

Original Article

Modeling the impact of population screening on breast cancer mortality in the United States[‡]

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SUMMARY

Objective: Optimal US screening strategies remain controversial. We use six simulation models to evaluate screening outcomes under varying strategies.

Methods: The models incorporate common data on incidence, mammography characteristics, and treatment effects. We evaluate varying initiation and cessation ages applied annually or biennially and calculate mammograms, mortality reduction (vs. no screening), false-positives, unnecessary biopsies and over-diagnosis.

Results: The lifetime risk of breast cancer death starting at age 40 is 3% and is reduced by screening. Screening biennially maintains 81% (range 67% to 99%) of annual screening benefits with fewer false-positives. Biennial screening from 50–74 reduces the probability of breast cancer death from 3% to 2.3%. Screening annually from 40 to 84 only lowers mortality an additional one-half of one percent to 1.8% but requires substantially more mammograms and yields more false-positives and over-diagnosed cases.

Conclusion: Decisions about screening strategy depend on preferences for benefits vs. potential harms and resource considerations.

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Introduction

Early randomized trials of mammography demonstrated significant reductions in breast cancer mortality associated with screening from ages 50 to 69 years.^{1,2} Unfortunately, there were only small numbers of women over age 70 who enrolled in the trials, so

results for this age group remain inconclusive. Recently the Age trial clearly demonstrated a small, but significant reduction in breast cancer mortality associated with screening women ages 40 to 49.^{3,4} All of these clinical trials varied in ages included, screening intervals, and use of invitations to screen vs. actual screening use, making it difficult to synthesize results to make public health recommendations.

In our prior work, we developed models of breast cancer incidence and mortality in the United States. These models are ideally suited for synthesizing data to estimate the effect of screening under a variety of policies^{5,6} since they can hold selected

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[‡] This work was done by 6 independent modeling teams from Dana-Farber Cancer Institute; Erasmus Medical Center; Georgetown University Medical Center, Lombardi Comprehensive Cancer Center (Dr. Mandelblatt, principal investigator); Harvard School of Public Health, Harvard Medical School, Harvard Pilgrim Health Care/University of Wisconsin (Dr. Stout, principal investigator); MD Anderson Comprehensive Cancer Center; (Dr. Berry, principal investigator) and Stanford University (Dr. Plevritis, principal investigator). Drs. Mandelblatt and Cronin were the writing and coordinating committee for the project; all other collaborators are listed in alphabetical order. Dr. Feuer was responsible for overall CISNET project direction.

conditions (e.g., screening intervals) constant, permitting direct comparison of strategies. Because all models make assumptions about unobservable events, we have collaborated to use of several models to provide a range of plausible results.⁶ Our prior results indicated that biennial screening of average-risk women captured the majority of mortality benefits of annual screening with approximately half as many mammography resources and false positive results.⁷ These data, together with a review of the most current trial results⁴ were used by the United States Preventive Services Task Force to inform their most recent guidelines for breast cancer screening. The Task Force recommended that screening occur every other year among average-risk women ages 50 to 74. The upper age limit was new from the prior guidelines. For women younger than 50, they suggested that women discuss individual risk and preferences for potential harms with their providers before considering routine screening.⁸

These recommendations generated considerable controversy^{9–14} and issuance of diverging guidelines from professional groups, generally to conduct more intensive screening than suggested by the Task Force.¹⁵ The majority of the controversy centered on the balance of benefits and harms of screening women ages 40 to 49, with little discussion about the upper age limit.

In this paper we present data from our established models comparing screening strategies,⁷ including new analyses of approaches that correspond to the different guidelines currently promoted in the US. These results are intended to make explicit the expected population outcomes for the strategies adopted.

Methods

The models were developed independently within the Cancer Intervention and Surveillance Modeling Network (CISNET) of the National Cancer Institute (NCI).^{6,7,16} Since there was no personal health information included, only population based de-identified data, the research was considered in the exempt category by the institutional review boards. The models have been described elsewhere.^{7,17–23} Briefly, the models share common features and inputs but differ in some ways. Several model include ductal carcinoma in-situ (DCIS) (model E [Erasmus], model G [Georgetown-Einstein], model M [MD Anderson Cancer Center], and model W [Wisconsin-Harvard] include DCIS. Models E and W specifically assume that some portions of DCIS are non-progressive and do not result in death; model W also assumes that some cases of small invasive cancer are non-progressive. Model S [Stanford] and model D [Dana-Farber] include only invasive cancer. Some groups model breast cancer in stages, but three (models E, S, and W) use tumor size and tumor growth. The models also differ by whether treatment affects the hazard for death from breast cancer (models G, S, and D), results in a cure for some fraction of cases (models E and W), or both (model M). Despite these differences, in previous collaborations⁷ the models came to similar qualitative estimates of the relative contributions of screening and treatment to observed decreases in deaths from breast cancer.

Model overview

The models begin with estimates of breast cancer incidence and mortality trends without screening and treatment and then overlay screening use and improvements in survival associated with treatment.⁷ We use a cohort of women born in 1960 and follow them beginning at age 25 years for their entire lives. Breast cancer is generally depicted as having a preclinical screening-detectable period (sojourn time) and a clinical detection point. On the basis of mammography sensitivity (or thresholds of detection), screening identifies disease in the preclinical screening-detection period and results in the identification of earlier-stage or smaller tumors than

might occur via clinical detection, resulting in reduction in breast cancer mortality. Age, estrogen receptor status, and tumor size- or stage-specific treatment have independent effects on mortality. Women can die of breast cancer or of other causes.

Model data parameters

We used a common set of age-specific variables for breast cancer incidence, mammography test characteristics, treatment algorithms and effects, and non-breast cancer competing causes of death.⁷ Each model also included additional model-specific inputs (or intermediate outputs) to represent preclinical detectable times, lead time, dwell time within stages of disease, and stage distribution in unscreened versus screened women on the basis of their specific model structure.^{7,17–23}

We use an age-period-cohort model to estimate what average breast cancer incidence rates would have been without screening.²⁴ This method considers the effect of age, temporal trends in risk by cohort, and time period. Because we do not have data on future incidence of breast cancer, we extrapolated forward assuming that future age-specific incidence increases as women age based on the last observed patterns. To isolate the effect of technical effectiveness of screening and to assess the effect of screening on mortality while holding treatment constant, models assume 100% adherence to screening and indicated treatment.

We used data on age-specific mammography sensitivity (and specificity) as observed in the Breast Cancer Surveillance Consortium (BCSC) program for initial and subsequent mammography performed at either annual or biennial intervals.²⁵

All women who have estrogen receptor-positive invasive tumors receive a hormonal treatment (tamoxifen if age at diagnosis is <50 years and anastrozole if ≥50 years) and non-hormonal treatment with an anthracycline-based regimen. Women with estrogen receptor-negative invasive tumors receive non-hormonal therapy only. Women with DCIS who have estrogen receptor-positive tumors receive hormonal therapy only.²⁶ Treatment effectiveness is based on synthesis of recent clinical trials and is modeled as a proportionate reduction in mortality risk or the proportion cured.^{27–29}

Benefits

We estimate the cumulative probability of unscreened women dying of breast cancer from age 40 years to death. Screening benefit is then calculated as the percentage of reduction in breast cancer mortality (vs. no screening). Benefits are cumulated over the lifetime of the cohort to capture reductions in breast cancer mortality (or life-years gained) occurring years after the start of screening, after considering non-breast cancer mortality.³⁰

Harms

Three different potential screening harms were examined: false-positive mammograms, unnecessary biopsies, and over-diagnosis. False-positive mammograms are the number of mammograms read as abnormal or needing further follow-up in women without cancer divided by the total number of positive screening mammograms based on the specificity reported in the BCSC.²⁵ Unnecessary biopsies are the proportion of women with false-positive screening results who receive a biopsy.³¹ Over-diagnosis is the proportion of cases in each strategy that would not have clinically surfaced in a woman's lifetime (because of lack of progressive potential or death from competing mortality among all cases arising from age 40 years onward).

Table 1Percent mortality reduction associated with different screening strategies ranked by number of screening exams by model^a

Screening strategy	# of screens/1000 women ^b	Percent breast cancer mortality reduction (vs. no screening) in model ^c					
		D	E	G	M	S	W
Efficient strategies (not dominated in 6 of 6 models)							
B 60–69	4263	11%	13%	11%	10%	9%	12%
B 55–69	6890	15%	18%	15%	14%	13%	19%
B 50–69	8947	16%	23%	17%	16%	15%	23%
B 50–74	11066	22%	27%	21%	21%	20%	28%
B 50–79	12366	25%	29%	24%	24%	25%	30%
B 50–84	13837	29%	31%	25%	27%	26%	33%
B 40–84	18708	31%	37%	28%	29%	27%	39%
A 40–84	36550	38%	49%	32%	29% ^d	35%	54%
Inefficient/dominated strategies (dominated in all 6 models) ^e							
A 60–69	8438	14%	18%	13%	12%	12%	17%
B 45–69	11694	18%	26%	20%	19%	17%	27%
A 55–69	13009	18%	25%	17%	15%	16%	26%
B 40–69	13831	18%	28%	20%	19%	16%	29%
A 50–69	17733	21%	31%	20%	18%	20%	33%
A 50–74	21330	27%	35%	24%	22%	26%	38%
A 45–69	22546	23%	35%	22%	20%	22%	39%
A 40–69	27428	24%	39%	23%	20%	22%	43%

A: Annual; B: Biennial.

^a Table adapted from results published by Mandelblatt et al., *Annals of Internal Medicine* 2009.⁷^b Average number of mammograms across models. Not all possible mammograms in the age interval are obtained in strategies that continue to the oldest age groups since many women die as the result of other causes before screening would occur.^c Model Group Abbreviations: D (Dana Farber Cancer Center), E (Erasmus Medical Center), G (Georgetown University), M (M.D. Anderson Cancer Center), S (Stanford University), W (University of Wisconsin/Harvard).^d Due to rounding, this strategy appears to be dominated, but the actual result is 29.4%.^e Strategies that are dominated (“inefficient”) within a specific model; a strategy is classified as dominated if there is another strategy that results in an equal or higher percent mortality decline with fewer average screening exams. Strategies that were dominated in some models but not others are not shown.

Analysis

We compared model results for 20 strategies among average-risk women. To rank the screening strategies, we first look at the results of each model independently. For a particular model, a strategy that requires more mammograms (our measure of resource use) but has a lower relative percentage of mortality reduction is considered inefficient or “dominated” by other strategies. We then evaluate strategies on the basis of results from all six models together.

After eliminating all dominated strategies, we represent the remaining strategies as points on a graph plotting the average number of mammograms versus the percentage of mortality reduction. We obtained the efficiency frontier for each graph by identifying the sequence of points that represent the largest incremental gain in percentage of mortality reduction per additional screening mammography. Screening strategies that fall on this frontier are the most efficient (that is, no alternative exists that provides more benefit for fewer mammograms performed).

Model validation and uncertainty

In our previous work, results of each model accurately projected independently estimated trends in the absence of intervention and closely approximated modern stage distributions and observed mortality trends.^{7,17–19,21–23} Using six models to project a range of plausible screening outcomes provides implicit cross-validation, with the range of results from the models as a measure of uncertainty.

Results

The cumulative probability of dying from breast cancer from age 40 to the end of a woman's life is a median of 3.0% across the models. Thus, if a particular screening strategy leads to a 10% breast cancer mortality reduction, then the probability of breast cancer death would be reduced from 3.0% to 2.7%, or 3 deaths averted per 1000 women screened.

The six models produce consistent results on the ranking of the strategies in terms of reduction in breast cancer mortality (Table 1). Eight approaches are “efficient” in all models (that is, not dominated, because they provide additional mortality reductions for added use of mammography); seven of these have a biennial interval, and all but two begin at age 50 years. In all models, biennial screening starting at age 50 years and continuing through ages 74 or 79 are of fairly similar efficiency. Strategies that include screening until age 84 years provide further improvements in outcome, albeit at a small added increment. (Fig. 1).

To examine the effect of screening interval, we calculated for each screening strategy and model the proportion of the annual benefit (in terms of mortality reduction) that could be achieved by biennial screening. Biennial screening maintains an average of 81% (range across strategies and models, 67% to 99%) of the benefits achieved by annual screening and yields only half as many false positives. The proportion of biopsies that occur because of these false-positive results that are retrospectively deemed unnecessary (that is, the woman did not have cancer) is about 7%; therefore, many more women will undergo unnecessary biopsies under annual screening than biennial screening (Table 2).

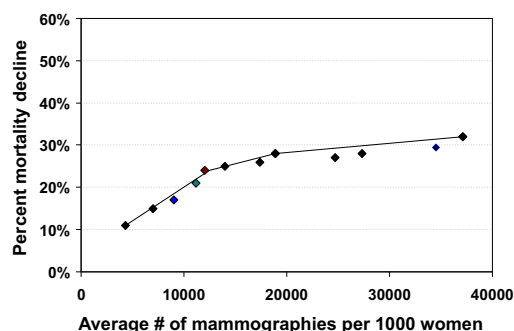


Fig. 1. Percentage of breast cancer mortality reduction versus number of mammographies per woman, by model and screening strategy. The panels show an efficiency frontier graph for an exemplar model (Model G). The graph plots the average number of mammograms performed per 1000 women against the percentage of mortality reduction for each screening strategy (vs. no screening). We plot efficient strategies (that is, those in which increases in use of mammography resources result in greater mortality reduction than the next least-intensive strategy). We also plot “borderline” strategies (approaches that are efficient in some models but not others). The line between strategies represents the “efficiency frontier.” Strategies on this line would be considered efficient because they achieve the greatest gain per use of mammography resources compared with the point (or strategy) immediately below it. Points that fall below the line are not considered as efficient as those on the line. When the slope in the efficiency frontier plot levels off, the additional reductions in mortality per unit increase in use of mammography are small relative to the previous strategies and could indicate a point at which additional investment (use of screening) might be considered as having a low return (benefit). Adapted from Mandelblatt et al., 2009.⁷

If screening begins at age 40 years (vs. 50 years) and continues to age 79 years (we did not model an ending age of 74 when starting at age 40), all models project additional, albeit small, reductions in breast cancer mortality for both annual and biennial screening (Table 3). However, more false-positive results occur in strategies that include screening from ages 40 to 49 years (Table 2). Continuing screening to age 84 years (vs. 50–74 years) results in a median increase in percentage of mortality reduction of 6.5% (range 4%–8%) and 5.5% (range 4% to 7%) under annual and biennial intervals, respectively. However, these benefits are accompanied by an accelerating rate of increase in the risk of over-diagnosis in the older age groups, mostly because of growing rates of competing causes of mortality (not shown).

To compare results for exemplar strategies that have been recommended in the US, we compared the incremental mortality reductions of the most intensive screening regimen (40 to 84 annually) to screening every other year from ages 50 to 74. Screening biennially from ages 50 to 74 lowers mortality by 23%. This means that the 3% chance of death without screening is reduced to 2.3% (3.0% minus 23.2% of 3%). Extending screening beyond biennial examinations from 50 to 74 to annual screening from ages 40 to 84 results in an additional 15.5% mortality reduction (Table 4), reducing the 2.3% probability of death further to 1.8%. In other words, most women do not die of breast cancer and the lifetime risk of death is only reduced by one half of one percent with more intensive screening.

Sensitivity analysis

Results for ranking of strategies and all conclusions were similar to the base analyses under different assumptions about test sensitivity (e.g., 10% increase in sensitivity).

Discussion

This collaborative modeling project demonstrates that the choice of optimal breast cancer screening strategies is complex. All six modeling groups concluded that the most efficient screening strategies are those that include a biennial screening interval. Initiation of screening at age 40 provides small added benefits but

is accompanied by a large increase in the number of screening examinations and a high false positive rate. Extending screening beyond age 74 yields moderate mortality reductions and lower false positive rates but at the expense of some women being over-diagnosed. The absolute difference in lifetime probability of death that would be expected under the Task Force recommendation for biennial screening from ages 50 to 74 compared to annual screening from age 40 to 84 is very small (0.5%).

Screening intervals are somewhat arbitrary. Screening every one month or every six months would detect the greatest number of cancers, but would be infeasible in terms of time, mammography resources and the weight of false positive exams. The finding in this study that biennial screening is more efficient than annual screening is consistent with previous screening trials, most of which used 2-year intervals.^{1,2} The efficiency of biennial screening is largely due to the biology of breast cancer and the specificity of mammography. Slow growing tumors are much more common than rapidly growing tumors, and the ratio of slow to fast growing tumors increases with age,³² so that little survival benefit is lost between screening every year versus every other year. For the small sub-set of younger women with aggressive, faster-growing tumors, even annual screening is not likely to confer a survival advantage. Since the specificity of mammography is less than 100%, the less often screening occurs, the lower the number of false positive results and unnecessary biopsies.

In all models, some reductions in breast cancer mortality, albeit small, were seen with strategies initiating screening at age 40 versus age 50. This is consistent with the recent Age trial in the United Kingdom.³⁴ The small magnitude of benefit is attributable to the low incidence of disease from age 40 to 49 and the low sensitivity of mammography in this age group. The same factors that lead to the small benefits in the younger age group also contribute to the harms of screening – high rates of false positive screens and unnecessary biopsies. In addition, since the proportion of DCIS is highest in younger women, screen detection of DCIS that may not be clinically significant could be considered a further harm. Thus, decisions to screen before age 50 largely depend on women's willingness to tolerate these harms for a small chance that cancer is present and that screening will reduce the probability of death from that cancer.

At the other end of the age spectrum, all six models found that screening beyond age 74 remains on the efficiency frontier. This result is consistent with previously reported results of screening benefit from observational and modeled data.^{33–36} As with the situation for younger women, any benefits of screening older women must be balanced against possible harms. For instance, the probability of over-diagnosis accelerates among women over age 74. Model estimates for the oldest age groups also have some uncertainty built in because of the limited primary data on natural history of breast cancer and the absence of screening trial data after age 74 years.

It is logical to assume that more screening will save substantially more lives. However, when comparing the strategy recommended by the US Preventive Services Task Force⁸ to the most intensive regimen we evaluated (annual screening from ages 40 to 84), there was only a one-half of one percent additional reduction in the lifetime probability of death from breast cancer. This somewhat counter-intuitive result is based on several factors, including the fact that most women never develop breast cancer and when cancer is diagnosed, treatment is very effective in avoiding death for most women. Additional variables, such as slow tumor growth rates and low incidence rates in young women, also mean that a less intensive screening schedule still maintains the majority of the benefits of more intensive strategies and use far fewer mammography screening and diagnostic resources.

Table 2Potential benefits and harms of mammography screening under different screening strategies varying by interval and age of initiation and cessation^a

Strategy	Average screens/1000	Potential benefits (vs. no screening)			Potential harms ^b	
		Mortality reduction (%) ^c	Cancer deaths averted/1000	Life years gained per 1000 ^c	# False positives/1000	# Unnecessary biopsies/1000
Biennial						
<i>Varying starting ages</i>						
B 40–69	13865	16% ³	6.1	120	1250	88
B 45–69	11771	17%	6.2	116	1050	74
B 50–69	8944	15%	5.4	99	780	55
B 55–69	6941	13%	4.9	80	590	41
B 60–69	4246	9%	3.4	52	340	24
<i>Varying stopping ages</i>						
B 50–69	8944	15%	5.4	99	780	55
B 50–74	11109	20%	7.5	121	940	66
B 50–79	12347	25%	9.4	130	1020	71
B 50–84	13836	26%	9.6	138	1130	79
Annual						
<i>Varying starting ages</i>						
A 40–69	27583	22%	8.3	164	2250	158
A 45–69	22623	22%	8.0	152	1800	126
A 50–69	17759	20%	7.3	132	1350	95
A 55–69	13003	16%	6.1	102	950	67
A 60–69	8406	12%	4.6	69	600	42
<i>Varying stopping ages</i>						
A 50–69	17759	20%	7.3	132	1350	95
A 50–74	21357	26%	9.5	156	1570	110
A 50–79	24439	30%	11.1	170	1740	122
A 50–84	26913	33%	12.2	178	1880	132

A: Annual; B: Biennial.

^a Table adapted from results published by Mandelblatt et al., *Annals of Internal Medicine* 2009.⁷ Results are from one model (Model S) for illustrative purposes; similar results were obtained from the other models.^b Over-diagnosis is another significant harm associated with screening. However, given the uncertainty in the knowledge base about DCIS and small invasive tumors, we felt that the absolute estimates are not reliable. In general, over-diagnosis increases with age across all age groups, but rises more sharply for women who are screened in their 70s and 80s.^c Strategies listed in grey are dominated by other strategies; the strategy that dominates may not be in the table.

The collaboration of six groups with different modeling philosophies and approaches to estimate the same end-points by using a common set of data provides an excellent opportunity to cross-replicate results and depicts uncertainty related to modeling assumptions and structure by providing a range of results. The resulting conclusions about the ranking of screening strategies were very robust and should provide greater credibility than inferences based on one model alone.

Despite our consistent results, our study had some limitations. First, our models project mortality reductions similar to those observed in clinical trials, but the range of results includes higher mortality reductions than seen in the trials because we model lifetime screening (vs. for the period of the trial and for invitations to screening) and assume adherence to all screening and treatment. The trials followed women for limited numbers of years and have some non-adherence. Second, we do not consider morbidity associated with surgery for screening-detected disease³⁷ or decrements in quality of life associated with false-positive results, living with earlier knowledge of a cancer

diagnosis, or over-diagnosis.³⁸ Third, in estimating lifetime results, we projected breast cancer trends from background incidence rates of a 1960 birth cohort extrapolated forward in time. However, future background incidence (and mortality) may change as the result of different forces, and/or results may vary for groups with higher than average risk.³⁹ We assumed 100% adherence to screening and treatment to evaluate program efficacy. Benefits will always fall short of the projected results because adherence is not perfect. If actual adherence varies systematically by age or other factors, the ranking of strategies could change. Finally, we did not include costs in our analysis, although the average number of mammograms per woman (and false-positive results) provides some proxy of resource consumption. Even with these acknowledged limitations, the models demonstrate meaningful, qualitatively similar outcomes.

Choices about optimal ages of initiation and cessation will ultimately depend on program goals, resources, weight attached to the balance of harms and benefits, and considerations of efficiency and equity.

Table 3

Incremental change in percent breast cancer mortality reduction by age of screening initiation and cessation

Model	Difference in % breast cancer mortality reduction			
	Start age 40 (vs. 50) ^a		Stop age 84 (vs. 74) ^b	
	Annual	Biennial	Annual	Biennial
D	2	2	8	7
E	7	6	6	4
G	3	2	4	4
M	1	1	6	6
S	3	1	7	6
W	3	6	7	5
Median	3	2	6.5	5.5

Model group abbreviations: D (Dana Farber Cancer Center), E (Erasmus Medical Center), G (Georgetown University), M (M.D. Anderson Cancer Center), S (Stanford University), W (University of Wisconsin/Harvard)

^a Incremental difference between screening from 40–79 vs. 50–79. Since the models did not evaluate 40–74, the gain from starting 10 years earlier is presented using the end age of 79.

^b Incremental difference between screening from 50–84 vs. 50–74.

Table 4

Incremental change in percent breast cancer mortality reduction from adopting different screening guidelines

Model	Difference in % breast cancer mortality reduction, screening annually 40–84 vs. biennially 50–74 years
Dana Farber Cancer Center	16%
Erasmus Medical Center	22%
Georgetown University	11%
M.D. Anderson Cancer Center	8%
Stanford University	15%
University of Wisconsin/Harvard	26%
Median	15.5%

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Conflict of interest statement

The authors have no conflict of interest to declare.

References

- Nystrom L, Andersson I, Bjurstram N, Frisell J, Nordenskjold B, Rutqvist LE. Long-term effects of mammography screening: updated overview of the Swedish randomised trials. *Lancet* 2002;359:909–19.
- Tabar L, Vitak B, Chen HH, Duffy SW, Yen MF, Chiang CF, et al. The Swedish Two-County Trial twenty years later. Updated mortality results and new insights from long-term follow-up. *Radiol Clin North Am* 2000;38:625–51.
- Moss SM, Cuckle H, Evans A, Johns L, Waller M, Bobrow L. Effect of mammographic screening from age 40 years on breast cancer mortality at 10 years' follow-up: a randomised controlled trial. *Lancet* 2006;368:2053–60.
- Nelson HD, Tyne K, Naik A, Bougatsos C, Chan B, Nygren P, et al. Screening for breast cancer: systematic evidence review update for the US preventive services task force. 2009.
- Mandelblatt JS, Fryback DG, Weinstein MC, Russell LB, Gold MR. Assessing the effectiveness of health interventions for cost-effectiveness analysis. Panel on Cost-Effectiveness in Health and Medicine. *J Gen Intern Med* 1997;12:551–58.
- Berry DA, Cronin KA, Plevritis SK, Fryback DG, Clarke L, et al. Effect of screening and adjuvant therapy on mortality from breast cancer. *N Engl J Med* 2005;353:1784–92.
- Mandelblatt JS, Cronin KA, Bailey S, Berry DA, de Koning HJ, Draisma G, et al. Effects of mammography screening under different screening schedules: model estimates of potential benefits and harms. *Ann Intern Med* 2009;151:738–47.
- United States Preventive Services Task Force. Screening for Breast Cancer. Agency for Health Care Research and Quality. Rockville, MD. <http://www.uspreventiveservicestaskforce.org/uspstf/uspssbrca.htm>. Accessed March 11, 2011.
- When evidence collides with anecdote, politics, and emotion: breast cancer screening. *Ann Intern Med* 2010;152:531–32.
- Woloshin S, Schwartz LM. The benefits and harms of mammography screening: understanding the trade-offs. *JAMA* 2010;303:164–65.
- Murphy AM. Mammography screening for breast cancer: a view from 2 worlds. *JAMA* 2010;303:166–67.
- Woolf SH. The 2009 breast cancer screening recommendations of the US Preventive Services Task Force. *JAMA* 2010;303:162–63.
- Kopans D. The 2009 U.S. Preventive Services Task Force Guidelines ignore important scientific evidence and should be revised or withdrawn. *Radiology* 2010;256 (1): 15–20.
- Hendrick RE, Helvie MA. United States Preventive Services Task Force screening mammography recommendations: science ignored. *AJR Am J Roentgenol* 2011;196:W112–W116.
- American Cancer Society Guidelines for the Early Detection of Cancer. <http://www.cancer.org/Healthy/FindCancerEarly/CancerScreeningGuidelines/american-cancer-society-guidelines-for-the-early-detection-of-cancer>. Accessed March 11, 2011.
- Cancer Intervention and Surveillance Modeling Network. Accessed at <http://cisnet.cancer.gov/breast/profiles.html> on 15 September 2008.
- Fryback DG, Stout NK, Rosenberg MA, Trentham-Dietz A, Kuruchittham V, Remington PL. The Wisconsin Breast Cancer Epidemiology Simulation Model. *J Natl Cancer Inst Monogr* 2006;37–47.
- Mandelblatt J, Schechter CB, Lawrence W, Yi B, Cullen J. The SPECTRUM population model of the impact of screening and treatment on U.S. breast cancer trends from 1975 to 2000: principles and practice of the model methods. *J Natl Cancer Inst Monogr* 2006;47–55.
- Berry DA, Inoue L, Shen Y, Venier J, Cohen D, Bondy M, et al. Modeling the impact of treatment and screening on U.S. breast cancer mortality: a Bayesian approach. *J Natl Cancer Inst Monogr* 2006;30–6.
- Clarke LD, Plevritis SK, Boer R, Cronin KA, Feuer EJ. A comparative review of CISNET breast models used to analyze U.S. breast cancer incidence and mortality trends. *J Natl Cancer Inst Monogr* 2006;96–105.
- Plevritis SK, Sigal BM, Salzman P, Rosenberg J, Glynn P. A stochastic simulation model of U.S. breast cancer mortality trends from 1975 to 2000. *J Natl Cancer Inst Monogr* 2006;86–95.
- Lee S, Zelen M. A stochastic model for predicting the mortality of breast cancer. *J Natl Cancer Inst Monogr* 2006;79–86.
- Tan SY, van Oortmarssen GJ, de Koning HJ, Boer R, Habbema JD. The MISCAN-Fadia continuous tumor growth model for breast cancer. *J Natl Cancer Inst Monogr* 2006;56–65.
- Holford TR, Cronin KA, Mariotto AB, Feuer EJ. Changing patterns in breast cancer incidence trends. *J Natl Cancer Inst Monogr* 2006;19–25.
- Performance Measures for Screening Mammography Examinations from 2002 to 2006 by Age & Time (Months) Since Previous Mammography. http://breastscreening.cancer.gov/data/performance/screening/2009/perf_age_time.html. Accessed March 11, 2011.
- National Comprehensive Cancer Network. NCCN Clinical Practice guidelines in oncology v.2.2008. at http://www.nccn.org/professionals/physician_gls/f_guidelines.asp Accessed January 2011.

27. Clarke M, Coates AS, Darby SC, Davies C, Gelber RD, Godwin J, et al. Adjuvant chemotherapy in oestrogen-receptor-poor breast cancer: patient-level meta-analysis of randomised trials. *Lancet* 2008;371:29–40.
28. Early Breast Cancer Trialists' Collaborative Group. Effects of chemotherapy and hormonal therapy for early breast cancer on recurrence and 15-year survival: an overview of the randomised trials. *Lancet* 2005;365:1687–717.
29. Cronin KA, Feuer EJ, Clarke LD, Plevritis SK. Impact of adjuvant therapy and mammography on U.S. mortality from 1975 to 2000: comparison of mortality results from the CISNET breast cancer base case analysis. *J Natl Cancer Inst Monogr* 2006;112–21.
30. Rosenberg MA. Competing risks to breast cancer mortality. *J Natl Cancer Inst Monogr* 2006;15–9.
31. Rosenberg RD, Yankaskas BC, Abraham LA, Sickles EA, Lehman CD, Geller BM, et al. Performance benchmarks for screening mammography. *Radiology* 2006;241:55–66.
32. Fracheboud J, Groenewoud JH, Boer R, Draisma G, de Bruijn AE, Verbeek AL, et al. Seventy-five years is an appropriate upper age limit for population-based mammography screening. *Int J Cancer* 2006;118:2020–5.
33. Mandelblatt J, Schechter C, Yabroff KR, et al. Towards optimal screening strategies for older women: Costs, benefits and harms of breast cancer screening by age, biology and health status. *J Gen Intern Med* 2005;20(6):487–96.
34. McCarthy EP, Burns RB, Freund KM, Ash AS, Shwartz M, Marwill SL, et al. Mammography use, breast cancer stage at diagnosis, and survival among older women. *J Am Geriatr Soc* 2000;48:1226–33.
35. Lash TL, Fox MP, Buist DS, Wei F, Field TS, Frost FJ, et al. Mammography surveillance and mortality in older breast cancer survivors. *J Clin Oncol* 2007;25:3001–6.
36. Boer R, de Koning HJ, van Oortmarssen GJ, van der Maas PJ. In search of the best upper age limit for breast cancer screening. *Eur J Cancer* 1995;31A:2040–3.
37. El Tamer MB, Ward BM, Schiffner T, Neumayer L, Khuri S, Henderson W. Morbidity and mortality following breast cancer surgery in women: national benchmarks for standards of care. *Ann Surg* 2007;245:665–71.
38. Bonomi AE, Boudreau DM, Fishman PA, Ludman E, Mohelnitzky A, Cannon EA, et al. Quality of life valuations of mammography screening. *Qual Life Res* 2008;17:801–14.
39. Ravdin PM, Cronin KA, Howlader N, et al. The decrease in breast-cancer incidence in 2003 in the United States. *N Engl J Med* 2007;356:1670–4.