# Cow-Calf Farm Management: Farm survey evidence from 2007 

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#### Abstract

This study describes and compares cow-calf operations and assesses their relative competitiveness, developing performance measures for a sample of U.S. farms. We find that larger operations tend to be significantly more scale and technically efficient than smaller operations. However, we do not find significant differences in net farm returns by size except on medium large operations-showing virtually no net return on farm assets in 2007. While larger operations are clearly more scale and technically efficient and have lower variable costs per cow, off-farm income makes smaller operations competitive as reflected in higher household returns than all size groups--except for very large cow-calf operations.


## Background:

Beef cow-calf operations vary considerably in size, available resources, profitability, and the use of technology. Opportunities remain to improve management practices, both production and financial, in many cow-calf operations in major cow-calf states (Beef Cattle Manuel). Beef cattle industry analyst Bill Helming recently outlined eight important trends occurring in the U.S. beef cattle industry that either directly or indirectly affect cow-calf operations: 1) consolidation accelerating due to excess capacity, 2) more direct cattle ownership in feedlots and less custom feeding, 3) cattle placement weights increasing due to high energy prices, 4) feedlot backgrounding (i.e. providing high energy rations to bigger calves on cow-calf sites in preparation for shipping at higher weights to feedlots) opportunities on cost-competitive feedlot operations given higher placement weights, 5) feedlot locations moving toward corn production locations, thus putting a greater premium on cutting transportation costs, 6) less flaked corn at the feedlot level and more dry corn and byproducts given high energy prices, 7) increasing domestic and export demand for beef, and 8) brand opportunities with feeding operations and beef packing companies partnering (Feedstuffs November 3, 2008).

In this study, we focus on the consolidation issue using stochastic production frontier (SPF) procedures to estimate the impact of size and off-farm income on competitiveness. We
hypothesize that increasing size and off-farm income from both the operator and the spouse enhance competitiveness.

Beef cow-calf production is relatively widespread and economically important in the United States. Figure 1 identifies the number of beef cows in important Agricultural Statistics Districts (ASDs) and Figure 2 characterizes the relative importance of these ASDs in cow-calf production. According to the 2002 Census of Agriculture, close to 800,000 farms held more than 33 million beef cows (Figure 3). Beef cow inventories ${ }^{1}$ are steady compared to 1997 while farm numbers dropped by about 100,000, suggesting consolidation trends.

Cow-calf operations are located throughout the United States, typically on land not suited or needed for crop production ( http://www.ers.usda.gov/briefing/cattle/Background.htm; Peel). In Figure 4 we see close to half of cow-calf operations are located in ASDs with farms averaging more than 500 acres of pasture. These operations are dependent upon range and pasture forage conditions, which are in turn affected by variations in the average level of rainfall and temperature for the area. Beef cows harvest forage from grasslands to maintain themselves and raise a calf with little, or no grain input, and are generally on lower priced land as shown in Figure 5. The cow is maintained on pasture year round, as is the calf until it is weaned. If additional forage is available at weaning, some calves may be retained for additional grazing and growth until the following spring when they are sold. The average beef cow herd is about 50 head, but operations with 100 or more beef cows comprise more than 9 percent of all beef operations (the same as 1997) and 61 percent of the beef cow inventory, compared to 49 percent in 1997. Operations with 50 or fewer head are largely part of multi-enterprises, or are supplemental to off-farm employment-i.e. hobby farms (USDA/ERS 2001).

[^0]Objectives: This study will: 1) identify the important economic and technical characteristics of cow-calf operations by region-Corn Belt, Northern Plains, Appalachia, Southeast, Delta, Southern Plains, Mountain, and Pacific for the 22 leading cow-calf states (footnote 1), 2) identify characteristics by size- 0 to 120 cows $^{2}, 121$ cows to 300 cows, 301 cows to 500 cows, 501 cows to 1,000 cows and greater than 1000 cows; and 3 ) calculate farm-level economic performance measures and assess factors influencing scale and technical efficiency in on operations with more than 30 beef cows using a stochastic production frontier approach.

Data Sources and Methods: This analysis is based on information from the recently released 2007 ARMS phase III survey, which collects information on the number of beef cows per farm and on costs and returns on these operations ${ }^{3}$. The ARMS data source allows a comparison of costs and returns by size and by region. The 2007 ARMS survey contains 3,915 observations on farms that report beef cows. We will also use recently developed regression techniques that allow us to relate several outputs to several inputs in a single equation to develop measures of technical (best practice production techniques) and scale efficiency scores by farm.

Table 1 presents information on cow-calf production by region in the 22 states analyzed. The western regions--Mountain, Pacific, and Southern Plains--account for close to one-third of cowcalf value of production, based on 2007 ARMS survey data, and along with the Northern Plains and Corn Belt dominant cow-calf production.

[^1]The comparison of summary data at the regional level shown in Table 1 and Figure 3 indicates that stocking rates (potential pasture acres per cow) are substantially higher and variable costs per cow are significantly lower in the three western regions---compared to the remaining regions. Table 1 also shows that relatively little corn production occurs on cow-calf operations in the western regions, on average. These observations suggest different production technologies in the Western regions compared to the eastern regions. However, to give an overview of the competitiveness by size group in the cow-calf industry, our econometric estimates of performance measures will include all regions. Finally, we chose to focus on cowcalf operations with greater than 30 cows. This allows us to capture performance issues in commercial operations while still including smaller operations that rely on off-farm income (see Figures 6 through 8 identifying the pervasiveness of off-farm income, particularly in the Southern regions) in addition to sales from cow-calf operations-thus recognizing the bimodal nature of the cow-calf industry from the Census of Agriculture data.

We use stochastic production frontier (SPF) measurement to econometrically estimate the input distance function $\operatorname{DI}(\mathbf{X}, \mathbf{Y}, \mathbf{R})$ where X refers to a vector of inputs, Y refers to a vector of outputs, and R refers to a vector of environmental or shift factors, such as soil texture and size groupings. Approximating this function by a translog functional form to limit a priori restrictions on the relationships among its arguments results in: where i denotes farm, t time period, $\mathrm{k}, \mathrm{l}$, outputs, $\mathrm{m}, \mathrm{n}$, inputs, and $\mathrm{q}, \mathrm{r}$ the technical/environmental (including for example age or rented land) variables.

This functional relationship, which embodies a full set of interactions among the $\mathbf{X}, \mathbf{Y}$ and $\mathbf{R}$ arguments of the distance function, can be more compactly written as $-\ln X_{1, i t}=T L\left(X / X_{1}, \boldsymbol{Y}, t\right)=$
$T L\left(\boldsymbol{X}^{*}, \boldsymbol{Y}, t\right)^{4}$. We append a symmetric error term, $\mathbf{v}$ to equation (1) to account for noise, and also change the notation "- $\ln$ Dit" to "u". The resulting $-\ln X 1=T L\left(\boldsymbol{X}^{*}, \boldsymbol{Y}, \boldsymbol{R}\right)+v-u$ function (with the sub-scripts suppressed for notational simplicity) may be estimated by maximum likelihood (ML) methods, to impute the TE measures as the distance from the frontier. For the SPF model -u thus represents inefficiency; the efficiency scores generated by FRONTIER ${ }^{5}$ essentially measure $\exp (-U)=D I\left(\boldsymbol{X}^{*}, \boldsymbol{Y}, \mathbf{R}\right)$. This is therefore our measure of technical efficiency.

A parametric input distance function approach is used to estimate performance measures, including RTS (returns to scale) and TE (technical efficiency). The input distance function is denoted as $D^{I}(X, Y, R)$, where $X$ refers to inputs, $Y$ to outputs, and $R$ to other farm efficiency determinants. For the analyses, three outputs developed from the ARMS data for cow-calf farms are: $\mathrm{Y}_{\mathrm{CROP}}=$ value of crop production, $\mathrm{Y}_{\mathrm{LIVE}}=$ value of livestock production, and $\mathrm{Y}_{\mathrm{OFF}}=$ off-farm income. Inputs are: $\mathrm{X}_{\mathrm{LAB}}=$ labor, $\mathrm{X}_{\mathrm{CAP}}=$ capital, $\mathrm{X}_{\mathrm{FEED}}=$ feed and miscellaneous including fertilizer and fuel, and $\mathrm{X}_{\mathrm{OLND}}=$ land.

Estimating $\mathrm{D}^{\mathrm{I}}(\mathrm{X}, \mathrm{Y}, \mathrm{R})$ requires imposing linear homogeneity in input levels (Färe and Primont), which is accomplished through normalization (Lovell et al.); $\mathrm{D}^{\mathrm{I}}(\mathrm{X}, \mathrm{Y}, \mathrm{R}) / \mathrm{X}_{1}=\mathrm{D}^{\mathrm{I}}\left(\mathrm{X} / \mathrm{X}_{1}, \mathrm{Y}, \mathrm{R}\right)=$ $\mathrm{D}^{1}\left(\mathrm{X}^{*}, \mathrm{Y}, \mathrm{R}\right)$. Approximating this function by a translog functional form to limit a priori restrictions on the relationships among its arguments results in:
(2a) $\ln D_{\mathrm{it}}^{1} / X_{1, \mathrm{it}}=\alpha_{0}+\Sigma_{\mathrm{m}} \alpha_{\mathrm{m}} \ln X^{*}{ }_{\mathrm{mit}}+.5 \Sigma_{\mathrm{m}} \Sigma_{\mathrm{n}} \alpha_{\mathrm{mn}} \ln X^{\star}{ }_{\mathrm{mit}} \ln X^{*}{ }_{\text {nit }}+\Sigma_{\mathrm{k}} \beta_{\mathrm{k}} \ln Y_{\mathrm{kit}}$

$$
\begin{aligned}
& +.5 \Sigma_{\mathrm{k}} \Sigma_{\mathrm{l}} \beta_{\mathrm{kl}} \ln Y_{\mathrm{kit}} \ln Y_{\mathrm{lit}}+\Sigma_{\mathrm{q}} \phi_{\mathrm{q}} R_{\mathrm{qit}}+.5 \Sigma_{\mathrm{q}} \Sigma_{\mathrm{r}} \phi_{\mathrm{qr}} R_{\mathrm{qit}} R_{\mathrm{rit}}+\Sigma_{\mathrm{k}} \Sigma_{\mathrm{m}} \gamma_{\mathrm{km}} \ln Y_{\mathrm{kit}} \ln X_{\mathrm{mit}}^{*} \\
& +\Sigma_{\mathrm{q}} \Sigma_{\mathrm{m}} \gamma_{\mathrm{qm}} \ln R_{\mathrm{qit}} \ln X_{\mathrm{mit}}^{*}+\Sigma_{\mathrm{k}} \Sigma_{\mathrm{q}} \gamma_{\mathrm{kq}} \ln Y_{\mathrm{kit}} \ln R_{\mathrm{qit}}+\mathrm{v}_{\mathrm{itt}}=\operatorname{TL}\left(\boldsymbol{X}^{\star}, \boldsymbol{Y}, \boldsymbol{R}\right)+\mathrm{v}_{\mathrm{it},} \text { or }
\end{aligned}
$$

$$
\begin{equation*}
-\ln X_{1, \mathrm{it}}=\mathrm{TL}\left(\boldsymbol{X}^{\star}, \boldsymbol{Y}, \boldsymbol{R}\right)+\mathrm{v}_{\mathrm{it}}-\ln \mathrm{D}_{\mathrm{it}}^{\prime}=\mathrm{TL}\left(\boldsymbol{X}^{\star}, \boldsymbol{Y}, \boldsymbol{R}\right)+\mathrm{v}_{\mathrm{it}}-\mathrm{u}_{\mathrm{it}} \tag{2b}
\end{equation*}
$$

4 By definition, linear homogeneity implies that $\mathrm{D}^{\mathrm{I}}(\omega \mathrm{X}, \mathrm{Y}, \mathrm{R})=\omega \mathrm{D}^{\mathrm{I}}(\mathrm{X}, \mathrm{Y}, \mathrm{R})$ for any $\omega>0$; so if $\omega$ is set arbitrarily at $1 / \mathrm{X}_{1}, \mathrm{D}^{1}(\mathrm{X}, \mathrm{Y}, \mathrm{R}) / \mathrm{X}_{1}=\mathrm{D}^{\mathrm{I}}\left(\mathrm{X} / \mathrm{X}_{1}, \mathrm{Y}, \mathrm{R}\right)$.

5 We used Tim Coelli's FRONTIER package for the SPF estimation, and computed the measures and $t$-statistics for measures using PC-TSP.
where $i$ denotes farm; t the time period; $\mathrm{k}, 1$ the outputs; $\mathrm{m}, \mathrm{n}$ the inputs; and $\mathrm{q}, \mathrm{r}$ the R variables. We specify $\mathrm{X}_{\text {OLND }}$ as land, so the function is specified on a per-acre basis, consistent with much of the literature on farm production in terms of yields.

The distance from the frontier, - $\ln D_{\mathrm{it}}^{\mathrm{it}}$ is explicitly characterized as the technical inefficiency error $-u_{i t}$. As in Battese and Coelli, we use maximum likelihood (ML) methods to estimate (2b) as an error components model. The one-sided error term $u_{i t}$ is a nonnegative random variable independently distributed with truncation at zero of the $N\left(m_{i t}, \sigma_{u}{ }^{2}\right)$ distribution, where $m_{i t}=\boldsymbol{R}_{\mathrm{it}} \delta, \boldsymbol{R}_{\mathrm{it}}$ is a vector of farm efficiency determinants (assumed here to be the factors in the $R$ vector), and $\delta$ is a vector of estimable parameters. The random error component $v_{i t}$ is assumed to be independently and identically distributed, $\mathrm{N}\left(0, \sigma_{\mathrm{v}}{ }^{2}\right)$. More precisely, we estimate a household model with three outputs, crops, livestock, and off-farm income, (measured as earned income relating to wages, agricultural and other rents, and earnings from another businesspassive income such as pensions and social security, interest income etc is not included), and four inputs-labor, miscellaneous expenses, capital, and land.

This function is estimated using SPF techniques. Technical efficiency is characterized assuming a radial contraction of inputs to the frontier (constant input composition). The econometric model includes two error terms to represent the distance from the frontier: a random (white noise) error term, $v_{i t}$, assumed to be normally distributed, and a one-sided error term, $u_{i t}$, assumed to be distributed as a half normal.

The productivity impacts (marginal productive contributions, MPC) of outputs or inputs can be estimated from this model by the first order elasticities, $\mathrm{MPC}_{\mathrm{m}}=-\varepsilon_{D l, Y \mathrm{~m}}=-\partial \mathrm{ln} \mathrm{D}^{\prime}(\boldsymbol{X}, Y, R) / \partial \mathrm{ln}$
$Y_{m}=\varepsilon_{X 1, Y m}$ and $M P C_{k}=-\varepsilon_{D 1, X^{*} m}=-\partial \ln D^{\prime}(X, Y, R) / \partial \ln X^{\star}{ }_{k}=\varepsilon_{X 1, X^{*} k} . M P C_{m}$ indicates the increase in overall input use when output expands (and so should be positive, like a marginal cost or output elasticity measure), and MPC ${ }_{\mathrm{k}}$ indicates the shadow value (Färe and Primont) of the $\mathrm{k}^{\text {th }}$ input relative to $X_{1}$ (and so should be negative, like the slope of an isoquant). Similarly, the marginal productive contributions of structural factors, including soil texture (TEXT), water holding capacity (WATHCA), and urban influences as measured by Nehring et al. (Popacc), can be measured through the elasticities, $\mathrm{MPC}_{R q}=-\varepsilon_{\mathrm{DI}, \mathrm{Rq}}=-\partial \mathrm{ln} \mathrm{D}^{\prime}(\boldsymbol{X}, \boldsymbol{Y}, \boldsymbol{R}) / \partial R_{\mathrm{q}}=\varepsilon_{\times 1, \mathrm{Rq}}$. If $\varepsilon_{\times 1, R q}<0$, an increased $\mathrm{R}_{\mathrm{q}}$ implies that less input is required to produce a given output, which implies enhanced productivity, and vice versa. ${ }^{7}$

Scale economies (SE) are calculated as the combined contribution of the $M$ outputs $Y_{m}$, or the scale elasticity $\mathrm{SE}=-\varepsilon_{\mathrm{DI}, \mathrm{Y}}=-\Sigma_{\mathrm{m}} \partial \mathrm{ln} \mathrm{D}^{\prime}(X, Y, R) / \partial \mathrm{ln} Y_{\mathrm{m}}=\varepsilon_{X 1, Y}$. That is, the sum of the input elasticities, $\Sigma_{\mathrm{m}} \partial \ln \mathrm{X}_{1} / \partial \ln \mathrm{Y}_{\mathrm{m}}$, indicates the overall input-output relationship and thus returns to scale. The extent of scale economies is thus implied by the short-fall of SE from 1 ; if $\mathrm{SE}<1$, inputs do not increase proportionately with output levels, implying increasing returns to scale. Finally, technical efficiency (TE) "scores" are estimated as TE $=\exp \left(-u_{\mathrm{it}}.\right)$. The impact of changes in $\mathrm{R}_{\mathrm{q}}$ on technical efficiency can also be measured by the corresponding $\delta$ coefficient in the inefficiency specification for $-u_{i t}$.

It is assumed that the inefficiency effects are independently distributed, and $u_{i t}$ arise by truncation (at zero) of the normal distribution with mean $\mu_{i t}$, and variance $\sigma^{2}$, where the mean of $\mu_{i t}$ is defined by
(3) $\mu_{\mathrm{it}}=\delta_{0}+\delta_{1}\left(\right.$ Popacc $\left._{\mathrm{it}}\right)+\delta_{2} \ln \left(\mathrm{OPLABOR}_{\mathrm{it}}\right)+\delta_{3} \ln \left(\mathrm{SPLABOR}_{\mathrm{it}}\right)+\delta_{4} \ln \left(\mathrm{TOTAU}_{\mathrm{it}}\right)$

In equation (3), variables are measured as follows: Popacc $_{i \text { t }}$, is an index measured as the degree of urbanization by county (see Nehring et al.), OPLABOR ${ }_{i t}$ represents hours of
operator hours worked off farm, SPLABOR ${ }_{i t}$ represents hours of spouse hours worked off farm, and TOTAU measures the total number of animal units on the farm. The $\delta_{1}-$ parameter, measuring the effect of urbanization on the inefficiency model in equation (3), is expected to have a negative effect on the size of the inefficiency effects. That is, higher urbanization is negatively related to technical efficiency. The sign on the $\delta_{2}$-parameter, the $\delta_{3}$-parameter, and the $\delta_{4}$-parameter, measuring the impacts of labor and total animal units, is less clear. Evidence in Fernandez et al. suggests that operator hours worked off farm are negatively related to technical efficiency-the argument being that off-farm work by the operator in particular is inimical to best practice farming on managerially intensive dairy operations. Evidence in Kompas relating to dairy farms suggests that total animal units are positively related to technical efficiency.

## Stochastic Frontier Results

More than one-half of the estimated coefficients from the input distance function are significant as shown in Table 3, including the own price on labor, and the own cross price effects for crops, livestock, and off-farm income. All of the measures of outputs and inputs have the expected signs, positive for outputs and negative for inputs, as shown in Table 4. All are significant or marginally significant except for capital. Among the inefficiency effects, we find that operator off-farm hours are positively associated with higher technical efficiency-spouse off-farm hours are only marginally significant, but tend to suggest the notion that spouse hours off-farm also boost technical efficiency. And, we also find that operations with more animal units (including all species) are more technically efficient than operations with smaller livestock populations.

## Conclusions

We find that larger operations tend to be significantly more scale and technically efficient than smaller operations. However, we do not find significant differences in net farm returns by size except on medium large operations-which showed virtually no net return on farm assets in 2007. While larger operations are clearly more scale and technically efficient and have lower variable costs per cow, off-farm income makes smaller operations competitive as reflected in higher household returns than all size groups except for very large cow-calf operations. In future research the availability of more detailed cost of production information will facilitate identifying competitiveness by region and size.

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Table1. Cost and Production Means and Statistics by Region all observations in the $\mathbf{2 2}$ most important cow-calf states, 2007

| Item | Corn Belt | Northern Plains | Appalachia | Southeast | Delta | Southern Plains | Mountain | Pacific |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Number of | 654 | 931 | 262 | 657 | 390 | 709 | 162 | 150 |
| Observations |  |  |  |  |  |  |  |  |
| Percent of farms | 14.0 | 12.4 | 15.7 | 10.4 | 6.7 | 31.4 | 5.0 | 4.2 |
| Percent of value of production | 17.2 | 27.7 | 7.1 | 9.7 | 7.9 | 17.0 | 9.1 | 4.4 |
| Percent of pasture acres | - 4.3 | 16.8 | 2.9 | 3.2 | 2.1 | 35.0 | 31.0 | 4.8 |
| Percent of corn acres | : 33.3 | 49.2 | 5.1 | 1.7 | 0.5 | 7.1 | 3.0 | 1.0 |
| Percent of hay acres | 13.7 | 30.4 | 12.4 | 3.5 | 4.6 | 21.1 | 11.1 | 3.1 |
| Beef Cows per Farm | $45.1{ }^{\text {BCGH }}$ | $91.1{ }^{\text {ACDEFGH }}$ | $27.8{ }^{\text {ABDEFGH }}$ | $40.2{ }^{\text {BCFGH }}$ | $43.4{ }^{\text {BCGH }}$ | $50.5^{\text {BCDG }}$ | $111.9^{\text {ABCDEFH }}$ | $61.2^{\text {ABCDG }}$ |
| Net Return on Assets Farm (\%) | $2.7{ }^{\text {JBDEF }}$ | $4.8{ }^{\text {ACDEFH }}$ | $1.8{ }^{\text {BDEF }}$ | $0.7{ }^{\text {ABCG }}$ | $0.4{ }^{\text {ABCG }}$ | 0.5 | $2.7{ }^{\text {DEF }}$ | $1.7^{\text {B }}$ |
| Net Return on Assets All In (\%) | $6.2{ }^{\text {DH }}$ | $7.7{ }^{\text {DH }}$ | $8.3{ }^{\text {DH }}$ | $4.7{ }^{\text {ABCEF }}$ | $7.4{ }^{\text {DH }}$ | $8.1{ }^{\text {DH }}$ | 5.4 | $4.4{ }^{\text {ABCEF }}$ |
| Variable Cost per Cow \$ | $1,343{ }^{\text {BCFGH }}$ | $1,077^{\text {ACDEFGH }}$ | 1,159 ${ }^{\text {AFGH }}$ | $1,345^{\text {FBFGH }}$ | $1,411{ }^{\text {BFGH }}$ | $768{ }^{\text {ABCDE }}$ | $663{ }^{\text {ABCDE }}$ | $775^{\text {ABCDE }}$ |
| Land price (\$/acre) | 2,484 ${ }^{\text {BCDFG }}$ | 861 ACDEFH | 3,354 ABEFGH | 4,102 ABEFGH | 2,233 ${ }^{\text {BCDEFG }}$ | 1,194 ABCDEGH | 696 ACDEFH | $2,053{ }^{\text {BCDFG }}$ |
| Off-farm income/ total Income (\%) | $26.8{ }^{\text {BCDEF }}$ | $15.6{ }^{\text {ACDEFH }}$ | $55.7{ }^{\text {ABCDGH }}$ | $38.8{ }^{\text {ABCFG }}$ | $50.1{ }^{\text {ABGH }}$ | $52.6{ }^{\text {ABCDGH }}$ | $26.4{ }^{\text {CDEF }}$ | $31.8{ }^{\text {BCEF }}$ |
| Contracts/total production (\%) | $21.5{ }^{\text {CDE }}$ | $16.7^{\text {CDE }}$ | $49.2{ }^{\text {ABEF }}$ | $60.3{ }^{\text {CDE }}$ | $76.0{ }^{\text {ABCDFGH }}$ | $14.0{ }^{\text {CDE }}$ | $28.0^{\text {DE }}$ | $39.3{ }^{\text {E }}$ |
| Operator hrs offfarm | $493{ }^{\text {CEF }}$ | $538{ }^{\text {CF }}$ | $836{ }^{\text {ABGH }}$ | $658{ }^{\text {G }}$ | $690^{\text {AG }}$ | $805{ }^{\text {ABGH }}$ | $363{ }^{\text {CDEF }}$ | $587{ }^{\text {CF }}$ |
| Spouse hours offfarm | $467{ }^{\text {B }}$ | $617^{\text {ACEFH }}$ | $436{ }^{\text {B }}$ | 428 | $399{ }^{\text {B }}$ | $488{ }^{\text {B }}$ | $753{ }^{\text {H }}$ | $317{ }^{\text {B }}$ |
| FORAGE |  |  |  |  |  |  |  |  |
| Potential pasture acres/cow | $2.60{ }^{\text {BDFGH }}$ | 4.91 ACDEFg | $2.99{ }^{\text {BDEFGH }}$ | $3.304^{\text {ABCFGH }}$ | $2.95{ }^{\text {BCFGH }}$ | $8.14{ }^{\text {ABCDEG }}$ | $21.53{ }^{\text {ABCDEFH }}$ | $7.23{ }^{\text {ACdeg }}$ |
| Purchased feed/ total costs | $18.5{ }^{\text {DEH }}$ | $20.2{ }^{\text {DEH }}$ | $21.4{ }^{\text {DEH }}$ | $35.3{ }^{\text {ABCEF }}$ | 53.0 ABCDFGH | $19.3{ }^{\text {DEH }}$ | $31.2{ }^{\text {E }}$ | $36.0{ }^{\text {ABCEF }}$ |
| Hay yield (tons/ac) | $2.35{ }^{\text {CH }}$ | $2.05{ }^{\text {EFH }}$ | $1.86{ }^{\text {ADEFH }}$ | $2.57{ }^{\text {C }}$ | $2.46{ }^{\text {BCH }}$ | $2.59{ }^{\text {BC }}$ | $2.17{ }^{\text {H }}$ | $3.44{ }^{\text {ABCEG }}$ |

Source: Authors' analysis of USDA Agricultural Resource Management Survey USDA (2007). a. The t-statistics are based on 3,951 observations using weighting techniques described in Dubman. A through J indicate significant differences in means across columns with $A=$ Corn Belt, $B=$ Northern Plains, $C$
= Appalachia, $D=$ Southeast, $E=$ Delta, $F=$ Southern Plains, $G=$ Mountain, $H=$ Pacific.
Table 2. Economic performance by size groupings, 2007
$\left.\begin{array}{llcccc}\hline \text { Item } & \begin{array}{l}30 \text { to } 120 \text { Beef } \\ \text { Cows }\end{array} & \begin{array}{l}121 \text { to } 300 \\ \text { Beef Cows }\end{array} & \begin{array}{l}301 \text { to } 500 \\ \text { Beef Cows }\end{array} & \begin{array}{l}501 \text { to } 1000 \\ \text { Beef Cows }\end{array} & \begin{array}{l}\text { Greater than } \\ 1000 ~ B e e f ~\end{array} \\ \text { Cows }\end{array}\right]$

S̄ource: Authors' analysis of USDA Agricultural Resource Management Survey USDA (2007). a. The t-statistics are based on 2,582 observations using weighting techniques described in Dubman. A through J indicate significant differences in means across columns with $\mathrm{A}=$ cow-calf operations with 30 to 120 cows, $B=$ cow-calf operations with 121 to 300 cows, $C=$ cow-calf operations with 301 to 500 cows, $\mathrm{D}=$ cow-calf operations with 501 to 1000 cows, and $\mathrm{E}=$ cow-calf operations with more than 1000 cows.

Table 3. Input Distance Function Parameter Estimates, 2007 Cow-calf

| Variable | Parameter t-test | Parameter t-test | Parameter t-test | Parameter t-test |
| :---: | :---: | :---: | :---: | :---: |
| $\alpha_{0}$ | 10.164 (11.45)*** |  |  |  |
| $\alpha_{\text {xLAB }}$ | -0.581 (-24.01)*** |  |  |  |
| $\alpha_{\text {XFEED }}$ | -0.166 (-4.14)**** |  |  |  |
| $\alpha_{\text {xCAP }}$ | -0.051 (-1.63) |  |  |  |
| $\beta_{\text {YCROP }}$ | -0.008 (-0.22) |  |  |  |
| $\beta_{\text {YIIVE }}$ | -0.259 (-1.73) |  |  |  |
| $\beta_{\text {YofF }}$ | -0.048 (-0.82) |  |  |  |
| $\beta_{\text {Ycrop, ycrop }}$ | 0.016 (9.69)*** |  |  |  |
| $\beta_{\text {Ylive, ylive }}$ | 0.031 (5.12)*** |  |  |  |
| $\beta_{\text {Yoff, YofF }}$ | 0.022 (5.92)** |  |  |  |
| $\beta_{\text {Ychop, YLive }}$ | -0.013 (-4.82)*** |  |  |  |
| $\beta_{\text {Ychop, yoff }}$ | -0.003 (-3.31)** |  |  |  |
| $\beta_{\text {YLive, yoff }}$ | -0.018 (-7.22)*** |  |  |  |
| $\gamma_{\text {YLIVE, text }}$ | 0.003 (0.58) |  |  |  |
| $\gamma_{\text {ylive, wathcap }}$ | -0.005 (-1.23) |  |  |  |
| $\gamma_{\text {Ycrop, URBan }}$ | 0.001 (0.93) |  |  |  |
| $\alpha_{\text {xLAB }, \text { хLAB }}$ | 0.060 (4.07)*** |  |  |  |
| $\alpha_{\text {XFEED , XFEED }}$ | 0.005 (0.55) |  |  |  |
| $\alpha_{\text {xCAP }, \text { xCAP }}$ | -0.001 (-0.13) |  |  |  |
| $\alpha_{\text {XLAB }, \text { XfeED }}$ | -0.046 (-3.45)*** |  |  |  |
| $\alpha_{\text {XLAB, }{ }^{\text {cCAP }}}$ | -0.027 (-3.04)** |  |  |  |
| $\alpha_{\text {XFEED, XCAP }}$ | 0.019 (2.04)* |  |  |  |
| $\alpha_{\text {xPassoum }}$ | 0.114 (2.41)** |  |  |  |
| $\alpha_{\text {xSMALL }}$ | 0.207 (4.02)*** |  |  |  |
| $\alpha_{\text {xneorum }}$ | 0.248 (3.24)** |  |  |  |
| $\alpha_{\text {xLARGE }}$ | 0.146 (1.19) |  |  |  |
| $\delta_{\text {INEFF EFFECTS }}$ | -4.670 (-1.71) |  |  |  |
| $\delta_{\text {POPACC }}$ | -0.050 (-0.09) |  |  |  |
| $\delta_{\text {OPABOR }}$ | -0.002 (-2.77)** |  |  |  |
| $\delta_{\text {SPLABOR }}$ | -0.001 (-1.42) |  |  |  |
| $\delta_{\text {totau }}$ | -0.002 (-6.55)** |  |  |  |
| $\delta^{2}$ | 2.134 (1.93)* |  |  |  |
|  | 0.920 (18.00)*** |  |  |  |
| Log-likelihood | -152,955 |  |  |  |

Notes: *** Significance at the 1\% level ( $\mathrm{t}=2.977$ ). ** Significance at the 5\% level ( $\mathrm{t}=2.145$ ). and * Significance at the 10\% level ( $\mathrm{t}=1.761$ ).
Source: USDA Agricultural Resource Management Study. USDA (2007).
The $t$-statistics are based on 2,582 observations, using weighting techniques described in Dubman.

Table 4: MPC's for outputs and inputs (t-statistics in parentheses)

| MPC ${ }_{\text {YCROP }}$ | 0.010 | (3.66)*** | MPC ${ }_{\text {xLAB }}$ | -0.530 | $(-12.18){ }^{* * *}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| MPC YıLVE | 0.200 | (8.55)*** | MPC XFEED | -0.190 | $(-3.07)^{\star * *}$ |
| MPC YOFF | 0.100 | (1.62) | MPC xCAP | -0.006 | (-0.97) |
|  |  |  | MPC XOLND | -0.220 | (-7.02)*** |

Notes: *** Significance at the $1 \%$ level ( $\mathrm{t}=2.977$ ). ** Significance at the $5 \%$ level ( $\mathrm{t}=2.145$ ). and * Significance at the $10 \%$ level ( $\mathrm{t}=1.761$ ) .

Source: USDA Agricultural Resource Management Study. USDA (2007).
The t -statistics are based on 2,852 observations using weighting techniques described in Dubman's CV15 program.


Figure 1. Average Number of Beef Cows per Farm by ASD (Agricultural Statistics District), based on 2007 ARMS phase III survey data.


Figure 2. Percent value of Production by ASD relative to the entire sample (value of all farm outputs on all cow-calf operations in an ASD-3\% in central California, e.g.-- relative to all production in the samplepercentages in the table sum to a 100), based on 2007 ARMS phase III survey data.


Figure 3. Beef cow inventory 2002, U.S. Census of Agriculture.


Figure 4. Average Pasture Potential (acres) per farm by ASD, (only beef cow-calf Operations), where Pasture Potential acres are equal to acres operated less harvested crop acres, based on 2007 ARMS phase III survey data.


Figure 5. Average Price of Land Per Acre by ASD, based on 2007
ARMS phase III survey data (value of acres operated/acres operated on cow-calf farms).


Figure 6. Percent of earned income relative to total income by ASD, based on 2007 ARMS phase III survey data, (on cow-calf operations).


Figure 7. Operator hours worked off-farm, average per farm by ASD, based on 2007 ARMS phase III survey data (on cow-calf operations).


Figure 8. Spouse hours worked off-farm, average per farm by ASD, based on 2007 ARMS phase III survey data (on cow-calf opeations).


[^0]:    1 Nationally pure bred cow-calf operations account for close to 6 percent of beef operations and mixed pure bred

[^1]:    and commercial operations account for more than 21 percent http://www.aphis.usda.gov/vs/ceah/ncahs/nahms/. In this study we do not differentiate between these operations and commercial cow-calf operations.
    2 Size groupings used in the tables were chosen to correspond to actual beef cows-including beef heifers that had calved----per farm and are arbitrary groupings. The SPF estimation also includes all other beef animals on the beef cow farm, and all other livestock on the farm. For example, in Table 2 the group with 30 to 120 beef cows, as defined above, averages 55.5 beef cows, 86.5 beef animals (including beef cows), 9.8 hogs, 0.9 dairy, and 1,744 poultry per farm.
    3 States and their designated regions included in this dataset include: NORTHERN PLAINS: KS, NE, ND, SD; DELTA: AR, LA; CORN BELT: IA, MO; APPALACHIA: KY, TN, VA; SOUTHEAST: AL, FL, GA; SOUTHERN PLAINS: OK,TX; MOUNTAIN WEST: AZ, CO, NM, WY; and PACIFIC: CA, OR. These 22 states will be included in the 2008 ARMS Cost of Production Survey.

