Optimization Targeting of the Annual Refiling Survey October 2015

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Abstract

The Annual Refilling Survey (ARS) is a survey of US business establishments that is part of the Quarterly Census of Employment and Wages (QCEW) program. The purpose of the ARS survey is to ensure the correct North American Industrial Classification System (NAICS) is assigned to those establishments listed on the QCEW longitudinal data base. A complete census of the entire population of establishments takes 3 years. On an annual basis 1/3 of all QCEW establishments are surveyed, with the sample chosen from those establishments which did not participate in the prior 2 years.

One measure of the quality of the refiling survey is the estimated percent of establishments misclassified by NAICS. We present an alternative survey technique based on operations research methodology which can potentially reduce the number of ARS surveyed establishments by up to 30% while significantly improving the quality of the survey with respect to the accuracy of industrial classification.

Key Words: NAICS, Quarterly Census of Employment and Wages, Census, Annual Refiling Survey

Introduction

The *Annual Refiling Survey* is a sample survey of establishments performed as a part of the Quarterly Census of Employment and Wages program. The main purpose of the ARS is to ensure correct NAICS classification for all establish ments listed on the QCEW longitudinal data base (QCEW-LDB). The ac curacy of NAICS classification is important for implementation of BLS programs which use NAICS classification, which currently are: Current Employment Statistics, (CES), Job Openings and Turnover Survey (JOLTS), Local Area Unemp loyment Statistics (LAUS), Business Employment Dynamics, (BED), and Occupational Safety and Health Survey (OSHS). Most of these programs produce leading economic indicators on monthly basis.

Presently approximately a third of establishments on the QCEW-LDB are sampled every year. This sample rate is uniform across N AICS clas sification. Thus a c omplete survey of the whole population of establishments takes 3 y ears. All QCEW-LDB establishments become part of the current sample provided they did not participate in the survey in the prior two samples.

A sufficiently high sam ple rate in a given year is important to maintain the accura cy of NAICS classification for the following year. Establishments which are not sampled and have changed their NAICS classification are left in the QCEW-LDB with their old classification in error. At the same

time, the cost of the survey depends on the number of sampled respondents. Both accuracy and cost depend on the sample rate.

It has been observed by BLS staff that the frequency of NAICS classification changes is not uniform across all NAICS classifications; rather the frequency of NAICS classification changes varies considerably by NAICS. This observation suggests that it may be worthwhile to adjust the sampling rate for individual NAICS classifications and thereby improve the quality of the survey as well as reducing its cost. In the industries where NAICS changes more frequently, the rate of sampling can be increased, while in the industries where NAICS changes less frequently the rate of sampling can be decreased.

We recommend that the adjustment of sam pling rate be d one in t hree different modes: establishments that are sampled annually, establishments that are sampled every three years, and establishments that are sam pled every six y ears. Thus, establishments in so me industries are recommended to be sampled more frequently, in some industries at the current rate, and in so me industries at a lesser rate. Establishments that are designated "new births" in the current year are to be sampled at the same rate as the existing establishments within the same NAICS code.

The method to be presented requires that every year the sample rate be recalculated for each 6 digit NAICS. The computation of sampling rates is to be performed for all NAICS separately for each state. The rates are to be computed for each state. In order to compute the rates, only four types of information are required

1. The total cost of the survey for a given state.

:

- 2. Total number of establishments for each NAICS for the current year on the QCEW-LDB.
- 3. Total number of establishments which moved from a given NAICS in the previous year to different NAICS in the current year on the QCEW-LDB.
- 4. The prior year recommended sampling rates for all NAICS.

Sampling rates from the previous year are known and need not to be re-computed. The frequency of NAICS changes between previous year and the current year can be directly computed from the QCEW-LDB.

To find *optimum* rates for each NAICS the m ethod uses *NAICS classification error for establishments*. The *NAICS classification error for establishments* is defined as a percent of all establishments which are misclassified by NAICS code.

The sampling rate for individual NAICS are adjusted to minimize the projected state-level NAICS classification error for the following year. The survey is optimized in the sense that, for a given fixed cost of the survey, the sampling rates are adjusted in such a way that the projected state-level NAICS classification error for establishments is minimized.

The presented methodology can be applied whenever a census is taken in parts over an extended period of time. The generalization of this methodology is beyond the scope of this paper.

Outline of the Paper

Part One. We introduce notation, develop methodology for estimating current year NAICS classification error given previous year rates of sampling, and develop methodology for estimating projected NAICS classification error for the next year given current year rates of sampling.

Part Two. We provide analysis of the problem, develop equations which are used for finding solutions, and provide a proof that the methodology for finding rates for a current year survey is indeed *optimal*. The sample rates are optimal in the sense that, for a given fixed cost of the survey, and other constrains of the survey the projected NAICS classification error is minimized. **Part Three.** We present 4 algorithms for finding an optimal solution, each achieving the desired result under given preferences which are selected by the user.

Part Fourth. We present tables with results from the first two algorithms for each state.

1. Methodology

1.1 Notation

 yr_t – current year: the year and a quarter in which the ARS is taken.

 yr_{t-1} – previous year: the preceding year and a quarter in which the ARS is taken.

 yr_{t+1} – next year: the following year and a quarter in which the ARS is taken.

 $N_t(naics)$ – true number of establishments existing in the year t, for a given NAICS.

 $N_t^{LDB}(naics)$ – number of establishments existing in the year t, for a given NAICS,

as it is recorded in QCEW-LDB.

 $\Delta_{t-1,t}(naics)$ – number of establishments classified by naics in the previous year *t*-1, with a different classification in the current year *t*.

 $\Delta_{t-1,t}^{LDB}$ (naics) – number of establishments classified by naics in the previous year t-1,

with different classification in the current year *t*, **as it is recorded in QCEW-LDB.** $\Delta_{t-1,t}^{\Lambda LDB}(naics)$ – number of establishments classified by NAICS in the previous year

with different classification in the currentyear t that are **not recorded on QCEW-LDB.** These are establishments which in the previous year t-1 had classification NAICS, and changed their classification in the current year, t, but were not given this new classification due to the fact that they were not interviewed in ARS.

From the above definitions it immediately foll ows that the total number of establishments which changed NAICS in the current year from the previous year is the sum of establishments which are recorded to change their NAICS, plus the establishments which are not recorded to change their NAICS, since they are not in the survey:

$\Delta_{t-1,t}(naics) = \Delta_{t-1,t}^{LDB}(naics) + \Delta_{t-1,t}^{\wedge LDB}(naics).$

Note that the quantity $\Delta_{t-1,t}^{LDB}(naics)$ may be directly found from the QCEW-LDB, whereas the quantities $\Delta_{t-1,t}(naics)$ and $\Delta_{t-1,t}^{\wedge}(naics)$ may only be estimated. The estimate of the second of these quantities is particularly important for this project. The value of $\Delta_{t-1,t}^{\wedge LDB}(naics)$ represents

a total of establish ments which are m isclassified by NAICS. An esti mate of this total will be denoted as $\Delta_{t-1,t}^{\wedge LDB}(naics)$.

Let ω denote sample ratio of establishments which are included in the survey. For example: if for a given year *t*, and a given NAICS, sa mple selection occurs annually, then $\omega_t(naics) = 1$, if sample selection occurs every 3 years, then $\omega_t(naics) = 1/3$, if sample selection occurs every 6 years, then $\omega_t(naics) = 1/6$. Note that under the current sy stem $\omega_t(naics) = 1/3$ for every NAICS, and every year *t*.

Similarly, define $\omega_{t-1}(naics)$ and $\omega_t(naics)$ as sample ratios of surve ys in the previous and current year respectively. Note that $\omega_{t-1}(naics)$ is always known from the past while $\omega_t(naics)$ is to be determined.

Let **Cost** denote the cost of one individual survey for only one establishment in a given state. Let **Total_Cost** denote the total cost of the survey in a given state. The **Total_Cost** is a total of all funds which are available in a given state to perform the survey.

1.2 Method to Measure Accuracy of NAICS Classification

The accuracy of NAICS classification for all establishments in a given state can be measured by estimating a percent of establishments with an in correct NAICS classification to the total of all establishments.

The percent of establishments which are **incorrectly** classified in the current year by NAICS code will be called the *NAICS classification error rate for establishments* calculated using the following formula:

(1)
$$naics_error(yr_t) = \frac{\sum_{\{NAICS\}} \Delta_{t-1,t}^{\wedge LD\overline{B}}(naics)}{\sum_{\{NAICS\}} N_t(naics)}$$

Note that the denom inator in (1) does not depend on whether est ablishments are summed with correct or incorrect NAICS, all that matters here is the total sum of establishments, calculated using the following formula:

$$\sum_{NAICS} N_t(naics) = \sum_{NAICS} N_t^{LDB}(naics).$$

Therefore, formula (1) can be written as follows:

(2)
$$naics_error(yr_t) = \frac{\sum_{\{NAICS\}} \Delta_{t-1,t}^{\land LDB}(naics)}{\sum_{\{NAICS\}} N_t^{LDB}(naics)}.$$

The denominator in (2) can be foun d directly in QCEW-LDB, The num erator part needs to be estimated.

Similarly, *NAICS classification error for employment* can be defined as a ratio of m isclassified employment to the total of all e mployment. The f ormulas for employment classification are analogous. In this work we limit our attention to the *NAICS classification error for establishments*.

1.3 Estimation of *naics_error* for the Current Year

In order to e stimate the value of naics_error in current year data it is necessary to estimate $\Delta_{t-1,t}^{\Lambda LDB}$ (naics) for each NAICS.

In the current survey a third of all establishments are sampled annually. Hence only one third of all changes of NAICS are being detected by the survey. Therefore, under the current system of survey the total number of establish ments which are changing NAICS classification in the current year, yr_t , can be estimated (Horvitz-Thompson estimator) as follows:

(3)
$$\Delta_{t,t-1}(naics) = 3 \times \Delta_{t-1,t}^{LDB}(naics).$$

The right hand side in (3) can be directly found from the QCEW-LDB.

In particular, in the current survey, the total number of establishments with changed NAICS which are not recorded in the QCEW-LDB can be estimated as the total of establishments which changed NAICS less the total of known establishments known from QCEW-LDB which changed NAICS. By formula:

(4)
$$\Delta_{t-1,t}^{\widehat{\Lambda LDB}}(naics) = \Delta_{t-1,t}(naics) - \Delta_{t,-1t}^{LDB}(naics) = 2 \times \Delta_{t-1,t}^{LDB}(naics).$$

Formula (4) implies that using the current sa mpling rate the n umber of est ablishments with incorrect NAICS can be estimated by the known number from QCEW-LDB changes, and it is twice as high as the known (from LDB) num ber of changes. (Infrequently, when the total number of establishments in the current quarter is less than $2 \times \Delta_{t-1,t}^{LDB}$ (naics), this estimation needs to be reduced to the number of total number of establishments in the current quarter.)

This result can be generalized for a situation where the sampling ratio $\omega_{t-1}(naics)$ is not equal to 1/3, but it is an arbitrary number from the interval (0, 1).

Following the same arguments as above, it can be shown that for an arbitrary ratio $\omega_{t-1}(naics)$ a Horvitz - Thompson estimate will yield that the total of establishments which changed their NAICS is

$$\widehat{\Delta_{t-1,y}(naics)} = \frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)$$

and the total number of establishments with incorrect NAICS classification in a current quarter can be estimated by the following formula:

(6)
$$\Delta_{t-1,t}^{\widehat{\Lambda LDB}}(naics) = \left(\frac{1}{\omega_{t-1}(naics)} - 1\right) \times \Delta_{t-1,t}^{LDB}(naics).$$

The NAICS classification error rate, $naics_error(yr_t)$, as defined in equation (1) can now be estimated from (2) and (5) as follows:

(7)
$$naics_error(yr_t) = \frac{\sum_{\{NAICS\}} (\frac{1}{\omega_{t-1}(naics)} - 1) \times \Delta_{t-1,t}^{LDB}(naics)}{\sum_{\{NAICS\}} N_t^{LDB}(naics)}$$

In particular, under the current sampling design:

(8)
$$naics_error_current(yr_t) = \frac{\sum_{\{NAICS\}} 2 \times \Delta_{t-1,t}^{LDB}(naics)}{\sum_{\{NAICS\}} N_t^{LDB}(naics)}$$
.

1.4 Estimation of Variance for the Current Year Estimator of naics_error

Having estimates of naics error without having estimates of their variance is not very useful. It can be shown that variance for naics error in (6) can be estimated by:

$$(9) \quad \widehat{V}(naics_error(yr_t)) = \left(\frac{1}{\sum_{\{\text{NAICS}\}} N_t^{LDB}(naics)}\right)^2 \sum_{\{\text{NAICS}\}} \left\{ (1 - \omega_{t-1}(naics)) \times (N_t^{LDB}(naics))^2 \\ \times \frac{\hat{p}_t(naics) \times (1 - \hat{p}_t(naics))}{\omega_{t-1}(naics) \times (N_t^{LDB}(naics) - 1)} \right\}$$

where

$$\hat{p}_t(naics) = \left(\frac{1}{\omega_{t-1}(naics)} - 1\right) \times \frac{\Delta_{t-1,t}^{LDB}(naics)}{N_t^{LDB}(naics)}.$$

In particular, under current sam pling design with $\omega_{t-1}(naics) = 1/3$, the estimator of variance of estimator in (8) is derived using:

(10)
$$\widehat{V}(naics_error(yr_t)) = \left(\frac{1}{\sum_{\{\text{NAICS}\}} N_t^{LDB}(naics)}\right)^2 \sum_{\{\text{NAICS}\}} \left\{2/3 \times (N_t^{LDB}(naics))^2 \times \frac{\hat{p}_t(naics) \times (1 - \hat{p}_t(naics))}{(1/3 \times N_t^{LDB}(naics) - 1)}\right\}.$$

1.5 Past NAICS error estimate and its variance

Note that the main goal of this project is NOT to find a design which minimize $\widehat{V(nacc_s_error)}$, as it is commonly performed in survey sampling. The goal is to minimize $nacc_s_error$. Note that

the top priority is minimization of this error rate, with estimation of the error rate as a secondar y goal. The design of the survey is to be made with the idea of achi eving $naics_error$ as s mall as possible. Also, note that $naics_error$ needs to be minimized NOT for the current year, but for the subsequent year. Thus, form ulas (7) and (10) can only serve to compute $naics_error$ and its variance in the current year and thereby serves to evaluate how effective the survey was in the preceding year. Specifical ly, the value $naics_error(yr_t)$ may only be used to evaluate how optimal was the choice of sampling rates ω for the preceding year.

The following section details the projection of the estimated naics error in the next year, given the selection of sampling rates ω in the current year for the survey.

1.6 Projected naics_error

The selection of sampling rates in the current sample year has its impact on the accuracy of NAICS classification for the following year. The projected NAICS classification error for the next year, given a specific selection of sampling rates ω in the current year is performed below. For projected NAICS classification error in the next year, using similar arguments as previously, the equations (5), (6) and (7) cited earlier can be adopted with small changes as follows:

(11)
$$\widehat{\Delta_{t,t+1}^{\wedge LDB}}(naics) = (1 - \omega_t(naics)) \times \Delta_{t,t+1}(naics)$$

where, $\omega_t(naics)$ is a rate for the current year.

The num ber of all establish ments that changed NAICS classification in the next year can be estimated by the estimate obtained from the previous year, by formula

(12)
$$\Delta_{t,t+1}(naics) = \Delta_{t-1,t}(naics)$$

The accuracy of estimation in (12) is outside the scope of this work and will not be performed here. From (11) it follows that:

(13)
$$\Delta_{t,t+1}^{\wedge LDB}(naics) = (1 - \omega_t(naics)) \times \Delta_{t-1,t}(naics)$$

then, as previously:

(14)
$$\Delta_{t,t+1}^{\wedge LDB}(naics) = (1 - \omega_t(naics)) \times \frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)$$

and thus the projected NAICS classification error for the next year is derived as follows:

(15)
$$naics_error(yr_{t+1}) = \frac{\sum_{\{NAICS\}} (1 - \omega_t(naics)) \times \frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)}{\sum_{\{NAICS\}} N_t^{LDB}(naics)}$$

where $\omega_{t-1}(naics)$ a rate used in the previous year, and $\omega_t(naics)$ is a rate to be used in the current year. For the purpose of optimization for the next year formula (15) will be used.

2. Optimization of the Survey

2.1 Problem of Finding Sampling Rates

The initial purpose of this project was t o design a survey which would produce the lowest error with the least resources. Finding a design which produces lower error means finding a set of rates $\{\omega_t(naics): naics \in NAICS\}$ which c onsume less resources while yielding t he smallest $naics_error(yr_{t+1})$. In this sense the process of finding the best rates at a given cost can be called survey optimization. The formula for projected NAICS error is given in (15). Ne xt, the formula for the survey cost is derived. The total is the cost of one individual unit multiplied by the number of survey units

(16) Total Cost =
$$C \times \sum_{\{NAICS\}} \omega_t(naics) \times N_t^{LDB}(naics)$$

where C is the cost of one individual survey, and $\omega_t(naics) \in \{1/6, 1/3, 1\}$. Therefore, we have arrived at the following optimization problem:

Find { ω_t (*naics*): *naics* \in *NAICS* } such that

MINIMIZE:

(•) naics_error
$$(yr_{t+1}) = \frac{\sum_{\{NAICS\}}(1-\omega_t(naics)) \times \frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)}{\sum_{\{NAICS\}} N_t^{LDB}(naics)}$$

under the following constrains:

(••) Total Cost =
$$C \times \sum_{\{NAICS\}} \omega_t(naics) \times N_t^{LDB}(naics)$$

and

(•••)
$$\omega_t(naics) \in \{1/6, 1/3, 1\}$$

where C is the cost of one survey unit. Here the Total Cost, is a fixed given value, which is assigned to each state. Note that the first sum obtains its minimum when the sum

$$\sum_{\{NAICS\}} \omega_t(naics) \times \frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)$$

obtains its maximum. Therefore, it is equivalent to find a maximum of the second sum in order to finding minimum of the first one. The last expression is mathematically and conceptually easier to work with than the first, and therefore the last expression is used for the further study. The above setup of finding minimum or maximum for a linear function under a set of constrains is very typical in operations research.

2.2 Analysis of the Optimization Problem

In or der to f ind a suitable solution t o the optimization pr oblem, a new sim pler notation is introduced. Assume that there are *n* different NAICS codes in a given state. Assume further that NAICS codes are given indices, i.e. { $\omega_t(naics) : naics \in \text{NAICS} \} = \{ \omega_i : 1 \le i \le n \}.$

Assume accordingly that indices are being assigned to the terms

$$\frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics)$$

and that these terms are denoted by the letter d, i.e.

$$\left\{\frac{1}{\omega_{t-1}(naics)} \times \Delta_{t-1,t}^{LDB}(naics) : naics \in \text{NAICS}\right\} = \left\{d_i : 1 \le i \le n\right\}.$$

The terms $N_t^{LDB}(naics)$ are given indices, and denoted by the letter e, i.e.

$$\{ N_t^{LDB}(naics) : naics \in NAICS \} = \{ e_i : 1 \le i \le n \}.$$

Next, let f and g be defined

$$f(\omega_1, \omega_2, \dots, \omega_n) = \sum_{i=1}^n (\omega_i \times d_i)$$

and

$$g(\omega_1, \omega_2, ..., \omega_n) = C \times \sum_{i=1}^n (\omega_i \times e_i).$$

Thus, the optimization problem can be stated as follows:

MAXIMIZE:

(*)
$$f(\omega_1, \omega_2, ..., \omega_n)$$

under the constraints:

(**)
$$Total_Cost = g(\omega_1, \omega_2, ..., \omega_n)$$

$$(***)$$
 $\omega_i \in \left\{\frac{1}{6}, \frac{1}{3}, 1\right\}$ $1 \le i \le n$.

Before finding a solution to the problem stated, making some useful observations is necess ary. Note that the function f is a linear function with respect to $\omega_1, \omega_2, \dots, \omega_n$. (Function f is therefore a harmonic function, and hence it assumes both minimum and maximum values on the edges of the closed region defined by (**)). The problem of finding the maximum of this function has a solution and the solution must be located on the edge of region defined by (**).

Through a series of sim ple algebraic computations it can be sho wn that condition (**) can be incorporated into expression (*) giving the following identity:

(17)
$$\sum_{i=1}^{n} (\omega_i \times d_i)$$
$$= \frac{1}{n} \sum_{i=1}^{n} \omega_i \left[(n-1) \times d_i - e_i \times \sum_{j=1, j \neq i}^{n} \frac{d_j}{e_j} \right] + \frac{1}{n} \times \frac{Total_Cost}{Cost} \times \sum_{i=1}^{n} \frac{d_i}{e_i}$$

for ω_1 , ω_2 , ..., ω_n satisfying the constraint (**).

In order to find a maximum of the function $f(\omega_1, \omega_2, ..., \omega_n)$ under the constraints (**) and (***), one can use the following procedure:

Order the series:

$$\{ (n-1) \times d_i - e_i \times \sum_{j=1, j \neq i}^n \frac{d_j}{e_j} ; 1 \le i \le n \}$$

to be a decreasing sequence. For simplicity, suppose that indices *i* are assigned in such a way that

(18)
$$(n-1) \times d_1 - e_1 \times \sum_{j=1, j \neq 1}^n \frac{d_j}{e_j} \ge (n-1) \times d_2 - e_2 \times \sum_{j=1, j \neq 2}^n \frac{d_j}{e_j} \ge \dots \ge (n-1) \times d_n - e_n \times \sum_{j=1, j \neq n}^n \frac{d_j}{e_j}.$$

It can be proven that, the first element is always positive and the last element is always negative, i.e., we have that:

(19)
$$(n-1) \times d_1 - e_1 \times \sum_{j=1, j\neq 1}^n \frac{d_j}{e_j} > 0$$
 and $(n-1) \times d_n - e_n \times \sum_{j=1, j\neq n}^n \frac{d_j}{e_j} < 0$.

and

Therefore, there exists an integer k such that 1 < k < n, which divides sequence (18) into all positive and and all negative parts.

From equation (17) it follows that the most eff ective way to maximize the value of the function $f(\omega_1, \omega_2, ..., \omega_n)$ is to assign, under the constraints given, the maximum values for ω 's with low indices, and assign the minimum values for ω 's with high indices. This way the largest values of ω 's are assigned to the positive part of the sequence (18) and the lowest to the negative part. That leads to relation:

(20) $\omega_1 \ge \omega_2 \ge \omega_3 \ge \ldots \ge \omega_n$.

These observations provide basis on which algorithm s for finding solutions to the stated problem can be made. The following simple algorithm can significantly reduce cost of survey and improve the quality of survey.

3. Proposed Algorithms

The algorithms below provide a solutio n to the st ated problem. Each has cert ain advantages and disadvantages depending on a particular use.

The first algorithm is the simplest one. This algorithm takes advantage of the fact that the sequence (18) can be divided into positive and negative part s by assigning 1's to the first part and 1/6' s to the second part. The first step is not needed for algorithm itself, nevertheless it is recommended to be made for a sake of being able to com pare achieved accuracy and cost to the present accuracy and cost.

Algorithm 1

Step 1 First, assume $\omega_1 = \omega_2 = \cdots = \omega_n = 1/3$, and compute:

(21)
$$M_0 = f(1/3, 1/3, ..., 1/3)$$
 and $C_0 = g(1/3, 1/3, ..., 1/3)$.

Step 2 Find an integer k such that:

$$d_k - \frac{1}{n-1} e_k \sum_{j=1, j \neq k}^n \frac{d_j}{e_j} \ge 0$$
 and $d_k - \frac{1}{n-1} e_{k+1} \sum_{j=1, j \neq k+1}^n \frac{d_j}{e_j} < 0.$

From equations (19), such an integer k always exists, and $k \in \{2, 3, \dots, n-1\}$.

Step 3

Compute $M_1 = f(\omega_1, \omega_2, \dots, \omega_n)$ and $C_1 = g(\omega_1, \omega_2, \dots, \omega_n)$ where:

 $\omega_1 = 1$, $\omega_2 = 1$, ..., $\omega_k = 1$ and $\omega_{k+1} = 1/6$, $\omega_{k+2} = 1/6$, ..., $\omega_n = 1/6$

That is, assume for ω_i value 1 when $i \le k$ and value 1/6 when i > k.

The next algorithm is meant to m ake a complete u se of available resources, defined as $C_0 = g(1/3, 1/3, ..., 1/3)$, and it is designed to slightly improve the quality of the survey in comparison to the previous algorithm. The main advantage of this algorithm versus the other one is that it is the most similar to the current system of surveying all establishments every 3 years. Thus, it is a smooth change between the current method and Algorithm 1.

Algorithm 2

Steps 1 – 3 are identical as in the previous algorithm. <u>Step 4</u> Assume $\omega_1 = \omega_2 = \cdots = \omega_n = 1/3$. Compute:

(21) $M_0 = f(1/3, 1/3, ..., 1/3)$ and $C_0 = g(1/3, 1/3, ..., 1/3)$.

Next compute the value $C_1 = g(\omega_1, \omega_2, ..., \omega_n)$. Compare C_1 to C_0 . There are two possibilities: $C_1 \ge C_0$ or $C_1 < C_0$.

If $C_1 \geq C_0$:

then consecutively decrease the values $\omega_1, \omega_2, \dots, \omega_k$ to 1/3 starting from the value ω_k , next ω_{k-1} , next ω_{k-2} , each time computing $g(\omega_1, \omega_2, \dots, \omega_n)$ until for the first time the value of $g(\omega_1, \omega_2, \dots, \omega_n)$ is less than C_0 .

If $C_1 < C_0$:

then consecutively increase the values ω_{k+1} , ω_{k+2} , ..., ω_n to 1/3 starting from the value ω_{k+1} , next ω_{k+2} , next ω_{k+3} , each time computing $g(\omega_1, \omega_2, ..., \omega_n)$ until the value of $g(\omega_1, \omega_2, ..., \omega_n)$ is greater than C_0 . Then chose the iteration which precedes the iteration when $g(\omega_1, \omega_2, ..., \omega_n)$ is greater than C_0 .

(In the second case, $C_1 < C_0$, one can also c hose an option of not increasing values ω_{k+1} , $\omega_{k+2}, \dots, \omega_n$.)

The described process can be defined more formally as follows:

If $C_1 \ge C_0$:

Define a matrix $[\omega_{i,j}]_{i \le k, j \le n}$ as follows:

$$\begin{split} & \omega_{1,1} = 1, \ \dots, \ \omega_{1,k-2} = 1, \\ & \omega_{1,k-1} = 1, \\ & \omega_{1,k-1} = 1, \\ & \omega_{1,k-1} = 1, \\ & \omega_{2,k-1} = 1/3 \\ & \text{and} \\ & \omega_{2,k+1} = 1/6, \\ & \dots, \\ & \omega_{2,n} = 1/6 \\ & 1/6 \end{split}$$

Compute g and f for each vector of ω 's. Let C_i and M_i be: $C_i = g(\omega_{1,i}, \omega_{2,i}, \dots, \omega_{n,i})$; $2 \le i \le k$ $M_i = f(\omega_{1,i}, \omega_{2,i}, \dots, \omega_{n,i})$; $2 \le i \le k$

Let $m = \min\{i: C_i \le C_0\}$, then $(\omega_{1,m}, \omega_{2,m}, \dots, \omega_{n,m})$ is the vector of ω 's with the maximum value for the function f, under the constraint $C_m < C_0$, i.e. $M_m = \max\{M_i: C_i \le C_0\}$.

If $C_1 < C_0$: Define a matrix $[\omega_{i,j}]_{i \le n-k,j \le n}$ as follows: $\omega_{1,1} = 1, \dots, \omega_{1,k} = 1$ and $\omega_{1,k+1} = 1/6, \omega_{1,k+2} = 1/6, \omega_{1,k+3} = 1/6, \dots, \omega_{1n} = 1/6$ $\omega_{2,1} = 1, \dots, \omega_{2,k} = 1$ and $\omega_{2,k+1} = 1/3, \omega_{2,k+2} = 1/6, \omega_{2,k+3} = 1/6, \dots, \omega_{2,n} = 1/6$ $\omega_{3,1} = 1, \dots, \omega_{3,k} = 1$ and $\omega_{3,k+1} = 1/3, \omega_{3,k+2} = 1/3, \omega_{3,k+3} = 1/6, \dots, \omega_{3,n} = 1/6$ \dots $\omega_{n-k,1} = 1, \dots, \omega_{n-k,k} = 1$ and $\omega_{n-k,k+1} = 1/3, \omega_{n-k,k+2} = 1/3, \omega_{n-k,n} = 1/3, \dots, \omega_{n,3} = 1/3$ Compute: $C_i = g(\omega_{1,i}, \omega_{2,i}, \dots, \omega_{n,i})$; $2 \le i \le k$ Let $m = \max\{i: C_i > C\}$, then $(\omega_{1,m-1}, \omega_{2,m-1}, \dots, \omega_{n,m-1})$ is the vector of ω 's with the maximum value for the function f, under the constraint $C_m < C_0$, i.e. $M_{m-1} = \max\{M_i: C_i \le C\}$.

In both cases, the obtained vector of ω 's is the final solution to the maximization of the function f, therefore it is the vector which gives the minimum NAICS classification error for the next year.

The third algorithm will maximize the function f, for the complete use of available resources, and it will divide all establishments into only two categories: Establishments which are surveyed every one year and establishments which are surveyed every 6 years.

Algorithm 3

Steps 1 - 3 are identical as in the previous algorithm. Step 4 Now, Step 4 is identical to the Step 4 in the pr evious algorithm, except that instead of using 1/3 rate when $C_1 \ge C_0$ the rate 1/6 is used, and instead 1/3 when $C_1 < C_0$ the rate 1 is used.

In Algorithm 2 and Alg orithm 3 the value of current expense for the whole survey, C_0 , is given and is matched by the process. In the fourth, and the last proposed algorithm the user defines any value the as the whole expense of the survey. Thus, it produces best results for any given available resources.

Algorithm 4

All steps are identical as in Algorithm 3, but now the value C_0 can be replaced by any fixed number between $g(1/6_{(1)}, 1/6_{(2)}, \dots, 1/6_{(k)}, 1/6_{(k+1)}, 1/6_{(k+2)}, \dots, 1/6_{(n)})$ and $g(1_{(1)}, 1_{(2)}, \dots, 1_{(k)}, 1_{(k+1)}, 1_{(k+2)}, \dots, 1_{(n)})$.

4. Results

The results are limited to the first two algorithms which by the opinion of the author are most suited to the BLS environment. The presented results were computed from QCEW-LDB, first quarters between previous year of 2011 and the current year of 2012.

Γ					Total	Estimated				
				Num. of	Total					
					Estab. Nu	m. of				
			Projected Pro	ojected	from the	Estab.				
			Total Total		Previous	from the			Projected Pr	ojected
			Number Nu	mber	Quarter	Previous		Estimated n	aics	naics
			Surveyed Su	rveyed	with diff.	Quarter	Standard Er		ro r	Error
	Total Under			Under	naics Code	with		Deviation U	Inder	Under
	Nur	nber	Algorithm 1	Algorithm 2	in the Curr.	Incorrect Es	ti mated	of	Algorithm 1	Algorithm 2
	Cur	rently	(no	(incudes	Quar. As	naics Code	Current	Current	(no	(includes
state	Perf	formed	3 years	3 years	Recorded	in the Curr.	naics	naics	3 year	3 years
	Survey		mode)	mode)	in LDB	Quarter	Error	Error	mode)	mode)
AL 3	616	4	21137	36132	1334	2604	0.024	0.00038	0.017674	0.014429
AK	6517 6		016 6514.4		62	831	0.0425	0.00121	0.012786	0.012096
AZ 4	693	3	29739	46825	4587	8318	0.05908	0.00047	0.043395	0.03668
AR 2	714	2	18930	27085	1071	2125	0.0261	0.00051	0.0203861	0.017248
CA 4	185	38	274415	410997	1071	2125	0.0261	0.00051	0.033621	0.026915
CO 5	567	9	33398	55413	28744	56465	0.04497	0.00016	0.023363	0.019103
CT 3	402	8	20000	33926	2730	4995	0.0299	0.00035	0.021551	0.017686

Table – Results for Algorithm 1 and 2. Previous year 2012, Current year 2013.

DE 8	939		5674	8845	1259	2445	0.02395	0.00043	0.02144	0.018121
DC		9780 7	651	9733 3	94	756	0.02819	0.00133	0.037661	0.033741
FL 2	007	35	125485	200503	1429	2830	0.09645	0.00179	0.043831	0.036153
GA 8	633	1	60021	86149	14383	27696	0.04599	0.00036	0.020435	0.017591
HI 1	216	6	11547	12164	7106	14141	0.0546	0.00049	0.014932	0.014576
ID 1	659	0	13181	16567	570	1137	0.03115	0.00122	0.02833	0.025336
IL 1	239	06	62874	122761	1151	2172	0.04364	0.0012	0.028341	0.022807
IN 5	079	2	31676	50768	4468	8814	0.02371	0.00035	0.023462	0.019357
IA 2	941	4	19847	29340	2043	4010	0.02632	0.00056	0.024903	0.020886
KS 2	586	8	21225	25839	1436	2810	0.03184	0.0008	0.026352	0.023807
KY 3	426	9	24464	34251	1504	2980	0.0384	0.00094	0.015077	0.01299
LA 3	946	7	22567	39118	1127	2242	0.02181	0.00062	0.027998	0.022761
ME 1	506	0	13481	15059	2024	3986	0.03367	0.00069	0.015604	0.01484
MD 5	258	9	31824	52447	680	1341	0.02968	0.00107	0.039963	0.032985
MA 7	028	9	39981	69925	3362	6629	0.04202	0.00069	0.02453	0.019631
MI 7	466	8	67590	74645	2624	5169	0.02451	0.00046	0.016026	0.015109
MN 5	280	3	31445	52560	7263	14525	0.06484	0.00057	0.028802	0.023635
MS 2	155	4	19947	21509	2536	4889	0.03086	0.00059	0.008119	0.007756
MO 5	587	3	35764	55654	535	1068	0.01652	0.0007	0.01569	0.012961
MT 1	349	1	10071	13464	1638	3197	0.01907	0.00045	0.021804	0.018975
NE 2	138	4	13983	21356	679	1337	0.03303	0.00119	0.02455	0.020116
NV 2	400	1	12894	23840	984	1914	0.02983	0.00091	0.023957	0.019429
NH 1	527	9	12260	15252	1557	2655	0.03687	0.00077	0.017235	0.01585
NJ 8	384	8	52832	83571	631	1243	0.02712	0.00103	0.031485	0.026119
NM 1	710	2	10606	17100	8195	16267	0.06467	0.00052	0.029235	0.024187
NY I	934	46	143025	193019	849	1656	0.03228	0.00107	0.02421	0.020953
NC 8	035	6	42502	79684	22520	44628	0.0769	0.00033	0.044831	0.03635
ND 9	279		8287	9278	4904	9635	0.03997	0.00055	0.005119	0.004832
OH 8	932	0	47160	88378	201	394	0.01415	0.00095	0.039782	0.032354
OK 3	281	9	21089	32664	4873	9649	0.03601	0.0005	0.031536	0.026316
OR 4	097	6	30011	40900	1804	3519	0.03574	0.00081	0.022228	0.019503
PA 1	103	04	59417	108165	1669	3300	0.02685	0.00064	0.038311	0.031206
RI		11367	7404	11323	5940	11748	0.0355	0.00044	0.010703	0.009134
SC 3	547	9	24155	35362	303	573	0.0168	0.00091	0.032202	0.027434
SD 9	604		7449	9595	1998	3955	0.03716	0.00079	0.010932	0.009943
TN 4	502	8	26234	44968	316	607	0.02107	0.00113	0.027002	0.022434
TX 1	880	90	96875	187138	1854	3681	0.02725	0.00061	0.040163	0.032237
UT 2	708	7	15857	27000	10011	19673	0.03487	0.00033	0.025535	0.020668

VT 7	428		5013	7417	1086	2140	0.02634	0.00077	0.01178	0.010254
VA 7	519	8	46811	75111	214	413	0.01853	0.00122	0.022906	0.01875
WA 7	689	4	52065	75987	3787	7519	0.03333	0.00047	0.018879	0.015484
WV 1	456	5	10641	14520	2502	4985	0.02161	0.00042	0.008696	0.007712
WI		48561	37476	48524	353	695	0.01591	0.0008	0.02073	0.018159
WY 7	899		8011	7566	5064	9219	0.06328	0.00058	0.004642	0.005401
	2,8	84,899 1	, 852,007	2,865,911						

As it is seen in the table he Algorithm 1 cut over 1,000,000 surveyed units (thus, can save hundreds of thousands of dollars!) while delivering for almost all states projected NAICS error about half of present estimated error. (Note that Algorithm 1 does not necessarily lead to decrease in the number of surveyed units for every state, for example, WY.) Algorithm 2 used all available resources while still improving NAICS error over Algorithm 1. The results for other years are similar.

Recommendations:

- 1. Perform more simulations
- 2. Examine bias and balance of flows in-and-out of industries
- 3. Consider cost and efficiency
- 4. Recommend an algorithm to replace the current methodology

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Disclaimer:

Views express in this document are those of the author and do not necessarily reflect the views or policies of the Bureau of Labor Statistics.

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