

## **Multimodal Artificial Intelligence (AI) Models Integrating Image, Clinical, and Molecular Data for Predicting Early and Late Breast Cancer Recurrence in TAILORx**

Joseph A. Sparano<sup>1</sup>, Norsang Lama<sup>2</sup>, Robert J. Gray<sup>3</sup>, Md Ashequr Rahman<sup>2</sup>, Victoria Wang<sup>3</sup>, Della F. Makower<sup>4</sup>, Yating Cheng<sup>2</sup>, Kathy S. Albain<sup>5</sup>, Ming Chen<sup>2</sup>, Daniel F. Hayes<sup>6</sup>, Anthony Helmstetter<sup>2</sup>, Charles E. Geyer, Jr.<sup>7</sup>, Casey Bales<sup>2</sup>, Elizabeth C. Dees<sup>8</sup>, Matthew P. Goetz<sup>9</sup>, John A. Olson, Jr.<sup>10</sup>, Sunil S. Badve<sup>11</sup>, Thomas J. Saphner<sup>12</sup>, Timothy J. Whelan<sup>13</sup>, Virginia G. Kaklamani<sup>14</sup>, Matthew Oberley<sup>2</sup>, Milan Radovich<sup>2</sup>, David Spetzler<sup>2</sup>, Eleftherios P. Mamounas<sup>15,16</sup>, Norman Wolmark<sup>16</sup>, George W. Sledge<sup>2</sup>.

Icahn School of Medicine at Mount Sinai, Tisch Cancer Institute, NY, NY<sup>1</sup>, Caris Life Sciences, Irving, TX<sup>2</sup>, ECOG-ACRIN Biostatistical Center, Dana-Farber Cancer Institute, Boston, MA<sup>3</sup>, Montefiore Einstein Comprehensive Cancer Center, Bronx, NY<sup>4</sup>, Loyola University Medical Center, Maywood, IL<sup>5</sup>, University of Michigan, Ann Arbor, MI<sup>6</sup>, University of Pittsburgh, Pittsburgh, PA<sup>7</sup>, University of North Carolina, Chapel Hill, NC<sup>8</sup>, Mayo Clinic, Rochester, MN<sup>9</sup>, Washington University School of Medicine, St. Louis, MO<sup>10</sup>, Winship Cancer Institute, Emory University, Atlanta, GA<sup>11</sup>, Aurora Medical Center, Two Rivers, WI<sup>12</sup>, McMaster University, Hamilton, CAN<sup>13</sup>, University of Texas Health, San Antonio, TX<sup>14</sup>, AdventHealth Cancer Institute, Orlando, FL<sup>15</sup>, National Surgical Adjuvant Breast and Bowel Project, Pittsburgh, PA<sup>16</sup>

# Disclosure Information

## Joseph A. Sparano, MD

I have the following relevant financial relationships to disclose:

Employee of: Icahn School of Medicine at Mount Sinai, Tisch Cancer Institute, New York, NY

Consultant for: Astra Zeneca, Genentech/Roche, Novartis, Pfizer,

Speaker's Bureau for: None

Grant/Research support from: National Cancer Institute, Department of Defense, Breast Cancer Research Foundation

Stockholder in: None

Honoraria from: None

# Background: Trial Assigning Individualized Options for Treatment (TAILORx)

## ■ Rationale:

- Oncotype DX (ODX) 21-gene RS was known to be prognostic for distant recurrence & predictive of chemotherapy benefit in ER+, HER2-neg, node-negative breast cancer

## ■ Design:

- TAILORx was designed to address uncertainty of CT benefit in mid-range RS 11-25, and prospectively confirm the favorable prognosis with ET alone for a very low RS 0-10

## ■ Results:

- Adjuvant ET is non-inferior to CET for 69% with a 21-gene RS 11–25 (primary endpoint)
  - ❑ Exploratory analysis: age  $\leq 50$  years had some CT benefit that varied with RS and clinical risk
- RS 0–10 is associated with low DR rates with ET alone
- Integration of RS with clinicopathologic features using the RSClin tool was more prognostic for distant recurrence than either used alone

# Objective and Methods

## ■ Objective:

- Develop a diagnostic test with better distant recurrence (DR) risk prognostication than the Oncotype DX (ODX) 21-gene recurrence score (RS)

## ■ Sample Analysis:

- Primary FFPE tumor samples analyzed for whole transcriptome sequencing (WTS) and whole exome sequencing (WES) by Caris MI Tumor Seek–Hybrid
- H&E images were digitized at 40X and analyzed at 10x by deep learning

## ■ Statistical Analysis:

- Models reflecting clinical (C), expanded molecular (M+), and pathomic imaging (I) features
- Model training & 5-fold nested cross validation (60/20/20) for distant recurrence
- Model performance was compared with the actual ODX RS
- Assessed using the concordance index (C-index), hazard ratios (HR), & log-rank tests
- Best performing models independently evaluated in holdout validation set

# Methods: REMARK Summary for Training/5-fold Cross-Validation and **Holdout Validation** Sets

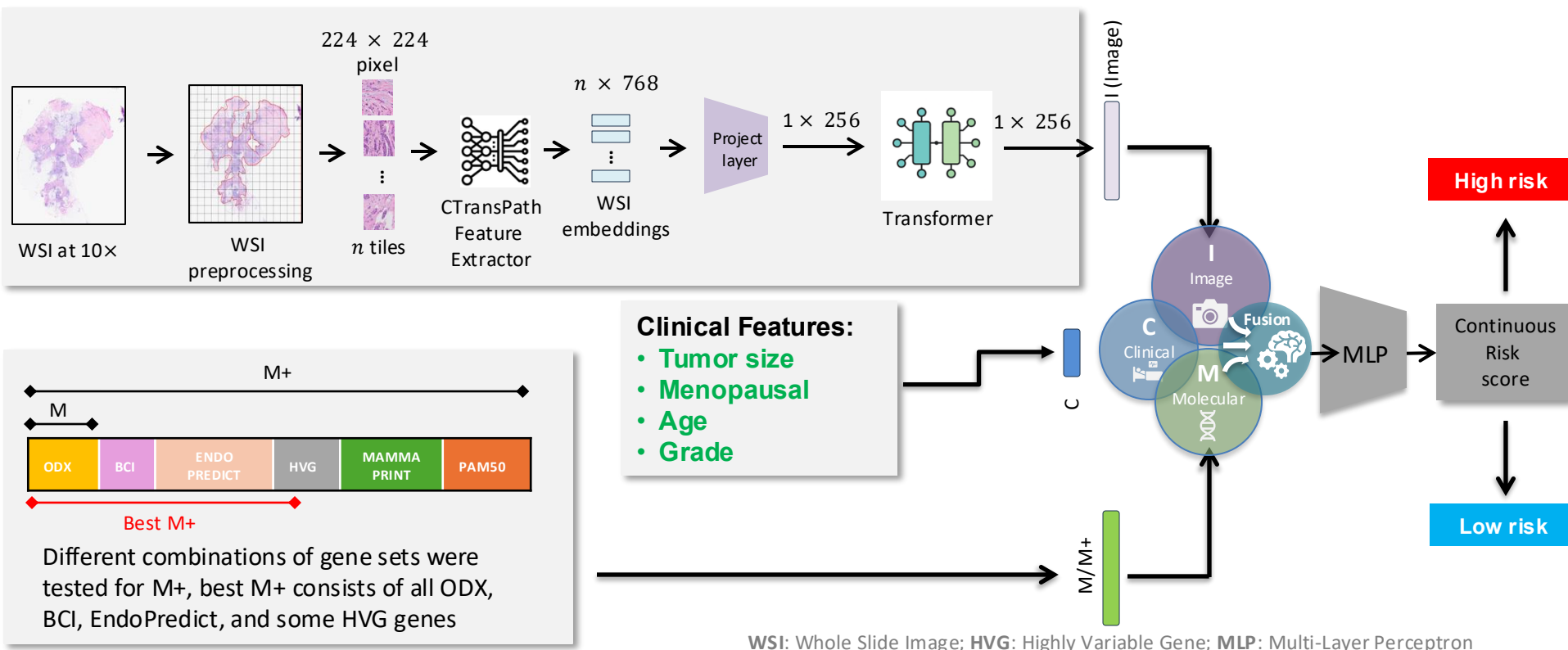
- **10,273 enrolled in TAILORx**
  - 8,062 (78%) with FFPE tissue blocks
  - 6,530 (64%) with evaluable H&E 40x WSI generated by Pramana scanner
  
- **4,462 (43%) cases with WSI & WTS**
  - **2,808 training/5-fold CV (63%)**
    - 2482 with WES used for exploratory modeling
  - **1,621 holdout validation (37%)**

Training set: ECOG-ACRIN, Alliance (ACOSOG, CALGB, NCCTG) & CCTG  
Holdout set : enrolled through SWOG, NRG (NSABP, RTOG, GOG) & CTSU

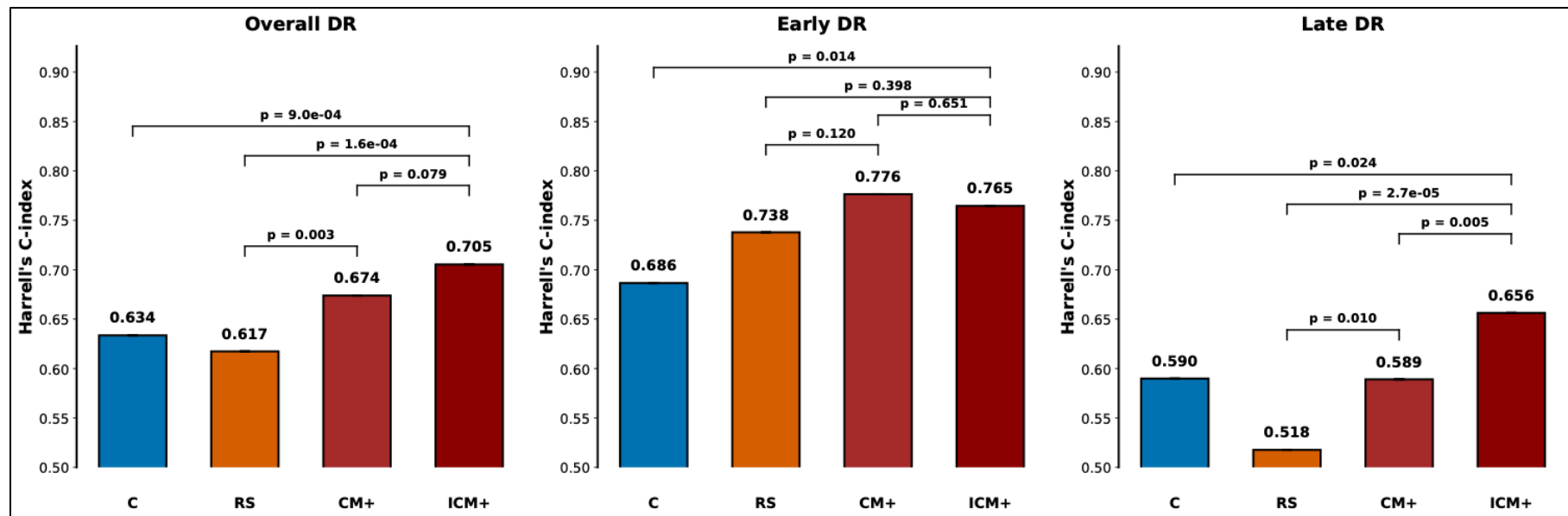
## Clinicopathologic characteristics

- **Training & validation set not significantly different:**
    - Median age: 56 years (range 23-75 years)
    - Postmenopausal: 66.2%
    - Tumor size  $\leq 2$  cm: 72.4%
    - PR positive: 90.1%
    - Low clinical risk: 66.9%
  - Statistically significant but clinically marginal differences:
    - High grade: 20.7% vs. 19.7% (p=0.004)
    - Low grade: 24.2% vs. 28.4%
- 
- **Included vs. excluded not significantly different:**
    - Age, menopausal status, PR, surgery, arm
  - Statistically significant but clinically marginal differences:
    - Tumor size  $\leq 2$  cm: 72.4% vs. 76.9% (p<0.001)
    - High grade: 19.7% vs. 17.2% (p=0.005)
    - Low grade: 25.8% vs. 26.7%
    - Low clinical risk: 66.9% vs. 71.9% (p<0.001)

# Multimodal Model Architecture



# Results: C-index for Optimal ICM+ and CM+ Models vs. Clinical Model and ODX RS in Training/5-fold CV Set



## Overall DR at 15 Years

- ICM+ & CM+ model performance significantly better than C model & RS
- ICM+ model had highest C-index, but not significantly better than CM+

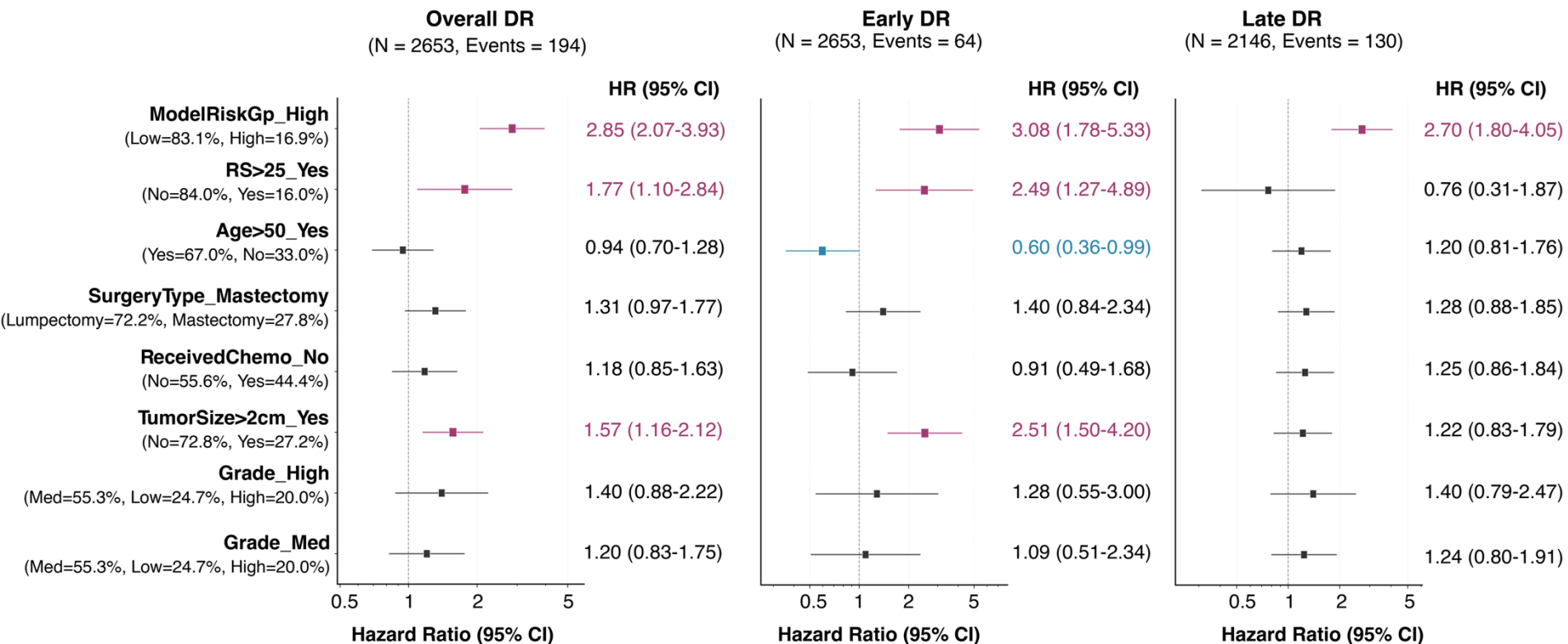
## Early DR $\leq 5$ Years

- ICM+ and CM+ models exhibited highest C-index, better than C model
- ICM+ and CM+ models numerically but not significantly better than RS

## Late DR $> 5$ Years

- ICM+ model performance significantly better than C, RS, and CM+

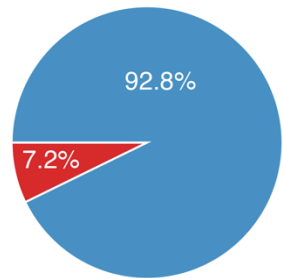
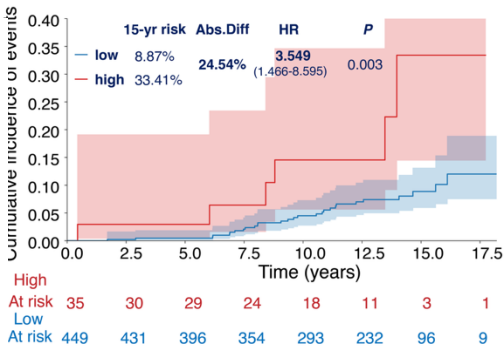
# Results: Multivariate Analysis for Distant Recurrence Including the ICM+ Model and Other Covariates in the Training/5-fold CV Set





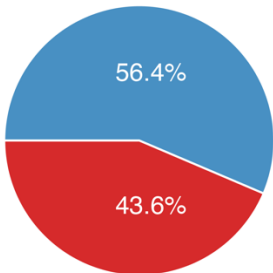
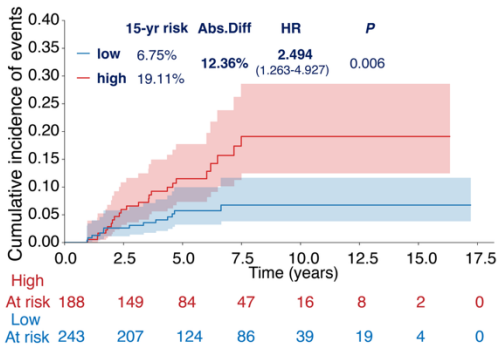
# Results: Distribution of ICM+ Risk Groups by RS-based Treatment Arms and 15-year Distant Recurrence Risk in Training/5-fold CV Set

RS0-10: ET



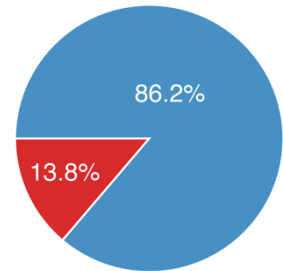
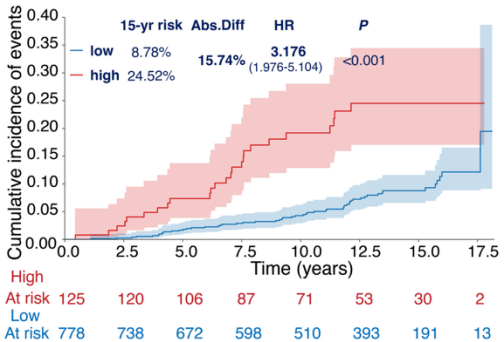
High Risk  
Low Risk

RS26-100: CET

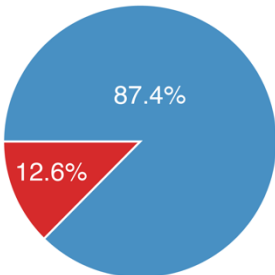
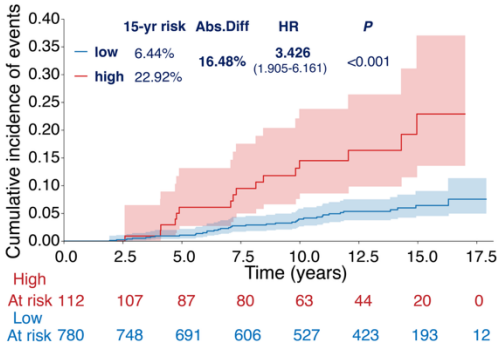


High Risk  
Low Risk

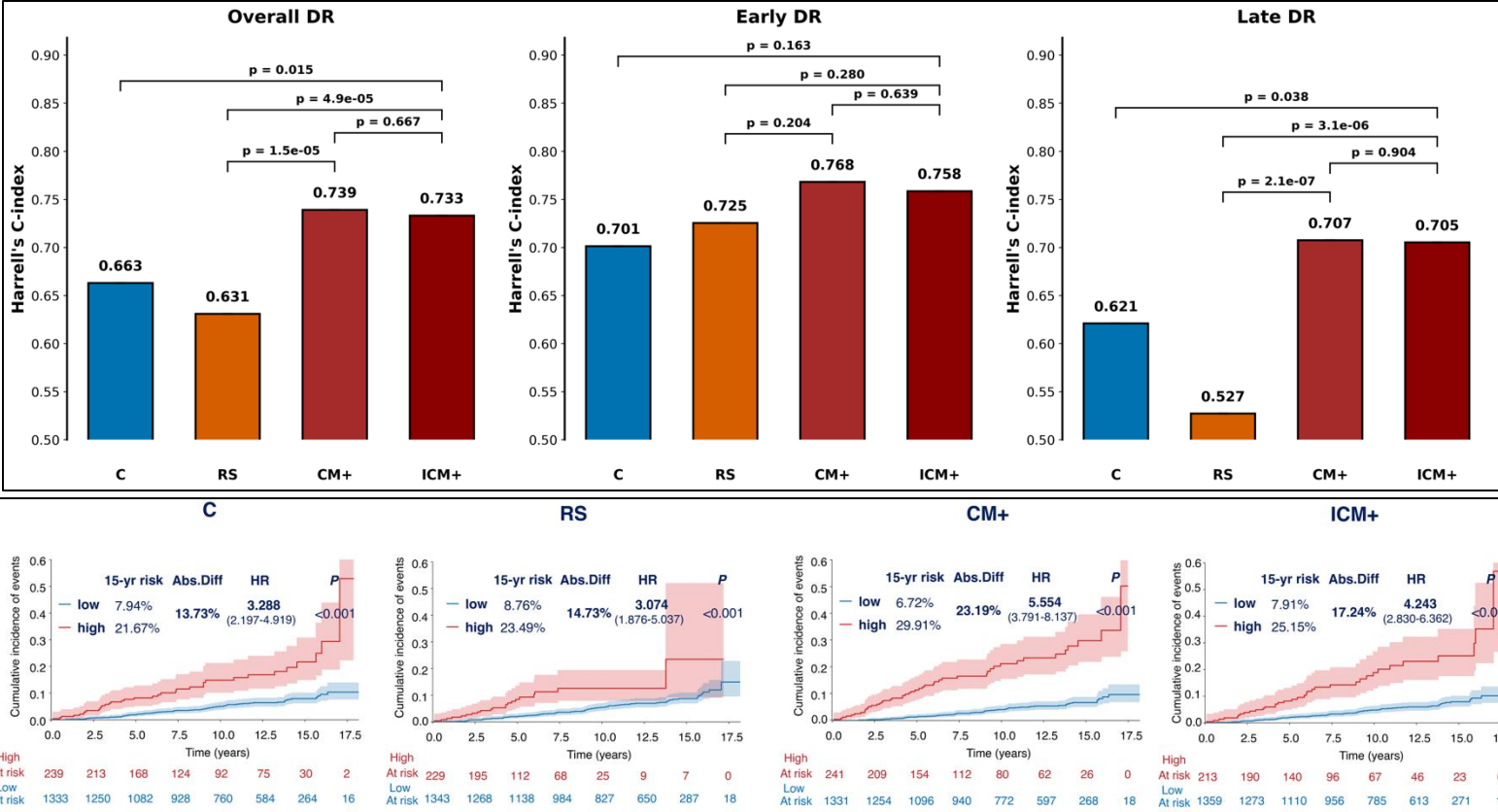
RS11-25: ET



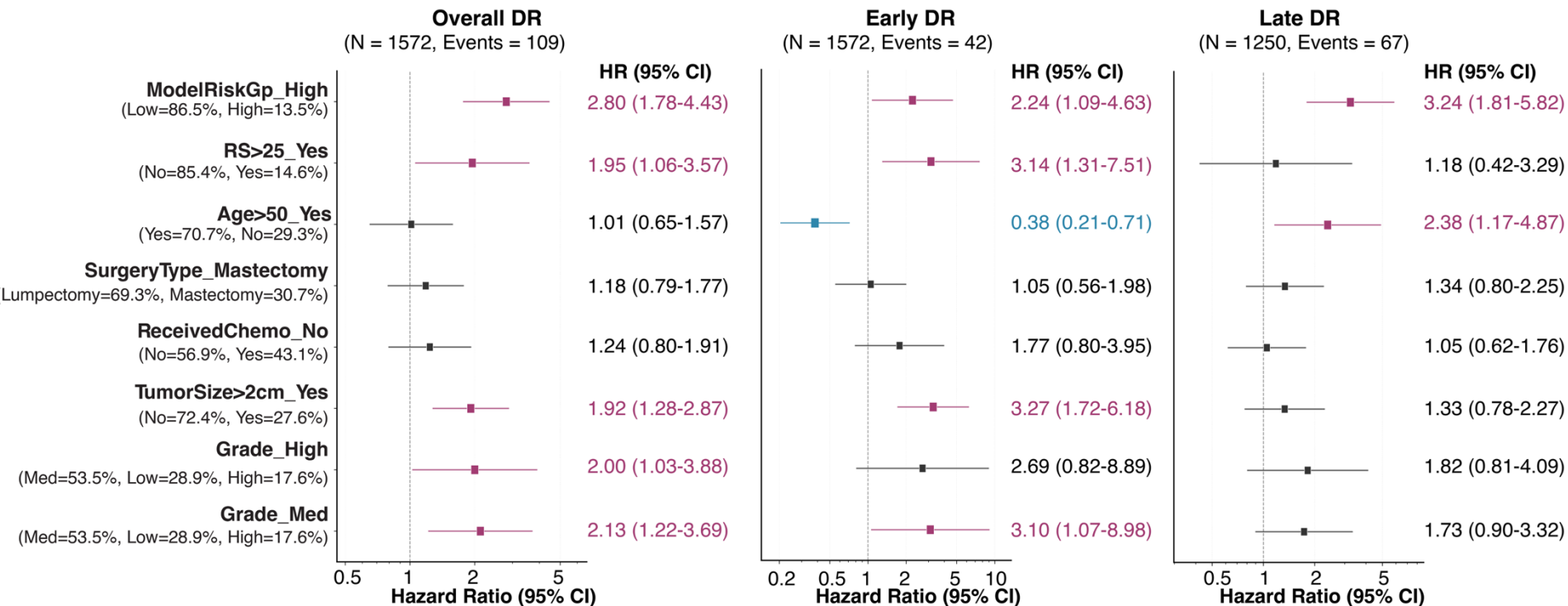
RS11-25: CET



# Results: C-index for Optimal ICM+ and CM+ Models vs. Clinical Model and ODX RS in **Holdout Validation Set**

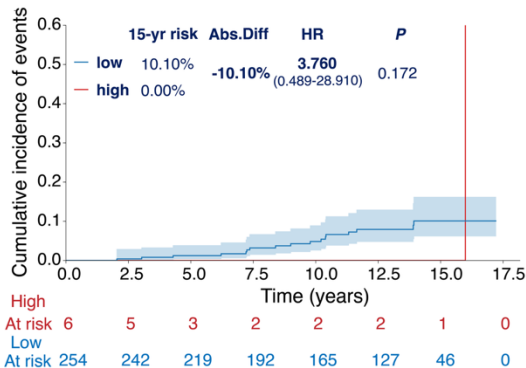


# Results: Multivariate Analysis for Distant Recurrence Including the ICM+ Model and Other Covariates in the **Holdout Validation Set**

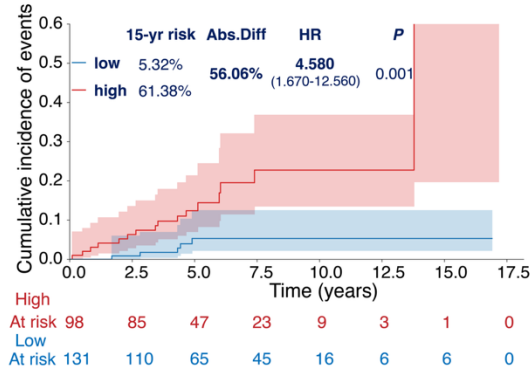


# Results: Distribution of ICM+ Risk Groups by RS-based Treatment Arms and 15-year Distant Recurrence Risk in **Holdout Validation Set**

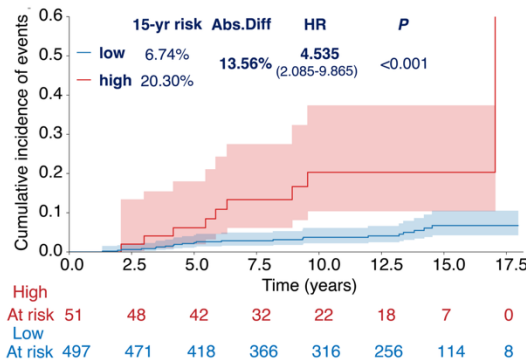
**RS0-10: ET**



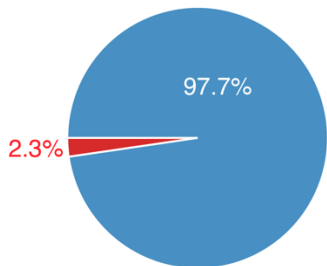
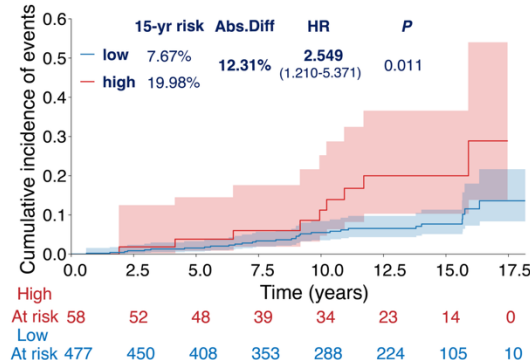
**RS26-100: CET**



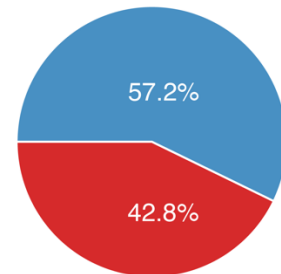
**RS11-25: ET**



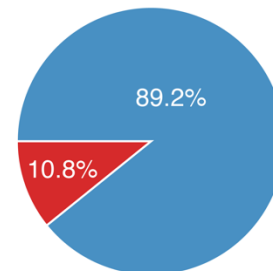
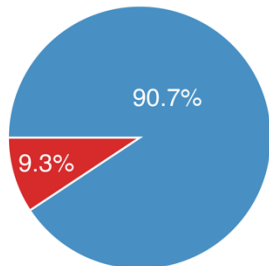
**RS11-25: CET**



High Risk  
Low Risk



High Risk  
Low Risk



# Conclusions

## ■ Training/5-fold cross validation set:

- **C Model:** Clinical features provided some prognostic information
- **M+ Model:** 42-gene molecular model including the 21-gene RS, BCI, and EndoPredict primarily drove prognostic stratification for early DR within 5 years
- **I Model:** Pathomic features strengthened prognostic stratification for late DR > 5 years
- **Multimodal ICM+ Model:**
  - ❑ Strongest prognostic performance for overall and late DR that were superior to the actual ODX RS
  - ❑ Provided statistically significant and clinically relevant prognostic stratification in the low (RS 0-25) and high (RS 26-100) genomic risk groups
  - ❑ Exhibited stronger performance than **CM+** for late recurrence

## ■ Holdout validation set:

- Superior prognostic performance of both the **ICM+** & **CM+** models were validated for overall and late DR
- Both **ICM+** & **CM+** models outperformed the actual Oncotype DX RS used to guide therapy

# Other TAILORx Presentations and EA1241

**RF3-01:** Clinical outcomes of invasive lobular carcinoma (ILC) versus non-lobular breast cancer (NLC) assessed by expert pathologists, an artificial intelligence (AI) CDH1 classifier, and AI-derived tumor microenvironment (TME) biomarkers in TAILORx.

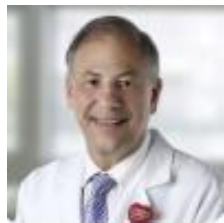
Roberto Salgado et al on behalf International Immuno-Oncology Biomarker Working Group and the TAILORx Investigators

**RF3-07:** A Multimodal-Multitask Deep Learning Model Trained in NSABP B-42 and Validated in TAILORx for Late Distant Recurrence Risk in HR+ Early Breast Cancer.

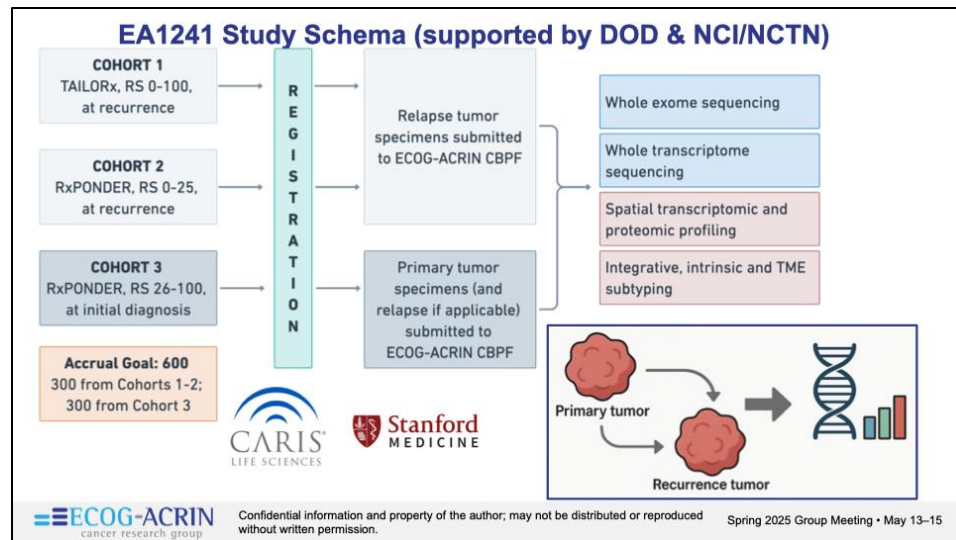
Eleftherios (Terry) Mamounas et al on behalf of NRG/NSABP and the TAILORx Investigators



**Roberto Salgado, MD**  
IO Biomarker WG



**Terry Mamounas, MD, MPH**  
NRG/NSABP



**Kevin Kalinsky, MD**  
SWOG



**Christina Curtis, PhD**  
Stanford Medicine



**Rima Patel, MD**  
ECOG-ACRIN

# Acknowledgements

- TAILORx trial volunteers
- Cancer advocacy community
- Health care & research professionals
- Funders
  - Breast Cancer Research Foundation
  - NIH/NCI: Award numbers: U10CA180820, U10CA180794, UG1CA189859, UG1CA233160, UG1CA233184, UG1CA233373, UG1CA232760, UG1CA233339, UG1CA233247, and UG1CA190140
  - U.S. Postal Service Stamp Fund