore, these studies used a much smaller number of SNPs to calculate the PRS. Li et al⁷ analysed a prospective dataset, and concluded that their 24-SNP PRS could have altered clinical management in up to 23% of women, regarding an MRI screening-threshold of 20% breast cancer lifetime risk. Evans et al.²⁹ performed a case-control study of women attending a familial risk clinic, and showed that their 18-SNP PRS moved 52% of the controls without a pathogenic variant in *BRCA1* or *BRCA2* to a different lifetime risk category based on the NICE guideline^{17,29}.

In our study, we adopted a conditional approach for association analysis because of the large heterogeneity between the families. Although our use of the BOADICEA_{FH} adjusts for family history, the HR is probably still underestimated given the strong selection criteria used in our study. Of note, this BOADICEA_{FH} is not a true family score in a clinical sense, given the retrospective nature of our family cohort. In clinical practice the risk scores are only calculated for unaffected family members, while in this study, we derive the BOADICEA_{FH} also for cases, assuming they were at age 1 and unaffected. With this definition, controls

have, in general, a larger $BOADICEA_{FH}$ than cases. Hence, adding the $BOADICEA_{FH}$ as a covariate in the model indirectly corrects the oversampling of cases of our design. The same definition of the $BOADICEA_{FH}$ is also used when computing $BOADICEA_{LTR}$ and the new individual score $BOADICEA_{sPRS}$ given by expression (2).

We found that 1.1% of the variance in the sPRS is explained by the $BOADICEA_{FH}$. Given that 18% of the familial relative risk for breast cancer is explained by the currently known SNPs, this is lower than expected. Nonetheless, other studies have also found a weak correlation or no correlation at all between the PRS and the $BOADICEA_{FH}$ or total $BOADICEA$ score^{12, 15}. Thus $BOADICEA$ appears to be a poor predictor of the PRS, underscoring the value of measuring the PRS for every individual in the family instead of using an estimated PRS based on the total family history.

It is estimated that a large number of SNPs just below the level of genome-wide significance, combined with the currently used 161 SNPs, are able to explain about 41% of the familial relative risk⁴. Addition of these SNPs could potentially further refine risk prediction and improve the discriminatory power of the PRS. Studies are now ongoing to find the best performing PRS, including also these SNPs. Khera et al.³⁰ found that a PRS of 5218 SNPs associated with breast cancer at a significance level of $<5.10^{-4}$, combined with age, had the best performance based on the area under the receiver-operator curve. Mavaddat et al.³¹ used a hard-thresholding approach to include 313 SNPs at a significance level of $<10^{-5}$. A further improvement for breast cancer risk prediction could come from information on pathogenic variants in non-BRCA high- or moderate-risk breast cancer genes (e.g. *PALB2*, *CHEK2*, *ATM*). Pathogenic variants in these genes are found in approximately 4-6% of women affected with breast cancer^{32, 33}. Recently, the $BOADICEA$ model has been extended with incorporation of the effects of truncating variants in *CHEK2*, *PALB2* and *ATM* and the 313-SNP based PRS to calculate breast cancer lifetime risks³⁴. A limitation of our study is that we had no ethical approval to test *CHEK2*, *PALB2* and *ATM* in the studied families. Extrapolating from expected prevalences of pathogenic variants in these genes, we estimate the total percentage of individuals that would have changed to another risk category by addition of the PRS to be 3-4% higher than the 20% we report here.

In summary, we showed that the PRS based on the most recently discovered breast cancer SNPs can be used for breast cancer risk prediction within high-risk breast cancer families. Individualising breast cancer risk prediction by adding the individual 161-SNP PRS to family history-based risk prediction may change screening recommendation in up to 20% of the individuals in these families. While this study illustrates the importance of clinical applicability of the PRS, our results must be interpreted with caution. The HR obtained in this family cohort cannot be translated directly to the clinic as the effect-size must be

validated in another larger familial breast cancer cohort. Further evaluation, preferably in prospective settings, will be needed.

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Disclosure of potential conflict of interests

The authors declare no potential conflicts of interest

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Supplementary methods

Quality Control

For the ORIGO cohort quality control was performed as part of association studies conducted by the Breast Cancer Association Consortium (BCAC)^{1, 2}. To summarise the thresholds used, individuals were excluded when they were genotypically not female, overall call-rate was <95%, low or high heterozygosity ($P < 1 \times 10^{-6}$), first-degree relatives determined by identity-by-state estimates or in the case of ancestry outliers by multidimensional scaling. SNPs were excluded with call rates <95% or deviation from Hardy-Weinberg equilibrium in controls at $P < 1 \times 10^{-7}$.

For the family-based cohort, quality control was performed with Plink version 1.7^{3,4}, which excluded 14342 SNPs with a call rate below 98%. For the remaining SNPs, there was no deviation from Hardy-Weinberg equilibrium in controls at $P < 1 \times 10^{-3}$. In total 27 individuals were excluded of which 19 individuals with a call rate below 96% and 6 individuals because of another degree of relatedness than expected based on identity-by-state estimates and pedigree information. Two individuals were genotypically not female and were excluded from further analysis.

Multidimensional scaling was performed to determine clustering of families, including the Hungarian families. There were no different clusters for families, therefore we could also include the Hungarian families.

Supplementary figures and tables

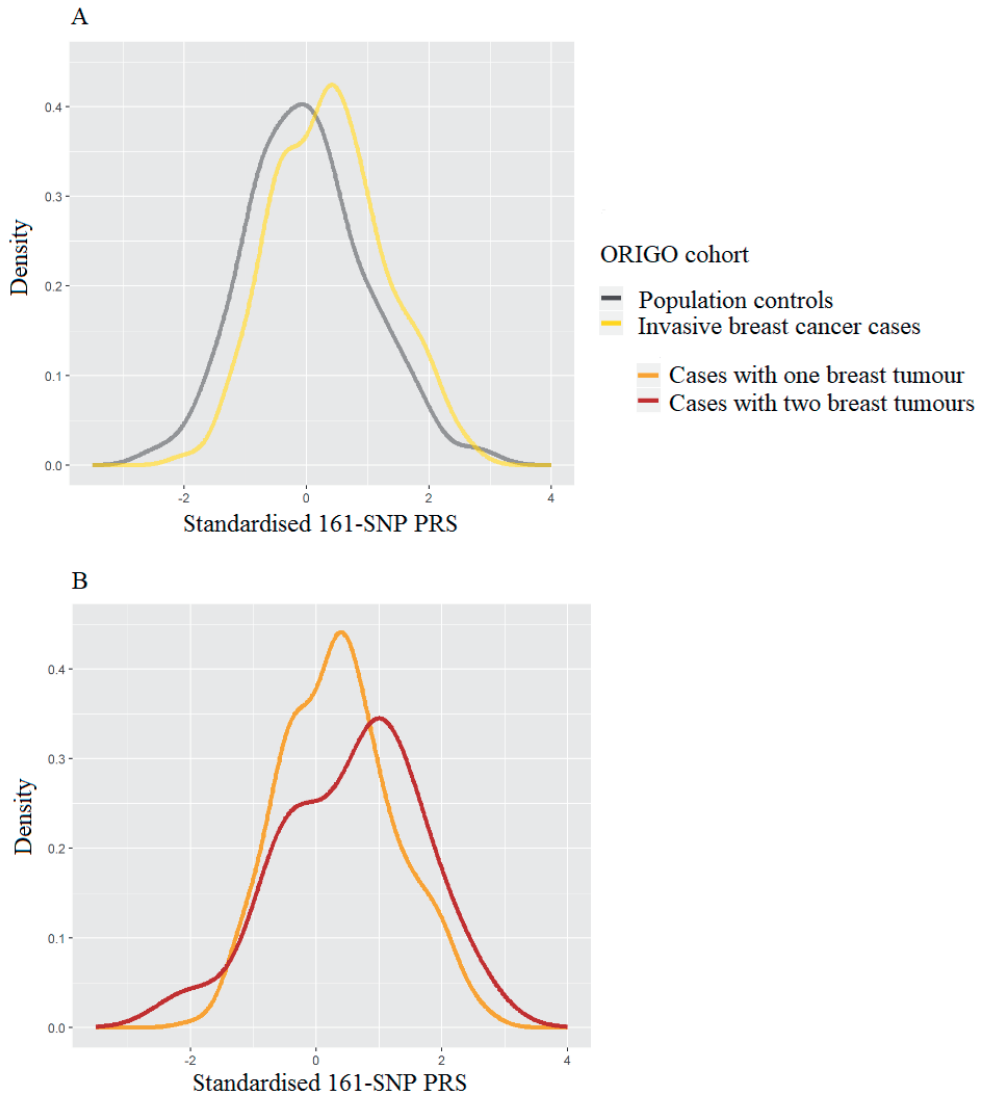


Figure S1: Distribution of the standardised 161-SNP PRS

The standardised 161-SNP PRS was plotted against the density in subgroups of the ORIGO cohort; (A) invasive breast cancer cases and population controls, (B) cases with one versus two breast tumours.

Table S1: 161 breast cancer associated SNPs used for calculating the Polygenic Risk Score²

See online material

Table S2: Dutch breast screening guideline (IKNL)⁵

	Low (<2)	Moderate (RR: 2-3)	High (RR: >3)
Life Time Risk	<20%	20-30%	>30%
Start screening	50 yr	40 yr	35 yr
Physical examination	-	-	+
Mammography	population screening*	<50 yr annual >50 yr population screening*	<60 yr annual >60 yr population screening*
MRI	-	-	-

*Biannual mammography

Table S3: Change in risk category for family breast cancer cases and unaffected relatives

	IKNL⁵		NICE⁶		NCCN⁷	
	Lower	Higher	Lower	Higher	Lower	Higher
Family breast cancer cases	7.4%	12.1%	5.0%	10.8%	5.0%	4.6%
Unaffected relatives	11.1%	9.2%	8.8%	4.6%	8.4%	5.3%

Percentages are based on the total number of family breast cancer cases and unaffected relatives, 323 and 262 respectively.

Following the Dutch IKNL guideline, cut off levels of 20% and 30% represent low, moderate and high risk categories. Following the NICE guideline, 17% and 30% represent low, moderate and high risk categories. Following the NCCN guideline, 20% represent a cut off level for the high risk category.

Table S4: Risk scores from individuals shown in Figure 3

	Individual	Standardised 161-SNP PRS	BOADICEA _{LTR}	BOADICEA _{sPRS}
Family A	III-1	0.62	0.35	0.37
	III-2	0.98	0.29	0.33
	III-3	0.86	0.40	0.44
	III-4	0.07	0.40	0.40
	IV-1	-1.54	0.24	0.20
	IV-2	-0.77	0.24	0.22
	IV-3	0.09	0.20	0.21
Family B	II-1	2.14	0.23	0.30
	II-2	0.20	0.32	0.33
	II-3	-0.66	0.32	0.29
	III-1	-0.62	0.20	0.18
	III-2	0.56	0.20	0.21
	III-3	-1.06	0.20	0.17
	III-4	0.81	0.20	0.22
	III-5	-0.56	0.23	0.21
	III-6	2.70	0.22	0.31
	III-7	0.16	0.21	0.22
Family C	III-1	1.06	0.27	0.31
	III-2	-0.50	0.31	0.29
	III-3	0.34	0.27	0.28
	III-4	0.12	0.27	0.28
	III-5	0.86	0.26	0.29
	IV-1	1.21	0.26	0.31
	IV-2	0.52	0.25	0.27
	IV-3	-1.42	0.21	0.18
	IV-4	-0.36	0.22	0.21
	IV-5	1.08	0.22	0.25
	IV-6	-0.27	0.22	0.21
	IV-7	0.95	0.22	0.25
	IV-8	0.63	0.22	0.24
	IV-9	-1.41	0.23	0.19
IV-10	0.17	0.30	0.31	
IV-11	0.28	0.31	0.32	
IV-12	-0.57	0.29	0.27	
IV-13	-0.14	0.29	0.29	
IV-14	-0.42	0.29	0.28	

161-SNP PRS, Polygenic Risk Score based on 161 breast cancer associated SNPs; BOADICEA_{LTR} breast cancer lifetime risk at age 80, based on BOADICEA alone; BOADICEA_{sPRS} 161-SNP PRS based individual breast cancer risk score.

Table S5: Mean and SD for ORIGO incident breast cancer cases subgroups

ORIGO cases subgroup	Number	Standardised 161-SNP PRS	
		Mean	SD
Non-invasive tumour*	44	0.21	0.95
Invasive tumour	313	0.37	0.91
1 invasive breast tumour	294	0.36	0.91
2 invasive breast tumours	19	0.56	1.11

*or unknown invasiveness

161-SNP PRS, Polygenic Risk Score based on 161 breast cancer associated SNPs; SD, Standard Deviation

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