

Prognosis of survival for breast cancer patients

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Outline

- Introduce the data and outcomes requested by the medics.
- Work through the stages of the analyses.

Imputation

Fractional Polynomials

Producing a simple score

Fitting Splines

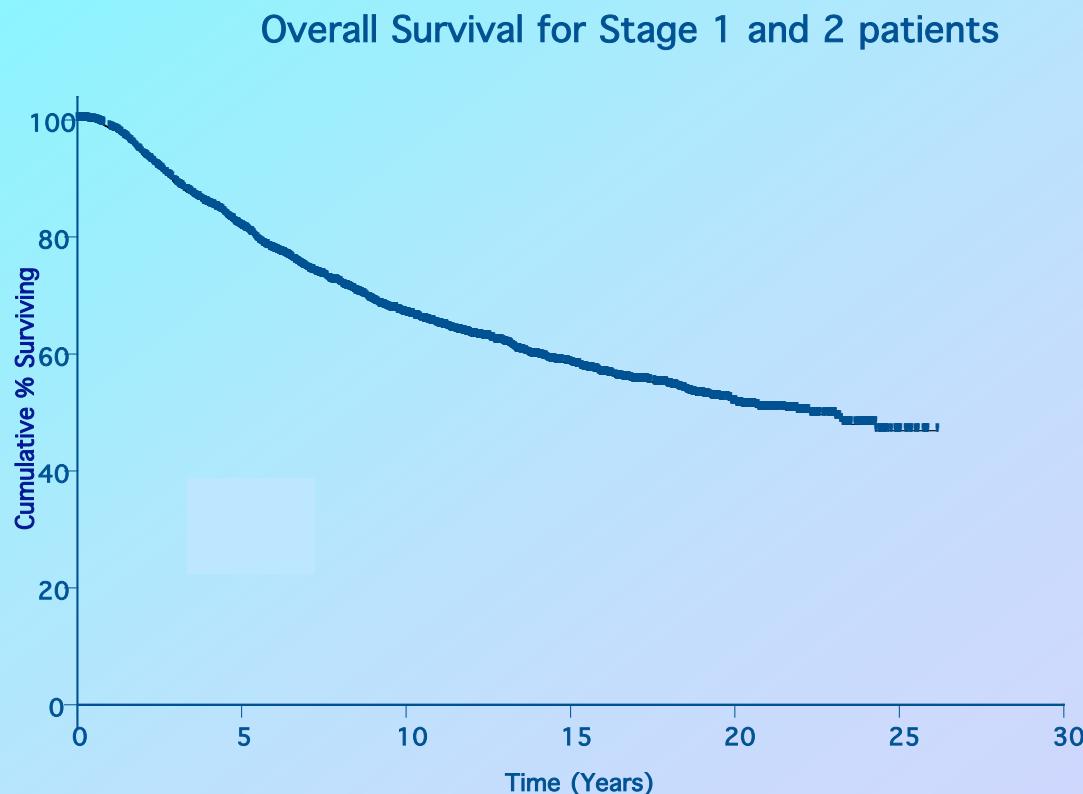
- Illustrate the menu call developed for everyday use.

The Problem

- Over 8000 patients with invasive Breast Cancer with 3083 operable stage 1 and 2 disease.
- Time of relapse-free survival
- Time to a distant relapse
- Overall survival
 - Death from breast cancer
 - All-cause mortality

The Problem

- **Kaplan-Meier Curves**



The Problem

Factors affecting outcome

- **Histological type (x_4)**
Ductal (Grade I, II & III), Lobular, others (Mucoid, Medullary, Tubular)
- **Tumour Size (x_3)**
(\square 2.0 cms, 2.0 - 5.0 cms, >5.0 cms)
- **Axilla Nodal status (x_5)**
(All negative, 1-3, 4 - 9, ≥ 10 positive)
- **Age (x_1)**
(Under 40, 41-50, 51-60, 61-70, Over 70)

The Problem Factors affecting outcome

- Postmenopausal status (x_2)
- Type of surgery (x_{11})
- Progesterone receptor status (x_{6a})
- Oestrogen receptor status (x_{7a})
- Adjuvant treatment
Chemotherapy (x_8)
- Hormonal therapy (x_9)
- Radiotherapy (x_{10})

The Task

- Given the information on the prognostic factors for a new patient
- To provide estimates of the probabilities of being relapse-free or surviving after 5, 10, 15, 20 years.

Prognostic Factors

	Code	Mean	% missing
Age (years)	x ₁	55	0
Postmenopausal status	x ₂	0.55	2.0
Clinical tumour size (mm)	x ₃	28	1.1
Histology Ductal Grade 2	x _{4a}	0.39	0
Histology Ductal Grade 3	x _{4b}	0.34	0
Histology Lobular	x _{4x}	0.13	0
Histology Other	x _{4y}	0.03	0
No of positive lymph nodes	x ₅	2.8	0
Progesterone receptor positive	x _{6a}	0.52	13.7
Oestrogen receptor positive	x _{7a}	0.70	11.5
Chemotherapy (any type)	x ₈	0.20	0
Hormonal therapy	x ₉	0.35	0
Radiotherapy	x ₁₀	0.42	0
Impalable tumour	x _{3x}	0.07	1.1

Imputation

van Buuren S., H. C. Boshuizen and D. L. Knook. 1999. Multiple imputation of missing blood pressure covariates in survival analysis. Statistics in Medicine 18:681-694.

mvisamp imputes missing values using "switching regression", an iterative multivariable regression technique.

```
mvisamp varlist [if exp] [in range] [weight], gen(name)
[ cc(ccvarlist) cmd(cmdlist) seed(#) ]
```

Imputation

Impute x6a and x7a on complete variables and transformed year and survival time

```
mvisamp x6a x7a  x1 x4a x4b x4x x4y x5 yr_1 yr_2  
yr_3 Int _d, gen(z1) seed(101)
```

Impute x3 on these variables and completed x6a, x7a

```
mvisamp x3  z1_1 z1_2 x1 x4a x4b x4x x4y x5 yr_1  
yr_2 yr_3 Int _d if x3>0,  gen(z2) seed(103)
```

Cox's Proportional Hazards model

Fractional Polynomials

- For the continuous variables, the Fractional Polynomials, based on X^{-2} , X^{-1} , $X^{-0.5}$, $X^{0.5}$, X^2 , X^3 , and $\log_e(X)$ are compared with the linear polynomial to determine if one or more of these polynomials provides a better fit.
- All other factors are (0,1) variables, including the interaction of Hormone Therapy with ER status.

Cox's Proportional Hazards model Fractional Polynomials

mfracpol selects the fractional polynomial (FP) model which best predicts the outcome variable, yvar, from the RHS variables, xvarlist. After execution, mfracpol leaves variables in the data named Ivar1_1, Ivar2_2, ...Ivar2_1, Ivar2_2 etc. The new variables contain the best-fitting fractional polynomial powers of var1, var2, .

```
mfracpol regression_cmd yvar xvarlist [in range] [if exp]  
[weight]      [, alpha(alpha_list) df(df_list)  
select(select_list)    regression_cmd_options ]
```

Cox's Proportional Hazards model Fractional Polynomials

* Force in ER and hormon, since know ER x hormon is significant.

```
mfracpol cox _t x1 ix2 ix3 ix3x x4a x4b x4x x4y x5 ix6a ix7a  
x7ah hormon chemo rt yr, dead(_d) sel(.05, ix7a hormon  
x7ah:0.9999) df(ix3 x5:2)
```

Cox's Proportional Hazards model

Fractional Polynomials

-> gen double lx1__1 = X^-2-.0327 if e(sample)

(where: $X = x1/10$)

-> gen double lix3__1 = ln(X)+1.243 if e(sample)

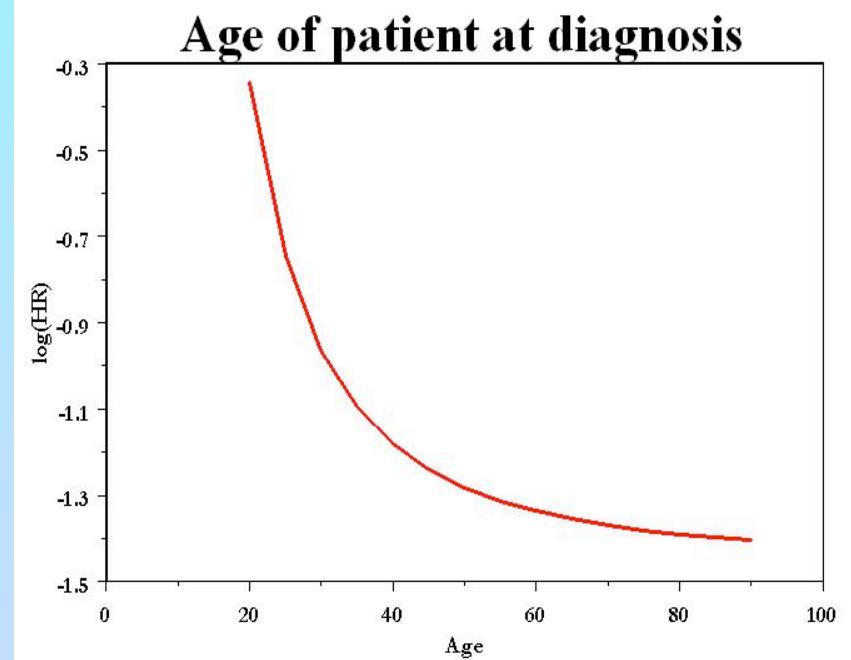
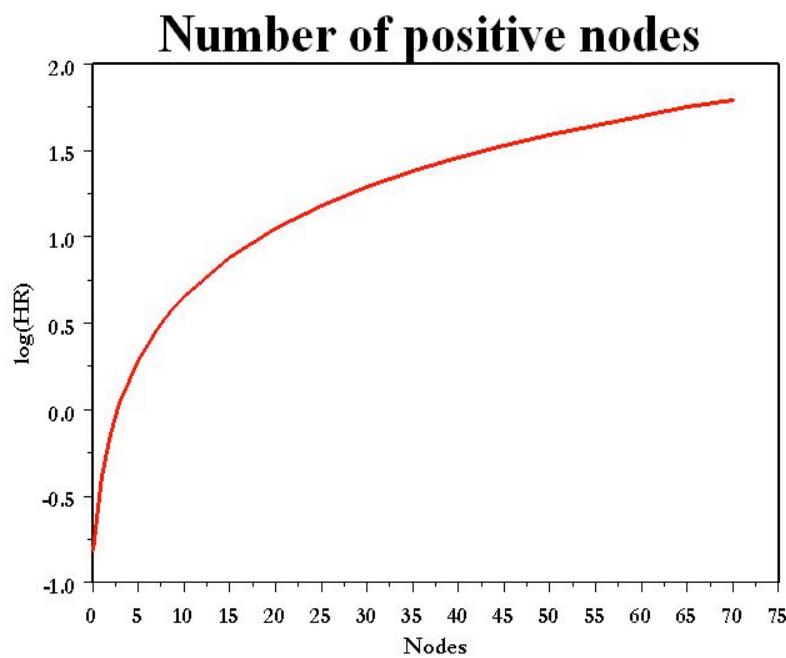
(where: $X = (ix3+1)/100$)

-> gen double lx5__1 = ln(X)+.974 if e(sample)

(where: $X = (x5+1)/10$)

Cox's Proportional Hazards model

Fractional Polynomials



Cox's Proportional Hazards model

Fractional Polynomials

Final multivariable fractional polynomial model for _t							
Variable	-----Initial-----			-----Final-----			
	df	Select	Alpha	Status	df	Powers	
x1	4	0.0500	0.0500	in	2	-2	
ix2	1	0.0500	0.0500	out	0		
ix3	2	0.0500	0.0500	in	2	0	
ix3x	1	0.0500	0.0500	in	1	1	
x4a	1	0.0500	0.0500	in	1	1	
x4b	1	0.0500	0.0500	in	1	1	
x4x	1	0.0500	0.0500	in	1	1	
x4y	1	0.0500	0.0500	out	0		
x5	2	0.0500	0.0500	in	2	0	
ix6a	1	0.0500	0.0500	in	1	1	
ix7a	1	0.9999	0.0500	in	1	1	
x7ah	1	0.9999	0.0500	in	1	1	
hormon	1	0.9999	0.0500	in	1	1	
chemo	1	0.0500	0.0500	in	1	1	
rt	1	0.0500	0.0500	out	0		
yr	4	0.0500	0.0500	out	0		

Cox's Proportional Hazards model

Fractional Polynomials

	<u>t</u>					
	<u>d</u>	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
Ix1_1	4.707014	1.439364	3.27	0.001	1.885911	7.528116
Iix3_1	.5143961	.0655541	7.85	0.000	.3859123	.6428798
ix3x	1.320571	.2790599	4.73	0.000	.7736239	1.867518
x4a	1.122357	.1557526	7.21	0.000	.8170873	1.427626
x4b	1.420355	.1588664	8.94	0.000	1.108983	1.731728
x4x	1.028325	.1704158	6.03	0.000	.6943161	1.362334
Ix5_1	.6070695	.0299867	20.24	0.000	.5482966	.6658423
ix6a	-.1314727	.066974	-1.96	0.050	-.2627393	-.000206
ix7a	.1902246	.0781319	2.43	0.015	.037089	.3433603
x7ah	-.708684	.1504531	-4.71	0.000	-1.003567	-.4138013
hormon	-.0381773	.1254731	-0.30	0.761	-.2841	.2077454
chemo	-.3198936	.0782234	-4.09	0.000	-.4732086	-.1665787

Deviance:16524.958.

Cox's Proportional Hazards model

Fractional Polynomials

- . global model `e(fp_fvl) ‘
- . stcox \$model, nolog sca(sca*) sch(sch*)
- . stphtest, rank detail

Test of proportional hazards assumption					
		rho	chi2	df	Prob>chi2
Ix1_1		-0.08739	10.11	1	0.0015
Ix3_1_1		0.03250	1.23	1	0.2666
x4a		-0.06038	4.17	1	0.0412
x4b		-0.11912	16.27	1	0.0001
x4x		-0.01873	0.40	1	0.5268
Ix5_1		-0.14572	23.72	1	0.0000
ix6a		0.06161	4.59	1	0.0321
ix7a		0.08467	8.35	1	0.0039
x7ah		0.03161	1.14	1	0.2847
hormon		-0.01300	0.19	1	0.6609
chemo		0.11522	16.23	1	0.0001
global test			145.80	11	0.0000

Simple scoring system

- A scoring system was developed so that a unit increase in the score was approximately an increase of 0.1 in the $\log_e(\text{HR})$.
- For the (0,1) binary variables the scores were taken to the nearest 0.1 of the $\log_e(\text{HR})$, but for the continuous variables the cut-off points were selected to have an increment of approximately 0.2 in the $\log_e(\text{HR})$.

Simple scoring system

Node status	Score	Tumour size	Score
0	0	$\leq 11\text{mm}$	0
1	4	12 – 19	2
2	6	20 – 27	4
3 – 4	8	28 – 38	6
5 – 6	10	≥ 39	8
7 – 9	12		
10 – 14	14		
15 – 20	16	ER ⁺ and HT	0
21 – 29	18	ER ⁺ and no HT	8
30 – 42	20	ER ⁻	7
43 – 52	22		
≥ 53	24		

Simple scoring system

Fitting splines

**stpm [varlist] [if exp] [in range] , model_complexity
scale(hazardlnormalodds) stratify(strat_varlist)**

stpm fits spline-based distributional models to right- or interval-censored survival-time data.

varlist is a set of covariates.

model_complexity given by defining the number of degrees of freedom to use or by defining the knot positions

scale specifies the scale of model

stratify

**stpm guysrisk, st(guysrisk) sc(h) knots(0.643 1.614)
nolog**

Simple scoring system

Score	1	2	5	10	20
10	0.989	0.973	0.942	0.914	0.881
14	0.984	0.960	0.914	0.873	0.827
18	0.976	0.940	0.873	0.815	0.751
22	0.964	0.911	0.815	0.735	0.650
26	0.947	0.869	0.735	0.628	0.522
30	0.921	0.809	0.629	0.497	0.376
34	0.884	0.727	0.498	0.348	0.229
38	0.830	0.619	0.350	0.204	0.109
42	0.755	0.485	0.205	0.092	0.035
46	0.656	0.337	0.092	0.027	0.006

Menu system

