

Yale University

EliScholar – A Digital Platform for Scholarly Publishing at Yale

Cowles Foundation Discussion Papers

Cowles Foundation

4-1-1993

Measuring the Impact of Global Warming in Agriculture

Robert Mendelsohn

William D. Nordhaus

Daigee Shaw

Follow this and additional works at: <https://elischolar.library.yale.edu/cowles-discussion-paper-series>



Part of the [Economics Commons](#)

Recommended Citation

Mendelsohn, Robert; Nordhaus, William D.; and Shaw, Daigee, "Measuring the Impact of Global Warming in Agriculture" (1993). *Cowles Foundation Discussion Papers*. 1288.

<https://elischolar.library.yale.edu/cowles-discussion-paper-series/1288>

This Discussion Paper is brought to you for free and open access by the Cowles Foundation at EliScholar – A Digital Platform for Scholarly Publishing at Yale. It has been accepted for inclusion in Cowles Foundation Discussion Papers by an authorized administrator of EliScholar – A Digital Platform for Scholarly Publishing at Yale. For more information, please contact elischolar@yale.edu.

COWLES FOUNDATION FOR RESEARCH IN ECONOMICS
AT YALE UNIVERSITY

Box 2125, Yale Station
New Haven, Connecticut 06520

COWLES FOUNDATION DISCUSSION PAPER NO. 1045

Note: Cowles Foundation Discussion Papers are preliminary materials circulated to stimulate discussion and critical comment. Requests for single copies of a Paper will be filled by the Cowles Foundation within the limits of the supply. References in publications to Discussion Papers (other than mere acknowledgment by a writer that he has access to such unpublished material) should be cleared with the author to protect the tentative character of these papers.

MEASURING THE IMPACT OF GLOBAL WARMING ON AGRICULTURE

Robert Mendelsohn, William D. Nordhaus and Daigee Shaw

April 1993

Measuring the Impact of Global Warming on Agriculture

Robert Mendelsohn

William D. Nordhaus

Daigee Shaw

February 23, 1993

Mendelsohn is Associate Professor of Forest Policy in the School of Forestry and Environmental Studies and Nordhaus is Professor of Economics in the Department of Economics and on the staff of the Cowles Foundation, Yale University, New Haven, Connecticut. Dr. Shaw is with the Institute of Economics, Academia Sinica, Taipei, Republic of China. We thank the National Science Foundation and Economic Research Service, USDA for funding this project although all views are the authors alone. We would also like to thank John Miranowski, Daniel Hellerstein, John Reilly, Katherine Segerson, Paul Waggoner, the Editor, and referees for their assistance and advice. Correspondence regarding the paper should be addressed to Yale FES, 360 Prospect Street, New Haven, CT 06511.

Abstract

This paper develops a "Ricardian" approach for measuring the economic impact of environmental factors such as climate by examining the direct impact of the environmental factor on land prices. Estimating the model using cross-sectional data on climate, farm-land prices, and other economic and geophysical data for almost 3000 counties in the United States, we find that higher temperatures in all seasons except autumn reduce average farm values in the United States and more precipitation in all seasons except autumn increases farm values. Applying the model to a global-warming scenario finds a range of impacts depending upon whether the model emphasize the grains through crop-land weights or a broader set of crops through crop-revenue weights. The results of the Ricardian approach show a significantly lower estimated impact of global warming than the traditional production-function approach.

I. INTRODUCTION

Over the last decade, scientists have extensively studied the greenhouse effect, which holds that the accumulation of carbon dioxide (CO₂) and other greenhouse gases (GHGs) is expected to produce global warming and other significant climatic changes over the next century. Numerous studies indicate major impacts on agriculture, especially if there is significant midcontinental drying and warming in the U.S. heartland.¹ Virtually every estimate of economic impacts relies on a technique we denote the production-function approach.

This study compares the traditional production-function approach to estimating the impacts of climate change with a new "Ricardian" approach that examines the impact of climate and other variables on land values and farm revenues. The traditional approach to estimating the impact of climate change relies upon empirical or experimental production functions to predict environmental damage (hence its label in this study as the production-function approach).² This approach takes an underlying production function and estimates impacts by varying one or a few input variables, such as temperature, precipitation, and carbon-dioxide levels. The estimates might rely on extremely carefully calibrated crop-yield models (such as CERES or SOYGRO) to

¹ See particularly the reports of the Intergovernmental Panel on Climate Change (IPCC [1990]) and the National Academy of Sciences Panel on Greenhouse Warming (NAS [1991]).

² Important studies include Adams et al. [1988], Adams [1989], Adams et al. [1990], Callaway et al. [1982], Decker et al. [1986], Rind et al. [1990], and Rosenzweig [1986]. For useful surveys, see NAS [1983], EPA [1989], NAS [1991], and Cline [1992].

determine the impact upon yields; the results often predict severe yield reductions as a result of global warming.

While providing a useful baseline for estimating the impact of climate change on farming, these studies have an inherent bias that will tend to overestimate the impact. This bias is sometimes called the "dumb-farmer scenario" to suggest that it omits a variety of the adaptations that farmers customarily make in response to changing economic and environmental conditions. Most studies assume little adaptation and simply calculate the impact of changing temperature on farm yields. Others allow limited changes in fertilizer application, irrigation, or cultivars (see Easterling et al. [1991]). None permits a full adjustment to changing environmental conditions by the farmer. For example, the literature does not consider the introduction of completely new crops (such as tropical crops in the south); technological change; changes in land use from farming to livestock, grassland, or forestry; or conversion to cities, retirement homes, campsites, or the 1001 other productive uses of land in a modern post-industrial society.

By not permitting a complete range of adjustments, previous studies have overestimated damages from environmental changes. Figure 1 shows the hypothetical values of output in four different sectors as a function of a single environmental variable, temperature, in order to illustrate the general nature of the bias. In each case, we assume that the production-function approach yields an accurate assessment of the economic value of the activity as a function of temperature. The four functions are a simplified example of how the value of wheat, corn, grazing, and retirement homes might look as a function of the

temperature. For example, the curve to the far left is a hypothetical "wheat production function," showing how the value of wheat varies with temperature, rising from cold temperatures such as point A, then peaking at point B, finally falling as temperatures rise too high. A production-function approach would estimate the value of wheat production at different temperatures along this curve.

The bias in the production-function approach arises because it fails to allow for economic substitution as conditions change. For example, when the temperature rises above point C, adaptive and profit-maximizing farmers will switch from wheat to corn. As temperature rises, the production-function approach might calculate that the yield has fallen to F in wheat, but wheat is in reality no longer produced; the realized value is actually much higher, at point D where corn is now produced. At a slightly higher temperature, the land is no longer optimally used for corn but switches to grazing, and production-function estimates that do not allow for this conversion will again overestimate the losses from climate change. Finally, at point E, even the best agricultural model will predict that the land is unsuitable for farming or grazing and that the damage is severe. A more complete approach might find that the land has been converted to retirement villages, to which old folks flock so they can putter around in the warm winters and dry climates.

All this is of course illustrative. But it makes the crucial point that the production-function approach will overestimate the damages from climate change because it does not, and probably cannot, take into account the infinite variety of substitutions, adaptations, and old and new activities that may displace

no-longer-advantageous activities as climate changes.

In this study, we develop a new technique that in principle can correct for the bias in the production-function technique by using economic data on the value of land. We call this the Ricardian approach, in which, instead of studying yields of specific crops, we examine how climate in different places affects the net rent or value of farm land. By directly measuring farm prices or revenues, we account for the direct impacts of climate on yields of different crops as well as the indirect substitution of different inputs, introduction of different activities, and other potential adaptations to different climates. If markets are functioning properly, the Ricardian approach will allow us to measure the economic value of different activities and therefore to verify whether the economic impacts implied by the production-function approach are reproduced in the field.

The results of the Ricardian approach can be seen in Figure 1. We assume that the "value" measured along the vertical axis is the net yield per acre of land; more precisely, it is the value of output less the value of all inputs (excluding land rents). Under competitive markets, the land rent will be equal to the net yield of the highest and best use of the land. This rent will in fact be equal to the heavy solid line in Figure 1. We label the solid line in Figure 1 the "best-use value function."

In general, we do not observe market land rents because land rent is generally a small component of the total profits, which includes also the return on capital items. However, with farms, land rents tend to be a large fraction of total rent. Farm value is the present value of future rent. If the interest rate, rate

of capital gains, and capital per acre are equal for all parcels, then farm value will be proportional to the land rent. Therefore, by observing the relationship of farm values to climatic and other variables, we can infer the shape of the solid, best-use value function in Figure 1.

This study measures the effect of climatic variables on agriculture. We examine both climatic data and a variety of fundamental geographical, geophysical, agricultural, economic, and demographic factors to determine the intrinsic value of climate on farming. The unit of observation is the U. S. county in the lower 48 states. We examine the effect of climatic variables as well as non-climatic variables on both land values and farm revenue, and the analysis includes a number of urban variables in order to measure the potential effect of development upon agriculture land values. The analysis suggests that climate has a systematic impact on agricultural rents through temperature and precipitation. These effects tend to be highly nonlinear and vary dramatically by season. The paper concludes with a discussion of the impacts of global warming on American farms.

II. MEASURING THE EFFECT OF CLIMATE ON AGRICULTURE

This section sketches the analytical apparatus that underlies the valuation of climate in this study. Assuming a set of utility-maximizing consumers with well behaved utility functions and linear budget constraints leads to a system of inverse demand functions for all outputs, $P_i = D_i^{-1}(Q_1, Q_2, \dots, Q_n, Y)$, where P_i and Q_i are respectively the price and quantity of good i , $i = 1, \dots, n$, and Y is aggregate income.

In addition, assume well-behaved production functions that link purchased inputs and exogenous environmental inputs with agricultural outputs of good i on site s , $Q_{i,s} = F^{i,s}(\mathbf{K}_{i,s}, \mathbf{E})$, $i = 1, \dots, n$, where $F^{i,s}$ is the production function. We here use bold face to denote vectors or matrices. $\mathbf{K}_{i,s}$ is a vector of purchased inputs in the production of good i at site s ($s = 1, \dots, S$), and \mathbf{E} is a vector of exogenous environmental inputs such as climate, water availability, and soil quality (we omit the subscript for each site as it is inessential to the analysis). Given environmental inputs, production functions, and a vector of factor prices, \mathbf{R} , cost minimization leads to a cost function, $C_{i,s} = \phi^{i,s}(Q_{i,s}, \mathbf{R}, \mathbf{E})$, where $C_{i,s}$ is the cost of production and $\phi^{i,s}$ is the cost function of good i at site s .

Profit maximization will lead to maximizing profits on each site. The rent on a site is equal to the maximum profit (excluding rent) for all possible production activities. The rent per acre is given by $\rho_s = \max[\pi(Q_{1,s}), \dots, \pi(Q_{i,s}), \dots, \pi(Q_{n,s})]$, where $\pi(Q_{i,s}) [= P_i Q_{i,s} - \phi^{i,s}(Q_{i,s}, \mathbf{R}, \mathbf{E})]$ is the maximum profit per acre earned on site s by production activity i . (These are the envelopes for each sector shown in Figure 1.) From elementary production theory, if z is the good for which $\rho_s = \max[\pi(Q_{1,s}), \dots, \pi(Q_{i,s}), \dots, \pi(Q_{n,s})]$, then $\pi(Q_{z,s}) = 0$ and $Q_{z,s} > 0$, and if specialization occurs $\pi(Q_{i,s}) < 0$ and $Q_{i,s} = 0$ for all $i \neq z$. Moreover, with a competitive land market, if q_s is the price per acre and r_s is the discount rate (which reflects interest rates, risk, capital gains, and taxes), then $q_s = \rho_s / r_s$.

The issue we are analyzing is the impact of exogenous changes in environmental variables on net economic welfare. Consider an environmental change from environmental state **A** to state **B**, which leads environmental inputs

to change from E^A to E^B . Given quantities of land L_s at each site, the change in economic welfare is given by:

$$(1) V(E^B) - V(E^A) = \int_0^{Q_B} [\sum_i D^{-1}_i(Q_i) dQ_i - \sum_{i,s} L_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^B)] \\ - \int_0^{Q_A} [\sum_i D^{-1}_i(Q_i) dQ_i - \sum_{i,s} L_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^A)]$$

where \int is the line integral evaluated between the initial vector of quantities and the zero vector. Q_k is the optimized output and cost for environment k , $k = A$ or B . If there is a single output, (1) is replaced by the conventional consumer-surplus integral for a single good.

The impact in (1) can be decomposed into two parts. First, costs per acre have changed for the production of good i from $\sum_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^A)$ to $\sum_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^B)$. Second, production has changed from Q_A to Q_B . If we examine changes in the environment which leave market prices unchanged,³ then (1) can be written as $V(E^B) - V(E^A) = [P Q_B - \sum_{i,s} L_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^B)] - [P Q_A - \sum_{i,s} L_s \phi^{i,s}(Q_{i,s}, \mathbf{R}, E^A)]$, where P is the vector of prices. Because rents are equal to the net revenues, we can substitute and obtain:

³If market prices are affected, one must also include measures of consumer surplus change. For the cross-sectional analysis performed here, the assumption of exogeneity of input prices seems realistic. On the other hand, studies by John Reilly and others suggest that including the linkages through international trade may have a significant effect on the impacts of global warming, generally buffering the impacts in individual regions by allowing for price and output adjustments.

$$(2) \quad V(E^B) - V(E^A) = \sum_s (\rho_s^B - \rho_s^A) L_s \quad .$$

where ρ_s^k is the competitive rent for site s in environmental condition $k = A, B$. Equation (2) provides the definition of the Ricardian value of environmental changes. Under the assumptions used here, the value of the change in the environmental variable is captured exactly by the change in the value of land rent between the different environmental conditions.

An important characteristic of markets is that firms adjust their purchased inputs in order to adapt to the changing environment, and the Ricardian measure of environmental impact incorporates this adaptive behavior. We can see the difference between adaptive and non-adaptive behavior by rewriting (1) in terms of revenues minus costs:

$$(3) \quad V(E^B) - V(E^A) = [\sum_{i,s} P_i F^{i,s}(K_{i,s,B}, E^B) L_s - \sum_{i,s} R K_{i,s,B} L_s] \\ - [\sum_{i,s} P_i F^{i,s}(K_{i,s,A}, E^A) L_s - \sum_{i,s} R K_{i,s,A} L_s]$$

As E deteriorates from E_A to E_B , one would expect that farmers would adjust their purchases of K from K_A to K_B to reduce some of the losses, although the exact form of the adaptation will generally be too complex to measure or observe. If one fails to incorporate these adjustments by firms and instead assumes that K is fixed at K_A , then (3) becomes:

$$(4) \quad V(E_B) - V(E_A) = \sum_{i,s} P_i [F^{i,s}(K_{i,s,A}, E^B) L_s - F^{i,s}(K_{i,s,A}, E^A) L_s] \quad .$$

Equation (4) is an approach which measures the impact as the partial effect of

the environmental change on the value of output. The difference between (3) and (4) is crucial to impact studies. (3) contains all the adaptations and adjustments to the changing environmental conditions, while (4) allows no (or, more generally, incomplete) adjustment. The latter measure illustrates the production-function approach in which all or a subset of inputs or adjustments are held constant while the environment changes. "Scientific" approaches use the approach represented in (4) when they tightly control all factors except for the environmental change. While the controlled experiment has the apparent attractiveness of isolating the influences of the environmental change, it actually is actually the opposite of the appropriate approach, which incorporates all the adjustments and adaptations by farmers.

The Ricardian measure in (3), which includes all optimizing adaptations, is a better predictor of environmental impact than the gross-revenue or production-function measure in (4), which contains only partial adaptation. Because the net economic welfare from adjusting all inputs and outputs is greater than that from incomplete adjustment, the estimated damage from climate change will always be greater under the production-function approach than under the Ricardian approach.

III. AN APPLICATION OF THE RICARDIAN TECHNIQUE TO AGRICULTURE

In this section, we apply the Ricardian technique by estimating the value of climate in U. S. agriculture. Agriculture is the most appealing application of the Ricardian technique both because of the significant impact of climate on agricultural productivity and because of the extensive county-level data on farm inputs and outputs.

Sources and methods⁴

The basic hypothesis is that climate shifts the production function for crops. Farmers at particular sites take environmental variables like climate as given and adjust their inputs and outputs accordingly. Moreover, we assume perfect competition in both product and input markets. Most important, we assume that the economy has completely adapted to the given climate so that land prices have attained the long-run equilibrium that is associated with each county's climate.

For the most part, the data are actual county averages, from the 1982 U. S. Census of Agriculture, so that there are no major issues involved in obtaining information on these variables. The County and City Data Book, and the computer tapes of those data, are the source for much of the agricultural data used here, including values of farm products sold per acre, farm land and building values,⁵ and information on market inputs for farms in every county in the United States. In addition, in many specifications, we include social, demographic, and economic data on each of the counties; these as well are drawn from the County and City Data Book.

Data about soils were extracted from the National Resource Inventory (NRI) with the kind assistance of Drs. Daniel Hellerstein and Noel Gollehon of the U. S. Department of Agriculture. The NRI is an extensive survey of land characteristics in the United States. For almost 800,000 sites, NRI has collected

⁴ Appendix A contains a complete description and definition of the variables used in this study.

⁵ The definition and source of the farm value variable is critical to this study and its derivation is described in Appendix B.

soil samples, or land characteristics, each providing a measure of salinity, permeability, moisture capacity, clay content, sand content, flood probability, soil erosion (K factor), rain erosion (R factor), slope length, wind erosion, whether or not the land is a wetland, and numerous other variables that are not used in this analysis. Each sample also contains an expansion factor, which is an estimate of the amount of land the sample represents in that county. Using these expansion factors, we aggregate these data to yield an overall county estimate for each soil variable.

Climatic data pose more difficult issues. They are available by meteorological station rather than by county, so it was necessary to estimate county-average climates. To begin with, climate data were obtained from the National Climatic Data Center, which gathers data from 5511 meteorological stations throughout the United States. These stations form a dense set of observations for most regions of the United States with the exception of some of the desert Southwest. The data include information on precipitation and temperature for each month from 1951 through 1980. Since the purpose of this study is to predict the impacts of climate changes on agriculture, we focus on the long-run impacts of precipitation and temperature on agriculture, not year-to-year variations in weather. We consequently examine the climatological normal variables -- the 30-year average of each climatic variable for every station. In this analysis, we collect data on normal daily mean temperatures and normal monthly precipitations for January, April, July, and October. We focus on these four months in order to capture seasonal effects of each variable. For example, cold January temperatures may be important as a control on insect pests, warm-but-not-hot summers may be good for crop growth, and warm October temperatures may assist in crop harvesting.

In order to link the agricultural data which are organized by county and the climate data which are organized by station, we conduct a spatial statistical analysis that examines the determinants of the climate of each county. Although the specific climatic variables we analyze in this study have been measured frequently, there are some counties with no weather stations and others with several. Some of the weather stations are not in representative locations, such as the station on the top of Mt. Washington. Furthermore, some counties are large enough or contain sufficient topographical complexity that there is variation of climate within the county. We therefore proceeded by constructing an average climate for each county.

First, we assume that all the weather stations within 500 miles of the geographic center of the county provide some useful climate information. The 500-mile circle invariably draws in many stations, so that our measure does not depend too heavily on any one station.

Second, we estimate a climate surface in the vicinity of the county by running a weighted regression across all weather stations within 500 miles. The weight is the inverse of the square root of a station's distance from the county center because we recognize that closer stations contain more information about the climate of the center. We estimate a separate regression for each county since the set of stations within 500 miles and the weights (distances) are unique for each county. The dependent variables are the monthly normal temperatures and precipitations for January, April, July, and October for the 30 year period. The independent variables include latitude, longitude, altitude, and distance from closest shoreline. The regression fits a second-order polynomial over these four basic variables, including interactive terms, so that there are 14 final variables in the regression, plus a constant term. Eight

regressions (4 seasons times 2 measures) for each of 3000 counties leads to over 24,000 estimated regressions.

Third, we calculate the predicted value of each climatic variable for the geographic center of the county. The predicted values of normal precipitation and temperature from the climate regressions are the independent variables for climate in the property value regressions. This complicated interpolation procedure is intended to provide geographically accurate estimates of the climatic variables for each county.

The estimates of the climate parameters for individual counties are too numerous to present, but we show two selected counties in Tables 1-A and 1-B. These show the independent variables as well as the coefficients and summary regression statistics for Fresno, California and Des Moines, Iowa. Note that more coefficients are significant in the Fresno than the Des Moines regressions. There is more variation across the sample in Fresno because of the effects of the coast and nearby mountain ranges. Although there are more significant coefficients in the California regression, the Iowa regression has a better overall fit and smaller standard errors. In general, the fit east of 100 degrees longitude (the east slope of the Rocky Mountains) was tighter than in the West.

In order to gain some sense of the reliability of this geographic approximation method, we predicted the climate for each of the weather stations. Dropping the weather station itself, we predicted the climatic variables for the station from all stations within 500 miles in the manner explained above. Comparing these results with the actual measurements from each station reveals that the approximation method predicts between 87% and 97% of the variation in precipitation in the continental United States and between 97% and 99% of

the variation in temperature. It should be noted that, even in a statistically stationary environment, the observations of "climate" themselves contain error because they contain only 30 observations. Depending upon the relative importance of idiosyncratic error in climate vs. misspecification error in our equation, the predictions might actually be superior to the recorded observations themselves. In any case, the predictions serve as sophisticated interpolations of the climate between stations.

IV. EMPIRICAL RESULTS

The Ricardian approach estimates the importance of climate and other variables on farm-land values. As noted above, land values are the expected present value of future rents. There is little reason for the riskless interest rate to vary across counties in the U. S., but the risk and capital-gains components of land value might vary considerably. For example, California agricultural land near growing cities might well have a larger capital-gains component than would rural land in an economically stagnant coal-mining region of Appalachia. Moreover, there are major potential errors in measurement of land values since values are estimated by farmers, and such estimates are often unreliable. However, there is no reason to believe that the errors of measurement are correlated with independent data such as temperature or precipitation. The major effect of measurement errors will be imprecision of the econometric estimates rather than bias in the estimation of the coefficients or in the estimate of the economic value of climate on agriculture.

We regress land values on climate, soil, and socioeconomic variables to estimate the best-value function across different counties. There are 2933 cross sectional observations. The means have been removed from the independent

variables in this regression. The quadratic climate variables are consequently easier to interpret. The linear term reflects the marginal value of climate evaluated at the U. S. mean while the quadratic term shows how that marginal effect will change as one moves away from the mean.

We present several regressions in Table 2. In order to give a sense of the importance of the non-farm variables in the model, we begin with a model which contains only climate variables. The first set of regressions in Table 2 is a quadratic model which includes the eight measures of climate (four months of precipitation and temperature). For each variable, linear and quadratic terms are included to reflect the nonlinearities that are apparent from field studies.

In the remainder of regressions, we include urban, soil, and other environmental variables to control for extraneous factors influencing land values and farm revenues. This raises the question of how the counties should be weighted. A first set of weights is the farm-land weights, which weights observations by the percent of each county in crop land. Counties with a large fraction of crop land should provide a better reading on price determination because other influences, such as cities or forests, are minimized; these results are particularly useful for the grain belt. A second set of regressions uses crop-revenue weights, that is, observations are weighted by the aggregate value of crop revenue in each county. This second weighting scheme emphasizes those counties which are most important to total agricultural production, even though some of the counties might have their land values affected by large neighboring cities; it also places greater weight on counties where more valuable crops are grown. On the whole, the crop-land measure tends to emphasize the corn, wheat, and soybean belt and therefore reflects the influence of climate on the grains. The crop-revenue weights, by contrast, give more influence to the truck farms

and citrus belt of the coast lands, and the crop-revenue regressions thus reflect a broader definition of agriculture.

The results of this analysis are shown in Table 2. The squared terms for most of the climate variables are significant, implying the observed relationships are nonlinear. However, some of the squared terms are positive, especially for precipitation, implying that there is a minimally productive level of precipitation and that either more or less precipitation will increase land values. The negative quadratic coefficient implies that there is an optimal level of a climatic variable from which the value function decreases in both directions.

The overall impact of climate as measured by the marginal impacts are largely the same across the different models although the quantitative estimates vary. All models suggest that higher winter and summer temperatures are harmful for crops; that higher fall temperatures and higher winter and spring rains are beneficial for crops; but that higher summer and fall rains are harmful. The two weighting schemes differ, however, in terms of their assessment of the relative importance of winter versus summer temperature. The crop-land model finds higher winter temperatures less harmful, valuing a one degree increase by between \$89 to \$103 per acre, whereas the crop-revenue model finds this effect more harmful, with estimated impacts between \$138 and \$160 per acre. However, a one-degree increase in summer temperature decreases farm values by only \$88 to \$132 according to the crop-revenue model but between \$155 and \$177 in the crop-land model. Except for spring rains, the crop-revenue model suggests that rain has a much larger effect on land value than the crop-land model. For example, the crop-revenue model suggests that winter rain increases farm values between \$172 and \$280 per monthly inch whereas the crop-land model suggests an effect between \$57 and \$85 per monthly inch.

The predicted overall effects from the existing climate across the United States are shown in Figures 2 and 3. These maps show the Ricardian values of climate by county in 1982 -- that is, the partial effect of climate on property values. To construct each map, we begin with the difference between the estimated climate for each county and the national average climate. We then multiply these differences by the estimated coefficients in Table 2 and sum them across the climate variables. Figures 2 and 3 show the estimated contribution of climate to the farm-land value in each county. The results match folk wisdom about farm values (for example, the infamous 100th meridian of American history can be seen sharply in Figure 2). The most valuable climates are along the west coast, the corn belt near Chicago, and the northeast. The least valuable areas are the southwest and southeast regions. Both figures show almost identical geographic patterns, indicating that the results are stable; similar results were also found using 1978 data.

The control variables in Table 2 provide a rich set of results in and of themselves. Economic and soil variables play a role in determining the value of farms. Farm values are higher in denser, growing, and wealthier counties because of higher local demand for food and the potential for conversion of land to non-farm uses. Farm values respond as expected to other environmental factors such as solar flux (latitude) and altitude. Salinity, likelihood of flooding, wetlands and soil erosion all act negatively as expected. Slope length was slightly beneficial to land values. Irrigation is left out of the regressions shown in Table 2 because irrigation is clearly an endogenous reaction to climate. However, when included, irrigation is a strongly positive variables, increasing land values substantially; which is not surprising, given the crucial importance

of irrigation in many areas of the arid West.⁶

One hypothesis suggested in the theory section is that the impacts of environmental effects would be exaggerated by a gross-revenue model. We explore this hypothesis in Table 3 by regressing the same climate and control variables on the gross revenue earned from crops. The marginal effects in Table 3 for the farm revenue model suggest similar seasonal patterns as the farm value equation with the exception of spring. Warmer Aprils reduce farm revenues whereas they increase farm values. Wetter springs, good for farm values, reduce farm revenues according to the crop-land model but increase farm revenues according to the crop-revenue model.

The magnitude of effects predicted by the gross-revenue model, however, are generally larger than the effects predicted by the Ricardian model. To compare the two approaches, we need to translate the annual rents into land value using the discount rate defined in section II. Based on asset returns and farm earnings, a real discount rate of 5 percent per annum appears most suitable.⁷ At this discount rate, the marginal coefficients in Table 3 should be multiplied by 20 to make them comparable with the present value estimates in Table 2. Making this adjustment, a one-degree increase in summer temperature decreases the present

⁶Including irrigation does not significantly change the results of the paper.

⁷According to Ibbotson and Brinson [1987], farmland prices over the period 1947-84 had a compound annual return (income and capital gains) of 9.6 percent while the GNP deflator rose at an average of 4.4 percent annually. This produces an average real yield of 4.99 percent per annum. By comparison, all real estate investments had an average real yield of 4.4 percent per annum over this period. Another comparison is the rate of profit on farms, defined as the net income of farms divided by total value of farms and farm land. For the three Census years of 1974, 1978, and 1982, the average rate of profit on farms was 5.02 percent per annum.

value of farms by between \$140 and \$540 according to the gross-revenue model but only between \$88 and \$177 according to the Ricardian models.

One concern with the Ricardian approach to climate effects is that the results may not be robust over time and that the weather and economic factors in a given year may have distorted the results. We consequently estimate the model again using data from 1978. These values have been converted to 1982 dollars using the GNP deflator obtained from the 1991 Economic Report of the President. The 1978 results are surprisingly similar to the findings using the 1982 data. The control variables have similar impacts in both years. Evaluating the marginal effects of climate in 1978 at the national mean and comparing the results with 1982 shows that the climatic variables are also similar in 1978 and 1982 with few exceptions. The pattern of climate effects on agriculture is stable over time but apparently some factors can alter the magnitude of the effects from year to year.

V. IMPLICATIONS FOR GREENHOUSE WARMING

The Ricardian analysis in the previous section shows that climate has complicated effects on agriculture, highly nonlinear and varying by season. An important application of this analysis is to project the impact of global warming on American agriculture. For this projection, we take a conventional CO₂ doubling scenario, which is associated with a 5 degree F increase in global mean surface temperature.⁸ According to most projections, such an increase will occur sometime in the second half of the next century if current trends continue. According to the survey by the IPCC, a 5 degree F temperature increase will be

⁸ See IPCC [1990] and NAS [1992].

accompanied by an 8 percent average increase in precipitation. These changes are applied uniformly by season and region to the U. S. in the calculations that follow. In principle, they show the impact of climate change including all adaptations, although they omit the impact of CO₂ fertilization and price effects.

Table 4 shows the results of this experiment for the two years and sets of weights. The first column shows the estimated impact of global warming on farmland values; the second column truncates these losses if they drive land values below zero. This second column is the preferred economic measure. The estimates diverge dramatically depending upon whether crop land or crop revenues are used for weighting. Under the crop-land weights, the loss in land value from warming ranges from \$119 to \$163 billion; assuming that the annual crop loss is 5 percent of this value,⁹ the annual loss ranges from \$6 to \$8 billion annually (in 1982 prices at 1978 or 1982 levels of output). This can be related to gross farm income in 1982 of \$164 billion. The crop-land model emphasizes the unattractiveness of a warmer climate for an agriculture that emphasizes grains, which have relatively low value per acre and thrive in the relatively cool climate of the northern U. S.

Strikingly different results emerge if we use the crop-revenue approach. For these, the net impact of warming (again without CO₂ fertilization) is slightly positive, ranging from a reduction of \$14 billion to an increase of \$35 billion in farmland values. To annualize these capital losses, we use the estimated real discount rate of 5 percent discussed above. At this rate, these changes represent a change of from -\$0.7 to plus \$2 billion per year. The differing results arise because the crop-revenue approach weights relatively more heavily the irrigated

⁹ See the discussion of this issue in the last section and in footnote 6.

lands of the West and South that thrive in a Mediterranean and subtropical climate, a climate that will become relatively more abundant with a warmer climate. Including this broader set of crops and adaptations paints a more optimistic picture because the gains from the sunbelt crops tend to offset the losses in the marginal grain regions.

The striking difference between the crop-revenue and crop-land approaches is a useful reminder of how we can be misled by our mental images. The specter of global warming calls up the a vista of corn blistering on the stalk or desiccated wheat fields. Yet the major grains so vulnerable to drought--wheat and corn--represented only \$22.5 billion of the \$143 billion of farm marketings in 1982. Our results suggest that the vulnerability of American agriculture to climate change may be may be exaggerated if the analysis is limited to the major grains. A broader vision should also include the warm-weather crops such as cotton, fruits, vegetables, rice, hay, and grapes in addition to the other sectors such as livestock and poultry. Whereas past production-function studies focus ominously on the vulnerable cool-weather grains, the comprehensive crop-revenue Ricardian model reminds us that the irrigated warm-weather crops may be a silver lining behind the climate-change cloud.

Figures 4 and 5 provide geographic detail for these global warming scenarios. According to the crop-land model shown in Figure 4, warming will be particularly harmful for the entire southern part of the U. S. and will only be beneficial to the northern fringe of the country. The crop-revenue model of Figure 5 suggests, by contrast, that global warming will be beneficial to California and the citrus belt of the southeast as well as the corn and wheat belts of the midwest. Global warming will be harmful, in this model, only to the relatively unimportant mountainous regions of Appalachia and the Rocky

Mountains.

It will be useful to compare these estimates with results from other studies. In its 1989 analysis, the U. S. Environmental Protection Agency [1989] surveyed a number of different climate and agriculture models to estimate the impact of CO₂ doubling. Omitting CO₂ fertilization, the EPA concluded that the impact would lie in the range of \$6 billion to \$34 billion per year (in 1982 prices). Cline [1992] used two different approaches, the EPA estimate and a modification of Rind et al. [1990], both of which produce losses of \$20 billion per year without CO₂ fertilization. It is instructive to note that all these studies rely on the production-function approach and apply it to grains; these estimates therefore are comparable to our crop-land model, and for that universe show, as was predicted in the theoretical section above, a higher estimate of damage than the Ricardian approach -- approximately triple the estimates in Table 4. By excluding the non-grain, warm-weather crops, these studies further bias upward the estimates of damage as is shown by the comparison between the crop-land and the crop-revenue models.

The results in Table 4 are based on a highly stylized global-warming scenario and are therefore highly tentative. In research underway, we are drawing actual global warming results from large-scale general circulation models; these should allow differentiation among broad regions of the United States. In addition, the effects of CO₂ fertilization should be included, for some studies indicate that this may produce a significant increase in yields. Other omitted variables are the effect of extremes and ranges in climatic variables as well as the effect of changes in irrigation. Notwithstanding these omissions, the present paper does provide a benchmark for projecting the impact of global warming on American agriculture. Using the narrow definition of crops, the

impact is estimated to lie between \$6 and 8 billion per year (in 1982 prices and output levels). Using a more inclusive definition that weighs more heavily warm-weather crops and irrigated agriculture, our projections suggest global warming may be slightly favorable to American agriculture.

REFERENCES

Adams, R., B. McCarl, D. Dudek, and J. Glycer [1988]. "Implications of Global Climate Change for Western Agriculture," Western Journal of Agricultural Economics 13, pp. 348-356.

Adams, R. [1989]. " Global Climate Change and Agriculture: An Economic Perspective," American Journal of Agricultural Economics, 71, pp. 1272-1279.

Adams, R., C. Rosenzweig, R. Pearl, J. Ritchie, B. McCarl, J. Glycer, R. Curry, J. Jones, K. Boote, and L. Allen [1990]. "Global Climate Change and U. S. Agriculture," Nature, 345, pp. 219-224.

Callaway, J., F. Cronin, J. Currie, J. Tawil [1982]. An Analysis of Methods and Models for Assessing the Direct and Indirect Economic Impacts of CO₂-Induced Environmental Changes in the Agricultural Sector of the U.S. Economy, Richland WA, Pacific Northwest Laboratory, PNL-4384.

Cline, William [1992]. The Economics of Global Warming, Washington, D. C., Institute of International Economics.

Decker, W. V. Jones, R. Achutuni [1986]. The Impact of Climate Change from Increased Atmospheric Carbon Dioxide on American Agriculture, Washington D.C., U. S. Dept. of Energy, DOE/NBB-0077.

Easterling, W.,III, M. McKenney, N. Rosenberg, and K. Lemon [1991]. Processes for Identifying Regional Influences of and Responses to Increasing Atmospheric CO₂ and Climate Change- The MINK Project- Report IIB Washington, D.C., U. S. Dept. of Energy, DOE/RL/01830T-H8.

EPA [1989]. The Potential Effects of Global Climate Change on the United States: Report to Congress, Washington D.C., U. S. Environmental Protection Agency, EPA-230-05-89-050.

Freeman, M. [1979]. The Benefits of Environmental Improvement, Baltimore, Resources for the Future, Johns Hopkins University Press.

Ibbotson, Roger G. and Gary P. Brinson [1987]. Investment Markets, New York, McGraw-Hill.

IPCC [1990]. Intergovernmental Panel on Climate Change, Climate Change: The IPCC Scientific Assessment, J.T. Houghton, G.J. Jenkins, and J.J. Ephraums, eds., New York, Cambridge University Press.

NAS [1991]. Policy Implications of Greenhouse Warming: Report of the Adaptation Panel, Washington D.C., National Academy Press.

National Research Council [1983]. Changing Climate, Washington, D. C., National Academy Press.

Rind, D., R. Goldberg, J. Hansen, C. Rosenzweig, and R. Ruedy [1990]. "Potential Evapotranspiration and the Likelihood of Future Drought," Journal of Geophysical Research, vol. 95, no. D7, pp. 9983-10004.

Rosenzweig, C. [1986]. "Effects on Agriculture" Chapter III in T. Tirpak, ed., Potential Effect of Future Climate Changes on Forests and Vegetation, Agriculture, Water Resources, and Human Health, Washington, D.C., U. S. EPA.

Appendix A. Definition of Major Variables Used in This Study

Variable

Definition

Constant: a term equal to one.

January temp: normal daily mean temperature from 1951-1980 in the month of January, Fahrenheit

Jan temp sq: January temp squared

April temp: normal daily mean temperature from 1951-1980 in the month of April, Fahrenheit

Apr temp sq: April temp squared

July temp: normal daily mean temperature from 1951-1980 in the month of July, Fahrenheit

July temp sq: July temp squared

October temp: normal daily mean temperature from 1951-1980 in the month of October, Fahrenheit

Oct temp sq: October temp squared

January rain: normal precipitation from 1951-1980 in the month of January, inches

Jan rain sq: January rain squared

April rain: normal precipitation from 1951-1980 in the month of April, inches

Apr rain sq: April rain squared

July rain: normal precipitation from 1951-1980 in the month of July, inches

July rain sq: July rain squared

October rain: normal precipitation from 1951-1980 in the month of October, inches

Oct rain sq: October rain squared

Income per capita: annual personal income per person in the county, 1984

Density: resident population per square mile, 1980

Density sq: density squared

Latitude: latitude measured in degrees from southern most point in U S.

Altitude: height from sea level in feet

Migration: net of incoming people minus outgoing people from 1980 to 1986 for the county

Salinity: percent of land which needs special treatment because of salt/alkaline in the soils

Flood prone: percent of land which is prone to flooding

Irrigated: percent of land where irrigation provides at least 50% of water needs

Wetland: percent of land considered wetland

Soil erosion: k factor-soil (erodibility factor) in hundredths of inches

Slope length: length of slope, number of feet (not steepness)

Wind erosion: measure of wind erosion in hundredths of inches

Farm value: estimate of the current market value of farm land including buildings for the county expressed in dollars per acre, 1982

Farm revenue: gross revenue from crops sold in 1982 for the county in dollars per acre

Sdist: distance in ruler from the nearest shoreline

Long: longitude measured in degrees from the eastern most point of the U.S.

Permeability: soil permeability in inches per hour

Moisture capacity: available water capacity in inches/lb

Appendix B. Data on farms and value of land and buildings¹⁰

The data on farms and on farm land values are central to this study. This appendix describes the definition and sources of the data. The current definition of a farm, first used for the 1974 Census of Agriculture final reports, is any place from which \$1,000 or more of agricultural products were sold or normally would have been sold during the census year. Land in farms is an operating-unit concept and includes land owned and operated as well as land rented from others. The acreage designated as "land in farms" consists primarily of agricultural land used for crops, pasture, or grazing. It also includes woodland and wasteland not actually under cultivation or used for pasture or grazing, provided it was part of the farm operator's total operation.

The land is defined to lie in the operator's principal county, that is, the county where the largest value of agricultural products was raised or produced. Irrigated land includes land watered by any artificial or controlled means, such as sprinklers, furrows or ditches, and spreader dikes. Crop land includes land from which crops were harvested or hay was cut, land in orchards, citrus groves, vineyards, nurseries, and greenhouses, land used only for pasture or grazing that could have been used for crops without additional improvement, and all land planted in crops that were grazed before the crops reached maturity. Also included were all crop land used for rotation pasture and land in government diversion programs that were pastured.

Respondents were asked to report their estimate of the current market value of land and buildings owned, rented, or leased from others, and rented or leased to others. Market value refers to the respondent's estimate of what the land and buildings would sell for under current market conditions. If the value of land and buildings was not reported, it was estimated during processing by using the average value of land and buildings from a similar farm in the same geographic area.

The value of products sold by farms represents the gross market value before taxes and production expenses of all agricultural products sold or removed from the place regardless of who received the payment. In addition, it includes the loan value received in 1982 for placing commodities in the Commodity Credit Corporation loan program.

¹⁰ This description is drawn from the City and County Data Book, and the underlying data are from U.S. Bureau of the Census, 1982 Census of Agriculture.

Table 1-A
Interpolation for County Climate Measures (Fresno, Ca)^a

	Temperature			Precipitation		
	April	July	October	April	July	October
Constant	131535	231764	124970	-58846	-184063*	16551
Longitude	-32.8*	-59.6*	-29.2	26.7	45.2*	1.96
Latitude	-13.2	-18.2	-16.8	-19.6	21.7*	-16.33
Lat sq	1.9E-4	2.8E-4	4.1E-4	1.6E-3	-3.1E-4	1.6E-3*
Long sq	2.0E-3*	3.8E-3*	1.7E-3	-2.3E-3	-2.7E-3*	-3.9E-4
Long*Lat	1.8E-3	2.8E-3	2.1E-3	1.5E-3	-2.9E-3*	1.1E-3
Altitude	-.56*	-1.44*	-1.00*	.525	1.28*	1.48*
Alt sq	-1.6E-6*	-3.0E-6*	-2.3E-6*	-3.7E-6*	-6.5E-7*	-2.4E-6*
Lat*Alt	4.3E-5	8.8E-5	7.7E-5*	-4.8E-5	-1.1E-4*	-1.1E-4*
Long*Alt	6.2E-5	1.8E-4*	1.1E-4*	-4.6E-5	-1.5E-4*	-1.7E-4*
Shore dist	-40.4*	-74.5*	-35.2	-5.47	59.4*	-26.6
Sdist sq	2.6E-3	4.2E-3	2.2E-3	2.9E-3	-4.9E-3*	4.8E-3*
Sdist*Long	5.2E-3*	9.6E-3*	4.2E-3	-1.3E-3	-6.7E-3*	2.6E-3
Sdist*Lat	2.0E-3	3.7E-3	2.3E-3	4.3E-3	-4.9E-3*	2.7E-3
Sdist*Alt	6.7E-5	1.3E-4	9.7E-5*	-1.9E-4	-7.0E-5*	-2.3E-4*
Adj R ²	.999	.998	.999	.796	.777	.706
Std err	.13	.24	.13	.54	.13	.30
Observations	331	331	331	525	525	525

^a Variables marked with an asterisk are significant at the 5% level. Temperature is measured in Fahrenheit and precipitation in inches per month.

Table 1-B

Interpolation for County Climate Measures (Des Moines, Iowa)^a

	Temperature			Precipitation		
	April	July	October	April	July	October
Constant	6425	5006	8967	-32243	77324*	41650
Longitude	-.919	-1.12	-2.55	7.72	-15.8*	-9.61
Latitude	-2.48	-.829	-1.55	10.0	-32.9*	-16.32
Lat sq	2.5E-4	2.0E-5	3.2E-5	-9.7E-4	3.2E-3*	1.6E-3
long sq	3.7E-5	8.1E-5	2.0E-4	-4.9E-4	6.8E-4	5.9E-4
Long*Lat	2.0E-4	1.0E-4	2.4E-4	-9.9E-4	3.8E-3*	1.8E-3
Altitude	-.13	.046	.34*	.353	3.02*	2.09*
Alt sq	-1.2E-6	-1.3E-6*	1.6E-6*	1.1E-5*	-1.5E-6	2.1E-5*
Lat*Alt	2.1E-5	-1.6E-5	-6.9E-5*	-1.2E-4	-5.7E-4*	-2.8E-4*
Long*Alt	1.1E-5	-9.7E-6	-4.9E-5*	-3.1E-5	-3.6E-4*	-3.2E-4*
Shore dist	1.14	-1.17	-.564	-.150	26.8	18.6
Sdist sq	1.8E-4	-3.1E-4	-1.9E-4	5.8E-4	-1.2E-3	1.4E-3
Sdist*Long	-4.4E-5	1.9E-4	-1.2E-4	-4.1E-4	-2.7E-3	-1.9E-3
Sdist*Lat	-3.6E-4	2.2E-4	9.0E-5	4.2E-4	-5.4E-3*	-3.8E-3
Sdist*Alt	-2.2E-5	3.2E-5	9.9E-5*	-1.7E-4	6.9E-4*	3.6E-4*
Adj R ²	.999	.999	.999	.989	.987	.976
Std err	.04	.04	.04	.14	.17	.15
Observations	928	928	928	1477	1477	1477

^a Variables marked with an asterisk are significant at the 5% level. Temperature is measured in Fahrenheit and precipitation in inches per month.

Table 2
Regression Models Explaining Farm Values^a

Independent Variables	(1) Crop-land Weights		(3)	(4) Crop-revenue weights		(5)
	1982	1982	1978	1982	1978	1978
Constant	1490. (71.20)	1329. (60.18)	1173. (57.95)	1451. (46.36)	1307. (52.82)	
January temp	-57.0 (6.22)	-88.6 (9.94)	-103. (12.55)	-160. (12.97)	-138. (13.83)	
Jan temp sq	-0.33 (1.43)	-1.34 (6.39)	-2.11 (11.03)	-2.68 (9.86)	-3.00 (14.11)	
April temp	-137. (10.81)	-18.0 (1.56)	23.6 (2.23)	13.6 (1.00)	31.8 (2.92)	
Apr temp sq	-7.32 (9.42)	-4.90 (7.43)	-4.31 (7.11)	-6.69 (9.44)	-6.63 (11.59)	
July temp	-167. (13.10)	-155. (14.50)	-177. (18.07)	-87.7 (6.80)	-132. (122.55)	
July temp sq	-3.81 (5.08)	-2.95 (4.68)	-3.87 (6.69)	-.30 (0.53)	-1.27 (2.82)	
October temp	351.9 (19.37)	192. (11.08)	175. (11.01)	217. (8.89)	198. (9.94)	
Oct temp sq	6.91 (6.38)	6.62 (7.09)	7.65 (8.93)	12.4 (12.50)	12.4 (15.92)	
January rain	75.1 (3.28)	85.0 (3.88)	56.5 (2.81)	280. (9.59)	172. (7.31)	
Jan rain sq	-5.66 (1.86)	2.73 (0.95)	2.20 (0.82)	-10.8 (3.64)	-4.09 (1.72)	
April rain	110. (4.03)	104. (4.44)	128. (5.91)	82.8 (2.34)	113. (4.05)	
Apr rain sq	-10.8 (1.17)	-16.5 (1.96)	-10.8 (1.41)	-62.1 (5.52)	-30.6 (3.35)	
July rain	-25.6 (1.87)	-34.5 (2.63)	-11.3 (0.94)	-116. (6.06)	-5.28 (0.34)	
July rain sq	19.5 (3.42)	52.0 (9.43)	37.8 (7.54)	57.0 (8.20)	34.8 (6.08)	
October rain	-2.30 (0.09)	-50.3 (2.25)	-91.6 (4.45)	-124. (3.80)	-135. (5.15)	
Oct rain sq	-39.9 (2.65)	2.28 (0.17)	.25 (0.02)	171. (14.17)	106. (11.25)	
Income per Capita Density		71.0 (15.25)	65.3 (15.30)	48.5 (6.36)	47.1 (7.39)	
Density		1.30 (18.51)	1.05 (16.03)	1.53 (18.14)	1.17 (17.66)	
Density sq		-1.72e-4	-9.33e-5	-2.04e-4	-9.38e-5	

	(5.31)	(3.22)	(7.47)	(4.57)
Latitude	-90.5 (5.12)	-94.4 (6.95)	-105. (5.43)	-85.8 (5.33)
Altitude	-0.167 (6.09)	-0.161 (6.41)	-0.163 (4.72)	-0.149 (5.20)
Salinity	-684. (3.34)	-416. (2.20)	-582. (2.59)	-153. (0.81)
Flood prone	-163. (3.34)	-309. (6.98)	-663. (8.59)	-740. (11.99)
Wetland	-58.2 (0.47)	-57.5 (0.51)	762. (4.41)	230. (1.72)
Soil erosion	-1258. (6.20)	-1513. (8.14)	-2690. (8.21)	-2944. (11.23)
Slope length	17.3 (2.91)	13.7 (2.49)	54.0 (6.24)	30.9 (4.54)
Sand	-139. (2.72)	-35.9 (0.77)	-288. (4.16)	-213. (3.95)
Clay	86.2 (4.08)	67.3 (3.47)	-7.90 (0.22)	-18.0 (0.63)
Moisture capacity	.377 (9.69)	.510 (14.21)	.206 (3.82)	.450 (10.07)
Permeability	-.002 (1.06)	-.005 (2.53)	-.013 (5.58)	-.017 (8.61)
Adj R ²	.671	.782	.784	.836
Observations	2938	2938	2941	2941

^a The dependent variable is the value of land and buildings per acre. All regressions are weighted. Values in parenthesis are t-statistics.

Table 3

Regression Models Explaining Farm Revenues^a

Independent Variables	(1) Crop-land Weights		(3) Crop-revenue Weights	
	1982	1978	1982	1978
Constant	180. (31.37)	143. (28.09)	213. (16.61)	186. (16.27)
January temp	-11.6 (5.00)	-6.65 (3.21)	16.1 (3.19)	16.4 (3.55)
Jan temp sq	-.048 (0.88)	.006 (0.13)	.867 (7.80)	.659 (6.71)
April temp	-23.5 (7.89)	-20.3 (7.63)	-47.7 (8.62)	-39.3 (7.83)
Apr temp sq	-1.31 (7.67)	-1.12 (7.43)	-2.74 (9.43)	-2.26 (8.55)
July temp	-27.2 (9.85)	-21.5 (8.66)	-10.0 (1.90)	-7.20 (1.49)
July temp sq	.053 (0.32)	-.166 (1.14)	1.27 (5.52)	.341 (1.65)
October temp	51.3 (11.43)	41.4 (10.43)	-2.12 (0.21)	2.92 (0.32)
Oct temp sq	.637 (2.62)	.598 (2.85)	-.025 (0.06)	.569 (1.58)
January rain	30.1 (5.29)	21.4 (4.26)	-28.9 (2.42)	-11.5 (1.06)
Jan rain sq	-4.10 (5.49)	-2.93 (4.49)	-4.08 (3.36)	-3.33 (3.04)
April rain	-22.5 (3.67)	-23.2 (4.29)	47.5 (3.28)	16.0 (1.24)
Apr rain sq	-2.46 (1.12)	4.65 (2.39)	-5.73 (1.24)	2.65 (0.63)
July rain	-3.29 (0.97)	2.12 (0.70)	-64.5 (8.25)	-33.3 (4.61)
July rain sq	10.8 (6.93)	6.74 (5.23)	22.8 (8.03)	13.2 (5.02)
October rain	-40.2 (6.93)	-16.1 (3.17)	-44.4 (3.32)	-16.3 (1.35)
Oct rain sq	27.2 (7.73)	17.4 (5.62)	33.8 (6.84)	9.32 (2.15)
Income per Capita Density	.568 (0.47)	.803 (0.73)	3.37 (1.08)	8.24 (2.81)
Density sq	.172 (9.46)	.133 (8.47)	.457 (13.28)	.280 (9.14)
Density sq	2.86e-6	2.92e-6	-4.47e-5	-1.92e-5

	(0.34)	(0.43)	(3.99)	(2.03)
Latitude	-24.3 (6.28)	-15.4 (4.44)	-72.6 (9.15)	-41.6 (5.59)
Altitude	-.049 (6.91)	-.033 (5.03)	-0.096 (6.78)	-0.059 (4.47)
Salinity	-156. (2.97)	-149. (3.23)	-502. (5.44)	-427. (4.90)
Flood prone	29.8 (2.36)	25.4 (2.27)	-40.7 (1.29)	-1.45 (0.05)
Wetland	70.9 (2.21)	64.8 (2.32)	234. (3.31)	115. (1.86)
Soil erosion	-169. (3.18)	-74.5 (1.60)	-413. (3.08)	-360. (2.98)
Slope length	-1.18 (0.73)	-1.21 (0.85)	-15.3 (4.33)	-13.5 (4.31)
Sand	28.7 (2.18)	32.3 (2.84)	70.3 (2.49)	46.7 (1.88)
Clay	11.1 (1.99)	12.3 (2.49)	-48.1 (3.32)	-31.8 (2.43)
Moisture	.062	.050	.101	.058
Capacity	(6.10)	(5.49)	(4.57)	(2.79)
Permeability	.001 (2.22)	.001 (2.15)	-.001 (6.94)	-.005 (5.30)
Adj R ²	.525	.509	.800	.762
Observations	2834	2443	2834	2443

^a The dependent variable is the gross value of crop revenue per acre per year. All regressions are weighted. Values in parenthesis are t-statistics.

Table 4

Predicted Impact of Global Warming on Farm-land Values
 [Billions of dollars, 1982 prices]

Year	Weight	Impact	Truncated Impact
1982	Crop land	-125.2	-118.8
1978	Crop land	-162.8	-141.4
1982	Crop revenue	34.5	34.8
1978	Crop revenue	-14.0	21.0

Note: The global-warming scenario is a uniform 5 degree F increase with a uniform 8% precipitation increase. The "impact" column shows the estimated loss in value of U. S. crop land; the "truncated impact" shows the impact when the loss in land value is limited to the original value of the land.

Figure 1
Bias in Production Function Studies

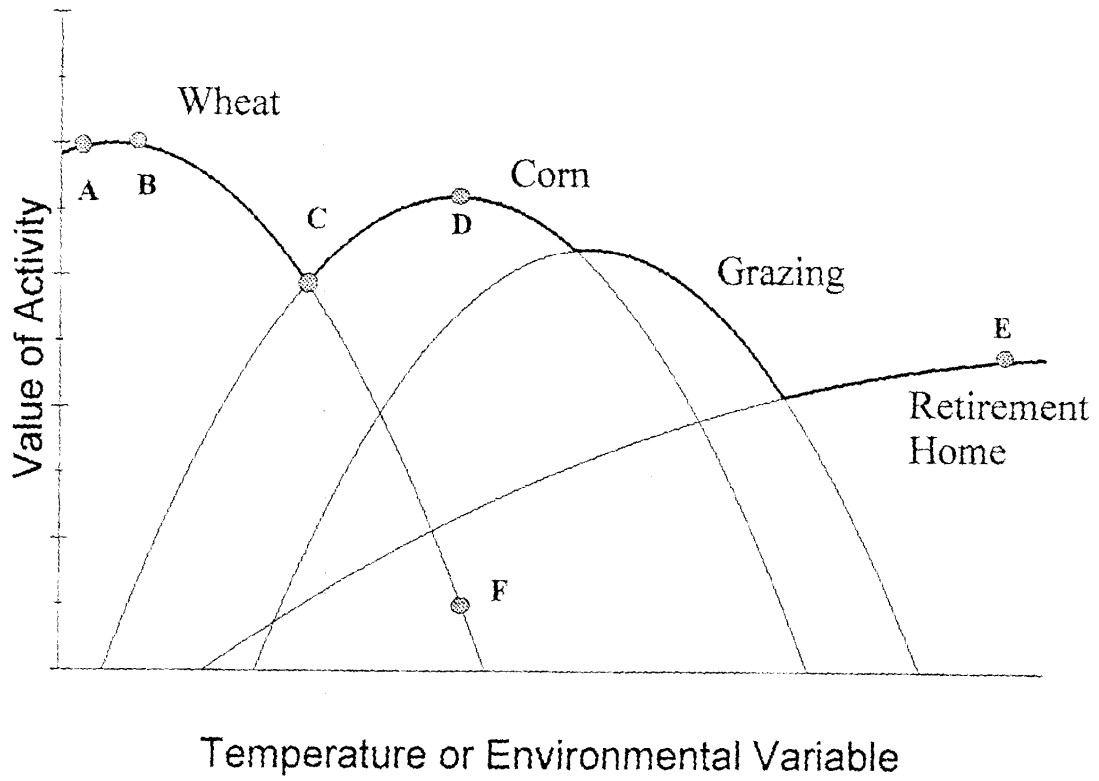


Figure 2
FARM VALUES FROM CURRENT CLIMATE WITH PERCENT OF COUNTY IN CROPLAND

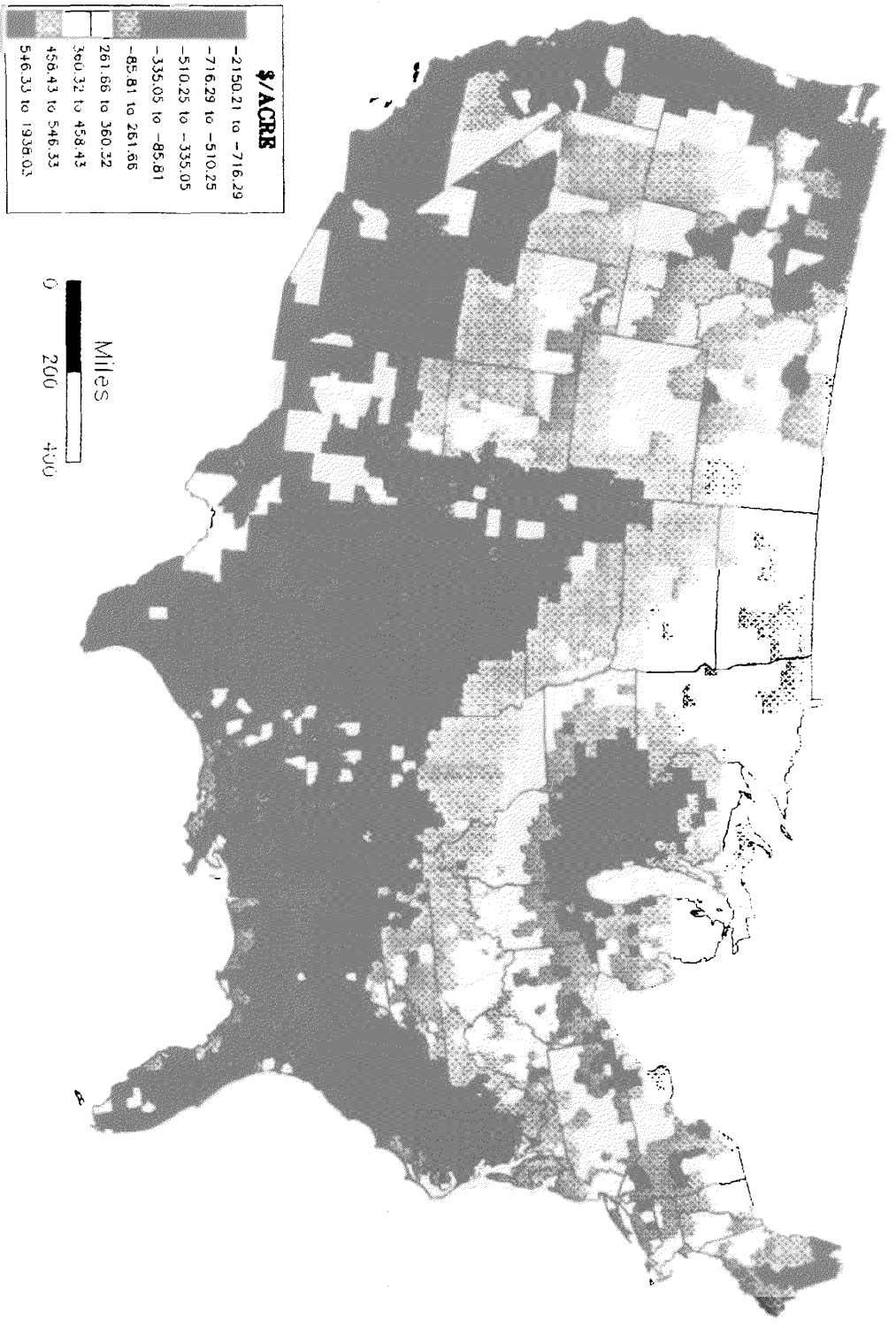


Figure 3
FARM VALUES FROM CURRENT CLIMATE WITH CROP REVENUE PER COUNTY

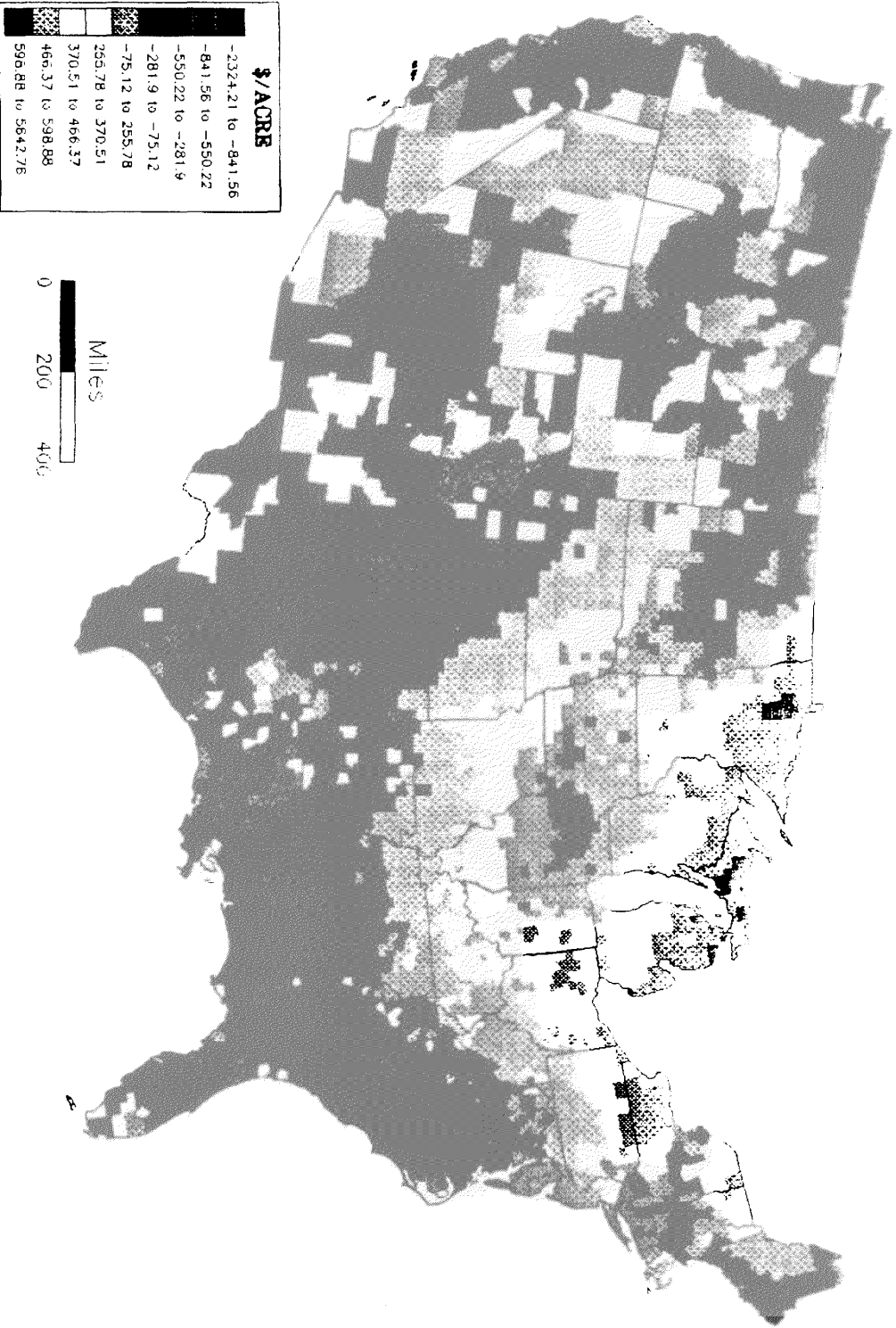


Figure 4
CHANGE IN VALUE DUE TO 5% UNIFORM INCREASE WEIGHTED BY % LAND IN CROPS

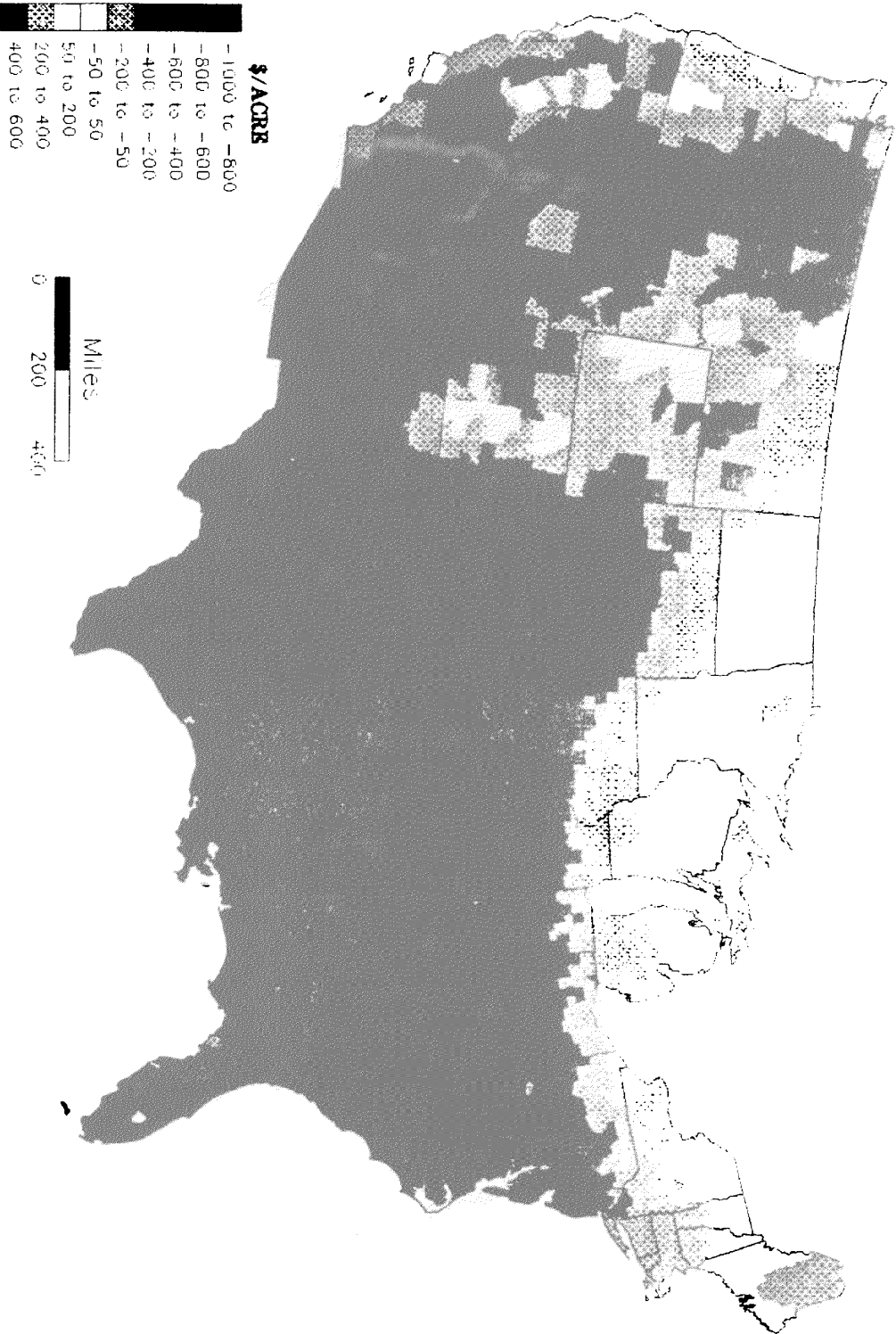


Figure 5
CHANGE IN VALUE DUE TO 5% UNIFORM INCREASE WEIGHTED BY CROP REVENUE/COUNTY

