

## Analiza cenzuriranih podataka i krivulje preživljjenja

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## Analiza preživljjenja

1. temeljni pojmovi
2. izračun vjerojatnosti preživljjenja
  - a) tablice preživljjenja
  - b) Kaplan-Meierov postupak
3. rizik umiranja
4. programska potpora
5. usporedba podataka o preživljjenju
6. statističko zaključivanje
7. regresijska analiza cenzuriranih podataka



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## Obrada podataka o preživljjenju bolesnika

- analiza preživljjenja
- *survival analysis*
- ponekad
  - analiza tablica preživljjenja
  - analiza osiguravateljskih (aktuarskih) podataka
  - *actuarial analysis*



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[www.aktuari.hr](http://www.aktuari.hr)

- aktuar – stručnjak koji se bavi problemima financijske neizvjesnosti i rizika koristeći matematičke metode teorije vjerojatnosti, statistike i financijske matematike
- posao – analiza podataka iz prošlosti, procjenu postojećih rizika i razvoj modela za projekciju budućih događaja
- zaposlenje – osiguranje i mirovinsko osiguranje
- znanja – matematika, ekonomija, praksa i zakoni države u kojoj radi, demografska i financijska kretanja, vještina komunikacije



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## Analiza preživljjenja

- Edmund Halley, 17. st
- engleski astronom, geofizičar, matematičar, meteorolog i fizičar
- [http://en.wikipedia.org/wiki/Edmond\\_Halley](http://en.wikipedia.org/wiki/Edmond_Halley)



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## Aktuarske tablice preživljjenja (tablice smrtnosti)

8 National Vital Statistics Reports, Vol. 54, No. 14, April 19, 2006

Table 1. Life table for the total population: United States, 2003

[Click here for spreadsheet version](#)

Age	Probability of dying between ages $x$ to $x+1$ $d_x$	Number surviving to age $x$ $l_x$	Number dying between ages $x$ to $x+1$ $d_{x+1}$	Person-years lived between ages $x$ to $x+1$ $t_{x+1}$	Total number of person-years lived above age $x$ $T_x$	Expectation of life at age $x$ $e_x$
					$d_x$	$l_x$
0-1	0.000865	100,000	687	99,394	774,865	77.5
1-2	0.000865	99,313	46	99,267	774,147	77.0
2-3	0.000831	99,267	33	99,234	7,550,161	76.1
3-4	0.000829	99,234	26	99,222	7,450,930	75.1
4-5	0.000818	99,209	20	99,199	7,351,709	74.1
5-6	0.000808	99,189	17	99,172	7,251,510	73.1
6-7	0.000815	99,172	15	99,165	7,153,329	72.1
7-8	0.000814	99,157	14	99,150	7,054,164	71.1
8-9	0.000819	99,143	14	99,136	6,955,017	70.2
9-10	0.000834	99,130	13	99,123	6,855,877	69.2
10-11	0.000815	99,116	16	99,108	6,756,754	68.2
11-12	0.000816	99,100	15	99,093	6,657,646	67.2
12-13	0.000817	99,085	17	99,077	6,558,533	66.2



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[http://en.wikipedia.org/wiki/Actuarial\\_table](http://en.wikipedia.org/wiki/Actuarial_table)

## Analiza preživljjenja

- psihijatrija – 1%
- patologija – 1%
- kirurgija – 12%
- onkologija – 14%
- izvorni znanstveni radovi u *The New England Journal of Medicine* – 32%
- podaci 1986.-2001., Dawson Saunders & Trapp, Basic and Clinical Biostatistics

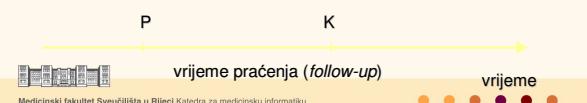


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## Analiza preživljjenja

- analiza podataka vezanih uz vremensko praćenje događaja
- dvije točke praćenja:
  - početak (P) (*time origin*)
  - kraj (K) (*end point*)



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## Početak praćenja

- rođenje
- pojava znaka bolesti
- postavljanje dijagnoze
- početak liječenja
- dan operativnog zahvata



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K

vrijeme praćenja



## Kraj praćenja

- smrt od osnovne bolesti
- smrt (svi ostali mogući uzroci)
- ponovno javljanje bolesti
- postizanje učinka liječenja
- gubitak iz uzorka (ispitne skupine)



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P

vrijeme praćenja



## Kraj praćenja

- smrt od osnovne bolesti
- smrt (svi ostali mogući uzroci)

uskladeno  
preživljenje  
engl. *adjusted survival rate*



ukupno preživljenje  
engl. *observed survival rate*

Manual for Staging of Cancer  
3rd ed., AJCC



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## Vrijeme praćenja

- raspodjela u pravilu nije simetrična
- podaci su nepotpuni, praćenje je nepotpuno, "cenzurirano" (*censored data*)
- podaci za primjere:
  - istraživanje karcinoma usne šupljine
  - MFK KBG
  - dr. Ivica Lukšić
  - n = 52; 1. siječnja 2000. – 31. prosinca 2004.
  - reprezentativni probani uzorak
    - dio populacije tog razdoblja
    - prva dg. karcinoma, bez regionalnih metastaza, itd.

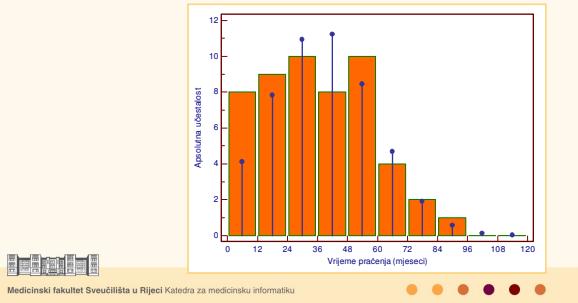


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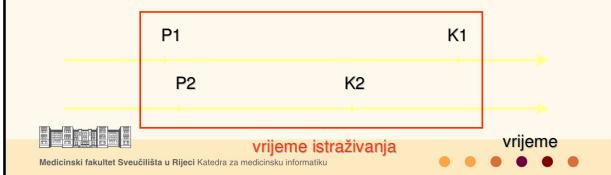
## Vrijeme praćenja (1)

- raspodjela u pravilu nije simetrična



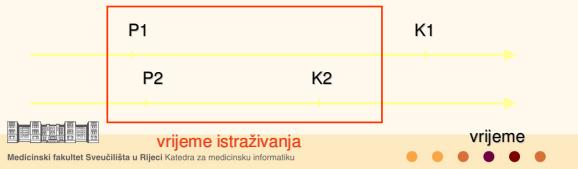
## Vrijeme praćenja (2)

- potpuni podaci (potpuno praćenje)



## Vrijeme praćenja (2)

- podaci su nepotpuni, praćenje je nepotpuno, "cenzurirano"
    - **cenzurirano vrijeme praćenja** = jedinka tijekom praćenja ne dostiže očekivani događaj



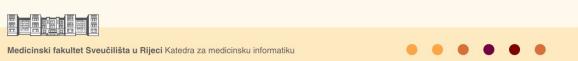
### Vrijeme praćenja (3)

- podaci su nepotpuni, praćenje je nepotpuno, "cenzurirano"
    - **cenzurirano vrijeme praćenja** = jedinka tijekom praćenja ne dostiže očekivani događaj

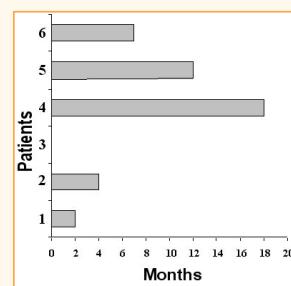


## Cenzuriranje

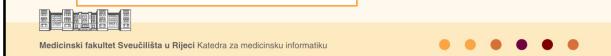
- događaj se ostvaruje = 1
  - sve ostalo = 0 (cenzurirani podaci)
    - kraj istraživanja (*end of the study*)
    - gubitak iz praćenja (*lost to follow-up*)
    - ostali događaji

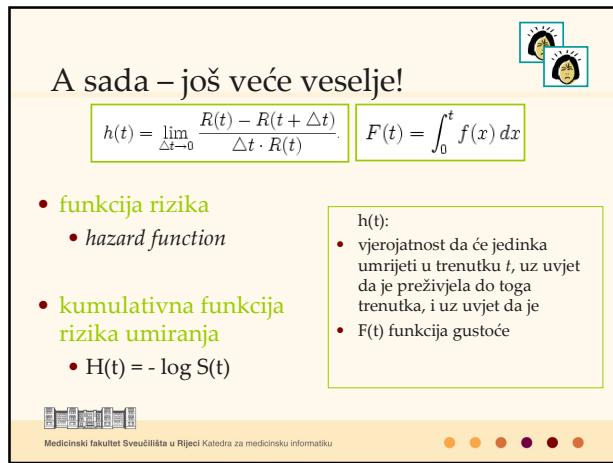
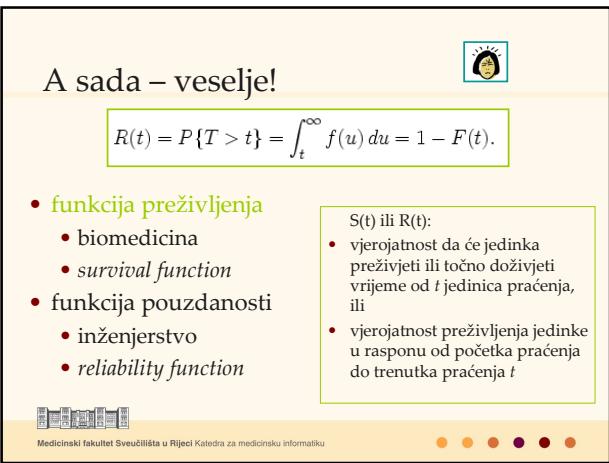
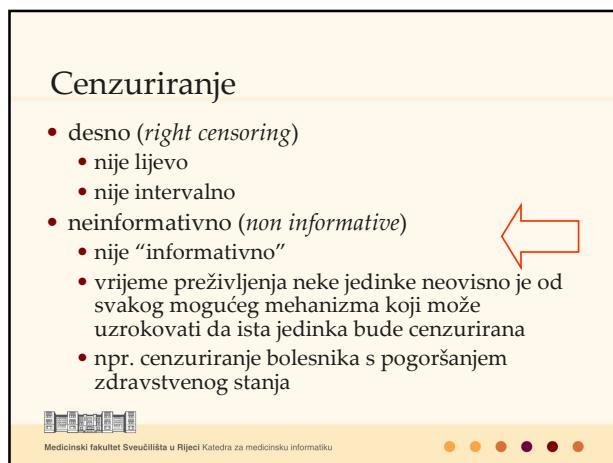
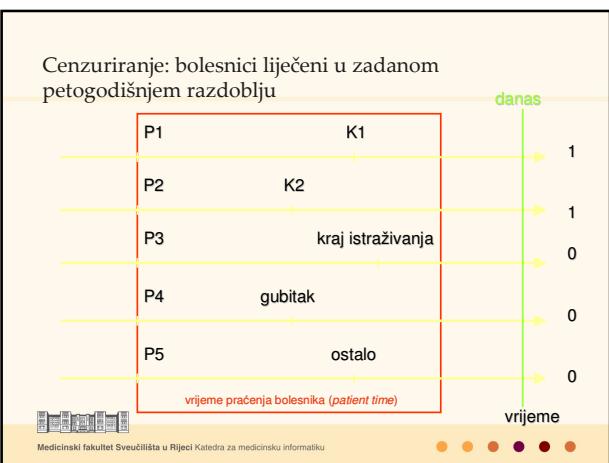
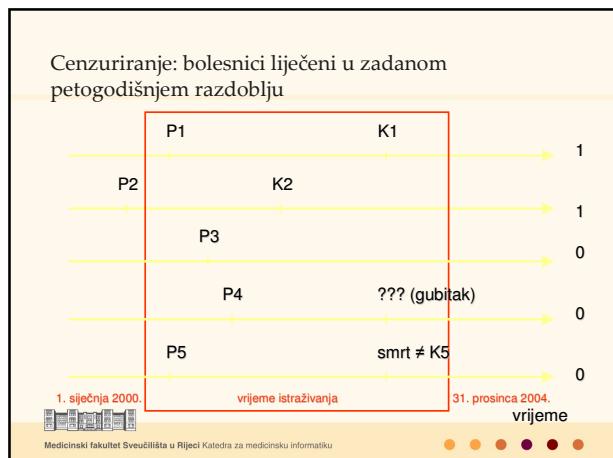
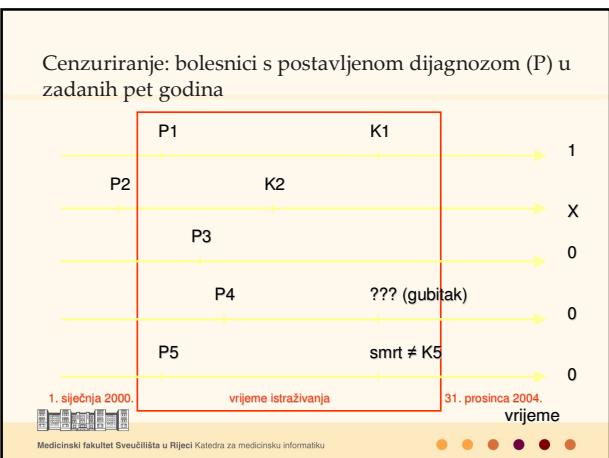


## “Izgubljen iz praćenja”



**Figure 3. Outcome**  
The mean length of survival for our patients was 7 months.  
Patient 2 was lost to follow-up.





## Zaključak: podaci o preživljjenju

- vjerojatnost preživljjenja
  - $S(t)$
- rizik umiranja
  - $H(t) = -\log S(t)$



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## Izračunavanje preživljjenja

- neparametrijski postupci
  - Cutler-Edererov postupak (tablice preživljjenja)
  - Kaplan-Meireov postupak
- parametrijski



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## I. Tablice preživljjenja

- osiguravateljske tablice
- tablice smrtnosti

vrijeme praćenja	vjerojatnost smrti $q=d/(n \cdot w)$	vjerojatnost preživljjenja $p=1-q$	kumulativna vjerojatnost preživljjenja $S(t)=Tp$
0-12 mј.	0,11	0,89	0,89
13-24 mј.	0,27	0,73	0,65
25-36 mј.	0	1	0,65
37-48 mј.	0,4	0,6	0,39
49-60 mј.	0	1	0,39



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## Kako do preživljjenja?

1. upis podataka
2. preuređenje podataka
3. izračun podataka



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## 1. Upis podataka, Excel®

	pacijent	datumop	datumkraj	cenzus	mjeseci
2	23456	23.6.2000	15.4.2007	0	81,8
3	24485	15.10.2003	8.11.2005	0	24,8
4	23080	25.7.2000	29.8.2004	0	49,2
5	23511	28.12.2001	15.2.2007	0	61,6
6	24188	20.2.2002	29.10.2004	0	32,3
7	22701	17.12.2003	8.6.2005	1	17,7
8	24241	17.7.2002	29.4.2007	0	57,4
9	23480	15.5.2003	20.8.2007	0	51,2
10	22823	5.10.2000	26.9.2002	1	23,7



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## 2. Preuređenje podataka

A	B	C	D	E
1 pacijent	datumop	datumkraj	cenzus	mjeseci
2 24485	23.6.2000	15.4.2007	0	81,8
3 23080	15.10.2003	8.11.2005	0	24,8
4 24188	25.7.2000	29.8.2004	0	49,2
5 23511	28.12.2001	15.2.2007	0	61,6
6 22701	20.2.2002	29.10.2004	0	32,3
7 24241	17.12.2003	8.6.2005	1	17,7
8 23480	17.7.2002	29.4.2007	0	57,4
9 22823	15.5.2003	20.8.2007	0	51,2
10 2309	5.10.2000	26.9.2002	1	23,7
11				

vrijeme praćenja	živi na početku intervala n	smrtni ishod u intervalu d	cenzurirani u intervalu w
0-12 mј.	10	1	1
13-24 mј.	8	2	1
25-36 mј.	5	0	2
37-48 mј.	3	1	1
49-60 mј.	1	0	1



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### 3. Izračun podataka

vrijeme praćenja	vjerojatnost smrti $q=d/(n-w/2)$	vjerojatnost preživljjenja $p=1-q$	kumulativna vjerojatnost preživljjenja $S(t)=\prod p$
0-12 mj.	0,11	0,89	0,89
13-24 mj.	0,27	0,73	0,65
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0-12 mj.	10	1	1
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25-36 mj.	5	0	2
37-48 mj.	3	1	1
49-60 mj.	1	0	1

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49-60 mj.	0	1	0,39

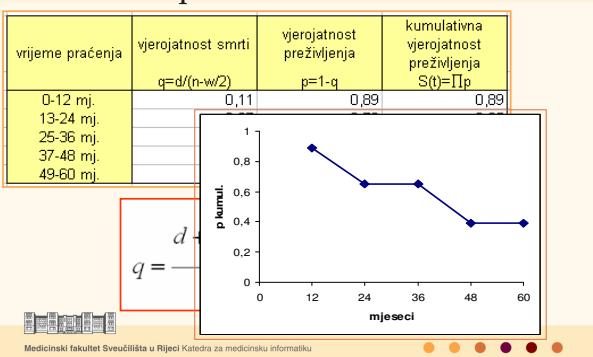
$$q = \frac{d + \frac{1}{2}wq}{n}$$

d – smrtni ishod u intervalu  
n – živi na početku intervala  
w – izgubljeni u intervalu

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### 3. Izračun podataka



### II. Kaplan-Meierov postupak

- nema zadane intervale praćenja
- vjerojatnost preživljjenja se izračunava za svakog bolesnika koji umre
- cenzurirani ispitanici nisu dio izračunavanja vjerojatnosti
- dugotrajno izračunavanje kod velikih skupina ispitanika



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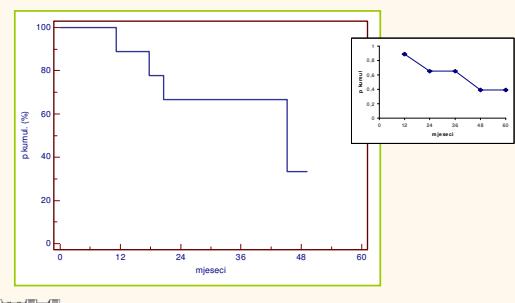
### KM podaci o preživljjenju

Survival time	A	B	C	D	E
Endpoint	mjeseći	datum pog.	datum kraj	cenzus	mjeseći
Sample size					10
Median survival					45,2
Survival time	Survival Proportion	Standard Error			
7,4	-	-			
11,2	0,889	0,105			
17,7	0,778	0,139			
20,6	0,667	0,157			
24,3	-	-			
24,8	-	-			
32,3	-	-			
37,2	-	-			
45,2	0,333	0,248			
49,2	-	-			

MedCalc® Version 9.3.0  
Windows 96/NT/Me/2000/XP/Vista  
<http://www.medcalc.be>  
Copyright © 1993-2007  
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This product is licensed to:  
Mladen Petrovski

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### KM krivulja preživljjenja



## Kaplan, Meier

- Kaplan EL, Meier P. Nonparametric estimation from incomplete observations. J Am Stat Assoc 1958;53:457-81.



**NONPARAMETRIC ESTIMATION FROM INCOMPLETE OBSERVATIONS\***  
Peter M. Lee  
University of California Radiation Laboratory  
Part I. Methods  
In this paper, methods of estimation of survival rates in death studies are presented. The method of estimation is based on the assumption that the survival rate at time  $t$  is the probability that a randomly selected individual in the study population is still alive at time  $t$ . The estimate is obtained by dividing the number of individuals surviving at time  $t$  by the number of individuals at risk at time  $t$ . The estimate is called the Kaplan-Meier product-limit estimate. The estimate is unbiased if all deaths in the study are observed. If some deaths are unobserved, the better results are obtained by dividing the number of individuals surviving at time  $t$  by the number of individuals at risk at time  $t$ , where the number of individuals at risk at time  $t$  is the number of individuals at risk at time  $t$  minus the number of deaths that have been observed up to time  $t$ . This estimate is the likelihood estimate associated with the product-limit estimate.  
Other estimates that are discussed are the actuarial estimate (which is the ratio of the number of individuals surviving at time  $t$  to the number of individuals at risk at time  $t$ ) and the estimate obtained by dividing the number of individuals surviving at time  $t$  by the number of individuals at risk at time  $t$ , where the number of individuals at risk at time  $t$  is the number of individuals at risk at time  $t$  minus the number of deaths that have been observed up to time  $t$ . The estimate of  $P(t)$  is also given for the case in which the proportion of deaths is constant.  
\* A portion of this work was done while the author was at the University of California Radiation Laboratory and later at the University of Southern California.  
The work was cited by a grant from the Office of Science Research.

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Journal of the American Statistical Association, Vol. 53, No. 282, June 1958  
ISSN 0162-1459/58/282457-05\$01.00/0

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## Kaplan, Meier

- među 5 najcitanijih radova u znanosti od trenutka objavlјivanja (M. Zhou, Kentucky University; <http://www.ms.uky.edu/~mai/>)
- prikaz krivulje u zavisnosti od N <http://www.ms.uky.edu/~mai/java/stat/KapMei.html>



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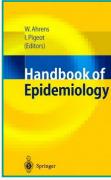
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## Rizik umiranja

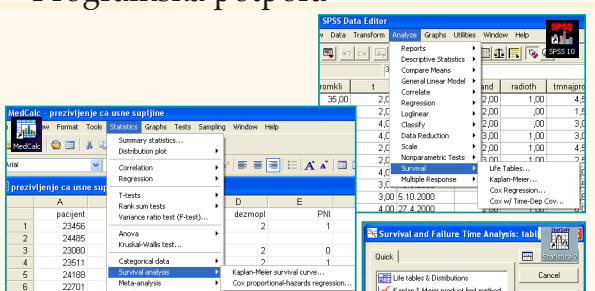
- $H(t) = - \log S(t)$
- snaga mortaliteta (epidemiologija)
- ...

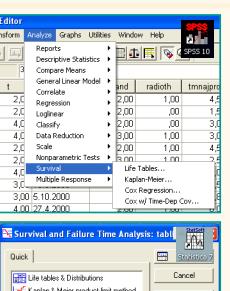


[http://www.bips.uni-bremen.de/handbook\\_of\\_epidemiology.php](http://www.bips.uni-bremen.de/handbook_of_epidemiology.php)

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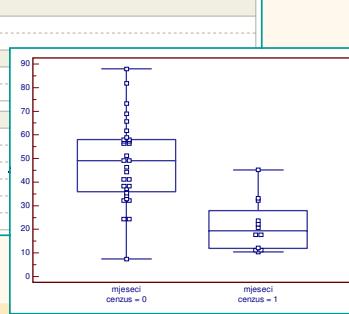
## Programska potpora





## Primjer...

Sample 1  
Variable mijeseci cenzus = 0  
Select  
Sample 2  
Variable mijeseci cenzus = 1  
Select  
Sample size  
Arithmetic mean  
95% CI for the mean  
Variance  
Standard deviation  
Standard error of the mean



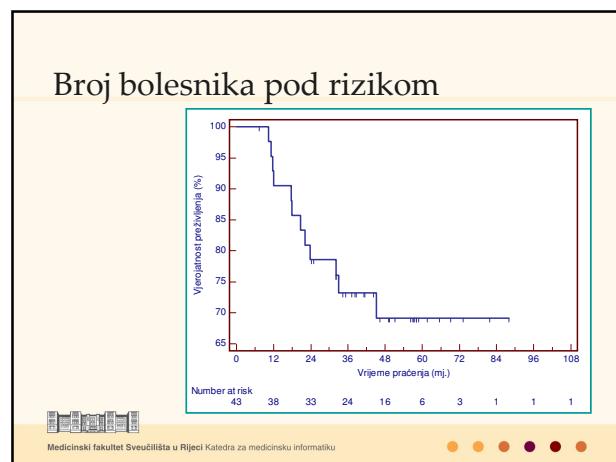
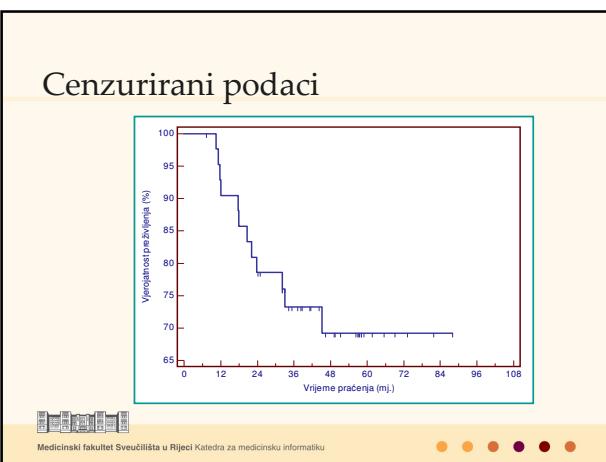
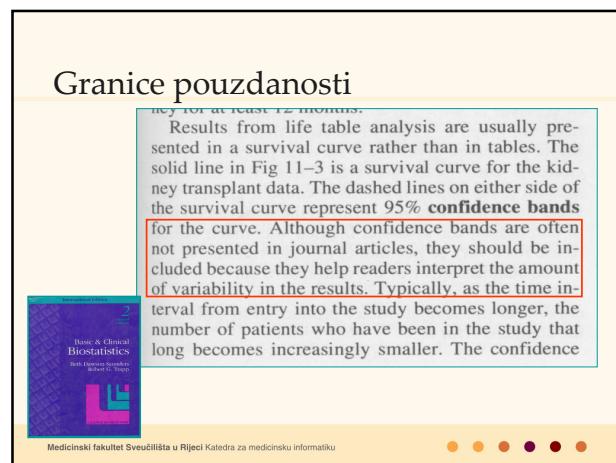
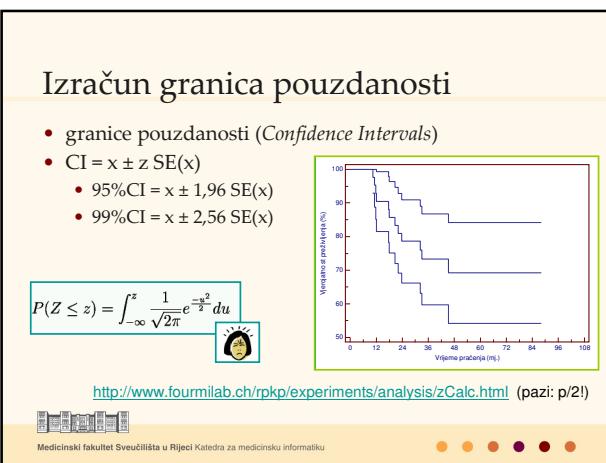
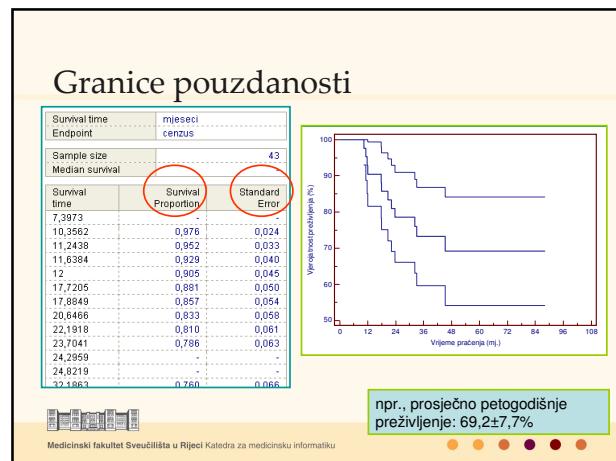
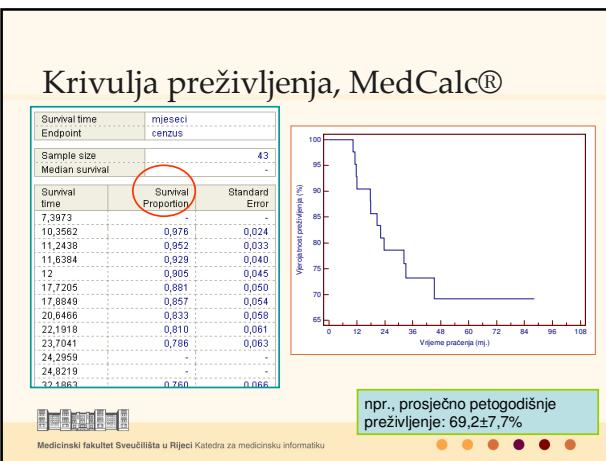
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## Preživljenje, MedCalc®

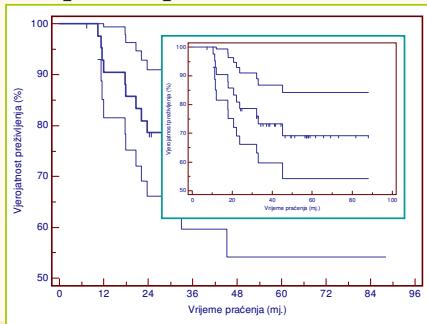
Survival time	mjeseci	cenzus	Sample size	Median survival	43
7,3973					
10,3562					
11,2438					
11,6384					
12					
17,7205					
17,8849					
20,6466					
22,1918					
23,7041					
24,2959					
24,8219					
32,1963					
	Survival Proportion				
	0,976	0,024			
	0,952	0,033			
	0,929	0,040			
	0,905	0,045			
	0,881	0,050			
	0,857	0,054			
	0,833	0,058			
	0,810	0,061			
	0,786	0,063			
	0,760	0,066			
	Standard Error				
	0,092	0,077			

npr., prosječno potugodišnje preživljenje:  $69,2 \pm 7,7\%$

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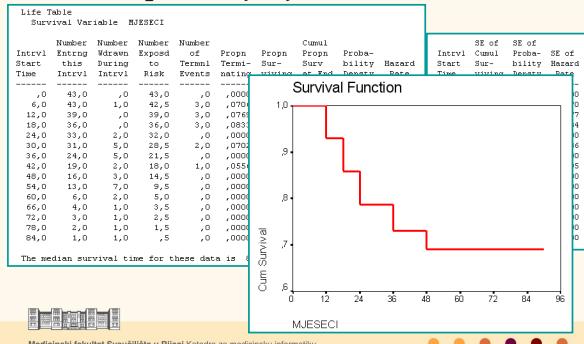


## Standardni prikaz podataka



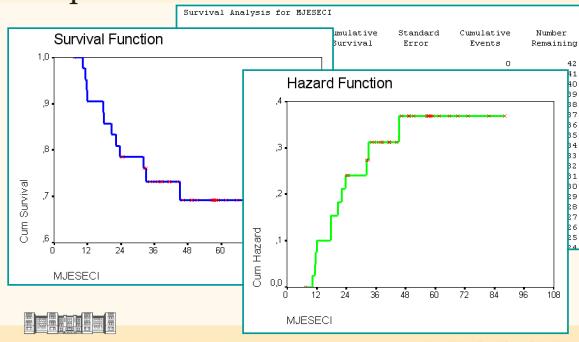
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## Tablice preživljjenja, SPSS®



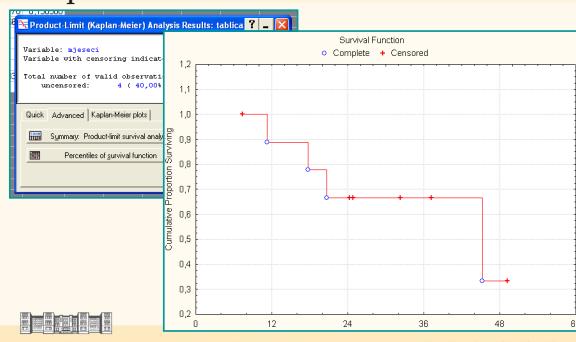
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## Kaplan-Meier, SPSS®



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## Kaplan-Meier, Statistica®



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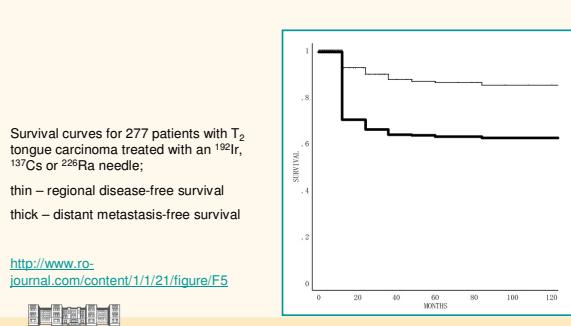
## "Preživljjenje"

- generalno: preživljjenje
  - survival
  - praćenje
  - početak
  - smrt od osnovne bolesti
- posebno: preživljjenje bez znakova bolesti
  - disease-free survival (DFS)
  - praćenje
  - početak
  - relaps bolesti ili smrt od osnovne bolesti



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## DFS, primjer



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## Usporedba podataka o preživljjenju

- usporedba dvije skupine podataka
  - log-rank (logrank) test*
  - Mentelov ili Mantel-Coxov test
- Wilcoxonov test
  - generalizirani Wilcoxonov test
  - Gehanov test
  - Gehan-Breslowljev test
  - opći Kruskal-Wallisov test za cenzurirane podatke
- Mantel-Haenszelov test
- Tarone-Wareov test
- usporedba triju i više skupina



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## Usporediti dvije skupine...

ceiving a kidney in 1984 is above the curve for patients receiving a kidney in 1978, indicating a higher proportion of patients retaining a functioning graft at any one point in time. However, variation in samples may be expected to occur simply by chance, and a reasonable question is whether the differences between the two patient cohorts is greater than expected by chance. To test this hypothesis, we need methods to compare survival distributions. If there are no censored observations, the **Wilcoxon rank-sum test** in-

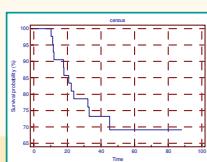
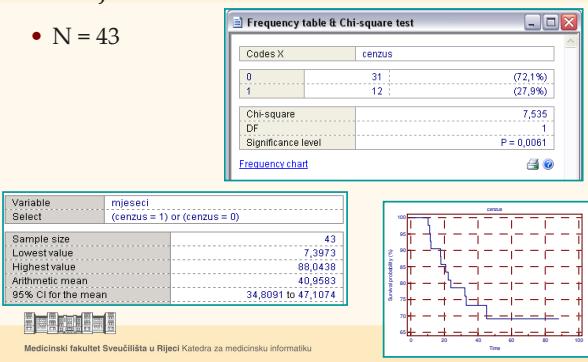


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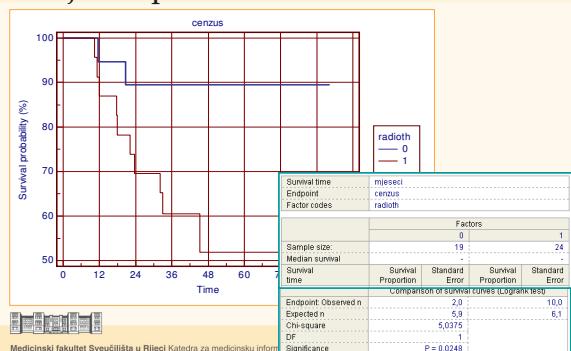


## Primjer

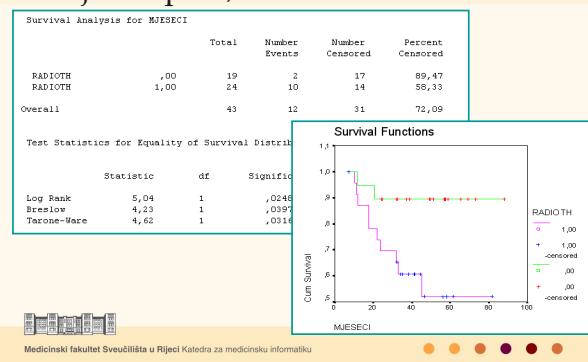
- N = 43



## Dvije skupine, MedCalc®

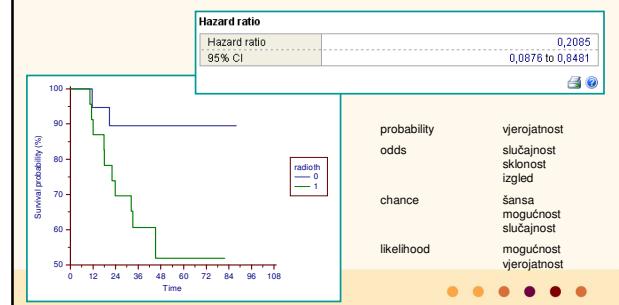


## Dvije skupine, SPSS®

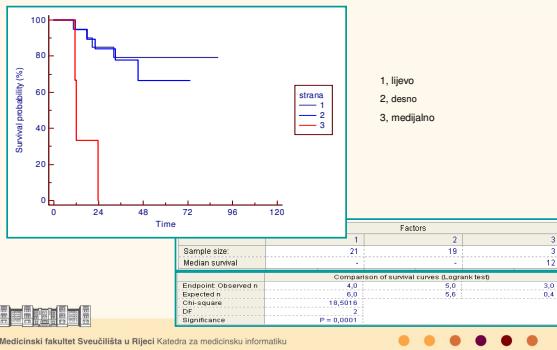


## Omjer izgleda

- odds ratio, hazard ratio*



## Usporedba triju skupina



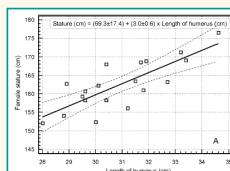
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## Zaključivanje

- granice pouzdanosti
- p-vrijednosti

## Regresijska analiza podataka

- korelacija
- pravocrtna regresija
- višestruka regresija
- logistička regresija
- Coxova regresija
- Passing-Bablokova regresija



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## Podatci &



- zavisni pokazatelji
  - vrijeme preživljavanja
  - podatak o ishodu (cenzuriranje)
- nezavisni pokazatelji
  - prediktori ili kovarijate (covariates)
  - sve mjerene ljestvice dopuštene
- rezultat
  - regresijski koeficijenti
    - ⇒ omjer rizika (hazard ratio (HR), ratio of the hazard function)
    - ⇒ mjera rizika (relativni rizik; relative risk (RR))

$$h_i(t) = h_0(t) \exp(\beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_k x_{ik})$$

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## Coxov regresijski test

- multivarijatni postupak
- Cox regression
  - Cox, 1972.
  - proportional hazard model

### 10.5.3 Importance of the Cox Model

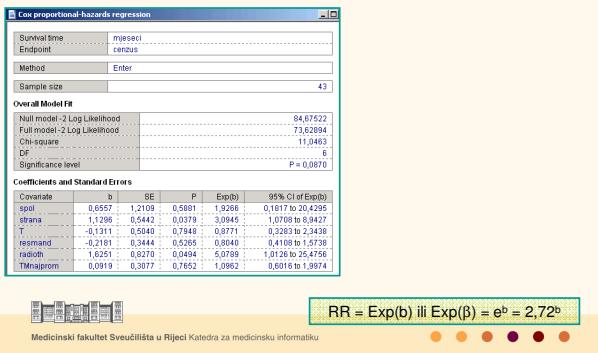
The Cox model is very useful in medicine, and it is easy to see why it is being used with increasing frequency. It provides the only valid method of predicting a time-dependent outcome, and many health-related outcomes are related to time. If the independent variables are divided into two categories (dichotomized), the exponential of the regression coefficient,  $\exp(b)$ , is the odds ratio, a useful way to interpret the risk associated with any specific factor. In addition, the Cox model provides a method for producing survival curves that are adjusted for confounding variables. The Cox model can be extended to the case of multiple events for a subject, but that

## Primjer, MedCalc® Coxov regresijski test

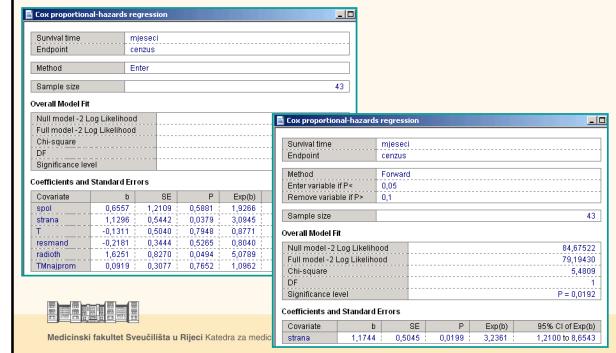
- analiza preživljavanja (smrt od osnovne bolesti)
- podaci – cenzurirani
- nezavisni pokazatelji
  - spol (M, Ž)
  - zahvaćena strana lica (L, D, M)
  - T-klasifikacija
    - resekcija donje čeljusti (1-5 kao nije, segmentalna, marginalna...)
    - liječenje zračenjem (da, ne)
  - najveći promjer tumora (cm)

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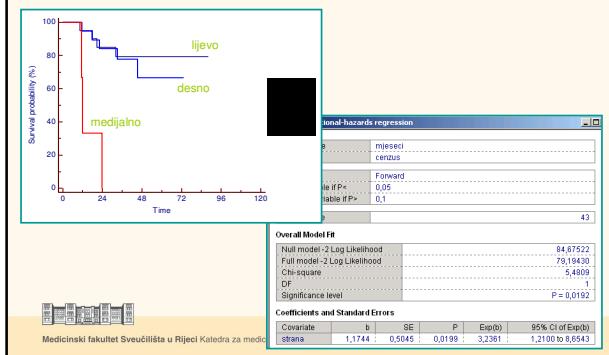
## Primjer, MedCalc® “svi pokazatelji”



## Primjer, MedCalc® “postupno biranje, unaprijed”



## Primjer, MedCalc® “postupno biranje, unaprijed”



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