Next Top Model: An Overview of Breast Cancer Risk Assessment Models

Description

Numerous risk assessment models are available to calculate a woman's lifetime risk of developing and/or carrying a gene mutation that may predispose her to developing breast cancer. Knowledge of risk may help to inform individual screening practices. Further understanding and evaluation of risk assessment models are needed to increase their utilization.

The purpose of this activity is to 1) introduce breast cancer risk assessment models for estimating an individual's risk of developing breast cancer or risk for carrying a gene mutation that may predispose to developing breast cancer, 2) review the strengths and limitations of each model and the model's applicability to underrepresented populations, and 3) provide an example that demonstrates the use of risk assessment models.

Learning Objectives

Upon completing this activity, the reader should be able to:

- Identify the available breast cancer risk assessment models.
- Describe the strengths and limitations of each model.
- Explain each model's applicability to underrepresented populations.

Target Audience

- Radiologists
- Related Imaging Professionals

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In 2023, an estimated 298,000 women in the United States were diagnosed with breast cancer.1 Incidence rates for breast cancer have been increasing by approximately 0.5% per year since the mid-2000s.² Significant racial disparities in breast cancer mortality exist, with mortality rates in Black women being approximately 40% higher than those for White women, despite similar incidence rates.1 Disparities are also prominent among young women. When comparing Black women age ≤50 years to White women in the same age group, mortality rates were 1.9-2.6 times higher in Black women versus 1.1-1.2 times higher in the groups aged ≥70 years.³

Identifying a diverse population of young women at high risk for breast cancer with dedicated risk assessment models can help to address these existing disparities in mortality. Clinicians perform breast cancer risk assessment by asking a series of

questions about such characteristics as family and breast health history and inputting the answers into an electronic tool, which calculates a woman's risk for breast cancer. Women identified as being at high risk (lifetime risk >20%) can be offered guideline-based early screening mammography and supplemental screening with MRI. This early or supplemental screening in high-risk women can identify cancers at earlier stages, prevent delays in diagnosis, and initiate earlier treatment. Ultimately, identification of a diverse population of young, high-risk women has the potential to improve mortality and address existing breast cancer disparities.

While numerous risk assessment models have been developed to identify women at high risk for developing breast cancer, these models remain underutilized in the clinical setting. The purpose of this activity is to

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 introduce breast cancer risk assessment models for estimating an individual's risk of developing breast cancer or risk for carrying a gene mutation that may predispose to developing breast cancer,

- 2) review the strengths and limitations of each model and the model's applicability to underrepresented populations, and
- provide a case study that demonstrates the use of risk assessment models.

Risk Factors for Breast Cancer

Many risk factors can increase an individual's likelihood of developing breast cancer (Figure 1).4 Nonmodifiable risk factors include increasing age, being born female, personal or family history of breast cancer, and inherited genetic changes in breast cancer susceptibility genes.4 Hormonal and reproductive risk factors include long menstrual history and having children later in life.4 Breastfeeding for at least one year can serve as a protective factor and decrease risk.4 Potentially modifiable risk factors include excess body weight, menopausal hormone therapy, physical inactivity, and excess alcohol consumption.4 Medical risk factors include high breast tissue density and history of radiation to

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Figure 1. Breast cancer risk factors include personal characteristics, hormonal/reproductive factors, personal breast history, family history.

Personal characteristics

Age Race/ethnicity BMI Obesity Diet Physical activity Alcohol intake Environmental exposure

Hormonal/ reproductive factors

Age at menarche Age at first live birth Age at menopause Hormone replacement therapy Oral contraceptive use

Personal breast history

Breast biopsies Atypical ductal hyperplasia Lobular carcinoma *in situ* Breast density

Family history

First/second/thirddegree relatives with breast cancer Age of onset Bilateral breast cancer Ovarian cancer Male breast cancer

the chest.⁴ Additional factors that can increase risk include history of breast biopsies and diagnosis of atypical hyperplasia or lobular carcinoma *in situ* (LCIS).⁴

Screening Guidelines Based on Calculated Lifetime Risk

The American College of Radiology (ACR) and Society of Breast Imaging (SBI) recommend risk assessment no later than age 25.5 For those at average risk, annual screening with mammography is recommended starting at age 40. If a patient is found to have a lifetime risk ≥20%, ACR guidelines recommend annual screening beginning at age 30 and annual breast MRI beginning at age 25-30. More specific guidelines are available for women with other factors that may increase personal risk for breast cancer (eg, history of chest radiation therapy; genetics-based increased risk; personal histories of breast cancer and dense breast tissue; family history of breast cancer at a young age; personal history of atypical ductal hyperplasia (ADH), atypical lobular hyperplasia, or LCIS).

National Comprehensive Cancer Network (NCCN) guidelines also recommend screening mammography for high-risk women at younger ages and supplemental screening with contrast-enhanced breast MRI.^{6,7} Specifically, patients with a lifetime risk ≥20% should receive an annual screening

mammogram beginning either at age 40 or 10 years prior to when the youngest family member was diagnosed with the disease, but not prior to age 30 (whichever comes first). Tomosynthesis is recommended. Additionally, patient with a lifetime risk ≥20% should either undergo annual breast MRI beginning at age 40 or 10 years prior to when the youngest family member was diagnosed with breast cancer, but not prior to age 25 (whichever comes first). If a patient cannot undergo MRI, then contrast-enhanced mammography or whole breast ultrasound should be considered. Additional guidelines are available for women with other high-risk factors (eg, those with thoracic radiation between ages 10 and 30, increased 5-year risk of invasive breast cancer, ADH and ≥20% lifetime risk, lobular neoplasia and ≥20% lifetime risk, or pedigree suggestive of genetic predisposition).

Overview of Risk Assessment Models

Gail Model

The Gail model is one of the earliest models of breast cancer risk assessment, first published in 1989.⁸ The data was derived from 243,221 White women in the Breast Cancer Detection Demonstration Project (BCDDP), a screening program conducted between 1973 and 1980 in the United States.⁹ This model was modified in

1992 by the National Surgical Adjuvant Breast and Bowel Project (NSABP) to estimate the absolute risk of developing only invasive breast cancer.10 Researchers have made additional updates to the model to provide more accurate estimates for Black women, Asian and Pacific Islander women, and Hispanic women. The modified model is used in the National Cancer Institute's (NCI) Breast Cancer Risk Assessment Tool. It estimates a patient's 5-year and lifetime risk of developing invasive breast cancer and is available at https://bcrisktool. cancer.gov/.11

Factors included in the Gail model are age, race/ethnicity, age at menarche, age at first live birth, number of previous breast biopsies, presence of atypical hyperplasia on biopsy, and number of affected first-degree relatives.

While easily accessible, wellcalibrated to provide moderate discriminatory accuracy in studies of predominantly White women, and updated to provide more accurate estimates among different populations, the Gail model may underestimate risk in certain populations, such as Black women with previous biopsies and Hispanic women born outside the United States.¹²⁻¹⁷ Additionally, owing to a lack of data, the model may be inaccurate among Native American and Alaskan Native women.

The Gail model has additional weaknesses that should be considered.

For example, it does not account for family history of breast cancer beyond affected first-degree relatives, thereby excluding paternal family history. For this reason, the NCCN guidelines do not list it as a model that should be used to identify candidates for supplemental screening with breast MRI. Instead, a patient found to have a 5-year risk of invasive breast cancer >1.7% in individuals age >35 per the Gail model should receive an annual screening mammogram (to begin when identified as being at increased risk), with tomosynthesis if available.

Furthermore, the Gail model does not include the age at onset of breast cancer among relatives or family history of other cancers. The model also does not consider variables such as mammographic density and should not be used in women under the age of 35. Additionally, the Gail model should not be used in women with known *BRCA1/ BRCA2* mutations, those with a previous history of breast cancer, those with prior treatment of Hodgkin lymphoma with radiation to the chest, or those with breast cancer-causing syndromes, including Li-Fraumeni syndrome.

International Breast Cancer Intervention Study [IBIS]/Tyrer-Cuzick (version 8)

The Tyrer-Cuzick model, also known as the IBIS model, was developed by scientists at the Wolfson Institute of Preventive Medicine, Queen Mary University of London.18,19 The Tyrer-Cuzick model was developed using data on first breast cancer diagnoses among women in the United Kingdom (Thames Cancer Registry) between 2005-2009. Familial risk is based on data from a Swedish population-based study.19 The latest version of the model (version 8) incorporates additional risk factors such as breast density and single nucleotide polymorphisms (SNPs). The model estimates an individual's 10-year and lifetime risk for developing breast cancer, as well as the likelihood of

Table 1: Summary of Risk Assessment Models (selected factors)					
	GAIL	TYRER-CUZICK	BCSC	BRCAPRO	CANRISK
Personal characteristics					
Age	Х	Х	Х	Х	х
Race/ethnicity	Х		Х	Х	
Body mass index		Х			х
Alcohol intake					х
Hormonal/reproductive					
Age at menarche	Х	Х			х
Age at first live birth	х	Х			Х
Age at menopause		Х			х
Hormone replacement therapy use		х			х
Oral contraceptive use					х
Personal breast history					
Prior breast biopsies	Х	х	Х		
Atypical hyperplasia	Х	Х	Х		
Lobular carcinoma in situ		Х	Х		
Breast density		Х	Х		х
Family history					
Number of affected first- degree relatives	х	х	х	х	х
Number of affected second- degree relatives		х		Х	х
Number of affected third- degree relatives		х			х
Age of onset of breast cancer		х		Х	х
Bilateral breast cancer		Х		х	х
Ovarian cancer		Х		Х	х
Male breast cancer		х		Х	х

carrying a *BRCA1* or *BRCA2* gene. It is available at https://ibis.ikonopedia.com/ and https://ems-trials.org/ riskevaluator/^{20,21}

Factors included in the Tyrer-Cuzick model are age, body mass index, reproductive history (age at menarche, age at first live birth, age at menopause), exogenous hormone exposure (hormone replacement therapy duration), results of previous breast biopsy (hyperplasia, presence of LCIS or atypical hyperplasia), breast density, family history (number and age of onset of first, second, or third-degree relatives with breast cancer, ovarian cancer diagnoses, male breast cancer diagnoses, unaffected relatives), Ashkenazi Jewish origin, and previous genetic test results (*BRCA1/2*).

External validation studies for the current version of the Tyrer-Cuzick model are ongoing. The Tyrer-Cuzick model has been found to be well-calibrated overall in non-Hispanic White women and Black women, with good calibration for Asian/Pacific Islanders and Native Americans, although sample sizes were small.²² The model may overestimate risk for Hispanic women.²² The addition of mammographic density was found to increase the discriminatory accuracy of the model.²³ One study suggests the Tyrer-Cuzick model may overestimate risk for women at the highest-risk decile.²⁴

A major strength of the Tyrer-Cuzick model is that it includes a diverse range of risk factors, including breast density and a comprehensive family history. Like the Gail model, it is easily accessible online and has undergone periodic updates to incorporate additional data on breast cancer incidence. Unlike the Gail model, the Tyrer-Cuzick model can be used in women ages <35 years and can calculate the risk for *BRCA1 or BRCA2* mutations.

The Tyrer-Cuzick model should not be used to assess risk in women who have already been diagnosed with breast cancer and may overestimate risk in women with atypical hyperplasia and LCIS.^{25,26}

The Breast Cancer Surveillance Consortium (BCSC) Risk Calculator version 2.0

The BCSC Risk Calculator model was developed in 2008 using data from 1,095,484 women in seven mammography registries participating in the NCI-funded BCSC in the United States.27 The study included women ages ≥35 with at least 1 mammogram with breast density measured using the Breast Imaging Reporting and Data System (BI-RADS) classification system. The model was updated in 2015 (version 2) to include benign breast diagnoses.²⁸ The BCSC Risk Calculator estimates a patient's 5- and 10-year risk of developing invasive breast cancer and is available at: https://tools.bcsc-scc.org/BC5yearRisk/calculator.htm.29

Factors included in the BCSC Risk Calculator are age, race/ethnicity, history of first-degree relatives with breast cancer (yes/no), history of a breast biopsy with benign breast disease diagnoses if known, and BI-RADS breast density.

The original model was externally validated among patients in the Mayo

Mammography Health Study (MMHS) cohort.³⁰ Version 2 of the model was validated in a cohort of women in Chicago and was well-calibrated but found to underestimate risk in younger women, Hispanic and non-Hispanic Black women, and those with almost entirely fat breast density.²⁸

The major strengths of the BCSC Risk Calculator are that it incorporates BI-RADS breast density and is easily accessible. This model cannot be used in women with a previous diagnosis of breast cancer or DCIS, prior breast augmentation, prior mastectomy, or those aged <35 or >74. Additionally, it does not account for a family history of breast cancer beyond affected first-degree relatives, thereby excluding paternal family history.

BRCAPRO

The BRCAPRO model was developed in 1997 based on estimates of BRCA1 mutation frequencies in the general population and age-specific incidence rates of breast and ovarian cancers in carriers and noncarriers of mutations.³¹ It was expanded in 1998 to include BRCA2.32 The model uses Mendelian genetics and Bayes' theorem to calculate a patient's likelihood of carrying a germline mutation in the BRCA1 or BRCA2 genes, developing invasive breast cancer, or developing contralateral breast cancer. Access to the model can be requested at https://projects. iq.harvard.edu/bayesmendel/bayesmendel-r-package.33

Factors included in the BRCA-PRO model are age, race/ethnicity, number/age at onset of first or second-degree relatives with breast cancer, family history of bilateral breast cancer or male breast cancer, personal or family history of ovarian cancer, and Ashkenazi Jewish origin.

Validation studies demonstrate variation in the performance of the BRCAPRO model with some studies demonstrating appropriate performance and other studies finding the model to underpredict risk.^{34,35} Like the Tyrer-Cuzick and the CanRisk models, one of the strengths of the BRCAPRO model is its ability to assess the likelihood of carrying a *BRCA1* and/or *BRCA2* gene mutation. The model also considers information about unaffected relatives and is routinely updated.

The BRCAPRO model does not include non-hereditary risk factors, such as age at menarche, age at first live birth, age at menopause, or specific results of prior breast biopsies. It also excludes family history of third-degree relatives.

Additionally, this model may underestimate risk in patients without *BRCA* gene mutations, as well as in families with prostate or ovarian cancer.³⁶ The model is not immediately accessible; however, it can be requested through an online form.

CanRisk (BOADICEA v5)

The Breast and Ovarian Analysis of Disease Incidence and Carrier Estimation Algorithm (BOADICEA) model calculates the probabilities of carrying rare loss-of-function variants in several breast or ovarian cancer susceptibility genes in addition to estimating the risk of developing breast and ovarian cancer.³⁷⁻³⁹ It has undergone numerous updates since its development in 2002 and incorporates the effects of common genetic variants (summarized as polygenic risk scores, PRS), pathogenic variants in other genes, mammographic density, and additional risk factors. The latest version of the model (v6) is available to use via a web tool called CanRisk (https://www. canrisk.org/).37,38,40-44

Factors included in the CanRisk Tool for breast cancer risk estimation are age, body mass index, height, daily alcohol intake, age at menarche, age at first live birth, use of menopause hormone therapy, use or oral contraception, parity, mammographic density, family and personal-proband history of breast, ovarian, and pancreatic cancer, rare pathogenic variants in moderate and high-risk susceptibility genes, age information on unaffected family members, information on year of birth to capture birth cohort, Ashkenazi Jewish origin, and common cancer genetic susceptibility variants (Polygenic Risk Scores).⁴⁵

The model has been validated in several studies, largely consisting of women of European ancestry, and has been found to be well-calibrated.^{24,46-48} However, it may not be as reliable in populations at lower risk for breast cancer or those of non-European ancestry.⁴⁹

Like the Tyrer-Cuzick model, the CanRisk Tool includes a diverse range of risk factors, including comprehensive family history. Additionally, this is the only model that includes lifestyle risk factors such as alcohol consumption. The model is easily accessible online but requires the user to create an account for access. Unlike other models, it can be used in patients with a previous diagnosis of breast cancer.

The CanRisk tool should not be used in patients with personal history ductal carcinoma *in situ* (DCIS). Additionally, it will underestimate risk in those with Ataxia-Telangiectasia or homozygous carriers of pathogenic CHEK2 pathogenic truncating variants and should not be used in these patients. The CanRisk tool does not incorporate information on prior breast biopsies (number or result).

S Father 70y 1953 Sister Patien Patien

Table 1 provides a summary of the factors included in several risk assessment models.

Case Study

47y 1976

breast ca: 44

A 37-year-old White female wishes to know her lifetime risk for developing breast cancer. She has no significant medical history and is not Ashkenazi Jewish. She has never been tested for genetic mutations, had a mammogram, or had a breast biopsy. Menarche was at age 14, and she is premenopausal. She had a daughter at the age of 22, who is currently 15 and healthy. The patient's mother (diagnosed at age 50, deceased age 65) and sister (diagnosed at age 44, alive (currently 47) had unilateral breast cancer (Figure 2). Her father is living, age 70, and healthy. There is no family history of ovarian cancer. Genetic testing for the patient's relatives has never been performed. She is 5 foot 4 inches, weighs 150 pounds, and does not drink alcohol. She has never used hormone replacement therapy or oral contraceptives. She has never had an SNP array/ PRS calculated.

15v 200

37v 1986

What is the patient's lifetime risk for developing breast cancer? What are the appropriate breast cancer screening recommendations?

Risk Model Assessment

The Tyrer-Cuzick and CanRisk models calculate the patient's lifetime

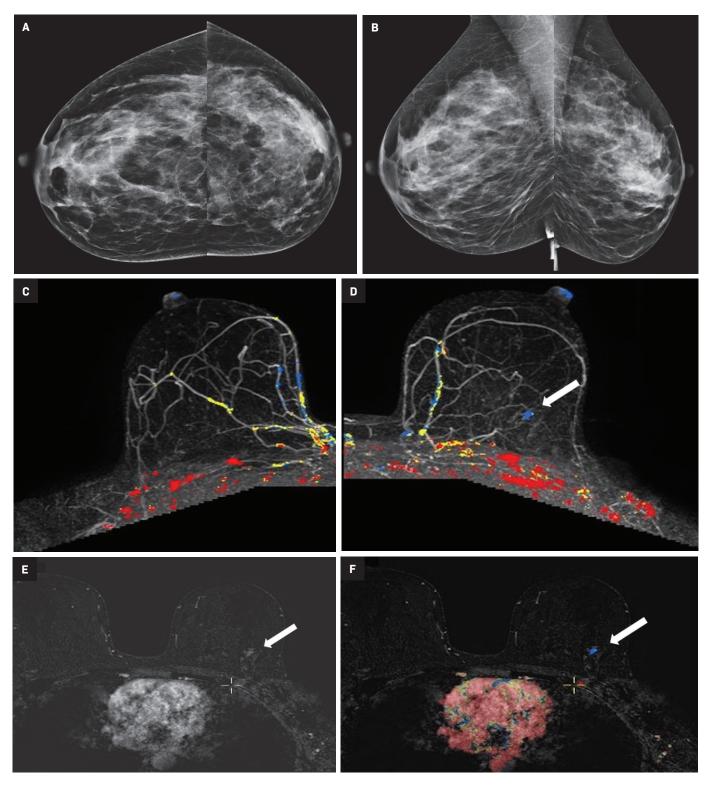
Figure 3. Lifetime breast cancer risk as calculated by the Tyrer-Cuzick (21.1%), BRCAPRO (13.0%), and CanRisk (22.3%) models for case study 1.

21.1%	Tyrer-Cuzick	 Considered age, BMI, age at menarche, age at first live birth, number of affected first and second-degree relatives with breast cancer, age of onset of breast cancer Did not consider race
13.0%	BRCAPRO	 Considered age, race, number of affected first and second-degree relatives with breast cancer, age of onset of breast cancer Did not consider BMI, age at menarche, age at first live birth
22.3%	CanRisk	 Considered age, BMI, age at menarche, age at first live birth, number of affected first and second-degree relatives with breast cancer, age of onset of breast cancer Did not consider race

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Figure 4. Bilateral craniocaudal (A) and mediolateral oblique screening mammogram (B) from the patient in the case study demonstrates heterogeneously dense breast tissue with no focal abnormality. Bilateral axial MIP images (C,D) demonstrate focal, clumped non-mass enhancement in the left outer breast. Bilateral axial postcontrast subtraction images (E,F) redemonstrate focal clumped non-mass enhancement with initial fast and delayed persistent kinetics in the left outer breast. MRI biopsy of the focal, clumped non-mass enhancement was performed. Pathology demonstrated DCIS with microinvasive component.



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risk for breast cancer as ≥20%. The BRCAPRO model calculates the patient's lifetime risk for breast cancer as <20%. Figure 3 depicts the risk assessment values for each model and the factors included in each model.

Based on the results of the Tyrer-Cuzick and CanRisk models, the patient is considered high risk. Per NCCN guidelines, they should consider screening mammography and screening MRI 10 years prior to the age of diagnosis of the youngest first-degree relative, but not before age 30. Because the patient's sister was diagnosed with breast cancer at age 44, screening mammography and screening MRI could have been considered as early as age 34.

The patient pursued screening mammography and MRI (Figure 4). The mammogram was normal, with dense breast tissue. Screening MRI demonstrated focal clumped non-mass enhancement in the left outer breast. MRI-guided biopsy was performed and revealed DCIS with a microinvasive component.

Conclusion

Numerous risk assessment models are available to calculate a woman's lifetime risk of developing and/or carrying a gene mutation that may predispose her to developing breast cancer. Knowledge of risk may help to inform individual screening practices. Risk assessment models have different strengths and weaknesses that may increase or limit use in certain populations. Further understanding and evaluation of risk assessment models are needed to increase their utilization. Increased breast cancer risk assessment among diverse populations can identify women who may be at high risk for breast cancer. Guideline-based breast cancer screening in these populations serves as an opportunity to address known breast cancer mortality disparities.

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