



# An Economic Analysis on the Transition to Dryland Production in Deficit-Irrigated Cropping Systems of the Texas High Plains

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The objective of this analysis was to integrate hydrologic, agronomic, and economic methods to evaluate various management strategies by changing crop acreage to better manage the declining resources of the Ogallala aquifer. A non-linear optimization model was used to estimate the optimal water use, crop mix, crop yield, and net returns over a 50 year period under dryland and deficit irrigation scenarios in the Texas High Plains. Results indicated that growers could maintain profitability by switching from fully irrigated center pivots to irrigating  $\frac{1}{2}$  and  $\frac{1}{4}$  pivots.

**Keywords:** ogallala, optimization, profitability, deficit irrigation, dryland

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## INTRODUCTION

The Ogallala aquifer is the largest unconfined aquifer in the United States, covering over 450,000 km<sup>2</sup> in parts of eight states: Nebraska, Texas, South Dakota, Wyoming, New Mexico, Kansas, Colorado, and Oklahoma (McGuire, 2017; TWDB, 2020). The Ogallala has an average 29 m of saturated thickness; however, excessive pumping has caused a depletion across much of the Great Plains. In Texas alone, the saturated thickness has declined >15 m since predevelopment in the 1950's (TWDB, 2020). The ability to irrigate crops has turned this semi-arid desert into one of the most productive agricultural regions in the world. The aquifer recharges through rainfall; however, in most counties recharge is 0–76 mm/year, far less than withdrawal (Deeds and Hamlin, 2015). As the saturated thickness of the aquifer has declined, this region has seen a shift in crop mix from more water-thirsty agricultural crops like corn to cotton (Colaizzi et al., 2009). Producers in some areas of the Texas High Plains are not able to fully irrigate entire pivots and are looking for deficit irrigation strategies to maximize water availability (Schneekloth et al., 2004; Rudnick et al., 2019). Developing new management practices that can maintain producer income will be required to adapt to reduced well capacities (Xue et al., 2017). A complete transition to dryland agriculture could potentially result in a loss of \$1.6 billion annually (Yates et al., 2010; Lascano et al., 2020).

Optimization methods have been used to model the interaction between groundwater pumping, profitable cropping systems, and government policy. In the Southern Ogallala Aquifer region, Feng (1992) created a profit-maximizing dynamic programming model to determine optimal irrigated water use and cropping patterns from water availability. The objective function maximized net present value over a series of crops, cropping practices and irrigation technologies. Wheeler et al. (2008) analyzed the impact of short-term water rights buyout policies where land was removed from agricultural production, but allowed irrigation to resume after an allocated amount of time,

and long-term buyout policy that switched to permanent dryland production. This study used the dynamic optimization model developed by Feng (1992) for nine counties in the Texas Southern High Plains. Yield estimates from the Environmental Policy Integrated Climate Model (EPIC) were used to build production functions (Williams et al., 1984). The results indicated that the long-term water buyout policy was more effective in reducing groundwater pumping, but the negative impact to the economy was greater than the short-term water buyout policy. Johnson et al. (2009) estimated the effect of a water pumping fee and a 50/50 quota policy that required 50% of the saturated thickness of the Ogallala aquifer to remain in 50 years. An IMPLAN input-output model was used to estimate the direct, indirect, and induced effects of each of the policy scenarios to determine the impact on regional economies (IMPLAN Group LLC, 2020). The 50/50 plan restricted water pumping more than the pumping fee, which resulted in a slower transition into dryland production.

The objective of this analysis was to successfully integrate hydrologic, agronomic, and economic methods to evaluate various management strategies by changing crop choices to better manage the dwindling resources of the Ogallala aquifer. This study builds on the models above by simulating aquifer drawdown to analyze the impact of declining water availability on the potential long-term economics of food and fiber production in the Texas High Plains using a non-linear dynamic optimization model by maximizing producer net returns projected over a 50 year time horizon. Management scenarios were created to analyze the profitability of reducing

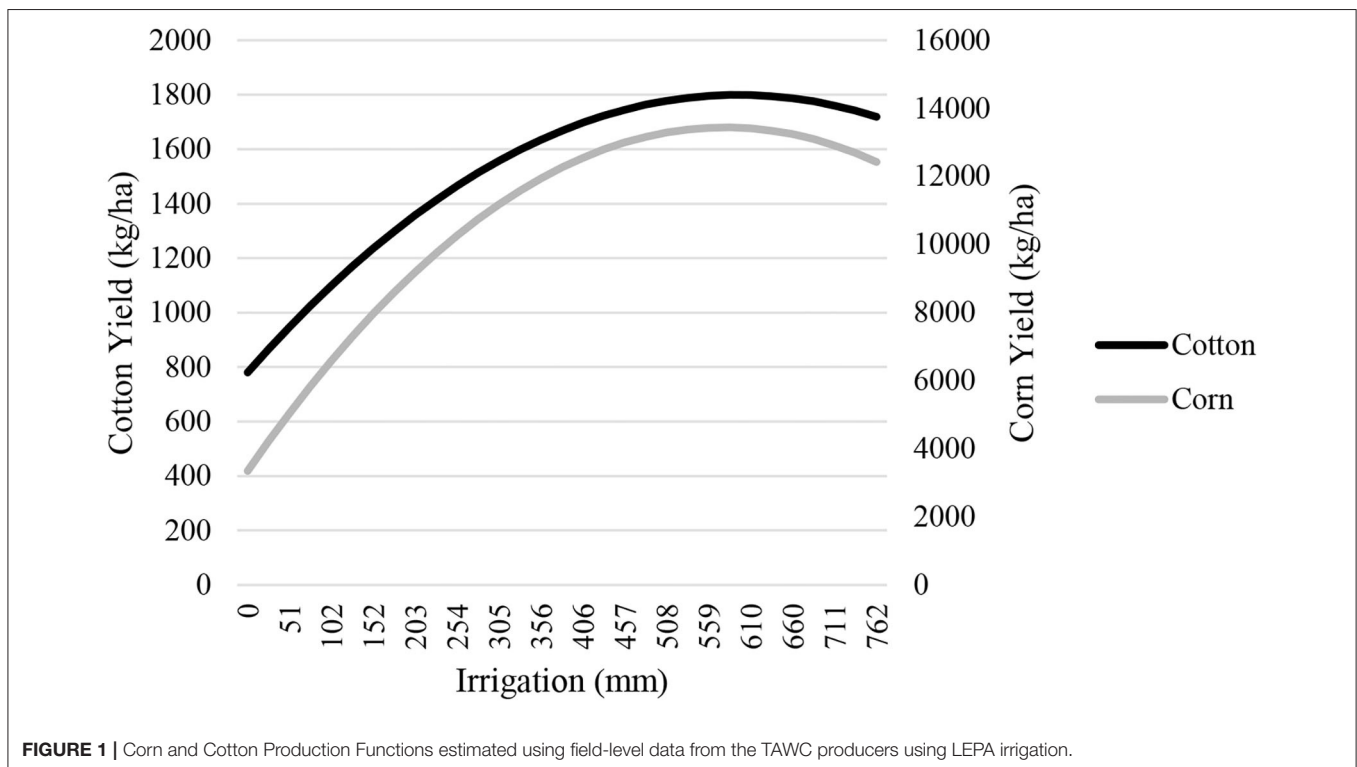
crop acreage and changing crop mix as a method to maximize water availability in a limited irrigation environment.

## MATERIALS AND METHODS

Due to the variability of saturated thickness and well capacity across the Southern Ogallala region, an aggregated model was estimated for one county in the Texas High Plains and a field-level analysis was conducted for a representative farm within that county to provide insight on producer profitability at a micro scale for various cropping systems. The model output provided an estimation of the optimal water use, crop mix, crop yield, and net returns over time.

### Case Study Location and Data

The Texas Alliance for Water Conservation (TAWC) is a demonstration project located in the Texas High Plains that has worked with 36 growers covering over 2,400 ha<sup>-1</sup> that began in 2005. Field sites were initially located in Hale and Floyd counties, but have expanded over time to Castro, Crosby, Deaf Smith, Lamb, Lubbock, Parmer, and Swisher Counties. As a consortium of growers, industry, universities, and government agencies, the TAWC seeks to maintain individual farm profitability and improve water use efficiency to extend the life of the Ogallala aquifer and the economic viability of rural communities. Detailed records of inputs and production practices are collected and analyzed across all TAWC field sites (Weinheimer et al., 2013; TAWC, 2020).



Hale County, Texas was selected as a case study location for this analysis because of the accessibility of grower data. This particular area represents an area transitioning from full irrigation to deficit and dryland production. From pre-development, saturated thickness in Hale County has declined more than 45 m (McGuire, 2017). While there are a few pockets of deeper saturated thickness, the northwestern part of the county has less than 15 years of useable life and the southeastern part of the county has already reached a saturated thickness of 9 m (Barbato, 2008). As saturated thickness declines, transitions into integrated and dryland systems are expected.

**TABLE 1** | Saturated thickness, net returns, and the percentage of irrigated acres in Hale County, Texas projected over a 50 year time period.

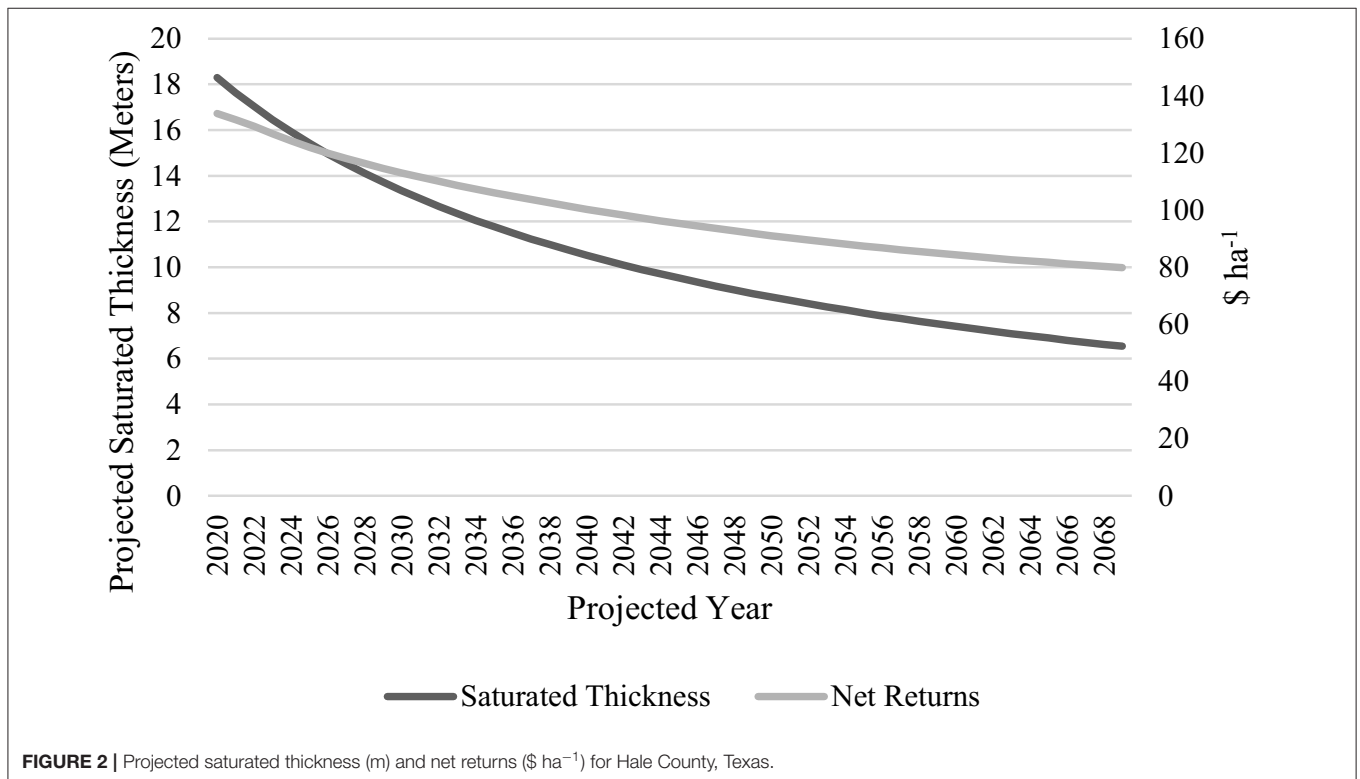
Period	Saturated thickness (m)	Net Returns (\$ ha <sup>-1</sup> )	Irrigated acres (%)
2020	18	134	73
2024	16	122	50
2029	14	116	36
2034	12	110	27
2039	11	104	21
2044	10	98	17
2049	9	92	14
2054	8	85	12
2059	8	85	10
2064	7	79	9
2069	7	79	8

Hydrologic, agronomic, and economic data was used to perform this analysis. Aquifer characteristics such as saturated thickness, recharge, specific yield, depth to water, and well yield were retrieved from the High Plains Water District (2018). Irrigated and dryland crop acreage and average dryland yield were based on 10 year averages from (United States Department of Agriculture, 2019) (2008–2018). To determine profitability across each crop, economic data regarding price and variable expenses were averaged over the 2014–2018 Texas A&M AgriLife Extension crop and livestock budgets (Smith, 2014–2018). Prices used were \$0.16 kg<sup>-1</sup>, \$1.43 kg<sup>-1</sup>, \$0.15 kg<sup>-1</sup>, and \$0.23 kg<sup>-1</sup> for corn, cotton, sorghum, and wheat, respectively.

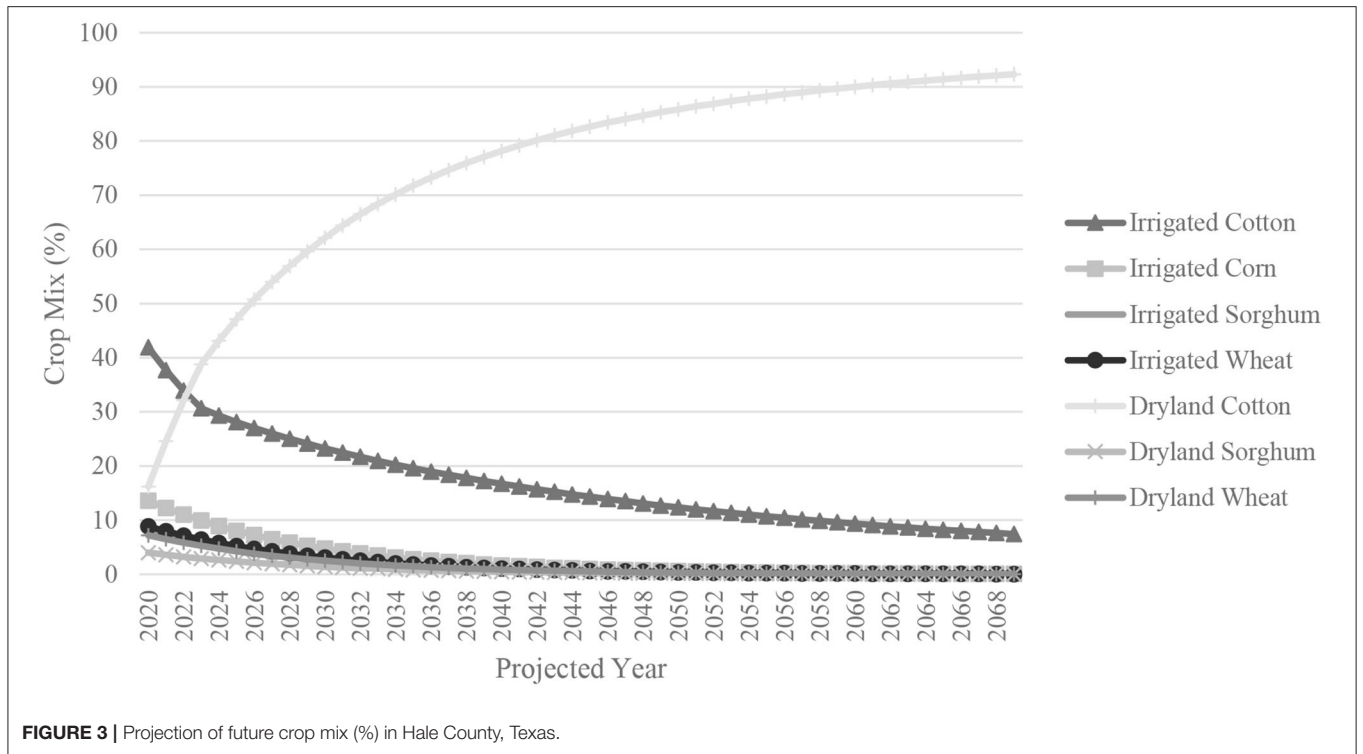
Quadratic production functions were created using a multiple regression analysis to estimate the yield-to-water relationship for each crop with TAWC grower data on yield and water applied. The production functions for corn and cotton are shown in **Figure 1**. Maximum corn yield occurs at 13,440 kg ha<sup>-1</sup> and 1,780 kg ha<sup>-1</sup> for cotton with 580 mm of irrigation.

### Economic Model

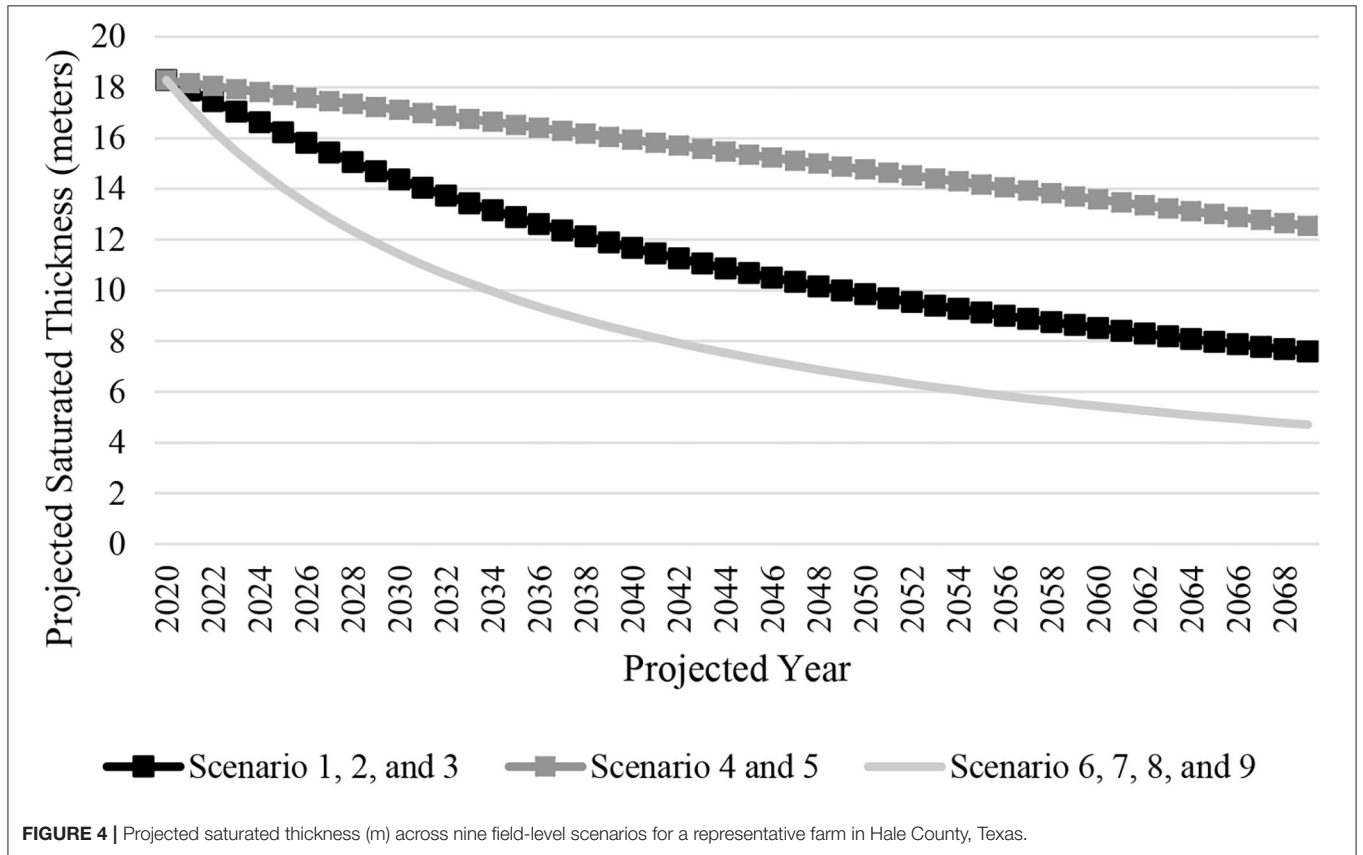
The production functions estimated from the TAWC producer data were input into a non-linear dynamic optimization model using the General Algebraic Modeling System (GAMS Development Corporation, 2019). The non-linear groundwater optimization model used in this analysis has an objective function to maximize producer profit over a 50 year projected period. The optimization routine runs iterations of possible yield and water combinations from the crop production functions and hydrologic characteristics of the aquifer to derive a solution that



**FIGURE 2** | Projected saturated thickness (m) and net returns (\$ ha<sup>-1</sup>) for Hale County, Texas.



**FIGURE 3** | Projection of future crop mix (%) in Hale County, Texas.



**FIGURE 4** | Projected saturated thickness (m) across nine field-level scenarios for a representative farm in Hale County, Texas.

satisfies the objective function to:

$$\text{Maximize NPV} = \sum_{t=1}^n NR_t(1+r)^{-t}, \quad (1)$$

where NPV is net present value,  $NR_t$  is net revenue in time  $t$ , and  $r$  is the social discount rate. Net revenue is defined as:

$$NR_t = \sum_i \sum_k \Theta_{ikt} \{P_i Y_{ikt} [WA_{ikt}, WP_{ikt}] - C_{ikt}(WP_{ikt})\}, \quad (2)$$

where  $i$  represents the crops grown,  $k$  represents irrigation technology,  $\Theta$  is the percentage of crop  $i$  produced using irrigation technology,  $k$ , during time  $t$ ,  $P_i$  is the price of crop  $i$ ,  $WA_{ikt}$  and  $WP_{ikt}$  are the water applied and water pumped from the aquifer,  $Y_{ikt}$  is crop yield, and  $C_{ikt}$  are the costs of production. The model equations and constraints are:

$$ST_{t+1} = ST_t - [(\sum_i \sum_k \Theta_{ikt} * WP_{ikt}) - R] PIA / SY, \quad (3)$$

$$X_{t+1} = X_t + [(\sum_i \sum_k \Theta_{ikt} * WP_{ikt}) - R] PIA / SY, \quad (4)$$

$$WT_t = \sum_i \sum_k \Theta_{ikt} * WP_{ikt}, \quad (5)$$

$$WT_t \leq GPC_t, \quad (6)$$

$$GPC_t = \left(\frac{ST_t}{IST}\right)^2 * (IGPM * \frac{WY}{AW}), \quad (7)$$

$$PC_{ikt} = \{[EF(X_t + 6.895 * PSI) EP] / EFF\} * WP_{ikt}, \quad (8)$$

$$C_{ikt} = VPC_{ik} + PC_{ikt} + HC_{ikt} + MC_k + DP_k + LC_k, \quad (9)$$

$$\sum_i \sum_k \Theta_{ikt} \leq 1 \text{ for all } t, \quad (10)$$

$$\Omega_{ikt} \geq 0 \quad (11)$$

Equations (3) and (4) are the two state variables, saturated thickness,  $ST$ , and pumping lift,  $X$ , where  $R$  is the recharge rate,  $PIA$  is the percentage of irrigated area expressed as the initial number of irrigated area in the county divided by the area of the county above the aquifer, and  $SY$  is the specific yield of the aquifer. Equation (5) calculates the total water pumped for each crop and Equation (6) restricts the total amount of water pumped to be less than or equal to the pumping capacity. Equation (7) represents the gross pumping capacity, where  $IST$  is the initial saturated thickness,  $WY$  is the well yield,  $AW$  is the area per well, and  $IGPM$  is the amount of water pumped per gallon per minute. Equation (8) calculates the pumping cost as a function of the energy use factor for electricity,  $EF$ , lift,  $X$ , price of electricity,  $EP$ , pump efficiency,  $EFF$ , and the water pumped,  $WP$ . The height of a column of water that will exert 1 psi = 6.895 kPa. Equation (9) is the total cost calculation, where  $VPC$  is the variable cost of production,  $PC$  is the pumping cost estimated in Equation (8),  $HC$  is the harvest cost,  $MC$  is the irrigation system maintenance cost,  $DP$  represents the depreciation of the irrigation system, and  $LC$  is the labor cost. Equation (10) restricts the crop area percentage over all crops over all irrigation systems to be <1 or = to 1, and Equation (11) is a non-negativity constraint, where  $\Omega$  represents all decision variables. A 3% discount rate was used in the NPV analysis.

**TABLE 2 |** Field-level scenarios ranked by net present value of net returns for a representative farm in Hale County, Texas.

Scenario	Description (ha <sup>-1</sup> )	NPV of net returns (\$ ha <sup>-1</sup> )
1	Irrigated Corn-12 ha <sup>-1</sup> , Irrigated Cotton-12 ha <sup>-1</sup> , Dryland Cotton-40 ha <sup>-1</sup>	3,566
2	Irrigated Cotton-24 ha <sup>-1</sup> , Dryland Cotton-40 ha <sup>-1</sup>	3,563
3	Irrigated Corn-24 ha <sup>-1</sup> , Dryland Cotton-40 ha <sup>-1</sup>	3,539
4	Irrigated Corn-12 ha <sup>-1</sup> , Dryland Cotton- 53 ha <sup>-1</sup>	3,254
5	Irrigated Cotton-12 ha <sup>-1</sup> , Dryland Cotton-53 ha <sup>-1</sup>	3,153
6	Irrigated Cotton-49 ha <sup>-1</sup> Dryland Cotton- 16 ha <sup>-1</sup>	2,946
7	Irrigated Corn-12 ha <sup>-1</sup> , Irrigated Cotton-36 ha <sup>-1</sup> , Dryland Cotton-16 ha <sup>-1</sup>	2,881
8	Irrigated Corn-24 ha <sup>-1</sup> , Irrigated Cotton-24 ha <sup>-1</sup> , Dryland Cotton-16 ha <sup>-1</sup>	2,795
9	Irrigated Corn-49 ha <sup>-1</sup> , Dryland Cotton-16 ha <sup>-1</sup>	2,528

## RESULTS

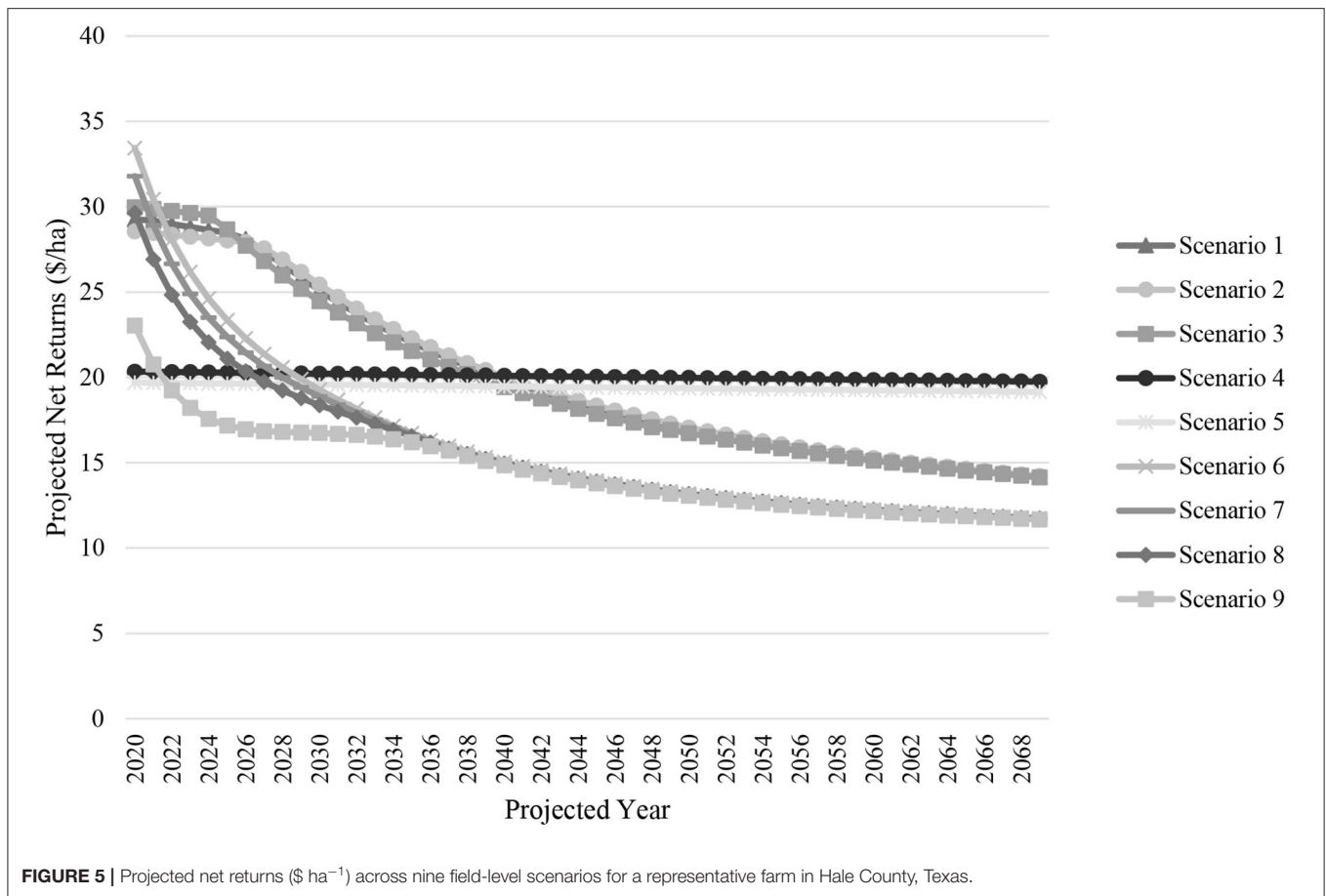
### County Baseline

Results from the aggregated county baseline model for saturated thickness, net returns, and the percentage of irrigated crops are shown in **Table 1** and **Figures 2–4**. Initial starting saturated thickness began at 18 m and declined to 7 m by the end of year 50 (2069) (**Table 1**, **Figure 2**). Across the 50 year time period, crop yields averaged 1,630 kg ha<sup>-1</sup> for cotton, 12,690 kg ha<sup>-1</sup> for corn, 7,790 kg ha<sup>-1</sup> for sorghum, and 2,130 kg ha<sup>-1</sup> for wheat. As saturated thickness declines, net returns across the time period declined from \$134 ha<sup>-1</sup> to \$79 ha<sup>-1</sup> with an NPV of \$2,799 ha<sup>-1</sup> over the entire projected period (**Table 1**, **Figure 2**). **Figure 3** depicts the optimal crop mix for the county. Initially, the crop mix was 42% irrigated cotton, 14% irrigated corn, 8% irrigated sorghum and wheat, 16% dryland cotton, 4% dryland sorghum, and 7% dryland wheat. By the end of the projected period the crop mix was 92% dryland cotton, 7% irrigated cotton, and the other 2% was a negligible combination of dryland sorghum and wheat.

### Field-Level Scenarios

A scenario analysis was performed to understand the long-term impact of various cropping systems at the field level. Based on the results from the county analysis, nine scenarios were created from combinations of corn and cotton production. Wheat and sorghum were not included in the field-level analysis due to a lack of impact at the county level. These scenarios are summarized in **Table 2**. These scenarios focus on one center pivot covering 49 ha<sup>-1</sup> and 16 ha<sup>-1</sup> of dryland cotton corners for a total of 65 ha<sup>-1</sup>. The scenarios were ranked by greatest amount of net present value.

While corn is a high-value agricultural crop, corn yields do not meet maximum yield potential in the Texas High Plains due to the semi-arid climate of the region, resulting in comparable profitability with cotton. The increased pumping cost for corn resulted in higher variable expenses and thus, lower profit. Since



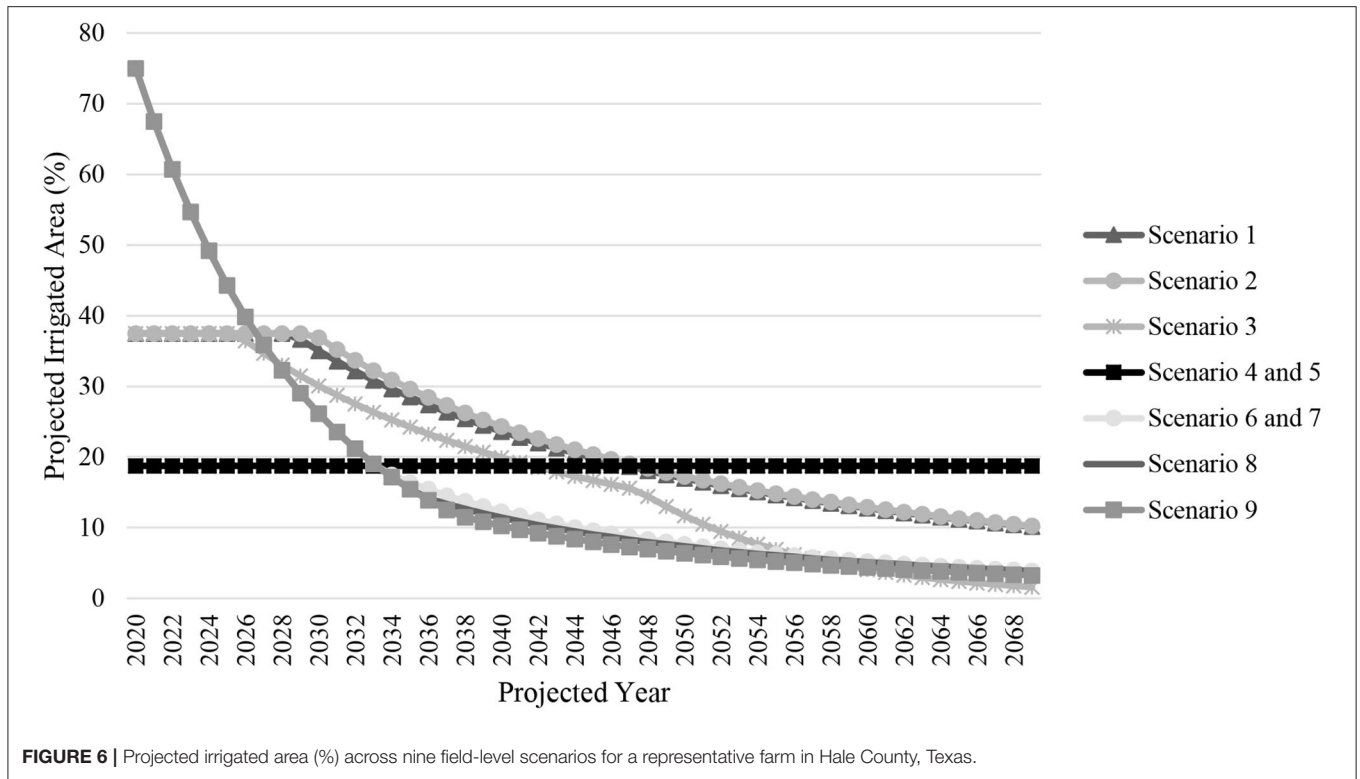
**FIGURE 5** | Projected net returns (\$ ha<sup>-1</sup>) across nine field-level scenarios for a representative farm in Hale County, Texas.

the profitability of corn and cotton were similar, the scenario results were primarily influenced by the amount of irrigated area, resulting in three distinct outcomes.

Scenarios 1, 2, and 3 were systems with 24 ha<sup>-1</sup> of irrigation and 40 ha<sup>-1</sup> of dryland cotton. These scenarios produced comparable results. Scenario 1 consisted of a ¼ pivot of irrigated corn (12 ha<sup>-1</sup>), a ¼ pivot of irrigated cotton (12 ha<sup>-1</sup>), and 40 ha<sup>-1</sup> of dryland cotton. Average corn yield across was 12,740 kg ha<sup>-1</sup> for corn and 1,650 kg ha<sup>-1</sup> for irrigated cotton. The yield for dryland cotton was 540 kg ha<sup>-1</sup> across all scenarios. This scenario generated the highest profitability over time with a net present value of net returns of \$3,566 ha<sup>-1</sup>. Scenario 2 consisted of 24 ha<sup>-1</sup> of irrigated cotton and 40 ha<sup>-1</sup> of dryland cotton, which generated a NPV of \$3,563/ha<sup>-1</sup>. Irrigated cotton yields averaged 1,650 kg ha<sup>-1</sup> with 356 mm of water applied. Scenario 3 consisted of 24 ha<sup>-1</sup> of corn and 40 ha<sup>-1</sup> of dryland cotton and generated a NPV of \$3,539 ha<sup>-1</sup> (Table 2). Irrigated corn yields averaged 12,740 kg ha<sup>-1</sup> with 432 mm of water applied. The nine crop scenarios resulted in three distinct water use patterns (Figure 4). Saturated thickness for Scenarios 1, 2, and 3 began at 18 m and declined to 8 m by year 50. Projected net returns over time are shown in Table 4 and Figure 5. As water availability declines, the amount of water that was able to be allocated to crops was also reduced, therefore resulting in lower net returns. In Scenario 1, 2, and 3, net returns were \$179 ha<sup>-1</sup>, \$174 ha<sup>-1</sup>, and \$183 ha<sup>-1</sup>

respectively, in the beginning of the time period. By the end of the period, net returns were \$87 ha<sup>-1</sup> for Scenarios 1 and 2 and \$86 ha<sup>-1</sup> for Scenario 3 (Table 4). The crop mix for Scenarios 1, 2, and 3 began with 38% irrigated ha<sup>-1</sup> and 62% dryland ha<sup>-1</sup>. Under these scenarios, irrigation remains constant for the first 15 years in Scenarios 1 and 2. By year 30 of the projection, the ability to irrigate sufficiently drops down to a ¼ of a pivot before essentially becoming a completely dryland farm. Scenario 3 maintains irrigation for the first 10 years into the forecast, and by year 25, can realistically irrigate ¼ of a pivot before becoming dryland (Table 5, Figure 6).

Scenarios 4 and 5 were systems with 12 ha<sup>-1</sup> of irrigation and 53 ha<sup>-1</sup> of dryland cotton. These scenarios produced similar results. Scenario 4 represented a ¼ pivot of irrigated area consisting of 12 ha<sup>-1</sup> planted to irrigated corn and 53 ha<sup>-1</sup> planted to dryland cotton. In this scenario, corn yields averaged 13,250 kg ha<sup>-1</sup> with 508 mm of water applied and generated a NPV of \$3,254 ha<sup>-1</sup>. Scenario 5 represented a ¼ pivot (12 ha<sup>-1</sup>) planted to irrigated cotton and 53 ha planted to dryland cotton. Irrigated cotton yields averaged 1,680 kg ha<sup>-1</sup> using 406 mm of water applied, which generated a NPV of NR of \$3,153 ha<sup>-1</sup> (Table 2). Since both scenarios used the least amount of irrigation, they had the least impact on saturated thickness, which only declined 5 m throughout the time period (Table 3, Figure 4). Net returns for both scenarios were also consistent over time



**TABLE 3 |** Projected saturated thickness (m) across nine field-level scenarios for a representative farm in Hale County, Texas.

Period	1	2	3	4	5	6	7	8	9
2020	18	18	18	18	18	18	18	18	18
2024	16	16	16	18	18	14	14	14	14
2029	14	14	14	17	17	11	11	11	11
2034	13	13	13	17	17	10	10	10	10
2039	12	12	12	16	16	8	8	8	8
2044	11	11	11	15	15	7	7	7	7
2049	10	10	10	15	15	7	7	7	7
2054	9	9	9	14	14	6	6	6	6
2059	9	9	9	14	14	5	5	5	5
2064	8	8	8	13	13	5	5	5	5
2069	8	8	8	13	13	5	5	5	5

**TABLE 4 |** Projected net returns (\$ ha<sup>-1</sup>) across nine field-level scenarios for a representative farm in Hale County, Texas.

Period	1	2	3	4	5	6	7	8	9
2020	179	174	183	124	120	204	194	181	141
2024	175	172	180	124	120	150	143	135	107
2029	158	160	154	124	120	121	119	115	102
2034	138	139	135	123	119	105	104	103	100
2039	124	125	121	123	119	94	93	93	92
2044	113	114	111	122	119	86	86	86	85
2049	105	106	103	122	118	81	81	81	81
2054	99	99	98	122	118	78	78	78	77
2059	94	94	93	121	118	75	75	75	75
2064	90	90	89	121	117	73	73	73	73
2069	87	87	86	121	117	72	72	72	71

with an annual average income of \$122-\$118 ha<sup>-1</sup>, respectively (Table 4, Figure 5). Irrigated area did change due to the fact that irrigation could continue in the long term (Table 5, Figure 6).

Scenarios 6–9 were systems with the greatest amount of irrigation with 49 ha<sup>-1</sup> and 16 ha<sup>-1</sup> of dryland cotton. Scenario 6 represented a whole pivot dedicated to irrigated cotton (49 ha<sup>-1</sup>) plus 16 ha<sup>-1</sup> dryland corners. This scenario generated average cotton yields of 1,610 kg ha<sup>-1</sup> using 330 mm of water applied. Scenario 7 was a ¼ pivot (12 ha<sup>-1</sup>) of corn, ¾ pivot (36 ha<sup>-1</sup>) of irrigated cotton and dryland corners (16 ha<sup>-1</sup>) planted to cotton. Irrigated corn averaged 12,560 kg ha<sup>-1</sup> with 406 mm of water applied and 1,600 kg ha<sup>-1</sup> of cotton irrigated with 330 mm.

Scenario 8 was ½ pivot (24 ha<sup>-1</sup>) of corn, ½ pivot (24 ha<sup>-1</sup>) of cotton, plus dryland cotton corners (16 ha<sup>-1</sup>). Irrigated corn in this scenario averaged 12,430 kg ha<sup>-1</sup> with 406 mm of irrigation and irrigated cotton averaged 1,580 kg ha<sup>-1</sup> with 330 mm. The least revenue-generating scenario assessed was Scenario 9, which was a whole pivot of corn (49 ha<sup>-1</sup>) and 16 ha<sup>-1</sup> of dryland corners. Corn yields averaged 12,240 kg ha<sup>-1</sup> with 381 mm of water. Scenarios 6–9 generated \$2,946, \$2,881, \$2,795, and \$2,528 ha<sup>-1</sup> of NPV, respectively (Table 2). These scenarios had the most irrigated acres, which therefore depleted saturated thickness with a total change of 14 m. Scenarios 6, 7, and 8 had the greatest amount of water use, which resulted in a decline in saturated

**TABLE 5** | Projected irrigated area (%) across nine field-level scenarios for a representative farm in Hale County, Texas.

Period	1	2	3	4	5	6	7	8	9
2020	38	38	38	19	19	75	75	75	75
2024	38	38	38	19	19	49	49	49	49
2029	37	38	31	19	19	29	29	29	29
2034	30	31	25	19	19	17	17	17	17
2039	24	25	21	19	19	13	13	12	11
2044	21	21	17	19	19	10	10	9	8
2049	17	18	13	19	19	8	8	8	7
2054	15	15	8	19	19	7	7	6	5
2059	13	13	5	19	19	5	5	5	5
2064	11	12	3	19	19	5	5	5	4
2069	10	10	2	19	19	4	4	4	3

thickness from 18 m to 5 m (Table 3, Figure 4). Net returns began at \$204, \$194, \$181, and \$141 ha<sup>-1</sup> and declined to \$71 ha<sup>-1</sup>. Each of these scenarios started as a fully irrigated pivot (Table 4, Figure 5). Within five years of heavy irrigation, a switch to half a pivot would have to occur. Between 2025 and 2035, there may be enough water availability to irrigate a quarter of a pivot before a shift to dryland (Table 5, Figure 6).

## DISCUSSION AND CONCLUSION

In a semi-arid environment such as the Texas High Plains, a dryland system of dryland row crops, forages and native grassland will be the most profitable as growers transition out of irrigated row crop production. This analysis produced future time paths to understand the potential food and fiber production possibilities as this region begins to transition to dryland production at both an aggregated county scale and at the micro-level. This research can help to determine the best management practices and technologies to maximize producer profit and increase water use efficiency in deficit irrigation and dryland scenarios. By identifying the comparative advantage of land use, this research will provide a strong foundation for the development of new evidence-based strategies to ensure understanding of areas that are most at risk due to water decline.

Results from this analysis indicate that growers irrigating entire pivots in Hale county will significantly exacerbate aquifer decline and increase the rate of transition to dryland crop production (Scenarios 6 and 9). Irrigating a full pivot (Scenarios 6–9) resulted in average corn yields of 12,240 kg ha<sup>-1</sup> and 1,610 kg ha<sup>-1</sup> for cotton using 381 mm of irrigation on corn and 330 mm of water applied to cotton. However, a grower could reduce irrigation to a ¼ pivot and maintain deficit irrigation in the long-term (Scenarios 4 and 5). The most profitable crop mixes in the long-term were from irrigating ½ of a pivot. Irrigating half of a pivot (Scenarios 1–3) resulted in an increase of corn yield to 12,740 kg ha<sup>-1</sup> on 432 mm of water applied and cotton yields of 1,650 kg ha<sup>-1</sup> on 356 mm of water applied. Results indicated the greatest crop yields were achieved when irrigated area was reduced to ¼ of a pivot (Scenarios 4–5), with corn

yields averaging 13,240 kg ha<sup>-1</sup> irrigated with 508 mm of water applied and 1,680 kg ha<sup>-1</sup> of cotton irrigated with 432 mm of water applied. Concentrating irrigation on a smaller area resulted in higher irrigated profitability, but when dryland ha<sup>-1</sup> were included as a system, it was actually the least profitable.

Understanding how drought impacts our water use and ability to grow crops/livestock will provide better predictions of human adaptation and inform new social and economic strategies to cope in our dryland future. Irrigated land will need to be monitored closely, but there are options available to extend irrigation crop production and maintain profitability in a dryland environment.

There are some limitations to this study which could impact the results. For example, this analysis assumed no government intervention and did not consider groundwater policies, which could change the rates of water withdrawal. Further, this study did not consider climate impacts or rainfall variability. In this analysis, dryland crop yields were held static over time. Naturally, dryland crop yield will change with regard to the amount of rainfall received and improved technological advancements in genetic productivity over time. Increased rainfall will also impact aquifer recharge, which was assumed to be zero. The crop production functions used in this analysis determine crop yield and thus, profitability. While these functions were estimated using actual producer data from 2005–2018, it might not be reflective of production across the entire county. While it was impossible to analyze all potential crop scenarios, wheat and sorghum rotations can be profitable enterprises. With regard to wheat, we did not analyze the impact of grazing, which should positively impact profitability and can be a good option for growers. Other alternative crops exist, which we did not attempt to analyze here.

## DATA AVAILABILITY STATEMENT

The datasets generated for this study are available on request to the corresponding author.

## AUTHOR CONTRIBUTIONS

DM-M was responsible for running the economic model and writing the manuscript. RM assisted with writing the manuscript. PJ provided the production functions. RW helped to conceptualize the paper and assist in writing the manuscript.

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**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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