```
EXST7034
```

Logistic regression Diagnostics

Regression Techniques

```
1
                                                              ***
2
          *** Logistic Regression - Disease outbreak example
          *** NKNW table 14.3 (Appendix C3)
                                                              ***:
3
          *** Study of a disease outbreak from a mosquito born
                                                             ***;
4
                                                              ***;
5
          *** disease within two sectors of a city.
          6
7
8
          dm'log;clear;output;clear';
9
          options nodate nocenter nonumber ps=512 ls=132 nolabel;
10
          ODS HTML style=minimal rs=none
body='C:\Geaghan\Current\EXST7034\Fall2005\SAS\DiseaseOutbreak01.html' ;
NOTE: Writing HTML Body file: C:\Geaghan\Current\EXST7034\Fall2005\SAS\DiseaseOutbreak01.html
11
          TITLE1 'Logistic Regression - NKNW Example 14.3';
12
13
          data Disease; infile cards missover;
14
             input case Age Status1 Status2 sector Disease;
15
             label case = 'case number'
16
                  age = 'Patients age'
17
                   status = 'Socioeconomic status upper, middle and lower'
18
                   disease = 'Disease present = 1';
          * Status classes are upper (0, 0), Middle (1, 0) and Lower (0, 1);
19
20
          Cards;
NOTE: Variable status is uninitialized.
NOTE: The data set WORK.DISEASE has 98 observations and 6 variables.
NOTE: DATA statement used (Total process time):
     real time
                 0.02 seconds
                       0.03 seconds
     cpu time
119
         ;
             ods html;
120
             ods graphics on;
121
NOTE: ODS Statistical Graphics will require a SAS/GRAPH license when it is declared production.
122
123
          proc logistic data=Disease DESCENDING alpha=0.05;
124
             TITLE2 'Logistic regression on Disease data';
125
             model Disease = Age Status1 Status2 Sector / lackfit RSQ iplots;
             output out=next1 PREDICTED=yhat Lower=lcl95 Upper=ucl95 dfbetas=_ALL_
126
                   resdev=resdev difdev=difdev;
127
128
          run;
NOTE: PROC LOGISTIC is modeling the probability that Disease=1.
NOTE: Convergence criterion (GCONV=1E-8) satisfied.
WARNING: Statistical graphics displays created with ODS are experimental in this release.
NOTE: There were 98 observations read from the data set WORK.DISEASE.
NOTE: The data set WORK.NEXT1 has 98 observations and 17 variables.
NOTE: At least one W.D format was too small for the number to be printed. The decimal may be
shifted by the "BEST" format.
NOTE: The PROCEDURE LOGISTIC printed page 1.
NOTE: PROCEDURE LOGISTIC used (Total process time):
                        4.77 seconds
     real time
     cpu time
                        3.12 seconds
Logistic Regression - NKNW Example 14.3
Logistic regression on Disease data
The LOGISTIC Procedure
              Model Information
                               WORK.DISEASE
Data Set
Response Variable
                               Disease
Number of Response Levels
                               2
Model
                               binary logit
Optimization Technique
                               Fisher's scoring
Number of Observations Read
                                       98
Number of Observations Used
                                       98
```

Re	esponse Profi				
Ordered		Tot	al		
Value	Disease	Frequenc	У		
1	1	3	1		
2	0	6	7		
Probability	modeled is I	Disease=1.			
	Model Conver	rgence Stat	us		
	Cor	nvergence c	riterion (GCONV=1E-8)	satisfied.
	Mod	del Fit Sta	tistics		
			Int	ercept	
		Intercep	t	and	
	Criterion	Onl	y Cova	riates	
	AIC	124.31	81	.11.054	
	SC	126.90	31	.23.979	
	-2 Log L	122.31	8 1	.01.054	
	R-Square	0.1950	Max-rescal	ed R-Square.	0.2736

Model fit statistics

1) Akaike Information Criterion $AIC = -2\log(L) + 2p$

where Log(L) is the log likelihood and p is the number of parameters

2) Schwarz Criterion
3) -2log L

$$SC = -2 \log(L) + p \log(\sum_{j} f_{j})$$

$$-2 \sum_{i=1}^{n} [Y_{i} \log_{e}(\hat{\pi}_{i}) + (1 - Y_{i}) \log_{e}(1 - \hat{\pi}_{i})]$$

This is analogous to the SSE in regression and is given in SAS as the "-2 Log L".

Two models (full and reduced) can be compared by calculating the difference in "-2 Log L" for both models. This difference follows a chi square distribution with a d.f. equal to the difference in d.f. for the two models.

4) Generalized R²
$$1 - \left(\frac{L(0)}{L(\theta)}\right)^{\frac{2}{n}}$$
, where L(0) is the intercept only model.

Since this value reaches its maximum of less than 1 for discrete models an adjustment has R^2

been proposed	This is called the Max-rescaled Requare in SAS	Λ
been proposed.	This is called the max-researed Requare in SAS.	$\overline{R_{\rm max}^2}$

Testing Glo	bal Null Hypothe	esis: BETA	A=0
Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	21.2635	4	0.0003
Score	20.4067	4	0.0004
Wald	16.6437	4	0.0023

Wald : used to test individual parameter estimates and to place confidence intervals. It is based on a large sample assumption of asymptotic normality.

Chi-square Test

 $\beta_i^2 / Var(\beta_i) = [\beta_i / Stderr(\beta_i)]^2$ $P\left(e^{(\hat{\beta}_i - 1.96S_{\hat{\beta}_i})} \le \beta i \le e^{(\hat{\beta}_i - 1.96S_{\hat{\beta}_i})}\right) = 0.95$

Confidence interval

Regression Techniques

	Analysi	is of Maxim	um Likeliho	od Estimates	
			Standard	Wald	
Parameter	DF I	Istimate	Error	Chi-Square	Pr > ChiSq
Intercept	1	-2.3127	0.6426	12.9545	0.0003
Age	1	0.0297	0.0135	4.8535	0.0276
Status1	1	0.4088	0.5990	0.4657	0.4950
Status2	1	-0.3051	0.6041	0.2551	0.6135
sector	1	1.5746	0.5016	9.8543	0.0017
	Odds Rati	lo Estimate	S		
	Point	9	5% Wald		
Effect	Estimate	Confi	dence Limit	S	
Age	1.030	1.00	3 1.0	58	
Status1	1.505	0.46	5 4.8	68	
Status2	0.737	0.22	6 2.4	08	
sector	4.829	1.80	7 12.9	07	
Associatio	n of Predi	icted Proba	bilities an	d Observed Res	sponses
Percent Co	ncordant	77.5	Somers' D	0.554	
Percent Di	scordant	22.1	Gamma	0.556	
Percent Ti	ed	0.3	Tau-a	0.242	
Pairs		2077	С	0.777	

Association of Predicted Probabilities and Observed Responses

Observations with different responses are paired and compared. If the one with the lower observed response has a lower predicted value it is said to be concordant. Otherwise they are discordant or tied. SAS reports concordant, discordant, ties and the number of pairs examined.

A number of other statistics are based on the same information of the number concordant (n_c) and the number discordant (n_d) . Where "t" is the total number of pairs with different responses and N is the sum of observation frequencies in the data then the following statistics can be derived. Note that ties are given by $t-n_c-n_d$. The statistic "c" is equal to $(n_c + 0.5(t-n_c-n_d))/t$. Somers' D is equal to $(n_c-n_d)/t$. The Goodman-Kruskal Gamma is $(n_c-n_d)/(n_c+n_d)$ and Kendall's Tau-a is $(n_c-n_d)/(0.5N(N-1)/t)$

	Partition	for the Ho	smer and Lem	eshow Test	
		Dise	ease = 1	Dise	ase = 0
Group	Total	Observed	Expected	Observed	Expected
1	10	0	0.79	10	9.21
2	10	1	1.02	9	8.98
3	11	2	1.51	9	9.49
4	10	1	1.78	9	8.22
5	10	3	2.34	7	7.66
6	10	4	3.09	6	6.91
7	10	7	3.91	3	6.09
8	11	3	5.51	8	5.49
9	10	5	6.32	5	3.68
10	6	5	4.75	1	1.25
Hosme	er and Lemesl	now Goodnes	s-of-Fit Tes	t	
Chi-S	Square	DF Pr	> ChiSq		
9	9.1871	8	0.3268		

Goodness-of-fit

Most Goodness-of-fit tests (Pearson and deviance) require replication in subpopulations. This is often a problem with continuous variables in the model. The Hosmer-Lemeshow Goodness-of-Fit Test can be used for sparser data. This test is only available for binary models.

In this approach the data are sorted on the basis of their response probability (default), and divided into approximately 10 groups (minimum = 3). See SAS help for details on the grouping.

Once in groups a Chi square statistic is calculated. For each group we have S_i as the observed number of "successes" in the group, n_i as the total number of observations in the group and $\overline{\pi_i}$ as the mean predicted probability in each group (from the model). For the "g" groups the Chi square statistic is then calculated as:

$$\chi^2 = \sum_{i=1}^{g} \frac{(S_i - n_i \overline{\pi}_i)^2}{n_i \overline{\pi}_i (1 - \overline{\pi}_i)}$$

The usual interpretations apply to this lack of fit test. Small values of P would indicate an inadequate model.

Deviance

The deviance in logistic regression can be calculated as

$$DEV(X_0, X_1, ..., X_{p-1}) = -2\sum_{i=1}^n \left[Y_i \log_e(\hat{\pi}_i) + (1 - Y_i) \log_e(1 - \hat{\pi}_i) \right]$$

Partial deviances can be calculated for reduced models and the difference in the two models (with n-p and n-p+diff d.f.) should follow a Chi square distribution with diff d.f. This test of partial deviances is also called the likelihood ratio test.

Deviance residuals

Residual analysis is not as simple with Logistic Regression as it was with Linear Regression. Since the

dependent variable is 0 or 1 the not normally distributed and in fact the distribution is not known. As a result residual analysis in Logistic Regression is done on "Deviance" residuals. These are calculated as:

$$dev_i = \pm \sqrt{\{-2[Y_i \log_e(\hat{\pi}_i) + (1 - Y_i) \log_e(1 - \hat{\pi}_i)]\}}$$

where the sign is + if $Y_i \ge \hat{\pi}_i$ and the sign is - if $Y_i < \hat{\pi}_i$. Note that the SS of these residuals will sum to the model deviance.

Pearson residuals

Given the residual $e_i = Y_i - \hat{\pi}_i$, the predicted probabilities $\hat{\pi}_i = Y_i - e_i$ and r_i = the number of events in a given observation with n_i trials, then the Pearson residual is given as:

$$\frac{r_i - n_i \hat{\pi}_i}{\sqrt{n_i \hat{\pi}_i (1 - \hat{\pi}_i)}}$$

Confidence interval displacement diagnostics

SAS provides two measures of confidence interval displacement as influence diagnostics, C for individual observations and CBAR for overall change in the parameter estimates when individual observations are removed.

EXST7034

Geaghan

Logistic regression Diagnostics

Logistic Regression - NKNW Example 14.3

Listing of one kept value for each by group from Logistic Reg

												D					
												F					
												в		D	D		
												Е		F	F	D	
												т		в	в	F	
												А		Е	Е	в	
												_	D	т	т	Е	
												I	F	А	А	т	
												n	в	_	_	А	
			s	s		D	_					t	Е	S	S	_	
			t	t	s	i	L				r	е	т	t	t	S	d
			a	а	е	s	Е		1	u	e	r	A	a	a	е	i
	С		t	t	С	е	v	У	с	С	S	с	_	t	t	С	f
0	a	А	u	u	t	а	Е	h	1	1	d	е	A	u	u	t	d
b	s	g	s	s	о	s	L	a	9	9	е	P	g	s	s	0	e
s	е	е	1	2	r	е	_	t	5	5	v	t	е	1	2	r	v
1	40	1	0	1	1	0	1	0.26630	0.09293	0.56251	-0.78696	-0.01939	0.09579	0.01031	-0.10891	-0.08726	0.65146
2	28	2	0	1	0	0	1	0.07187	0.02346	0.19976	-0.38621	-0.02297	0.02517	0.00406	-0.02016	0.01472	0.15105
3	37	3	1	0	1	0	1	0.44026	0.20277	0.70864	-1.07729	-0.03653	0.13065	-0.15334	-0.00929	-0.09777	1.23059
4	34	4	0	0	0	0	1	0.10031	0.03273	0.26867	-0.45980	-0.06299	0.03831	0.03101	0.03402	0.03349	0.21527
5	62	5	0	1	0	0	1	0.07805	0.02670	0.20716	-0.40315	-0.02281	0.02410	0.00395	-0.02219	0.01582	0.16460
6	3	6	0	0	0	0	1	0.10581	0.03573	0.27423	-0.47295	-0.06449	0.03745	0.03228	0.03561	0.03517	0.22774
7	91	7	0	1	0	0	1	0.08244	0.02904	0.21258	-0.41482	-0.02260	0.02318	0.00385	-0.02364	0.01660	0.17427
8	18	8	0	0	1	0	1	0.37751	0.18144	0.62395	-0.97368	-0.12774	0.11676	0.10849	0.08850	-0.07308	0.98811
9	14	9	0	0	0	1	1	0.11456	0.04062	0.28333	2.08167	0.51506	-0.27619	-0.26478	-0.29462	-0.29239	4.59669
10	24	11	1	0	1	0	1	0.49946	0.25967	0.73950	-1.17650	-0.00272	0.08844	-0.18022	-0.01563	-0.11255	1.46058
11	46	12	1	0	1	0	1	0.50690	0.26696	0.74370	-1.18915	0.00209	0.08229	-0.18374	-0.01651	-0.11446	1.49154
12	42	13	0	0	1	1	1	0.41304	0.21594	0.64260	1.32982	0.16963	-0.13835	-0.16158	-0.13291	0.11495	1.85198
13	25	14	0	1	1	0	1	0.34824	0.14743	0.62277	-0.92529	0.01835	0.05833	0.00400	-0.14741	-0.11664	0.89961
14	20	15	0	0	1	1	1	0.42754	0.23031	0.65085	1.30362	0.15429	-0.11829	-0.15489	-0.12785	0.11266	1.77482
15	23	16	0	0	1	1	1	0.43483	0.23757	0.65515	1.29058	0.14686	-0.10860	-0.15161	-0.12536	0.11150	1.73728
16	39	17	1	0	1	1	1	0.54397	0.30330	0.76572	1.10351	-0.02357	-0.04073	0.16920	0.01774	0.10412	1.27659
17	5	18	0	1	0	1	1	0.11082	0.04422	0.25134	2.09755	0.15482	-0.11485	-0.02231	0.26748	-0.17253	4.60704
18	96	19	0	1	0	0	1	0.11378	0.04577	0.25578	-0.49152	-0.01878	0.01312	0.00263	-0.03438	0.02201	0.24494
19	30	20	0	1	0	0	1	0.11682	0.04734	0.26041	-0.49845	-0.01822	0.01186	0.00247	-0.03545	0.02252	0.25194
20	21	21	1	0	1	1	1	0.57330	0.33160	0.78442	1.05485	-0.03820	-0.01341	0.16141	0.01876	0.09840	1.16292
21	16	22	0	0	1	1	1	0.47909	0.28142	0.68354	1.21315	0.10549	-0.05528	-0.13272	-0.11097	0.10445	1.52577
22	11	23	0	0	0	0	1	0.16403	0.06903	0.34176	-0.59860	-0.07437	0.01912	0.04453	0.05175	0.05261	0.36518
23	43	24	0	0	1	0	1	0.49395	0.29583	0.69399	-1.16715	-0.09070	0.03837	0.12371	0.10385	-0.09963	1.40958
24	81	25	1	0	0	0	1	0.23861	0.09829	0.47395	-0.73839	-0.03390	0.00011	-0.08578	0.01437	0.07126	0.56267
25	6	26	0	1	0	0	1	0.13653	0.05707	0.29233	-0.54184	-0.01385	0.00251	0.00128	-0.04256	0.02582	0.29819
26	13	27	0	0	0	0	1	0.18100	0.07840	0.36473	-0.63194	-0.07550	0.01099	0.04777	0.05627	0.05763	0.40740
27	70	28	0	0	0	0	1	0.18545	0.08079	0.37099	-0.64051	-0.07569	0.00867	0.04860	0.05745	0.05895	0.41866
28	68	30	0	0	0	0	1	0.19461	0.08559	0.38416	-0.65792	-0.07592	0.00366	0.05028	0.05987	0.06166	0.44206
29	8	31	1	0	0	1	1	0.27253	0.11409	0.52147	1.61246	0.06279	0.06227	0.27211	-0.03847	-0.21553	2.76692
30	22	32	1	0	1	1	1	0.65079	0.40153	0.83810	0.92689	-0.06762	0.04432	0.13975	0.02029	0.08316	0.89443

EXS Log	ST70. jistic 1	34 regre	ssio	n Di	iagn	osti	cs		Regressio	n Techniqu	es			Geaghan Page 6			
31	1	33	0	0	0	0	1	0.20898	0.09281	0.40556	-0.68473	-0.07590	-0.00487	0.05285	0.06363	0.06591	0.47951
32	2	35	0	0	0	0	1	0.21898	0.09759	0.42094	-0.70308	-0.07561	-0.01128	0.05460	0.06624	0.06888	0.50612
33	10	37	1	0	0	0	1	0.30931	0.12945	0.57423	-0.86030	-0.00925	-0.05360	-0.12060	0.01410	0.09116	0.77259
34	50	38	1	0	1	1	1	0.69019	0.43337	0.86648	0.86115	-0.07776	0.06620	0.12804	0.02044	0.07527	0.77207
35	57	39	0	0	1	0	1	0.60396	0.39037	0.78411	-1.36107	-0.02646	-0.08139	0.13312	0.11555	-0.12830	1.92752
36	49	40	1	0	1	1	1	0.70277	0.44296	0.87546	0.83992	-0.08033	0.07216	0.12419	0.02040	0.07271	0.73465
37	35	44	0	1	1	0	1	0.56602	0.28366	0.81117	-1.29210	0.19641	-0.15575	-0.02924	-0.26635	-0.20543	1.79985
38	69	46	0	0	0	0	1	0.28001	0.12255	0.51991	-0.81059	-0.06920	-0.05866	0.06465	0.08206	0.08725	0.67970
39	58	50	0	0	1	0	1	0.67901	0.43957	0.85086	-1.50754	0.04032	-0.19919	0.13590	0.12156	-0.15094	2.39781
40	84	51	0	1	0	0	1	0.24960	0.09562	0.51133	-0.75782	0.03299	-0.08525	-0.01037	-0.08886	0.04451	0.59705
41	74	52	0	0	0	1	1	0.31736	0.13470	0.58133	1.51506	0.13190	0.20421	-0.15128	-0.19742	-0.21258	2.45114
42	41	53	1	0	1	1	1	0.77682	0.49439	0.92531	0.71070	-0.08867	0.09691	0.10008	0.01925	0.05719	0.52764
43	65	59	0	1	0	1	1	0.29676	0.10555	0.60143	1.55875	-0.15028	0.32987	0.04183	0.26448	-0.12459	2.65895
44	4	60	0	0	0	0	1	0.37100	0.14908	0.66506	-0.96293	-0.04443	-0.15768	0.07794	0.10593	0.11613	0.98523
45	29	61	0	1	0	0	1	0.30932	0.10784	0.62395	-0.86032	0.07270	-0.15548	-0.01986	-0.11808	0.05478	0.78738
46	48	65	0	1	1	0	1	0.70895	0.35026	0.91671	-1.57116	0.40560	-0.43105	-0.07048	-0.36607	-0.27791	2.80759
47	17	67	0	0	1	1	1	0.77815	0.49016	0.92752	0.70829	-0.04972	0.12152	-0.03751	-0.03561	0.05310	0.52488
48	51	68	1	0	1	1	1	0.84467	0.53631	0.96236	0.58104	-0.08489	0.10238	0.07521	0.01660	0.04190	0.35400
49	44	70	0	0	1	1	1	0.79317	0.49700	0.93704	0.68076	-0.05250	0.12278	-0.03380	-0.03250	0.05013	0.48565
50	52	74	0	0	1	1	1	0.81201	0.50546	0.94806	0.64535	-0.05521	0.12315	-0.02928	-0.02867	0.04631	0.43731

122 proc sort data=next1 nodupkey; by Age; run;

```
NOTE: There were 98 observations read from the data set WORK.NEXT1.
NOTE: 48 observations with duplicate key values were deleted.
NOTE: The data set WORK.NEXT1 has 50 observations and 17 variables.
NOTE: PROCEDURE SORT used (Total process time):
                          0.01 seconds
     real time
      cpu time
                          0.01 seconds
123
           proc print data=next1;
124
              TITLE2 'Listing of one kept value for each by group from Logistic Reg';
125
           run;
NOTE: There were 50 observations read from the data set WORK.NEXT1.
NOTE: The PROCEDURE PRINT printed page 2.
NOTE: PROCEDURE PRINT used (Total process time):
     real time
                          0.21 seconds
      cpu time
                          0.07 seconds
126
127
           options ps=56 ls=111;
128
           proc sort data=Disease; by Age; run;
NOTE: There were 98 observations read from the data set WORK.DISEASE.
NOTE: The data set WORK.DISEASE has 98 observations and 6 variables.
NOTE: PROCEDURE SORT used (Total process time):
     real time
                         0.01 seconds
      cpu time
                          0.01 seconds
```

```
EXST7034
                                  Regression Techniques
                                                                              Geaghan
Logistic regression Diagnostics
                                                                                Page 7
129
           proc sort data=next1; by Age; run;
NOTE: Input data set is already sorted, no sorting done.
NOTE: PROCEDURE SORT used (Total process time):
      real time
                           0.00 seconds
      cpu time
                           0.00 seconds
           proc means data=Disease noprint; by Age; var Disease;
130
131
                output out=next3 n=n mean=mean var=var; run;
NOTE: There were 98 observations read from the data set WORK.DISEASE.
NOTE: The data set WORK.NEXT3 has 50 observations and 6 variables.
NOTE: PROCEDURE MEANS used (Total process time):
      real time
                           0.03 seconds
      cpu time
                           0.03 seconds
132
133
           data three; set next1 next3; run;
NOTE: There were 50 observations read from the data set WORK.NEXT1.
NOTE: There were 50 observations read from the data set WORK.NEXT3.
NOTE: The data set WORK.THREE has 100 observations and 22 variables.
NOTE: DATA statement used (Total process time):
                           0.01 seconds
      real time
                           0.01 seconds
      cpu time
134
           proc plot data=three; plot yhat*Age='x' mean*Age='o' / overlay;
                TITLE2 'Plot of observed means (o) and predicted values (p)';
135
136
           run;
138
              ods graphics off;
              ods html close;
139
NOTE: There were 100 observations read from the data set WORK.THREE.
NOTE: The PROCEDURE PLOT printed page 3.
NOTE: PROCEDURE PLOT used (Total process time):
                           0.24 seconds
      real time
                           0.07 seconds
      cpu time
```

Logistic Regression - NKNW Example 14.3 Plot of observed means (o) and predicted values (p)

> Plot of yhat*Age. Plot of mean*Age. Symbol used is Symbol used is '*': vhat 1.0 000 ο 0 0 00 ο 0 00 0 ο x 0.8 x х x v x х x x x 0 0.6 x x x x x o x x o o o 00 v xx 0.4 x x x x o x x x x x 0 0 x 0.2 x x 0 x xx хх xxx x х xx 0.0 00000 00 00 00 о o 0 000 0 0 00 00 10 70 20 30 4 Ó 50 6Ò 8Ò Age

NOTE: 100 obs had missing values.

A number of new graphics are available, but are still reported as "experimental". They are activated with ODS statements as follows.

120	ods	html;	
121	ods	graphics	on;

The graphics are then places in the HTML output.



EXST7034 Logistic regression Diagnostics

Geaghan Page 9

