ICPSR 2015 "Advanced Maximum Likelihood": Survival Analysis Day Seven

August 11, 2015

Separation

"Separation" = "perfect prediction"

Intuition:

$$\Pr(Y = 1 | X = 0) = ?$$

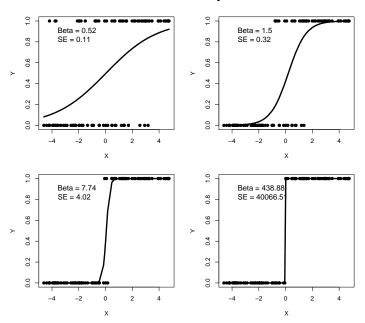
Separation: Effects

•
$$\hat{\beta}_X = \pm \infty$$

•
$$\widehat{\mathsf{s.e.}}_\beta = \infty$$

$$\bullet \left. \frac{\partial^2 \ln L}{\partial X^2} \right|_{\hat{\beta}} = 0$$

Separation Illustrated



Separation: What Happens

```
> summary(glm(Y~W+Z+X,family="binomial"))
Call:
glm(formula = Y ~ W + Z + X, family = "binomial")
Coefficients:
           Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.363
                    0.111 -3.26 0.00111 **
W
             0.424 0.119 3.57 0.00036 ***
Z
             -0.412 0.112 -3.67 0.00024 ***
X
             18.746 541.835 0.03 0.97240
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 684.41 on 499 degrees of freedom
Residual deviance: 464.69 on 496 degrees of freedom
ATC: 472.7
Number of Fisher Scoring iterations: 17
```

Separation: What Happens (Stata Remix)

. logit Y W Z X

```
note: X != 0 predicts success perfectly
    X dropped and 136 obs not used
```

```
Iteration 0: log likelihood = -245.53269
Iteration 1: log likelihood = -232.41173
Iteration 2: log likelihood = -232.34436
Iteration 3: log likelihood = -232.34436
```

Logistic regression

Number of obs

					[95% Conf.	_
W Z	.4242117	.1187869 .1123154	3.57	0.000	.1913936 6321628	.6570298
	03626348		-3.26	0.001	5805892	1446803

364

Solution (?): Exact Logistic Regression

- Cox (1970, Ch. 4); Hirji et al. (1987 JASA);
 Mehta & Patel (1995 Stat. Med.)
- Conditions on permutations of covariate patterns
- Always has finite solutions;
- Computational issues...

Firth's (1993) Correction

Firth proposed:

$$L(\boldsymbol{\beta}|Y)^* = L(\boldsymbol{\beta}|Y) |\mathbf{I}(\boldsymbol{\beta})|^{\frac{1}{2}}$$

$$\ln L(\boldsymbol{\beta}|\boldsymbol{Y})^* = \ln L(\boldsymbol{\beta}|\boldsymbol{Y}) + 0.5 \ln |\mathbf{I}(\boldsymbol{\beta})|$$

"Penalized likelihood":

- Consistent
- Eliminates small-sample bias
- Exist given separation
- Bayesians: "Jeffreys' prior"

Potential Drawbacks

- "Profile" (= "concentrated")
 likelihood
- $L(\hat{\beta})$ can be asymmetrical...
- ullet \rightarrow inference...

Software

- R
- elrm (exact logistic regression via MCMC)
- brlr ("bias-reduced logistic regression")
- logistf ("Firth's logistic regression")

- Stata
 - exlogistic (exact logistic regression)
 - firthlogit (Firth corrected logit)

Example: Pets as Family

- CBS/NYT Poll, April 1997
- Standard political/demographics, plus
- "Do you consider your pet to be a member of your family, or not?"
- Yes = 84.4%, No = 15.6%

Pets as Family: Data

```
> summary(Pets)
                                  married
  petfamily
                 female
Min.
       :0.00 Female:403
                          Divorced/Sep:118
                          Married
                                     :442
 1st Qu.:1.00 Male :321
Median:1.00
                                      :118
                          NBM
Mean
       :0.84
                          Widowed
                                      : 46
3rd Qu.:1.00
Max. :1.00
                        education
       partyid
           : 58
                 < HS
                            : 71
Democrat
           :224
                 College Grad: 131
GOP
           :228
                 HS diploma :244
 Independent:214
                 Post-Grad : 96
                 Some college: 182
```

Pets as Family: Basic Model

> Pets.1<-glm(petfamily~female+as.factor(married)+as.factor(partyid) +as.factor(education).data=Pets.familv=binomial) > summary(Pets.1) Call: glm(formula = petfamily ~ female + as.factor(married) + as.factor(partyid) + as.factor(education), family = binomial, data = Pets) Coefficients: Estimate Std. Error z value Pr(>|z|) (Intercept) 2.0133 0.5388 3.74 0.00019 *** femaleMale -0.6959 0.2142 -3.25 0.00116 ** as factor(married) Married -0.0657 0.2911 -0.23 0.82147 as.factor(married)NBM 0.4599 0.3957 1.16 0.24504 as.factor(married)Widowed 0.4921 -0.32 0.75007 -0.1568 as.factor(partyid)Democrat -0.1241 0.4286 -0.29 0.77213 as.factor(partyid)GOP -0.0350 0.4321 -0.08 0.93537 as.factor(partyid)Independent -0.1521 0.4299 -0.35 0.72338 as.factor(education)College Grad 0.2511 0.4121 0.61 0.54228 as.factor(education)HS diploma 0.0595 0.3685 0.16 0.87182 as.factor(education)Post-Grad 0.1946 0.4331 0.45 0.65321 as.factor(education)Some college 0.0587 0.3867 0.15 0.87928 Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 (Dispersion parameter for binomial family taken to be 1) Null deviance: 627.14 on 723 degrees of freedom Residual deviance: 612.76 on 712 degrees of freedom

Number of Fisher Scoring iterations: 4

ATC: 636.8



Pets as Family: More Complicated Model

```
> summarv(Pets.2)
Coefficients:
                                   Estimate Std. Error z value Pr(>|z|)
(Intercept)
                                     2.2971
                                                0.6166
                                                         3.73
                                                                0.0002 ***
femaleMale
                                    -1.1833
                                               0.5305
                                                       -2.23
                                                                0.0257 *
as factor(married) Married
                                               0.4470
                                                       -0.72
                                                                0.4716
                                    -0.3218
as.factor(married)NBM
                                     0.1854
                                                0.6140
                                                       0.30
                                                                0.7628
as.factor(married)Widowed
                                    -0.7415
                                               0.5780
                                                       -1.28
                                                                0.1995
as.factor(partvid)Democrat
                                    -0.1575
                                               0.4297
                                                        -0.37
                                                                0.7140
as.factor(partyid)GOP
                                    -0.0445
                                               0.4334
                                                        -0.10
                                                                0.9182
as.factor(partyid)Independent
                                    -0.1757
                                               0.4312
                                                        -0.41
                                                                0.6837
as.factor(education)College Grad
                                     0.2332
                                                0.4137
                                                         0.56
                                                                0.5730
as.factor(education)HS diploma
                                     0.0558
                                               0.3703
                                                         0.15
                                                                0.8801
as.factor(education)Post-Grad
                                               0.4342
                                                         0.50
                                                                0.6171
                                     0.2171
as.factor(education)Some college
                                     0.0358
                                               0.3890
                                                         0.09
                                                                0.9266
femaleMale:as.factor(married)Married
                                     0.4853
                                                0.5908
                                                         0.82
                                                                0.4114
femaleMale:as.factor(married)NBM
                                     0.5260
                                                0.8051
                                                         0.65
                                                                0.5136
femaleMale:as.factor(married)Widowed 15.2516
                                              549.3719
                                                         0.03
                                                                0.9779
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 627.14 on 723 degrees of freedom
Residual deviance: 607.42 on 709 degrees of freedom
ATC: 637.4
```

Number of Fisher Scoring iterations: 14

What's Going On?

```
> with(Pets, xtabs(~petfamily+as.factor(married)+female))
, , female = Female
         as.factor(married)
petfamily Divorced/Sep Married NBM Widowed
                            28
                    67
                           199
                                58
                                        32
, , female = Male
         as.factor(married)
petfamily Divorced/Sep Married NBM Widowed
                    11
                    33
                           168 47
```

Pets as Family: Firth Model

> Pets Firth

Model fitted by Penalized ML Confidence intervals and p-values by Profile Likelihood

	coef	se(coef)	lower 0.95	upper 0.95	Chisq	р
(Intercept)	2.1589	0.60	1.05	3.40	16.17636	0.000058
femaleMale	-1.1387	0.52	-2.19	-0.14	5.04186	0.024742
as.factor(married)Married	-0.2739	0.43	-1.19	0.53	0.41518	0.519353
as.factor(married)NBM	0.1589	0.59	-0.99	1.37	0.07322	0.786705
as.factor(married)Widowed	-0.7263	0.56	-1.84	0.38	1.67233	0.195947
as.factor(partyid)Democrat	-0.1182	0.42	-0.99	0.66	0.08159	0.775159
as.factor(partyid)GOP	-0.0078	0.42	-0.89	0.78	0.00034	0.985289
as.factor(partyid)Independent	-0.1364	0.42	-1.01	0.65	0.10813	0.742278
as.factor(education)College Grad	0.2390	0.40	-0.57	1.02	0.34480	0.557069
as.factor(education)HS diploma	0.0753	0.36	-0.67	0.76	0.04289	0.835933
as.factor(education)Post-Grad	0.2184	0.43	-0.63	1.05	0.26307	0.608019
as.factor(education)Some college	0.0524	0.38	-0.72	0.78	0.01888	0.890698
femaleMale:as.factor(married)Married	0.4558	0.58	-0.66	1.61	0.63550	0.425347
femaleMale:as.factor(married)NBM	0.5233	0.78	-1.02	2.05	0.45133	0.501702
femaleMale:as.factor(married)Widowed	2.4017	1.68	-0.14	7.37	3.37453	0.066212

Likelihood ratio test=17 on 14 df, p=0.24, n=724

Summary

- Separation → dropping covariates!
- Firth's approach > ELR
- Can also be applied to other sparse-data situations...

From the coxph Documentation

Convergence

In certain data cases the actual MLE estimate of a coefficient is infinity, e.g., a dichotomous variable where one of the groups has no events. When this happens the associated coefficient grows at a steady pace and a race condition will exist in the fitting routine: either the log likelihood converges, the information matrix becomes effectively singular, an argument to exp becomes too large for the computer hardware, or the maximum number of interactions is exceeded. (Nearly always the first occurs.) The routine attempts to detect when this has happened, not always successfully. The primary consequence for he user is that the Wald statistic = coefficient/se(coefficient) is not valid in this case and should be ignored; the likelihood ratio and score tests remain valid however.

Separation in Survival Data

```
> set.seed(7222009)
> X<-rep(0:1,times=100)
> T<-abs(rweibull(200,shape=1.2,
        scale=1/(exp(0+0.2*X)))
> C < -rbinom(200,1,0.2)
> C<-ifelse(X==0,0,C)
> table(C,X)
   χ
  0 100 81
      0 19
```

Cox Results

```
> cox.fit<-coxph(Surv(T,C)~X,method="efron")</pre>
Warning message:
In fitter(X, Y, strats, offset, init, control, weights = weights, :
 Loglik converged before variable 1; beta may be infinite.
> summary(cox.fit)
Call:
coxph(formula = Surv(T, C) ~ X, method = "efron")
 n= 200, number of events= 19
      coef exp(coef) se(coef) z Pr(>|z|)
X 2.038e+01 7.112e+08 5.630e+03 0.004 0.997
 exp(coef) exp(-coef) lower .95 upper .95
X 711225014 1.406e-09
                                     Tnf
Concordance= 0.761 (se = 0.064)
Rsquare= 0.137 (max possible= 0.583)
Likelihood ratio test= 29.37 on 1 df, p=5.994e-08
Wald test
                    = 0 on 1 df, p=0.9971
Score (logrank) test = 22.11 on 1 df, p=2.58e-06
```

Parametric Model = No Help

```
> weib.fit<-survreg(Surv(T,C)~X,dist="weibull")</pre>
> summary(weib.fit)
Call:
survreg(formula = Surv(T, C) ~ X, dist = "weibull")
              Value Std. Error
(Intercept) 61.5536 0.337 182.851 0.000
Х
           -60.1634 0.000 -Inf 0.000
Log(scale) -0.0184 0.190 -0.097 0.923
Scale= 0.982
Weibull distribution
Loglik(model) = -45.9 Loglik(intercept only) = -60.8
Chisq= 29.94 on 1 degrees of freedom, p= 4.5e-08
Number of Newton-Raphson Iterations: 9
n = 200
```

Firth \rightarrow Cox

Heinze and Schemper (2001):

$$\ln PL(\boldsymbol{\beta}|\boldsymbol{Y})^* = \ln PL(\boldsymbol{\beta}|\boldsymbol{Y}) + 0.5 \ln |\boldsymbol{I}(\boldsymbol{\beta})|$$

with
$$\mathbf{I}(\boldsymbol{\beta}) = -E\left[\frac{\partial^2}{\partial \theta^2} \ln PL(X, \boldsymbol{\beta}) \middle| \boldsymbol{\beta}\right].$$

Also, software: coxphf...

Firth-Corrected Cox

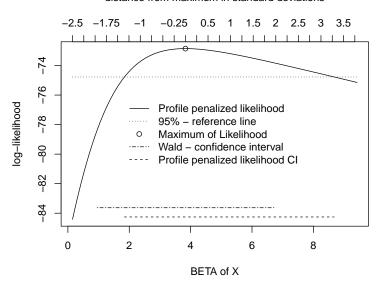
Likelihood ratio test=26.08802 on 1 df, p=3.262011e-07, n=200

> SIM<-cbind(T,C,X)
> SIM<-data.frame(SIM)</pre>

Examining the Profile Likelihood

Profile likelihood

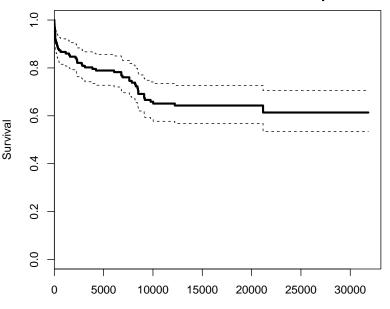
distance from maximum in standard deviations



Example: Lo et al. (2008)

- Outcome: "Cease-Fire Duration"
- Key covariate: Foreign-Imposed Regime Change ("FIRC")
- Data: Annual data on cease-fires, 1914-2001 (expanding Fortna 1998)
- Hypotheses:
 - · FIRCs \rightarrow more durable cease-fires
 - · Pacifying influence of FIRCs declines over time

Lo et al.: Kaplan-Meier



Time (in days)

From Lo et al. (2008)

Variables	Model 1 (ARCHIGOS data)	Model 2 (ARCHIGOS data)
FOREIGN-IMPOSED REGIME CHANGE	-161***	_
FIRC*ln(t)	(29.3) 16.8***	_
DIADDET FING	(3.03)	-161***
PUPPET-FIRC	_	(29.4)
PUPPET-FIRC* $ln(t)$	_	16.8*** (3.04)
CHANGE IN CAPABILITIES	.272	.274
BATTLE CONSISTENCY	(.376) 796**	(.376) 809**
2.11.122 001.01212.101	(.336)	(.342)

Simplified Cox Model

```
> LHR.Cox<-coxph(LHR.S~archigosFIRC+archigosFIRClnt,data=LHR,method="efron",
               iter.max=10000)
Warning message:
In fitter(X, Y, strats, offset, init, control, weights = weights, :
 Ran out of iterations and did not converge
> LHR. Cox
Call:
coxph(formula = LHR.S ~ archigosFIRC + archigosFIRClnt, data = LHR,
   method = "efron", iter.max = 10000)
                 coef exp(coef) se(coef) z
archigosFIRC -44.11 6.99e-20 20.92 -2.11 0.035
archigosFIRClnt 4.69 1.09e+02 2.24 2.09 0.036
Likelihood ratio test=20.8 on 2 df, p=3.04e-05 n= 6368, number of events= 54
```

Firth-Corrected Cox Model

```
> LHR.CoxF<-coxphf(LHR.S~archigosFIRC+archigosFIRClnt,data=LHR,maxit=1000)
> LHR.CoxF
coxphf(formula = LHR.S ~ archigosFIRC + archigosFIRClnt, data = LHR,
   maxit = 1000
Model fitted by Penalized ML
Confidence intervals and p-values by Profile Likelihood
                    coef se(coef) exp(coef) lower 0.95
archigosFIRC -55.591223 26.330771 7.19513e-25 4.808432e-65
archigosFIRClnt 5.848163 2.775927 3.46597e+02
                                                      NaN
               upper 0.95 Chisq
              NaN 9.738246 0.001804731
archigosFIRC
archigosFIRClnt 7320112 8.647141 0.003275750
Likelihood ratio test=15.09131 on 2 df, p=0.0005284014, n=6368
```

What Is Going On?

> table(LHR\$archigosFIRC,LHR\$X_d)

```
0 1
0 5265 52
1 1049 2
```

$\mathsf{Days} \to \mathsf{Years} = \mathsf{Little} \; \mathsf{Help}$

	Cox	Firth-Corrected
FIRC	-53.83	-21.08
	(51.30)	(9.99)
$FIRC \times In(T)$	14.57	5.85
	(13.54)	(2.78)
AIC	494.07	496
Num. events	54	54