25) Cox's Proportional Hazard Model

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Tables, Graphics, and Figures from:

1) https://lifelines.readthedocs.io/

Survival Regression

2) Cameron and Trivedi (2005)

Ch 17 - Transition Data: Survival Analysis

Standard Parametric Models

Parametric Model	Hazard Function	Survivor Function
Exponential	γ	$\exp(-\gamma t)$
Weibull	$\gamma \alpha t^{\alpha-1}$	$\exp(-\gamma t^{\alpha})$
Generalized Weibull	$\gamma \alpha t^{\alpha-1} S(t)^{-\mu}$	$[1-\mu\gamma t^{\alpha}]^{1/\mu}$
Gompertz	$\gamma \exp(\alpha t)$	$\exp(-(\gamma/\alpha)(e^{\alpha t}-1))$
Log-normal	$\frac{\exp(-(\ln t - \mu)^2 / 2\sigma^2)}{t\sigma\sqrt{2\pi}[1 - \Phi((\ln t - \mu)/\sigma)]}$	$1 - \Phi\left(\left(\ln t - \mu\right)/\sigma\right)$
Log-logistic	$\alpha \gamma^{\alpha} t^{\alpha - 1} / \left[(1 + (\gamma t)^{\alpha}) \right]$	$1/\left[1+(\gamma t)^{\alpha}\right]$
Gamma	$\frac{\gamma(\gamma t)^{\alpha-1} \exp[-(\gamma t)]}{\Gamma(\alpha)[1 - I(\alpha, \gamma t)]}$	$1 - I(\alpha, \gamma t)$



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Fully Parametric Analysis: MLE

$$Pr[T > t] = \int_{t}^{\infty} f(u|x,\theta) du = 1 - F(t|x,\theta) = S(t|x,\theta)$$
 $f(t_{i}|x_{i},\theta)^{\delta_{i}} S(t_{i}|x_{i},\theta)^{1-\delta_{i}}$
 $\delta_{i} \begin{cases} 1 & no & censoring \end{cases}$

$$\delta_i \begin{cases} 1 & no & censoring \\ 0 & right & censoring \end{cases}$$

$$Ln(\theta) = \sum_{i=1}^{N} \left[\delta_i Inf(t_i|x_i,\theta) + (1-\delta_i) \right] InS(t_i|x_i,\theta)$$

$$Ln(\theta) = \sum_{i=1}^{N} \left[\delta_i ln \lambda(t_i | x_i, \theta) + \Lambda(t_i | x_i, \theta) \right]$$

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Cox's Proportional Hazard Model

$$\lambda(t|x,\beta) = \lambda_0(t)\phi(x,\beta)$$

$$\lambda(t|x,\beta) = \lambda_0(t)\exp(\beta_1x_1 + \dots + \beta_kx_k)$$

$$\eta_i = \beta_1x_{i1} + \dots + \beta_kx_{ik}$$

$$\eta_j = \beta_1x_{j1} + \dots + \beta_kx_{jk}$$

$$\frac{\lambda_j(t|x,\beta)}{\lambda_j(t|x,\beta)} = \frac{\lambda_0(t)e^{\eta_j}}{\lambda_0(t)e^{\eta_j}} = \frac{e^{\eta_j}}{e^{\eta_j}}$$

McCall (1996)

Current Population Survey's Displaced Workers Supplements (DWS) for the years 1986, 1988, 1990, and 1992

spell: if person is re-employed at a full-time job

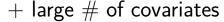
UI: the subject filed an unemployment claim or not

Replacement: weekly benefit amount divided by the amount of weekly earnings in the lost job

Disregard: the threshold amount up to which recipients of unemployment insurance who accept part-time work can earn without any reduction in unemployment benefits

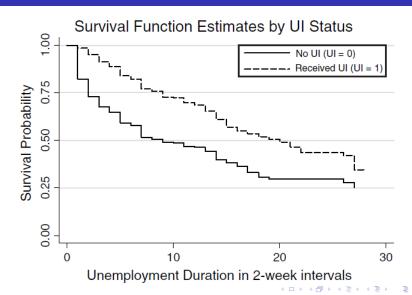
Key Economic Covariates to Explain Joblessness Duration

Variable Name	Variable Label				
spell	periods jobless: two-week interval	6.248			
CENSOR1	1 if reemployed at full-time job	0.321			
CENSOR2	1 if reemployed at part-time job	0.102			
CENSOR3	1 if reemployed but left job: pt-ft status unknown	0.172			
CENSOR4	1 if still jobless	0.375			
UI	1 if filed UI claim	0.553			
RR	eligible replacement rate	0.454			
DR	eligible disregard rate	0.109			
TENURE	tenure years in lost job	4.114			
LOGWAGE	log weekly earnings	5.693			



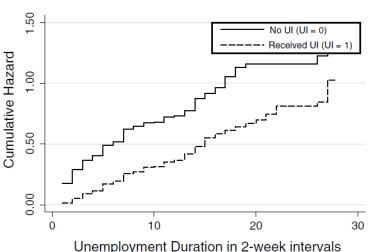


Survival Function: Unemployment Duration by Unemployment Insurance



Cumulative Hazard Functions: Unemployment Duration by Unemployment Insurance

Cumulative Hazard Estimates by UI Status



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Estimated Parameters

	Exponential		Weibull		Gompertz		Cox PH	
Var	coeff.	t	coeff.	t	coeff.	t	coeff.	t
RR	0.472	0.79	0.448	0.70	0.472	0.78	0.522	0.91
DR	-0.576	-0.75	-0.427	-0.53	-0.563	-0.74	-0.753	-1.04
UI	-1.425	-5.71	-1.496	-5.67	-1.428	-5.69	-1.317	-5.55
RRUI	0.966	0.92	1.105	1.57	0.969	1.58	0.882	1.52
DRUI	-0.199	-0.20	-0.299	-0.28	-0.211	-0.21	-0.095	-0.10
LOGWAGE	0.35	3.03	0.37	2.99	0.35	3.03	0.34	3.03
CONS	-4.079	-4.65	-4.358	-4.74	-4.097	-4.65	_	_
α			1.129					
−ln L	2700.7		2687.6		2700.6		_	

Positive state dependence ($\alpha=1.129>1$): the probability of the spell terminating increases as the spell lengthens

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Hazard Ratios

Var	Exponential		We	eibull	Gompertz		Cox PH	
	β	t	β	t	β	t	β	t
RR	1.603	0.63	1.565	0.57	1.604	0.62	1.686	0.71
DR	0.562	-1.02	0.653	-0.66	0.570	-0.99	0.471	-1.55
UI	0.241	-12.65	0.224	-13.12	0.240	-12.65	0.268	-11.53
RRUI	2.626	1.01	2.760	0.99	2.635	1.01	2.416	1.01
DRUI	0.819	-0.22	0.742	-0.33	0.810	-0.23	0.909	-0.10
LOGWAGE	1.420	2.56	1.441	0.08	1.42	2.55	1.40	2.57
α			1.129					
−ln L	2700.7		2687.6		2700.6		_	

Cox PH model has no intercept and makes no assumption about the shape of the baseline hazard

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Rossi et al. (1980)

Experimental Study of Recidivism

week: week of first arrest after release, or censoring time

arrest: 1 for those arrested during the period of the study

fin: financial aid was a randomly assigned

race: 1 for black

wexp: full-time work experience prior to incarceration

mar: individual was married

paro: individual was released on parole

prio: # of prior convictions

from lifelines.datasets import load_rossi

```
rossi\_dataset = load\_rossi()
rossi\_dataset.sample(9)
```

	week	arrest	fin	age	race	wexp	mar	paro	prio
308	52	0	0	19	1	1	0	1	3
129	52	0	0	22	0	0	0	1	4
93	52	0	1	24	1	0	0	0	1
261	52	0	0	22	1	0	0	0	1
103	49	1	1	19	1	0	0	1	1
243	52	0	0	20	1	1	0	1	1
211	52	0	1	30	1	0	0	1	1
410	27	1	0	20	0	1	0	0	1
40	52	0	1	33	1	1	1	0	9

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from lifelines import CoxPHFitter

```
cph = CoxPHFitter()
cph.fit(rossi dataset, duration col='week', event col='arrest')
cph.print summary()
n=432, number of events=114
      coef
           exp(coef) se(coef) z
                                   p lower 0.95 upper 0.95
fin -0.3794
             0.6843
                    0.1914 -1.9826 0.0474
                                        -0.7545
                                                 -0.0043
age -0.0574
             0.9442 0.0220 -2.6109 0.0090 -0.1006
                                                 -0.0143
             1.3688 0.3080 1.0192 0.3081 -0.2898 0.9176
race 0.3139
             wexp -0.1498
                                                  0.2662
    -0.4337
             0.3147
mar
paro -0.0849 0.9186 0.1958 -0.4336 0.6646 -0.4685
                                                  0.2988
prio 0.0915
             1.0958
                   0.0286 3.1938 0.0014
                                       0.0353
                                                  0.1476
            0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
```

Concordance = 0.640 Likelihood ratio test = 33.266 on 7 df, p=0.00002

cph.plot_covariate_groups('prio', [0, 5, 10, 15])

