

Univariable survival analysis

S8

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Survival analysis

- Cohort studies (incl. clinical trials & patient series)
- Survival time outcomes (also called failure time, time-to-event), censoring
- Kaplan-Meier survival estimate
- Logrank test (2-sample, 2-sample stratified, >2-sample, trend)
- Cox regression (next session)
- Proportional hazards assumption (next session)

TUMOR VOLUME AS PROGNOSTIC FACTOR IN CHEMORADIATION FOR ADVANCED HEAD AND NECK CANCER

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- 360 patients (pts) w/ advanced head & neck squamous cell carcinoma (oral cavity, oropharynx, hypopharynx)
- Diagnosed 1997–2006 in 6 hospitals
- Chemoradiation for functionally/anatomically unresectable disease
- Primary tumor volume measured by pretreatment MRI or CT
- 72% male, mean age 56 yrs (range, 25–85)
- Followed for local recurrence (median, 20 months)

Data set tumorvolume.sav

- Includes several transformations of original variables based on code in `transformations_syntax.txt`
- Codebook `labels.doc` describes variable labels

Quantitative aspects of a cohort study

Exposed/Treated	Subjects	Events	Person-time
Yes	N_1	D_1	PYR_1
No	N_0	D_0	PYR_0
Total	N	D	PYR

- Cumulative incidence: proportion of subjects with event during given interval D/N
- Incidence rate (incidence density): #events per person-time (rate) D/PYR
- Incidence rate ratio: ratio of incidence rates by exposure/treatment (approximates relative risk for rare diseases)

$$\frac{D_1/PYR_1}{D_0/PYR_0}$$

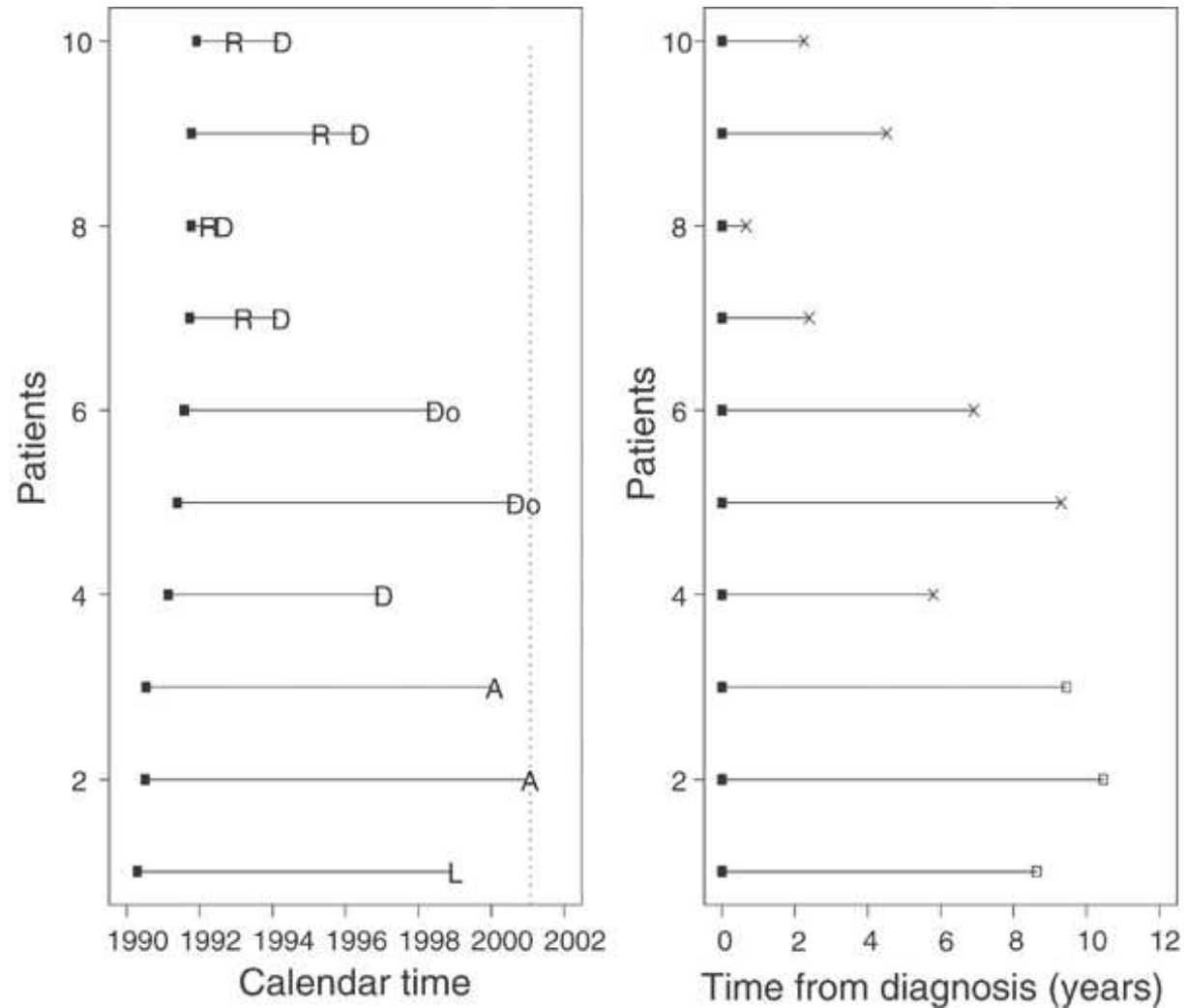
Tumor volume study with first local recurrence as event¹

	Tumor volume (cm ³)		Overall
	<50	≥50	
#Pts	279	84	363
#Person-months observed	7141	1455	8595
#Events	71	34	105
#Pts w/ event	71	34	105
Mean #person-months/pt	25.6	17.3	$\frac{8595}{363}=23.7$
Incidence rate/year	11.9%	28.0%	$\frac{105}{8595/12}=14.7\%$
Incidence rate ratio	1.0 (ref)	$\frac{28.0}{11.9}=2.4$	

Survival outcome: time to an event of interest

- From cancer treatment to relapse/progression of disease (clinical trial)
- From birth or entry into a cohort to death (epidem. mortality study)
- From exposure to death (rodent carcinogenesis bioassay)
- From cancer diagnosis (DX) to death from underlying cancer or all causes (pt series)
- From exposure to palpable tumor (rodent carcinogenesis bioassay)

Schematic of survival data



R=relapse, D=death from disease, Do=death from other cause, A=attended last clinic visit (alive),
L=loss to follow-up, X=death, open square=censored

Why is this special?

- If event occurred in all study subjects, many methods of analysis would be applicable: comparison of two or more continuous distributions
- In cancer trials, survival times rarely normally distributed but skewed with many early & relatively few late events → nonparametric test
- Survival times are censored when event of interest can no longer be observed (e.g., at end of follow-up)
 - Loss to follow-up
 - Death before event of interest
 - Other event which precludes observation of event of interest
- Special statistical methods necessary, logistic regression inappropriate

Length of follow-up

- Needs to be long enough so that sufficiently many events occur
- Needs to be tailored to event of interest (short-term, long-term)
- Interpretation of findings needs to take length of follow-up into account (conclusions beyond length of follow-up are speculation)
- In terms of statistical power, size of survival study is #events, not #pts
- Other indicator of size of study: total #person-yrs contributed by all pts

Censoring

- Examples
 - Pt has not (yet) experienced event by the close of the study
 - Loss to follow-up during study period
 - Different event makes further follow-up impossible (death from other cause, amputation)
 - Non-compliance with treatment
 - Mouse sacrificed due to other disease or death without tumor
- True time to event unknown, only: no event happened until censoring
- Censored survival time underestimates true (unknown) time to event

Important

Always state event of interest & start/end of period of observation (including censoring)

” Local control, locoregional control & disease-free survival were calculated from the time of randomization until DX of a local recurrence, regional recurrence, distant metastasis, new tumor, death, loss to follow-up or end of follow-up, whichever came first. Pts with distant metastasis, new tumor, or death as the first event were censored at that time as well as pts lost to follow-up or event-free at the end of follow-up. For calculation of local control, pts with regional recurrence were censored as well.

Overall survival was the time from randomization until death, loss to follow-up, or end of follow-up, whichever came first. Pts alive at end of follow-up or lost to follow-up were censored at that time, respectively.

Disease-specific survival & metastasis-free survival were calculated accordingly.” (Rasch et al. Cancer 2010)

Tumor volume study

- Primary outcome: first local recurrence
- Represented by 2 variables
 - Survival time: time (in months) from start of treatment until DX of local recurrence, regional recurrence, distant metastasis, new tumor, death, loss to follow-up or end of follow-up, whichever came first
 - Event indicator: binary variable indicating whether local recurrence was diagnosed at survival time (event of interest occurred, value=1) or not (censored, value=0)

Survival & hazard

- Survival probability (survivor function)

$$S(t) = P(T \geq t)$$

Probability that survival time T is at least t , i.e., that subject survives from time origin (e.g., cancer DX) to future time t

- Hazard

$$h(t) = \lim_{\Delta t \rightarrow 0} \frac{P(t \leq T < t + \Delta t | T \geq t)}{\Delta t}$$

Probability that subject who is at risk at time t has event at that time (instantaneous event rate for subject who has already survived to t), difficult to interpret but useful for modeling

- Hazard relates to incident current event rate, survival reflects cumulative non-occurrence

Kaplan-Meier (product-limit) survival estimate

- Nonparametric estimate of survival $S(t)$ from observed (censored & uncensored) survival times
- k pts have events at distinct times $t_1 < t_2 < \dots < t_k$
- Assume censoring is unrelated to event probability & events occur independently from each other

$$S(t_j) = S(t_{j-1}) \left(1 - \frac{d_j}{n_j}\right)$$

$S(t_{j-1})$ probability of being alive at t_{j-1}

n_j #pts alive just before t_j

d_j #events at t_j

$t_0 = 0$

$S(0) = 1$

Survival time calculation²

volume2=1 if tumor volume < 50 cm³
and 2 otherwise

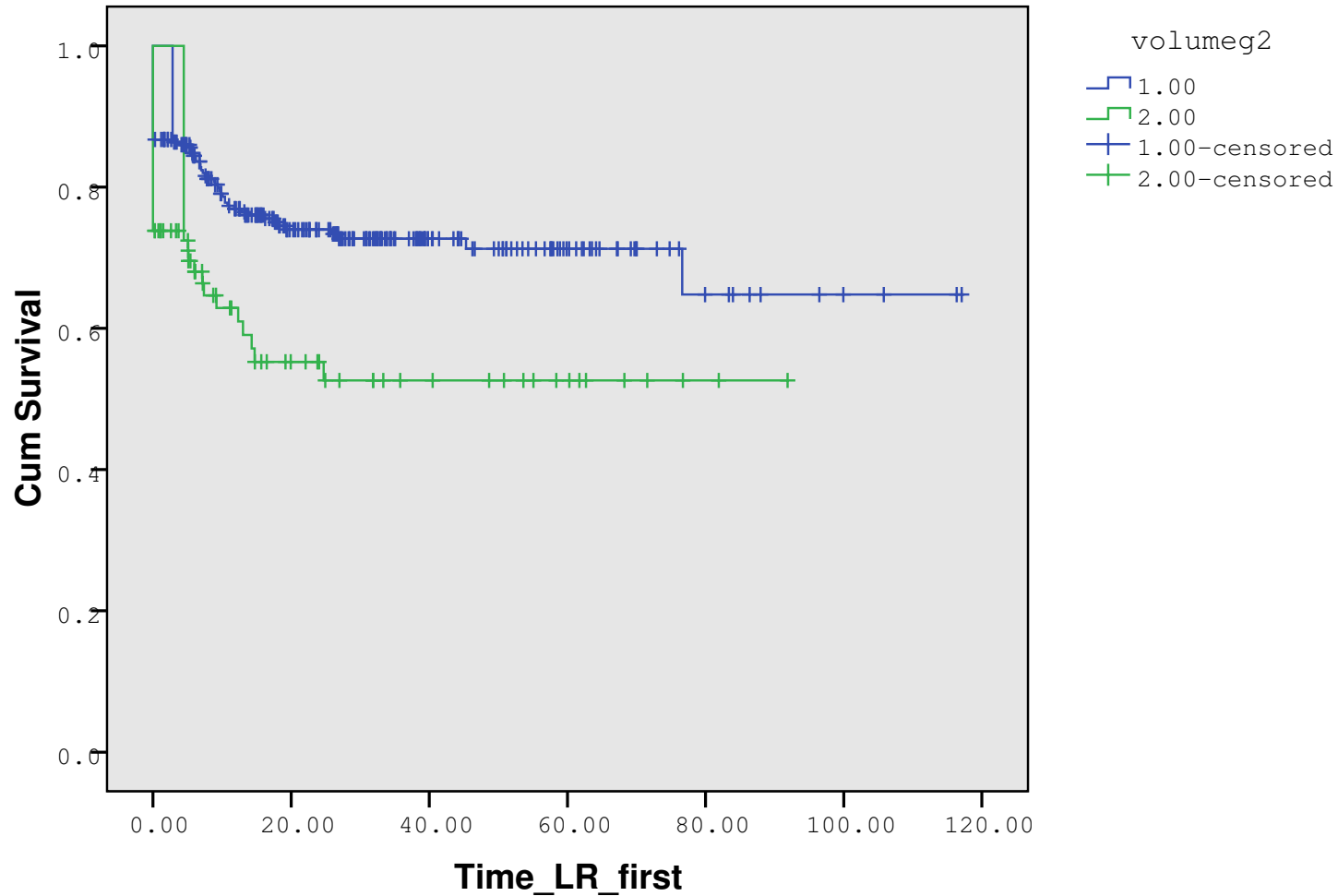
Here volume2=1:

T		KM
0	37 events	$1 * (1 - 37/279) = .867$
.3	1 censored	$.867 * (1 - 0/242) = .867$
...
2.63	1 censored	$.867 * (1 - 0/234) = .867$
2.83	1 event	$.867 * (1 - 1/233) = .864$
3.06	1 censored	$.864 * (1 - 0/232) = .864$
...
3.45	1 censored	$.864 * (1 - 0/229) = .864$
3.78	1 event	$.864 * (1 - 1/228) = .860$
...

volume2	Time	Status	Cumulative Proportion Surviving at the Time		N of Cumulative Events	N of Remaining Cases	
			Estimate	Std. Error			
1.00	1	.000	1	.	1	278	
	2	.000	1	.	2	277	
	3	.000	1	.	3	276	
	4	.000	1	.	4	275	
	5	.000	1	.	5	274	
	6	.000	1	.	6	273	
	7	.000	1	.	7	272	
	8	.000	1	.	8	271	
	9	.000	1	.	9	270	
	10	.000	1	.	10	269	
	11	.000	1	.	11	268	
	12	.000	1	.	12	267	
	13	.000	1	.	13	266	
	14	.000	1	.	14	265	
	15	.000	1	.	15	264	
	16	.000	1	.	16	263	
	17	.000	1	.	17	262	
	18	.000	1	.	18	261	
	19	.000	1	.	19	260	
	20	.000	1	.	20	259	
	21	.000	1	.	21	258	
	22	.000	1	.	22	257	
	23	.000	1	.	23	256	
	24	.000	1	.	24	255	
	25	.000	1	.	25	254	
	26	.000	1	.	26	253	
	27	.000	1	.	27	252	
	28	.000	1	.	28	251	
	29	.000	1	.	29	250	
	30	.000	1	.	30	249	
	31	.000	1	.	31	248	
	32	.000	1	.	32	247	
	33	.000	1	.	33	246	
	34	.000	1	.	34	245	
	35	.000	1	.	35	244	
	36	.000	1	.	36	243	
	37	.000	1	.867	.020	242	
	38	.300	0	.	.	37	241
	39	1.180	0	.	.	37	240
	40	1.510	0	.	.	37	239
	41	1.640	0	.	.	37	238
	42	1.680	0	.	.	37	237
	43	2.140	0	.	.	37	236
	44	2.140	0	.	.	37	235
	45	2.630	0	.	.	37	234
	46	2.830	1	.864	.021	38	233
	47	3.060	0

Survival curve

Survival Functions

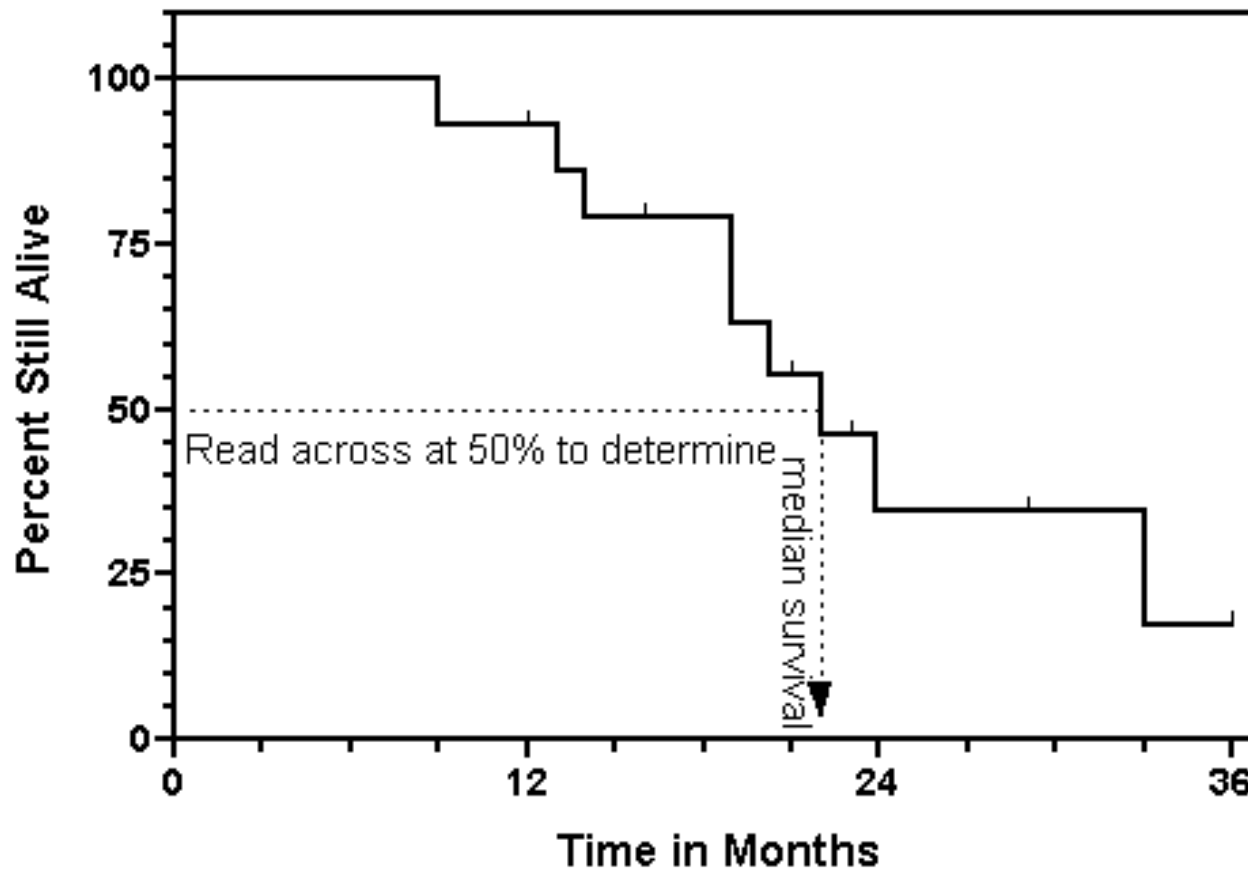


- 5-yr local recurrence-free survival for pts w/ lower volume tumors is about .71 (71%)
- Steep decline in first yr indicates generally poor prognosis
- #Lost to follow-up can be calculated from total #pts & #events, and #pts at risk at end of follow-up

Properties of KM estimator

- Value of $S(t)$ constant between event times
- Step function that changes value only at times of events
- Pts contribute information to calculations for as long as they are known to be event-free
- If all pts experience event: ratio of #pts event-free at t divided by #pts in the study
- KM survival curve is plot of KM survival probability against time
- KM plot can be used to estimate median survival (mean survival not often used due to large skew in distribution of most survival data)

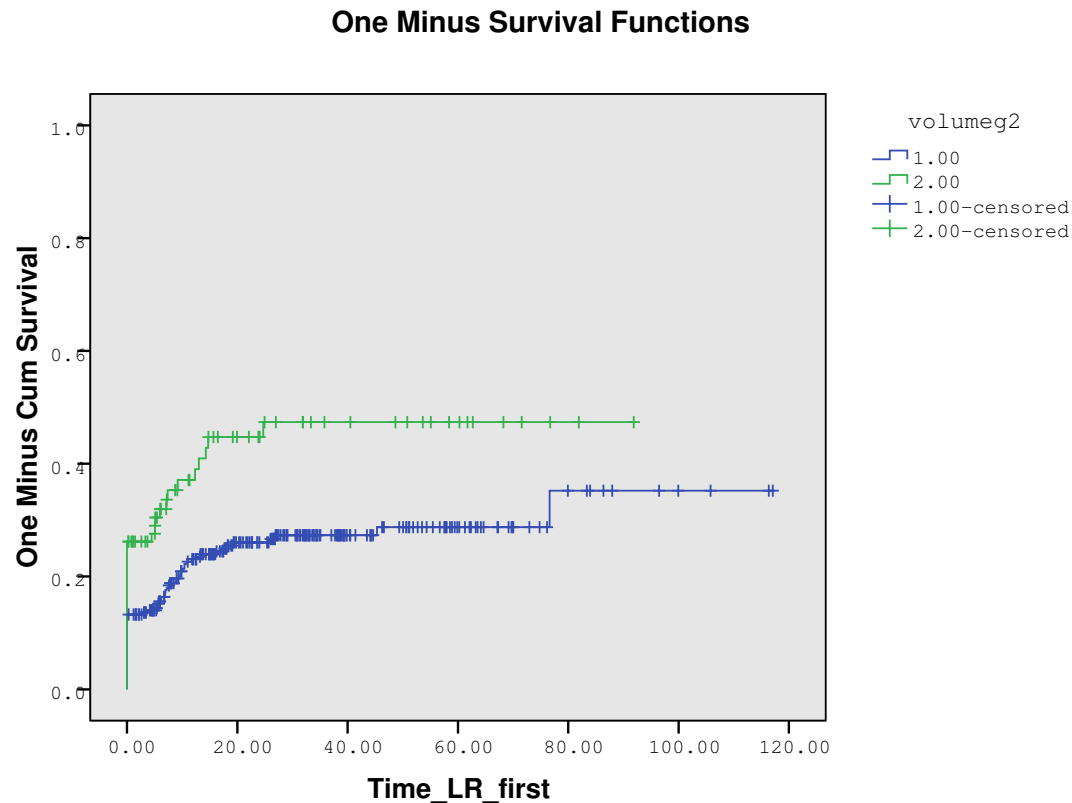
Median survival time



Median local recurrence-free survival unknown for tumor volume study because follow-up is too short (or recurrence rate too low)

Cumulative incidence of local recurrence curve

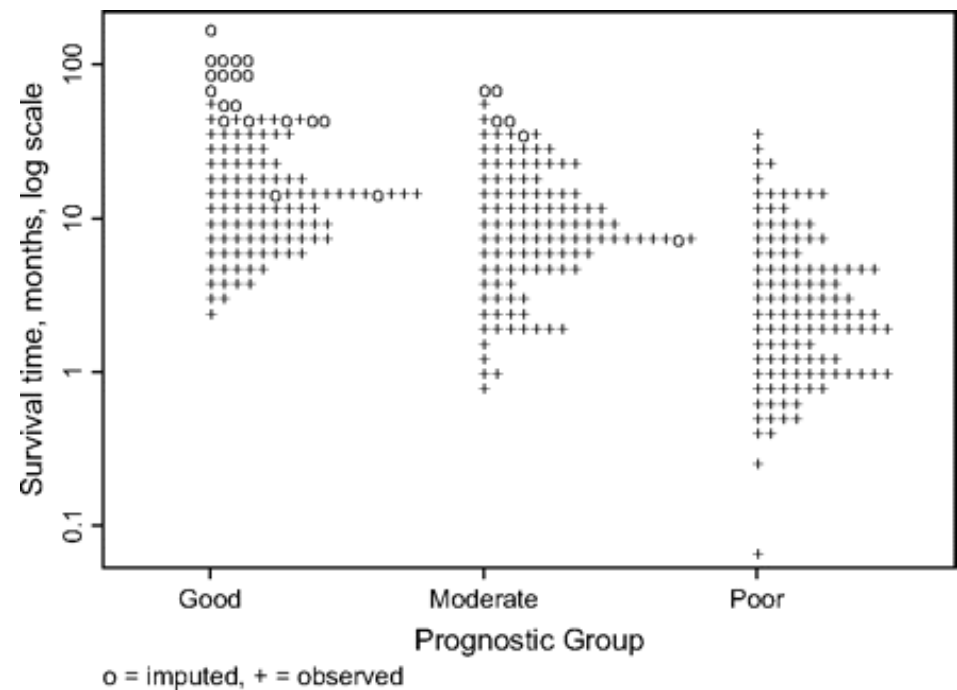
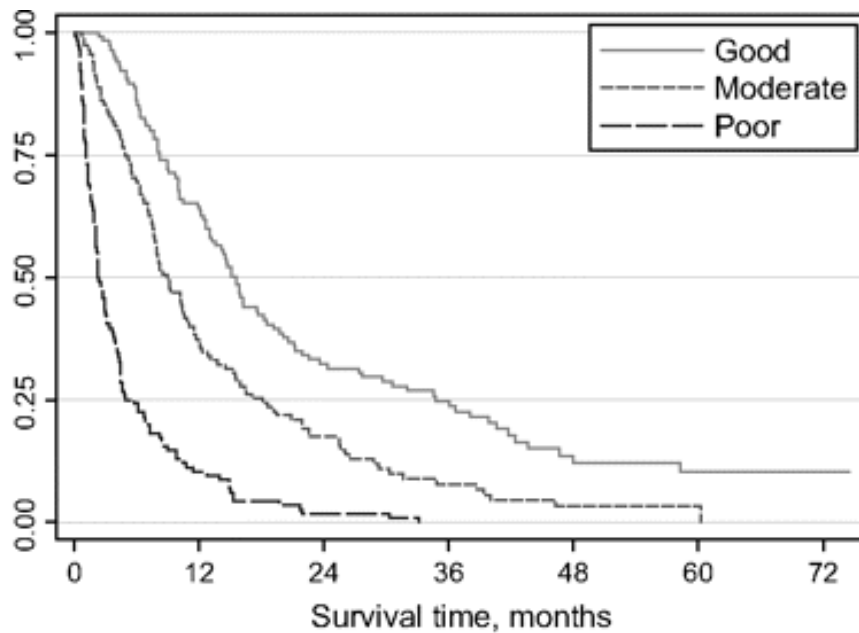
$$\text{Cumulative incidence} = 1 - S(t)$$



Cumulative incidence among pts w/ lower volume tumors is 29% for the 1st 5 yrs (5-yr survival 71%)

Separation of survival curves often overestimated by KM plots

Survival of pts by prognostic group in the Medical Research Council RE01 trial



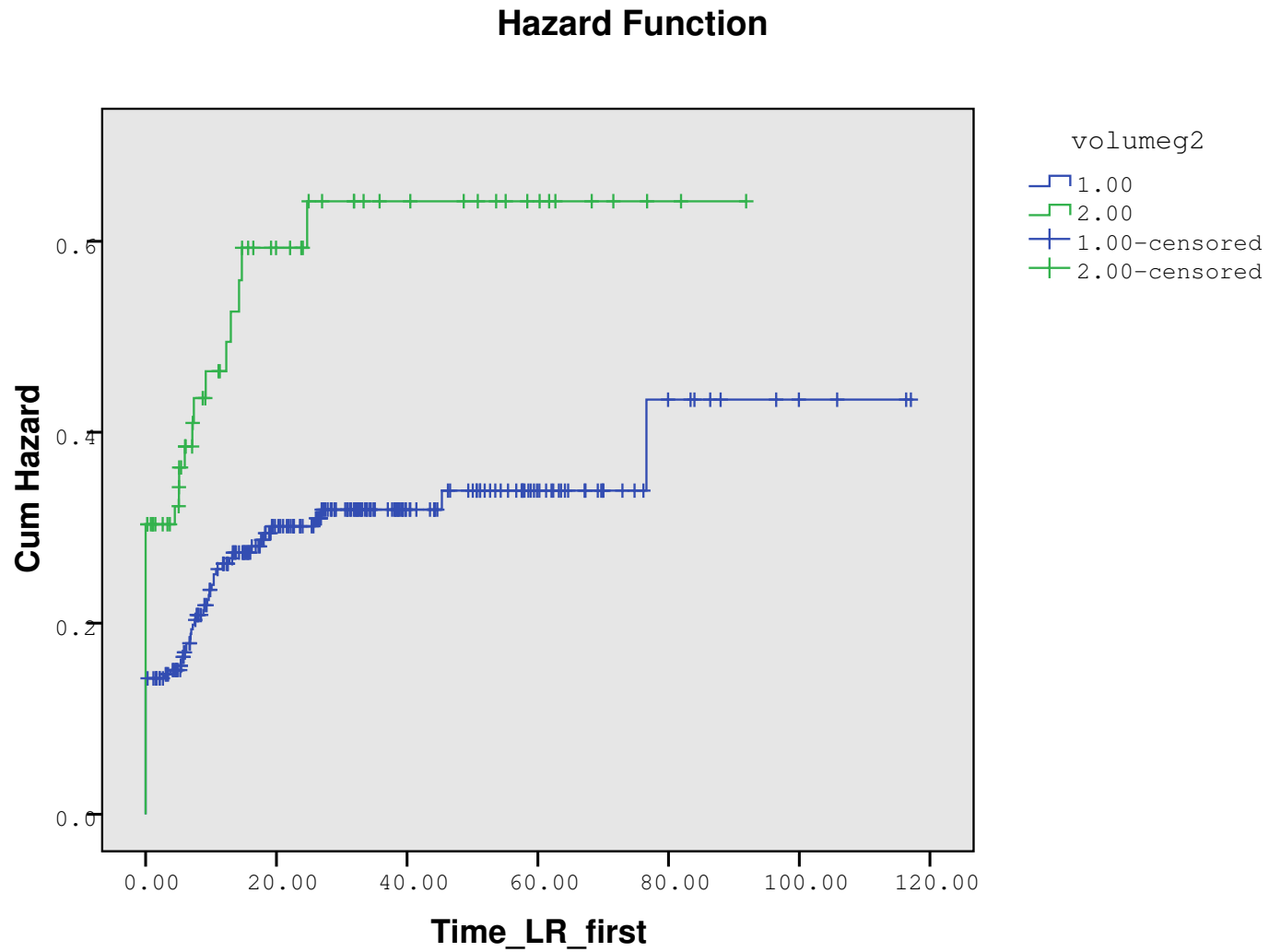
Number at risk				
	0	12	24	36
Good	115	74	36	8
Moderate	116	42	20	3
Poor	116	12	2	0

Royston et al., *J Natl Cancer Inst* 2008

Hazard & cumulative hazard

- Relationship between $S(t)$ & $h(t)$: $h(t) = -\frac{d}{dt} \log S(t)$
- $h(t)$ cannot be easily estimated
- Instead, estimate cumulative hazard $H(t) = \int_0^t h(t)dt = -\log S(t)$
"area under the hazard function between times 0 and t "
- Interpretation difficult: cumulative force of mortality or expected number of events for each subject if event was repeatable process
- $H(t)$ is used as intermediary measure for estimating $h(t)$ & for model diagnostics

Cumulative hazard curves



Two sample logrank (Mantel-Haenszel) test

- Comparing 2 survival curves nonparametrically (e.g., treatment arms)
- $H_0: h_A(t) = h_B(t)$ vs. $H_1: h_A(t) = \theta * h_B(t)$
- 2-by-2 table at each failure time

Treatment	Died at time t_j	Alive at time t_j	At risk just before t_j
A	d_{Aj}	$n_{Aj} - d_{Aj}$	n_{Aj}
B	d_{Bj}	$n_{Bj} - d_{Bj}$	n_{Bj}
Total	d_j	$n_j - d_j$	n_j

- Cell frequencies conditional on marginals under $H_0 \rightarrow d_{Aj}$ determines the rest of the table

Example: 60% group A, 40% group B, under H_0 , at each failure time 60% of all observed events are expected in group A

Test statistic

Treatment	Died at time t_j	Alive at time t_j	At risk just before t_j
A	d_{Aj}	$n_{Aj} - d_{Aj}$	n_{Aj}
B	d_{Bj}	$n_{Bj} - d_{Bj}$	n_{Bj}
Total	d_j	$n_j - d_j$	n_j

$$\frac{\left\{ \sum_j [d_{Aj} - E(d_{Aj})] \right\}^2}{\text{VAR}} = \frac{\left[\sum_j (O_i - E_i) \right]^2}{\text{VAR}} \sim \chi^2(1 \text{ DF})$$

At each event time & given the observed #events which occurred, the #events expected in the 1st group is calculated under the null hypothesis of no group differences, then summed up over all event times for the expected total #events in the first group, test compares observed #events with expected

Does local recurrence differ by tumor volume?³

Overall Comparisons

	Chi-Square	df	Sig.
Log Rank (Mantel-Cox)	10.689	1	.001

Test of equality of survival distributions for the different levels of volume².

Significantly better survival for pts w/ lower volume tumors (p=.001)

Stratified two sample logrank test

- Similar to Cochran-Mantel-Haenszel test for stratified 2-by-2 tables
- Test statistic

$$\frac{\left[\sum_{\text{strata}} \sum_j (O_{ij} - E_{ij}) \right]^2}{\text{VAR}} \sim \chi^2(1 \text{ DF})$$

- Assumption: homogeneity of treatment effects across strata
- Evaluate with Breslow-Day test of homogeneity for stratified 2-by-2 tables (M-1 DF for M strata)

Does volume measurement modality (CT vs. MRI) confound the results?⁴

Case Processing Summary

MRI_CT	volumeg2	Total N	N of Events	Censored	
				N	Percent
1	1.00	205	49	156	76.1%
	2.00	70	28	42	60.0%
	Overall	275	77	198	72.0%
2	1.00	73	21	52	71.2%
	2.00	14	6	8	57.1%
	Overall	87	27	60	69.0%
Overall	Overall	362	104	258	71.3%

Overall Comparisons^a

	Chi-Square	df	Sig.
Log Rank (Mantel-Cox)	12.066	1	.001

Test of equality of survival distributions for the different levels of volumeg2.

a. Adjusted for MRI_CT.

Volume measurement modality does apparently not confound the association between tumor volume & local recurrence

More than 2 samples

- $H_0: h_1(t) = h_2(t) = \dots = h_G(t)$

- Test statistic

$$\frac{\left[\sum_{\text{samples}} \sum_j (O_{ij} - E_{ij}) \right]^2}{\text{VAR}} \sim \chi^2(G - 1 \text{ DF})$$

O_{ij} observed #failures at event time i in group j

E_{ij} expected #failures at event time i in group j

Local recurrence-free survival by 4 tumor volume categories⁵

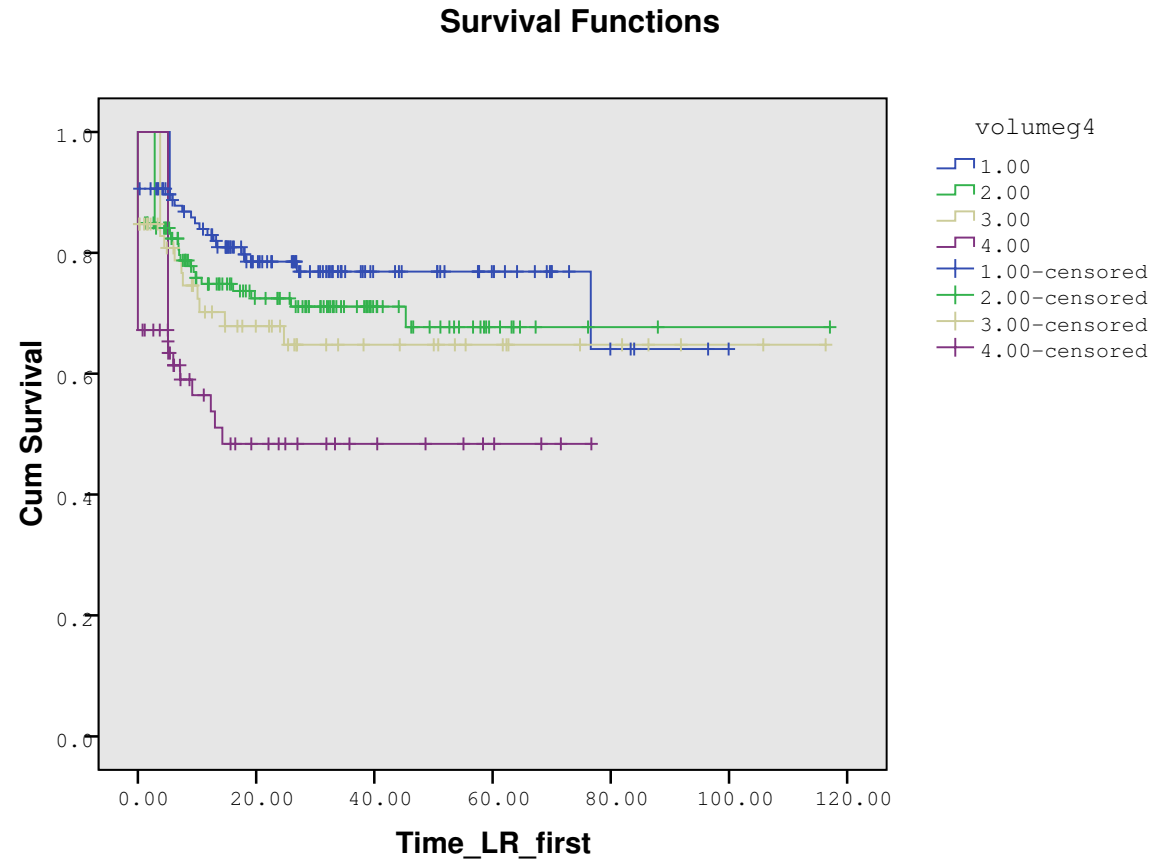
Case Processing Summary

volumeg4	Total N	N of Events	Censored	
			N	Percent
1.00	117	25	92	78.6%
2.00	126	34	92	73.0%
3.00	59	18	41	69.5%
4.00	61	28	33	54.1%
Overall	363	105	258	71.1%

Overall Comparisons

	Chi-Square	df	Sig.
Log Rank (Mantel-Cox)	17.966	3	.000

Test of equality of survival distributions for the different levels of volumeg4.



Strong evidence of heterogeneity between 4 tumor volume groups ($p < .001$)

Test of trend

- If groups naturally ordered (age groups, stages of cancer, lo-med-hi, none-moderate-severe, doses of treatment/exposure), consider trend in survival across groups
 - null hypothesis the same as above
 - ordered alternative, more powerful test, only 1 DF
- Scores w_1, w_2, \dots, w_G represent groups (1, 2, 3, ... or actual values)
- Test statistic

$$\frac{\left[\sum_{\text{samples}} w_j \sum_j (O_{ij} - E_{ij}) \right]^2}{\text{VAR}} \sim \chi^2(1 \text{ DF})$$

O_{ij} observed #failures at event time i in group j

E_{ij} expected #failures at event time i in group j

Survival trend across 4 tumor volume categories?⁶

Overall Comparisons

	Chi-Square	df	Sig.
Log Rank (Mantel-Cox)	15.518	1	.000

The vector of trend weights is -3, -1, 1, 3. This is the default.

- Significant trend in survival with tumor volume ($p < .001$)
- Compare with k-sample logrank test

Overall Comparisons

	Chi-Square	df	Sig.
Log Rank (Mantel-Cox)	17.966	3	.000

Test of equality of survival distributions for the different levels of volume⁴.

Comments on logrank test

- Assumptions same as for KM survival estimates
 - Censoring unrelated to event probability (uninformative)
 - Survival similar for subjects recruited early & late in study

Deviations matter most when different across groups

- Purely test of significance, no estimate of magnitude of difference
- Most likely detects difference when risk of event is consistently greater in one group

- Unlikely to detect difference if survival curves cross → always plot survival curves
- For 2 groups, logrank test tests H_0 that ratio of the hazard rates in the two groups equals 1
- Hazard ratio (HR) is measure of relative survival experience, can be estimated by

$$HR = \frac{O_1/E_1}{O_2/E_2}$$

where O_i/E_i is estimated relative (excess) hazard in group i

- In practice, better estimate HR by Cox regression

Uninformative censoring

- Subjects who are censored at a given point in time should be as likely to have a subsequent event as those individuals who remain in the study
- Examples for informative censoring: withdrawal from clinical trial because of drug toxicity or worsening clinical condition
- If censoring occurs in a small #pts, impact of informative censoring is likely small

SPSS code (syntax and clicking)

1. Tumor volume study with first local recurrence as event
Click Analyze – Descriptive Statistics – Frequencies and select variable volumeg2. Click Analyze – Compare Means – Means and select variables Time_LR_first and volumeg2. Click Analyze – Descriptive Statistics – Crosstabs and select variables First_LR and volumeg2.

```
FREQUENCIES VARIABLES=volumeg2
  /ORDER=ANALYSIS.
MEANS TABLES=Time_LR_first BY volumeg2
  /CELLS MEAN SUM.
CROSSTABS
  /TABLES=First_LR BY volumeg2
  /FORMAT=AVALUE TABLES
  /CELLS=COUNT
  /COUNT ROUND CELL.
```

2. Survival time calculation

Click Analyze – Survival – Kaplan-Meier and select variable Time_LR_first as Time, First_LR as Status with 1 indicating an event and volumeg2 as Factor. Under Options, select Survival and One Minus Survival in Plots.

```
KM Time_LR_first BY volumeg2
  /STATUS=First_LR(1)
  /PRINT TABLE MEAN
  /PLOT SURVIVAL OMS.
```

3. Does local recurrence differ by tumor volume?

Click Analyze - Survival - Kaplan-Meier and select variables as above. Request logrank test under Compare Factors.

```
KM Time_LR_first BY volumeg2
  /STATUS=First_LR(1)
  /PRINT NONE
  /TEST LOGRANK
  /COMPARE OVERALL POOLED.
```

4. Does volume measurement modality (CT vs. MRI) confound the results?

Request Kaplan-Meier plot and logrank test above, but select variable MRI_CT as stratification variable.

```
KM Time_LR_first BY volumeg2
/STRATA=MRI_CT
/STATUS=First_LR(1)
/PRINT NONE
/TEST LOGRANK
/COMPARE OVERALL POOLED.
```

5. Local recurrence-free survival by 4 tumor volume categories
Click Analyze - Survival - Kaplan-Meier and select variables as above.
Request logrank test under Compare Factors.

```
KM Time_LR_first BY volumeg4
/STATUS=First_LR(1)
/PRINT NONE
/TEST LOGRANK
/COMPARE OVERALL POOLED.
```

6. Survival trend across 4 tumor volume categories

Click Analyze - Survival - Kaplan-Meier and select variables as above. Under Compare Factors, request logrank test and linear trend test.

```
KM Time_LR_first BY volumeg4  
  /STATUS=First_LR(1)  
  /PRINT NONE  
  /TEST LOGRANK  
  /TREND  
  /COMPARE OVERALL POOLED.
```