CONSTRUCTING WEIGHTS TO USE IN ANALYZING PAIRS OF INDIVIDUALS FROM ADD HEALTH DATA

Kim Chantala Carolina Population Center University of North Carolina at Chapel Hill

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Investigating certain research questions may involve linking the responses of one adolescent to those of another adolescent interviewed in the Add Health Survey. The goal of the analysis will be to characterize the behavior of a*pair of adolescents* rather than the behavior of an *individual adolescent*. Unfortunately, the available sampling weights are appropriate only when the analysis seeks to investigate behaviors of individual adolescents. However, an appropriate sampling weight can be constructed for each pair of adolescents and used to correct for clustering and unequal probability of selection of the pair. This paper covers the types of pairs of adolescents that can have a pair weight constructed, describes the necessary formulas and data, and concludes with an example showing how to compute pair weights for romantic partners from the Wave II In-home Interview.

Types of Linked Pairs of Adolescents

The most common types of linked pairs in the Add Health data are:

- Friends: Respondents and the friends they listed (available from the In-School, Wave I and Wave II In-home Interviews).
- Couples: Respondents and the romantic partners they listed (available from both the Wave I and Wave II In-home Interviews).
- Genetic Pairs: Siblings and unrelated pairs who live in the same household.

Most of the pairs in the friends and couples categories were obtained serendipitously as part of the probability sample (figure 1). Pairs that are in the probability sample have the needed information to compute a pair weight.



Figure 1. Schematic showing overlap of the sample collected for use in making national estimates (Probability Sample) and the sample collected for genetic analysis (Genetic Sample).

There is no information available to compute sampling weights for pairs of adolescents that are not included in the probability sample. These pairs are additional siblings who did not attend any of the 132 schools selected from the sampling frame and were interviewed to increase the sample size for genetic analysis. The number of pairs of friends or couples that cannot have pair weights computed is small and should not affect having sufficient sample size for analyses. However, 35.83% of the pairs in the genetic sample do not have weight information for both members of the pair (table 1). This is a large enough percentage that you may decide to do your analysis without any correction for unequal probability of selection of genetic pairs. Since your results may not be nationally representative unless you use sampling weights, a thorough description of the characteristics of the genetic sample would be very helpful to someone wanting to generalize your results to their own group of interest.

		Number without	
Type of Pair in Genetic	Number of	Weight	Percentage without
Sample	Pairs	Information	Weight Information
Full-Siblings	1,251	227	18.15%
Half-Siblings	442	175	39.59%
Twins	784	212	27.04%
Non-Related living together	662	511	77.19 %
Total	3,139	1,125	35.83%

Table 1. The percent of pairs that do not have information for constructing pair weights.

Linking Data for Pairs of Adolescents

Adolescents were asked to identify their friends and partners using the combined rosters from the high school and feeder school. The Add Health identification number of the friend (or partner) was saved in the data set. To link the records of two adolescents, you will need to merge the data from the respondent with the data collected from their friend (or partner) to create a data set as illustrated in Table 2. The respondent with AID=923 nominated five friends, but only two of the friends were interviewed. Similarly, only one of the friends nominated by AID=952 was interviewed. Only these three of the eight friendship pairs are available to be in the data set used to describe the target population of teenage friends.

AID	AGE	PVT	School ID	GSWGT2	Friend AID	Friend Age	Friend PVT	Friend School ID	Friend GSGWT2
923	16	106	96	1043	943	17	102	96	98
923	16	106	96	1043	891				
923	16	106	96	1043	419				
923	16	106	96	1043	160	16	110	196	5068
923	16	106	96	1043	331				
952	14	98	196	211	922				
952	14	98	196	211	221	16	95	196	32
952	14	98	196	211	887				

Table 2. Example of data representative of Add Health friendship pairs from Wave II.

Computing Pair Weights

The first step is to compute an initial pair weight. This can produce some pairs of adolescents whose computed weights will be extremely large, thereby increasing the variability of the weights and any estimates computed from them. To minimize the variance and bias of estimates, a trimming procedure is used to limit the value of extreme weights. This is an iterative procedure that allows you to find the optimal value for trimming the weights and construct a desired set of pair weights. To understand the details of this technique, pair weights will be computed for romantic partners from the Wave II In-Home Survey.

Formulas for Pair Weights

There are two basic formulas for computing the initial pair weights. Appendix A shows how the formulas were derived. The only information needed for these formulas is the sampling weight of each adolescent (*WEIGHT*) and the sampling weight for the school from which they were sampled (*SCHOOLWT*). The formula used if the pair of adolescents attended the same school differs from the formula used when they came from a high school and the associated feeder school. The formulas are as follows:

1) The Pair of Adolescents (*i* and *j*) were sampled from the same feeder or high school (*k*):

$$PAIRWT_{i,j} = \frac{WEIGHT_i * WEIGHT_j}{SCHOOLWT_i}$$

2) The Pair of Adolescents (*i* and *j*) were sampled from a high school and the associated feeder school:

$$PAIRWT_{i,j} = \frac{WEIGHT_i * WEIGHT_j}{SCHOOLWT_{High_School}}$$

The primary sampling unit becomes the High School. This means that to correct for clustering of respondents, the students that were sampled from a feeder school will be assigned to the associated high school.

Data Used in the Pair Weight Formulas

Use SCHWT1, the Add Health sampling weight variable for the school attended, in place of *SCHOOLWT* in the previous two formulas. You may need to request this variable from the Add Health data manager. The variables you need to use in place of WEIGHT_i and WEIGHT_j in the previous formulas are the variables for each adolescent's sampling weight that you normally use in analysis for individuals. These are listed in Table 3. Notice that you will use the adolescents' weight variable from the data set you used to link the records.

Table 3. Variable names for an adolescent's sampling weight in each data set.

Linked records are from:	Use this variable for WEIGHT:
In-School Data	SCHWGTPS
Wave I In-home Data	GSWGT1
Wave II In-home Data	GSWGT2

Example 1. In the probability sample, there are 983 adolescents from Wave II who nominated 1067 opposite-sex partners who were also interviewed at Wave II. These relationships can be broken down as shown in Table 4. The couples have been divided into two groups: respondents whose partners also nominated them (reciprocating partners) and respondents whose partners did not nominate them (non-reciprocating

partners). While there were 1,067 nominated relationships, there are only 865 unique couples. This is because the 202 boys with their reciprocating partners make up the same couples as the 202 girls with their reciprocating partners.

	Wave II Romantic Couples	Number of Pairs	Number of Unique Pairs
Non-reciprocating partners	Girls and boyfriends who did not nominate the girl as a partner	300	663
	Boys and girlfriends who did not nominate the boy as a partner	363	
Reciprocating partners	Girls and boyfriends who nominated the girl as a partner	202	202
	Boys and girlfriends who nominated the boy as a partner	202	
Total		1067	865

Table 4. Wave II couples.

The task is to use the pair weight formulas to construct sampling weights for the 865 unique couples from the Wave II In-home Survey.

A data set has been constructed linking the identification numbers of the school(s) attended, the schools' weight, and the grand sample weights from Wave II for the respondent and the partners they nominated. Table 5 shows what the data might look like for three of these couples.

Table 5. Hypothetical data for three couples.

	School			Partner	Partner	Partner	Partner
AID	ID	SCHWT1	GSWGT2	AID	School ID	SCHWT1	GSGWT2
923	96	68	1043	943	96	68	98
923	96	68	1043	160	196	124	5068
952	196	124	211	221	196	124	32

Substituting the values of GSWGT2 for WEIGHT in the equations yields the following values for constructed weights:

PAIRWT_{923,943} =
$$\frac{1043 * 98}{68}$$
 = 1,503.15
PAIRWT_{923,160} = $\frac{1043 * 5068}{68}$ = 77,734.18
PAIRWT_{952,221} = $\frac{211 * 32}{124}$ = 54.45

All three pairs would be assigned to the primary sampling unit (PSU) corresponding to school 96 (the high school). Notice that the large individual weights for PAIRWT_{923,160} have created a very large pair weight.

Using these same formulas, initial pair weights have been computed for the 865 unique couples from Wave II. Here are the descriptive statistics for these initial pair weights computed using PROC UNIVARIATE in SAS:

UNI VARI ATE	STATISTICS	ON INITIAL	PAIR WEIGHTS
	Ν	865	
	Mean	10933.56	
	Std Dev	30393. 84	
	Vari ance	9. 2379E8	
	Sum	9457532	
	PERCENT	ILE VALUES	
	100% Max	672945.8	
	75% Q 3	9507.123	
	50% Med	3378. 266	
	25% Q1	700. 0378	
	0% Mi n	20. 06924	
	99% 1	110294. 9	
	95%	47011.17	
	90% 2	28557.53	
	10% 1	159. 5563	
	5%	84. 89495	
	1% 4	40. 42081	

Each pair weight estimates the number of teenage couples represented by the pair we interviewed and ranges from 20.1 to 672,945.8 couples. The sum of the pair weights

means that our sample represents an estimated 9,457,532 teenage couples from the U.S. The average pair weight is 10,933 and the median value is 3,378. This indicates an extremely skewed distribution, as shown in Figure 2.



Figure 2. Box plot and frequency chart showing the distribution of initial pair weights.

The box-plot at the left of the figure shows the range and quartiles of the distribution. From this diagram it is easy to see than most of the pair weight values are small 25% are below 660 and 50% are below 3,378. The interquartile range contains all the values between the 25^{th} (value of 660) and the 75^{th} percentile (value of 9,507).

The top bar shows there are 71 initial pair weights with values between 33,000 and 672,946. All the other bars are in increments of 1,000.

Procedure for Trimming the Initial Pair Weights

This section gives an overview of a procedure that can be used to find a set of adjusted weights that minimizes both the variance and the bias in estimates (Potter, 1990 and Potter, 1993). The details of how to perform each step will be illustrated using the couples initial pair weights computed in the previous example. This procedure can be broken down into the following steps.

- Select a set of values based on percentile cut-points of the initial pair weights that will be used to trim the weights. For example, you may select the values from the 80th percentile to the 99th percentile. Each value will be used to create a set of adjusted weights.
- 2. Using each cut-point, truncate the initial pair weights at the cut-point and redistribute the excess sample weight uniformly over all pairs below the cut-point to create a set of *adjusted weights*. The excess sample weight is the difference between the sum of the untrimmed weights and the sum of the trimmed weights. To redistribute the excess weight uniformly, divide the excess weight by the number of untrimmed weights, and add this value to the initial weights below the cut-point.
- 3. Select a set of Add Health data items to use in evaluating which set of adjusted weights is the best to use.
- 4. Using each set of adjusted weights, compute the Mean Square Error (MSE) for each data item as follows:

BIAS = ESTIMATE Initial Wt - ESTIMATE Adjusted Wt

 $MSE = Variance_{Adjusted Wt} + BIAS^2$

The estimate can be a mean, a proportion, or a regression parameter.

- 5. For each selected data item, rank the MSEs across all adjusted weights.
- 6. For each set of adjusted weights, compute the average of the ranks. Choose the set of adjusted weights with the smallest value of average rank MSE to use in subsequent analyses.

The wisdom of choosing the smallest rank of the MSE will become clear in the example that follows.

Example 2. Use the trimming procedure to determine a set of adjusted pair weights that will give minimum bias and variance for estimates.

Step 1. Choose values for trimming the initial weights. We will need to choose enough values so that we observe the average rank of the mean square error decrease and then increase again. For this example, we will use the 70^{th} to 99^{th} percentiles of the

initial pair weight distributions as cut-points for trimming. Table 6 shows these values as computed using PROC UNIVARIATE in SAS.

Percent	Cut-point	Percent	Cut-point	Percent	Cut-point
70%	7,134.34	80%	12,638.35	90%	28,557.53
71%	7,475.42	81%	13,614.24	91%	31,107.67
72%	7,941.88	82%	14,622.08	92%	33,257.42
73%	8,470.67	83%	15,423.71	93%	37,879.85
74%	9,100.29	84%	16,718.79	94%	43,522.64
75%	9,507.12	85%	17,973.99	95%	47,011.17
76%	10,526.52	86%	19,914.61	96%	53,162.82
77%	11,421.12	87%	21,736.53	97%	67,640.99
78%	11,978.12	88%	22,914.57	98%	76,869.29
79%	12,198.29	89%	25,804.63	99%	110,294.94

Table 6. Percentile cut-points for trimming Initial pair weights for Wave II couples.

Here we see that 99% of the pair weights are below 110,295; 85% are below 17,974; and 70 percent are below 7,134.

Step 2. Compute a set of adjusted weights for each of the 30 cut-points in Table 6. Show the calculations for the five smallest and five largest initial pair weights using the 99th percentile cut-point.

The entries in Table 7 show this calculation for the couples with the five smallest and five largest initial pair weight values. The calculations for the other 855 (see p. 10) unique couples are not listed in the table, but the values for these couples have been included in the computations in row b through row e.

- a) Set all initial weights over the value of the 99% cut-point to the corresponding value of 110,294.94 (Table 7, column a).
- b) Sum the initial weights and trimmed weights (row b).
- c) Determine the excess weight by subtracting the sum of the trimmed weights from the sum of the initial weights (row c).
- d) Determine how many of the initial weights did not need to be trimmed (row d).
- e) Compute the amount that needs to be added (row e) to each of the untrimmed weights by dividing the excess weight computed in row c by the number of initial weights that did not need to be trimmed listed in row d.
- f) Add this amount to each of the trimmed weights (columna) to compute the adjusted weights (column f). Note that weights that needed to be trimmed in column a have remained unchanged in column f.

The sum of the initial weights is equal to the sum of the adjusted weights. This means that the average weight does not depend on the trimming level and remains at 10,933.56.

Table 7. Calculation the set of adjusted pair weights trimmed at the 99^{h} % cut-point using procedure on page 9.

		Calculation for 9	9 th % cut-point	
	Initial Pair Weight	a) Trim at 99 th % (110,294.94)	f) Adjusted weight 99 th %	Notes
	20.07	20.07	977.95	957.88 added to each
The five	20.81	20.81	978.69	weight under the 99 th % cut-
smallest	20.81	20.81	978.69	point value of 110,294.94
Initial weights	22.70	22.70	980.58	
	22.86	22.86	980.74	
	127,941.5	110,294.94	110,294.94	The initial pair weights over
The five	132,343.6	110,294.94	110,294.94	the 99 th % cut-point were set
largest	134,268.0	110,294.94	110,294.94	to 110,294.94 (the value of the 99 th % cut-point)
Initial weights	279,515.3	110,294.94	110,294.94	
	672,945.8	110,294.94	110,294.94	
	•	•	•	
855 not listed	•	•	•	
b) Sum of weights; includes 855 weights not listed (see p. 9)	9,457,532	8,637,591	9,457,532	The sum of the initial weights is equal to the sum of the adjusted weights.
c) Excess weight (subtract trimmed from initial)		9,457,532 <u>- 8,637,591</u> 819,941		
d) # out of 865 initial weights not trimmed (see p. 9)		856		
e) Amount to be added to untrimmed weights		819,941 / 856 = 957.88		

Repeat this process for all other cut-points in Table 6. The resulting distributions for each of the 30 sets of adjusted weights are shown in Figure 3.



Figure 3. Box-plots for each set of adjusted weights.

Step 3. Select several Add Health variables to use in evaluating the adjusted weights.

The choice of data items should represent the possible relationships between the weights and the data expected in the full set of data items. It is important to choose a set of variables where most of the continuous variables do not have values of zero, because a value of zero may mask the influence of an extreme weight (Potter, 1990, page 225). Some variables should also be related to the weights because a negative correlation between the weights and the data can imply that the variation in the weight is good (reduces the sampling variance). However, your data may not be correlated with the weights if the sampling design did not have much over-or under-sampling of populations. For the Add Health data, we will select both continuous and binary variables. For continuous variables we will want to estimate the mean value and for binary variables we will estimate proportions.

Table 8 shows the variables that were selected. The first six variables are from the section of the Wave II questionnaire that asked respondents to select events in their relationship from a set of cards. These are binary variables with a value of 1 indicating the card was selected and a value of 0 indicating the card was refused. A t-test showed

there was no association between the initial pair weights for respondents who selected the card versus those that did not select the card. The binary variable BLACK showed a strong association with the initial pair weights. The last six are continuous variables constructed from various Add Health data items. Pearson correlation coefficients indicated only a small association between the sampling weights and these variables and ranged from -0.04 to 0.05.

Variable Name	Description
INLOVE	Card H: Say In love, 1=selected, 0=not selected
THNKCOUP	Card I: Think of as a Couple, 1=selected, 0=not selected
BC_STD	Card J: Talk about contraceptive/STD, 1=selected, 0=not selected
KISS	Card K: Kissed, 1=selected, 0=not selected
TOUCH	Card L: Touch Each Other Under Clothes, 1=selected, 0=not selected
FONDLE	Card N: Touch Each Others Genitals, 1=selected, 0=not selected
BLACK	Black Race, 1=Yes, 0=No
LONGDATE	How Long Relationship has lasted, 0.2 to 145 months
BMI	Body Mass Index, 15 to 45 kg/(m*m)
GPA	GPA from English, Math, History, Science, 1=D to 4=A
GOODLOOK	Attractiveness, 1=Very Bad, 5=Very Good
DEPRESD2	Depress Score - Ave of Depression questions, 0 to 3
PVT_PER1	ADD Health PVT Percentile Rank, 1% to 100%

Table 8. Variables selected to use in evaluating the adjusted weights.

The data items we have chosen are available for each individual. Since each person in a reciprocated partnership reported the relationship, each can be included in this procedure to find the optimal set of adjusted weights. Hence, we will use the response from each adolescent reporting the relationship for the 1,067 couples for the computations in the remaining steps.

Steps 4-5. Use each set of adjusted weights to compute and rank the MSE for each selected variable.

The computation will be shown for the BMI (Body Mass Index) variable using the adjusted weight computed at the 85^{th} % trim level. Using a package designed for survey data analysis with the initial pair weight as our sampling weight gives an estimate of 22.388 as the mean value of BMI. Using the adjusted weight computed at the 85^{th} % trim level gives an estimated mean of 22.605 and standard error of 0.17105 for BMI.

The bias is:

The mean square error can then be computed as:

MSE = VARIANCE + BIAS²
=
$$(STD ERR)^2$$
 + BIAS²
= $(0.17105)^2$ + $(-0.217)^2$
= $0.0.076$ (step 4).

Using BMI, repeat this computation for each set of adjusted weights. The results are displayed in the second column of Table 9.

Examining the formula for MSE shows why it is a good criterion to use in evaluating the effect of the adjusted pair weight. The MSE is composed of two elements crucial to testing any hypothesis: the BIAS and the VARIANCE. A small BIAS indicates our estimate is "close" to the true value. A small variance guarantees that in repeated sampling a higher fraction of our estimates will be "close" to the true value of what we are trying to measure. This means that the smaller the variance the more certain we are that we have estimated the true value and have more evidence from our data to support our hypothesis. We can select the optimal set of adjusted weights for each variable by finding the set that gives us the smallest MSE for all of the data items.

The next step is to rank the values of MSE across all of the adjusted weights as shown in the third column of Table 9. The smallest MSE computed used the adjusted weight for the 81^{st} % cut-point (assigned a rank of 1) and the largest MSE was computed using the adjusted weight for the 98^{th} % cut-point (assigned a rank of 30). The value we computed using the adjusted weight at the 85^{th} % cut-point has a rank of 6.

Adjusted Weight at cut-point	Step 6) MSE for BMI	Step 7) Rank of MSE
70	0.0896	22
71	0.0872	19
72	0.0890	21
73	0.0846	18
74	0.0822	17
75	0.0800	13
76	0.0795	12
77	0.0807	14
78	0.0770	7
79	0.0717	2
80	0.0748	4
81	0.0717	1 – smallest
82	0.0787	11
83	0.0776	9
84	0.0815	16
85	0.0761	6 ¬ example
86	0.0783	10
87	0.0774	8
88	0.0729	3
89	0.0760	5
90	0.0814	15
91	0.0879	20
92	0.0914	23
93	0.0923	24
94	0.0948	27
95	0.0946	26
96	0.0955	28
97	0.0973	29
98	0.0982	30 – largest
99	0.0940	25

Table 9. MSE computed using each of the adjusted weights and rank of MSE for BMI.

Steps 4 and 5 should be repeated for all of the other variables in Table 8. The resulting MSE computations (omitted here) for the other variables show that there is a wide range of values of MSE. For example, the values of MSE for the TOUCH variable range from 0.0004 to 0.0021 while those for the PVT variable range from 5.75 to 14.74. By ranking the MSE values for each set of variable values, the results from one variable will not dominate over the influence of any of the others.

Step 6. Average the ranks of the variables for each of the adjusted weights.

The ranks for all of ourseleted variables are listed in Table 10. The last column is the average of the ranks for each row and can be used to compare the overall effect of each of the 30 adjusted weights. Examining this column shows that the smallest average rank is for the set of adjusted weights at the 85th % cut-point. This is the set that can be used in any analyses on romantic pairs from Wave II.

	Ranking of MSE values													
Adjusted weight cut-point	BC_STD	BLACK	BMI	DEPRESD 2	FONDLE	GOODLOOK	GPA	INLOVE	E KISS	LONGDATE	PVT_PER1	THNKCOU	P TOUCH	Ave. Rank
70	24	30	22	20	20	30	30	1	22	29	30	23	21	23.23
71	28	29	19	22	14	29	29	2	27	28	29	22	19	22.84
72	30	28	21	26	6	28	28	3	29	25	28	20	18	22.31
73	29	26	18	24	4	27	27	4	30	23	27	21	17	21.31
74	26	27	17	27	5	26	26	5	26	13	26	19	15	19.85
75	27	25	13	25	2	24	25	6	28	12	25	17	14	18.69
76	25	24	12	28	1	25	24	7	24	15	23	12	10	17.69
77	23	23	14	29	3	23	23	13	25	16	24	14	3	17.92
78	22	22	7	30	15	22	22	9	20	9	21	9	5	16.38
79	20	21	2	23	12	21	21	14	21	8	22	11	2	15.23
														İ
80	21	20	4	21	13	20	20	12	23	7	20	8	1	14.62
81	18	19	1	19	19	19	19	8	19	4	19	6	9	13.76
82	17	18	11	18	18	17	18	10	16	3	18	3	8	13.46
83	13	17	9	17	16	16	17	11	18	2	17	4	6	12.54
84	10	16	16	15	11	18	16	15	17	5	16	1	4	12.31
														İ
85	8	15	6	12	9	15	15	16	14	6	15	2	7	10.77
86	6	14	10	11	8	14	14	18	13	24	14	5	11	12.46
87	4	13	8	14	10	13	13	23	12	22	13	7	13	12.69
88	7	12	3	16	7	12	12	26	11	21	12	10	12	12.38
89	2	11	5	13	17	11	11	24	10	18	11	13	16	12.46
90	3	10	15	10	21	7	10	21	9	17	10	15	20	12.92
91	1	9	20	9	22	6	9	19	7	14	9	16	22	12.54
92	5	8	23	8	23	4	8	17	5	11	8	18	23	12.38
93	9	7	24	7	24	2	7	22	4	19	7	24	24	13.85
94	11	6	27	6	25	1	6	25	3	27	6	25	25	14.85
95	12	5	26	5	26	3	5	28	2	30	4	26	26	15.23
96	14	4	28	4	27	5	4	27	1	26	5	27	27	15.31
97	16	3	29	3	28	8	3	30	6	20	3	28	28	15.76
98	15	2	30	2	29	9	2	29	8	10	2	29	29	15.08
99	19	1	25	1	30	10	1	20	15	1	1	30	30	14.15

Table 10. Ranks for MSE computed using all 30 adjusted weights for each variable.

Table 11 shows the changes in the estimates for the 13 selected data items when computed with the initial weights versus the optimal adjusted weights trimmed at the 85th % cut-point. The proportion of couples reporting "in love" changed from 0.55 to 0.63 and the length of time dating changed from eight to nine months. The largest percent change occurred in the proportion of black adolescents reporting relationships. The estimate computed with the adjusted weight, 0.16, matches the population estimate of black adolescents computed with the full Wave II In-home sample. All other changes were very small.

		Estimate from		
Variable Name	Estimate from Initial Weight	Adjusted Weight at 85 th % Cut-point	Difference in Estimates	Percent Change in Estimate
BC_STD	0.31	0.32	-0.02	-4.93
BLACK	0.10	0.16	-0.06	-38.58
BMI	22.39	22.60	-0.22	-0.96
DEPRESD2	0.54	0.56	-0.02	-3.91
FONDLE	0.52	0.50	0.02	3.99
GOODLOOK	3.59	3.69	-0.11	-2.93
GPA	2.97	2.88	0.09	2.99
INLOVE	0.55	0.63	-0.08	-11.97
KISS	0.82	0.84	-0.02	-2.47
LONGDATE	8.00	9.28	-1.28	-13.75
PVT_PER1	58.55	57.04	1.51	2.65
THNKCOUP	0.78	0.79	-0.01	-1.37
TOUCH	0.56	0.55	0.01	1.49

Table 11. Estimates from initial weights compared to estimated from adjusted weight at the 85th % cut-point.

The improvement in the design effect (DEFF) is shown in Table 12 for the thirteen data items used to evaluate the adjusted weights. The DEFF is a ratio of the variance obtained under our complex survey design compared to the variance that would have been obtained if the data had been collected through simple random sampling. The DEFF from the initial pair weights averages 11.8 and ranges from 3.5 to 20.7.

Using the adjusted weights at the 85th % cut-point reduced the average DEFF to 3.5 with the range going from 1.0 to 13.9. Two data items (GPA and BLACK) had an increase in DEFF. By computing the ratio of the values in the two DEFF columns, we can see how much larger the variance is if the initial pair weights are used. In general, the improvement in variance is substantial. The variance of estimates from nine of the thirteen data items computed with the initial weights would be over twice as large as the variance of estimates computed with the adjusted weight at the 85th %. The variance for the proportion of couples reporting being in love would be 21 times larger if the initial weights rather than the optimum adjusted weight were used.

Variable Name	DEFF _{Initial} =	DEFF _{Adjusted} = DEFF from Adjusted	
	DEFF from Initial Weight	Weight at 85 ^{°°} % Cut-point	DEFF _{Initial} / DEFF _{Adjusted-}
BC_STD	5.1	1.8	2.8
BLACK	11.8	13.9	0.8
BMI	8.3	2.1	4.0
DEPRESD2	4.4	2.8	1.6
FONDLE	16.6	2.1	7.8
GOODLOOK	17.2	3.1	5.5
GPA	3.5	4.8	0.7
INLOVE	20.7	1.0	21.3
KISS	13.0	1.7	7.8
LONGDATE	7.9	1.7	4.7
PVT_PER1	10.0	8.1	1.2
THNKCOUP	19.2	1.1	17.3
TOUCH	15.3	1.5	10.3

Table 12. Comparison of design effect for estimates computed from initial pair weights to adjusted weights at 85th % cut-point.

Effects of Using the Best Adjusted Weight vs. the Initial Weight

To understand how the sampling weight can affect the results, we will compare the proportion of respondents who reported sexual relations for reciprocating versus non-reciprocating partners. The proportions have been reported separately for girls and boys in table 13. The standard errors are larger when the initial weight is used in the analysis. Using the initial weight finds no difference in proportions for girls having reciprocating versus non-reciprocating partners (p=0.12) while the adjusted weights reject the hypothesis of no difference ($p \le 0.01$).

Table 13. Proportion of boys and girls reporting sexual relations with partner, by reciprocating partner status.

		Adjusted weight at
Group	Initial Weight*	85 ^{°°} % cut-point*
Non-Reciprocating Girlfriend	0.25 (0.043)	0.28 (0.030)
Reciprocating Girlfriend	0.61 (0.121)	0.47 (0.046)
Difference in proportions	0.36 (0.130)	0.19 (0.052)
t-test for no difference	t=2.78, p≤0.01	t=3.71, p≤0.01
Non-Reciprocating Boyfriend	0.17 (0.041)	0.22 (0.028)
Reciprocating Boyfriend	0.37 (0.120)	0.48 (0.043)
Difference in proportions	0.20 (0.130)	0.26 (0.045)
t-test for no difference	t=1.58, p=0.12	t=5.69, p≤0.01
	Group Non-Reciprocating Girlfriend Reciprocating Girlfriend Difference in proportions t-test for no difference Non-Reciprocating Boyfriend Reciprocating Boyfriend Difference in proportions t-test for no difference	GroupInitial Weight*Non-Reciprocating Girlfriend $0.25 (0.043)$ Reciprocating Girlfriend $0.61 (0.121)$ Difference in proportions $0.36 (0.130)$ t-test for no difference $t=2.78, p \le 0.01$ Non-Reciprocating Boyfriend $0.17 (0.041)$ Reciprocating Boyfriend $0.37 (0.120)$ Difference in proportions $0.20 (0.130)$ t-test for no difference $t=1.58, p=0.12$

* Value in parenthesis is the standard error.

For boys with reciprocating partners, a higher proportion are estimated to have had sexual relations with their partner when computed with the initial weights (0.61) rather than with the adjusted weights (0.47). Conversely, a lower proportion of girls with reciprocating partner are estimated to have had sexual relations with their partners when computed with the initial weights (0.37) versus the adjusted weights (0.48). Since the boys with reciprocating girlfriend are the same couples as the girls with reciprocating boyfriends, these proportions are comparing the responses from the male partner to those from the female partner. Hence, we would expect these proportions to be approximately the same as suggested by the estimates computed with the set of adjusted weights at the 85th % cut-point.

Discussion

Adjusting the weights has been shown to reduce the variance and give a more believable estimate for the example in the previous section. This best adjusted set of weights has been constructed for doing a global analysis on the 865 relationships reported. Analyses can also be done on several sub-populations, such as the 202 reciprocating partners, the 502 girls and boyfriends, as well as the 565 boys and girlfriends.

If the researcher had indicated that the analysis of a certain group was more important than global analysis, more emphasis might have been placed on finding the optimal set of weights for this group. For example, there may be little interest in analyzing the202 reciprocating partners. In this case, the 502 girls and their boyfriends might always be analyzed separately from the 565 boys and their girlfriends. In this case, an optimal set of adjusted weights could be found for girls and boyfriends and the process repeated for boys and girlfriends. This was done for the Wave II couples and the optimal set of pair weights was still found to be that adjusted at the 85^{h} % cut-point for both of these sub-populations.

Appendix A. Derivation of Pair Weight Formulas

This appendix shows the derivation of the formulas used to calculate pair weights.

Available Data

School Weights

For each school in the probability sample we know the final school weight, SCHWT1. For high schools this weight is the inverse of the probability of selecting the high school:

SCHWT1 = $[Pr{high school}]^{-1}$

For feeder schools it is the inverse of the joint probability of selecting the feeder school and high school:

SCHWT1 = [Pr{feeder school, high school}]⁻¹

Adolescent Weights

Sampling weights are available for everyone in the probability sample for the various panels of data:

PANEL OF		
DATA	WEIGHT	MEANING
In-School	SCHWGTPS	Final In-school Weight
Wave I	GSWGT1	Final Grand Sample Weight for Wave I participants
Wave II	GSWGT2	Final Grand Sample Weight for Wave II participants

Assume the probability of selecting one adolescent in the pair is not influenced by the selection of the other adolescent.

For adolescents from a high school (including high schools with an 8th grade) the weight represents the inverse of the joint probability of selecting the student and selecting the high school:

WEIGHT = [Pr{student, high school}]⁻¹

For adolescents from a feeder school the weight represents the inverse of the joint probability of selecting the student, the feeder school and the associated high school:

WEIGHT = [Pr{student, feeder school, high school}]⁻¹

Note that GSWGT1 for Wave I was computed as the average of a set of weights. The set of weights for each adolescent was computed from the different samples for which the adolescent was eligible.

COMPUTATION OF PAIR WEIGHTS

Students from the Same School

The weight for any pair of adolescents is the inverse of the probability of selecting both adolescents and their schools into the sample. For two adolescents denoted by and j from the same school k:

$$\Pr \{kid_i, kid_j, school_k\} = \Pr \{kid_i \mid school_k\} * \Pr \{kid_j \mid school_k\} * \Pr \{school_k\}$$
(1)

By assuming the joint probability of selecting the kid and the school can be represented by:

and substituting the result into equation 1 gives:

$$Pr \{kid_i, kid_j, school_k\} = [WEIGHT_i * Pr\{school_k\}]^{-1} * [WEIGHT_j * Pr\{school_k\}]^{-1} * Pr\{school_k\}]^{-1} * Pr\{school_k\}$$

=
$$[WEIGHT_i * WEIGHT_j * Pr{school_k}]^{-1}$$

= SCHWT1_k* $[WEIGHT_i * WEIGHT_j]^{-1}$

So the pair weight for two students from the same school will be:

$$PAIRWT_{i,j} = WEIGHT_i^* WEIGHT_j / SCHWT1_k$$
(2)

Students from High School and Associated Feeder School

Let the adolescent from the feeder school be denoted by f and from the high by h. The joint probability of selecting adolescent f from feeder school f and adolescent h from high school h is:

Pr {kid_f, kid_h, school_f, school_h}

= Pr\$chool_f, school_h} * Pr{kid_f | school_f} * Pr{kid_h | school_h} (3)

The first term following the equality sign is the joint probability of both the feeder and high school combination and is multiplied by the product of the conditional probability that each adolescent was selected given that his or her school was selected.

Note for kids from feeder schools:

 $WEIGHT_{f}^{-1} = Pr\{ kid_{f}, school_{f}, school_{h} \} = Pr\{ kid_{f} | school_{f} \} * Pr\{ school_{f}, school_{h} \}$

So $Pr \{kid_f, kid_h, school_f, school_h\} =$

 $P_{f} \text{ school}_{h}, \text{ school}_{h} \in [WEIGHT_{f} * Pr\{\text{school}_{f}, \text{school}_{h}\}]^{-1} * [WEIGHT_{h} * Pr\{\text{school}_{h}\}]^{-1}$

- = [WEIGHT_f* WEIGHT_h* Pr{school_h}]⁻¹
- = $[WEIGHT_{f}*WEIGHT_{h}]^{-1}*SCHWT1_{h}$

The pair weight for a pair composed of a student from the feeder school and a student from the high school is:

$$PAIRWT_{f,h} = WEIGHT_{f}^{*} WEIGHT_{h} / SCHWT1_{h}$$
(4)

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