# **Supplementary Online Content**

Caswell-Jin JL, Sun LP, Munoz D, et al. Analysis of breast cancer mortality in the US— 1975 to 2019. *JAMA*. Published January 16, 2024. doi:10.1001/jama.2023.25881

**eMethods.** Model Inputs and Assumptions

**eReferences.** 

**eTable 1.** Breast Cancer Model Input Parameters

**eTable 2.** NCCN Outcomes Database Cohort Characteristics

**eTable 3.** Breast Cancer Treatments and Their Efficacy, 1975-2019

**eTable 4.** Model M Standardized Hazard Reductions for Overall Survival After Metastasis by ER/ERBB2 Subtype and Year Based on First-Line Metastatic Therapy

**eTable 5.** Estimated Absolute Breast Cancer Age-Adjusted Mortality Rate and Its Reduction Relative to No Intervention by Subtype Across Eight Scenarios in 2019, by CISNET Model

**eTable 6.** Estimated Median Breast Cancer–Specific Survival After Distant Recurrence Over Time by Estrogen Receptor/ERBB2 Subtype, by CISNET Model

**eTable 7.** Estimated Five and Ten-Year Distant Recurrence-Free Survival Over Time by Estrogen Receptor/ERBB2 Subtype

**eFigure 1.** Breast Cancer–Specific Survival After Distant Recurrence by Subtype in the NCCN Outcomes Database as Compared to Model S

**eFigure 2.** Distant Recurrence-Free Survival by Subtype in the NCCN Outcomes Database as Compared to Model S

**eFigure 3.** Survival After Metastasis as Observed in Clinical Trials Versus Predicted From Model S

**eFigure 4.** Metastatic Therapy Usage by Year of Diagnosis of Metastatic Recurrence

**eFigure 5.** Summary of 127 Approaches to Calculate Contributions of Interventions to Breast Cancer Mortality Reduction

**eFigure 6.** Comparison of Symmetrical to Asymmetrical Approaches to Calculate Contributions of Interventions to Breast Cancer Mortality Reduction

**eFigure 7.** Estimated Age-Adjusted Breast Cancer Mortality Over Time Across Eight Scenarios Compared to Observed Mortality, by CISNET Model

**eFigure 8.** Estimated Breast Incidence, Average Across All Models and by Model, as Compared to SEER Breast Cancer Incidence

**eFigure 9.** Associations With Overall Breast Cancer Mortality Reduction of Screening, Stage I-III Treatments, and Metastatic Treatments in 2019, by CISNET Model and Within Model M

**eFigure 10.** Associations With Overall Breast Cancer Mortality Reduction of Screening, Stage I-III Treatments, and Metastatic Treatments Over Time, by CISNET Model

This supplementary material has been provided by the authors to give readers additional information about their work.

### eMethods**. Model inputs and assumptions**

### *Modeling the occurrence of metastatic recurrence and subsequent baseline survival*

In the original CISNET breast cancer models, we used estimates of a stage- and ER/ERBB2 category-specific baseline breast cancer survival curve, denoted by *S0*, which captures breast cancer survival for patients who are symptomatically detected, in the absence of screening, and who are treated with surgery and radiation, per NCCN guidelines. To maintain consistency with this framework and enable modeling distant recurrence, we deconstruct *S0* into two components: (i) baseline survival from initial diagnosis of Stage I-III breast cancer to the detection of distant recurrence, denoted by *S10*, and also referred to as baseline distant recurrence-free survival; and (ii) baseline survival from detection of distant recurrence to breast cancer death, denoted by *S20*, assumed to be the same as the baseline progression-free survival curve during metastasis.

To estimate *S10 and S20*, we leveraged the NCCN breast cancer recurrence data, together with the Model S, and made the following assumptions:

- Treatment benefits follow a proportional hazards model (as assumed in prior work).
- Metastatic patients receive all the available treatments sequentially, per standard of care.
- Baseline survival from detection of distant recurrence to breast cancer death and baseline progression-free survival curve during metastasis are equal.
- *S1* and *S2* are dependent. Assuming independence between S1 and S2 generated *S2* >> S1 for many patients, contradicting real world observations.
- *S2* does not depend on mode of detection. When we compared *S2* curves from the NCCN dataset between clinically diagnosed and screened patients, we found no significant differences.

Step 1 – Initial estimate of *S20*: We first simulate a virtual cancer registry using the original Model S in the absence of screening but in the presence of Stage I-III treatment. Sampling from Kaplan-Meier estimates of *S2* from the NCCN dataset, we compute dates of distant recurrence in the virtual registry. We then assign metastatic treatments depending on each patient's date of distant recurrence using our new treatment dissemination inputs (**eTable 3**). Finally, to generate the baseline curve *S20*, we remove the estimated metastatic treatments benefits (**eTable 3**) from the observed *S2* curves, assuming proportional hazards.

Step 2 – Initial estimate of  $SI_0$ : We simulate a virtual registry using the original Model S in the absence of screening and Stage I-III treatment, then subtract survival times from distant recurrence for each patient using initial estimates of *S20* (Step 1). From the simulated events we compute Kaplan-Meier estimates of *S10*.

Step 3 – Calibration of *S20*: We incorporate the initial estimates of *S10* (Step 2) and *S20* (Step 1) into the revised Model S, which we use to generate  $S2<sub>0</sub>$  through microsimulation and calibrate  $S2<sub>0</sub>$  so that simulated results match previously reported *S2* medians in the 1970s where the treatment options were limited and NCCN-observed *S2* curves for the calendar years 1997-2012. These are the final estimates of *S20*.

Step  $4$  – Calibration of  $SI_{\theta}$ : We repeat Step 2, except using the calibrated estimates of  $S2_{\theta}$  from Step 3.

We validate the final estimates of  $SI_0$  and  $S_0$  by comparing the simulated outputs of the revised Model S with various observed real-world data, namely the incidence and mortality rates from SEER overall and by ER subtype, *S2* from the NCCN Outcomes Database (**eFigure 1),** *S1* from the NCCN Outcomes Database (**eFigure 2**), and *S2* in the clinical trials of patients with metastatic breast cancer (**eFigure 3**). For comparison with clinical trials of firstline treatments, we assumed that overall survival from trial enrollment would approximate breast cancer–specific survival after metastasis in the model.

Three models (Models D, M, and S) used these baseline curves  $SI_0$  and  $S_0$  as generated through the above inference process. In contrast, Model W uses a nonparametric mixture cure modeling approach to represent post-diagnosis events. Upon diagnosis and initial treatment, Model W assumes a proportion of patients are cured of breast cancer and their tumor natural history and progression are interrupted and these women are destined to die of non-breast cancer causes. For the remaining who are uncured, their tumors continue to progress according to the underlying tumor growth and progression trajectory. In Model W, the tumor natural history model includes individualized

growth model that determine sizes and a progression model that includes spread to lymph nodes and metastases. When a tumor reaches the metastatic state, it is assigned a subtype-specific remaining distant survival time based on the CISNET common-input. Breast cancer death occurs through progression to the metastatic state. In Model W, overall survival times are derived functions of the previously calibrated growth model parameters, new re-calibrated cure model parameters and CISNET common inputs on distant survival, treatment use and effectiveness. In order to incorporate diagnosis and treatment of distant recurrence into Model W, they first re-estimated the proportion of the patients who are cured of breast cancer and the proportion of the patients who are not cured in the absence of neoadjuvant, adjuvant, and metastatic therapy treatment. They then adjusted the proportion of the cured patients depending on the use and effectiveness of treatments. This model was calibrated to match observed mortality in the US from 1975 forward.

### *Modeling the treatment of Stage I-III breast cancer*

Modeling Stage I-III treatment was similar to our previous publications <sup>1</sup>, except that benefits were applied to the curve from diagnosis to recurrence instead of from diagnosis to death. All models used the same inputs for Stage I-III treatment efficacy and dissemination.

We used the dissemination patterns for Stage I-III treatments as in our prior publication, using the Surveillance Epidemiology and End Results (SEER) patterns-of-care special studies for 1975-1996 and the National Comprehensive Cancer Network (NCCN) Outcomes Data Base from 1997 onwards <sup>1</sup>.

Four treatments entered clinical practice after 2012, later than the scope of our prior publication: ovarian suppression for premenopausal women with ER+ disease, pertuzumab for ERBB2+ disease, capecitabine for triple-negative disease, and neratinib for ERBB2+ disease. In the absence of data available for the usage patterns of these recent drugs, their dissemination was estimated using a combination of expert opinion (authors JLC and AWK) and available data on chemotherapy usage patterns 2-4.

The probability of receiving ovarian suppression was set as 64% of simulated patients with regional ER+/ERBB2 disease; 21% with local ER+/ERBB2- disease; and 64% with ER+/ERBB2+ disease. These numbers were extrapolated from chemotherapy usage patterns, given that, based on results of subgroup analyses from the SOFT/TEXT trials<sup>5</sup>, ovarian suppression is typically given to patients with disease at sufficiently high risk to warrant the receipt of chemotherapy. In an analysis of the Georgia and Los Angeles SEER registries in 2015, 64% of patients with regional ER+/ERBB2- disease and 21% of patients with local ER+/ERBB2- disease received chemotherapy 2 . The estimate for ER+/ERBB2+ disease assumed similar chemotherapy usage patterns as for regional ER+/ERBB2- disease.

The probability of receiving pertuzumab was set as 90% of patients with regional ERBB2+ disease, based on expert opinion.

The probability of receiving capecitabine was set as 18% of patients with ER-/ERBB2- disease. This number was extrapolated from data about chemotherapy usage patterns for ER-/ERBB2- disease, assuming that patients with ER- /ERBB2- disease treated with neoadjuvant chemotherapy with residual disease would receive capecitabine. A study used data from the National Cancer Data Base (NCDB) to report that in 2010-2011, 81% of ER-/ERBB2- patients received chemotherapy<sup>3</sup>. We estimated that at time of introduction of adjuvant capecitabine, 50% of ER-/ERBB2patients would receive neoadjuvant therapy. We assumed a pathogenic complete response (pCR) to neoadjuvant therapy rate of 44%, in between the pCR rate of dose-dense doxorubicin/cyclophosphamide and paclitaxel (AC-T, 41%) and the pCR rate of AC-T with carboplatin (54%) in the CALGB 40603 trial <sup>6</sup>, based on data from the Georgia and California statewide SEER registries 4 .

The probability of receiving neratinib was set as 10% of patients with regional ER+/ERBB2+ disease, based on expert opinion.

For each Stage I-III treatment, we identified the hazard ratio for recurrence-free survival (if available) or diseasefree survival (if recurrence-free survival was not reported) from the most recent report from the randomized phase 3 clinical trial that led to its approval (**eTable 3**). Hazard ratios from the intention-to-treat populations were used with the following exceptions, where the drug appeared to be more efficacious in a subpopulation and then was

recommended to be used generally in that subpopulation: neratinib for ER+/HER2+ tumors, pertuzumab for nodepositive tumors, and capecitabine for ER-/HER2- tumors.

### *Modeling the treatment of metastatic disease*

Three models (Models D, S, and W) assumed that patients with metastatic disease were treated with sequential specific treatment regimens (**Figure 1B**), with specific benefits derived as below.

We compiled a list of available drugs from the current National Comprehensive Cancer Network (NCCN) guidelines for the management of metastatic breast cancer  $^7$  and identified their years of approval from publicly available reports of the U.S. Food and Drug Administration <sup>8</sup>. We identified the hazard ratios for overall survival with these drugs from the most recent reports of the randomized phase 3 clinical trials that led to their approval (**eTable 3**); we did not include drugs that had demonstrated a progression-free survival, but not an overall survival, benefit. As no data were available comparing the earliest available chemotherapies (e.g. anthracyclines) or endocrine therapies (e.g. tamoxifen) to no treatment in the metastatic setting, we estimated these hazard ratios based on the range of benefits observed for chemotherapy and endocrine therapy to no treatment in the Stage I-III setting as well as the range of benefits observed for newer chemotherapies and endocrine therapies compared to older in the metastatic setting (**eTable 3**)

Simulated patients with metastatic disease received lines of therapy according to standard of care at the time (**Figure 1B**). Baseline (in the absence of treatment) overall survival from diagnosis of metastatic recurrence was assumed to be equal to baseline progression-free survival after metastasis; that is, we assume that in the absence of available treatment, death typically occurs shortly after progression. Hazard ratios for overall survival for metastatic drugs were assumed to be comparable to their hazard ratios for progression-free survival. When a line of therapy was given, its benefits were applied to the simulated patient's baseline survival curve to the next progression, or death if all lines of therapy were exhausted. "Standard of care" was determined from NCCN guidelines in 2019 and extrapolated to previous years using expert opinion (authors JLC and AWK), with reference to control arms of the pivotal clinical trials of newly approved drugs. Because models applied benefits of only those regimens with demonstrated overall survival benefit, the complexity of treatment selection for later-line therapy was narrowed in the models. For example, no simulated patients with metastatic disease received everolimus, alpelisib, lapatinib, or single-agent cyclophosphamide, as these agents have not demonstrated overall survival benefit. Moreover, in a simplified approximation of real-world practice, all simulated patients received lines of therapy in the same prespecified order.

In those regimens that include multiple agents (for example, taxane/trastuzumab/pertuzumab for ERBB2+ disease), the benefit received was the product of the benefit for each drug individually, based on reports of clinical trials. Similarly, in those regimens in which at least one drug replaced a prior drug that also had an overall survival benefit, the benefit received was the product of the benefit of the prior drug and the benefit of the new drug: for example, a simulated patient receiving trastuzumab emtansine for metastatic ERBB2+ disease received the overall survival benefit of capecitabine (used in the control arm of the EMILIA trial) multiplied by the overall survival benefit of trastuzumab emtansine (as reported by the EMILIA trial)<sup>9</sup>. In this scenario, the drug whose overall survival benefit was already applied could not be used in a later line of therapy (e.g., after the introduction of trastuzumab emtansine, the benefit of capecitabine was given with trastuzumab emtansine, and not again in a later line of therapy). In the years after a treatment had moved from the metastatic to the Stage I-III setting, we halved its efficacy in the metastatic setting if the patient had already received it in the Stage I-III setting.

The probability that a simulated patient would receive a given number of lines of therapy was estimated for 2010- 2019 from IBM MarketScan US insurance claims for 2007-2014 10. In these data, of 6,180 women with metastatic breast cancer, 100% received at least one line of therapy, 72% at least two lines of therapy, 44% at least three lines of therapy, and 23% at least four lines of therapy. Adjusting for the fact that this study used receipt of at least one line of therapy for metastatic disease for cohort definition, we used the following parameters in the models: 90% probability of receiving at least one line of therapy, 68% at least two lines of therapy, 51% at least three lines of therapy (if available), and 25% four lines of therapy (if available). We assumed that before 1990, half of eligible simulated patients received each available line of therapy, and between 1990 and 2010, dissemination was interpolated to increase linearly to 2010 levels. These assumptions produced drug usage patterns over time as illustrated in **eFigure 4.** 

Model M used different assumptions about the treatment of metastatic disease from those described above. Instead of simulating the receipt of specific treatment regimens, Model M applied subtype-specific benefits to the curve from metastasis to death that captured the effects of metastatic treatments in a given year of diagnosis of metastatic disease. To do so, they introduced four parameters to represent the overall survival benefits of metastatic treatments <sup>1,11</sup>. They denote these four parameters as  $\alpha_{-}$ ,  $\alpha_{+}$ ,  $\alpha_{+}$ ,  $\alpha_{++}$ . They represent the hazard ratios of metastatic treatments for overall survival following distant metastasis in year 2020 for the primary disease subtypes ER- /ERBB2-, ER-/ERBB2+, ER+/ERBB2- and ER+/ERBB2+, respectively. As with all unknown parameters in the Bayesian approach, they have probability distributions. These distributions are updated based on the evidence, which in this case is the fit of the Model M results to observed (from SEER) breast cancer mortality. The baseline distributions (years 1975 to 1990) for the breast-cancer survival following distant metastasis are assumed to be exponentially distributed with a median of 1.35 and 1.70 years for the ER- and ER+ subtypes, respectively. These baseline distributions include any benefits of real-world pre-1990 chemotherapy and, for ER+ subtypes, tamoxifen.

Model M derives the hazard ratios of metastatic treatments for ER/ERBB2 subtypes in years 1975 to 2019 using  $\alpha_{-}, \alpha_{-}, \alpha_{+}, \alpha_{+}$ , and  $\alpha_{++}$ , based on the inputs on metastatic treatment dissemination and efficacy as used by the other models. The procedure involves two steps. Step 1 is to construct an approximate "standardized" hazard reduction table by subtype and year for metastatic treatments, based on the calculated raw hazard reductions. Step 2 is to discount the raw hazard reductions proportionally using the standardized hazard reduction table and the four  $\alpha$ parameters. The details are given below.

Step 1: For each ER/ERBB2 subtype and in each year from 1975-2019, the raw hazard reduction is calculated as the product of the dissemination probability of the therapy and the hazard reduction (i.e., 1 - hazard ratio) due to the therapy for overall survival. Only first-line therapy benefits were included. For combination therapies, the hazard ratio is the product of the hazard ratios of the component therapies. We then standardize the hazard reduction table by dividing each raw hazard reduction in each year by the raw hazard reduction in year 2020 for each ER/ERBB2 subtype. The hazard reductions in 2020 equal 1.0. The full set of calculated standardized hazard reductions is presented in eTable 4.

Step 2: The hazard ratio in each year from 1975-2019 for each ER/ERBB2 subtype is then calculated as  $1.0 - (1.0 \alpha_{ij}$ <sup>\*</sup>r where *i* and *j* equal + or – and *r* is the corresponding standardized hazard reduction in eTable 4. For example, the hazard ratio for a patient of ER-/ERBB2- subtype in 2010 is  $1.0 - (1.0 - \alpha_{-})*0.80684$ . Similarly, the hazard ratio for a patient of ER+/ERBB2- subtype in 2010 is  $1.0 - (1.0 - \alpha_{++})^*0.7099$ .

Using the same approximate Bayesian computation method as in references <sup>1,11,12</sup>, Model M obtains the posterior distributions of all the parameters in the model. These include the parameters in the earlier versions of Model M as well as the  $\alpha$  parameters. However, the treatment efficacy parameters (i.e., hazard ratios) for survival in the previous versions of Model M are now replaced by hazard ratios for time from diagnosis to disease recurrence. The resulting posterior distribution of breast-cancer mortality over time in Model M was based on 172 accepted parameter sets. These formed the basis of their simulations of breast cancer incidence and mortality in the eight counterfactual scenarios.

### **Computation of the relative contributions to mortality reduction associated with each cancer control intervention**

### *Previous work*

In our prior work <sup>1,13</sup>, we considered the effect of two cancer control interventions on breast cancer mortality reduction, namely screening and treatment of stage I-III breast cancer. To compute their relative contributions to mortality reduction, we used the following notation:

- *MR*(*scr*): mortality reduction associated with screening only
- $MR(tx)$ : mortality reduction associated with treatment only
- *RC*(*scr*): relative contribution to mortality reduction associated with screening only

### $R_C(tx)$ : relative contribution to mortality reduction associated with treatments only

We computed relative contributions as follows:

$$
RC(scr) = \frac{MR(scr)}{MR(scr) + MR(tx)} = 1 - RC(tx)
$$

We refer to this approach as the "symmetrical approach."

Because  $MR(scr) + MR(tx) \neq MR(scr, tx)$ , there is an interaction between screening and treatment. Consequently, two other approaches to computing the relative contributions are possible, which we call the "asymmetrical approaches" as their results vary depend on which intervention we consider first. If screening is considered first, then the relative contributions are:

$$
RC'(scr) = \frac{MR(scr)}{MR(scr, tx)} = 1 - RC'(tx)
$$

If adjuvant treatment is considered first, then the relative contributions are:

$$
RC''(tx) = \frac{MR(tx)}{MR(scr, tx)} = 1 - RC''(scr)
$$

We previously showed that the symmetrical approach is roughly equal to the average of the two asymmetrical approaches. For this reason, we report the results from the symmetrical approach as our main findings.

#### *Current work*

After incorporating metastatic recurrence into the models, there are now three interventions contributing to mortality reduction, namely: screening, Stage I-III adjuvant treatment and metastatic treatment, which we denote as *scr*, *tx\_early* and *tx\_met, respectively*. Now there are 127 possible approaches to compute, which are summarized in **eFigure 5**.

The most straightforward approach is the symmetrical one, which is computed as follows:

$$
RC(scr) = \frac{MR(scr)}{MR(scr) + MR(tx\_early) + MR(tx\_met)}
$$

$$
RC(tx\_early) = \frac{MR(tx\_early)}{MR(scr) + MR(tx\_early) + MR(tx\_met)}
$$

$$
RC(tx\_met) = \frac{MR(tx\_met)}{MR(scr) + MR(tx\_early) + MR(tx\_met)}
$$

For the other approaches, we will go through one example to demonstrate how they differ. Of note, the first key distinction is how we choose to compute the total mortality reduction associated with all three interventions. Indeed, one could say it is simply equal to  $MR(scr, tx\_early, tx\_met)$  (mortality reduction associated with the counterfactual scenario with all three interventions), however in the world of counterfactuals, we could also say it is equal to  $MR(scr) + MR(tx\_early, tx\_met).$ 

The difference between these approaches is the way we handle interaction terms between each intervention. In actuality, we are choosing one partition of a set with three elements. For instance, the symmetrical approach is derived from the partition with singleton subsets.

Let's say we choose  $MR(tx\_early) + MR(sr, tx\_met)$ . Now the order with which we decide to compute relative contributions matters (unlike the symmetrical approach…). If we decide to consider screening first, we can either compute it first, or last, which gives, respectively:

$$
RC(scr) = \frac{MR(scr)}{MR(tx\_early) + MR(scr, tx\_met)} = 1 - RC(tx\_early, tx\_met)
$$

$$
RC(scr) = 1 - \frac{MR(tx\_early, tx\_met)}{MR(tx\_early) + MR(scr, tx\_met)} = 1 - RC(tx\_early, tx\_met)
$$

Note that if we consider adjuvant treatment, computing it first or last yields the same results.

Next, to separate the relative contribution associated with adjuvant and metastatic treatments, we can proceed as in the previous work when only two interventions were considered: we can either select one of the two asymmetrical approaches or the symmetrical approach. For example:

$$
RC(tx\_early) = RC(tx\_early, tx\_met) \times \frac{MR(tx\_early)}{MR(tx\_early, tx\_met)}
$$

However, we could also consider these three computations in the presence of screening. Using the same example as above:

$$
RC(tx\_early) = RC(tx\_early, tx\_met) \times \frac{MR(tx\_early, scr)}{MR(tx\_early, tx\_met, scr)}
$$

With this final step, we covered all possible approaches to compute the relative contributions.

These approaches yield differing estimates of the component of the mortality reduction attributed to each intervention because of interactions between the three interventions. For example, if we consider *MR(tx\_early, tx\_met, scr)* as the denominator – that is, the mortality reduction in the presence of all three interventions, or the "real world" scenario counterfactual – we find that, across all models, each of the interventions taken alone represents a higher proportion of this overall mortality reduction than we report in the symmetrical approach. By the symmetrical approach, Model D estimates Stage I-III treatment to account for 35% of the overall mortality reduction, but when the scenario of Stage I-III treatment is considered in the absence of screening or metastatic treatment, it alone achieves 49% of that overall mortality reduction (for Model M it is 60% vs 70%; for Model S 44% vs 57%; and for Model W 47% vs 63%). This is not surprising: without screening, disease is diagnosed at a later stage and therefore associated with a worse baseline survival curve, and so the impact of treatment is greater given that the treatment effect is modeled as a proportional hazard. Similarly, by the symmetrical approach Model D estimates metastatic treatment to account for 33% of the overall mortality reduction, but when the scenario of metastatic treatment is considered in the absence of screening or Stage I-III treatment, it alone achieves 46% of that overall mortality reduction (for Model M it is 19% vs 24%; for Model S 31% vs 40%; and for Model W 32% vs 43%). Here, the absence of Stage I-III treatment means both that more patients develop metastatic disease and that the impact of that treatment is greater, because of the absence of resistance to previously received therapies. In total, the sum of the mortality reductions from the three single interventions exceeds the total mortality reduction in all four models – 141% for Model D, 127% for Model M, 130% for Model S, and 134% for Model W – underscoring the importance of the interactions between the interventions.

In the main paper, we reported the symmetrical approach, because it is relatively close to the average of all approaches (**eFigure 6**) and it maintains consistency with our previous work.

## **Online-Only References**

1. Plevritis SK, Munoz D, Kurian AW, et al. Association of Screening and Treatment With Breast Cancer Mortality by Molecular Subtype in US Women, 2000-2012. *JAMA*. 2018;319:154. doi:10.1001/jama.2017.19130

2. Kurian AW, Bondarenko I, Jagsi R, et al. Recent Trends in Chemotherapy Use and Oncologists' Treatment Recommendations for Early-Stage Breast Cancer. *J Natl Cancer Inst*. May 1 2018;110(5):493-500. doi:10.1093/jnci/djx239

3. Killelea BK, Yang VQ, Wang SY, et al. Racial Differences in the Use and Outcome of Neoadjuvant Chemotherapy for Breast Cancer: Results From the National Cancer Data Base. *J Clin Oncol*. Dec 20 2015;33(36):4267-76. doi:10.1200/JCO.2015.63.7801

4. Kurian AW, Abrahamse P, Hamilton AS, et al. Chemotherapy Regimens Received by Women With BRCA1/2 Pathogenic Variants for Early Stage Breast Cancer Treatment. *JNCI Cancer Spectr*. Jul 1 2022;6(4)doi:10.1093/jncics/pkac045

5. Francis PA, Pagani O, Fleming GF, et al. Tailoring Adjuvant Endocrine Therapy for Premenopausal Breast Cancer. *N Engl J Med*. Jul 12 2018;379(2):122-137. doi:10.1056/NEJMoa1803164

6. Sikov WM, Berry DA, Perou CM, et al. Impact of the addition of carboplatin and/or bevacizumab to neoadjuvant once-per-week paclitaxel followed by dose-dense doxorubicin and cyclophosphamide on pathologic complete response rates in stage II to III triple-negative breast cancer: CALGB 40603 (Alliance). *J Clin Oncol*. Jan 1 2015;33(1):13-21. doi:10.1200/JCO.2014.57.0572

7. Gradishar WJ, Anderson BO, Abraham J, et al. Breast Cancer, Version 3.2020, NCCN Clinical Practice Guidelines in Oncology. *J Natl Compr Canc Netw*. Apr 2020;18(4):452-478. doi:10.6004/jnccn.2020.0016

8. Orange Book: Approved Drug Products with Therapeutic Equivalence Evaluations. U.S. Food & Drug Administration. 2022. https://www.accessdata.fda.gov/scripts/cder/ob/index.cfm

9. Dieras V, Miles D, Verma S, et al. Trastuzumab emtansine versus capecitabine plus lapatinib in patients with previously treated HER2-positive advanced breast cancer (EMILIA): a descriptive analysis of final overall survival results from a randomised, open-label, phase 3 trial. *Lancet Oncol*. Jun 2017;18(6):732-742. doi:10.1016/S1470-2045(17)30312-1

10. Caswell-Jin JL, Callahan A, Purington N, et al. Treatment and Monitoring Variability in US Metastatic Breast Cancer Care. *JCO Clin Cancer Inform*. May 2021;5:600-614. doi:10.1200/CCI.21.00031

11. Huang X, Li Y, Song J, Berry DA. A Bayesian Simulation Model for Breast Cancer Screening, Incidence, Treatment, and Mortality. *Med Decis Making*. Apr 2018;38(1 suppl):78S-88S. doi:10.1177/0272989X17714473

12. Berry DA, Inoue L, Shen Y, et al. Modeling the impact of treatment and screening on U.S. breast cancer mortality: a Bayesian approach. *J Natl Cancer Inst Monogr*. 2006;(36):30-6. doi:10.1093/jncimonographs/lgj006

13. Berry DA, Cronin KA, Plevritis SK, et al. Effect of screening and adjuvant therapy on mortality from breast cancer. *N Engl J Med*. Oct 27 2005;353(17):1784-92. doi:10.1056/NEJMoa050518

14. Munoz DF, Xu C, Plevritis SK. A Molecular Subtype-Specific Stochastic Simulation Model of US Breast Cancer Incidence, Survival, and Mortality Trends from 1975 to 2010. *Med Decis Making*. Apr 2018;38(1\_suppl):89S-98S. doi:10.1177/0272989X17737508

15. Gangnon RE, Sprague BL, Stout NK, et al. The contribution of mammography screening to breast cancer incidence trends in the United States: an updated age-period-cohort model. *Cancer Epidemiol Biomarkers Prev*. Jun 2015;24(6):905-12. doi:10.1158/1055-9965.EPI-14- 1286

16. Rosenberg MA. Competing risks to breast cancer mortality. *J Natl Cancer Inst Monogr*. 2006;(36):15-9. doi:10.1093/jncimonographs/lgj004

17. Gangnon RE, Stout NK, Alagoz O, Hampton JM, Sprague BL, Trentham-Dietz A. Contribution of Breast Cancer to Overall Mortality for US Women. *Med Decis Making*. Apr 2018;38(1\_suppl):24S-31S. doi:10.1177/0272989X17717981

18. Mandelblatt JS, Near AM, Miglioretti DL, et al. Common Model Inputs Used in CISNET Collaborative Breast Cancer Modeling. *Med Decis Making*. Apr 2018;38(1\_suppl):9S-23S. doi:10.1177/0272989X17700624

19. Early Breast Cancer Trialists' Collaborative G. Effects of chemotherapy and hormonal therapy for early breast cancer on recurrence and 15-year survival: an overview of the randomised trials. *Lancet*. May 14-20 2005;365(9472):1687-717. doi:10.1016/S0140- 6736(05)66544-0

20. Early Breast Cancer Trialists' Collaborative G, Peto R, Davies C, et al. Comparisons between different polychemotherapy regimens for early breast cancer: meta-analyses of longterm outcome among 100,000 women in 123 randomised trials. *Lancet*. Feb 4 2012;379(9814):432-44. doi:10.1016/S0140-6736(11)61625-5

21. Perez EA, Romond EH, Suman VJ, et al. Trastuzumab plus adjuvant chemotherapy for human epidermal growth factor receptor 2-positive breast cancer: planned joint analysis of overall survival from NSABP B-31 and NCCTG N9831. *J Clin Oncol*. Nov 20 2014;32(33):3744-52. doi:10.1200/JCO.2014.55.5730

22. Piccart M, Procter M, Fumagalli D, et al. Adjuvant Pertuzumab and Trastuzumab in Early HER2-Positive Breast Cancer in the APHINITY Trial: 6 Years' Follow-Up. *J Clin Oncol*. May 1 2021;39(13):1448-1457. doi:10.1200/JCO.20.01204

23. Masuda N, Lee SJ, Ohtani S, et al. Adjuvant Capecitabine for Breast Cancer after Preoperative Chemotherapy. *N Engl J Med*. Jun 1 2017;376(22):2147-2159. doi:10.1056/NEJMoa1612645

24. Chan A, Moy B, Mansi J, et al. Final Efficacy Results of Neratinib in HER2-positive Hormone Receptor-positive Early-stage Breast Cancer From the Phase III ExteNET Trial. *Clin Breast Cancer*. Feb 2021;21(1):80-91 e7. doi:10.1016/j.clbc.2020.09.014

25. Jassem J, Pienkowski T, Pluzanska A, et al. Doxorubicin and paclitaxel versus fluorouracil, doxorubicin and cyclophosphamide as first-line therapy for women with advanced breast cancer: long-term analysis of the previously published trial. *Onkologie*. Sep 2009;32(8- 9):468-72. doi:10.1159/000226210

26. Gibson L, Lawrence D, Dawson C, Bliss J. Aromatase inhibitors for treatment of advanced breast cancer in postmenopausal women. *Cochrane Database Syst Rev*. Oct 7 2009;(4):CD003370. doi:10.1002/14651858.CD003370.pub3

27. O'Shaughnessy J, Miles D, Vukelja S, et al. Superior survival with capecitabine plus docetaxel combination therapy in anthracycline-pretreated patients with advanced breast cancer: phase III trial results. *J Clin Oncol*. Jun 15 2002;20(12):2812-23. doi:10.1200/JCO.2002.09.002

28. Slamon DJ, Leyland-Jones B, Shak S, et al. Use of chemotherapy plus a monoclonal antibody against HER2 for metastatic breast cancer that overexpresses HER2. *N Engl J Med*. Mar 15 2001;344(11):783-92. doi:10.1056/NEJM200103153441101

29. Lee CI, Goodwin A, Wilcken N. Fulvestrant for hormone-sensitive metastatic breast cancer. *Cochrane Database Syst Rev*. Jan 3 2017;1:CD011093. doi:10.1002/14651858.CD011093.pub2

30. Cortes J, O'Shaughnessy J, Loesch D, et al. Eribulin monotherapy versus treatment of physician's choice in patients with metastatic breast cancer (EMBRACE): a phase 3 open-label randomised study. *Lancet*. Mar 12 2011;377(9769):914-23. doi:10.1016/S0140-6736(11)60070-6

31. Swain SM, Miles D, Kim SB, et al. Pertuzumab, trastuzumab, and docetaxel for HER2 positive metastatic breast cancer (CLEOPATRA): end-of-study results from a double-blind, randomised, placebo-controlled, phase 3 study. *Lancet Oncol*. Apr 2020;21(4):519-530. doi:10.1016/S1470-2045(19)30863-0

32. Hortobagyi GN, Stemmer SM, Burris HA, et al. Overall Survival with Ribociclib plus Letrozole in Advanced Breast Cancer. *N Engl J Med*. Mar 10 2022;386(10):942-950. doi:10.1056/NEJMoa2114663

33. Hurvitz SA, McAndrew NP, Bardia A, et al. A careful reassessment of anthracycline use in curable breast cancer. *NPJ Breast Cancer*. Oct 8 2021;7(1):134. doi:10.1038/s41523-021- 00342-5

# **eTable 1. Breast cancer model input parameters**





SEER = Surveillance, Epidemiology, and End Results Registry. AJCC = American Joint Committee on Cancer. BCSC = Breast Cancer Surveillance Consortium. ER = estrogen receptor. NCCN = National Comprehensive Cancer Network.

Italicized parameters are new in these models.

## **eTable 2. NCCN Outcomes Database cohort characteristics**







## **eTable 3. Breast cancer treatments and their efficacy, 1975-2019.**



<sup>a</sup> Hazard ratio is for disease-free or recurrence-free survival (as available) for Stage I-III treatments, and for breast cancer–specific survival for treatments after metastasis.

b Dissemination of these treatments follows data from Surveillance, Epidemiology, and End Results (SEER) special patterns of care studies and the National Comprehensive Cancer Network database<sup>1</sup>.

<sup>c</sup> Chemotherapy prior to 1975 included cyclophosphamide, methotrexate, 5-FU and anthracycline-containing regimens <sup>33</sup>.

5-FU = 5-fluorouracil; ER, estrogen receptor; TNBC, triple-negative breast cancer

<sup>d</sup> There were no clinical trials comparing chemotherapy to placebo or tamoxifen to placebo in the metastatic setting. These hazard ratios were selected

to fit within the range of benefits observed for chemotherapy and tamoxifen compared to no treatment in stage I-III breast cancer, as well as the range of

benefits observed for new chemotherapies and endocrine therapies compared to prior in metastatic breast cancer.

**eTable 4. Model M raw hazard reductions for overall survival after metastasis by ER/ERBB2 subtype and year based on first-line metastatic therapy**





**eTable 5. Estimated absolute breast cancer age-adjusted mortality rate and its reduction relative to no intervention by subtype across eight scenarios in 2019, by CISNET model** 



\* "Abs" (absolute) refers to the predicted age-adjusted breast cancer mortality rate per 100,000 women under each scenario. † "Rel" (relative) refers to the percentage reduction in that mortality rate as compared to predicted mortality with no interventions ("none"). ER = estrogen receptor, met = metastatic.

**eTable 6. Estimated median breast cancer–specific survival after distant recurrence over time by estrogen receptor/ERBB2 subtype, by CISNET model** 











\* Year is the year of diagnosis of metastatic recurrence. ER = estrogen receptor. Figure 3A of the main paper is a graphical display of the model mean results from 2000 to 2019.

## **eTable 7. Estimated five and ten-year distant recurrence-free survival over time by estrogen receptor/ERBB2 subtype**











\* Year is the year of diagnosis of stage I-III breast cancer. ER = estrogen receptor. Figure 3B of the main paper is a graphical display of the average results from 2000-2019.

## **eFigures**





ER = estrogen receptor.

## **eFigure 2. Distant recurrence-free survival by subtype in the NCCN Outcomes Database as compared to Model S**



To allow comparability of distant recurrence-free survival between the model and the NCCN Outcomes Database, model outputs were censored in accordance with the pattern of time of last follow-up for patients in the NCCN Outcomes Database. ER = estrogen receptor.

### **eFigure 3. Survival after metastasis as observed in clinical trials vs predicted from Model S**



### A. Estrogen receptor-positive, ERBB2-negative

Survival curves from five first-line clinical trials for metastatic breast cancer (solid lines) and the outputs of Model S (Stanford) (dotted lines) are shown across three ER/ERBB2 subtypes (A-C). Overall survival from trial enrollment is compared to model-estimated breast cancer-specific survival after metastasis. Trial control arms, reflecting standard-of-care therapy at the time of trial conduct, are shown. The simulated patients included in the model outputs were diagnosed with metastatic disease during the years the trial was conducted, received first-line therapy, and were sampled based on age at diagnosis of metastatic disease (all trials), receipt of endocrine and/or chemotherapy for Stage I-III disease (all trials), recurrence-free interval (SWOG 0026, MONALEESA-2, MONALEESA-7), and estrogen receptor status (CLEOPATRA) to correspond to the trial population.



### **eFigure 4. Metastatic therapy usage by year of diagnosis of metastatic recurrence**

The probability of a simulated patient with metastatic recurrence receiving the benefit of each drug available for the treatment of metastatic disease is plotted against the year of diagnosis of metastatic recurrence. These results are produced by the assumptions about metastatic therapy uptake outlined in the **eMethods**. Because simulated patients may receive multiple (or no) lines of treatment, the proportions do not add up to 1.0. Notably, some of these benefits are given to simulated patients in the absence of them receiving that drug, if the drug they do receive demonstrated its benefit over another drug; for example, the benefit of tamoxifen is given to any simulated patient who receives aromatase inhibitor, as aromatase inhibitors demonstrated their benefit over tamoxifen. These results apply to Models D, S, and W. A = anthracycline, Tamox = tamoxifen, Pacli = Paclitaxel (or any taxane), AI = aromatase inhibitor, Capec = capecitabine, Fulv = fulvestrant, Erib = eribulin, Abema = abemaciclib (or any CDK4/6 inhibitor), Trastuz = trastuzumab, Pertuz = pertuzumab, TDM1 = trastuzumab emtansine.

## **eFigure 5. Summary of 127 approaches to calculate contributions of interventions to breast cancer mortality reduction**



The three interventions are screening (scr), Stage I-III treatment (tx\_early), and metastatic treatment (tx\_met). For more details, please see eMethods.

**eFigure 6. Comparison of symmetrical to asymmetrical approaches to calculate contributions of interventions to breast cancer mortality reduction** 



Screening is on the left, Stage I-III treatment in the middle, and metastatic treatment on the right for each of the four models. Calculated relative contributions are shown over time for all 127 approaches, with the symmetrical approach shown in thick blue. For more details, please see eMethods.

## **eFigure 7. Estimated age-adjusted breast cancer mortality over time across eight scenarios compared to observed mortality, by CISNET model**







Calendar year

Models D, S, and W use a shared age-period-cohort model <sup>1</sup> to estimate breast cancer incidence in the absence of screening. Model M uses a linear model of increasing annual incidence starting in 1975, whose parameters were selected from Bayesian inference to match observed incidence (SEER = Surveillance, Epidemiology, and End Results). Observed incidence is in the presence of screening.

**eFigure 9. Associations with overall breast cancer mortality reduction of screening, Stage I-III treatments, and metastatic treatments in 2019, by CISNET model and within Model M** 



The percentage of the estimated overall mortality reduction in 2019 attributable to each of the three categories of intervention from the four models is plotted. The point estimates from the four models are shown in large colored dots. The smaller gray dots represent pairs of estimates based on the 172 simulations of Model M; gray contours represent the contour lines of two-dimensional kernel estimation (n=100, h=25) for the contour levels of 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 0.95, and 0.99. Model M uses a Bayesian approach to incorporate parameter uncertainty, including in treatment efficacy, that is reflected in this distribution of estimates. Of note, the within-model variability from model M is comparable to the between-model variability.

## **eFigure 10. Associations with overall breast cancer mortality reduction of screening, Stage I-III treatments, and metastatic treatments over time, by CISNET model**



All interventions (Stage I-III treatment, metastatic treatments, and screening) are in addition to standard treatments available in 1975. Because local therapy was part of standard-of-care treatment for Stage I-III disease in 1975, the benefit of screening occurs in the presence of standard local therapy.