

2009-2010 National Adult Tobacco Survey

Weighting Methodology Report

Prepared by

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Introduction¹

Background and Purpose of the National Adult Tobacco Survey

In 1999, the Office on Smoking and Health (OSH), a division in the Centers for Disease Control and Prevention (CDC), created the National Tobacco Control Program (NTCP) to encourage coordinated efforts nationwide to reduce tobacco-related diseases and deaths. The program provides funding and technical support to State and territorial health departments for comprehensive tobacco control programs. The four goals of the NTCP are to

- (1) prevent initiation of tobacco use among young people,
- (2) eliminate non-smokers' exposure to second-hand smoke,
- (3) promote cessation among adults and young people, and
- (4) identify and eliminate tobacco-related health disparities.

The NATS is the first adult tobacco survey designed within the framework provided by the Key Outcome Indicator (KOI) report. As such, it both includes the established measures identified in the KOI report and introduces measures for the new or revised indicators proposed in the KOI report. The NATS also establishes a comprehensive framework for evaluating both the national and State-specific tobacco control programs. As described below, the NATS sample design prescribes a roughly equal target number of completes for each State in order to allow analyses by State. Prior to the NATS, 25 States had independently conducted an Adult Tobacco Survey (ATS) with technical assistance and support from CDC's OSH. After a pilot program in conjunction with the States, OSH started regularly supporting ATS's in 2002. See (Centers for Disease Control and Prevention, 2010) for State ATS data from 2003 to 2007. Thus, the primary purpose of the NATS is to evaluate the CDC's NTCP. OSH developed the NATS to assess the prevalence of tobacco use and the factors promoting and impeding tobacco use among adults. Specifically, the NATS is meant to:

- (1) Estimate the extent to which adults engage in tobacco use behaviors.
- (2) Assess the degree to which tobacco use behaviors among adults vary as a function of gender, age, and race/ethnicity.

¹ This introduction is reprinted from the 2009-2010 NATS Methodology Report prepared by CDC's OSH and ICF Macro

- (3) Estimate the accomplishment of key short-, intermediate-, and long-term tobacco prevention and control outcome indicators found in the *Key Outcome Indicators for Evaluating Comprehensive Tobacco Control Programs* report.
- (4) Estimate the degree to which exposure to influences expected to promote or impede tobacco use has its expected effects.
- (5) Assess the degree to which response to influences expected to promote or impede tobacco use varies as a function of gender, age, and race/ethnicity.

NATS Methodology at a Glance

The NATS was designed as a stratified, national, landline and cell phone survey of non-institutionalized adults 18 years of age and older. It was designed to yield data representative and comparable at both national and State levels. Each state is divided into at least three strata: a listed landline stratum, a not-listed landline stratum, and a cell phone stratum. Some states have additional landline strata based on counties or county-equivalents. The OSH target number of landline completes per state was 1863. OSH's target number of cell phone completes per state varied in proportion to each state's population. Some states independently added respondents to the OSH targets.

The NATS was implemented from October 20, 2009 to February 28, 2010. Respondent selection varied by phone type. For landline telephone numbers, one adult age 18 and over was randomly selected from households with at least one adult age 18 or over. Adults age 18 and over reached via a cell phone telephone number were selected if a cell phone was the only way they could be reached by telephone at home. We assumed that a cell phone was used only by the person who answered. The exact question was "In your home, is a cell phone the only way you can be reached by telephone?" The data collection protocol for landline telephone numbers specified that telephone numbers without a final disposition code after at least 15 call attempts could be assigned a final disposition code. For cell phone telephone numbers, a final disposition code could be assigned to a telephone number after a minimum of 6 call attempts. The calls had to be appropriately distributed among weekdays, weeknights, and weekends. A total of 118,581 interviews—110,634 landline interviews and 7,947 cell phone interviews—of non-institutionalized adults aged 18 and older were completed.

The landline data were first weighted by the inverse of the probability of selection of the telephone number, a non-response adjustment, and adjustments for number of landlines and number of eligible subjects in a household. The cell phone data were initially weighted by the inverse of the probability of selection of the telephone number and a nonresponse adjustment. Next, the data were poststratified by state to the distributions of various demographic variables and phone type (cell-phone-only users and all others).

National Adult Tobacco Survey Sample Design²

² This section is reprinted from the 2009-2010 NATS Methodology Report prepared by CDC's OSH and ICF Macro

The NATS’s target population was non-institutionalized adults age 18 and older residing in the 50 States and D.C. The sample was designed with two main objectives:

- (1) Precise overall State-level estimates; and
- (2) National estimates for subgroups defined by gender, age, and race/ethnicity.

The sample allocation focused on the need for precise estimates at the State level since this required the larger overall sample size between the two main objectives.

Sampling Frames

Respondents were selected from two sampling frames, one for landlines and one for cell phones. Each state was divided into at least three strata: a listed landline stratum, a not-listed landline stratum, and a cell phone stratum. Some states had additional landline strata based on counties or county-equivalents. The listed stratum consists of landline telephone numbers listed in residential directories or in other source databases. The not-listed stratum consists of landline telephone numbers not listed as a residential number in any source database.

The NATS’s landline sampling population is listed and not-listed telephone numbers from hundred-blocks with one or more listed telephone numbers. A “hundred block” is a set of 100 telephone numbers with the same area code, prefix, first two digits of the suffix, and all permutations of the last two digits of the suffix, from 00 to 99. A “one-plus block” telephone number is a telephone number from a hundred block with one or more listed household telephone numbers. The landline sampling frame for the national adult tobacco survey is the set of all one-plus block telephone numbers in the US. The cell phone sampling frame is telephone numbers from cellular-dedicated, thousand blocks, sets of telephone numbers with the same area code and prefix. The blocks originated from the Telcordia® LERG. The cellular-dedicated banks were then identified by coding provided on the LERG.

Sample Selection

Telephone numbers listed in residential directories and other database sources are most often working residential numbers, whereas unlisted telephone numbers include large numbers of non-working and non-residential telephone numbers. To leverage this information, the listed stratum was oversampled at a 1.5-to-1 ratio relative to the not-listed stratum. This oversampling increases sampling efficiency by raising the percentage of working residential numbers selected in the sample.

For the landline strata, each State was allocated an equal sample size to ensure adequate precision for survey estimates at the State level. The total target landline sample size of 95,013 was equally distributed among the states, 1,863 each. For the cell phone strata, each state was allocated a sample size in proportion to its population. The total cell phone number of completes was 7,947.

Overview of the weighting strategy

The landline sample was allocated so that each state had a least 1,863 landline respondents. In contrast, the cell phone sample was allocated proportional to the population size of the state. Consequently, some states have very few cell phone respondents. For states with a small number of cell phone respondents, using both landline and cell phone data results in a large unequal weighting effect, and therefore, large estimated variances of survey estimates and small effective sample sizes.³

To accommodate this design, three sets of weights were created. One weight is used when calculating national estimates. It is called WT_NATIONAL on the data file. This weight uses all of the respondents in all of the states. Another weight, called WT_STATE, is used for the state level estimates. In this weight cell phone respondents are assigned a non-zero weight for states with 200 or more cell phone respondents whereas cell phone respondents are assigned a weight of zero in states with less than 200 cell phone respondents. As just described, the cell phone sample members are excluded from the weights in the states with small cell phone sample sizes because their inclusion would result in much larger variances and smaller effective sample sizes. Table 1 contains the number of cell phone respondents in each state; and, states with 200 or more cell phone respondents are in bold. The third weight, called WT_LANDLINE, only uses landline respondents. See Appendix A for a discussion on why the cutoff of 200 was implemented.

An earlier version of the sample weights was released. We recommend using the weights described in this document rather than the first iteration of the weights. This second iteration has a number of refinements that will reduce bias. In the first iteration of the weights, the nonresponse adjustments were simply ratio adjustments within stratum and population density category. In contrast, in this iteration of the weights, a logistic model was used to predict response propensity. This model can distinguish the different response propensities across frame members, not just within broad categories. Consequently, more nonresponse bias is removed. This is particularly important for telephone studies in general because of low response rates and the NATS specifically which has a 37.6% response rate. Also, this iteration of the weights uses many more variables in the poststratification. In the original weights the only categories used in the poststratification were gender, age category and type of telephone usage (cell-only, landline). In addition to these categories, the weights were also poststratified to race/ethnicity, marriage status and educational attainment. These additional poststratification categories are important because many of the important study variables, such as smoking status, are correlated with these additional variables. Inclusion of these variables in the poststratification will reduce coverage bias and nonresponse bias. Finally, another advantage of this weighting approach is that a more current estimate of the percent of the population that is “cell only” was used. These estimates came from the National Health Statistics Report (Blumberg 2011) dated April 20, 2011, and cover from July 2009-June 2010 (which corresponds closely with the NATS data collection period), whereas the old weights used the National Health Statistics Report (Blumberg 2009) dated March 11, 2009, and cover from January 2007-December 2007.

The steps in the calculation of the weights are discussed in the following sections, and are the following:

³ See the section on Unequal Weighting Effect for a definition of unequal weighting effect and effective sample size.

- Calculation of the initial weight for the landline and cell phone samples,
- Adjustment for nonresponse,
- Adjustment for household size and number of landlines,
- Adjustment for undercoverage using poststratification, and
- Evaluation of the unequal weighting effect nationally and for each state.

Calculate initial weights for the landline and cell sample:

Respondents and nonrespondents

For respondents and nonrespondents, the initial design weight is calculated:

$$W_{ij}^1 = \frac{N_i}{n_i},$$

W_{ij}^1 = the initial weight for the j^{th} sample member in stratum i .

N_i = the number of records in stratum i ; and

n_i = the number of records selected in stratum i .

There are 229 total strata. Within the landline frame, some states are broken into county groups. These geographical strata are broken into listed and unlisted numbers. Cell phones were only stratified by state.

Within each strata there were between 4 and 8 different samples selected. To calculate the total sample selected in each stratum (n_i) the individual samples, the individual samples were summed. But, the total frame count (N_i) varied over the different samples. The maximum value over all the samples in a stratum was used to calculate the frame count.

Unknown response status

An adjustment for unknown eligibility status was calculated as follows:

$$U_i^{adj} = \frac{n_i^r + n_i^n}{n_i^r + n_i^n + n_i^i},$$

U_i^{adj} = the unknown eligibility weight adjustment.

n_i^r = the number of responders in stratum i ,

n_i^n = the number of nonresponders in stratum i , and

n_i^i = the number of ineligible in stratum i .

Next, the telephone design weight for sampled frame members with unknown eligibility status was calculated as follows:

$$W_{ij}^1 = \frac{U_i^{adj} * N_i}{n_i}.$$

Ineligible frame members were then removed.

Appendix B contains a table displaying, for each stratum, the sample selected, the frame count, the initial weight, the unknown adjustment and the initial weight with the unknown adjustment which is applied to the sample members with unknown response status.

Nonresponse adjustment:

The nonresponse adjustment involves the following steps:

- Ancillary data are appended to the sample frame
- A logistic model is fit in which the outcome is response and the independent variables are the ancillary data
- The nonresponse adjustment, the inverse of the predicted probability of response, is applied to the respondents.

Append ancillary data to the sample frame

The sample frame contains geographical information for each sample member. This information was used to append ancillary data which was then used to model the probability of response. The most precise geographical information that is available for all of the sample members from the cell phone frame is the area code, and for sample members from the landline frame it is county FIPS code.

Geolytics, a private company that sells the ancillary data used in the nonresponse adjustment, uses census data and a modeling algorithm to create estimates for various domains at various geographical levels. The county level data sets from Geolytics were used in this weight adjustment, and a weighted average (the weight is the population count) was used to aggregate the blockgroup level data to area code level data.

The following variables are used in the nonresponse adjustment procedure.

- Population density
- Proportion white
- Proportion African-American
- Proportion Hispanic
- Proportion of families below 150% of the poverty line
- Proportion that are high school graduates
- Proportion that completed a Bachelor's degree

The above variables were made categorical by converting the distribution of each variable into its quintiles.

Fit the logistic model

For each of the 102 state by frame type (cell phone, land line) combinations, a weighted logistic model was used. The weights (W_{ij}^1) calculated up to this step and SAS software (PROC SURVEYLOGISTIC) were used to fit the following model:

$$\text{logit}(p_i) = \beta_{i,0} + \beta_{i,1}X_{i,1} + \cdots + \beta_{i,7}X_{i,7}.$$

The index i refers to one of the 102 models. The independent variables are the seven variables ($X_1 \cdots X_7$) in the list above. Then we calculate the probability of response for each sample member:

$$P_{i,j} = \frac{1}{1 + e^{-\text{logit}(p_{i,j})}}$$

The index i refers to one of the 102 models. The index j refers to one of the respondents within one of the 102 models.

Apply the nonresponse adjustment

A new weight was calculated as:

$$W_{i,j}^2 = W_{i,j}^1 * \frac{1}{P_{i,j}}.$$

$W_{i,j}^2$ = nonresponse adjusted probability for the j th respondent,

$P_{i,j}$ = the predicted probability of response for the j th respondent from the logistic model.

Because the sum of the nonresponse adjusted weights ($\sum W_{i,j}^2$) are not exactly equal to the sum of ($\sum W_{i,j}^1$), the following ratio adjustment was made.

$$W_{i,j}^3 = \frac{\sum W_{i,j}^1}{\sum W_{i,j}^2} W_{i,j}^2.$$

Adjust for household size and number of phone lines per household:

The landline sample uses a two-stage selection process. First, the household is selected. Then, within a household one subject is randomly chosen from all eligible household members. Additionally, the probability of selecting a landline household is a function of the number of residential lines. So, for respondents sampled on landline phones the weight was adjusted for the number of eligible household members and for the number of residential phone lines in the household. The maximum number of eligible subjects in a household and the maximum number of residential phone lines was truncated in the adjustment is 5, in order to prevent sample members from extremely large households from having a large impact on the weights.

$$W_{i,j}^4 = \frac{W_{i,j}^3 * a_{i,j}}{l_{i,j}}$$

where, $a_{i,j}$ = the number of adults in the household of respondent for the j th listed frame member in stratum i , and

$l_{i,j}$ = the number of landlines in the household of respondent for the j th listed frame member in stratum i .

The cell phone frame uses a one-stage selection process. Cell phones are generally considered as single user devices, and were treated in this manner for weighting. This is expressed in the sampling methodology by sampling the person who answers the cell phone. Consequently, there is no adjustment for cell phones at this stage. So, for cell phone respondents the fourth weight is equal to the third weight.

$$W_{i,j}^4 = W_{i,j}^3$$

Poststratification and trimming

Weight trimming is applied to the state level weights before the weights are poststratified to population totals. Very large weights increase the variance of the parameter estimates. The purpose of weight trimming is to constrain the most extreme weights, and thereby reduce variance. The concern with weight trimming is that bias will be introduced. However, when weight trimming is applied, one is willing to accept some increase in bias to achieve a reduction in variance. The following rule was applied for weight trimming. Within each state, the maximum weight was defined as the median weight plus 4 times the inter-quartile range of the weights.

$$\text{Max weight} = \text{Median}(\text{WT}) + 4 * (\text{Q3} - \text{Q1})$$

Weights greater than the max weight were truncated to the max weight. This weight trimming rule is similar to the rule applied to data from the CDC study Behavioral Risk Factor Surveillance System and is described in the paper (Izreal 2009).

The purpose of poststratification is to constrain the weights to known population totals. This could remove nonresponse and coverage bias. In the NATS, 60.8% of all respondents are female, whereas 51.4% of the US population 18+ is female. The reason that the NATS has a higher percentage of females than the population is differential nonresponse and coverage error. Males are more likely to be nonrespondents than females, and males are more likely to be cell phone only; consequently, males are in the respondent sample at a lower rate than females. If gender is correlated with a study outcome then nonresponse and coverage error will result in nonresponse bias and coverage bias in the parameter estimate for that outcome. Adjusting the weights to sum to the population totals for gender (and other demographic categories that are correlated with study outcomes) can reduce these sources of bias. Within each state, the weights were adjusted to sum to the distributions of the following demographic variables: age category, gender, race/ethnicity, marriage status and educational attainment. The weights were also adjusted to sum to the distributions of landline and cell phone only telephone usage. The distributions of these variables are displayed in tables 2-6.

The phone usage data were derived from the National Health Statistics Report, Number 39, Blumberg et al (AAPOR 2010). The other data used in the poststratification were obtained from the 2005-2009 American Community Survey 5-year Summary File (ACS 2011).

The SUDAAN software procedure WTADJUST was used to apply a model based approach to poststratify the weights to the population totals.

Imputing the poststratification variables

Imputation was used to replace missing values for the survey responses for the variables used in poststratification. The following table shows the variables and the number of respondents with missing values.

Variable	Categories	Number of missing values
Age	18-24, 25-34, 35-44, 45-54, 55-64, 65+	2,840
Gender	Male, female	173
Race	Hispanic, white alone (non-Hispanic), black alone (non-Hispanic), other race	1,664
Marriage status	Married, Divorced, Widowed, Other	1,080
Educational Attainment	Less than high school graduate, , High school graduate but not bachelor's degree, Bachelor's degree or higher	797

For the binary response variable (gender) a logistic model using a logit link function was used. For the other variables with more than two categories (polytomous response) logistic models with cumulative logits were used. The fitted model was used to predict, for subjects with missing values, the probability that subject would respond to the various categories of the variable. These categories were sampled with probability proportion to the predicted probability, and the imputed value is the category selected.

The following are the independent variables used in the imputation modeling procedures:

- State of residence
- Call attempts
- Number of males in household
- Number of females in household
- Gender (not used to impute gender)
- General health
- Smoked 100 cigarettes in life
- Ever smoked
- Employed
- Marital status (not used to impute marital status)
- Income

- Education
- Race/ethnic category (not used to impute race)
- Age category (not used to impute age)

Some of the independent variables (number of call attempts, number of men in the household and number of woman in the household) were collapsed into categories.

The weights will be delivered in a SAS data set containing one record for each respondent. In addition to three weight variables, the imputed poststratification categories will be included on the weight file.

Redefining the stratum for “stratum jumpers”

“Stratum jumpers” is a term used to describe survey respondents that are in a stratum outside of their state of residence because of frame error. For example, it could be the case that a telephone number is sampled in New Jersey, but, the subject resides in Georgia. Then when the data is used to make state level estimates for Georgia one respondent is from a stratum in New Jersey. But, the algorithm for calculating the variance of the point estimate requires 2 subjects from each stratum. Consequently, the algorithm will fail to converge. To avoid this problem we reassign subjects with a stratum that differs from their state of residence. The stratum variable for these “stratum jumpers” is changed to the stratum in their state of residence with most respondents and with the same phone type (listed, unlisted or cell phone). Redefining the stratum variable for “stratum jumpers” will not affect means. For example, there will be no effect on estimates of smoking prevalence. However, it will affect the estimate of variance. But, this affect will be very small if it is detectable at all.

Unequal Weighting Effect

The unequal weighting effect (UWE) is an upper bound of the ratio of the variance of an estimate calculated from a survey to the variance one would obtain from a simple random sample with the same sample size. The concept of UWE is described by Paul Biemer in the International Handbook of Survey Methodology. The following is an excerpt from this book.

“Kish (1965, p.427) derived a formula for determining the maximum increase in variance of an estimate of a population mean due to a weight variation. His formula assumes there is no correlation between the survey weights and the characteristic whose mean is to be estimated. This may be a good approximation for many survey variables because the survey design and weight adjustments are optimized for only a few key characteristics out of hundreds that may be collected in a survey. The actual variance increase will vary across characteristics in the survey and will be smaller for characteristics where the covariance between the observations and the weights are larger. Under these assumptions, Kish obtained the following expression for the unequal weighting effect (UWE) defined as the ratio of the variances of the weighted mean to the variance of the unweighted mean:

$$\text{UWE} = 1 + cv^2$$

“Where cv is the coefficient of variance of the weights or the sample standard deviation of the weights divided by the sample average weight.” (Leeuw, 2008, p. 337)

The effective sample size (ESS) is the sample size divided by the UWE. The ESS is the sample size a simple random sample needs to be to have an equal variance, for variables uncorrelated with the weight, to our design. Let us use Alaska to illustrate UWE and ESS. To make estimates for Alaska we use the 1,836 landline respondents. The UWE is 2.2. Consequently, for estimates not correlated with the weight, the variance will be 2.2 times greater than a simple random sample with 1,836 respondents; the standard error of an estimate will be $\sqrt{2.2} = 1.48$ times greater than what we would observe with a simple random sample with the same sample size. The ESS is 839 $(1,836/2.2)^4$. That is, parameters that are uncorrelated with the weights will have the same variance as a simple random sample with a sample size of 839.

The UWE does not take into account the effect of the stratification on the estimates. Stratification usually reduces variance.

Table 1 contains the unequal weighting effects (UWE) and the effective sample sizes (ESS) for the three weights (WT_NATIONAL, WT_STATE, WT_LANDLINE) for the national domain and each state domain.

For national estimates using the weight WT_NATIONAL, the results are in columns UWE and EFF. For national estimates using the weight WT_LANDLINE, the results are in columns Landline only UWE and Landline only EFF.

For state estimates using the weight WT_STATE, the results are in columns UWE and EFF. For national estimates using the weight WT_LANDLINE, the results are in columns Landline only UWE and Landline only EFF.

For states with less than 200 cell phone respondents, the weight WT_STATE only contains non-zero weights for landline respondents. As a result, for states with less than 200 cell phone respondents, the weights in WT_STATE are identical to the weights in WT_LANDLINE, and the UWE and EFF for STATE_WT and LANDLINE_WT are also identical. States with more than 200 cell phone respondents are displayed in **bold**. As discussed earlier, cell phone respondents in states with fewer cell phone interviews were not included in the state-level weight because their inclusion resulted in a much larger unequal weighting effect in those states.

Table 1: Unequal Weighting Effect

State FIPS	State	Cell phone respondents	Landline respondents	UWE	Effective Sample Size	Landline only UWE	Landline only Effective Sample Size
N/A	National	7,947	110,634	4.9	24,443	6.9	16,099
1	AL	68	1,902	2.5	770	2.5	770
2	AK	13	1,836	2.2	839	2.2	839
4	AZ	111	1,799	3.8	476	3.8	476
5	AR	50	2,818	2.6	1,065	2.6	1,065

⁴ This is approximate since 2.2 is a rounded value

State FIPS	State	Cell phone respondents	Landline respondents	UWE	Effective Sample Size	Landline only UWE	Landline only Effective Sample Size
6	CA	723	1,849	2.4	1,072	2.9	646
8	CO	145	1,816	2.6	686	2.6	686
9	CT	56	1,839	2.7	690	2.7	690
10	DE	18	1,966	2.4	816	2.4	816
11	DC	26	1,870	4.0	468	4.0	468
12	FL	408	1,863	1.9	1,186	3.1	595
13	GA	231	4,683	3.0	1,616	2.8	1,650
15	HI	33	1,788	2.5	730	2.5	730
16	ID	53	1,774	2.8	623	2.8	623
17	IL	245	1,805	2.0	1,039	2.8	639
18	IN	152	1,873	2.1	879	2.1	879
19	IA	76	2,051	2.4	856	2.4	856
20	KS	70	1,850	2.2	843	2.2	843
21	KY	73	1,777	2.5	707	2.5	707
22	LA	307	6,044	2.9	2,212	2.8	2,186
23	ME	36	1,995	2.1	928	2.1	928
24	MD	134	1,841	2.8	647	2.8	647
25	MA	134	1,818	2.7	667	2.7	667
26	MI	184	1,820	2.4	766	2.4	766
27	MN	116	1,788	2.2	821	2.2	821
28	MS	57	1,754	2.9	603	2.9	603
29	MO	140	1,859	2.2	833	2.2	833
30	MT	36	1,826	2.2	819	2.2	819
31	NE	47	1,829	2.3	784	2.3	784
32	NV	51	1,803	2.6	694	2.6	694
33	NH	27	1,934	2.0	964	2.0	964
34	NJ	801	3,294	2.3	1,820	3.0	1,110
35	NM	62	1,791	3.0	594	3.0	594
36	NY	401	1,838	2.0	1,099	2.6	699
37	NC	204	1,815	2.5	728	2.5	728
38	ND	29	2,192	2.4	903	2.4	903
39	OH	289	1,856	1.5	1,393	2.2	862
40	OK	526	3,123	1.6	2,288	2.2	1,390
41	OR	158	1,869	2.5	757	2.5	757
42	PA	252	3,181	1.7	2,046	2.0	1,579
44	RI	26	1,906	2.5	749	2.5	749
45	SC	88	5,078	3.1	1,614	3.1	1,614
46	SD	32	1,993	2.3	864	2.3	864
47	TN	127	1,832	2.4	750	2.4	750
48	TX	476	1,882	2.0	1,155	3.2	594
49	UT	83	2,025	2.6	777	2.6	777
50	VT	23	2,041	2.1	985	2.1	985
51	VA	177	2,271	2.6	887	2.6	887
53	WA	196	1,850	2.2	822	2.2	822
54	WV	45	1,770	2.2	787	2.2	787
55	WI	112	1,825	2.1	855	2.1	855
56	WY	20	1,732	2.5	695	2.5	695

Table 2: Population Totals by State: Age Categories (age 18+)

State	Total	18-24	25-34	35-44	45-54	55-64	65+
Alabama	3,509,577	459,914	601,870	630,875	665,660	525,127	626,131
Alaska	499,977	80,919	97,084	96,196	106,926	71,216	47,636
Arizona	4,652,197	589,846	919,448	858,473	819,130	651,464	813,836
Arkansas	2,136,963	271,198	376,918	375,125	393,246	322,222	398,254
California	26,868,769	3,649,552	5,284,454	5,339,190	5,062,646	3,561,732	3,971,195
Colorado	3,646,549	490,872	699,987	711,876	726,949	519,250	497,615
Connecticut	2,672,367	328,175	403,751	519,164	548,747	397,735	474,795
Delaware	658,603	81,515	112,525	121,689	125,193	98,832	118,849
DC	474,441	69,898	105,778	86,657	78,308	64,251	69,549
Florida	14,167,647	1,635,367	2,291,105	2,517,243	2,564,133	2,088,829	3,070,970
Georgia	6,983,323	952,166	1,354,887	1,443,696	1,326,631	958,667	947,276
Hawaii	990,953	125,426	182,813	175,834	179,290	147,004	180,586
Idaho	1,086,071	159,304	198,968	191,246	202,411	158,794	175,348
Illinois	9,591,923	1,282,601	1,756,017	1,818,828	1,849,893	1,333,538	1,551,046
Indiana	4,755,941	637,155	827,413	878,969	925,329	687,675	799,400
Iowa	2,268,969	320,716	354,546	385,827	438,321	330,823	438,736
Kansas	2,079,386	303,959	354,568	363,202	402,408	294,743	360,506
Kentucky	3,241,564	415,946	565,564	600,914	625,607	485,532	548,001
Louisiana	3,297,180	482,389	583,680	587,240	634,469	475,091	534,311
Maine	1,036,654	118,355	147,678	184,858	216,823	171,675	197,265
Maryland	4,271,121	532,517	738,127	845,785	865,672	625,841	663,179
Massachusetts	5,061,250	645,115	839,719	975,711	997,328	733,667	869,710
Michigan	7,600,237	988,620	1,225,685	1,419,233	1,534,653	1,139,998	1,292,048
Minnesota	3,926,248	523,046	675,511	733,404	794,863	556,360	643,064
Mississippi	2,158,108	315,158	381,340	383,363	403,457	310,117	364,673
Missouri	4,473,226	589,319	757,402	801,084	871,891	656,852	796,678
Montana	736,558	103,470	112,416	117,203	149,498	119,199	134,772
Nebraska	1,326,139	194,946	224,517	228,914	254,072	187,582	236,108
Nevada	1,887,203	223,022	378,436	372,973	347,285	279,016	286,471
New Hampshire	1,017,239	127,782	148,323	197,151	217,309	157,715	168,959
New Jersey	6,578,265	735,507	1,102,731	1,316,847	1,329,184	953,332	1,140,664
New Mexico	1,460,839	201,713	263,031	253,610	275,840	218,283	248,362
New York	14,925,614	1,877,930	2,605,989	2,836,735	2,881,804	2,162,451	2,560,705
North Carolina	6,839,619	899,371	1,207,653	1,321,189	1,284,241	1,006,095	1,121,070
North Dakota	496,341	87,872	77,660	75,892	92,765	68,508	93,644
Ohio	8,750,969	1,079,419	1,468,399	1,591,383	1,746,072	1,302,614	1,563,082
Oklahoma	2,709,105	384,844	479,169	464,155	504,889	394,707	481,341
Oregon	2,862,879	356,297	503,583	506,038	551,749	454,323	490,889
Pennsylvania	9,703,855	1,193,167	1,502,121	1,728,848	1,916,548	1,451,607	1,911,564
Rhode Island	823,719	109,997	132,505	153,275	160,925	118,568	148,449

State	Total	18-24	25-34	35-44	45-54	55-64	65+
South Carolina	3,355,523	439,558	579,940	610,614	627,305	518,605	579,501
South Dakota	600,114	86,327	97,856	98,075	116,095	87,542	114,219
Tennessee	4,683,923	573,728	834,703	875,602	895,080	707,833	796,977
Texas	17,185,928	2,463,010	3,488,231	3,387,812	3,201,378	2,243,645	2,401,852
Utah	1,826,327	340,982	414,891	320,042	302,377	213,558	234,477
Vermont	489,395	68,299	67,537	85,347	102,398	80,347	85,467
Virginia	5,890,260	797,902	1,044,124	1,146,581	1,140,502	853,293	907,858
Washington	4,918,282	628,844	895,386	923,053	971,842	739,072	760,085
West Virginia	1,424,575	170,890	222,254	241,313	274,982	233,643	281,493
Wisconsin	4,279,113	585,806	687,508	785,727	860,734	620,738	738,600
Wyoming	397,980	58,505	67,402	64,513	81,914	62,042	63,604
	227,279,008	29,838,236	40,443,203	42,748,574	43,646,772	32,601,353	38,000,870

Table 3: Population Totals by State: Gender (18) and percent cell-only

State	Total	Male	Female	Cell only %
Alabama	3,509,577	1,668,871	1,840,706	25.3
Alaska	499,977	261,423	238,554	19.9
Arizona	4,652,197	2,312,791	2,339,406	29.4
Arkansas	2,136,963	1,029,846	1,107,117	35.2
California	26,868,769	13,323,473	13,545,296	18.2
Colorado	3,646,549	1,825,856	1,820,693	30.4
Connecticut	2,672,367	1,282,748	1,389,619	13.6
Delaware	658,603	314,393	344,210	16.5
DC	474,441	219,579	254,862	27.7
Florida	14,167,647	6,875,871	7,291,776	27.3
Georgia	6,983,323	3,378,409	3,604,914	26.5
Hawaii	990,953	497,438	493,515	21.8
Idaho	1,086,071	540,975	545,096	31.7
Illinois	9,591,923	4,658,495	4,933,428	24.4
Indiana	4,755,941	2,309,439	2,446,502	25.2
Iowa	2,268,969	1,104,622	1,164,347	29.2
Kansas	2,079,386	1,019,073	1,060,313	28.7
Kentucky	3,241,564	1,566,212	1,675,352	31.5
Louisiana	3,297,180	1,574,713	1,722,467	26.8
Maine	1,036,654	499,276	537,378	22.9
Maryland	4,271,121	2,032,738	2,238,383	18.4
Massachusetts	5,061,250	2,418,058	2,643,192	16.8
Michigan	7,600,237	3,688,862	3,911,375	29.2
Minnesota	3,926,248	1,934,180	1,992,068	25.2
Mississippi	2,158,108	1,025,515	1,132,593	35.1
Missouri	4,473,226	2,149,145	2,324,081	22.4
Montana	736,558	365,078	371,480	19.4

State	Total	Male	Female	Cell only %
Nebraska	1,326,139	649,688	676,451	30.4
Nevada	1,887,203	957,373	929,830	24.2
New Hampshire	1,017,239	495,942	521,297	16.0
New Jersey	6,578,265	3,170,753	3,407,512	12.8
New Mexico	1,460,839	715,267	745,572	27.2
New York	14,925,614	7,124,681	7,800,933	17.0
North Carolina	6,839,619	3,294,267	3,545,352	25.2
North Dakota	496,341	247,618	248,723	32.3
Ohio	8,750,969	4,201,587	4,549,382	25.6
Oklahoma	2,709,105	1,318,499	1,390,606	30.1
Oregon	2,862,879	1,405,986	1,456,893	30.6
Pennsylvania	9,703,855	4,646,983	5,056,872	16.5
Rhode Island	823,719	392,485	431,234	12.8
South Carolina	3,355,523	1,606,879	1,748,644	25.8
South Dakota	600,114	296,675	303,439	15.6
Tennessee	4,683,923	2,247,275	2,436,648	27.9
Texas	17,185,928	8,484,856	8,701,072	32.5
Utah	1,826,327	908,967	917,360	24.4
Vermont	489,395	237,762	251,633	20.3
Virginia	5,890,260	2,857,831	3,032,429	21.2
Washington	4,918,282	2,431,688	2,486,594	26.4
West Virginia	1,424,575	688,377	736,198	20.5
Wisconsin	4,279,113	2,104,516	2,174,597	25.3
Wyoming	397,980	200,873	197,107	22.3
	227,279,008	110,563,907	116,715,101	

Table 4: Population Totals by State: Race Category (18+)

State	Total	Hispanic	White alone Not-Hispanic	Black alone	Other race
Alabama	3,509,577	76,935	2,483,494	868,684	80,464
Alaska	499,977	25,251	345,760	18,220	110,746
Arizona	4,652,197	1,175,706	2,977,043	155,777	343,671
Arkansas	2,136,963	90,938	1,678,665	305,400	61,960
California	26,868,769	8,490,040	12,577,310	1,644,941	4,156,478
Colorado	3,646,549	613,010	2,730,336	132,681	170,522
Connecticut	2,672,367	269,419	2,048,239	235,453	119,256
Delaware	658,603	35,381	470,397	127,167	25,658
DC	474,441	36,907	167,198	248,619	21,717
Florida	14,167,647	2,741,857	9,013,851	1,965,310	446,629
Georgia	6,983,323	453,100	4,266,375	1,991,431	272,417
Hawaii	990,953	68,458	265,283	23,747	633,465
Idaho	1,086,071	87,788	954,046	5,251	38,986
Illinois	9,591,923	1,201,520	6,566,298	1,324,794	499,311
Indiana	4,755,941	197,314	4,056,116	382,236	120,275
Iowa	2,268,969	72,109	2,093,437	47,648	55,775
Kansas	2,079,386	148,754	1,729,732	110,328	90,572
Kentucky	3,241,564	61,097	2,888,960	228,950	62,557
Louisiana	3,297,180	102,561	2,125,255	981,907	87,457
Maine	1,036,654	10,545	991,938	8,243	25,928
Maryland	4,271,121	246,029	2,559,800	1,192,830	272,462
Massachusetts	5,061,250	357,882	4,110,189	284,037	309,142
Michigan	7,600,237	254,306	6,062,236	991,968	291,727
Minnesota	3,926,248	122,820	3,455,964	146,816	200,648
Mississippi	2,158,108	39,963	1,332,811	746,959	38,375
Missouri	4,473,226	113,924	3,758,236	462,263	138,803
Montana	736,558	16,607	663,898	3,095	52,958
Nebraska	1,326,139	80,337	1,153,739	49,347	42,716
Nevada	1,887,203	398,681	1,170,794	134,160	183,568
New Hampshire	1,017,239	21,354	958,274	9,491	28,120
New Jersey	6,578,265	957,560	4,233,155	852,460	535,090
New Mexico	1,460,839	606,122	671,367	29,734	153,616
New York	14,925,614	2,230,565	9,358,112	2,205,737	1,131,200
North Carolina	6,839,619	381,404	4,808,321	1,386,589	263,305
North Dakota	496,341	8,107	454,854	4,062	29,318
Ohio	8,750,969	188,161	7,379,820	956,341	226,647
Oklahoma	2,709,105	164,350	2,027,338	186,212	331,205
Oregon	2,862,879	234,718	2,393,753	45,242	189,166
Pennsylvania	9,703,855	365,475	8,123,472	924,639	290,269
Rhode Island	823,719	76,959	673,567	39,955	33,238
South Carolina	3,355,523	111,708	2,268,186	900,408	75,221
South Dakota	600,114	11,820	533,491	4,953	49,850

State	Total	Hispanic	White alone Not-Hispanic	Black alone	Other race
Tennessee	4,683,923	138,166	3,714,208	717,891	113,658
Texas	17,185,928	5,522,882	8,912,545	1,943,311	807,190
Utah	1,826,327	181,360	1,541,689	15,748	87,530
Vermont	489,395	5,771	468,795	3,208	11,621
Virginia	5,890,260	340,498	4,072,549	1,110,826	366,387
Washington	4,918,282	372,609	3,867,888	157,553	520,232
West Virginia	1,424,575	12,932	1,341,875	44,675	25,093
Wisconsin	4,279,113	167,974	3,752,732	219,307	139,100
Wyoming	397,980	25,272	353,823	3,592	15,293
	227,279,008	29,715,006	156,607,214	26,580,196	14,376,592

Table 5: Population Totals by State: Marriage Categories (18+)

State	Total	Married	Divorced	Widowed	Other marriage status
Alabama	3,509,577	1,813,174	433,399	277,038	985,966
Alaska	499,977	244,156	62,745	18,654	174,422
Arizona	4,652,197	2,294,389	594,503	281,095	1,482,210
Arkansas	2,136,963	1,143,860	277,078	168,633	547,392
California	26,868,769	12,708,913	2,712,733	1,519,321	9,927,802
Colorado	3,646,549	1,919,565	439,676	178,170	1,109,138
Connecticut	2,672,367	1,378,271	276,237	181,617	836,242
Delaware	658,603	330,138	73,074	46,361	209,030
District of Columbia	474,441	114,927	48,484	30,125	280,905
Florida	14,167,647	6,989,891	1,801,827	1,095,985	4,279,944
Georgia	6,983,323	3,400,617	825,315	428,175	2,329,216
Hawaii	990,953	494,171	95,466	63,985	337,331
Idaho	1,086,071	639,041	132,435	58,687	255,908
Illinois	9,591,923	4,779,384	953,558	639,202	3,219,779
Indiana	4,755,941	2,545,013	595,839	319,094	1,295,995
Iowa	2,268,969	1,286,265	236,859	159,908	585,937
Kansas	2,079,386	1,151,549	240,546	138,437	548,854
Kentucky	3,241,564	1,730,404	412,711	241,427	857,022
Louisiana	3,297,180	1,563,424	386,127	251,511	1,096,118
Maine	1,036,654	557,550	137,443	73,008	268,653
Maryland	4,271,121	2,081,923	427,851	273,373	1,487,974
Massachusetts	5,061,250	2,416,752	483,634	339,993	1,820,871
Michigan	7,600,237	3,911,926	877,378	504,107	2,306,826
Minnesota	3,926,248	2,163,131	384,932	225,977	1,152,208
Mississippi	2,158,108	1,026,650	249,551	173,843	708,064
Missouri	4,473,226	2,350,357	543,473	314,138	1,265,258
Montana	736,558	389,664	94,569	48,657	203,668
Nebraska	1,326,139	733,118	136,179	88,631	368,211

State	Total	Married	Divorced	Widowed	Other marriage status
Nevada	1,887,203	923,518	268,622	105,948	589,115
New Hampshire	1,017,239	556,689	121,232	62,344	276,974
New Jersey	6,578,265	3,360,892	563,035	467,722	2,186,616
New Mexico	1,460,839	707,287	186,856	92,946	473,750
New York	14,925,614	6,678,549	1,325,138	1,045,881	5,876,046
North Carolina	6,839,619	3,555,099	719,334	467,785	2,097,401
North Dakota	496,341	278,409	42,037	35,270	140,625
Ohio	8,750,969	4,489,281	1,069,842	630,508	2,561,338
Oklahoma	2,709,105	1,437,187	359,631	198,659	713,628
Oregon	2,862,879	1,487,900	373,393	173,093	828,493
Pennsylvania	9,703,855	4,935,263	915,354	778,025	3,075,213
Rhode Island	823,719	386,466	91,287	61,141	284,825
South Carolina	3,355,523	1,662,171	349,433	252,066	1,091,853
South Dakota	600,114	328,201	63,865	42,161	165,887
Tennessee	4,683,923	2,435,291	609,650	339,623	1,299,359
Texas	17,185,928	8,749,535	1,920,476	1,007,953	5,507,964
Utah	1,826,327	1,065,668	174,478	76,061	510,120
Vermont	489,395	252,008	62,566	30,505	144,316
Virginia	5,890,260	3,045,131	586,202	365,648	1,893,279
Washington	4,918,282	2,566,946	623,640	269,354	1,458,342
West Virginia	1,424,575	779,622	172,125	122,299	350,529
Wisconsin	4,279,113	2,315,825	441,861	270,005	1,251,422
Wyoming	397,980	222,974	51,773	23,460	99,773
	227,279,008	114,378,135	25,025,452	15,057,609	72,817,812

Table 6: Population Totals by State: Educational Attainment (18+)

State	Less than HS degree	HS degree	Some college but no Bachelors degree	Bachelors Masters or professional degree
3,509,577	676,341	1,116,021	1,028,366	688,849
499,977	53,411	151,760	179,451	115,355
4,652,197	783,642	1,231,273	1,554,383	1,082,899
2,136,963	396,574	758,823	611,908	369,658
26,868,769	5,211,569	6,224,048	8,235,718	7,197,434
3,646,549	436,432	896,254	1,144,534	1,169,329
2,672,367	319,077	776,283	714,172	862,835
658,603	92,574	213,322	186,002	166,705
474,441	68,925	100,573	99,501	205,442
14,167,647	2,207,978	4,375,463	4,251,474	3,332,732
6,983,323	1,233,360	2,109,813	1,932,069	1,708,081
990,953	102,070	305,607	320,877	262,399

				Bachelors Masters or professional degree
State	Less than HS degree	HS degree	Some college but no Bachelors degree	
1,086,071	139,925	327,574	389,683	228,889
9,591,923	1,388,771	2,719,527	2,865,711	2,617,914
4,755,941	710,920	1,704,813	1,385,631	954,577
2,268,969	239,524	766,076	759,250	504,119
2,079,386	236,412	611,036	690,812	541,126
3,241,564	628,725	1,115,240	901,233	596,366
3,297,180	650,696	1,141,152	893,534	611,798
1,036,654	114,963	368,032	304,453	249,206
4,271,121	543,632	1,176,074	1,169,282	1,382,133
5,061,250	596,039	1,395,097	1,308,451	1,761,663
7,600,237	990,656	2,396,918	2,512,277	1,700,386
3,926,248	370,443	1,109,115	1,329,761	1,116,929
2,158,108	456,155	676,558	656,108	369,287
4,473,226	661,627	1,468,563	1,334,098	1,008,938
736,558	76,898	238,115	242,969	178,576
1,326,139	142,663	397,665	460,628	325,183
1,887,203	322,645	578,760	614,611	371,187
1,017,239	101,850	309,766	303,199	302,424
6,578,265	877,221	1,980,157	1,629,477	2,091,410
1,460,839	270,522	411,086	453,410	325,821
14,925,614	2,360,205	4,268,184	3,892,795	4,404,430
6,839,619	1,168,119	1,988,791	2,073,745	1,608,964
496,341	53,593	141,964	187,651	113,133
8,750,969	1,190,435	3,143,817	2,517,056	1,899,661
2,709,105	424,080	890,799	848,778	545,448
2,862,879	354,544	767,617	1,003,132	737,586
9,703,855	1,285,244	3,634,828	2,447,231	2,336,552
823,719	132,571	236,726	231,029	223,393
3,355,523	602,352	1,068,802	967,289	717,080
600,114	71,603	196,867	198,940	132,704
4,683,923	846,125	1,587,619	1,284,642	965,537
17,185,928	3,561,894	4,660,099	5,063,947	3,899,988
1,826,327	187,417	482,070	710,261	446,579
489,395	49,002	157,856	137,141	145,396
5,890,260	834,565	1,606,165	1,655,190	1,794,340
4,918,282	565,887	1,257,212	1,721,986	1,373,197
1,424,575	259,296	579,710	358,815	226,754
4,279,113	482,758	1,459,393	1,345,345	991,617
397,980	38,635	128,379	148,351	82,615
227,279,008	35,570,565	67,407,462	67,256,357	57,044,624

Recoded Variables

The following 21 variables are included in the final dataset. These variables are coded as described below.

Smokstatus_r: For respondents who reported smoking at least 100 cigarettes in their entire life (smok100), those who reported smoking everyday (smoknow) were recoded as “current everyday smokers”, those who reported smoking some days were recoded as “current some day smokers”, and those who reported not smoking at all were recoded as “former smokers”. Respondents who answered no to smoking at least 100 cigarettes in their entire life were recoded as “never smokers”. Those who answered “don’t know” or “refused” to smoknow or smok100 were recoded as “unknown”.

Smokstatus2_r: This recode combines levels from variable smokstatus_r. Those recoded as “current everyday smokers” and “current some day smokers” were combined into “current everyday or some day smokers.” Those recoded as “former smokers” and “never smokers” were combined into “never smoker or former smoker.” Those recoded “unknown” remained “unknown.”

Smokever_r: If respondents answered “yes” to question “Have you ever smoked 100 cigarettes in your entire life?” (smokever), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Mentholcigs2_r: This recode is non-missing for current smokers, defined as respondents who reported smoking at least 100 cigarettes in their entire life (smok100) and currently smoke everyday or some days (smoknow). Current smokers who reported smoking menthol cigarettes during past 30 days (mentholcigs2) were recoded “yes”. Else if they reported not smoking menthol cigarettes during the past 30 days, they were recoded “no”. Current smokers with missing, “don’t know” and “refused” responses were recoded as “unknown.”

Cigflavor_r: This recode is non-missing for current smokers, defined as respondents who reported smoking at least 100 cigarettes in their entire life (smok100) and currently smoke everyday or some days (smoknow). Current smokers who reported smoking flavored cigarettes during past 30 days (cigflavor) were recoded “yes”. Else if they reported not smoking flavored cigarettes during the past 30 days, they were recoded “no”. Current smokers with missing, “don’t know” and “refused” responses were recoded as “unknown.”

Sltever2_r: If respondents answered “yes” to question “Have you ever tried chewing tobacco, snuff, or dip, even just one time in your entire life?” (sltever2), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Currentslt_r: This recode is non-missing for respondents who answered “yes” or “no” to question “Have you ever tried chewing tobacco, snuff, or dip...?” (sltever2). Respondents who reported 1-30 days of chewing tobacco, snuff, or dip use (sltndays) were recoded as “yes”, respondents who answered “none” were recoded as “no” and “don’t know” and “refused” responses were recoded as “unknown.”

Snusever_r: If respondents answered “yes” to question “Have you ever tried snus, even just one time in your entire life?” (snusever), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Cigarever_r: If respondents answered “yes” to question “Have you ever tried smoking cigars, cigarillos, or very small cigars in your entire life, even one or two puffs?” (cigarever), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Currentcigar_r: This recode is non-missing for respondents who answered “yes” or “no” to question “Have you ever tried smoking cigars, cigarillos, or very small cigars that look like cigarettes in your entire life?” (cigarever). Respondents who reported 1-30 days of cigar use (cigarnodays) were recoded as “yes”, respondents who answered “none” were recoded as “no”, and “don’t know” and “refused” responses were recoded as “unknown.”

Pipewtrever_r: If respondents answered “yes” to question “Have you ever tried smoking tobacco in a water pipe in your entire life, even one or two puffs?” (pipewtrever), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Pipeothever_r: If respondents answered “yes” to question “Have you ever tried smoking tobacco in a pipe other than a water pipe in your entire life, even one or two puffs?” (pipeothever), recoded “yes”. If respondents answered “no”, recoded “no”. Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Qtlineawr_r: Non-users of tobacco (qtlineawrnt) and users (qtlineawrt) were asked “Are you aware of any telephone quitline services that are available to help people quit using tobacco?” Those who responded “yes” for variable qtlineawrnt or qtlineawrt were recoded “yes.” Those who responded “no” for variable qtlineawrnt or qtlineawrt were recoded as “no”. Else respondents were recoded as “unknown.”

Qtattempt_r: This recode is non-missing for current smokers, defined as respondents who reported smoking at least 100 cigarettes in their entire life (smok100) and currently smoke everyday or some days (smoknow). Current smokers who reported 1-76 times stopped smoking for one day or longer (qt12ms) were recoded as “yes”, those who answered “none” were recoded as “no”, and “don’t know” and “refused” were recoded as “unknown.”

Qtteffect_r: Respondents who answered “yes” to at least one of the following questions: “The last time you tried to quit smoking, did you call a telephone quitline?” (qtline), “The last time you tried to quit smoking, did you use a class or program, to help you quit?” (qtclasspgm), “The last time you tried to quit smoking, did you use one-on-one counseling from a health professional to help you quit?” (qtcounsel), or “The last time you tried to quit smoking, did you use any of the following medications...?” (qtmed2) were recoded “yes”. If respondents answered “no” to all above questions, recoded “no”. Else respondents were recoded as missing.

Qtwant_r: This recode is non-missing for current smokers, defined as respondents who reported smoking at least 100 cigarettes in their entire life (smok100) and currently smoke everyday or some days (smoknow). Current smokers who reported wanting to quit smoking cigarettes for good (qtwant) were recoded “yes”. Else if they reported not wanting to quit smoking cigarettes for good, they were recoded “no”. Missing, “don’t know” and “refused” responses were recoded as “unknown.”

Hcwaskadv_r: If respondents answered “yes” to questions “In the past 12 months did any doctor, dentist, nurse, or other health professional advise you to quit smoking cigarettes or using any other tobacco products?” (hchwadvise2) or “In the past 12 months, did any doctor, dentist, nurse, or other health professional ask if you smoke cigarettes or use any other tobacco products?” (hchwask) then respondents were recoded “yes.” If respondents answered “no” to both above questions or “no” to hchwask then respondents were recoded as “no.” Else respondents were recoded as “unknown.”

Homerules2_r: When asked “Not counting decks, porches, or garages, inside your home is smoking...?” (homerules2), respondents who answered “Always allowed” were recoded as “Always allowed.” Those who responded “Allowed only at some times or in some places” were recoded as “Allowed only at some times or in some places” and those who responded “Never allowed” were recoded as “Never allowed.” Missing values, “don’t know” and “unknown” responses were recoded “unknown.”

Worksmokind_r: When asked “At your workplace, is smoking in indoor areas...?” (worksmokind), respondents who answered “Always allowed” were recoded as “Always allowed.” Those who responded “Allowed only at some times or in some places” were recoded as “Allowed only at some times or in some places” and those who responded “Never allowed” were recoded as “Never allowed.” Missing values remained coded as missing values and “don’t know” and “unknown” responses were recoded “unknown.”

Worksmokout_r: When asked “At your workplace, is smoking in outdoor areas...?” (worksmokout), respondents who answered “Always allowed” were recoded as “Always allowed.” Those who responded “Allowed only at some times or in some places” were recoded as “Allowed only at some times or in some places” and those who responded “Never allowed” were recoded as “Never allowed.” Missing values remained coded as missing values and “don’t know” and “unknown” responses were recoded “unknown.”

Shsvehpol_r: When asked “Not counting motorcycles, in the vehicles that you or family members who live with you own or lease, is smoking ...?” (shsvehpol), respondents who answered “Always allowed in all vehicles” were recoded as “Always allowed in all vehicles.” Those who responded “Sometimes allowed in at least one vehicle” were recoded as “Sometimes allowed in at least one vehicle” and those who responded “Never allowed in any vehicle” were recoded as “Never allowed in any vehicle.” Missing values, “don’t know,” “Respondent’s family does not own or lease a vehicle,” and “unknown” responses were recoded “unknown.”

The following 14 variables are used in the weighing process:

SEQNO_C: this is the ID variable, a unique identifier for each respondent

STATERES: 2 digit state abbreviation

STATE_FIPS: State FIPS code

STRATUM: Identifies the stratum. The strata are combinations of county group and phone type

GEOSTRS_SAMP: County group—some states have only 1 county group

DENSTRS_SAMP: Phone type (1=listed, 2=unlisted, 9=cell)

CELL: indicates cell phone

IMP_EDUCATION: imputed education variable used in poststratification

IMP_MARITAL: imputed marital variable used in poststratification

IMP_RACE: imputed race variable used in poststratification

IMP_GENDER: imputed gender variable used in poststratification

IMP_AGE_CAT: imputed age category variable used in poststratification

WT_NATIONAL: Weight used for making national estimates

WT_STATE: Weight used for making state estimates—state with 200 or more cell phone respondents use both landline and cell phone respondents, whereas, states with less than 200 cell phone respondents only use landline respondents

WT_LANDLINE: Weight that uses only landline respondents

Formats

The weights will be delivered in a SAS data set containing one record for each respondent. In addition to weight variables, the imputed poststratification categories will be included.

The following are the formats for the various levels of the imputed poststratification variables.

```
proc format;
  value imp_age_catf
    1="18-24"
    2="25-34"
    3="35-44"
    4="45-54"
    5="55-64"
    6="65+";
  value imp_genderf
    1="Male"
    2="Female";
  value imp_racef
    1="White only(non-Hispanic)"
    2="Black only(non-Hispanic)"
    3="Hispanic"
    4="Other";
  value imp_maritalf
    1="Married"
    2="Divorced"
    3="Widowed"
    4="Other";
  value imp_education
    1="Less than HS degree"
    2="HS degree"

    4="BS or higher";
run;
```

References

AAPOR Cell Phone Task Force Report 2010, New Considerations for Survey Researchers When Planning and Conducting RDD Telephone Surveys in the U.S. With Respondents Reached via Cell Phone Numbers, P.J. Lavarakas, et al.

http://www.aapor.org/Cell_Phone_Task_Force_Report.htm

ACS 2011, The 2005-2009 ACS 5-Year Summary File Technical Documentation, American Community Survey Office, Data Products, Version 2, February 8, 2011.

http://www2.census.gov/acs2009_5yr/summaryfile/ACS_2005-2009_SF_Tech_Doc.pdf

Blumberg SJ, Luke JV, Ganesh N, et al. Wireless substitution: State-level estimates from the National Health Interview Survey, January 2007–June 2010. National health statistics reports; no 39. Hyattsville, MD: National Center for Health Statistics. 2011. <http://www.cdc.gov/nchs/data/nhsr/nhsr039.pdf>

Blumberg SJ, Luke JV, Davidson G, Davern ME, Yu T, Soderberg K, et al. Wireless substitution: State-level estimates from the National Health Interview Survey, January 2007–December 2007. National health statistics reports; no 14. Hyattsville, MD: National Center for Health Statistics. 2009.

Izrael D, Battaglia MP, Frankel MR, Extreme Survey Weight Adjustment as a Component of Sample Balancing (a.k.a Raking), Paper 247-2009, SAS Global Forum 2009.

<http://support.sas.com/resources/papers/proceedings09/247-2009.pdf>

Leeuw ED, Hox J, Dillman D, International Handbook of Survey Methodology, Psychology Press, 2008.

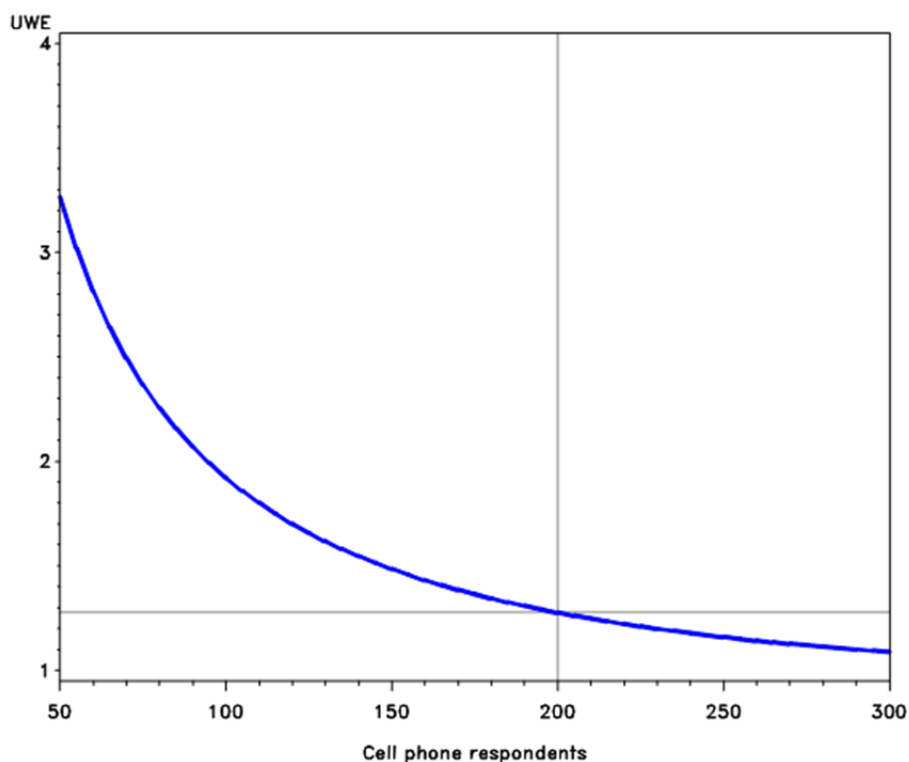
Kish L, Survey Sampling, Wiley-Interscience Publication, 1965.

Appendix A: Cutoff of 200 cell phone respondents

For states with very few cell phone respondents, including cell phone respondents would result in large unequal weighting effects, and consequently, large variance for population estimates. For example, In Alaska there were only 13 cell phone respondents and 20% of Alaskans are cell phone only.

Consequently, if we included cell phones, we would use 13 respondents to make estimates on 20% of the population of Alaska. The resulting unequal weighting effect (UWE) exclusively attributed to this disproportionate sampling of cell phones would be 6.3. This UWE will be multiplied by the other causes of UWE resulting in a huge UWE and poor precision of population estimates. To avoid this problem we do not include cell phone respondents when we make state level estimates for Alaska.

We are tasked with creating a cutoff point when we will accept the increase in variance, from the under sample of cell phones to avoid the coverage error from excluding cell phones. We choose 200. This number is arbitrary, as any cutoff will be, but, it is not without good reason. Each state has a different distribution of cell-only and landline population members. But, if a state had 1,863 total respondents and 27% cell-only population percentage then the graph below displays the increase in UWE attributed exclusively to the disproportionate sampling of cell phones as a result of sampling cell phones as the number of cell phones goes from 50 to 300. Consequently, selecting a lower bound of 200 limits the increase in UWE to about 1.28. In this example the UWE is only the UWE attributed to the disproportionate sampling of cell phone only and landline subjects. That is, it does not account for the other sources of UWE.



Appendix B: Initial weights

State FIPS	Geographical stratum	Phone type Listed=1; unlisted=2; cell=9	n selected	Frame count	WT1	Unknown adjustment	WT1 with Unknown adjustment
1	1	1	15,957	1,353,056	84.8	0.408	34.6
1	1	2	24,753	3,146,019	127.1	0.028	3.6
1	1	9	3,503	9,010,000	2572.1	0.097	249.1
2	2	1	4,876	94,337	19.3	0.580	11.2
2	2	2	26,024	753,303	28.9	0.078	2.3
2	2	9	594	1,586,000	2670.0	0.086	229.4
4	3	1	11,068	1,199,738	108.4	0.464	50.3
4	3	2	25,292	4,022,899	159.1	0.056	8.8
4	3	9	4,723	7,476,000	1582.9	0.182	288.4
5	4	1	18,648	695,778	37.3	0.463	17.3
5	4	2	32,862	1,840,991	56.0	0.030	1.7
5	4	9	2,148	4,925,000	2292.8	0.101	231.3
6	5	1	10,818	6,294,268	581.8	0.448	260.6
6	5	2	28,512	24,875,620	872.5	0.086	75.1
6	5	9	44,593	45,938,000	1030.2	0.209	215.4
8	6	1	10,420	1,136,197	109.0	0.471	51.4
8	6	2	22,580	3,691,218	163.5	0.043	7.1
8	6	9	6,169	6,419,000	1040.5	0.195	202.4
9	7	1	15,357	1,065,623	69.4	0.464	32.2
9	7	2	22,593	2,350,477	104.0	0.037	3.8
9	7	9	4,458	4,104,000	920.6	0.176	161.8
10	8	9	1,166	1,153,000	988.9	0.154	152.4
10	9	1	2,972	41,987	14.1	0.472	6.7
10	9	2	3,748	80,231	21.4	0.050	1.1
10	10	1	7,571	149,352	19.7	0.437	8.6
10	10	2	13,189	389,120	29.5	0.036	1.1
10	11	1	3,448	69,002	20.0	0.472	9.5
10	11	2	4,022	123,141	30.6	0.067	2.0
11	12	1	11,739	153,894	13.1	0.392	5.1
11	12	2	40,491	795,506	19.6	0.027	0.5
11	12	9	875	1,494,000	1707.4	0.235	400.8
12	13	1	16,165	4,810,088	297.6	0.391	116.3
12	13	2	29,945	13,372,612	446.6	0.039	17.3
12	13	9	23,388	24,916,000	1065.3	0.175	186.5
13	14	9	11,685	14,560,000	1246.0	0.159	197.8
13	15	1	2,461	68,755	27.9	0.319	8.9
13	15	2	4,469	186,351	41.7	0.024	1.0
13	16	1	1,851	129,988	70.2	0.417	29.3
13	16	2	3,549	373,512	105.2	0.037	3.9
13	17	1	1,922	223,103	116.1	0.394	45.7
13	17	2	3,628	630,474	173.8	0.022	3.8
13	18	1	1,780	155,560	87.4	0.385	33.7
13	18	2	4,520	597,340	132.2	0.017	2.3
13	19	1	2,310	123,272	53.4	0.466	24.9
13	19	2	3,900	313,843	80.5	0.035	2.8

State FIPS	Geographical stratum	Phone type Listed=1; unlisted=2; cell=9	n selected	Frame count	WT1	Unknown adjustment	WT1 with Unknown adjustment
13	20	1	2,207	263,760	119.5	0.371	44.3
13	20	2	3,523	632,938	179.7	0.028	5.0
13	21	1	1,625	245,146	150.9	0.361	54.4
13	21	2	4,885	1,103,367	225.9	0.015	3.4
13	22	1	2,220	216,674	97.6	0.412	40.2
13	22	2	2,730	400,402	146.7	0.035	5.1
13	23	1	2,048	132,768	64.8	0.466	30.2
13	23	2	2,962	289,595	97.8	0.041	4.0
13	24	1	1,784	110,565	62.0	0.468	29.0
13	24	2	2,926	271,935	92.9	0.038	3.5
13	25	1	2,297	154,751	67.4	0.442	29.8
13	25	2	2,683	273,046	101.8	0.045	4.6
13	26	1	1,923	111,513	58.0	0.482	27.9
13	26	2	2,427	212,272	87.5	0.034	3.0
13	27	1	2,192	139,654	63.7	0.423	27.0
13	27	2	2,938	281,774	95.9	0.036	3.4
13	28	1	1,520	34,062	22.4	0.516	11.6
13	28	2	2,860	95,938	33.5	0.048	1.6
13	29	1	1,930	60,959	31.6	0.486	15.4
13	29	2	5,649	150,747	26.7	0.048	1.3
13	30	1	1,784	83,876	47.0	0.514	24.2
13	30	2	2,956	212,942	72.0	0.051	3.7
13	31	1	1,764	95,530	54.2	0.443	24.0
13	31	2	3,426	278,913	81.4	0.031	2.5
13	32	1	1,790	88,063	49.2	0.441	21.7
13	32	2	4,000	295,437	73.9	0.030	2.2
15	33	1	11,133	186,839	16.8	0.422	7.1
15	33	2	42,627	1,072,851	25.2	0.053	1.3
15	33	9	1,757	1,566,000	891.3	0.264	235.2
16	34	1	9,723	315,850	32.5	0.461	15.0
16	34	2	21,417	1,035,850	48.4	0.041	2.0
16	34	9	1,908	2,117,000	1109.5	0.143	158.5
17	35	1	12,971	3,066,102	236.4	0.446	105.5
17	35	2	28,129	9,965,150	354.3	0.032	11.3
17	35	9	15,915	19,360,000	1216.5	0.146	178.2
18	36	1	10,777	1,818,823	168.8	0.507	85.5
18	36	2	17,243	4,358,395	252.8	0.031	7.7
18	36	9	7,874	9,020,000	1145.5	0.156	178.7
19	37	1	11,406	865,906	75.9	0.521	39.5
19	37	2	22,314	2,535,694	113.6	0.019	2.2
19	37	9	3,805	5,039,000	1324.3	0.113	149.9
20	38	1	9,432	741,140	78.6	0.525	41.3
20	38	2	16,308	1,920,660	117.8	0.031	3.6
20	38	9	3,497	4,923,000	1407.8	0.087	122.3
21	39	1	12,283	1,159,487	94.4	0.493	46.5
21	39	2	21,137	2,980,215	141.0	0.030	4.2
21	39	9	5,413	6,705,000	1238.7	0.117	144.3
22	40	9	27,979	8,160,000	291.6	0.097	28.4

State FIPS	Geographical stratum	Phone type Listed=1; unlisted=2; cell=9	n selected	Frame count	WT1	Unknown adjustment	WT1 with Unknown adjustment
22	41	1	5,404	195,909	36.3	0.424	15.4
22	41	2	17,366	944,591	54.4	0.016	0.9
22	42	1	6,575	168,967	25.7	0.465	11.9
22	42	2	12,655	486,978	38.5	0.026	1.0
22	43	1	6,597	102,502	15.5	0.455	7.1
22	43	2	12,783	298,908	23.4	0.024	0.6
22	44	1	6,512	148,065	22.7	0.435	9.9
22	44	2	11,398	388,981	34.1	0.024	0.8
22	45	1	5,540	79,262	14.3	0.464	6.6
22	45	2	10,600	228,510	21.6	0.020	0.4
22	46	1	6,027	80,502	13.4	0.443	5.9
22	46	2	10,803	217,539	20.1	0.027	0.5
22	47	1	6,097	156,726	25.7	0.419	10.8
22	47	2	11,303	436,249	38.6	0.025	1.0
22	48	1	6,469	100,771	15.6	0.406	6.3
22	48	2	12,011	280,430	23.3	0.026	0.6
22	49	1	6,379	145,465	22.8	0.468	10.7
22	49	2	8,231	283,811	34.5	0.034	1.2
23	50	1	9,673	400,943	41.4	0.494	20.5
23	50	2	17,987	1,101,275	61.2	0.047	2.9
23	50	9	1,822	1,747,000	958.8	0.127	121.5
24	51	1	12,854	1,563,478	121.6	0.441	53.6
24	51	2	22,366	4,077,838	182.3	0.040	7.2
24	51	9	7,098	7,453,000	1050.0	0.188	197.8
25	52	1	12,820	1,996,144	155.7	0.471	73.3
25	52	2	18,200	4,240,129	233.0	0.033	7.8
25	52	9	8,364	9,044,000	1081.3	0.156	168.2
26	53	1	12,496	2,933,616	234.8	0.427	100.3
26	53	2	22,544	7,917,843	351.2	0.024	8.3
26	53	9	12,460	16,169,000	1297.7	0.130	169.3
27	54	1	9,871	1,486,305	150.6	0.512	77.1
27	54	2	18,749	4,228,188	225.5	0.028	6.3
27	54	9	6,531	7,398,000	1132.8	0.150	169.4
28	55	1	16,751	688,174	41.1	0.375	15.4
28	55	2	29,419	1,810,598	61.5	0.028	1.7
28	55	9	3,613	5,865,000	1623.3	0.073	117.7
29	56	1	10,195	1,665,251	163.3	0.508	83.0
29	56	2	16,595	4,061,560	244.7	0.036	8.9
29	56	9	7,372	8,151,000	1105.7	0.149	165.2
30	57	1	7,805	222,714	28.5	0.518	14.8
30	57	2	19,465	826,553	42.5	0.048	2.0
30	57	9	1,307	1,987,000	1520.3	0.072	109.1
31	58	1	11,043	454,837	41.2	0.531	21.9
31	58	2	23,667	1,460,663	61.7	0.022	1.4
31	58	9	2,200	2,932,000	1332.7	0.095	126.6
32	59	1	11,144	449,045	40.3	0.432	17.4
32	59	2	26,926	1,615,755	60.0	0.072	4.3
32	59	9	3,266	3,250,000	995.1	0.218	216.6

State FIPS	Geographical stratum	Phone type Listed=1; unlisted=2; cell=9	n selected	Frame count	WT1	Unknown adjustment	WT1 with Unknown adjustment
33	60	1	10,635	398,721	37.5	0.479	18.0
33	60	2	16,545	923,368	55.8	0.046	2.5
33	60	9	1,763	1,862,000	1056.2	0.110	116.0
34	61	1	24,695	2,255,431	91.3	0.443	40.5
34	61	2	51,685	7,080,669	137.0	0.047	6.5
34	61	9	78,380	11,830,000	150.9	0.148	22.4
35	62	1	11,299	376,304	33.3	0.437	14.6
35	62	2	26,351	1,304,660	49.5	0.048	2.4
35	62	9	2,516	2,750,000	1093.0	0.167	182.2
36	63	1	12,631	4,575,154	362.2	0.461	166.8
36	63	2	26,549	14,371,041	541.3	0.045	24.2
36	63	9	24,615	25,435,000	1033.3	0.198	205.1
37	64	1	11,707	2,405,509	205.5	0.454	93.3
37	64	2	20,153	6,188,794	307.1	0.042	13.0
37	64	9	11,455	12,788,000	1116.4	0.160	178.1
38	65	1	10,488	178,662	17.0	0.571	9.7
38	65	2	24,132	611,216	25.3	0.028	0.7
38	65	9	883	1,738,000	1968.3	0.091	180.0
39	66	1	10,398	3,054,737	293.8	0.521	153.2
39	66	2	20,832	9,170,081	440.2	0.038	16.6
39	66	9	14,351	16,113,000	1122.8	0.160	179.1
40	67	1	18,005	916,510	50.9	0.503	25.6
40	67	2	32,305	2,463,593	76.3	0.047	3.5
40	67	9	28,254	6,109,000	216.2	0.113	24.5
41	68	1	7,739	778,542	100.6	0.516	51.9
41	68	2	19,351	2,909,148	150.3	0.055	8.3
41	68	9	4,860	4,620,000	950.6	0.211	201.0
42	69	1	16,852	3,587,338	212.9	0.521	110.9
42	69	2	26,888	8,582,049	319.2	0.051	16.2
42	69	9	15,861	15,949,000	1005.5	0.150	151.1
44	70	1	11,908	304,734	25.6	0.530	13.6
44	70	2	16,832	646,460	38.4	0.045	1.7
44	70	9	1,398	1,261,000	902.0	0.185	167.0
45	71	9	5,632	6,261,000	1111.7	0.146	162.1
45	72	1	4,924	129,319	26.3	0.498	13.1
45	72	2	7,616	301,657	39.6	0.039	1.6
45	73	1	4,935	252,199	51.1	0.474	24.2
45	73	2	7,545	579,981	76.9	0.033	2.5
45	74	1	4,512	249,242	55.2	0.503	27.8
45	74	2	7,428	614,959	82.8	0.043	3.5
45	75	1	5,311	143,344	27.0	0.490	13.2
45	75	2	8,999	365,514	40.6	0.048	1.9
45	76	1	5,189	84,071	16.2	0.503	8.1
45	76	2	8,791	215,810	24.5	0.048	1.2
45	77	1	5,372	108,613	20.2	0.497	10.0
45	77	2	10,678	326,215	30.6	0.041	1.2
45	78	1	4,740	157,239	33.2	0.461	15.3
45	78	2	8,550	426,317	49.9	0.040	2.0

State FIPS	Geographical stratum	Phone type Listed=1; unlisted=2; cell=9	n selected	Frame count	WT1	Unknown adjustment	WT1 with Unknown adjustment
45	79	1	4,973	70,898	14.3	0.467	6.7
45	79	2	9,307	200,872	21.6	0.057	1.2
46	80	1	10,159	215,449	21.2	0.522	11.1
46	80	2	25,031	791,278	31.6	0.020	0.6
46	80	9	1,104	1,602,000	1451.1	0.105	151.8
47	81	1	13,116	1,718,449	131.0	0.453	59.4
47	81	2	20,784	4,088,561	196.7	0.029	5.7
47	81	9	7,819	9,003,000	1151.4	0.180	207.6
48	82	1	15,817	5,576,892	352.6	0.387	136.6
48	82	2	29,153	15,392,508	528.0	0.031	16.4
48	82	9	28,696	34,521,000	1203.0	0.179	215.2
49	83	1	10,418	507,525	48.7	0.495	24.1
49	83	2	20,422	1,492,375	73.1	0.038	2.8
49	83	9	3,203	3,343,000	1043.7	0.177	184.3
50	84	1	8,003	184,063	23.0	0.531	12.2
50	84	2	16,117	549,729	34.1	0.056	1.9
50	84	9	894	843,000	943.0	0.104	98.3
51	85	9	9,697	10,694,000	1102.8	0.167	184.2
51	86	1	2,731	313,760	114.9	0.510	58.6
51	86	2	3,779	647,340	171.3	0.053	9.1
51	87	1	3,144	565,530	179.9	0.519	93.4
51	87	2	6,366	1,732,370	272.1	0.033	8.9
51	88	1	2,682	369,060	137.6	0.532	73.3
51	88	2	4,128	848,142	205.5	0.041	8.5
51	89	1	2,702	375,428	138.9	0.453	63.0
51	89	2	4,708	984,117	209.0	0.031	6.5
51	90	1	2,748	440,018	160.1	0.488	78.1
51	90	2	5,232	1,251,769	239.3	0.036	8.6
53	91	1	9,724	1,512,291	155.5	0.479	74.4
53	91	2	21,026	4,896,262	232.9	0.049	11.3
53	91	9	8,197	7,838,000	956.2	0.221	211.0
54	92	1	9,651	442,605	45.9	0.531	24.4
54	92	2	15,519	1,066,514	68.7	0.075	5.2
54	92	9	2,448	2,235,000	913.0	0.117	107.0
55	93	1	10,109	1,710,594	169.2	0.523	88.4
55	93	2	16,021	4,061,045	253.5	0.027	6.8
55	93	9	7,097	8,704,000	1226.4	0.100	122.6
56	94	1	8,522	113,289	13.3	0.494	6.6
56	94	2	21,028	414,717	19.7	0.043	0.9
56	94	9	743	1,601,000	2154.8	0.063	135.0