

Producing up-to-date survival predictions from prognostic models

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26th September 2019



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PhD Project

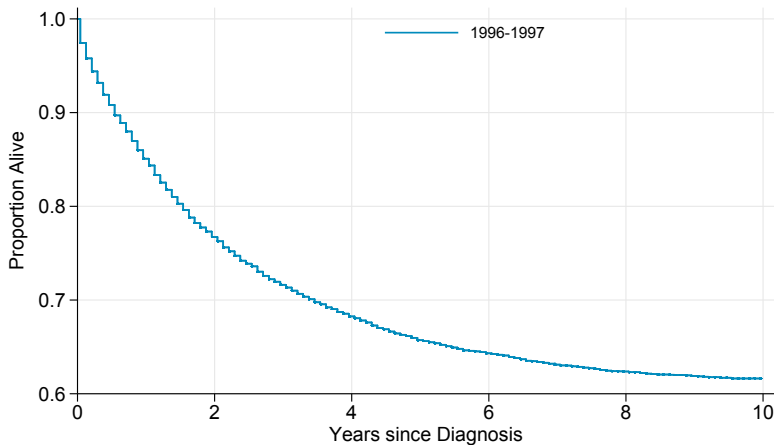
Prognostic Models

- Prognostic models can be used to inform patients and aid treatment decisions
- Often built using data collected over a long time period
- Improvements in survival may lead to out-dated survival predictions

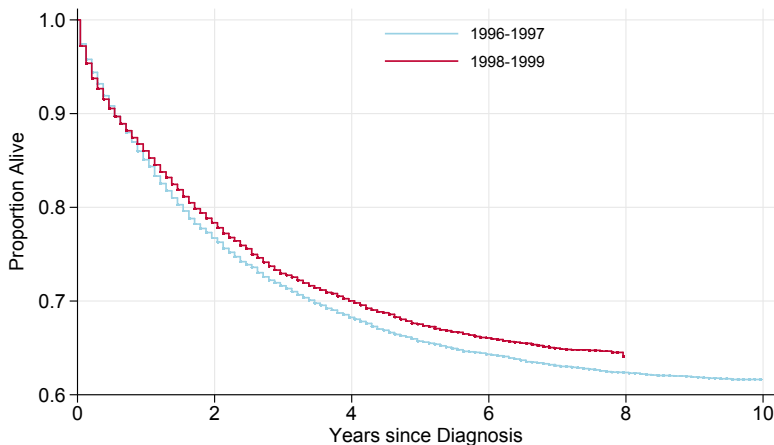
Methods

- Developed temporal recalibration which combines period analysis with recalibration techniques
- Alternative approach involving modelling calendar time

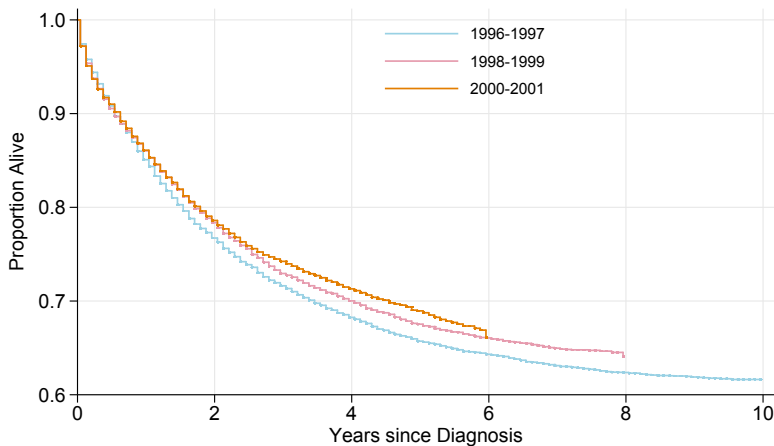
Assessing Improvements in Survival



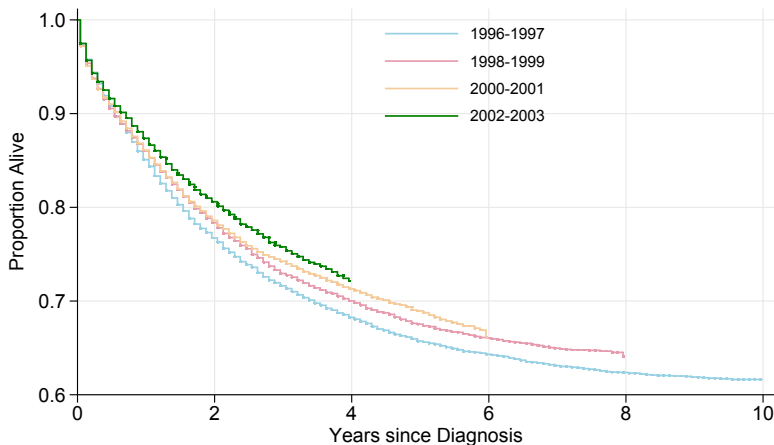
Assessing Improvements in Survival



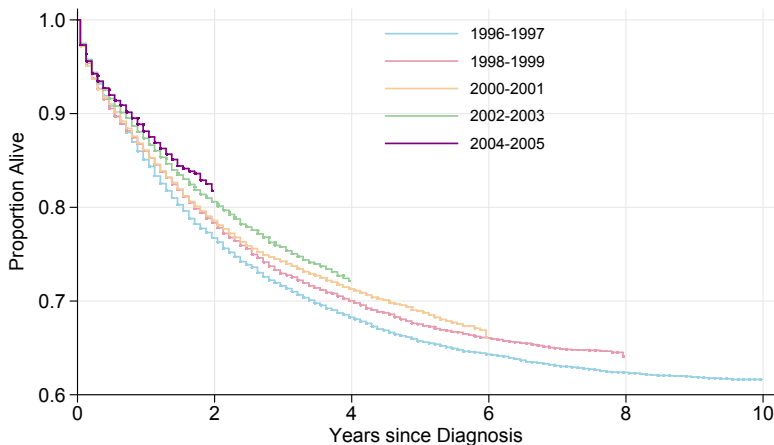
Assessing Improvements in Survival



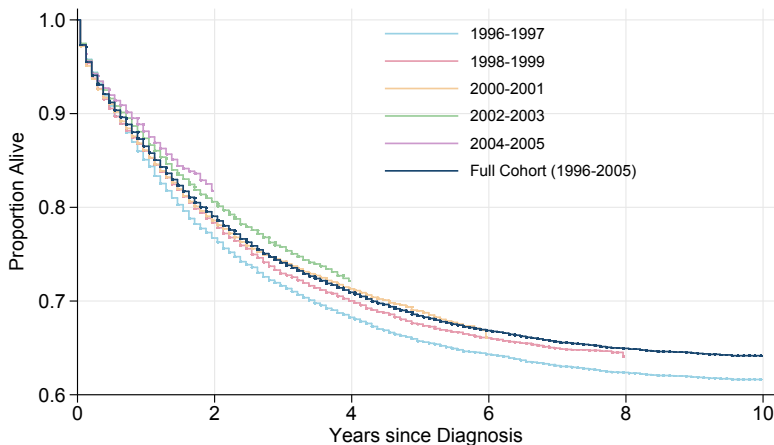
Assessing Improvements in Survival



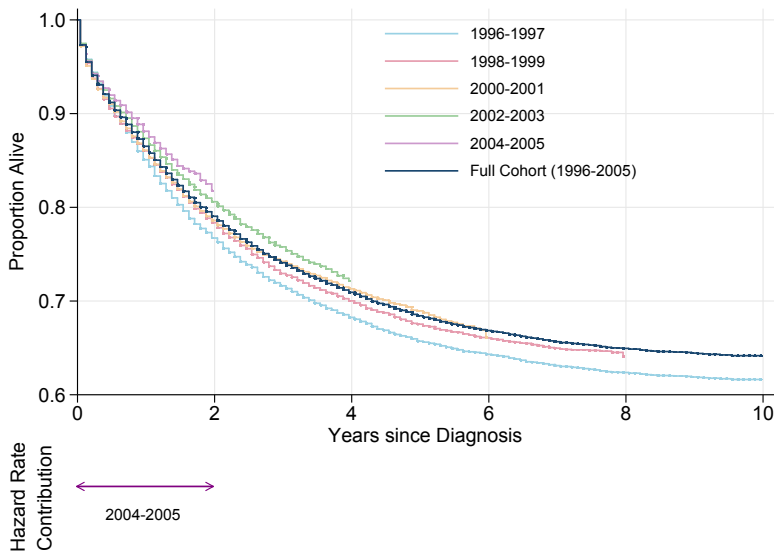
Assessing Improvements in Survival



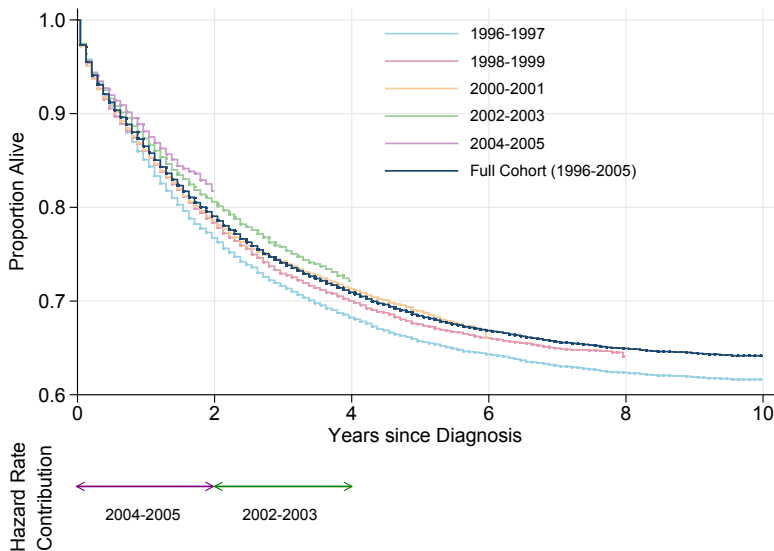
Assessing Improvements in Survival



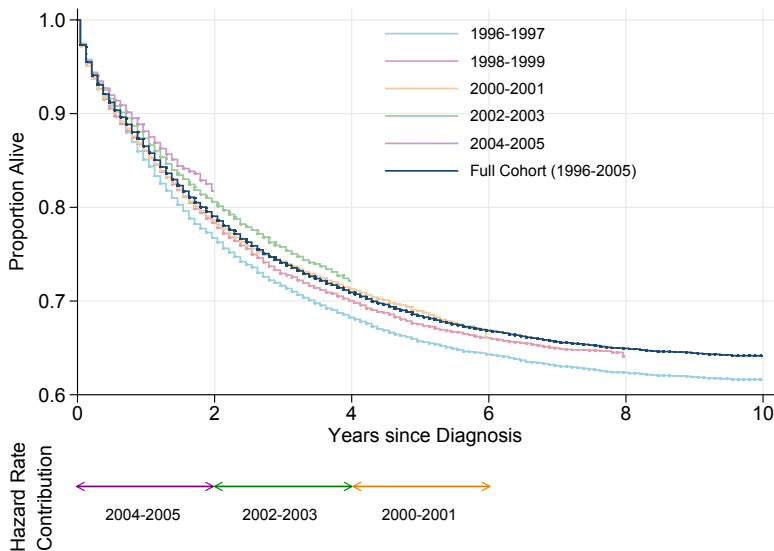
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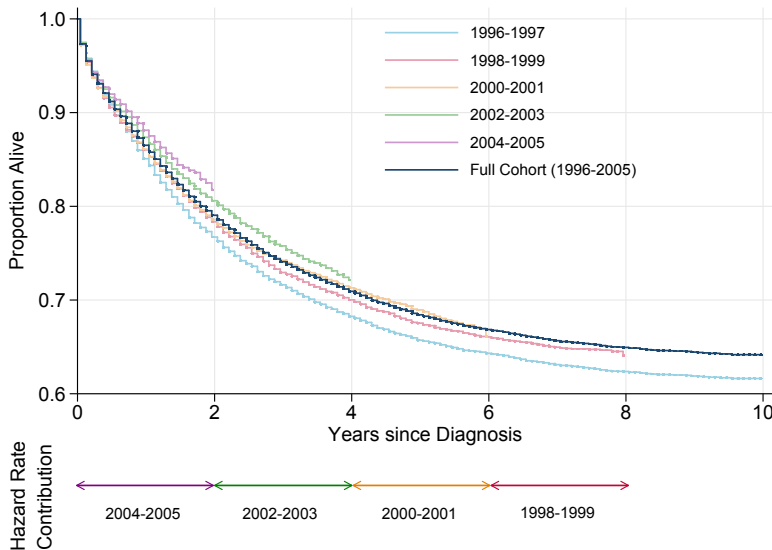
Assessing Improvements in Survival



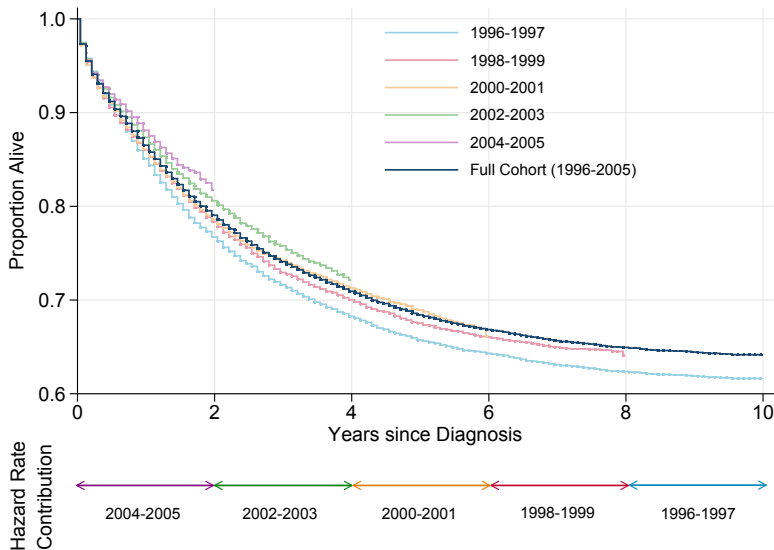
Assessing Improvements in Survival



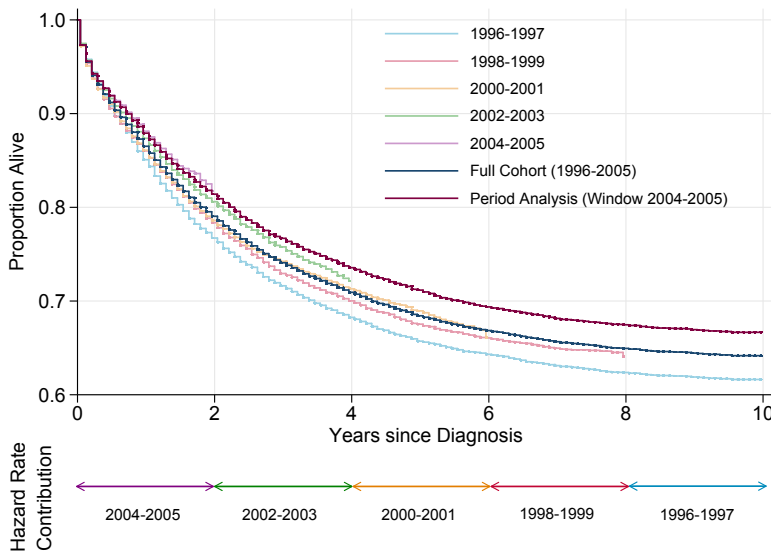
Assessing Improvements in Survival



Assessing Improvements in Survival



Assessing Improvements in Survival



Full Cohort Analysis

Participant	Year of Diagnosis & Follow-Up									
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005
A	1	2	3	4	5	6	7	8	9	10
B		1	2	3	4	5				
C						1	2	3	4	5
D									1	2

Full Cohort Analysis

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- Standard method used to develop prognostic models
- All 4 participants would be included

Period Analysis

Participant	Year of Diagnosis & Follow-Up									
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005
A	1	2	3	4	5	6	7	8	9	10
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- Define a recent time window (2004-2005)

H. Brenner and O. Gefeller, "An alternative approach to monitoring cancer patient survival," *Cancer*, vol. 78, no. 9, pp. 2004–2010, 1996.

Period Analysis

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- Define a recent time window (2004-2005)
- Only include those who are alive at some point in the window

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- Define a recent time window (2004-2005)
- Only include those who are alive at some point in the window
- Only analyse the risk-time and events within the window

H. Brenner and O. Gefeller, "An alternative approach to monitoring cancer patient survival," *Cancer*, vol. 78, no. 9, pp. 2004–2010, 1996.

Period Analysis

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Advantages

- More up-to-date survival estimates, people diagnosed many years ago only contribute to long-term hazard rates

H. Brenner and O. Gefeller, "An alternative approach to monitoring cancer patient survival," *Cancer*, vol. 78, no. 9, pp. 2004–2010, 1996.

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Advantages

- More up-to-date survival estimates, people diagnosed many years ago only contribute to long-term hazard rates

Disadvantages

- Reduces sample size and number of events
- Larger standard errors

H. Brenner and O. Gefeller, "An alternative approach to monitoring cancer patient survival," *Cancer*, vol. 78, no. 9, pp. 2004–2010, 1996.

Temporal Recalibration

Participant	Year of Diagnosis & Follow-Up									
	1996	1997	1998	1999	2000	2001	2002	2003	2004	2005
A	1	2	3	4	5	6	7	8	9	10
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Method

- Fit a full cohort model

Temporal Recalibration

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Method

- Fit a full cohort model
- Use a period analysis sample to recalibrate the model

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Method

- Fit a full cohort model
- Use a period analysis sample to recalibrate the model
- The predictor effects are constrained to be the same (i.e hazard ratios for age, sex, stage are the same)
- The baseline hazard function is allowed to vary which can capture any improvements in survival

Summary of Methods

Type of Analysis	Predictor Effects	Baseline
Full Cohort	All	All
Period Analysis	Recent	Recent
Temporal Recalibration	All	Recent

Models

Type of Analysis	Year of Diagnosis & Follow-Up					Follow-Up Only
	1996-2002	2003	2004	2005	2006	2007-2015
Full Cohort						
Temporal Recalibration						
Period Analysis						
Validation						

Surveillance, Epidemiology, and End Results (SEER) Program (www.seer.cancer.gov) Research Data (1973-2015), National Cancer Institute, DCCPS, Surveillance Research Program, released April 2018, based on the November 2017 submission

Models

Type of Analysis	Year of Diagnosis & Follow-Up					Follow-Up Only
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Period Analysis						
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Temporal Recalibration						
Period Analysis						
Validation						

- US colon cancer registry data from the Surveillance, Epidemiology, and End Results (SEER) Program
- Cause-specific survival models

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Models

Type of Analysis	Year of Diagnosis & Follow-Up					Follow-Up Only
	1996-2002	2003	2004	2005	2006	2007-2015
Full Cohort						
Temporal Recalibration						
Period Analysis						
Validation						

- US colon cancer registry data from the Surveillance, Epidemiology, and End Results (SEER) Program
- Cause-specific survival models
- No variable selection: Age (modelled with splines), sex, race, stage and grade

Surveillance, Epidemiology, and End Results (SEER) Program (www.seer.cancer.gov) Research Data (1973-2015), National Cancer Institute, DCCPS, Surveillance Research Program, released April 2018, based on the November 2017 submission

Survival Models

Cox Proportional Hazards Models

$$h(t; x_i) = h_0(t)e^{\beta x_i}$$

- Post-estimation of baseline using restricted cubic splines

D. R. Cox, "Regression models and life-tables," Journal of the Royal Statistical Society. Series B (Methodological), vol. 34, no. 2, pp. 187–220, 1972.

P. Royston and M. K. B. Parmar, "Flexible parametric proportional-hazards and proportional-odds models for censored survival data, with application to prognostic modelling and estimation of treatment effects," Statistics in Medicine, vol. 21, no. 15, pp. 2175–2197, 2002.

Survival Models

Cox Proportional Hazards Models

$$h(t; x_i) = h_0(t)e^{\beta x_i}$$

- Post-estimation of baseline using restricted cubic splines

Flexible Parametric Survival Models

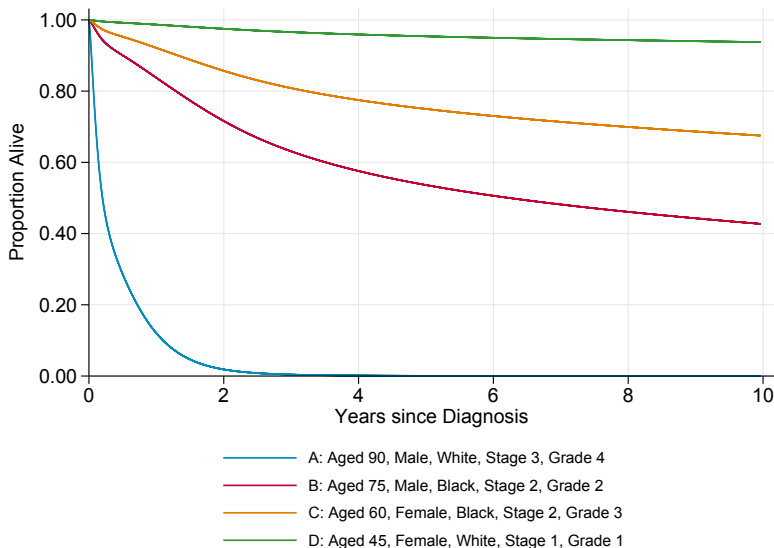
$$\ln[H(t|x_i)] = \gamma_0 + \gamma_1 z_{1i} + \gamma_2 z_{2i} + \gamma_3 z_{3i} + \dots + \beta x_i$$

- z_i = derived variables for the restricted cubic splines

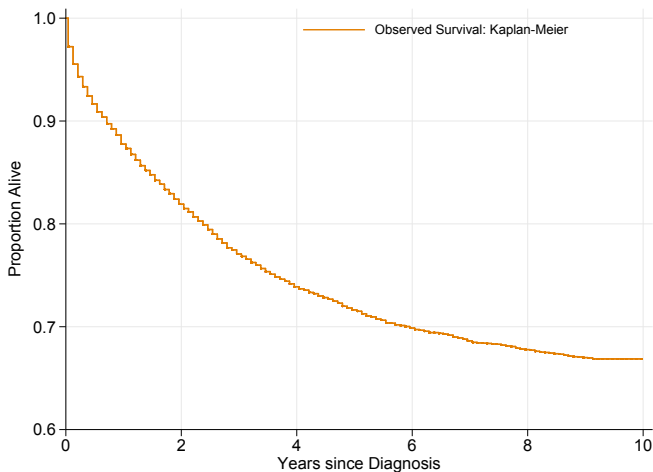
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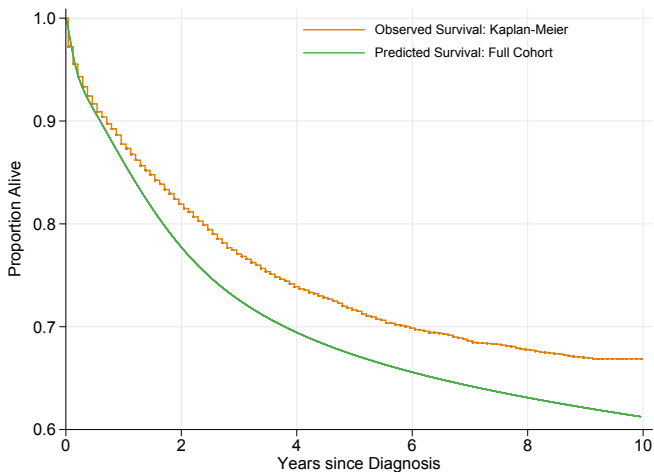
Individual Survival Predictions



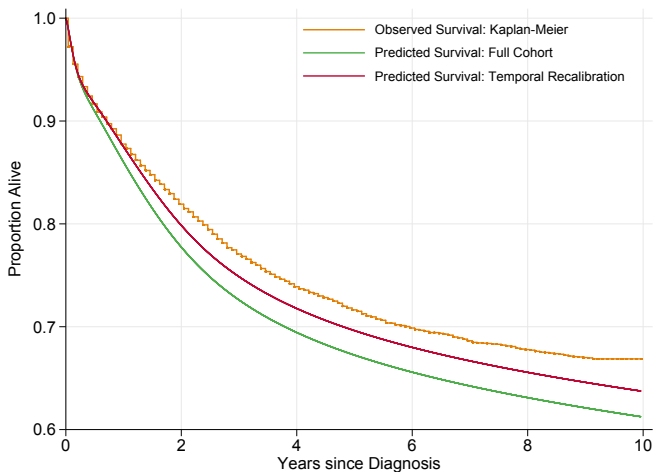
Model Validation: Individuals diagnosed in 2006



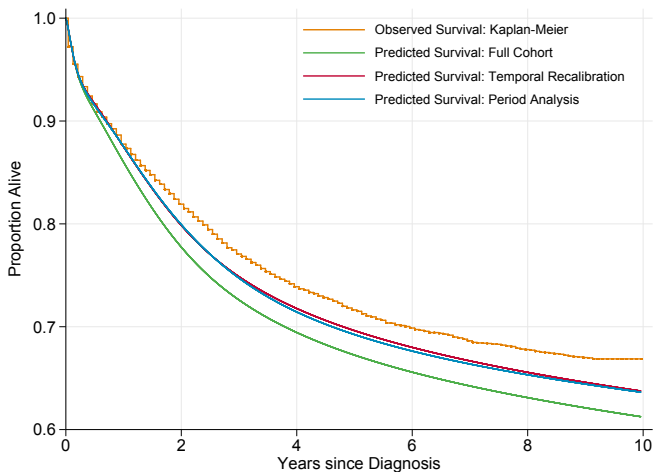
Model Validation: Individuals diagnosed in 2006



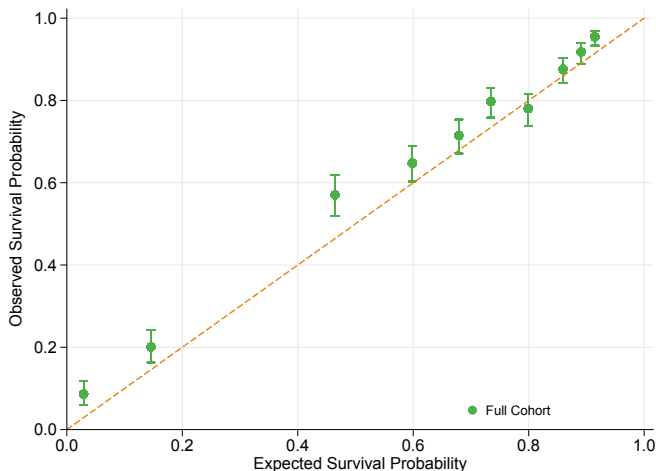
Model Validation: Individuals diagnosed in 2006



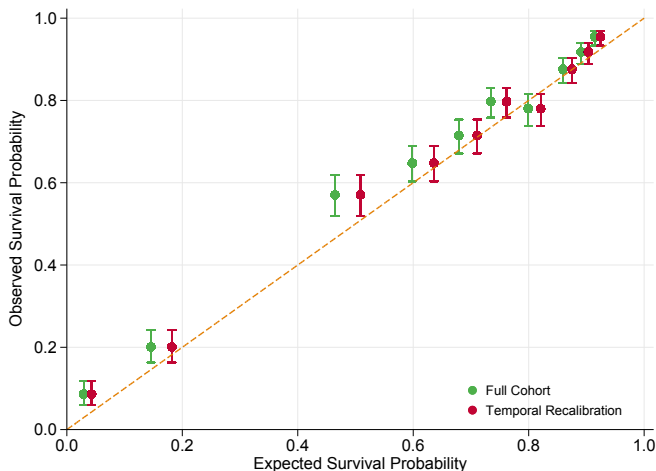
Model Validation: Individuals diagnosed in 2006



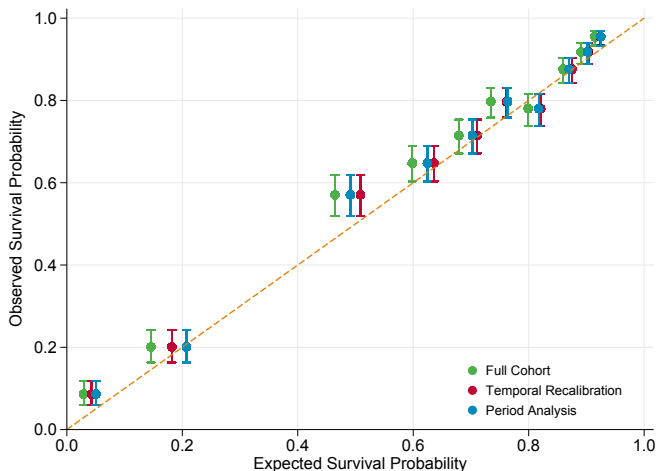
Calibration of Models



Calibration of Models



Calibration of Models



Period Analysis Window

Type of Analysis	Sample Size	Number of Events
Full Cohort	48,861	12,040
2 Year Window	33,197	2,900
1 Year Window	30,427	1,500
6 Month Window	28,896	774
3 Month Window	28,099	383

- Width of window could be determined using sample size criteria by Riley et al.

R. D. Riley et al. "Minimum sample size for developing a multivariable prediction model: PART II - binary and time-to-event outcomes," Statistics in Medicine, 2018.

Hazard Ratio

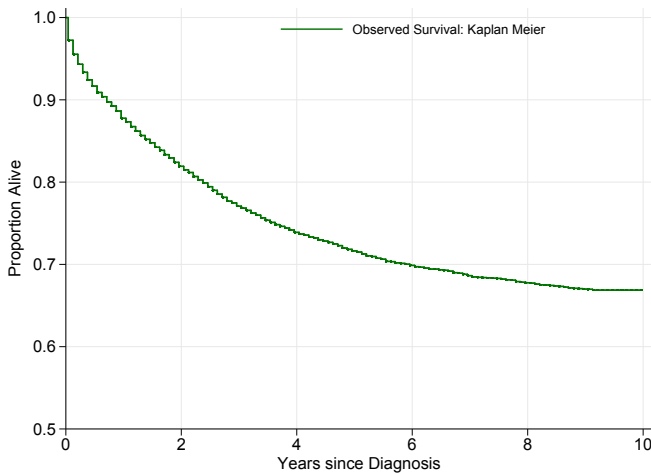
Type of Analysis	log HR Female	Standard Error
Full Cohort	-0.05	0.018
2 Year Window	-0.10	0.038
1 Year Window	-0.13	0.053
6 Month Window	-0.07	0.073
3 Month Window	-0.06	0.104

Updating Prognostic Models

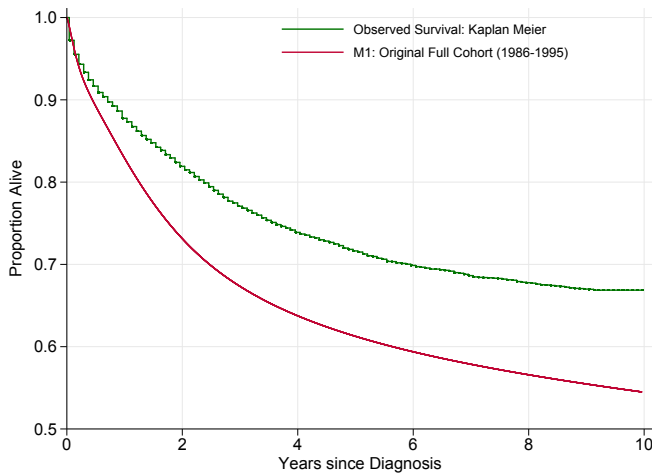
	Year of Diagnosis & Follow-Up			Follow-Up Only
	1986-1995	1996-2005	2006	2007-2015
Development				
Updating				
Validation				

- M1: Original full cohort model (1986-1995)
- M2: Full cohort model with all available data (1986-2005)
- M3: Full cohort model with most recent data (1996-2005)
- M4: Temporal recalibration of M1 (period window 2004-2005)
- M5: Temporal recalibration of M3 (period window 2004-2005)

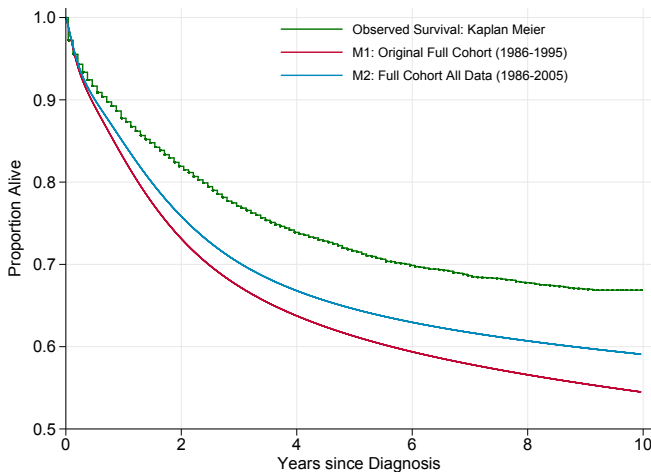
Observed Survival: Individuals diagnosed in 2006



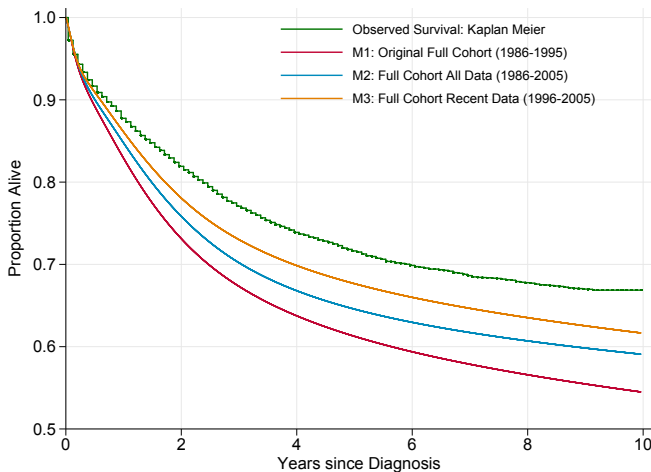
M1: Original full cohort model (1986-1995)



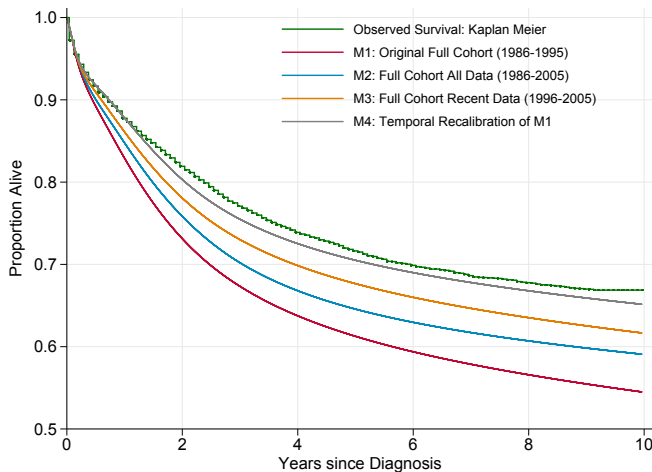
M2: Full cohort model with all available data (1986-2005)



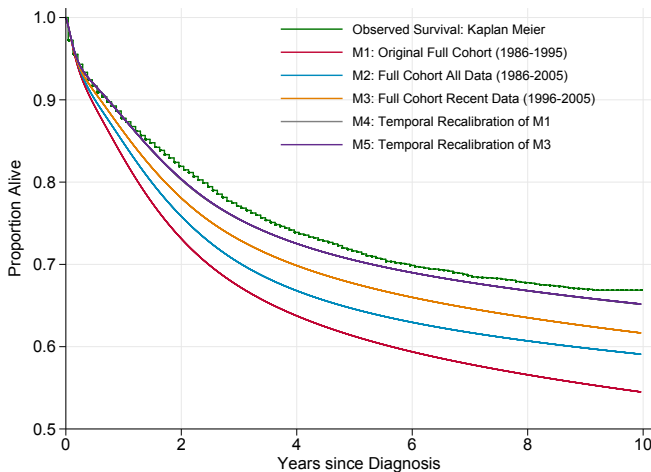
M3: Full cohort model with most recent data (1996-2005)



M4: Temporal recalibration of M1



M5: Temporal recalibration of M3



Summary

- Full cohort models often underestimate survival if there have been recent improvements in survival
- Period analysis uses a subset of data to create more up-to-date survival predictions
- Temporal recalibration also produces more up-to-date survival predictions but all the data is used to estimate the predictor effects

Discussion: Modelling calendar time (year of diagnosis)

Functional form

- Linear, categorical, restricted cubic splines
- Incorporate month of diagnosis for a smoother function
- Time dependent effects
- Interactions between predictor effects and year of diagnosis

Survival predictions e.g. new patient diagnosed in 2007 and the model is fitted using data from 1996-2005

- Use the most recent year (2005) included in the model
- Extrapolate the trend to 2007
- Update the prognostic model every year