The SAS RELRISK9 Macro

Sally Skinner, Ruifeng Li, Ellen Hertzmark, and Donna Spiegelman

November 15, 2012

Abstract

The %RELRISK9 macro obtains relative risk estimates using PROC GENMOD with the binomial distribution and the log link. This is particularly useful when the odds ratio is not a good approximation to the rate ratio (e.g., because of high prevalence of the outcome or large relative risks).

Keywords: SAS, macro, proc genmod, repeated measures, relative risk, initial values NOTE: do we want to have initial values as a keyword?

Contents

1	Des	scription	2				
2	Invo	ocation and Details	2				
3	Exa	amples	5				
	3.1	Example 1. Basic macro call using default parameters: One observation per subject	5				
	3.2	Example 2. <i>INITMETH</i> =none, demonstrating <i>DICHEXP</i> , <i>ESTIMATE1</i>	6				
	3.3	Example 3. <i>INITMETH</i> =force	9				
	3.4	Example 4. Multiple observations per subject, also showing <i>DICHEXP</i> and printing the results of the Poisson model (<i>POISPRINT</i> =T)	10				
	3.5	Example 5. Errors in calls to %RELRISK9	14				
4	Lim	litations	16				
5	Kno	own Bugs	16				
6	Credits						
7	Free	quently Asked Questions	16				
	7.1	Q: My output does not say that I am running a GEE model, but I have repeated measures (multiple observations per subject).	16				

8	Refe	erences	17
	7.4	Q: How can I get .rtf output?	17
	7.3	Q: I have tried a lot of models and they do not converge properly. What should I do?	17
	7.2	Q: My .saslog says 'Out of Memory.'	16

1 Description

 $\% \rm RELRISK9$ is a SAS macro that uses PROC GENMOD with the binomial distribution and the log link to compute relative risk estimates. Because the macro uses PROC GENMOD, it can handle repeated measures.

If the user specifies EMPCAL=T, the confidence intervals based on the empirical/robust estimates of the standard errors are given. Otherwise, model-based estimates are shown.

There are two main problems with this type of model: (1) it does not always converge; (2) the estimated probabilities derived from the model may not all be between 0 and 1. To help with convergence, the macro starts with initial values that are expected to be close to the "correct" values. This is done by running another procedure first, then using the results as initial values for the estimation procedure of proc genmod. Since we have found empirically that PROC GENMOD with the Poisson distribution and log link, using the emprical variance produces estimates fairly close to those that would be produced by PROC GENMOD with the binomial distribution and the log link, the default for the macro is to use this to obtain the initial values.

After the model is run, the macro checks whether all the predicted probabilities are below 1.

Error messages are printed if parameters or data do not conform to requirements. All error messages begin with "ERR'OR in macro". If any error occurs, the program terminates and no output is produced. If the ERROR message does not contain the word 'macro,' it is a SAS error, not a macro error.

2 Invocation and Details

In order to run this macro, your program must know where to look for it. You can tell SAS where to look for macros by using the options

This will allow you to use all the SAS read macros for the data sets as well as other public SAS macros, such as %LGTPHCURV9, %MIXCURV8, %PM, %INDIC3, %EXCLUDE, %MPHREG, %CALADJ, and %PCTL in /usr/local/channing/sasautos.

In the rest of this section, we will list all the input parameters, some of which are required and some of which are optional. If a parameter has a default value, it is given after the '='.

REQUIRED PARAMETERS

	====	
DATA =	1	the name of the input dataset
DEPEND =	1	the name of the dependent variable, for the left side of
	r	nodel equation in proc genmod
		This should be a binary variable coded
	() for 'no' and 1 for 'yes.'
INDEPEND	= 1	the list of independent variable(s), for the right
	5	side of the model equation in proc genmod
INITMETH	= P(DISSON the method for finding the initial values
		for each parameter
	Tł	nere are six choices:
	1	. POISSON: uses PROC GENMOD with the POISSON
		distribution.
	2	. NLIN: uses PROC NLIN to determine the initial values.
	3	. PREV: uses the previous estimates to initialize genmod.
		It can only be used if the previous run of RELRISK has
		converged and one or two variables have been replaced
		or added.
	4	. FORCE: allows the user to force either the intercept,
		the initial estimates, or both, using FINTERCEP
		and FINITVAL. Here is an example of the call:
		INITMETH=force, FINTERCEPT=-5, FINITVAL=-1 -1 2
		This sets the initial value of the intercept to -5 and
		the initial estimates
		for the first three variables in INDEPEND to -1, -1 and 2,
		respectively.
		One might want to use this if a previous attempt did not
		converge, or if it converged, but resulted in predicted
	_	probabilities greater than 1.
	5	. LUGISTIC: uses PRUC GENMUD with the logit link.
	6	. NUNE: goes directly to the estimation of
		the log binomial model.
		This is be useful (and a CPU-time-saver) if you know
		that a model will converge without problem to a set of
		legitimate values.
		Delault=Polsson
ΠΡΤΤΟΝΔΙ	PAR	AMETERS RELATING TO THE DATA
==========		

WHERE = A subsetting clause to restrict the data used

EXTRAVAR = Additional variables to be kept in the working data set, primarily for use in the WHERE parameter NOTE: If you subset using a variable not in the model and do not list the variable in EXTRAVAR, the macro will not keep the variable in the working dataset and the working dataset is likely to have 0 observations.

OPTIONAL PARAMETERS RELATING TO THE MODEL

- CLASS = A list of class/categorical variables used in the model. If EMPCAL is T, the SUBJECT will automatically be included in CLASS. NOTE: Use of CLASS variables can save typing time (and avoid errors), but, because of the possibly strange parameterization used for CLASS variables, it is advisable to use them only for covariates for which the parameter estimates are not of interest. For covariates whose parameter estimates are of interest, make indicators.
- EMPCAL = F Whether you want the empirical variance (GEE model)
- SUBJECT = The subject identifier used for SUBJECT= in the REPEATED
 statement (REQUIRED if EMPCAL=T)
- NOINT = F If T, no-intercept model is run. This cannot be run with INITMETH=NLIN
- REPTYPE = IND The covariance structure of the data, such as IND, EXCH, UN, MDEP)

If you have repeated measures, you probably want a covariance structure other than IND. Use of any covariance structure other than IND or EXCH

requires the specification of a WITHINVAR (see next item) to fit observations properly into the data structure.

WITHINVAR = the index variable for within-group records REQUIRED if EMPCAL=T and REPTYPE is not IND or EXCH. WEIGHT = The name of the weight variable OPTIONAL

OPTIONAL PARAMETERS RELATING TO THE OUTPUT

DICHEXP =	The name of the dichotomous exposure, if any of interest.
	If you give this a value, the macro will produce a table
	giving the numbers of subjects, observations, and cases,
	as well as the percent of cases for each value of the
	dichotomous exposure, and for the study group overall.
COV = F	If T, the covariance matrix is printed out and sent to ods
	with the name covb
CORR = F	If T, the correlation matrix is printed out and sent to ods
	with the name corrb
	Default=F
CONTRAST1 =	First contrast statement
CONTRAST2 =	Second contrast statement
CONTRAST3 =	Third contrast statement
CONTRAST4 =	Fourth contrast statement
ESTIMATE1 =	First estimate statement
ESTIMATE2 =	Second estimate statement
ESTIMATE3 =	Third estimate statement
ESTIMATE4 =	Fourth estimate statement
GMPRINT = F	If T, the macro will print whatever results the PROC GENMOD
	gets. This is useful if you are having convergence problems,

INITMETH=FORCE.

OUTDAT = The name of a sas data set to keep the results, if desired						
POISPRINT = F Whether to print the results of the Poisson model						
(if using INITMETH=POISSON)						
TITLENUM = 4 The first "free" title number, after the titles						
you already have and wish to keep during this procedure ,						
More titles have been added to make it easier to navigate						
through the output. In order to avoid having these titles						
knock out your titles, use TITLENUM to state what line						
the MACRO titles should begin in. The default is 4.						
For example, if you are using title1 and title2, TITLENUM						
should be set to 3.						

3 Examples

For the examples, we are using data from NHS II. The main dataset is N2DAT. In addition, we have dataset AGE18, which has one observation for each nurse and is primarily interested in the nurse's BMI at age 18. The outcomes of interest will be

OBESE	whether	the	nurse	has	bmi	ge	30					
OVERWT	whether	the	nurse	has	bmi	ge	25	(includes	ob	ese	e)	
OVWT18	whether	the	nurse	repo	orted	l be	eing	overweigh	nt	at	age	18

The predictors are

AGE group	as 5-year age group indicators; reference level
	is 25-29
BIRTH YEAR group	as 5-year cohorts; reference level is 1945-1949
	(macro variable name is &byrgp_)
REGION	as indicators for MIDWEST SOUTH WEST; reference
	level is NORTHEAST

3.1Example 1. Basic macro call using default parameters: One observation per subject

Here we model overweight at age 18 based on region and birth year group.

title2 'bare bones call (using default parameters) for one obs per subject'; %relrisk9(data=age18, depend=ovwt18, independ=midwest south west &byrgp_);

The macro output is

_____ /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:57 stleh 1 bare bones call (using default parameters) for one obs per subject

NOTE: 5259 observations deleted because of missing values for dependent, independent, subject, or within variables.

```
_____
```

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:57 stleh 2 bare bones call (using default parameters) for one obs per subject

Results of running RELRISK9 on data age18 Outcome is ovwt18 . 12100 events in 111412 observations (10.9 %).

		Standard	Р		Relative
Parameter	ESTIMATE	Error	value	UNITS	risk
Tatomoont	0 1160	0 0021	< 0001	1	0 10040
Intercept	-2.1109	0.0231	<.0001	T	0.12040
MIDWEST	-0.0369	0.0203	0.0698	1	0.96381
SOUTH	-0.1413	0.0249	<.0001	1	0.86823
WEST	-0.3544	0.0293	<.0001	1	0.70158
BYRGP2	-0.0044	0.0252	0.8616	1	0.99562
BYRGP3	-0.0376	0.0254	0.1399	1	0.96314
BYRGP4	-0.0283	0.0290	0.3295	1	0.97213
Scale	1.0000	0.0000	-	1	2.71828
95% CI	95% CI				
Low	high	RR (95% C	I)		
0.11507	0.12598	0.12 (0.12,	0.13)		
0.92616	1.00299	0.96 (0.93,	1.00)		

0.87 (0.83, 0.91)

0.662460.743010.70 (0.66, 0.74)0.947711.045951.00 (0.95, 1.05)0.916301.012380.96 (0.92, 1.01)0.918441.028960.97 (0.92, 1.03)2.718282.718282.72 (2.72, 2.72)

0.91173

The macro first printed out a notice, much like those printed by SAS procedures, saying how many observations were deleted because of missing values. The regression output is preceded by information on the name of the dataset, the outcome variable, the number and percent of events, and the number of observations.

The space between the lines beginning 'bare bones' and 'Results' on page 2 of the output results from the fact that *TITLENUM*=4, so there is no title3.

3.2 Example 2. INITMETH=none, demonstrating DICHEXP, ESTIMATE1

The macro call is

0.82680

```
title2 'overweight at age 18 by region and birth cohort';
title3 'initmeth=none';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south west &byrgp_,
dichexp=west,
estimate1=%str('south/west' south 1 west -1 / exp),
initmeth=none);
```

Note that we did not give values for the parameters that have default values, except for the one we wanted to change (*INITMETH*).

Although west is just one of the set of region indicators, formally it is a binary variable, and we can use it to demonstrate *DICHEXP*.

In the *ESTIMATE1* parameter, we used %str to avoid problems that might arise because of the quotation marks or the slash. It is especially important to do this when there is a comma in the 'name' of the estimate. We used exp to get the relative risk of south compared to west.

_____ /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:52 stleh 1 overweight at age 18 by region and birth cohort initmeth=none NOTE: 5259 observations deleted because of missing values for dependent, independent, subject, or within variables. /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:52 stleh 2 overweight at age 18 by region and birth cohort initmeth=none Results of running RELRISK9 on data age18 Outcome is ovwt18 . 12100 events in 111412 observations (10.9 %). Counting cases of ovwt18 by west Obs WEST NSNCASES PCTCASES 1 0 94615 10704 11.3 16797 8.3 2 1 1396 3 111412 12100 10.9 . /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:52 stleh 3 overweight at age 18 by region and birth cohort initmeth=none Results of running RELRISK9 on data age18 Outcome is ovwt18 . 12100 events in 111412 observations (10.9 %).

```
Standard P Relative
```

Parame	eter ESTI	MATE	Error	value	UNITS	risk		
Inter	cept -2.	1169	0.0231	<.0001	1	0.12040		
MIDWES	ST -0.	0369	0.0203	0.0698	1	0.96381		
SOUTH	-0.	1413	0.0249	<.0001	1	0.86823		
WEST	-0.	3544	0.0293	<.0001	1	0.70158		
BYRGP	2 -0.	0044	0.0252	0.8616	1	0.99562		
BYRGP	3 -0.	0376	0.0254	0.1399	1	0.96314		
BYRGP	4 -0.	0283	0.0290	0.3295	1	0.97213		
Scale	1.	0000	0.0000	-	1	2.71828		
95% (CI 95% C	I						
Low	high	ı R	R (95% C	CI)				
0.115	07 0.1259	0.12	(0.12.	0.13)				
0.926	16 1.0029	0.96	(0.93,	1.00)				
0.8268	80 0.9117	3 0.87	(0.83,	0.91)				
0.6624	46 0.7430	0.70	(0.66,	0.74)				
0.947	71 1.0459	5 1.00	(0.95,	1.05)				
0.9163	30 1.0123	0.96	(0.92,	1.01)				
0.9184	44 1.0289	0.97	(0.92,	1.03)				
2.718	28 2.7182	2.72	(2.72,	2.72)				
<pre>/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 12:52 stleh 4 overweight at age 18 by region and birth cohort initmeth=none Results of running RELRISK9 on data age18 Outcome is ovwt18 . 12100 events in 111412 observations (10.9 %).</pre>								
Obs 1	LABEL	MEA	NESTIMAT	'E MEANL	OWERCL M	EANUPPERCL	LBETAESTIMATE	
1 :	south/west		1.2375	1.	1602	1.3200	0.2131	
2 1	Exp(south/we	st)	-	-		-	1.2375	
Obs	STDERR	ALPHA	LBETALC	WERCL	LBETAUPPER	CL CHISQ	PROBCHISQ	
1	0.0329	0.05	0.1	.486	0.2776	41.93	<.0001	
2	0.0407	0.05	1.1	.602	1.3200	_	_	

The results of DICHEXP came before those of the model.

The results for ESTIMATE1 print last. The first line gives all the results you need. the 'MEAN' variables are the relative risk and its 95% confidence limits. the 'LBETA' variables are the estimate of the log(RR) and its 95% confidence bounds. PROBCHISQ is the p-value.

3.3 Example 3. *INITMETH*=force

Here are three (3) examples using *INITMETH*=force. The first gives initial values for all the effects. The second gives initial values for the intercept and the first effect. The third gives an initial value for the intercept only (and note that it is much lower than those in the first two macro calls). Note also that, since *INITMETH*=force just gives proc genmod a place to start, it is not necessary to be precise with the values used.

```
title3 'initmeth=force, all parameters initialized';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south west &byrgp_,
initmeth=force, fintercept=-5, finitval=.1 0 -.2 0 -.1 0);
```

```
title3 'initmeth=force, intercept and first parameter initialized';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south west &byrgp_,
initmeth=force, fintercept=-5, finitval=.1 );
```

```
title3 'initmeth=force, only intercept initialized ';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south west &byrgp_,
initmeth=force, fintercept=-10);
```

Since all three macro calls gave the same results, we show only one.

```
/udd/stleh/doctn/relrisk Program exrelrisk9 110CT2011 11:14
                                                          stleh
                                                                   1
initmeth=force, all parameters initialized
NOTE: 5259 observations deleted because of missing values for
    dependent, independent, subject, or within variables.
_____
/udd/stleh/doctn/relrisk Program exrelrisk9 110CT2011 11:14
                                                                   2
                                                          stleh
initmeth=force, all parameters initialized
Results of running RELRISK9 on data age18
Dutcome is ovwt18 . 12100 events in 111412 observations (10.9 %).
                     Standard
                                    Ρ
                                                Relative
Parameter
          ESTIMATE
                      Error
                                       UNITS
                               value
                                                 risk
Intercept
           -2.1169
                      0.0231
                               <.0001
                                         1
                                                0.12040
                      0.0203
MIDWEST
                               0.0698
           -0.0369
                                         1
                                                0.96381
SOUTH
           -0.1413
                      0.0249 <.0001
                                         1
                                                0.86823
WEST
                      0.0293
                               <.0001
           -0.3544
                                         1
                                                0.70158
BYRGP2
           -0.0044
                      0.0252 0.8616
                                         1
                                                0.99562
                      0.0254 0.1399
BYRGP3
           -0.0376
                                         1
                                                0.96314
                             0.3295
BYRGP4
           -0.0283
                      0.0290
                                         1
                                                0.97213
Scale
            1.0000
                       0.0000
                                         1
                                                2.71828
```

95% CI	95% CI	
Low	high	RR (95% CI)
0.11507	0.12598	0.12 (0.12, 0.13)
0.92616	1.00299	0.96 (0.93, 1.00)
0.82680	0.91173	0.87 (0.83, 0.91)
0.66246	0.74301	0.70 (0.66, 0.74)
0.94771	1.04595	1.00 (0.95, 1.05)
0.91630	1.01238	0.96 (0.92, 1.01)
0.91844	1.02896	0.97 (0.92, 1.03)
2.71828	2.71828	2.72 (2.72, 2.72)

The results of *DICHEXP* came before those of the model.

3.4 Example 4. Multiple observations per subject, also showing *DICHEXP* and printing the results of the Poisson model (*POISPRINT*=T)

In this example, we examine obesity over time in our longitudinal dataset, N2DAT. Since we have repeated measures, we need to use the *EMPCAL*, *SUBJECT*, and *REPTYPE* parameters. In addition, we get two contrasts for the global effects of the region and birth cohort variables, use a *WHERE* parameter, get the fit statistics, and show the effect of *TITLENUM*. Furthermore, we show the similarity/difference between the Poisson and the log-binomial estimates.

title2 'longitudinal obesity by region and birth cohort'; title3 'among those not overweight at age 18'; title4 'using where parameter, empcal=t'; title5 'showing use of contrast, fitprint, and titlenum parameters'; %relrisk9(data=n2dat, depend=obese, independ=midwest south west &byrgp_, empcal=t, reptype=exch, subject=id, fitprint=t, titlenum=7, poisprint=t, dichexp=west, contrast1=%str('effect of region' midwest 1, south 1, west 1), contrast3=%str('effect of birth cohort' byrgp2 1, byrgp2 1, byrgp4 1), where=ovwt18 eq 0);

Here are the results.

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:07 stleh 1
longitudinal obesity by region and birth cohort
among those not overweight at age 18
using where parameter, empcal=t
showing use of contrast, fitprint, and titlenum parameters

NOTE: 1615 observations deleted because of missing values for dependent, independent, subject, or within variables.

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:07 stleh 2 longitudinal obesity by region and birth cohort among those not overweight at age 18 using where parameter, empcal=t showing use of contrast, fitprint, and titlenum parameters

Results of running RELRISK9 on data n2dat Results of Poisson regression with empirical variance Outcome is obese . 53419 events in 451954 observations (11.8 %).

		Empirical					
		Standard					05% at
		Error	Р		relative	95% CI	95% CI
Parameter	ESTIMATE	Estimates	value	UNITS	risk	Low	high
Intercept	-1.8768	0.0186	<.0001	1	0.15308	0.14759	0.15878
MIDWEST	0.1942	0.0181	<.0001	1	1.21438	1.17212	1.25816
SOUTH	0.1586	0.0211	<.0001	1	1.17192	1.12440	1.22144
WEST	-0.1846	0.0250	<.0001	1	0.83142	0.79173	0.87309
BYRGP2	-0.1949	0.0193	<.0001	1	0.82292	0.79234	0.85469
BYRGP3	-0.5019	0.0207	<.0001	1	0.60537	0.58131	0.63042
BYRGP4	-0.8100	0.0264	<.0001	1	0.44484	0.42239	0.46849

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:07 stleh 3
longitudinal obesity by region and birth cohort
among those not overweight at age 18
using where parameter, empcal=t
showing use of contrast, fitprint, and titlenum parameters

Results of running RELRISK9 on data n2dat Outcome is obese . 53419 events in 451954 observations (11.8 %). where ovwt18 eq 0 Counting cases of obese by west

					Cases as
					a percent
Obs	WEST	NS	NOBS	NCASES	of obs
1	0	83893	382745	46841	12.2
2	1	15395	69209	6578	9.5
3		99288	451954	53419	11.8

_____ /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:07 stleh 4 longitudinal obesity by region and birth cohort among those not overweight at age 18 using where parameter, empcal=t showing use of contrast, fitprint, and titlenum parameters Results of running RELRISK9 on data n2dat Outcome is obese . 53419 events in 451954 observations (11.8 %).where ovwt18 eq 0 Obs CRITERION VALUE QIC 1 323667.0955 2 QICu 323634.8028 _____ /udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:07 stleh 5 longitudinal obesity by region and birth cohort among those not overweight at age 18 using where parameter, empcal=t showing use of contrast, fitprint, and titlenum parameters Results of running RELRISK9 on data n2dat Outcome is obese . 53419 events in 451954 observations (11.8 %).where ovwt18 eq 0 Empirical Standard Р Error Parameter ESTIMATE Estimates value UNITS Intercent -1 8765 0 0187 < 0001 1

Intercept	-1.0705	0.0107	<.0001	T
MIDWEST	0.1940	0.0181	<.0001	1
SOUTH	0.1575	0.0211	<.0001	1
WEST	-0.1847	0.0250	<.0001	1
BYRGP2	-0.1951	0.0193	<.0001	1
BYRGP3	-0.5018	0.0207	<.0001	1
BYRGP4	-0.8098	0.0264	<.0001	1

Relative	95% CI	95% CI	
risk	Low	high	RR (95% CI)
0.15313	0.14763	0.15883	0.15 (0.15, 0.16)
1.21404	1.17175	1.25786	1.21 (1.17, 1.26)

1.17055	1.12303	1.22008	1.17	(1.12,	1.22)			
0.83137	0.79166	0.87308	0.83	(0.79,	0.87)			
0.82277	0.79220	0.85453	0.82	(0.79,	0.85)			
0.60544	0.58139	0.63049	0.61	(0.58,	0.63)			
0.44494	0.42248	0.46860	0.44	(0.42,	0.47)			
/udd/stl longitud among th using wh showing	eh/doctn/relr inal obesity i ose not overwe ere parameter use of contra	isk Program by region an eight at age , empcal=t st, fitprint	n exrel nd birt e 18 t, and 1	risk9 h coho: titlen	12JUN rt um para	N2012 13:07	stleh	6
Results Outcome where ov	of running RE is obese . 5 wt18 eq 0	LRISK9 on da 3419 events	ata n2da in 4519	at 954 ob:	servati	ions (11.8 %).		
Obs	CONTRAS	Г	DF	(CHISQ	PROBCHISQ	TYPE	
1 e	ffect of regi	on	3	3:	26.46	<.0001	Score	

The space between the lines beginning 'showing' and 'Results' comes from the fact that TI-TLENUM=7.

95.80

<.0001

Score

1

The first part of the output (page 2) tells you that it is based on the Poisson estimates (because we set POISPRINT=T). The second part (page 4) is based on the log-binomial estimates. Note that when there are multiple observations per subject, DICHEXP produces a table based on observations, not subjects. The results of DICHEXP printed before the log-binomial results.

Below we do a trend test on birth cohort using the ordered categorical variable. We set TI-TLENUM=6 to avoid writing over the titles we gave before the macro call.

title2 'longitudinal obesity by region and birth cohort'; title3 'among those not overweight at age 18'; title4 'using where parameter, empcal=t'; title4 'trend test for birth cohort'; %relrisk9(data=n2dat, depend=obese, independ=midwest south west bygp, empcal=t, reptype=exch, subject=id, titlenum=6, where=ovwt18 eq 0);

Here are the results.

2

effect of birth cohort

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:10 stleh 1
longitudinal obesity by region and birth cohort
among those not overweight at age 18

trend test for birth cohort

NOTE: 1615 observations deleted because of missing values for dependent, independent, subject, or within variables.

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 13:10 stleh 2 longitudinal obesity by region and birth cohort among those not overweight at age 18 trend test for birth cohort

Results of running RELRISK9 on data n2dat Outcome is obese . 53419 events in 451954 observations (11.8 %). where ovwt18 eq O

		Empirical		
		Standard		
		Error	Р	
Parameter	ESTIMATE	Estimates	value	UNITS
Intercept	-1.5734	0.0216	<.0001	1
MIDWEST	0.1938	0.0181	<.0001	1
SOUTH	0.1575	0.0211	<.0001	1
WEST	-0.1838	0.0250	<.0001	1
BYGP	-0.2678	0.0075	<.0001	1

Relative	95% CI	95% CI	
risk	Low	high	RR (95% CI)
0.20733	0.19873	0.21630	0.21 (0.20, 0.22)
1.21382	1.17156	1.25761	1.21 (1.17, 1.26)
1.17055	1.12303	1.22009	1.17 (1.12, 1.22)
0.83211	0.79236	0.87386	0.83 (0.79, 0.87)
0.76503	0.75380	0.77644	0.77 (0.75, 0.78)

3.5 Example 5. Errors in calls to %RELRISK9

In this macro call, a required parameter has been omitted;

%relrisk9(data=n2dat, independ=midwest south west);

The diagnostics in the .log and the .lst are

```
/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 14:13 stleh 1
ERR''OR in macro call: The name of the dependent variable was omitted.
The RELRISK9 macro stopped.
_____
In the macro call below, a variable is misspelled.
title2 'a variable name misspelled';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south wes);
The diagnostics in the .log and the .lst are
      _____
/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 14:16 stleh 1
a variable name misspelled
ERR''OR in macro run:
      An error occurred during the creation of the working dataset,
      while keeping DEPEND and INDEPEND from age18
      The names of your variables may be incorrect.
        dependent vbl is ovwt18 .
       predictors are midwest south wes .
The RELRISK9 macro stopped.
Finally, we show an example of setting EMPCAL=T without giving a value for SUBJECT.
title2 'empcal=t without specifying a subject variable';
%relrisk9(data=age18, depend=ovwt18, independ=midwest south west,
empcal=t);
The diagnostics in the .log and the .lst are
  _____
/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 14:17 stleh
                                                                 1
empcal=t without specifying a subject variable
ERR''OR in macro call: You asked for a GEE model (empcal=T),
   but did not specify a variable to serve as the subject (SUBJECT).
   The macro will run a non-GEE model (with REPTYPE=IND)
_____
```

/udd/stleh/doctn/relrisk Program exrelrisk9 12JUN2012 14:17 stleh 2
empcal=t without specifying a subject variable

NOTE: 5259 observations deleted because of missing values for dependent, independent, subject, or within variables.

The diagnostics in the .log and the .lst are

4 Limitations

These models do not always converge. They are especially likely to have trouble for very large data sets, very high overall prevalences of the outcome, large numbers of covariates, or continuous covariates.

5 Known Bugs

At the moment, *INITMETH*=prev and *INITMETH*=nlin do not work. *INITMETH*=poisson and *INITMETH*=force should be adequate substitutes.

6 Credits

This program was written by Sally Skinner from Tufts University and modified by Ruifeng Li and Ellen Hertzmark for the Channing Laboratory. The documentation was written by Ruifeng Li and Ellen Hertzmark. Questions can be directed to Ruifeng Li, strui@channing.harvard.edu, 617-432-6321.

7 Frequently Asked Questions

7.1 Q: My output does not say that I am running a GEE model, but I have repeated measures (multiple observations per subject).

A: You probably did not specify EMPCAL=T.

7.2 Q: My .saslog says 'Out of Memory.'

A: Run your job with extra memory by submitting with the following shell command: qbs -q l -o memsize=2048M program name

7.3 Q: I have tried a lot of models and they do not converge properly. What should I do?

A: Indicators that represent small numbers of observations can be a problem. If you have these, either combine categories (if that makes sense) or delete the observations in the small categories (this may make sense if these are 'missing' categories). You may also need to pare down your models to the essential variables. If you can get those running, you can try adding one of the other variables at a time, and using the *INITMETH*=FORCE option.

If you have made a serious effort and you still don't have acceptable results, you may use the Poisson model with the empirical variance. *INITMETH*=POISSON, *POISPRINT*=T, *EMPCAL*=T. The macro will still try to fit a log-binomial model, but before it does, it will print the results of the Poisson model.

7.4 Q: How can I get .rtf output?

A: Make an output dataset OUTDAT=name of output dataset you want. Then ods rtf file='name of .rtf to make'; proc print data=name of output dataset; ods rtf close;

Note: you might prefer to make a permanent sas dataset in outdat, so you can determine which variables you want to print and in which order.

8 References

References for using the log-binomial model are

```
AUTHOR = Wacholder, Sholom
TITLE = Binomial regression in GLIM: Estimating risk ratios and risk
differences
JOURNAL = American Journal of Epidemiology
VOLUME = 123
PAGES = 174-184
YEAR = 1986
```

Skov T, Deddens J, Petersen MR, Endahl L. Prevalence proportion ratios: estimation and hypothesis testing. International Journal of Epidemiology 1998:27;91-95.

In the methods section for a cross-sectional study, you write

Prevalence ratios and 95% confidence intervals were estimated by binomial regression with the log link function.

If you are using log-Poisson models with the empirical covariance structure because some of the log-binomial models did not converge, then use the following reference, in addition to the other two.

Zou G. A modified Poisson regression approach to prospective studies with binary data. Am J Epidemiol 2004; 159: 702-706.

The methods section should read

Log-binomials models were used to estimate relative risks and confidence intervals (ref). In a few instances, the models didn't converge and log-Poisson models, which provide consistent but not fully efficient estimates of the relative risk and its confidence intervals, were used (ref Zou paper).