

Evaluating Polygenic Risk Scores for Breast Cancer in Women of African Ancestry

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Abstract

Background: Polygenic risk scores (PRSs) have been demonstrated to identify women of European, Asian, and Latino ancestry at elevated risk of developing breast cancer (BC). We evaluated the performance of existing PRSs trained in European ancestry populations among women of African ancestry. **Methods:** We assembled genotype data for women of African ancestry, including 9241 case subjects and 10 193 control subjects. We evaluated associations of 179- and 313-variant PRSs with overall and subtype-specific BC risk. PRS discriminatory accuracy was assessed using area under the receiver operating characteristic curve. We also evaluated a recalibrated PRS, replacing the index variant with variants in each region that better captured risk in women of African ancestry and estimated lifetime absolute risk of BC in African Americans by PRS category. **Results:** For overall BC, the odds ratio per SD of the 313-variant PRS (PRS₃₁₃) was 1.27 (95% confidence interval [CI] = 1.23 to 1.31), with an area under the receiver operating characteristic curve of 0.571 (95% CI = 0.562 to 0.579). Compared with women with average risk (40th-60th PRS percentile), women in the top decile of PRS₃₁₃ had a 1.54-fold increased risk (95% CI = 1.38-fold to 1.72-fold). By age 85 years, the absolute risk of overall BC was 19.6% for African American women in the top 1% of PRS₃₁₃ and 6.7% for those in the lowest 1%. The recalibrated PRS did not improve BC risk prediction. **Conclusion:** The PRSs stratify BC risk in women of African ancestry, with attenuated performance compared with that reported in European, Asian, and Latina populations. Future work is needed to improve BC risk stratification for women of African ancestry.

Inherited genetic variation contributes to breast cancer (BC) risk, with approximately 30% of the total variability in liability due to genetic factors (1). Genome-wide association studies (GWAS) have discovered approximately 180 common risk variants for BC at genome-wide statistical significance levels (2-4), and polygenic risk scores (PRSs) comprised of multiple common variants have been demonstrated to stratify women at different levels of risk of developing BC (5). In the Breast Cancer Association Consortium (BCAC), a 313-variant PRS constructed using data from European-ancestry populations identified 1% of women with 4.4-fold and 2.8-fold increased risks of estrogen receptor (ER)-positive and ER-negative BC, respectively (6), compared with the population average (40th-60th PRS percentile). The incorporation of PRS into existing risk prediction models with nongenetic factors also improved risk stratification, with predicted lifetime risk over 30% for women in the top 1% of the PRS (7-9).

The discovery of risk variants for BC and subsequent PRS development has been based on studies primarily conducted among women of European ancestry. Broad clinical application of the PRS will require performance evaluation and enhancement across diverse racial and ethnic populations.

In this study, we assembled the largest genetic data set of BC in women of African ancestry to date, including 9241 cases (4299 with ER-positive and 2636 with ER-negative disease) and 10 193 controls, to examine the performance of the most current PRS panels for BC [179 (3,4) and 313 (6) variants] in stratifying risk in this population. We also recalibrated the PRSs by replacing index variants with markers that better captured BC risk in women of African ancestry and evaluated performance in this population.

Methods

Study Participants

This includes women of African ancestry from 4 BC consortia and 1 additional study, each genotyped with a different GWAS

array. Detailed descriptions of each consortium and study are provided in [Supplementary Table 1](#) (available online). The African American Breast Cancer consortium (10,11) includes genetic data from 3007 cases (1518 ER-positive, 987 ER-negative) and 2720 controls in the analysis; the African American Breast Cancer Epidemiology and Risk consortium (12) includes data from 1407 cases (952 ER-positive, 385 ER-negative) and 2408 controls; the BCAC/Genetic Associations and Mechanisms in Oncology (GAME-ON) OncoArray consortium (13) includes 2271 cases (1130 ER-positive, 613 ER-negative) and 1406 controls; the GWAS of Breast Cancer in the African Diaspora consortium (14) includes data for 1657 cases (403 ER-positive, 374 ER-negative) and 2029 controls, and the Ghana Breast Health Study (GBHS) includes 899 cases (296 ER-positive, 277 ER-negative) and 1630 controls. The study was approved by institutional review boards at each of the study sites.

Risk Variant Characteristics

We evaluated 2 PRSs. The first, PRS₃₁₃, is a 313-variant PRS developed by BCAC in European-ancestry populations (6). Briefly, the 313 variants were determined by *P* values and linkage disequilibrium threshold-based filtering followed by a stepwise forward selection. The second, PRS₁₇₉, included 179 known common risk variants that reached genome-wide statistical significance in GWAS analyses (3,4). Comparing these 2 sets of variants, 54 variants overlapped and another 52 were highly correlated ($r^2 \geq 0.8$ in the 1000 Genomes Project (1KGP) European-ancestry populations). In total, 163 of the 179 (91.1%) variants and 224 of the 313 (71.6%) variants had imputation scores ($r^2 \geq 0.8$ across all consortia and GBHS ([Supplementary Table 2](#), available online). For missing variants (not genotyped or imputed) in 1 study, we assigned each individual in that study the expected dosage derived from the remainder of studies; details are provided in the [Supplementary Methods](#) (available online).

Statistical Analysis

Association Testing of Individual Risk Alleles. Risk Allele Frequencies (RAFs) were derived by averaging the RAF in controls of each consortium and GBHS, weighted by the corresponding control numbers. We excluded variants with minor allele frequency (MAF) less than 0.1% or imputation quality score less than 0.3 in each consortium or study. Power calculations were conducted using odds ratios (ORs) in previous European GWAS (4) and RAFs in women of African ancestry. The odds ratio and *P* value of each variant were estimated using unconditional logistic regression adjusting for relevant covariates (Supplementary Table 1, available online) in each consortium and in GBHS, and the results were combined using a fixed-effect meta-analysis with inverse variance weights. Consistent directionality of effect was determined as alleles with odds ratios in the same direction as those previously reported (ie, $OR > 1$). A nominal *P* value of .05 was used to determine statistical significance.

PRS Analyses. For each individual *i*, a PRS was constructed as $PRS_i = \sum_{m=1}^C \beta_m g_m$, where g_m is the risk allele dosage at variant *m*, β_m is the weight for variant *m*, and *C* defines a set of risk loci. The weights for PRS₃₁₃ were those derived by Mavaddat et al. (6) in the development of PRS₃₁₃, while the weights for the 179 variant PRS (PRS₁₇₉) were the log odds ratios from GWAS of BC in women of European ancestry (4). For PRSs of BC subtypes, we used the corresponding subtype-specific weights from women of European ancestry (4,6). An ER-negative specific PRS was constructed using 15 risk alleles, which were identified to have stronger associations with triple-negative or nonluminal BC subtypes than ER-positive BC through a cluster analysis (PRS₁₅) (15). We categorized PRSs by percentile (<1%, 1%-5%, 5%-10%, 10%-20%, 20%-40%, 40%-60%, 60%-80%, 80%-90%, 90%-95%, 95%-99%, ≥99%) in controls, and the risk for each category was estimated relative to the 40%-60% reference group using logistic regression adjusting for the first 10 principal components (PCs), age, and study. We estimated odds ratios per unit SD and the area under the receiver operating characteristic curves (AUC) to compare with results from previous studies. We also computed theoretical AUC and the precision-recall curve (AUPRC) (16); details are provided in the Supplementary Methods (available online). In sensitivity analyses, excluding variants based on their imputation quality did not affect the performance of PRSs (Supplementary Table 3, available online).

We examined whether age modified the association between PRS and BC risk in stratified models by age category (<40 years, 40-49 years, 50-59 years, 60-69 years, ≥70 years) and the interaction with age. We compared distributions of PRSs in controls and the performance of PRSs between African countries (Ghana, Nigeria in Breast Cancer in the African Diaspora consortium, and the Women of African Ancestry Breast Cancer Study (WAABCS) study in the OncoArray consortium) and those of admixed US and Barbadian origin. We also assessed the interaction between PRS with ancestry (PC1) and family history on BC risk in African Americans and Barbadians; family history information was not available for samples from African countries.

The lifetime absolute risk of BC by PRS category among women with African and European ancestry were estimated taking into account the competing risk of dying from causes other than BC, which was described elsewhere (5). Inputs included the odds ratios of PRS₃₁₃ estimated in women of African

ancestry and European ancestry (6); age-specific BC incidence rates from the Surveillance, Epidemiology and End Results program (2000-2016); and mortality rates from the National Center for Health Statistics, Centers for Disease Control and Prevention (CDC) (2000-2016). The absolute risk for BC subtypes did not account for the competing risk of other subtypes. We also computed the lifetime risk with the odds ratios of continuous PRS₃₁₃ (per SD) within the age range of 35 to 85 years using the R package iCARE 1.18.0 (17).

Recalibrated PRS. For the 179-variant PRS, we examined whether the performance could be improved by constructing recalibrated PRSs. To avoid overestimating performance, we implemented 150 repeated fourfold cross-validations in which samples in each consortium or study were randomly split into 4 nonoverlapping parts; in each round, 1 part was left out as the testing set and the other 3 parts were used as training sets. We estimated log odds ratios (weights) and selected markers in training sets and then applied the results to testing sets. First, we examined weights based on the marginal log odds ratios derived from women of African ancestry (PRS_{179.AFR}) and from multi-ethnic populations of African and European ancestry (PRS_{179.ME}). Second, for each risk variant, we considered alternative markers within each region that might be more informative than the index variant ("better markers") in women of African ancestry (PRS_{179.AFR-better}) or in the multi-ethnic populations (PRS_{179.ME-better}). We assessed the variability of log odds ratios of the 179 risk loci in training sets by calculating the root-mean-squared deviation from effect sizes estimated using all samples. Details are provided in the Supplementary Methods (available online).

All statistical analyses were conducted using R v.4.0.0. All tests for statistical significance used a 2-sided alpha of .05.

Results

Participant Characteristics

The analysis included 9241 BC cases (4299 ER-positive and 2636 ER-negative) and 10193 controls. The characteristics of the participants by study are described in Supplementary Table 1 (available online). The mean age of cases ranged from 45 years to 71 years across studies.

Association Testing of Individual Risk Alleles

The comparison of RAF and effect size of individual variants between women of European and African ancestry are summarized in Supplementary Figure 1 and Supplementary Tables 4 and 5 (available online). We had 80% power to detect the reported European effect sizes of overall BC for 42 of the 179 variants and for 48 of the 313 variants. Of the 179 previously reported BC risk loci, 177 were polymorphic with MAF of at least 0.1% and imputation quality score of at least 0.3 in at least 2 of the 4 consortia or studies. Of the 177, 133 (75.1%) had consistent direction of effect on overall BC risk, and 29 (16.4%) were nominally statistically significant ($P < .05$). Of the 313 variants in PRS₃₁₃, 311 had a MAF greater than 0.1% and imputation score greater than 0.3 in at least 2 of the 4 consortia or studies, among which 215 (69.1%) showed directional consistency and 47 (15.1%) were nominally statistically significantly associated with overall BC risk.

Table 1. Association between PRS and overall breast cancer risk in women of African ancestry

PRS and AUC ^a	Controls, No.	PRS ₁₇₉			PRS ₃₁₃		
		Cases, No.	OR (95% CI) ^b	P ^c	Cases, No.	OR (95% CI) ^b	P ^c
PRS category, %							
<1	106	58	0.62 (0.44 to 0.89)	.009	60	0.63 (0.44 to 0.89)	.009
1-5	406	248	0.70 (0.58 to 0.84)	1.43 × 10 ⁻⁴	235	0.70 (0.58 to 0.85)	2.29 × 10 ⁻⁴
5-10	508	306	0.67 (0.56 to 0.79)	2.25 × 10 ⁻⁶	309	0.72 (0.61 to 0.85)	9.51 × 10 ⁻⁵
10-20	1020	687	0.74 (0.65 to 0.84)	2.84 × 10 ⁻⁶	681	0.74 (0.65 to 0.84)	3.52 × 10 ⁻⁶
20-40	2038	1598	0.89 (0.81 to 0.99)	.03	1560	0.84 (0.76 to 0.93)	6.92 × 10 ⁻⁴
40-60	2038	1808	1.00 (Referent)	—	1802	1.00 (Referent)	—
60-80	2038	2033	1.10 (1.00 to 1.22)	.045	1952	1.03 (0.94 to 1.14)	.50
80-90	1019	1093	1.17 (1.04 to 1.32)	.008	1197	1.29 (1.15 to 1.45)	1.22 × 10 ⁻⁵
90-95	508	645	1.42 (1.23 to 1.65)	1.80 × 10 ⁻⁶	643	1.38 (1.20 to 1.60)	1.28 × 10 ⁻⁵
95-99	406	593	1.60 (1.37 to 1.87)	2.15 × 10 ⁻⁹	602	1.61 (1.38 to 1.87)	1.32 × 10 ⁻⁹
>99	106	172	1.83 (1.39 to 2.40)	1.56 × 10 ⁻⁵	200	2.01 (1.53 to 2.63)	3.69 × 10 ⁻⁷
Continuous PRS per 1 SD	10 193	9241	1.26 (1.22 to 1.30)	2.59 × 10 ⁻⁴⁵	9241	1.27 (1.23 to 1.31)	4.23 × 10 ⁻⁴⁹
AUC			0.568 (0.56 to 0.576)	—		0.571(0.562 to 0.579)	—

^aAUCs were adjusted for study and the first 10 principal components. AUC = area under the receiver operating characteristic curve; CI = confidence interval; GBHS = Ghana Breast Health Study; OR = odds ratio; PRS = polygenic risk score.

^bOdds ratios were estimated using unconditional logistic regression model, adjusting for age, study, and the first 10 principal components in each consortium and in the GBHS and then combined using a fixed-effect meta-analysis with inverse variance weights.

^cTwo-sided P values were Wald P value from fixed-effect meta-analysis.

Polygenic Risk Score

Both PRS₁₇₉ and PRS₃₁₃ were associated with risk, with slightly greater effect sizes and AUCs observed for PRS₃₁₃ (Table 1). The actual threshold of percentiles of PRSs was shown in Supplementary Table 6 (available online). Women in the top 10% and 1% of the PRS₃₁₃ had a 1.54-fold (95% CI = 1.38-fold to 1.72-fold) and a 2.01-fold (95% CI = 1.53-fold to 2.63-fold) elevated risk compared with women at average risk (PRS in 40th-60th percentiles), respectively (Table 1; Figure 1). The odds ratio per 1 SD of PRS₃₁₃ was 1.27 (95% CI = 1.23 to 1.31), the AUC was 0.571 (95% CI = 0.562 to 0.579), and the AUPRC was 0.539 (baseline = 0.476).

For ER-positive BC, compared with the population average, women in the top 10% and 1% of PRS₃₁₃ had a 1.85-fold (95% CI = 1.61-fold to 2.13-fold) and a 2.16-fold (95% CI = 1.56-fold to 3.00-fold) increased risk, respectively (Table 2). The odds ratio per 1 SD of PRS₃₁₃ was 1.37 (95% CI = 1.32 to 1.43), the AUC was 0.588 (95% CI = 0.577 to 0.599), and the AUPRC was 0.368 (baseline = 0.297). The theoretical AUC of PRS₃₁₃ was 0.588 for African-ancestry women and 0.643 for women of European ancestry. For ER-negative BC, compared with women at average risk, those in the top 10% and 1% of PRS₃₁₃ had a 1.47-fold (95% CI = 1.25-fold to 1.74-fold) and a 2.18-fold (95% CI = 1.50-fold to 3.16-fold) increased risk, respectively (Table 3). The PRS₁₇₉ performed slightly better than PRS₃₁₃, with an AUC of 0.578 vs 0.562, an AUPRC of 0.256 vs 0.246 (baseline = 0.205), and an odds ratio per 1 SD of 1.31 vs 1.21, respectively. The PRS₁₅ performed similarly to PRS₁₇₉ (Table 3). The theoretical AUC of PRS₃₁₃ was 0.554 for African-ancestry women and 0.604 for European-ancestry women.

We did not observe a statistically significant interaction between either PRS and age at diagnosis for overall or subtype-specific BC risk (Supplementary Table 7; Supplementary Figure 2, available online). The average PRS was greater in controls from studies in Africa compared with those from studies of US and Barbadian origins ($P < 2.2 \times 10^{-16}$ for both PRS₃₁₃ and PRS₁₇₉; Supplementary Figure 3, available online). The performance of the PRSs by country of origin, ancestry proportion

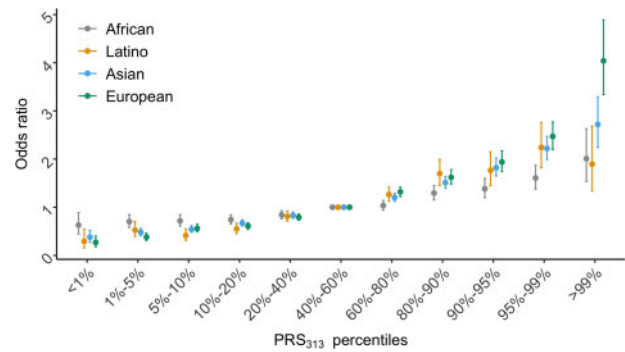


Figure 1. Association between the categorical 313-variant polygenic risk score (PRS₃₁₃) and overall breast cancer risk by population. The x-axis indicates the PRS₃₁₃ percentiles. The y-axis represents odds ratio (OR) values for the indicated PRS₃₁₃ percentiles compared with the 40%-60% category of PRS₃₁₃ as the reference (population average risk). Dots represent odds ratios and error bar lines represent standard error estimated in each population. The grey line represents results for women of African ancestry, which were estimated in this study. The yellow line represents results for Latinas obtained from a previous Latino PRS study (10). The blue line represents results for Asian women obtained from an Asian PRS study (18). The green line represents results for women of European ancestry obtained from a previous European ancestry PRS study (6). African-ancestry results are also provided in Table 1.

defined by PC1, or family history are shown in Supplementary Tables 7-9.

The distribution and root-mean-square deviation (RMSD) of log odds ratios of the “better markers” and the 179 known risk variants in the 600 training sets are shown in Supplementary Figure 4 and Supplementary Table 10 (available online). When estimating the 4 recalibrated PRSs for overall BC, we found the PRS_{179,ME-better}, which used “better markers” and variant weights derived from the multi-ethnic training sets, had the largest AUC but did not differ from that of PRS₁₇₉ (average AUC = 0.569 vs AUC = 0.566; Supplementary Figure 5, available online).

The estimated lifetime absolute risks for African American and European-ancestry women by PRS₃₁₃ categories for overall and subtype-specific BC are shown in Figure 2. By age 85 years,

Table 2. Association between PRS and ER-positive breast cancer risk in women of African ancestry

PRS and AUC ^a	PRS ₁₇₉				PRS ₃₁₃			PRS ₃₁₃ in European women ^d
	Controls, No.	Cases, No.	OR (95% CI) ^b	P ^c	Cases, No.	OR (95% CI) ^b	P ^c	OR (95% CI)
PRS category, %								
<1	106	25	0.54 (0.33 to 0.87)	.01	28	0.60 (0.37 to 0.95)	.03	0.16 (0.09 to 0.30)
1-5	406	93	0.51 (0.39 to 0.66)	4.75 × 10 ⁻⁷	89	0.56 (0.43 to 0.73)	1.92 × 10 ⁻⁵	0.32 (0.25 to 0.40)
5-10	508	152	0.62 (0.50 to 0.77)	2.10 × 10 ⁻⁵	128	0.63 (0.50 to 0.79)	7.12 × 10 ⁻⁵	0.50 (0.42 to 0.60)
10-20	1020	323	0.71 (0.60 to 0.83)	4.11 × 10 ⁻⁵	288	0.69 (0.59 to 0.82)	3.17 × 10 ⁻⁵	0.61 (0.53 to 0.69)
20-40	2038	736	0.83 (0.73 to 0.95)	.007	723	0.86 (0.75 to 0.98)	.02	0.77 (0.70 to 0.85)
40-60	2038	856	1.00 (Referent)	—	807	1.00 (Referent)	—	1.00 (Referent)
60-80	2038	918	1.07 (0.94 to 1.21)	.32	934	1.13 (1.00 to 1.29)	.05	1.40 (1.28 to 1.52)
80-90	1019	496	1.13 (0.97 to 1.31)	.11	541	1.30 (1.12 to 1.51)	6.18 × 10 ⁻⁴	1.59 (1.44 to 1.76)
90-95	508	302	1.43 (1.19 to 1.72)	1.71 × 10 ⁻⁴	350	1.75 (1.46 to 2.10)	1.52 × 10 ⁻⁹	2.17 (1.93 to 2.44)
95-99	406	306	1.80 (1.48 to 2.19)	2.56 × 10 ⁻⁹	315	1.89 (1.56 to 2.29)	7.97 × 10 ⁻¹¹	2.68 (2.37 to 3.03)
>99	106	92	2.21 (1.58 to 3.09)	3.31 × 10 ⁻⁶	96	2.16 (1.56 to 3.00)	4.12 × 10 ⁻⁶	4.37 (3.59 to 5.33)
Continuous	10 193	4299	1.33 (1.27 to 1.38)	4.35 × 10 ⁻⁴¹	4299	1.37 (1.32 to 1.43)	8.63 × 10 ⁻⁵¹	1.74 (1.66 to 1.82)
PRS per 1 SD								
AUC			0.576 (0.566 to 0.585)	—		0.588 (0.577 to 0.599)	—	0.651

^aAUCs were adjusted for study and the first 10 principal components. AUC = area under the receiver operating characteristic curve; CI = confidence interval; GBHS = Ghana Breast Health Study; OR = odds ratio; PRS = polygenic risk score.

^bOdds ratios were estimated using unconditional logistic regression model, adjusting for age, study, the first 10 principal components in each consortium and in the GBHS, and then combined using a fixed-effect meta-analysis with inverse variance weights.

^cTwo-sided P values were Wald P value from fixed-effect meta-analysis.

^dResults were obtained from a previous PRS study in women of European ancestry (6).

the absolute risk of overall BC was 19.6% for women in the top 1% of PRS₃₁₃ and 6.7% for women in the lowest 1%. The density plots of lifetime absolute risk of ER-positive and ER-negative BC in European and African American women are shown in [Supplementary Figure 6](#) (available online). The average lifetime risk of ER-positive BC from 35 years to 85 years is 9.1% (SD = 4.6%) for European-ancestry women and 6.6% (SD = 2.0%) for African American women, and for ER-negative BC the risks are 1.9% (SD = 0.7%) for European-ancestry women and 2.9% (SD = 0.6%) for African American women.

Discussion

We evaluated the performance of 2 PRSs developed in women of European ancestry in women of African ancestry and found both PRSs to be statistically significantly associated with overall and subtype-specific BC risk, with odds ratios per SD of 1.21 to approximately 1.37 and AUCs ranging between 0.57 and 0.59. Women in the top 10% of either PRS had a 1.54-fold elevated risk of BC. Women in the top 1% of the PRSs had a 1.83-fold to 2.01-fold increase in risk and were estimated to have an 18% to 20% lifetime risk of developing BC. However, these estimates are markedly lower than what have been reported for other racial or ethnic populations.

The authors of the largest PRS study of BC in women of European ancestry examined multiple PRSs using between 77 and 3820 risk variants, reported odds ratios per SD between 1.49 and 1.71, and AUCs between 0.61 and 0.64 for overall BC (6). The authors of the largest PRS study in Latinas examined the performance of a 71- and 180-variant PRS for overall BC and reported odds ratios per SD from 1.51 to 1.58 and AUCs of 0.61 to 0.63 (19) ([Figure 1](#)). Researchers examined in East Asians a 67-variant PRS for overall BC and reported an odds ratio per SD of 1.44 and an AUC of 0.61 (20,21). The authors of the largest PRS study in

Asians examined a 287-variant PRS (derived from the 313 variants in the European ancestry study) and reported an odds ratio per SD of 1.51 and AUC of 0.62 for overall BC (18). The weaker performance of PRS in women of African ancestry is consistent with observations for other cancers and other chronic diseases (22,23). Factors likely to be underlying the difference include variation in linkage disequilibrium patterns, allele frequencies, and potential effect heterogeneity between populations. The reduction in performance is also in agreement with previous studies that have demonstrated the decline in PRS performance with increasing genetic divergence from the training population (24-26).

To note, even though the AUC was worse in African-ancestry women than that in European-ancestry women, the risk stratification in the population, which depends also on the underlying rates of disease, may not be worse. This was demonstrated in comparison of ER-negative BC in our study: the AUC was greater in European-ancestry women (theoretical AUC = 0.604) than in African-ancestry women (theoretical AUC = 0.554); however, because of the greater baseline risk of ER-negative disease in African-ancestry women than in European-ancestry women (2.9% vs 1.9% average lifetime risk, respectively), for any given risk threshold, a larger percentage of African-ancestry women than European-ancestry women would be identified as being at elevated risk of this disease.

For BC subtypes, PRS₃₁₃ performed better in the prediction of ER-positive BC than ER-negative BC, which is consistent with the previous study in women of European ancestry (6). Both PRS₁₇₉ and PRS₁₅ performed better than PRS₃₁₃, which suggests PRS₃₁₃ is not optimal for ER-negative disease in women of African ancestry. The previous investigations of the PRS in women of European ancestry reported a weak nonlinear decline in effect with increasing age for ER-positive BC. However, we did not observe an interaction with age for overall or BC subtypes in this study. The previous BCAC study reported an attenuated odds ratio of PRS₃₁₃

Table 3. Association between PRS and ER-negative breast cancer risk in women of African ancestry

PRS and AUC ^a	PRS ₁₇₉			PRS ₁₅			PRS ₃₁₃			PRS ₃₁₃ in European women ^d
	Controls, No.	Cases, No.	OR (95% CI) ^b	P ^c	Cases, No.	OR (95% CI) ^b	P ^c	Cases, No.	OR (95% CI) ^b	
PRS category, %										
<1	106	17	0.68 (0.39 to 1.20)	.19	15	0.63 (0.35 to 1.12)	.12	22	0.86 (0.52 to 1.42)	.55
1-5	406	56	0.58 (0.42 to 0.80)	7.78 × 10 ⁻⁴	71	0.73 (0.55 to 0.97)	.03	90	0.94 (0.72 to 1.23)	.66
5-10	508	75	0.57 (0.44 to 0.76)	9.26 × 10 ⁻⁵	87	0.74 (0.57 to 0.96)	.02	86	0.71 (0.54 to 0.92)	.01
10-20	1020	192	0.75 (0.61 to 0.91)	.003	185	0.78 (0.64 to 0.95)	.01	191	0.78 (0.64 to 0.94)	.01
20-40	2038	462	0.90 (0.77 to 1.04)	.16	463	0.95 (0.82 to 1.11)	.54	454	0.87 (0.75 to 1.01)	.08
40-60	2038	523	1.00 (Referent)	—	492	1.00 (Referent)	—	517	1.00 (Referent)	—
60-80	2038	537	1.06 (0.91 to 1.22)	.46	563	1.14 (0.98 to 1.32)	.09	551	1.02 (0.88 to 1.18)	.80
80-90	1019	339	1.30 (1.10 to 1.55)	.002	341	1.33 (1.12 to 1.58)	.001	324	1.20 (1.01 to 1.43)	.03
90-95	508	190	1.42 (1.15 to 1.75)	.001	201	1.59 (1.29 to 1.96)	1.27 × 10 ⁻⁵	165	1.20 (0.96 to 1.50)	.11
95-99	406	187	1.80 (1.44 to 2.24)	1.51 × 10 ⁻⁷	172	1.84 (1.47 to 2.30)	1.01 × 10 ⁻⁷	175	1.66 (1.32 to 2.07)	9.12 × 10 ⁻⁶
>99	106	58	2.05 (1.41 to 2.97)	1.72 × 10 ⁻⁴	46	1.93 (1.31 to 2.86)	9.72 × 10 ⁻⁴	61	2.18 (1.50 to 3.16)	4.08 × 10 ⁻⁵
Continuous PRS per 1 SD	10193	2636	1.31 (1.24 to 1.37)	5.53 × 10 ⁻²⁷	2636	1.28 (1.22 to 1.34)	9.29 × 10 ⁻²⁵	2636	1.21 (1.15 to 1.27)	3.39 × 10 ⁻¹⁵
AUC			0.578 (0.564 to 0.591)	—		0.571 (0.557 to 0.583)	—		0.562 (0.551 to 0.573)	0.611

^aAUCs were adjusted for study and the first 10 principal components. AUC = area under the receiver operating characteristic curve; CI = confidence interval; GBHS = Ghana Breast Health Study; OR = odds ratio; PRS = polygenic risk score.

^bOdds ratios were estimated using unconditional logistic regression model, adjusting for age, study, the first 10 principal components in each consortium, and in the GBHS, and then combined using a fixed-effect meta-analysis with inverse variance weights.

^cTwo-sided P values were Wald P value from fixed-effect meta-analysis.

^dResults were obtained from a previous PRS study in women of European ancestry (6).

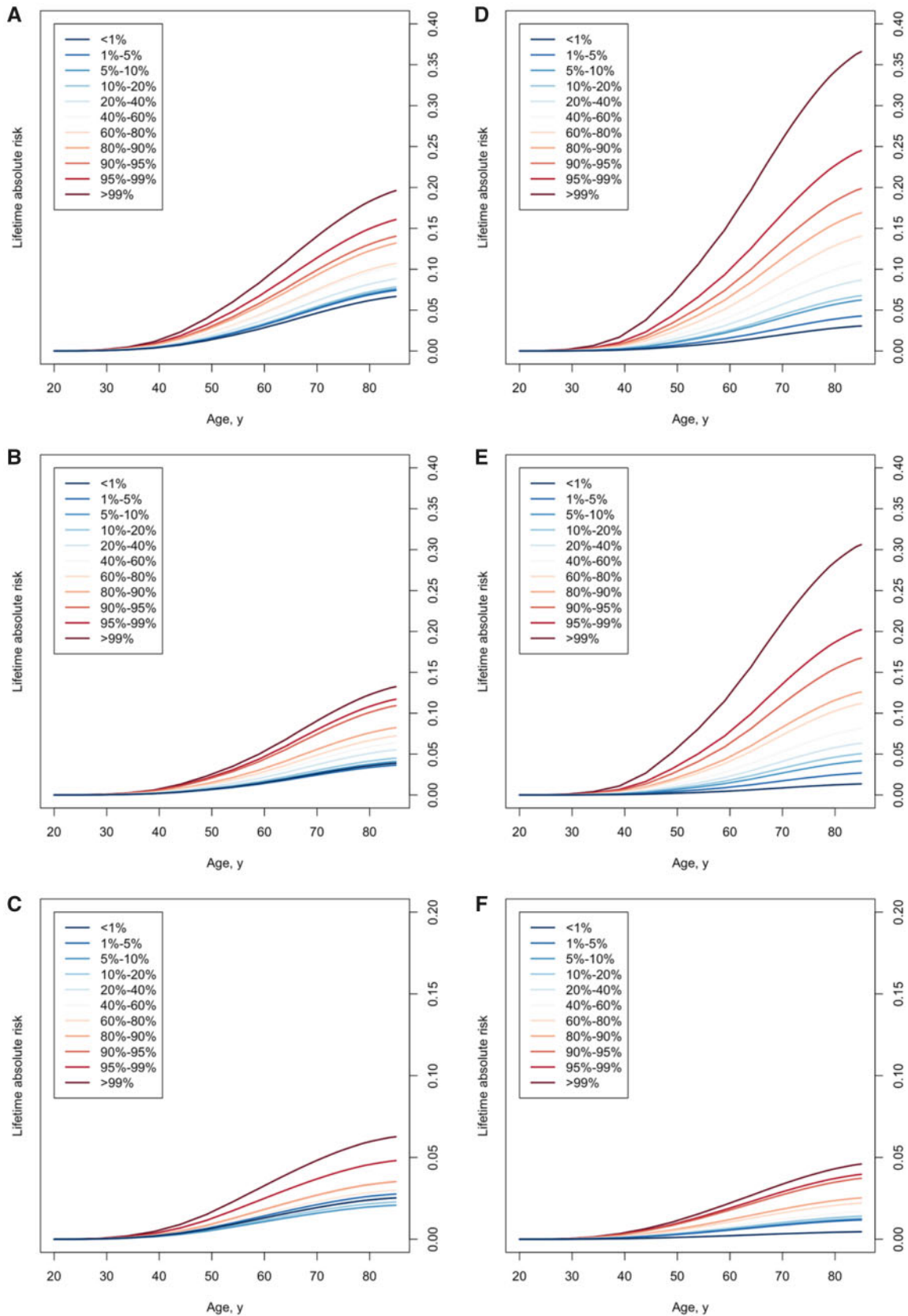


Figure 2. Lifetime absolute risk of breast cancer by polygenic risk score (PRS) category in African American women. Lifetime absolute risk of developing breast cancer for the 313-variant PRS (PRS₃₁₃) in African American women for overall breast cancer (A), estrogen receptor (ER)-positive breast cancer (B) and ER-negative breast cancer (C) and in European-ancestry women for overall breast cancer (D), ER-positive breast cancer (E), and ER-negative breast cancer (F). The x-axis represents age and the y-axis is the absolute risk of breast cancer by a given age. The different colored lines represent the corresponding PRS strata. See Methods for details about the calculation of absolute risks.

for ER-positive BC in women with family history (6). We detected a suggestive interaction of the PRS with family history, although larger studies will be needed to improve the power to detect interactions in women of African ancestry.

Based on the current sample size of African-ancestry women in this study, we found European weights provided optimal PRS performance—the performance could not be improved using weights estimated in women of African ancestry or by replacing an index variant with a “better African marker” in each risk region. This could be due to the training set not being large enough to provide accurate estimates of effect or to distinguish causal variants from correlated markers in each risk region. We observed a minor improvement in PRS using multi-ethnic weights, which is corroborated by a simulation study showing that multi-ethnic training populations substantially outperformed PRS using a single training population (27). The current multi-ethnic samples predominantly consisted of women of European ancestry. We expect that PRS performance may be further improved when including more samples of African ancestry.

The current guideline for BC screening classifies those at high risk based on BRCA1 or BRCA2 mutations and family history-based lifetime risk assessment (28–30). Several studies have reported that incorporating PRS into these existing risk models could increase discrimination accuracy and improve calibration (8,9,31–34). These studies and models were predominantly conducted in European ancestry populations, and similar assessments are needed in larger African-ancestry studies to assess the value of incorporating common germline variation into the decision-making process of screening recommendations. Of note, incidence rates of BC in West African countries are substantially lower than that in the United States (35). And given the differences in access to health care, the harm-to-benefit ratio and cost-effectiveness of introducing risk-stratified BC screening programs at the population level should be carefully assessed.

In conclusion, although PRSs developed in women of European ancestry can identify women at elevated risk of developing BC, they currently substantially underperform in African-ancestry women compared with women of European, Asian, and Latino ancestry.

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Data Availability

The datasets are publicly available via dbGaP for AABC (phs000851.v1.p1), ONCO (phs001265.v1.p1), AMBER (phs000669.v1.p1), GBHS (phs002387.v1.p1) and ROOT (phs000383.v1.p1).

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