
Appendix C.

Statistical Methodology

THE CENSUS MAIL LIST AND SCREENER PHASE

The National Agricultural Statistics Service (NASS) maintains a list of farmers and ranchers from which the census mail list (CML) is compiled. The goal is to build as complete a list as possible of agricultural places that produce and sell, or would normally sell, \$1,000 or more of agricultural products per year. This is the same list used to define sampling populations for NASS surveys conducted for the agricultural estimates program. Each record on the list includes name, address, and telephone number plus additional information used to efficiently sample and administer the NASS census of agriculture and its agricultural estimates programs.

NASS builds and improves the list on an ongoing basis by obtaining outside source lists. Sources include state and federal government lists, producer association lists, seed grower lists, pesticide applicator lists, veterinarian lists, marketing association lists, and a variety of other agriculture-related lists. NASS occasionally obtains special commodity lists to address specific list deficiencies. In 2000, NASS began an intensive push to increase list coverage in preparation for the census.

Most names on a newly acquired list are already on the list sampling frame. Those found on the list are set aside. Those not found are treated as potential farms until NASS can confirm their existence as a qualifying farm. Field offices routinely contact these potential farms to determine their status, however, the increased pre-census list building activity generated much more followup work.

Beginning in April 2002, NASS conducted the 2002 Farm Identification Survey to screen 591,288 potential farms before placing them on the CML. These records were mailed a one-page report form and a nonresponse

followup mailing was made in May 2002. A second mailing to a group of 568,692 additional potential farm records was conducted in mid-July 2002. There was no followup mailing. The entire screener phase confirmed 349,664 qualifying farms that were added to the CML. A total of 282,901 names were confirmed as out of scope and were dropped from the list. Names returned as undeliverable-as-addressed totaled 92,203 and they were excluded from further census mailings. The remaining 435,212 names did not respond and were mailed census forms although they were not added to the CML as active farms.

During the spring and summer of 2002, measures were taken to improve name and address quality. Checks were made to detect and remove duplication both within states and across states. List addresses were processed through the National Change of Address registry and the Locatable Address Conversion System to ensure they were correct and complete. Records on the mail list with missing or invalid phone numbers were matched against a nationally available telephone database to obtain as many phone numbers as possible.

Records requiring special handling for census data collection or for analysis and summarization were identified. These were mostly farms considered unique because of their size or because they produced specialty commodities.

The official CML was established on September 1, 2002. The list contained 2,841,788 records. There were 1,839,533 records that were thought to meet the NASS farm definition and 1,002,255 potential farm records.

CENSUS SAMPLE DESIGN

All name and address records on the final CML received a 2002 Census of Agriculture report form. Two different types of census report forms, sample and

nonsample, were used to collect data. Sections 1 through 16 and 22 through 25 of the sample form were identical to sections on the nonsample census form. Sections 17 through 21 of the sample form contained additional questions on usage of fertilizers and chemicals, farm production expenditures, value of machinery and equipment, value of land and buildings, and hired workers. There were 12 regional versions of the nonsample form and 13 regional versions of the sample form with listings of crops varying by region.

The sample form was mailed to all mail list records in Alaska and Rhode Island and to a sample of records in other states. Mail list records were selected into the sample with certainty if they (1) were expected to have large total value of agricultural products sold or large acreage, (2) were in a county with less than 100 farms in 1997, or (3) had other special characteristics (e.g., abnormal farms such as institutional farms, experimental and research farms, Indian reservations, etc.). Mail list records in counties containing 100 to 199 farms in 1997 were systematically sampled at a rate of 1 in 2; counties containing 200 to 299 farms in 1997 were systematically sampled at a rate of 1 in 4; counties containing 300 to 399 farms in 1997 were systematically sampled at a rate of 1 in 6; and counties containing 400 or more farms in 1997 were systematically sampled at a rate of 1 in 8. The mail list records not chosen to receive the sample form received the nonsample form. This differential sampling scheme was used to provide reliable data for the sample sections of the report form for all counties.

The regional report form versions and the sampling scheme were used to provide reliable data for a large number of items/commodities at the county level, while reducing response burden.

EDITING DATA AND IMPUTING FOR ITEM NONRESPONSE

The mailing label on all forms returned to the National Processing Center (NPC) were scanned using bar code readers to capture identifiers and for check-in purposes. Forms determined to represent qualifying, in-scope farms were submitted for imaging. A snapshot was taken of each page of every report form and optical mark recognition (OMR) and intelligent character recognition (ICR) techniques were used to capture reported data from the images. The ICR engine

determined a confidence level for every cell read. Any cell with a confidence level below a prescribed value was referred to analysts to review and correct from the image, when necessary. The images and the captured data were transferred to NASS on a flow basis. Data collected by telephone were captured using computer-assisted telephone interview software.

Captured data were processed through a format program. This program verified that record identifiers were valid and checked the basic integrity of the data fields. Rejected records were referred to analysts for correction. Accepted records were posted to the database.

All 2002 census data were passed through a complex computer edit. Data were batched by state for submission to the computer edit. The edit determined whether a reporting operation met the minimum criteria to be counted as a farm in the census. Operations failing to meet the minimum criteria were referred to analysts for verification. The edit examined each report for reasonableness and completeness and determined whether to accept, delete, impute (supply), or alter the reported value for each data record item.

Whenever possible, imputations, deletions, and changes made by the editing system were based on related data on the respondent's report form. For some items, such as operator characteristics, available data for that farm from the previous census were used. Values reported on previous NASS surveys were used, where applicable.

When these and similar methods were not available and values had to be supplied, the imputation process used information reported for another farm operation in the same state or in a neighboring state with characteristics similar to those of the farm operation with incomplete data. For example, a farm operation that reported acres of corn harvested, but did not report bushels of corn harvested, was assigned the same bushels of corn per acre harvested as that of another farm from that region having similar characteristics and reporting an acceptable yield. Assigned values for one operation could come from more than one respondent because imputation for missing items in each section of the report form was conducted separately.

Each execution of the computer edit consisted of records from only one state. Successfully edited records were made available as potential "donors," to supply values needed in subsequent imputations. These records were accumulated into pools of donors according to geographic location, so that each pool might be used during the computer edit of any reports from appropriate states. When imputation was required, a report's collective imputation needs for a section were used to identify a group of matching variables for the report which contained acceptable data relating to the missing items. For example, acres of corn harvested would be a matching variable for bushels of corn harvested, in consideration of the high correlation between the two items.

Similarity to the report being edited was evaluated for the matching variables for all farms in the appropriate donor pool. Values were imputed from the donor report considered most similar, referred to in this context as the "nearest neighbor" to the report being edited. Similarity between the edited record and a donor was calculated as the Euclidean distance between their selected matching variables. As part of the distance computation, the values of the matching variables were normalized to have the same variance within each donor pool. Latitude and longitude were consistently included in all imputation requests as matching variables, so that geographic proximity played a role in all donor selections.

Imputation conformed to logic provided by the complex edit. When appropriate, only donors able to contribute a nonzero imputed value were considered. For a farm reporting harvested corn acreage, for example, imputed bushels of corn harvested would be taken only from farms with harvested corn. In addition, imputed values were often adjusted. In some cases, acceptable data in another field of the edited report were used to establish a ratio between the edited report and the donor report. This proportion was applied to the imputed value as a scale factor. In the corn example, total bushels of corn from the donor would be scaled by the ratio of the acres of corn in the edited report to those in the donor report.

To maintain consistency with the complex edit, the imputed values in most sections of the report were tested to ensure they satisfied critical relationships among items within the section. If any of these

constraints were not met, alternative donors were considered in order of their similarity to the edited report, until all the constraints for the module were satisfied.

In some cases, nearest-neighbor imputation was not possible. The requirement of a positive imputed value might rule out all available donors, resulting in an imputation failure. However, if some members of the donor pool were found to satisfy this requirement, then as many as 25 nearest neighbors were given further consideration. But if none of the candidate donors could provide qualifying data, the result was also noted as an imputation failure. Processing of records that encountered these imputation failures was suspended at the section where the failure occurred. These records were made available for analyst review and later reconsidered by the automated edit as a followup to corrective actions taken by the analyst.

The donor pool for each region was frequently updated with records from its area which had completed the editing process. As records were added to the donor pool, the records became available to donate values to incomplete reports subsequently edited for that region. Prior to editing, all donor pools were empty and no donors were available. Initial donor pools were created by giving special treatment to the first batches of data received from each state. Similar to the way that imputation failures were resolved through analyst review of the reports, early reports from initial batches were reviewed and adjusted manually by teams of analysts. This process was employed until each donor pool became self-sufficient in consistently providing imputed values for its region through the automated nearest-neighbor selection process.

To streamline editing once they had reached a mature stage in their growth, donor pools for some regions were not expanded in size beyond a chosen plateau. This provided assurance that computer edits would not exceed a reasonable processing time for nearest-neighbor searches. Although their size was limited, these donor pools did not become static. They were regularly recreated with representative samples of all records available from their regions. Within a given region, all successfully edited sample form records were included in the appropriate donor pool. Successfully edited nonsample form records were ordered by farm size and sales volume for a given

region, and then systematically sampled. Every “ith” record from the nonsample form list was joined to the complete list of sample forms for its region to form a refreshed donor pool. The steady renewal of donor pools for regions with large numbers of records assured a more diverse selection of donors over time.

All records with data changes were resubmitted to the edit to verify that acceptable corrections were made. Records with imputation failures were referred to an analyst for resolution. Corrected data were posted and the record was re-edited.

The complex edit ensured the full internal consistency of the record. Analysts were provided an additional set of tools to review record-level data across farms. These examinations detected extreme outliers or unique data distribution patterns that were possibly a result of reporting, recording, or handling errors. Potential problems were researched and, when necessary, corrections were made and the record re-edited.

NONRESPONSE AND SAMPLE ESTIMATION

Statistical estimation procedures were used to account for whole farm nonresponse and sample data collection. The procedures for nonresponse were necessary because some farm operators did not respond to the census despite numerous attempts to contact them. Statistical estimates for sample-form-only data items had to be calculated since, by design, the data were not collected from every farm. Nonresponse and sample estimation procedures were not applied in Alaska and Rhode Island because all farms received the sample form and data were collected from all farms.

Treatment of Farms Selected for the Screener Phase

The screener phase and followup strategies resulted in several possible outcomes depending on whether the screener name responded and was in or out of scope. Each of these outcomes was handled differently to adjust for nonresponse.

Names responding to the screener as out of scope (nonfarms) were excluded from the CML. If the

respondent answered the screener as in scope, the respondent was added to the CML and received a census form. If this in-scope screener respondent answered the census form, the operation’s report was eligible to be used to help account for nonrespondents to the census. If the in-scope screener respondent failed to respond to the census form, that operation’s data were accounted for by census respondents.

Records for operations that did not respond to any of the three screener mailings were not considered to be part of the CML, but they were sent a census form. Screener nonrespondents that responded as in-scope operations on the census were assigned a fixed nonresponse weight of 1 for census tabulations. Screener nonrespondents that failed to respond to the census form were treated in summarization as if they never existed on a mail list.

Whole Farm Nonresponse Estimation

Whole farm nonresponse to the census occurred when no data were received from an operation on the CML. Records deemed to represent either a large farm, as defined by the total value of production or acreage, or a unique farm operation received intensive telephone or personal followup during census processing to obtain a response. If these attempts failed, data were imputed for the record. These large and/or unique records were designated as “Must” records and were assigned a fixed nonresponse weight of 1, meaning their data were not used for nonresponse adjustment. Screener respondents with reported sales above a certain level automatically became Must records.

During mail list development, the field offices, in an effort to reduce respondent burden, identified operations that participated in multiple NASS surveys, and those that had special reporting relationships with an enumerator. The records for these operations were “Tagged.” The field offices assumed full responsibility for the data collection for any Tagged operations, including imputing data for them if a response was not obtained. Tagged records became Must records. They had a nonresponse weight of 1 and the reports were not used for nonresponse adjustments.

Whole farm nonresponse that occurred within the remaining universe of records, called non-Musts, was

accounted for by a statistical weighting procedure. All responding non-Musts in a state were put into mutually exclusive weighting groups based on their size and county as recorded on the CML database. Statistical models were used to estimate the number of nonresponse farms that were in scope for each weighting group. The weights of the responding farms in each weighting group were increased to account for nonresponding farms in that group.

Throughout the data collection period, changes and additions were made to the CML. Records added after the initial CML was created on September 1, 2002 were designated as new adds, treated like screener nonrespondents, and given a nonresponse weight of 1. New adds responding as in-scope records to the census were subsequently subtracted from the measurement of undercoverage. New adds linked to operations originally on the CML were not considered new adds. New adds occurred any time after the CML creation and before final weighting in February, 2004.

Some operators were sent more than one census form. These operators were required to fill out a separate form for each operation. Also, an operator may have had an operation for which a census form was not received, but the existence of which was noted on the form of the known operation. That operator was sent a new census form or enumerated by telephone to obtain data for that previously unknown operation. If a response was obtained for the previously unknown operation, the nonresponse weight for the new record was set equal to the nonresponse weight for the original operation reporting its existence. If no response was obtained for the previously unknown operation, it was treated as out of scope.

Some large farms operating in more than one county were treated as distinct county-specific operations to more accurately allocate data to counties. Similarly, large farms operating in more than one state were treated as distinct state-specific operations. Split add records were created for these operations and they were assigned the same nonresponse weight as the original CML operation. Controls ensured the calculated and nonresponse weights never exceeded 2. The nonresponse weights were systematically rounded to integers and an integerized weight of either 1 or 2 was assigned to each record. The integerization process eliminated any impact rounding would have

had on census farm counts and totals in each county and in cross tabulations.

Tables A and C quantify the effect of the nonresponse estimation procedures on selected census data items. These tables contain percentages of the census aggregates that were contributed by nonresponse adjustments. As noted earlier, names included in the screener sample that never responded were treated as if they never existed on a mail list. Any in-scope farm in this group was missed and, consequently, “attributed” to the coverage adjustment. This is shown in Table C. For selected items, estimates of what was attributed were reallocated to nonresponse to obtain “corrected” values, which appear in Table A. This was possible at the state level only. The differences between state-level nonresponse adjustment numbers in the first line of Table C and their counterparts in Table A represent the amount reallocated.

There was no such reallocation in Hawaii because records in that state were not adjusted to account for coverage errors. No tables appear for Alaska because records were not adjusted for nonresponse or coverage.

The estimates provided in Tables A and C do not reflect the effect of item nonresponse on individual census data items. The effect of this item nonresponse is discussed in the “Nonmeasurable Census Error” section.

Sample Estimation

All Must records were preselected to receive the census sample form. Non-Must records were sampled to determine which would receive the sample form and which the nonsample form. All records in some small counties automatically received the census sample form but these records were not necessarily Must records. Nonresponse adjustment was allowed for the non-Musts.

Weights applied to the sample items appearing on the sample form only (Sections 17 through 21) were calculated by multiplying the farm’s coverage-adjusted weight, which is described later, by the sample factor (e.g, 6 for a farm sampled with a 1-in-6 rate, 1 for a Must). An adjustment was made that ensured the number of farms operating in a county as estimated from the sample matched the number estimated from

the full census. Before computing published tabulations based on the sample, each record's sample weight was integerized to eliminate the impact rounding would have had on census farm counts and totals.

Operators with more than one operation were sampled as one record and received the same census form for each operation. Operations added after sampling were treated differently depending on whether or not the record was linked to a record on the original CML. Added operations that linked to a record on the original CML were mailed the same census form as the original CML operation. Added operations that were **not** linked to a record on the original CML were mailed the sample form.

MEASURABLE CENSUS ERROR

The root mean squared error of an estimated data item from the census provides a measure of the error a field office associated with completing a census. It measures the variation in the value of that estimated data item based on all possible outcomes of the census collection, including variants as to who was on the census list, who returned a census form and who was selected to fill out the sample form.

Data items were classified as either complete count items or sample count items. Sample count items were collected only on the sample version of the census report form. Complete count items were collected from all respondents. Variability in the estimates of complete count items was due only to the nonresponse and coverage estimation adjustment procedures. Variability in the estimates of sample count items was due to both the adjustment procedures and the census sample selection and estimation procedure. Therefore, variability in the sample count item estimates tends to be larger than the variability in the complete count item estimates.

Table B presents the fully adjusted total with the root mean squared error for selected items. The relative root mean squared error is obtained by dividing the root mean squared error by the value of the estimate and then multiplying by 100. The table also includes the percent contribution to the mean squared error (the square of the root mean squared error) from nonresponse adjustment and sampling and from

coverage adjustment. Mean squared errors for Hawaii are entirely due to nonresponse adjustment.

Nonsampling error due to mail list incompleteness and duplication as well as misclassification of records on the mail list is called coverage error. The section titled "Classification Error Study" addresses attempts to assess, at least qualitatively, the impact of classification error on the census results.

NONMEASURABLE CENSUS ERROR

The accuracy of the census counts is affected jointly by the measurable errors described in the previous section and by nonmeasurable errors (nonmeasurable in the sense of not being included in root mean squared error estimates). Extensive efforts were made to compile a complete and accurate mail list for the census, to design an understandable report form with instructions, and to minimize processing errors through the use of quality control measures. Despite these efforts, nonmeasurable errors are inevitable and arise from many sources, including respondent or enumerator error, incorrect data capture, editing, and imputing for missing data. These errors are discussed in this section.

Respondent and Enumerator Error

Incorrect or incomplete responses to the census report form or to the questions posed by an enumerator can introduce error into the census data. To reduce reporting error, detailed instructions for completing the report form were provided to each respondent. Questions were phrased as clearly as possible based on previous tests of the report form. Computer-assisted telephone interviewing software included immediate integrity checks of recorded responses so suspect data could be verified or corrected. In addition, each respondent's answers were checked for completeness and consistency by the complex edit and imputation system.

Item Nonresponse

As information flowed from data collection to tabulation, various types of item nonresponses were identified on the census report forms. Nonresponse to particular questions on the form that logically should have been present created a type of nonsampling error

in both complete count and sample count data. In this case, information from a similar farm was used to impute for these missing data items. The resulting data may have been biased if the characteristics of the nonreporting farms were different from those of reporting farms for those items. The section titled “Editing Data and Imputing for Item Nonresponse” provides a detailed explanation of item imputation procedures.

Processing Error

All phases of processing for each census report form were potential sources of nonsampling error. An automated check-in procedure recorded that the report had been returned and excluded it from further followup mailings. Approximately one-third of the mail returns were reviewed to resolve questions dealing with multiple reports, respondent remarks, or no reported data. The remaining mail returns (about two-thirds), along with some of the reviewed cases containing farm data, were batched and sent directly to imaging and data capture. Data were transmitted, formatted, and run through the complex edit and imputation system to ensure within record consistency. About one-fifth of all forms edited were clerically reviewed for inconsistencies, omissions, or questionable values. While reviewing these forms, staff determined if the action taken by the computer edit and imputation system was correct. Additional analysis tools were used to examine data across records for distributional irregularities and extreme values. Edited records were tabulated to the county level. Each county was reviewed and, when necessary, individual records were corrected prior to publication.

Developing accurate processing methods is complicated by the complex structure of agriculture. Among the complexities are the many places to be included, the variety of arrangements under which farms are operated, the continuing changes in the relationship of operators to the farm operated, the expiration of leases and the initiation or renewal of leases, the problem of obtaining a complete list of agriculture operations, the difficulty of contacting and identifying some types of contractor/contractee relationships, the operator’s absence from the farm during the data collection period, and the operator’s opinion that part or all of the operation does not qualify and should not be included in the census.

During data collection and processing of the census, all operations underwent a number of quality control checks to ensure results were as accurate as possible.

COVERAGE ADJUSTMENT

Although much effort was expended making the CML as complete as possible, the coverage of farms was not complete. NASS’s goal was to produce agricultural census totals for publication that were fully adjusted for list undercoverage at the county level. Estimates of the undercoverage for a specified set of farm characteristics, called calibration variables, were computed using an area-frame sample. Initial weights were assigned to census respondents to account for nonresponse, and these weights were further adjusted to compensate for estimated state-level undercoverage for each of the calibration variables based on the area frame sample. Each farm with census data was assigned a fully-adjusted weight by this process and county-level totals were generated for every census variable, not just the calibration variables. The section titled “Calibration Algorithm” provides a list of the area frame based calibration variables.

To further improve coverage adjustment, a second set of targets and ranges were added to the calibration effort. These were well established commodity totals for which excellent check data were available for validation. The introduction of these commodity targets strengthened the overall coverage adjustment process by limiting the possible adjustments produced by the area frame based targets to ensure major commodity totals remained within reasonable bounds of established benchmarks.

Most targets were determined at the state level. The one exception was the New England states - Connecticut, Maine, Massachusetts, New Hampshire, Rhode Island, and Vermont - which were combined into one “calibration region.” In what follows, “state” refers to the calibration region for New England. Coverage adjustments were not made in Alaska and Hawaii.

Measuring Mail List Undercoverage

Census mail list undercoverage was measured using an independent survey of land segments selected from the NASS area frame. The NASS area frame covers all

land in the U.S. and includes all farms. Each June, NASS conducts a survey that enumerates area frame segments for agricultural activity. The sampled segments are allocated to provide accurate measures of acres planted to widely grown crops and inventories of hogs and cattle.

The 2002 June Agricultural Survey (JAS) was supplemented with the 2002 Agricultural Coverage Evaluation Survey (ACES) to better estimate CML incompleteness. The ACES used a sample of segments allocated in a way that, when pooled with the JAS, ensured accurate measures of number of farms and land utilization could be obtained. Enumerators visited all segments, identified all farms operating land in each segment, and obtained basic data about those farms.

The names and addresses collected in the 2002 JAS and 2002 ACES were matched to the census mail list. Farms that did not match were re-contacted after the census mailout to confirm that they did not receive a census form. Farms that had not received a census form represented the farms not on the mail list (NML). Those who received a census form had been erroneously classified as NML and were removed.

The percentage of farms missed in the census varied considerably by state. In general, farms not on the mail list tended to be small in acreage, production, and sales of agricultural products. Farm operations were missed for various reasons, including the possibility that the operation started after the mail list was developed, the operation was so small as not to appear in any agriculture-related source lists, or the operation was falsely classified as a nonfarm prior to mailout.

Determining Targets to Correct for Undercoverage

The 2002 June Agricultural Survey consisted of 11,075 land segments and the 2002 Agricultural Coverage Evaluation Survey (ACES) added 2,400 segments. Data values a field office associated with NML tracts were used to estimate the state-level undercoverage of the CML for the first set of calibration variables. The state-level totals for these variables were then summed to yield national totals.

The national NML estimate for the number of farms

was used directly in determining calibration targets (CML + NML). State-level farm-count estimates based on the NML sometimes had unacceptably high standard errors, as well as apparent systematic biases. These estimates were smoothed across states based on separate NASS surveys and previous analysis.

Other calibration targets were derived from the NML-estimated fractions of farms of certain types (i.e. in a particular sales class or with a principal operator of a particular race). Most of these had unacceptably high state-level standard errors. As a result, more reliable national-level NML estimates were used to smooth state estimates. The smoothed state NML-estimated fraction was computed by taking a weighted average of the actual state estimate and a prediction for the state based on national- and state-level numbers (i.e., the number of NML farms in the state, the fraction of farms with black owners on the state's CML, and the national relative difference between the fraction of black owners on the NML and CML). The weighting factor was chosen to approximately minimize mean squared error under a random effects model. The smoothed NML-estimated fractions were multiplied by the corresponding smoothed NML farm-count estimates described above and added to corresponding CML estimates to obtain coverage-adjusted state-level totals, that served as calibration targets.

Tolerance Ranges

Although full calibration would assure that the weighted total among census respondents equaled its target for each calibration variable in either set, it was not always possible to calibrate to such a large number of target values while keeping all farm weights within a reasonable range (for example, the weight for any farm cannot be less than one). Because of this and because calibration targets are estimates themselves subject to uncertainty, NASS allowed some tolerance in the determination of coverage-adjusted weights. Rather than forcing the total for each calibration variable computed using the coverage-adjusted weights to equal a specific amount, NASS allowed the estimated total to fall within a tolerance range. This tolerance strategy sometimes made it possible for the calibration algorithm to produce a set of satisfactory coverage-adjusted weights that it would not have otherwise.

Ranges for the first set of calibration variables used to adjust for undercoverage were determined differently from the second set used to adjust for measurement error. The number of farms had no tolerance range. The tolerance range for every other variable in the first set was the estimated state total for the variable (CML + NML) plus or minus one-half of one estimated standard error. This choice limited the cumulative deviation from the estimated total for a variable when state-level totals were combined to create a U.S.-level total. These ranges did not have to be symmetric around the target value.

Calibration Algorithm

Coverage adjusted weights were obtained by an algorithm based on the restricted regression algorithm referred to by Singh and Mohl (1996) as the Linear Truncated Method. Coverage adjustments began with the nonresponse weights before integerization. The final coverage-adjusted weights were restricted to the interval [1,6].

The calibration variables were based on the following reported items:

1. Total market value of agricultural products sold and government payments.

\$0	\$5,000 - \$24,999
\$1 - \$999	\$25,000 - \$99,999
\$1,000 - \$2,499	\$100,000 - \$499,999
\$2,500 - \$4,999	\$500,000 and above

2. Age of principal operator.

Less than 25 years old
25 - 34
35 - 44
45 - 54
55 and older

3. Sex of principal operator.

Female

4. Race of principal operator (selected categories).

Black, American Indian or Alaska Native, Asian, and Native Hawaiian or Other Pacific Islander

5. Principal operators of Spanish, Hispanic, or Latino origin.
6. Number of farms and land in farms.
7. Number of extreme operations (very large or unusual farms).
8. Selected types of farms by commodity produced.

All cattle farms
 Dairy farms
 Hog/pig farms
 Horse/Equine farms
 Poultry farms
 Sheep/goat farms

Fruit/nut/berry farms
 Nursery/horticulture farms
 Tobacco farms
 Vegetable and melon farms

9. Various commodity acreage and production statistics (varies by state).

Beef cow inventory
 Broiler production
 Cattle on feed inventory
 Layer inventory
 Milk cow inventory
 Total cattle and calves inventory
 Total hog and pig inventory

Alfalfa acres harvested (South Dakota only)
 Apples acres harvested
 Corn acres harvested
 Cotton bales produced
 Grape acres harvested
 Hay acres harvested
 Lettuce acres harvested
 Potato acres harvested
 Rice acres harvested
 Soybean acres harvested
 Sugarbeet acres harvested
 Sugarcane acres harvested
 Tobacco acres harvested
 Tomatoes acres harvested
 Total orange acres
 Wheat acres harvested
 Durum wheat acres harvested (North Dakota only)

Other spring wheat acres harvested (North Dakota only)

Integerization and Sample Weights

Coverage-adjusted weights were integerized to eliminate the need for rounding estimated counts computed with coverage-adjusted weights. The integerization process minimized county-level impact on the nonresponse and coverage adjustment of number of farms and total land in farms.

Sample weights were computed by multiplying coverage-adjusted weights before integerization with the appropriate sampling factors and adjusting the results to add up to matched census counts as described previously. Sample weights were then integerized.

Measuring the Amount of Coverage Adjustment

Tables A and C display the proportions of selected census data items that are due to nonresponse and coverage adjustments. The section of this appendix on whole farm nonresponse adjustment explained how the nonresponse adjustment values were determined. The coverage adjustment values account for the rest of the differences between the weighted and unweighted totals for these data items. Some estimated coverage adjustments are negative. The use of commodity targets in calibration indirectly exposed some duplication on the census list resulting in negative coverage adjustments.

CLASSIFICATION ERROR STUDY

The 2002 Classification Error Study (CES) was conducted for the conterminous U.S. to study the potential impact of classification error on the census results. The study used data from the 2002 June Agricultural Survey (JAS) and the 2002 Agricultural Coverage Evaluation Survey to examine farms incorrectly classified as nonfarms (undercount), nonfarms incorrectly classified as farms (overcount), and duplication of farms (overcount) in the 2002 Census of Agriculture. The CES was not intended to adjust census farm counts, but rather, to evaluate procedures and to identify potential improvements in list building, data collection, and other activities in

preparation for future censuses.

For the evaluation, additional name, address, and telephone information were collected on both the JAS and ACES by adding the following three questions:

1. During the past two years, has the operator received mail for this operation at any address other than the one shown on the face page?
2. Excluding partners and landlords, were any other names associated with this operation in the past two years? (For example, other business names, spouses names, etc).
3. Is any of the land inside the blue tract boundary rented from others? (Include land for which you paid cash rent, land used rent free, or land rented on shares).

The CES consisted of a two phase review process. The initial phase, Review of Possible Matches, used Probabilistic Record Linkage (PRL) to match the additional information collected on the area surveys to the name and addresses on the 2002 Census Mail List (CML) including late adds. PRL is a technique used to identify records that are believed to correspond to a CML record. Records were brought together into link groups, with each link group consisting of all records that possibly represented the same operation. Each link group was classified into one of three distinct types: matches, possible matches and nonmatches. The nonmatches were represented in estimation as part of the undercoverage measure. The CES was primarily concerned with the matches and possible matches. Each field office reviewed the possible matches and determined match or nonmatch status.

Upon completion of the PRL review, the field offices conducted a Farm Classification Resolution review of two additional sets of records. The first of these was comprised of area records matching two or more census records. Reviewing these records helped identify duplication on the CML. The second set consisted of groups of records (area and census) within which the reported acreage differed by more than 25 percent. Analysts reviewed the cases in the second phase. Upon completion of both phases, data were compiled to estimate undercount, overcount and duplication.

The analysis of these data will provide insight into census processes used to accurately determine farm status and identify duplication. Any weaknesses

identified in the findings will be addressed for future censuses.

Table A. Summary of State Nonresponse and Coverage Adjustments: 2002

[For meaning of abbreviations and symbols, see introductory text]

Item	Total	Percent from nonresponse adjustment, corrected	Percent from coverage adjustment, corrected	Item	Total	Percent from nonresponse adjustment, corrected	Percent from coverage adjustment, corrected
Farms number	83,300	14.8	17.0	Tenure - Con.			
Land in farms acres	33,661,826	11.7	0.8	Part owners farms	24,497	15.1	8.3
Farms by size:				acres	19,835,231	10.8	-2.7
1 to 9 farms	3,091	15.8	29.3	Tenants farms	5,986	17.1	17.6
acres	14,835	15.6	29.7	acres	2,448,899	14.0	2.3
10 to 49 farms	17,072	14.5	28.4	Principal operator characteristics by-			
acres	461,588	14.3	27.9	Sex of operator:			
50 to 179 farms	28,776	14.3	22.2	Male farms	74,580	14.8	16.7
acres	3,110,778	14.3	21.5	acres	31,411,761	11.6	0.7
180 to 499 farms	19,357	15.3	14.0	Female farms	8,720	15.3	19.3
acres	5,876,656	15.5	13.1	acres	2,250,065	13.2	3.2
500 to 999 farms	7,668	17.8	1.2	Primary occupation:			
acres	5,317,775	18.0	0.3	Farming farms	46,053	14.8	13.5
1,000 to 1,999 farms	4,119	18.9	-14.3	Other farms	37,247	14.9	21.4
acres	5,603,196	18.7	-14.5	Spanish, Hispanic, or Latino origin (see text) farms	1,498	10.4	45.5
2,000 or more farms	3,217	5.1	-4.8	acres	380,903	9.8	32.1
acres	13,276,998	3.9	-3.7	Race:			
Market value of agricultural products sold \$1,000	4,456,404	6.5	4.1	White farms	76,603	14.6	17.6
Farms by value of sales:				acres	31,990,936	11.6	0.9
Less than \$1,000 farms	20,060	14.5	29.3	Black or African American farms	840	15.7	19.4
\$1,000 farms	2,257	20.0	3.7	acres	139,659	16.1	7.4
\$1,000 to \$2,499 farms	9,867	15.3	22.4	American Indian or Alaska Native farms	4,546	17.7	9.7
\$2,500 to \$4,999 farms	16,412	15.1	22.2	acres	1,101,801	14.3	-1.0
\$4,999 to \$9,999 farms	9,925	15.8	12.1	Native Hawaiian or Other Pacific Islander farms	12	33.3	8.3
\$9,999 to \$19,999 farms	35,590	15.9	11.8	acres	311	46.0	-11.3
\$19,999 to \$24,999 farms	12,452	14.6	17.4	Asian farms	52	19.2	3.8
\$24,999 to \$29,999 farms	87,909	14.7	17.0	acres	4,765	22.7	-7.1
\$29,999 to \$34,999 farms	10,056	16.4	7.3	More than one race reported farms	1,247	17.2	8.8
\$34,999 to \$39,999 farms	139,858	16.6	6.6	acres	424,354	12.3	-3.5
\$39,999 to \$44,999 farms	2,777	17.4	4.1	Reporting primary occupation as farming by age group:			
\$44,999 to \$49,999 farms	61,345	17.3	4.4	Under 25 years farms	416	18.5	25.7
\$49,999 to \$54,999 farms	5,437	14.5	19.6	25 to 34 years farms	2,003	19.8	11.1
\$54,999 to \$59,999 farms	169,164	14.5	19.2	35 to 44 years farms	5,720	15.5	17.9
\$59,999 to \$64,999 farms	1,936	15.2	11.2	45 to 54 years farms	8,611	14.3	15.2
\$64,999 to \$69,999 farms	85,330	15.2	11.2	55 to 64 years farms	10,898	14.7	11.6
\$69,999 to \$74,999 farms	4,340	16.4	5.1	65 years and over farms	18,405	14.2	12.5
\$74,999 to \$79,999 farms	302,570	16.4	4.7	Reporting primary occupation as other than farming by age group:			
\$79,999 to \$84,999 farms	3,647	11.9	8.0	Under 25 years farms	366	18.3	31.1
\$84,999 to \$89,999 farms	568,762	11.6	8.0	25 to 34 years farms	2,220	21.0	15.1
\$89,999 to \$94,999 farms	1,609	8.5	2.6	35 to 44 years farms	7,698	15.5	24.1
\$94,999 to \$99,999 farms	561,561	8.3	2.2	45 to 54 years farms	11,940	13.8	23.9
\$99,999 to \$104,999 farms	810	4.4	2.5	55 to 64 years farms	9,283	14.9	18.7
\$104,999 to \$109,999 farms	558,053	4.3	2.4	65 years and over farms	5,740	13.9	18.5
\$109,999 to \$114,999 farms	384	2.3	2.6	All operators by age group ¹ :			
\$114,999 to \$119,999 farms	1,867,593	0.7	1.1	Under 25 years farms	2,637	16.6	22.0
Farms by type of organization:				25 to 34 years farms	7,797	19.9	13.0
Family or individual farms	78,197	14.9	17.5	35 to 44 years farms	21,657	15.2	20.6
acres	27,938,071	12.6	1.6	45 to 54 years farms	30,312	14.1	19.6
Partnership farms	3,392	14.8	8.9	55 to 64 years farms	27,885	14.9	14.8
acres	3,686,707	8.8	-3.6	65 to 74 years farms	19,168	13.9	13.9
Corporation:				75 years and over farms	10,053	14.8	12.5
Family held farms	978	10.8	10.3				
acres	1,312,382	6.2	-1.6				
Other than family held farms	138	8.7	22.5				
acres	90,748	3.9	2.8				
Other - cooperative, estate or trust, institutional, etc farms	595	11.4	11.9				
acres	633,918	5.2	-0.1				
Tenure:							
Full owners farms	52,817	14.4	21.0				
acres	11,377,696	12.8	6.8				

¹ Data were collected for a maximum of three operators per farm.

Table B. Reliability Estimates of State Totals: 2002

[For meaning of abbreviations and symbols, see introductory text]

Item	Total	Root mean squared error (RMSE)	Relative RMSE (percent)	Nonresponse and sampling contribution to MSE (percent)	Coverage adjustment contribution to MSE (percent)
Farms number	83,300	2,060	2.5	0.1	99.9
Land in farms acres	33,661,826	1,031,903	3.1	0.3	99.7
Farms by size:					
1 to 9 farms	3,091	122	3.9	4.7	95.3
10 to 49 acres	14,835	597	4.0	5.5	94.5
50 to 179 farms	17,072	510	3.0	1.1	98.9
180 to 499 acres	461,588	13,501	2.9	1.5	98.5
500 to 999 farms	28,776	706	2.5	0.9	99.1
1,000 to 1,999 acres	3,110,778	76,387	2.5	1.1	98.9
2,000 or more farms	19,357	540	2.8	1.2	98.8
1,000 to 1,999 acres	5,876,656	167,227	2.8	1.3	98.7
2,000 or more farms	7,668	282	3.7	1.8	98.2
1,000 to 1,999 acres	5,317,775	198,880	3.7	1.8	98.2
2,000 or more farms	4,119	185	4.5	1.8	98.2
1,000 to 1,999 acres	5,603,196	251,775	4.5	1.8	98.2
2,000 or more farms	3,217	125	3.9	0.9	99.1
2,000 or more acres	13,276,998	431,366	3.2	0.8	99.2
Market value of agricultural products sold \$1,000	4,456,404	94,269	2.1	0.7	99.3
Farms by value of sales:					
Less than \$1,000 farms	20,060	1,029	5.1	0.2	99.8
\$1,000 to \$2,499 \$1,000	2,257	167	7.4	0.7	99.3
\$2,500 to \$4,999 farms	9,987	588	6.0	0.3	99.7
\$5,000 to \$9,999 \$1,000	16,412	977	6.0	0.4	99.6
\$10,000 to \$19,999 farms	9,925	448	4.5	0.6	99.4
\$20,000 to \$24,999 \$1,000	35,590	1,601	4.5	0.7	99.3
\$25,000 to \$39,999 farms	12,452	444	3.6	1.0	99.0
\$40,000 to \$49,999 \$1,000	87,909	3,147	3.6	1.1	98.9
\$50,000 to \$99,999 farms	10,056	397	3.9	1.2	98.8
\$100,000 to \$249,999 \$1,000	139,858	5,559	4.0	1.2	98.8
\$250,000 to \$499,999 farms	2,777	115	4.2	4.6	95.4
\$500,000 to \$999,999 \$1,000	61,345	2,542	4.1	4.6	95.4
\$1,000,000 to \$2,499,999 farms	5,437	233	4.3	2.1	97.9
\$2,500,000 to \$4,999,999 \$1,000	169,164	7,286	4.3	2.1	97.9
\$5,000,000 to \$9,999,999 farms	1,936	97	5.0	4.7	95.3
\$10,000,000 to \$24,999,999 \$1,000	85,330	4,283	5.0	4.7	95.3
\$25,000,000 to \$49,999,999 farms	4,340	205	4.7	2.0	98.0
\$50,000,000 to \$99,999,999 \$1,000	302,570	14,145	4.7	2.1	97.9
\$100,000,000 to \$249,999,999 farms	3,647	131	3.6	2.8	97.2
\$250,000,000 to \$499,999,999 \$1,000	568,762	20,503	3.6	3.0	97.0
\$500,000,000 to \$999,999,999 farms	1,609	52	3.2	6.3	93.7
\$1,000,000,000 or more \$1,000	561,561	17,603	3.1	6.1	93.9
1,000,000,000 or more farms	810	43	5.3	2.0	98.0
1,000,000,000 or more \$1,000	558,053	28,037	5.0	2.2	97.8
1,000,000,000 or more farms	384	12	3.2	5.1	94.9
1,000,000,000 or more \$1,000	1,867,593	22,055	1.2	5.4	94.6
Farms by type of organization:					
Family or individual farms	78,197	1,934	2.5	0.1	99.9
Partnership acres	27,938,071	884,504	3.2	0.3	99.7
Corporation: farms	3,392	106	3.1	5.7	94.3
Family held acres	3,686,707	107,049	2.9	2.8	97.2
Other than family held farms	978	41	4.2	10.0	90.0
Other - cooperative, estate or trust, institutional, etc acres	1,312,382	49,541	3.8	5.0	95.0
Other - cooperative, estate or trust, institutional, etc farms	138	12	8.8	19.8	80.2
Other - cooperative, estate or trust, institutional, etc acres	90,748	4,995	5.5	14.9	85.1
Other - cooperative, estate or trust, institutional, etc farms	595	25	4.2	17.3	82.7
Other - cooperative, estate or trust, institutional, etc acres	633,918	15,744	2.5	13.8	86.2
Tenure:					
Full owners farms	52,817	1,304	2.5	0.3	99.7
Part owners acres	11,377,696	314,716	2.8	0.8	99.2
Tenants farms	24,497	718	2.9	0.7	99.3
Tenants acres	19,835,231	668,206	3.4	0.5	99.5
Tenants farms	5,986	184	3.1	4.0	96.0
Tenants acres	2,448,899	86,862	3.5	3.8	96.2
Principal operator characteristics by-					
Sex of operator:					
Male farms	74,580	1,851	2.5	0.1	99.9
Female acres	31,411,761	973,466	3.1	0.3	99.7
Male farms	8,720	240	2.8	3.1	96.9
Female acres	2,250,065	71,032	3.2	4.5	95.5
Primary occupation:					
Farming farms	46,053	1,185	2.6	0.4	99.6
Other farms	37,247	947	2.5	0.6	99.4
Spanish, Hispanic, or Latino origin (see text) farms	1,498	301	20.1	0.2	99.8
Spanish, Hispanic, or Latino origin (see text) acres	380,903	98,870	26.0	0.5	99.5
Race:					
White farms	76,603	1,942	2.5	0.1	99.9
Black or African American acres	31,990,936	1,000,394	3.1	0.3	99.7
Black or African American farms	840	39	4.6	13.4	86.6
American Indian or Alaska Native acres	139,659	9,097	6.5	14.5	85.5
American Indian or Alaska Native farms	4,546	355	7.8	0.5	99.5
Native Hawaiian or Other Pacific Islander acres	1,101,801	91,653	8.3	1.2	98.8
Native Hawaiian or Other Pacific Islander farms	12	4	33.7	16.3	83.7
Native Hawaiian or Other Pacific Islander acres	311	104	33.6	20.0	80.0

See footnote(s) at end of table.

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Table B. Reliability Estimates of State Totals: 2002 - Con.

[For meaning of abbreviations and symbols, see introductory text]

Item	Total	Root mean squared error (RMSE)	Relative RMSE (percent)	Nonresponse and sampling contribution to MSE (percent)	Coverage adjustment contribution to MSE (percent)
Principal operator characteristics by- Con.					
Race - Con.					
Asian farms	52	7	14.0	18.9	81.1
25 to 34 years acres	4,765	791	16.6	23.9	76.1
More than one race reported farms	1,247	101	8.1	2.2	97.8
acres	424,354	28,868	6.8	3.7	96.3
Reporting primary occupation as farming by age group:					
Under 25 years farms	416	107	25.7	0.7	99.3
25 to 34 years farms	2,003	72	3.6	8.4	91.6
35 to 44 years farms	5,720	275	4.8	1.4	98.6
45 to 54 years farms	8,611	354	4.1	1.2	98.8
55 to 64 years farms	10,898	324	3.0	1.9	98.1
65 years and over farms	18,405	547	3.0	1.0	99.0
Reporting primary occupation as other than farming by age group:					
Under 25 years farms	366	87	23.9	0.9	99.1
25 to 34 years farms	2,220	79	3.6	8.1	91.9
35 to 44 years farms	7,698	360	4.7	1.0	99.0
45 to 54 years farms	11,940	459	3.8	0.9	99.1
55 to 64 years farms	9,283	275	3.0	2.5	97.5
65 years and over farms	5,740	181	3.1	3.7	96.3
All operators by age group ¹:					
Under 25 years farms	2,637	224	8.5	1.1	98.9
25 to 34 years farms	7,797	230	3.0	4.1	95.9
35 to 44 years farms	21,657	843	3.9	0.5	99.5
45 to 54 years farms	30,312	1,045	3.4	0.5	99.5
55 to 64 years farms	27,885	776	2.8	1.0	99.0
65 to 74 years farms	19,168	554	2.9	1.3	98.7
75 years and over farms	10,053	304	3.0	2.3	97.7
Net cash farm income of operations (see text) ²:					
Farms with gains of ³ -					
Less than \$1,000 farms	5,199	325	6.3	54.5	45.5
\$1,000 farms	2,469	175	7.1	57.3	42.7
\$1,000 to \$4,999 farms	11,524	505	4.4	39.4	60.6
\$1,000 farms	30,613	1,442	4.7	43.0	57.0
\$5,000 to \$9,999 farms	5,252	312	5.9	49.4	50.6
\$1,000 farms	37,603	2,265	6.0	50.2	49.8
\$10,000 to \$24,999 farms	5,832	332	5.7	44.1	55.9
\$1,000 farms	92,790	5,469	5.9	44.5	55.5
\$25,000 to \$49,999 farms	3,953	259	6.5	45.6	54.4
\$1,000 farms	138,075	9,097	6.6	45.3	54.7
\$50,000 or more farms	4,615	212	4.6	37.4	62.6
\$1,000 farms	846,420	29,270	3.5	31.1	68.9
Farms with losses of -					
Less than \$1,000 farms	7,485	401	5.4	52.5	47.5
\$1,000 farms	3,905	235	6.0	56.4	43.6
\$1,000 to \$4,999 farms	18,587	717	3.9	32.3	67.7
\$1,000 farms	50,261	2,055	4.1	37.1	62.9
\$5,000 to \$9,999 farms	10,098	495	4.9	46.1	53.9
\$1,000 farms	71,898	3,562	5.0	46.7	53.3
\$10,000 to \$24,999 farms	7,532	406	5.4	50.7	49.3
\$1,000 farms	115,077	6,382	5.5	51.6	48.4
\$25,000 to \$49,999 farms	2,078	185	8.9	58.9	41.1
\$1,000 farms	69,809	6,149	8.8	58.6	41.4
\$50,000 or more farms	1,127	100	8.9	59.0	41.0
\$1,000 farms	152,414	10,935	7.2	61.0	39.0

¹ Data were collected for a maximum of three operators per farm.
² Data are based on a sample of farms.
³ Farms with zero net cash income are included as farms with gains of less than \$1,000.

Table C. Summary of Nonresponse and Coverage Adjustments by County: 2002

[For meaning of abbreviations and symbols, see introductory text]

Geographic area	All farms			Land in farms			Sales		
	Total (number)	Nonresponse adjustment, attributed (percent)	Coverage adjustment (percent)	Total (acres)	Nonresponse adjustment, attributed (percent)	Coverage adjustment (percent)	Total (\$1,000)	Nonresponse adjustment, attributed (percent)	Coverage adjustment (percent)
STATE TOTAL									
Oklahoma	83,300	10.9	20.9	33,661,826	9.9	2.7	4,456,404	5.5	5.1
COUNTIES									
Adair	1,130	10.7	21.3	237,874	10.9	6.5	80,437	6.6	7.4
Alfalfa	666	12.3	14.3	461,288	10.2	0.4	71,478	5.7	2.0
Atoka	1,206	11.9	17.9	492,174	9.0	3.0	22,332	12.1	6.7
Beaver	960	9.1	13.5	1,018,626	6.0	0.3	119,841	1.8	0.1
Beckham	1,012	12.7	16.4	533,250	11.7	-0.3	35,179	7.2	8.5
Blaine	825	12.6	13.6	537,314	10.6	0.1	77,215	6.5	2.5
Bryan	1,673	12.1	20.1	458,275	13.4	4.3	54,508	6.9	4.7
Caddo	1,504	13.5	16.6	710,833	14.4	1.1	88,896	10.5	8.6
Canadian	1,360	12.3	22.4	500,872	14.6	3.2	84,799	10.4	11.2
Carter	1,353	12.5	23.2	430,706	11.3	4.4	23,847	9.0	14.2
Cherokee	1,221	9.5	22.4	220,739	9.6	10.6	94,069	1.8	4.8
Choctaw	1,095	14.1	19.5	337,443	14.3	3.9	29,019	12.5	6.5
Cimarron	545	9.0	10.6	1,121,690	4.3	-0.9	182,042	0.7	0.9
Cleveland	1,294	8.7	30.5	165,483	11.5	12.0	13,222	9.8	15.5
Coal	617	12.6	17.3	262,626	10.9	1.7	13,925	10.1	2.5
Comanche	1,188	11.3	22.4	425,146	12.5	3.4	32,892	9.3	12.3
Cotton	488	12.1	13.1	334,328	9.8	-0.5	37,889	9.0	1.3
Craig	1,289	11.9	20.7	435,804	9.4	3.0	57,640	6.8	3.7
Creek	1,838	9.5	27.5	365,810	11.2	8.1	17,313	10.3	14.2
Custer	802	11.2	15.5	544,615	9.9	-0.9	44,457	7.6	2.6
Delaware	1,393	10.5	19.3	282,106	11.8	5.7	131,608	4.9	5.2
Dewey	774	11.1	12.8	584,368	9.0	-0.8	26,137	8.7	0.2
Ellis	727	10.5	14.0	672,764	7.3	(Z)	42,566	3.8	2.6
Garfield	1,083	12.3	17.6	631,939	10.9	1.0	74,711	10.4	1.6
Garvin	1,637	10.7	24.1	468,880	10.8	6.9	38,297	7.7	9.3
Grady	1,804	11.3	22.3	601,607	11.3	4.5	95,632	6.7	8.3
Grant	744	10.9	13.7	594,809	8.3	-0.1	58,048	6.9	0.4
Greer	515	14.6	10.9	324,987	14.5	-4.7	17,942	13.3	-5.2
Harmon	397	9.6	14.4	296,368	7.9	2.1	27,408	5.0	6.6
Harper	517	9.5	12.8	601,162	6.1	-0.5	142,047	1.3	0.7
Haskell	901	11.1	18.0	274,562	11.5	2.7	66,666	6.0	2.2
Hughes	955	13.1	18.6	374,192	11.1	3.8	54,665	3.2	8.2
Jackson	732	10.4	18.7	453,654	8.4	2.5	67,832	4.4	19.6
Jefferson	483	12.8	16.1	407,194	7.4	2.4	50,080	3.6	7.8
Johnston	682	11.0	18.9	325,983	7.4	2.4	26,208	3.4	8.0
Kay	1,003	11.7	19.0	479,927	13.5	-1.3	47,444	11.5	2.1
Kingfisher	1,063	13.2	19.7	552,561	12.8	4.5	88,192	9.2	7.9
Kiowa	662	14.5	11.3	580,490	11.0	-2.5	49,978	8.8	0.8
Latimer	738	12.3	21.8	205,652	10.6	5.7	14,069	10.5	17.5
Le Flore	1,927	9.8	21.1	410,923	9.1	6.5	169,171	6.2	0.4
Lincoln	2,218	10.2	23.7	472,446	11.9	9.5	25,251	8.8	12.8
Logan	1,205	9.4	22.3	365,671	9.6	4.7	41,461	5.2	18.3
Love	725	11.6	22.1	244,465	10.5	6.0	16,590	12.2	11.3
McClain	1,273	9.7	25.0	307,330	10.7	6.6	34,853	6.7	14.5
McCurtain	1,855	11.3	22.1	357,991	12.4	7.5	149,161	5.7	3.1
McIntosh	944	10.7	19.9	266,403	11.0	4.4	17,559	11.0	6.7
Majors	879	12.3	16.0	508,689	11.9	0.6	72,427	4.5	8.6
Marshall	477	8.2	23.1	164,275	7.6	5.6	11,749	5.4	21.6
Mayes	1,552	10.2	23.1	302,172	11.8	8.9	45,361	7.3	12.8
Murray	525	9.7	20.2	202,136	9.4	1.8	20,414	5.8	5.5
Muskogee	1,740	12.0	23.3	351,895	11.2	7.3	36,181	9.8	7.7
Noble	776	12.4	16.0	395,344	12.3	-0.3	34,552	9.6	5.4
Nowata	887	9.8	21.4	311,198	9.7	4.5	29,176	5.6	8.7
Okfuskee	901	12.5	21.0	289,778	11.7	6.8	23,728	7.0	7.9
Oklahoma	1,268	9.0	32.0	172,203	12.2	13.9	21,377	11.3	27.2
Okmulgee	1,268	11.0	24.5	288,969	12.3	7.2	14,607	13.2	8.5
Osage	1,420	10.0	22.5	1,186,354	5.3	1.6	62,550	4.8	5.0
Ottawa	1,137	10.7	22.2	226,436	10.7	10.0	71,321	3.9	3.1
Pawnee	826	8.6	25.2	280,768	9.6	8.6	24,620	6.7	10.5
Payne	1,445	10.4	23.3	340,629	11.1	7.2	28,765	8.5	15.0
Pittsburg	1,687	11.0	20.7	505,047	13.0	0.7	29,901	9.6	9.5
Pontotoc	1,368	8.5	23.5	368,306	9.2	5.5	26,011	7.3	7.3
Pottawatomie	1,663	10.7	24.4	343,119	12.3	8.4	21,842	13.4	12.2
Pushmataha	780	11.8	19.2	309,855	10.3	1.1	9,536	11.0	5.5
Roger Mills	677	11.5	10.2	738,683	9.3	-2.7	27,294	13.8	4.2
Rogers	1,803	8.3	27.6	310,415	8.4	9.1	25,957	7.6	12.9
Seminole	1,167	11.0	22.2	279,262	12.0	7.8	18,483	7.4	11.3
Sequoyah	1,259	10.6	24.9	222,350	12.0	9.7	25,213	7.4	10.7
Stephens	1,359	10.2	23.5	419,923	10.9	6.3	33,984	7.6	18.5
Texas	1,002	10.8	15.3	1,181,025	6.2	0.9	662,508	0.6	0.6
Tillman	592	13.7	13.7	484,533	11.6	-2.6	53,469	8.5	0.5
Tulsa	1,146	8.4	32.1	151,070	9.0	14.2	23,432	4.8	23.4
Wagoner	1,217	8.5	27.1	259,978	7.6	10.5	35,421	6.4	15.6
Washington	847	9.0	26.8	222,882	7.6	7.2	19,726	7.7	10.8
Washita	1,006	12.2	15.3	568,343	11.2	1.1	69,411	11.1	1.7
Woods	761	11.3	12.6	816,386	7.1	0.2	55,551	4.2	2.9
Woodward	842	10.8	14.4	726,473	9.0	-3.4	67,263	2.4	0.8