

# **Assessment of prediction error of risk prediction models**

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# Outline

- Situation
- Measures of prediction error
- Application to prediction of breast cancer survival
- General conclusion
- Considerations for breast cancer risk prediction

## Situation (1)

- **Prediction:**  $\pi(t|X)$

predicted probability that an individual will be event-free up to  $t$  units of time based on covariate information  $X$  available at  $t = 0$

- **Outcome:**

$$1\{T > t\} = \begin{cases} 0, & \text{event occurred before } t \\ 1, & \text{otherwise} \end{cases}$$

$T$  denotes time to event of interest

- **Goal:** Assessment of predictions  $\pi(t|X_i)$  based on a comparison with actually observed outcomes  $T_i$  in a sample of  $n$  individuals ( $i = 1, \dots, n$ )

## Situation (2)

### – Prediction: $\pi(t|X)$

- can be defined for a fixed time  $t$  or for a time range
- should have the properties of a survival probability function
- is ideally externally derived
- but otherwise, can be anything: produced by statistical model building, by machine learning techniques or may constitute expert guesses

## Measures of prediction error (1)

- General loss function approach

$$E(L(T, X, \pi))$$

- Common choices:

$$L(T, X, \pi) = [1\{T > t\} - \pi(t|X)]^2$$

$$L(T, X, \pi) = -[1\{T > t\} \log \pi(t|X) + 1\{T \leq t\} \log(1 - \pi(t|X))]$$

## Measures of prediction error (2)

- Expected quadratic or Brier score

$$E\left( [ 1\{T > t\} - \pi(t|X) ]^2 \right)$$

"Mean Squared Error of Prediction (MSEP)"

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$$E\left( [ 1\{T > t\} - \pi(t|X) ]^2 \right)$$

"Mean Squared Error of Prediction (MSEP)"

- Decomposition

$$\begin{aligned} & [ 1\{T > t\} - \pi(t|X) ]^2 \\ &= [ 1\{T > t\} - S(t|X) ]^2 + [ \pi\{T > t\} - S(t|X) ]^2 \end{aligned}$$

$S(t|X)$  denotes the "true" probability that an individual with covariate  $X$  will be event-free up to  $t$

## Measures of prediction error (3)

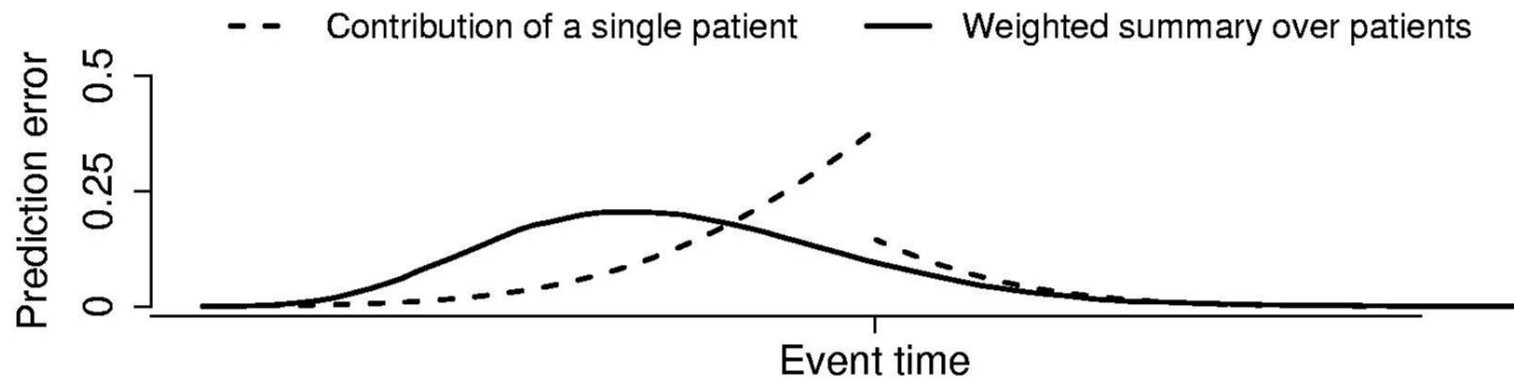
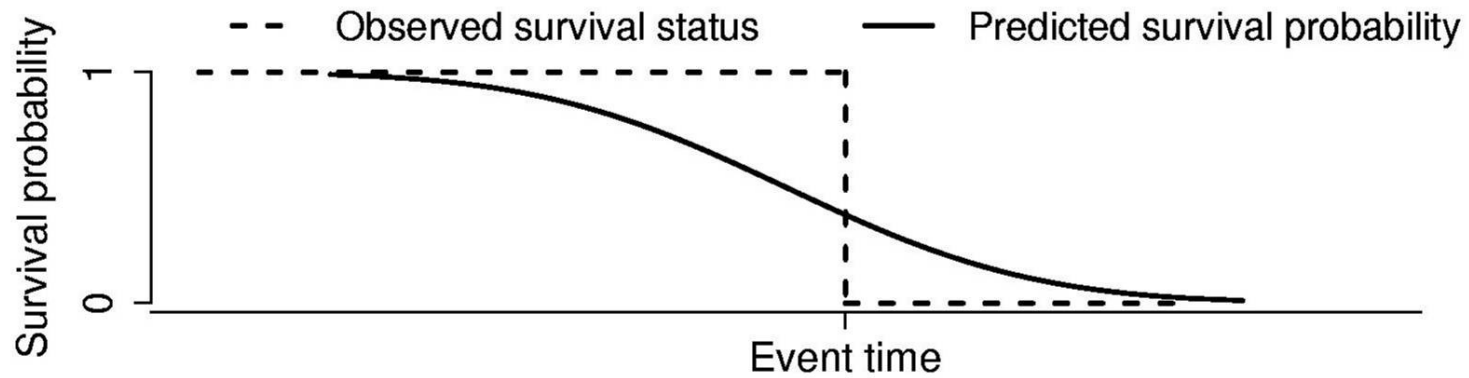
- Empirical quadratic or Brier Score

$$\frac{1}{n} \sum_{i=1}^n [ \mathbf{1}\{\tau_i > t\} - \pi(t|X_i) ]^2$$

"Residual Sum of Squares (RSS)"

- MSE and RSS are time-dependent in survival problems
- Graphical tool: plotting RSS over time





## Measures of prediction error (3)

- Empirical quadratic or Brier Score

$$\frac{1}{n} \sum_{i=1}^n [ \mathbf{1}\{T_i > t\} - \pi(t|X_i) ]^2$$

"Residual Sum of Squares (RSS)"

- Incorporation of censored observations

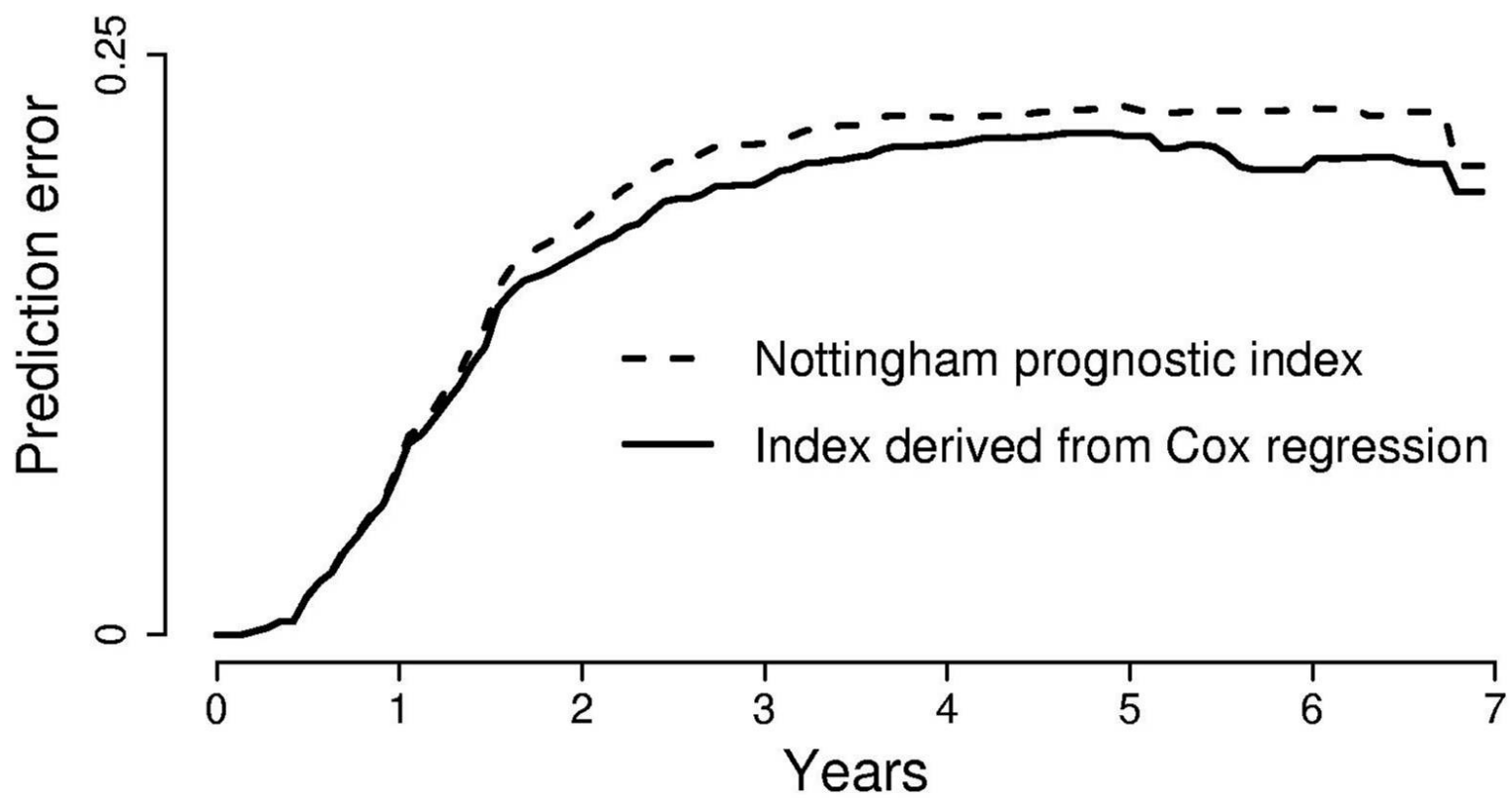
$$\frac{1}{n} \sum_{i=1}^n w_i(t) [ \mathbf{1}\{T_i > t\} - \pi(t|X_i) ]^2$$

"Weighted Residual Sum of Squares (WRSS)"

# Application to prediction of breast cancer survival

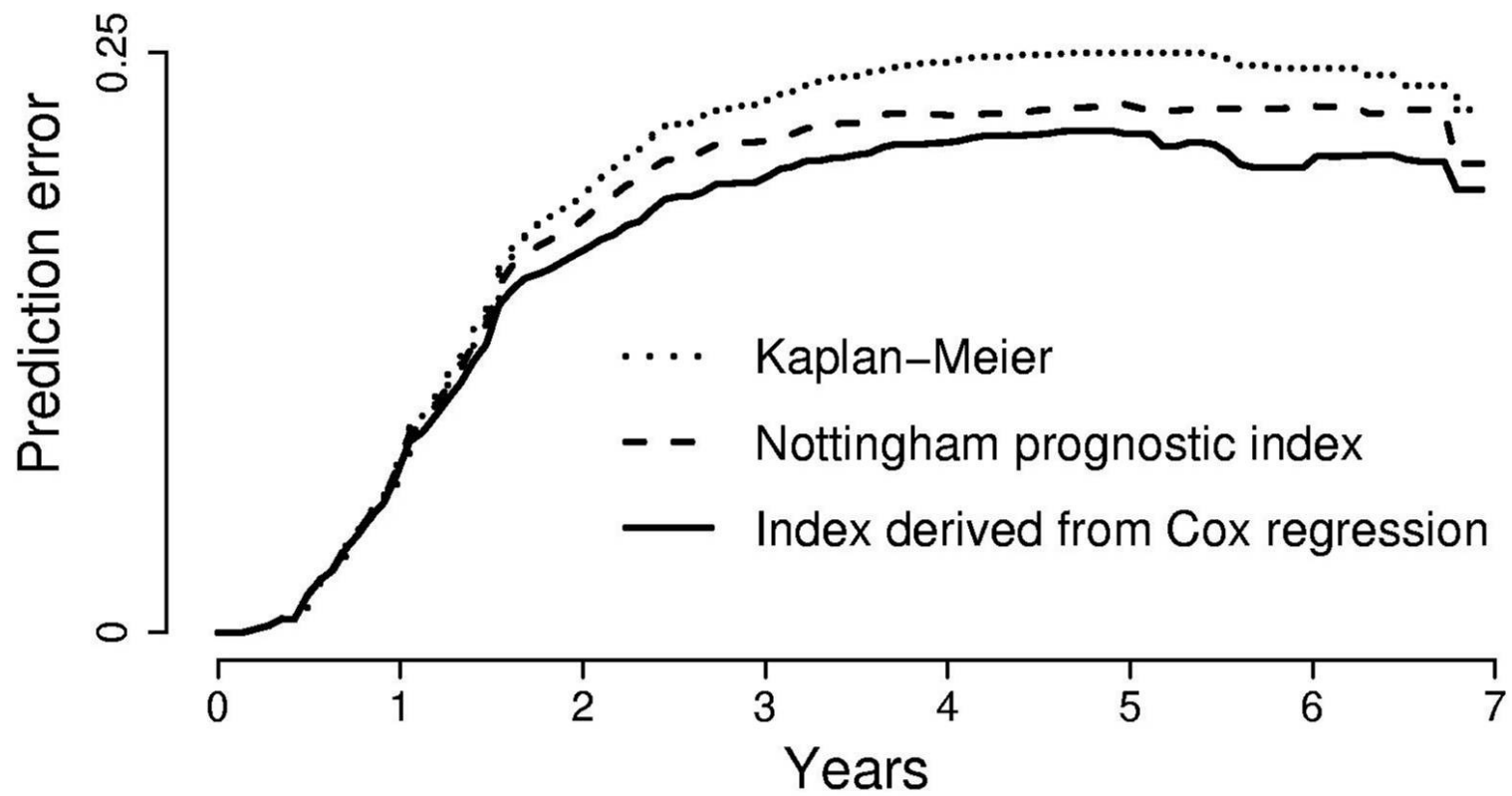
## GBSG-2-study (German Breast Cancer Study Group)

- 686 patients with complete information on prognostic factors
- Two thirds are randomized, otherwise standardized treatment
- Median follow-up 5 years, 299 events for event-free survival
- Prognostic factors considered: age, tumor size, tumor grade, number of positive lymph nodes, progesterone receptor, estrogen receptor
- Predictions for individual patients are derived in terms of conditional event-free probabilities given the covariate combination by means of the Nottingham Prognostic Index and a Cox regression model with all six prognostic factors



## Which benchmark value?

- "Naive" prediction  $\pi(t|X) = 0.5$  for all  $t$  and  $X$  gives a Brier score value of 0.25
- Common prediction  $\pi(t)$  for all individuals ignoring the available covariate information ("pooled Kaplan Meier")

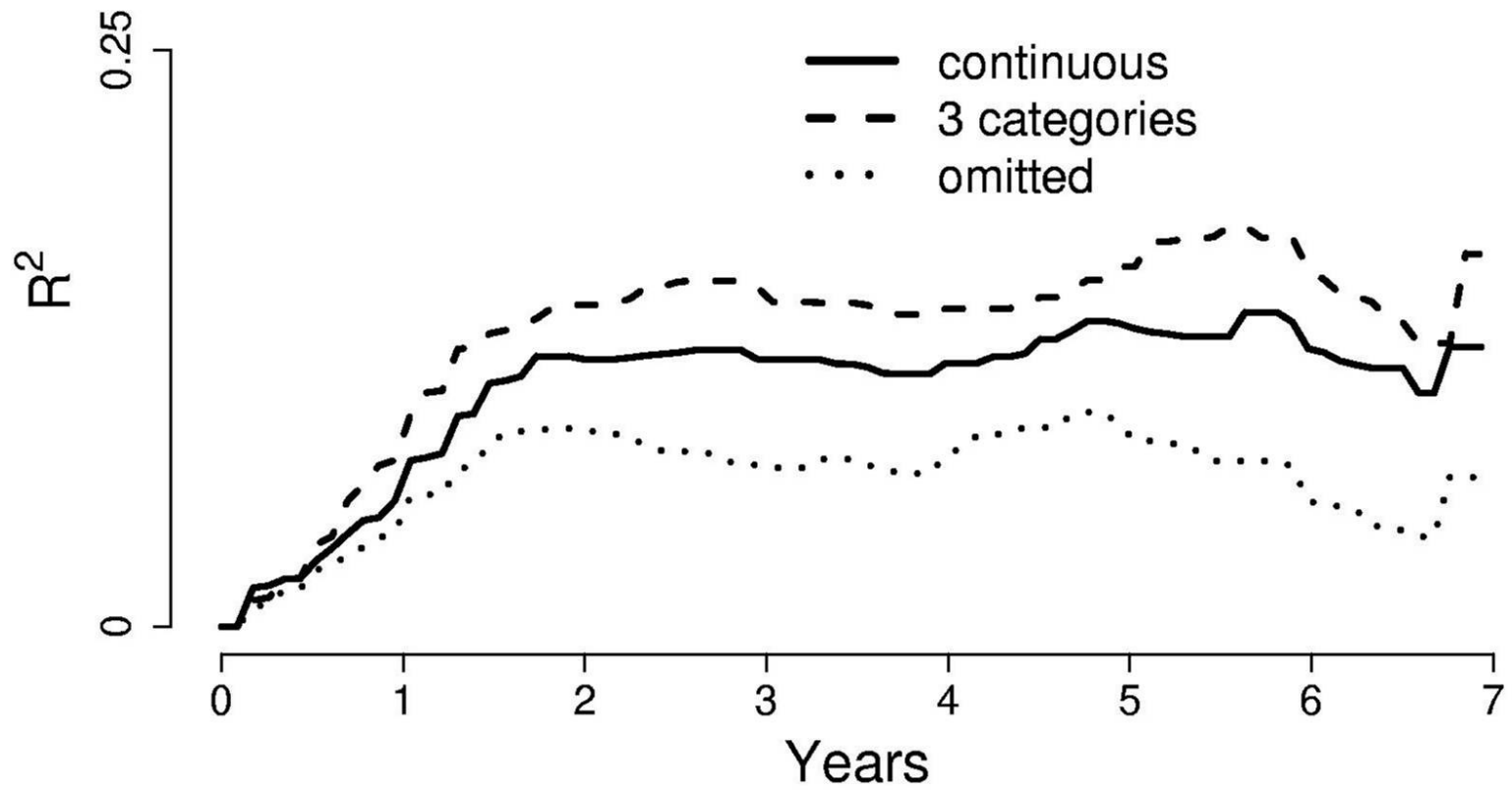


## Which benchmark value?

- "Naive" prediction  $\pi(t|X) = 0.5$  for all  $t$  and  $X$  gives a Brier score of 0.25
- Common prediction  $\pi(t)$  for all individuals ignoring the available covariate information ("pooled Kaplan Meier")
- Calculation of  $R^2$ -measures for checking various aspects of prediction models

$$R^2(t) = 1 - \frac{\text{WRSS}(t, \pi(t|X))}{\text{WRSS}(t, \pi(t))}$$

## Adjusting the number of positive lymph nodes to the Cox model



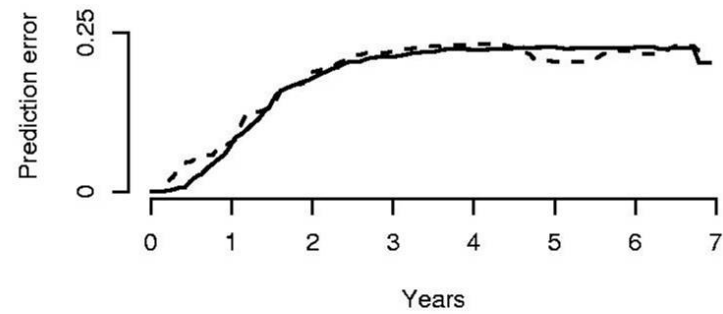


— Same data  
- - - Independent test data

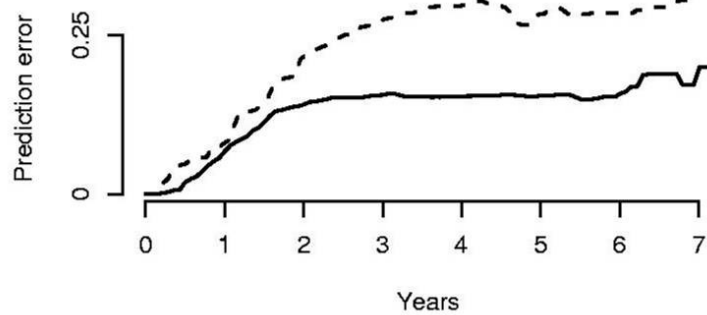
Cox regression model



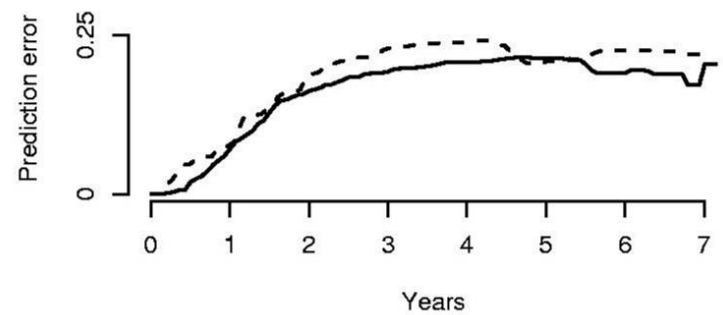
Nottingham prognostic index



Neural net (20 hidden nodes)



Neural net (weight decay)



## General conclusion

### The quadratic or Brier score

- is the mean squared error of prediction (MSEP) when predictions are made in terms of event(-free) probabilities
- allows the assessment of any kind of predictions based on individual covariate values
- can be estimated even in the presence of right censoring by a weighted residual sum of squares in a nonparametric way
- is a valuable tool to detect overfitting
- allows the calculation of  $R^2$ -measures
- can be adapted to the situation of competing risks and dynamic updating of predictions

## Considerations for breast cancer risk prediction

- Prediction:  $\pi(t|X)$

predicted probability that a woman will develop breast cancer up to time  $t$  based on covariate information including age available at  $t = 0$  (entry into program or study; time when prediction is performed)

- Outcome:

$$1\{T \leq t\} = \begin{cases} 0, & \text{development of breast cancer before } t \\ 1, & \text{otherwise} \end{cases}$$

$T$  denotes time from entry into program to development of breast cancer

- Intention: Assessment of predictions for  $t = 5y$  based on aggregated data published by Costantino et al. JNCI 1999; constant prediction ignoring all covariate information is used as benchmark value

**Table 3.** Comparison of the expected cases of total breast cancer (invasive and all *in situ*) predicted from model 1 to the observed cases among white women in the placebo arm of the Breast Cancer Prevention Trial

Age group, y	Predicted 5-year risk, %	No. of women	Observed ( <i>O</i> ) breast cancers	Expected ( <i>E</i> ) breast cancers	<i>E/O</i>	95% confidence intervals
≤49	<2.32	111	1	1.93	1.93	0.35–76.25
	2.32–2.65	499	11	9.60	0.87	0.49–1.75
	2.66–3.28	521	25	12.89	0.52	0.35–0.80
	3.29–4.73	614	17	19.32	1.14	0.71–1.95
	>4.73	587	29	31.42	1.08	0.75–1.62
	Total		2332	83	75.16	0.91
50–59	<2.32	304	8	5.35	0.67	0.34–1.55
	2.32–2.65	468	14	9.80	0.70	0.42–1.28
	2.66–3.28	362	6	8.47	1.41	0.65–3.85
	3.29–4.73	326	13	10.43	0.80	0.47–1.15
	>4.73	347	13	17.69	1.36	0.80–2.56
	Total		1807	54	51.75	0.96
≥60	<2.32	784	21	9.75	0.46	0.30–0.75
	2.32–2.65	232	8	4.77	0.60	0.30–1.38
	2.66–3.28	308	12	7.61	0.63	0.36–1.23
	3.29–4.73	244	9	8.30	0.92	0.49–2.02
	>4.73	262	17	14.01	0.82	0.51–1.41
	Total		1830	67	44.44	0.66
All ages	<2.32	1199	30	17.03	0.57	0.40–0.84
	2.32–2.65	1199	33	24.17	0.73	0.52–1.06
	2.66–3.28	1191	43	28.97	0.67	0.50–0.93
	3.29–4.73	1184	39	38.05	0.98	0.71–1.37
	>4.73	1196	59	63.13	1.07	0.83–1.41
Grand total		5969	204	171.34	0.84	0.73–0.97

Costantino et al., Journal of the National Cancer Institute, Vol. 91, No. 18, September 15, 1999

## Estimated prediction error based on aggregated data

Age group, y	Brier Score		Logarithmic score	
	model 1	const. pred.	model 1	const. pred.
$\leq 49$	0.03428	0.03432	0.15289	0.15367
50 – 59	0.02900	0.02899	0.13446	0.13434
$\geq 60$	0.03519	0.03527	0.15625	0.15702
All ages	0.03296	0.03301	0.14834	0.14897

## Estimated relative risk (RR) for predicted risk quintiles (model 1, all ages)

Predicted 5-year, % risk	No. of women	Observed breast cancer	RR
< 2.32	1199	30	1
2.32 – 2.65	1199	33	1.1
2.66 – 3.28	1191	43	1.436
3.29 – 4.73	1184	39	1.316
> 4.73	1196	59	1.972

## "Diagnostic" properties of predicted risk quintiles (model 1, all ages)

<b>Cutpoint Pred. 5-year risk, %</b>	<b>Sensitivity</b>	<b>Specificity</b>	<b>Pos. pred. value</b>	<b>Neg. pred. value</b>
<b>2.32</b>	<b>0.853</b>	<b>0.203</b>	<b>0.036</b>	<b>0.975</b>
<b>2.66</b>	<b>0.690</b>	<b>0.405</b>	<b>0.039</b>	<b>0.974</b>
<b>3.29</b>	<b>0.480</b>	<b>0.604</b>	<b>0.041</b>	<b>0.971</b>
<b>4.73</b>	<b>0.289</b>	<b>0.803</b>	<b>0.049</b>	<b>0.970</b>

