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Climate Change Impacts on Investment in Crop Sowing Machinery

Ross Kingwell^{1,2} and Imma Farré¹

¹ Department of Agriculture and Food, Western Australia

² School of Agricultural and Resource Economics, University of Western Australia

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Abstract

Down-scaled global circulation modelling is combined with wheat growth simulation modelling to generate yield responses to times of sowing under current and projected climatic conditions for several locations in the grainbelt of Western Australia. A model for investment in crop sowing machinery draws on these simulated yield relationships at each location and is used to determine a farmer's optimal investment in crop sowing work rate capacity under current and projected climate regimes. The key finding is that at most locations the projected change in climate has marked impacts on profit distributions from grain production, yet mostly modest changes in the farmer's investment in machinery work rate form part of the profit-maximising response to climate change at each location. There is also a divergence in machinery investment response between high versus low rainfall locations, with increases and decreases in work rates respectively being forecast. However, as illustrated for a few locations, the changes in investment in work rate within a broadly similar rainfall region are not uniform; principally due to climate change differently affecting the pattern of yield response to time of sowing at each location.

Keywords: climate change, farm machinery, farm management, machinery investment

Introduction¹

There is a burgeoning literature on climate change and its possible impacts on agriculture. For Australian agriculture, several studies have reported or discussed possible impacts of climate change on agricultural industries and regions (Pittock 2003, Kingwell 2006, Heyhoe *et al.* 2007). Some studies (e.g. Pittock 2003; Howden and Jones 2001, 2004; Kokic *et al.* 2005) have focused on wheat production, a principal enterprise of broadacre farming in Australia, and these studies often point to a range of spatially diverse projected impacts of climate change on wheat yields.

Howden and Jones (2004) report how the value of Australia's wheat production could be affected by a projected climate regime towards 2070. Their results suggest that the projected climate regime towards 2070 poses a significant downside risk for the Australian wheat industry, although adaptation strategies could substantially reduce this risk. The findings of Howden and Jones (2001, 2004) show marked spatial variation across sites in Australia, with wheat production in south-western Australia deleteriously being affected while southern Queensland and higher rainfall regions of New South Wales benefit.

Although these studies report possible impacts of climate change on wheat yields and possible adaptation strategies to lessen adverse impacts on crop yields (Heyhoe *et al.* 2007, Blackwell *et al.* 2007), an issue yet to be discussed is: what is the likely impact of climate change on farmers' investment in crop machinery for wheat production? The impact of climate change on farmers' capital investment decisions regarding crop sowing capacity has so far been over-looked; so this paper focuses on this issue.

¹ This paper has benefited from some earlier investigations by Reid Seaby, an honours student in the School of Agricultural and Resource Economics at the University of Western Australia and from discussions with Doug Abrecht. Amir Abadi wrote the Visual Basic code to automate the regression analyses.

The paper is structured as follows. The next section examines briefly the nature and role of crop sowing machinery in wheat production in Australia. Then a conceptual model of investment in crop sowing machinery is outlined. Data and relationships derived from climate and wheat production simulation models are used to apply the conceptual model to wheat growing locations in south-western Australia in order to illustrate how climate change may affect investments in crop sowing machinery. Lastly a set of conclusions is provided.

Farm Machinery for Wheat Production in Australia

Broadscale mechanised operations underpin wheat production in Australia, with wheat being by far the main crop sown on Australian broadacre farms. Each year, on average, around 13 million hectares are sown to wheat and average wheat production is around 23 million tonnes (ABARE 2006). Most wheat grown in Australia is a winter temperate crop (Foster 2000). Sowing usually occurs anywhere from late April to late June, depending on seasonal conditions. Typically crops are sown after opening rains using reduced tillage methods and herbicide technology. Large tractors pulling wide cultivator bars, airseeders and spray rigs are commonplace. Often the sowing of all crops is completed within a three week period and typically involves a single pass operation.

The need for speedy crop establishment arises firstly from the desire to provide plants with favourable growing conditions for as long as possible and secondly, the sheer size of some farmer's cropping programs dictates their need to sow quickly in order to then switch to other tasks such as post-emergent spraying of weeds. Crops sown late tend to have worsening light and temperature conditions that limit their early vigour and reduce their potential yield (Anderson and Garlinge 2000). By illustration, Shackley (2000) reported time of sowing trials for wheat grown at several sites in Western Australia over a 10 year period and found that the average yield penalty for delays to sowing ranged from 17 to 35 kg/ha/day across the sites. In some years at some sites the yield penalty was as high as 50 kg/ha/day which, at a farm-gate wheat price of \$240 per tonne, translates to foregone revenue of \$12/ha/day. With farmers' large investments in crop sowing machinery being commonplace in broadacre cropping in Australia, the question arises as to how may climate change affect yield and time of sowing relationships and thereby affect farmers' investments in crop sowing machinery?

An Investment Model for Crop Sowing Machinery

Kingwell (1996) provides a conceptual model for investment in crop sowing machinery. His model is extended here to consider the likely consequences of climate change and to allow for variation in the size of a farmer's cropping program. In Kingwell's original model the size of a farmer's cropping program was fixed and wheat yield was assumed to be a non-stochastic linear function of the day of sowing. We alter and extend his model by relaxing several of its assumptions. The linear yield response:

$$Y_x = Y_s - bx \quad (1)$$

where

Y_x is the yield (t/ha) of wheat sown on day x
 Y_s is the yield (t/ha) of wheat at commencement of sowing (day = 0) and
 b is the rate of yield decline per day's delay in sowing (t/ha/day)

is replaced by the more commonly observed quadratic form where, in year t :

$$Y_{xt} = Y_{st} + b_t x - c_t x^2 \quad (2)$$

noting that Y_{st} , b_t and $c_t \geq 0$

The fixed size of the sowing program is replaced by a state-contingent specification. In years that in prospect are favourable (unfavourable), due to already observed weather and price conditions, the size

of the cropping program increases (decreases). These tactical changes in the size of a cropping program are known to be an important feature of profitable farm management in various regions of Australia (Kingwell *et al.* 1993, Kingwell 1994).

The size of the farmer's intended cropping program (in hectares) in year t is S_t such that:

$$S_t = \alpha + \beta p_t Y_{st} \quad \text{with } S_{\min} \leq S \leq S_{\max} \quad (3)$$

S_{\min} and S_{\max} are the minimum and maximum sizes of the cropping program respectively, as determined by issues such as rotational considerations, land suitability for cropping and farm size. Also p_t is the anticipated farm-gate price of wheat (\$/t) in year t as may be provided through a forward contract.

The profit from crop production in year t is:

$$\pi_t = p_t q_t - h_t S_t - m - g \quad (4)$$

where

π_t is the profit (\$) from sowing S_t hectares of wheat in year t

p_t is the farm-gate price of wheat (\$/t) (e.g. a forward price contract in year t)

q_t is the quantity (tonnes) of wheat harvested in year t

h_t is the variable production costs (\$/ha) associated with wheat growing in year t

m are the fixed or overhead costs (\$) incurred when growing S_t hectares of wheat (apart from seeding machinery investment)

g is the cost of investment in seeding machinery (\$)

In equation (4) the size of the cropping program (S_t) and the variable costs of production (h_t) are assumed to depend on the anticipated revenue at the start of sowing. Also q_t is a function of the yield parameters in equation (2) and the work rate (R) of the farm machinery for crop sowing (ha/day). g is assumed to be a linear function of R ; that is, $g = d + eR$, making the marginal cost of work rate constant. In other words the cost of acquiring a further unit of sowing capacity (one ha/day) is identical across the range of sowing gear likely to be considered by the farmer.

In equation (4) q_t can be re-expressed as:

$$q_t = \int_0^{\frac{S_{\min}}{R}} R(Y_{st} + b_t x - c_t x^2) dx \quad \text{for years where } S_t = S_{\min};$$

$$\frac{(\alpha + \beta p_t S_t)}{R} \int_0^R R(Y_{st} + b_t x - c_t x^2) dx \quad \text{for years where } S_t = \alpha + \beta p_t Y_{st} \text{ and}$$

$$\int_0^{\frac{S_{\max}}{R}} R(Y_{st} + b_t x - c_t x^2) dx \quad \text{for years where } S_t = S_{\max}$$

The farmer's decision problem is to select an investment in sowing capacity (i.e. work rate R) that maximises profit from wheat production across a range of production years; noting that the size of the cropping program S_t can vary from S_{\min} to S_{\max} , conditional on a production year's revenue prospects and that similarly, variable costs of production can range from h_{\min} to h_{\max} , dependent on revenue prospects at the commencement of sowing. The probabilities of occurrence across T years for cropping programs, S_{\min} and S_{\max} hectares, are u/T and w/T respectively; each arbitrarily set at one sixth. Hence one third of years are classed as being in prospect high or low revenue years, with crop

areas and variable costs of production accordingly being set at various limits, characteristic of the location, in response to these prospects.

The farmer's expected profit from wheat production across T years, where $T = u + v + w$, is:

$$\begin{aligned} \pi = & \frac{u}{T} \left(\int_1^u \left(p_t \int_0^{\frac{S_{\min}}{R}} R(Y_{st} + b_t x - c_t x^2) dx - d - eR - h_{\min} S_{\min} - m \right) dt \right) + \\ & \frac{w}{T} \left(\int_1^v \left(p_t \int_0^{\frac{S_{\max}}{R}} R(Y_{st} + b_t x - c_t x^2) dx - d - eR - h_{\max} S_{\max} - m \right) dt \right) + \\ & \frac{v}{T} \left(\int_1^v \left(p_t \int_0^{(\alpha + \beta p_t Y_{st})/R} R(Y_{st} + b_t x - c_t x^2) dx - d - eR - (\alpha_h + \beta_h p_t Y_{st})(\alpha_s + \beta_s p_t Y_{st}) - m \right) dt \right) \end{aligned} \quad (5)$$

To find the level of R that maximises profit requires differentiation of equation (5) with respect to work rate and solving for the R_{opt} that causes $\frac{\partial \pi}{\partial R} = 0$. Given the complicated nature of equation (5),

R_{opt} in this study is found through a non-linear programming re-formulation of the decision problem as outlined later.

The next section draws on crop simulation modelling to identify likely changes in Y_{st} , b_t and c_t at different locations and subsequently equation (5) is applied to gauge the possible impact of climate change on optimal investment in crop sowing work rate at those locations. The locations considered are in south-western Australia. This region is used as the study region for several reasons. Firstly, it is a major source of Australia's cereal grain production, supplying almost 40 per cent of the nation's cereal production over the last decade. Secondly, it is a region acknowledged as a climate change hot-spot, with crop production in this region viewed as particularly exposed to the impacts of projected climate change (ACG 2005, Hennessy and Jones 2007), although much uncertainty remains about the timing, size and variation of impacts of the climate change (Reyenga *et al.* 2001, Howden and Hayman 2005, Farré *et al.* 2007, Heyhoe *et al.* 2007). Thirdly, the region is well-served by validated crop simulation models of wheat production that draw on long time series of daily weather data available for many locations. The region also has down-scaled global circulation model datasets for key climate scenarios. Sites within the region selected for this study are shown in Figure 1.

Crop Growth Simulation Model Output

Crop yield estimates for each location in Figure 1 were generated by the APSIM-Wheat model (Meinke *et al.* 1998, Keating *et al.* 2003, Wang *et al.* 2003, Asseng *et al.* 2004, Farré *et al.* 2007). This model simulates a wheat crop's development, yield, water uptake and nitrogen accumulation in response to temperature, radiation, day length, soil water and nitrogen supply. The model is a daily time-step crop growth simulation model. It calculates the climate-limited yield of each characterised site. Some other factors that may limit yield are not considered including trace elements, weeds, pests and diseases.

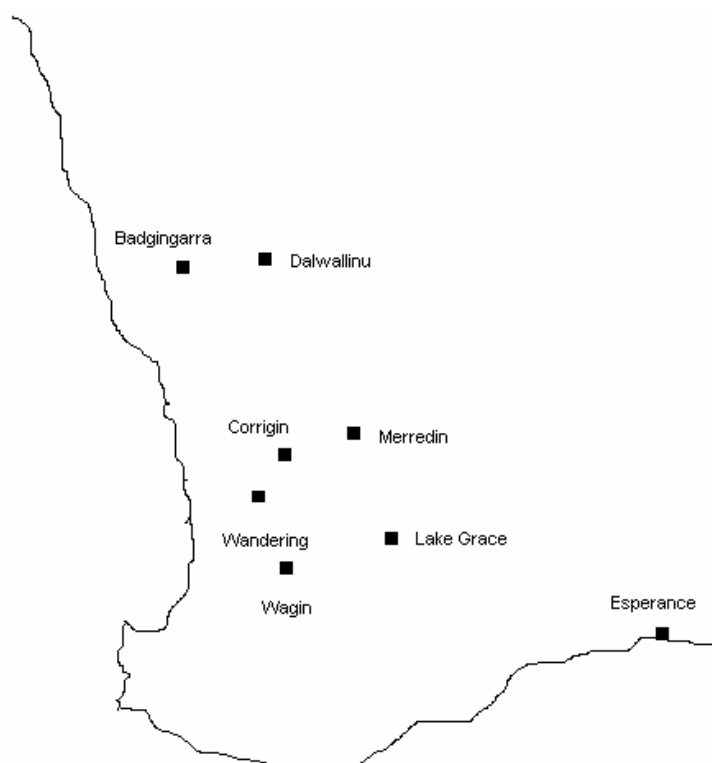


Figure 1 Study sites in the grainbelt of the south-west of Australia

The APSIM-Wheat model used in this study is the same as that used by Farré *et al.* (2007) which has been validated against experimental data from various sites in Western Australia. Farré *et al.* used the APSIM-Wheat model to consider yield impacts of two climate scenarios. Using 30 year periods, the model was run with climate data representing firstly the period 1976-2005 with current concentrations of CO₂ (350 ppm) and secondly, the period 2035-2064 with an expected higher CO₂ level (440 ppm). To generate the daily weather data used in the 30 year simulation periods the Cubic Conformic model (CCAM), which is a higher-resolution nested model of the CSIRO GCM MK3, was downscaled to provide samples of daily climate data for the current (1976-2005) and future (2035-2064) periods at each of the eight sites (Table 1). These sites represent the range of rainfall zones and agricultural sub-regions that form the wheatbelt of Western Australia.

Table 1 Characteristics of selected sites in the wheatbelt of Western Australia

Site	Agricultural zone	Annual rainfall (current climate 1976-2005) mm	Annual rainfall (future climate 2035 to 2064) mm	Reduction in annual rainfall associated with climate change %
Badgingarra	north	576	524	9.0
Dalwallinu	north	386	355	8.0
Merredin	central	353	329	6.8
Corrigin	central	389	368	5.4
Wandering	central	597	534	10.6
Lake Grace	south	367	336	8.4
Wagin	south	462	413	10.6
Esperance	south	664	622	6.3

Source: adapted from Farré *et al.* (2007)

At all sites the future climate scenario simulations pointed toward a decline in average annual rainfall, with the percentage decline ranging from 5.4% to 10.6%. Farré *et al.* examined the seasonal composition of these declines in rainfall and reported that the greatest rainfall reductions (up to 30%) were simulated to occur in the period April to June which is the period during which crops usually are sown.

In dryland wheat production, rainfall is a key determinant of grain yield. Hence, it may be anticipated that a decline in annual, and more particularly growing season rainfall, caused by climate change would lessen grain yield. However, other features of the future climate scenario, such as higher concentrations of CO₂ and changes in temperatures, also affect plant growth and grain yield. How these factors interact to alter grain yield is captured by the APSIM-Wheat model.

Using APSIM-Wheat to identify possible yield impacts of climate change requires invoking a range of assumptions regarding soil characteristics, starting conditions, varietal characteristics and crop sowing rules. The data used in this study rely on the following assumptions. Firstly, only clay soils at each site are considered and the soil at each site is assumed to be dry at 1 January each year. The first available day for sowing in any year is determined by the following sowing rule:

Between 25 April and 19 June, sow if rain over 3 consecutive days is greater or equal to $(25\text{mm} - 0.36\text{mm} \cdot (\text{day} - 115))$ where day is the numeric calendar day (e.g. Jan 1 = 1).

Effectively this rule is a linearly decreasing rain requirement as the growing season progresses. After 19 June, the sowing rule becomes; sow if there is at least 5mm of rain over 3 consecutive days.

The variety sown is a function of the date of sowing whereby a long season cultivar is assumed to be planted if sowing occurs before 20 May, whilst a medium season cultivar is sown between 21 May and 9 June, and a short season cultivar is sown after that date. Hence, in using this varietal sowing rule a degree of adaptation to climate change is assumed insofar as it is considered that farmers will alter the portfolio mix of varieties sown in accordance with changes to the dates of sowing.

Using these assumptions the APSIM-Wheat model was used to generate estimates of the potential Y_{st} in equation (5) for each year at each of the eight locations for the two climate scenarios. The averages of these estimates are listed in Table 2 while Table 3 lists the average date of the commencement of crop sowing in terms of calendar days for each climate scenario.

Table 2 Average potential yields on the first day of sowing at each site (kg/ha)

Site	Average potential yield (current climate 1976-2005) kg/ha	Average potential yield (future climate 2035 to 2064) kg/ha	Change in average potential yield %
Badgingarra	3527	3615	↓2.4
Dalwallinu	1806	1627	↓10.0
Merredin	1308	1160	↓11.3
Corrigin	1775	1698	↓4.3
Wandering	3143	3314	↑5.4
Lake Grace	1569	1388	↓11.5
Wagin	2584	2294	↓11.2
Esperance	4061	4100	↓0.9

The data in Tables 2 and 3 point to two main consequences of the projected change in climate for wheat grown on clay soils at several of the sites. Firstly at almost all sites, projected climate change causes crops to be sown a few days later with the delay to the start of sowing being up to almost a week at Badgingarra, Merredin and Wagin. Secondly, at all sites except Wandering the average

potential yield, based on the first day of sowing, declines in the face of climate change. Wandering is a high rainfall site on the western edge of the wheatbelt and the reduction in waterlogging, combined with higher CO₂ concentrations and warmer winter temperatures, that form part of the suggested climate change, allow grain yield to be higher on average. The data in Tables 2 and 3 show that it is the change in climate, rather than just the delay in sowing, that is the principal cause of the yield reductions at most sites. The projected reductions in initial yield at a majority of sites range from 200 to 600 kg/ha.

Table 3 Average date of the commencement of crop sowing at each site (calendar days) and estimates of the average delay to sowing associated with projected climate change (days).

Site	Calendar day of start of sowing		Delay to sowing ^a (days)
	current climate 1976 to 2005	future climate 2035 to 2064	
Badgingarra	139	146	6
Dalwallinu	152	155	3
Merredin	153	157	5
Corrigin	153	152	-1
Wandering	142	145	3
Lake Grace	153	155	2
Wagin	148	153	5
Esperance	140	142	2

^a This is the mean difference (rounded to whole days) of the calendar day starts to sowing for each climate scenario at each location.

Data from Farré *et al.* (2007) were complemented by re-running the model with identical specifications except that a range of sowing days were considered. This provided a dataset of yields according to day of sowing for each year in sequences of years that formed the two climate scenarios. A quadratic form was fitted to this data on yield versus time of sowing at each location. Examples of the simulation data for one year at four locations, assuming current climate are shown in Figure 2. Estimates of Y_{st} , b_t and c_t were generated at each location for each year t in the period of current climate 1976 to 2005 and for the projected climate, 2035 to 2064. In virtually all years and at all sites the adjusted R² values were greater than 0.95 for the fitted quadratic form.

The APSIM output for the current climate scenario is consistent with field experiments at a range of locations within the wheatbelt of Western Australia. These field trials show variation across years in the initial yield and response to later sowing. On average these trials suggest a rate of yield decline associated with a delay to sowing of between 15 to 35 kg/ha/day (Garlinge 2005).

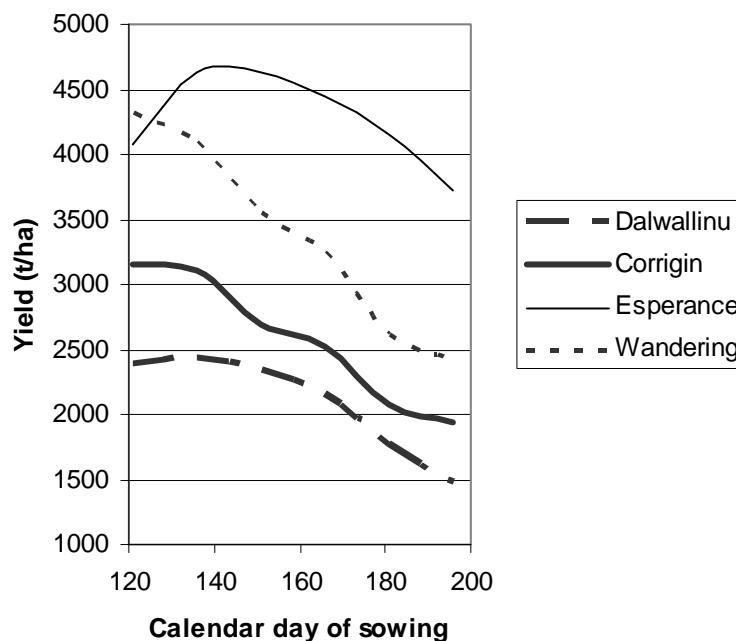


Figure 2 Examples of the APSIM output of wheat yield versus time of sowing for one year at four locations

Determining optimal investments in crop sowing machinery

R_{opt} was found by including equation 5 in a non-linear programming problem where the profit, specified by equation 5, was maximised through choice of R , subject to usual non-negativity and logical constraints. Re-expressing the right hand side terms of equation 5 by the symbol φ , then the non-linear programming problem at each location, for each climate scenario, is to find the level of R that maximizes profit from wheat production.

$$\text{Max}_R \varphi$$

s.t.

$$T = 30, \text{ (i.e. a sample of 30 years in each climate scenario)}$$

$$T = u + v + w$$

$$Y_{st} + b_t x - c_t x^2 \geq 0 \text{ (i.e. wheat yield cannot be negative)}$$

$$u \geq 0, v \geq 0, w \geq 0, R \geq 0, p_t \geq 0$$

As part of this optimization process values for p_t (wheat price in year t) at each location were specified by fitting a probability density function to farm-gate wheat prices at each location. These prices were based on the location's dominant grade (e.g. Australian Standard White or Australian Premium White) and its pool prices over the period 1995/6 to 2007/8. The farm gate price at each location in each year of the period was calculated as the published (and forecast for 2007/8) pool price less Cooperative Bulk Handling (CBH) outturn costs plus port differentials minus Australian Wheat Board Ltd finance costs minus CBH receival and grain assessment costs less grain freight to port less grain cartage to the receival point (also known as silo or bin) less levies (e.g. skeleton weed levy, grain research levy).

The historical handling and silo to port freight costs were obtained for each location from Cooperative Bulk Handling as reported in various issues of the Farm Budget Guide. The farm to silo grain freight was calculated assuming a 20 kilometre journey. For most locations farm-gate prices were well-approximated by lognormal distributions, generated by the Excel-Add in, *BestFit* (Palisade

Corporation 2002). For each year in the climate scenarios, at each location, 20 random price draws from the location's lognormal price distribution were used in calculating profit.

The cost of work rate, depicted as g in equation (4) and d and e in equation (5) (noting that $g = d + eR$) was estimated from farm survey data displayed in Figure 3.

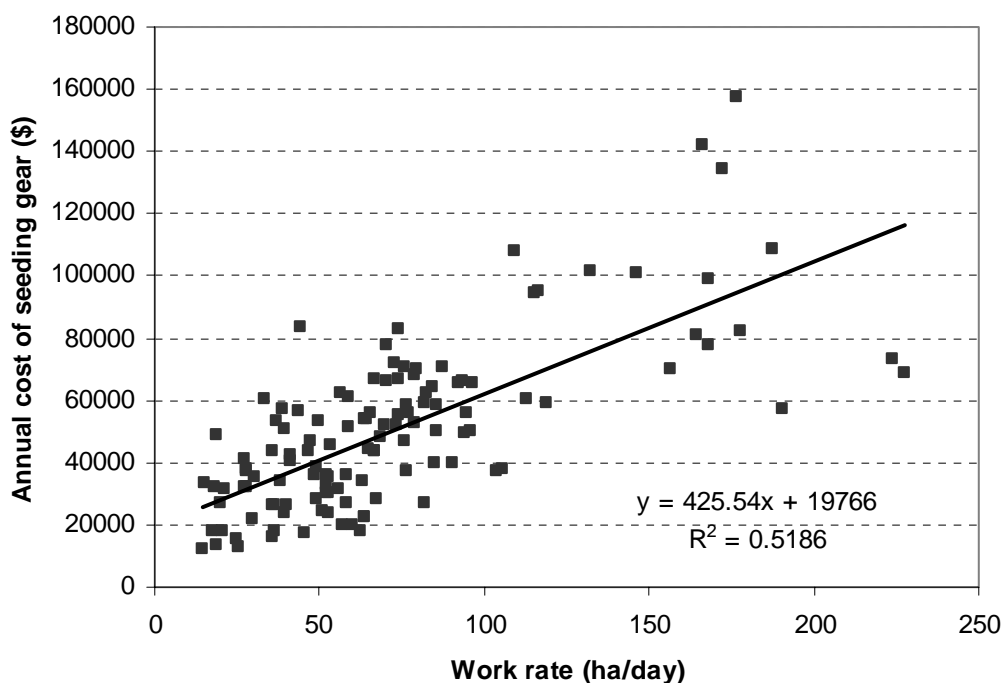


Figure 3 The relationship between the annual cost of investment in seeding machinery and sowing machinery work rate. Based on 2005/6 data for 122 broadacre farms in Western Australia

At each location the relationship $S_t = \alpha_s + \beta_s p_t Y_{st}$ was also estimated by drawing on firstly, the APSIM yield simulation data generated for each location; secondly, the price draws from the lognormal distributions of the farm-gate wheat prices at each location and thirdly, the current size of cropping programs at each location as reported by BankWest farm survey data (BankWest 2006). In calculating profits from crop production, the variable costs of wheat production, h_{min} , h_{max} and $h_t = \alpha_h + \beta_h p_t Y_{st}$, and other overhead costs (apart from fixed costs of machinery), m , were based on BankWest (2007) and DAFWA (2005) wheat gross margins for the locations, or nearest locations, in the grainbelt of Western Australia.

Of the 30 years that constituted the sample of years in each climate scenario, u and w years were associated with sowing programs of size S_{min} and S_{max} hectares respectively. At each location values for S_{min} and S_{max} were set based on BankWest data, historical data from farm management consultants and local farm adviser opinion. A diagrammatic representation of an example of the relationship $S_t = \alpha_s + \beta_s p_t Y_{st}$ is shown in Figure 4. The variable cost of wheat production is a similar shaped function, based on variable costs of wheat production at each location as specified in DAFWA (2005) wheat gross margins and adviser opinion. The treatment of variable costs assumes that in years where wheat revenues at the start of crop sowing are anticipated to be high (low) then, within limits, variable costs are increased (decreased).

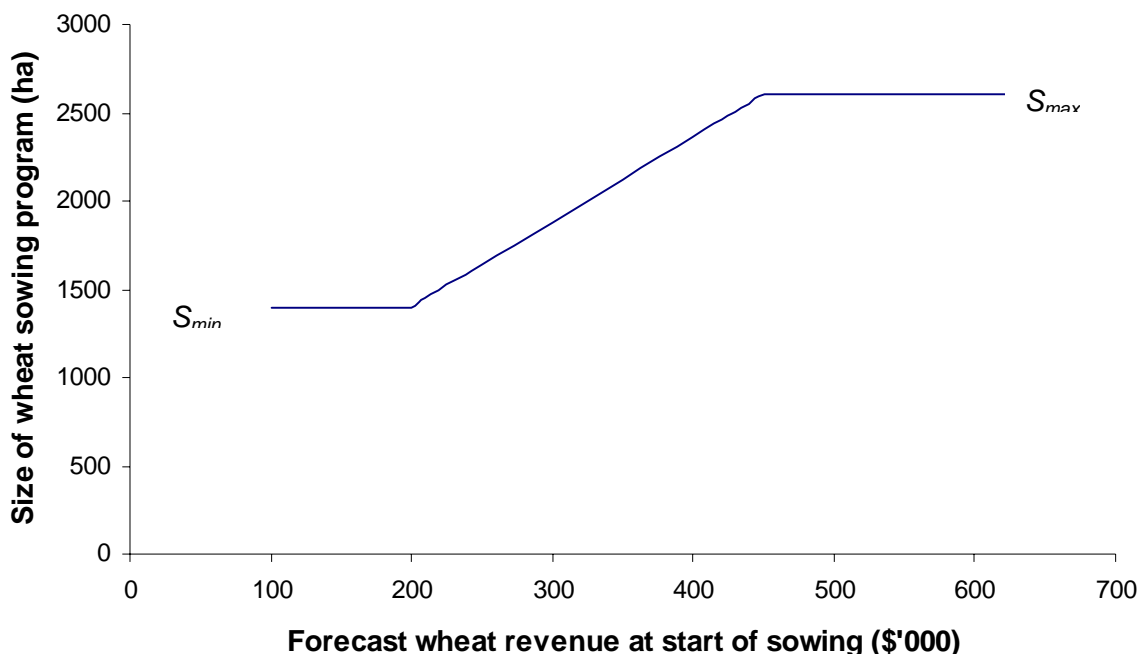


Figure 4 An example of the relationship between the size of the wheat sowing program (S_t) and the anticipated wheat revenue at the start of crop sowing ($p_y Y_{st}$).

Results and Discussion

The optimal investments in work rate, R_{opt} , for an average farm at each location were derived for current and projected climate regimes and results are presented in Table 4. At five of the eight locations optimal work rates are reduced by climate change impacts, whilst at Esperance, Wandering and Badgingarra work rates increase. The latter locations are in high rainfall parts of the grainbelt and benefit from climate change or experience only small declines in yield potentials (see Table 2).

Lake Grace and Merredin record the largest declines in work rate, 7.6 and 7.4 per cent respectively. At Lake Grace, Merredin and Dalwallinu farms are large and crop dominant, and these locations are all subject to very adverse impacts of climate change with the average commencement of crop sowing being delayed by several days and average yield potentials at the start of crop sowing being around 10 per cent less than average yield potentials at the start of crop sowing under the current climate regime. These low rainfall regions are the main sources of wheat production in Western Australia so the adverse impacts of climate change have potentially important ramifications for the State's and nation's wheat production..

Directly linked to the very adverse impacts of climate change at these locations are their large reductions in cumulative profit from wheat production. At Merredin, Lake Grace and Dalwallinu where large crop sowing machinery supports wheat production, the cumulative profit from wheat production is reduced in the presence of climate change by 35, 24 and 18 per cent respectively. The projected climate change causes a leftwards shift in the yield and profit distributions at these locations.

Given the assumption that the area sown to wheat is a state-contingent function of anticipated revenue at the commencement of sowing, then one of the impacts of climate change is a reduction in wheat plantings at these locations. Accordingly, as shown in Table 5, at Merredin, Lake Grace and Dalwallinu the average area sown to wheat is likely to decrease by around 3 per cent, *ceteris paribus*. This reduction in the average area sown to wheat in the presence of climate change is part of the reason for the decline in work rate. The other reasons for the decline in work rate are the differences

in yield responses to time of sowing with current versus future climate conditions and the lesser capital costs associated with owning crop sowing machinery of smaller size. As an illustration, consider the case of Merredin in Figures 5a (current climate) and 5b (future climate). There are greater yield penalties for late sowing in high return years under the current climate regime compared to the future climate scenario. Hence, there is greater incentive to invest in work rate under the current climate in order to lessen foregone profit in the 'high' return years. In Figures 5a and 5b the years that generate annual profits in excess of \$250,000 are shown by the symbol (■). The average yield penalty from late sowing in these high return years is 23.6 kg/ha/day and 15.9 kg/ha/day in the current and climate scenarios respectively.

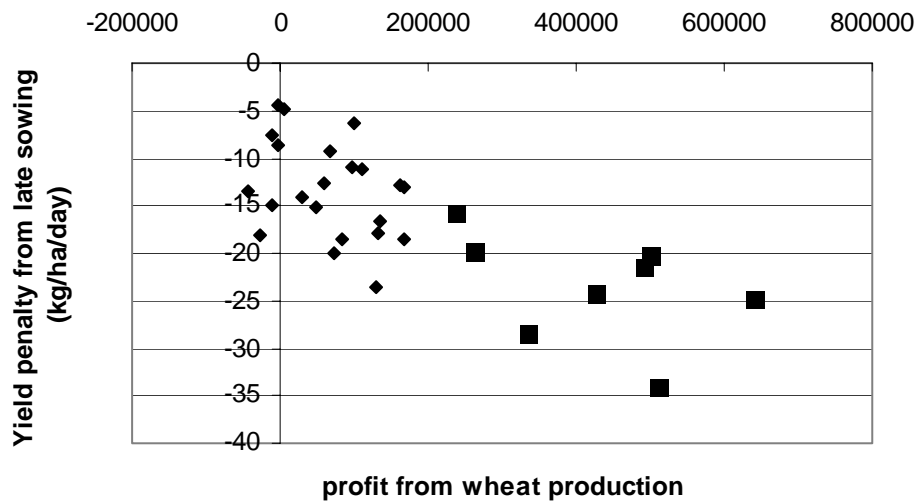


Figure 5a Yield penalty from late sowing versus profits from wheat production in the current climate scenario at Merredin. The symbol (■) refers to years that generate annual profits in excess of \$250,000.

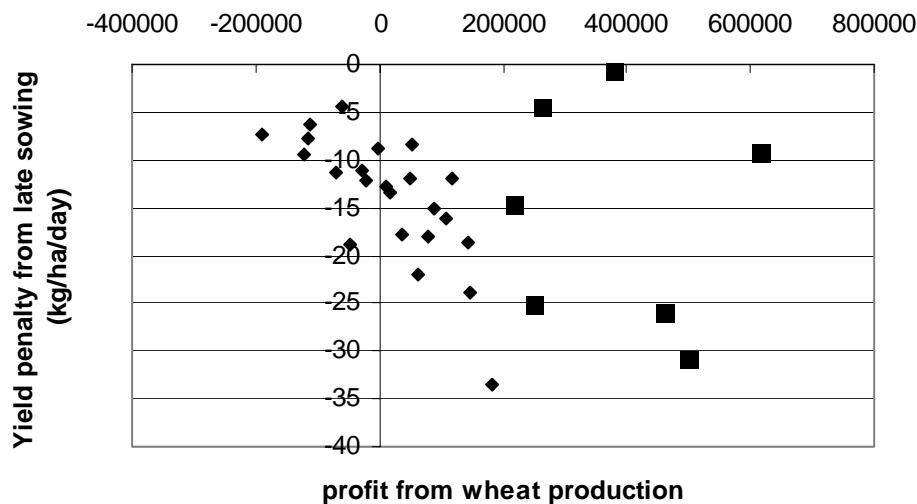


Figure 5b Yield penalty from late sowing versus profits from wheat production in the future climate scenario at Merredin. The symbol (■) refers to years that generate annual profits in excess of \$250,000.

Table 4 Optimal investment in crop sowing machinery work rate on an average farm at various sites for current and projected climate regimes; and estimates of cumulative profit from wheat production at each location for both climate regimes.

Optimal investment crop sowing machinery (in terms of work rate (ha/day))	Dalwallinu	Badgingarra	Merredin	Corrigin	Lake Grace	Wagin	Wandering	Esperance
Current climate 1976-2005	152	56	115	85	123	75	81	65
Future climate 2035-2064	151	58	106	83	113	71	86	71
Percentage change in work rate	↓ 0.8	↑ 2.3	↓ 7.4	↓ 2.2	↓ 7.6	↓ 5.3	↑ 5.8	↑ 9.0
Cumulative profit from wheat production (\$mill)								
Current climate 1976-2005	414.1	371.2	120.7	131.9	217.9	247.6	350.5	708.1
Future climate 2035-2064	341.3	356.9	78.7	117.8	165.0	193.4	386.6	714.2
Percentage change in cumulative profit	↓ 17.6	↓ 3.8	↓ 34.8	↓ 10.7	↓ 24.3	↓ 19.9	↑ 19.3	↑ 0.8

Table 5 Average areas sown to wheat on an average farm at each location under current and projected climate regimes.

	Unit	Dalwallinu	Badgingarra	Merredin	Corrigin	Lake Grace	Wagin	Wandering	Esperance
Average area sown in current climate 1976-2005 regime	ha	3098	1088	2625	1668	2588	1344	1348	1674
Average area sown in future climate 2035-2064 regime	ha	3004	1079	2552	1657	2508	1289	1378	1681
Percentage change in area sown	%	↓ 3.0	↓ 0.8	↓ 2.8	↓ 0.7	↓ 3.1	↓ 4.1	↑ 2.2	↑ 0.4

The general pattern of results is that at locations that are projected to be particularly adversely affected by climate change, a reduction in the work rate of sowing machinery is an element of the profit-maximising strategy. Conversely, at the few locations projected to benefit from climate change then the work rate of sowing machinery increases (see Figure 6). However, there are exceptions. At Dalwallinu there is a projected decline in cumulative profit from wheat production of almost 18 per cent yet there is only a very small reduction in work rate. The explanation for the results at Dalwallinu are that although climate change is associated with a 3 per cent reduction in the size of cropping program, thereby suggesting a slight reduction in work rate, there is a much greater offsetting yield penalty for late sowing after 20 days. In years with large programs when forecast profits are often relatively high, then the costs of late sowing are particularly high, and so investment in higher work rates is warranted. Also at Badgingarra where cumulative profit from wheat production is projected to decline by about 4 per cent, a small increase in work rate is recorded.

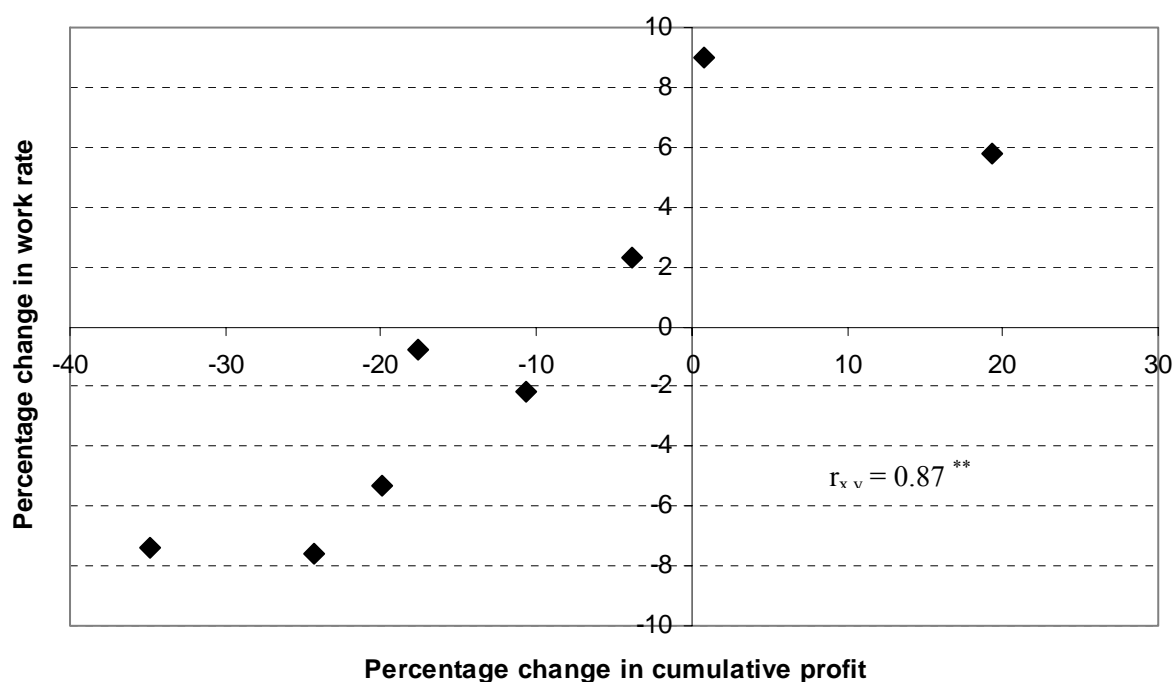


Figure 6 A plot of the eight locations' percentage changes in work rate versus percentage changes in cumulative profit associated with climate change. The correlation coefficient, $r_{x,y}$ is significant at the 1% level of significance.

The situation regarding yield penalties for late sowing at Dalwallinu is shown in Figures 7a and 7b. At Dalwallinu the rates of yield decline associated with larger sowing programs are much greater under the future climate scenario. This necessitates maintenance of the investment in sowing capacity as years in which cropping programs are large tend to be years of 'higher' returns and consequently the costs of late sowing are sufficiently large to justify investment in sowing machinery with large work rates. These findings are consistent with those of Kingwell (1996) who showed how yield penalties affect the use of and investment in crop sowing gear at a farm level.

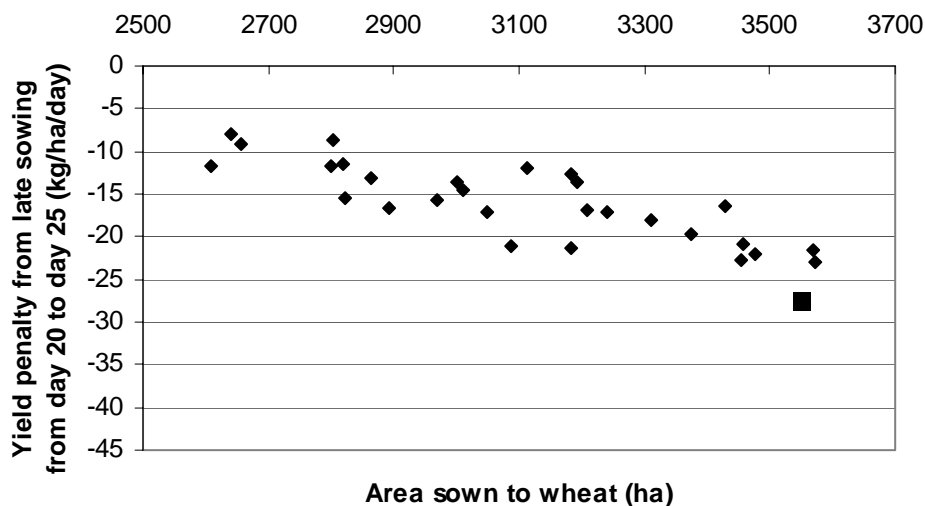


Figure 7a Yield penalty from late sowing versus area sown to wheat at Dalwallinu in the current climate scenario. The symbol (■) refers to years that generate yield penalties greater than 25 kg/ha/day.

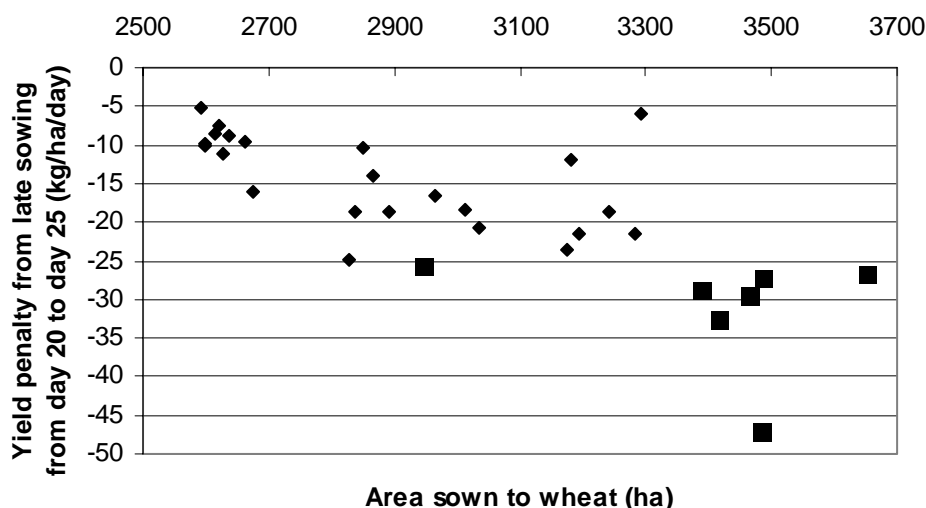


Figure 7b Yield penalty from late sowing versus area sown to wheat at Dalwallinu in the future climate scenario. The symbol (■) refers to years that generate yield penalties greater than 25 kg/ha/day.

There are several caveats to this analysis. Firstly, it is a partial analysis inasmuch as the focus of the analysis has been on the impact of climate change on grain yield and the subsequent impact on investment in crop sowing machinery. However, although the future climate regime may on average be less conducive to wheat production at several of the locations, nonetheless the rate and degree of yield decline could be altered by technical and managerial innovation.

Higher yielding short season varieties may ameliorate some of the potential loss of yield. Plant breeding may deliver greater drought tolerant varieties. Better systems of weed management (Loi *et al.* 2006 & 2007) may allow later sown crops to suffer less from weed competition or enable crops to be sown earlier. Management innovations such as grain soaking (Blackwell *et al.* 2007) will allow a more rapid emergence of crops, compensating for some delays to sowing. Technical innovations in

various aspects of crop preparation and sowing (e.g. knife-point, press-wheel, deep tillage, variable rate and spray technology, GM technologies) are likely to generate yield and profit improvements thereby affecting the returns from investment in crop sowing machinery. Innovation and adaptation to ameliorate or capitalise on unfolding climate change will impact on optimal choice of work rate.

Secondly, by concentrating on the likely yield implications of climate change at the various locations the analysis overlooks other important economic impacts and policy responses to climate change that will affect the relative profitability of wheat production. For example, biofuel subsidy policies, petroleum and natural gas pricing, carbon taxation, regional impacts of climate change and climate-related adaptation in other grain-producing regions or competing and complementary enterprises in concert will affect the prices of key inputs to wheat production such as fertilisers and fuel and will also affect wheat prices. It is the combination of all such factors that ultimately will govern the costliness of delays to crop sowing. However, to canvass all these possibilities is understandably beyond the scope of this paper.

Thirdly, the study does not consider the backdrop of the trend of increasing farm size that is usually associated with further investments in large machinery and associated labour-saving technologies. In this study the upper and lower limits on the sizes of cropping programmes at each location are fixed under each climate scenario. Yet economies of size benefits may complement or offset partially or wholly the adverse impacts of climate change and thereby affect a farmer's investment in work rate.

Fourthly, the in prospect division of weather-years into a third being 'good' or 'poor', and cropping programs being determined by in prospect conditions, is a key set of assumptions that influence the study's findings. Altering those assumptions will generate different findings, although interestingly fixing the crop area at each location generates similar differences in work rate as those reported in Table 4.

Lastly, this study is based on profit-maximising behaviour by a farmer and so risk aversion is overlooked. Australian farmers are mostly acknowledged to be slightly risk averse (Bond and Wonder 1980; Bar-Shira 1991; Abadi 2000) and thus their attitude to risk will affect their investment decisions regarding work rate.

Notwithstanding these important caveats, this study does suggest that in a major wheat growing region of Australia, climatic challenges to traditional wheat production are likely to emerge in coming decades. Wheat production, especially in currently low rainfall parts of this grainbelt will face environmental conditions less conducive to wheat production. Hence, maintaining the profitability of wheat production in this region through innovation and productivity gain will be a major technical challenge. If there is no off-setting movement in wheat prices or marked technological advances then a leftwards shift in the distribution of profits from wheat production is very likely at many locations in the low rainfall marginal parts of this grainbelt.

Conclusions

This study examines projected climate change at several locations in the grainbelt of Western Australia. Of particular interest in this study is how might farmers respond to the projected climate change at these locations, in terms of their investment in work rates for crop sowing? Farmers' investment in crop sowing machinery is a major component of their business expenses of grain production so it is important to understand how climate change may affect these costs.

Climate change at almost all locations considered in this study cause a slightly later sowing on average, given current technologies, and yield responses across times of sowing are altered. Accordingly, the interesting question arises: Will farmers increase or decrease their investment in work rate in the presence of climate change; and is there a consistent response across the grainbelt?

Farmers who maximise profits from wheat production will adjust their investments in crop sowing machinery in the face of climate change. Although subject to a range of caveats, the findings in this study suggest that at most locations in the grainbelt of Western Australia, fairly modest changes in investment in work rate, often less than 8 per cent of work rates under the climate regime of 1976-2005, are likely to form part of the profit-maximising response to climate change. The main finding is that there is a likely divergence in investment response between high versus low rainfall locations, with increases and decreases in work rates respectively being forecast. However, as illustrated by a few locations, the investment response to changing work rate in response to altered profit distributions, within a region of broadly similar rainfall, is not uniform. To forecast the possible impacts of climate change on machinery investments in work rate at a particular location requires knowledge of how climate change affects the pattern of yield response to time of sowing at that location, as well as knowledge of all the other ingredients in the farmer's profit function for grain production.

References

- Abadi, A. (2000) *Risk, uncertainty and learning in farmer adoption of a crop innovation*, Unpublished PhD thesis, Faculty of Agriculture, University of Western Australia.
- ABARE (2006) *Australian Commodity Statistics 2006*, Australian Bureau of Agricultural and Resource Economics, Canberra, ISSN 1325-8109, pp. 349.
- ACG (2005) *Climate Change Risk and Vulnerability*, A report by the Allen Consulting Group commissioned by the Australian Greenhouse Office, Department of Environment and Heritage, Canberra, pp. 159.
- Anderson, W.K. and Garlinge, J.R. (2000) *The Wheat Book: Principles and Practice*, Bulletin No. 4443, Department of Agriculture, Western Australia, pp.332.
- Asseng, S., Jamieson, P., Kimball, B., Pinter, P., Sayre, K., Bowden, J. and Howden, S. (2004) Simulated wheat growth affected by rising temperature, increased water deficit and elevated atmospheric CO₂. *Field Crops Research* 85: 85–102.
- BankWest (2006) BankWest Benchmarks: 2005-6, BankWest Agribusiness Centre, West Perth, pp. 87.
- Bar-Shira, Z. (1991) An approach to the econometric estimation of attitudes to risk in agriculture: comment. *Australian Journal of Agricultural Economics* 35:315-318.
- Blackwell, P., Riethmuller, G., Sharma, D. and Collins, M. (2007) Crop survival in challenging times. Paper presented to the Agribusiness Crop Updates, Burswood Entertainment Complex, Perth Western Australia, 14-15 February 2007.
- Bond, G. and Wonder, B. (1980) Risk attitudes amongst Australian farmers, *Australian Journal of Agricultural Economics* 24:16-34.
- CSIRO (2001) Climate change projections for Australia. Melbourne: CSIRO Atmospheric research. <http://www.dar.csiro.au/publications/projections2001.pdf> [accessed 11 April 2005]
- DAFWA (2005) Gross Margins Guide 2005: representative gross margins for crop and livestock enterprises of the agricultural regions of Western Australia, Department of Agriculture and Food, Western Australia, South Perth.
- Farré, I., Foster, I. and Charles, S. (2007) Future wheat yields in the West Australian wheatbelt. Paper presented to the Agribusiness Crop Updates, Burswood Entertainment Complex, Perth Western Australia, 14-15 February 2007.
- Foster, I. (2000) Environment. Chapter 1, In *The Wheat Book: Principles and Practice*, (Eds: W. Anderson and J. Garlinge), Bulletin No. 4655, Department of Agriculture, Western Australia.
- Garlinge, J. (2005) 2005 Crop Variety Sowing Guide for Western Australia, Bulletin 4482, ISSN 1448-0352, pp.205, Department of Agriculture and Food, Western Australia.
- Hennessy, K. and Jones, R. (2007) IPCC report on impacts, adaptation and vulnerability. Media briefing 10 April 2007, Climate Impacts and Risk Group, CSIRO Marine and Atmospheric Research.
- Heyhoe, E., Kim, Y., Kokoc, P., Levantis, C., Ahammad, H., Schneider, K., Crimp, S., Nelson, R., Flood, N. and Carter, J. (2007) Adapting to climate change issues and challenges in the agriculture sector. *Australian Commodities* 14 (1): 167-178.

- Howden, M. and Hayman, P. (2005) The distribution of cropping under climate change: Goyder's line. Contributed paper presented at the conference *Greenhouse2005*, Carlton Crest Hotel, Melbourne, 14-17 Nov, 2005.
- Howden, M. and Jones, R. (2001) Costs and benefits and CO₂ increase and climate change on the Australian wheat industry. A report prepared for the Australian Greenhouse Office, Canberra, Australia, pp.25. Downloadable at <http://www.greenhouse.gov.au/science/wheat>
- Howden, M. and Jones, R. (2004) Risk assessment of climate change impacts on Australia's wheat industry. In Proceedings for the 4th International Crop Science Congress, Brisbane, Australia, 26 September – 1 October 2004. Available at: www.cropscience.org.au
- Keating, B.A., Asseng, S., Brown, S., Carberry, P.S., Chapman, S., Dimes, J.P., Freebairn, D.M., Hammer, G.L., Hargreaves, J.N.G., Hochman, Z., Holzworth, D., Huth, N.I., Meinke, H., McCown, R.L., Mclean, G., Probert, M.E., Robertson, M.J., Silburn, M., Smith, C.J., Snow, V., Verburg, K., and Wang, E. (2003) The Agricultural Production Systems Simulator (APSIM): its history and current capability. *European Journal of Agronomy* 18: 267-288.
- Kingwell, R. (1994) Effects of tactical responses and risk aversion on farm wheat supply. *Review of Marketing and Agricultural Economics* 62:29-42
- Kingwell, R. (1996) Technical factors affecting investment in crop sowing machinery. *Australian Journal of Agricultural Economics* 40: 175-188.
- Kingwell, R. (2006) Climate change in Australia: agricultural impacts and adaptation. *Australian Agribusiness Review*, Paper 1, Vol.14, pp.30. ISSN 1442-6951 Downloadable at <http://www.agrifood.info/review/2006/Kingwell.htm>
- Kokic, P., Heaney, A., Pechey, L., Crimp, S. and Fisher, B. (2005) Climate change: predicting the impacts on agriculture: a case study. *Australian Commodities* 12(1): 161-170.
- Loi, A., Nutt, B. and Revell, C. 2006, '*Trifolium dasyurum* L. (Eastern star clover): Exploiting a novel pasture species for innovative strategies to control weed seed banks.' in *Proceedings of the 13th Australian Agronomy Conference*, Australian Society of Agronomy, Perth, Western Australia.
- Loi, A., Nutt, B. J. and Revell, C. K. 2007, 'AGWEST Sothis eastern star clover: a new pasture legume for weed management in cropping systems', in *2007 Agribusiness Crop Updates*, Burswood Entertainment Complex, Perth, Western Australia.
- Meinke, H., Hammer, G.L., van Keulen, H. and Rabbinge, R. (1998) Improving wheat simulation capabilities in Australia from a cropping systems perspective III. The integrated wheat model (I_WHEAT). *European Journal of Agronomy* 8:101-116.
- NSWDPI (2007) Guide to tractor and implement costs. Downloadable at <http://www.agric.nsw.gov.au/reader/machine+water/242kw-2007.pdf>
- Palisade Corporation (2002) Guide to using *BestFit*, version 4.5, February, 2002, Palisade Corporation, 31 Decker Road, Newfield, New Yor, USA, pp. 118.
- Pittock, B. (ed) (2003) *Climate change - An Australian Guide to the Science and Potential Impacts*, Report compiled for the Australian Greenhouse Office, Canberra, Australia. pp. 239. Available at <http://www.greenhouse.gov.au/science/guide/index.html>
- Reyenga, P.J., Howden, S.M, Meinke, H. and McKeon, G.M. (1999) Impacts of global change on cropping in SW Queensland. *Environmental Modelling and Software* 14: 297-306.
- Reyenga, P.J., Howden, S.M., Meinke, H. and Hall, W.B. (2001) Global impacts on wheat production along an environmental gradient in south Australia. *Environment International* 27: 195-200.
- Shackley, B.J. (2000) Crop Management. Chapter 7 In *The Wheat Book: Principles and Practice*, (Eds: W. Anderson and J. Garlinge), Bulletin No. 4443, Department of Agriculture, Western Australia.
- Wang, E., van Oosterom, E., Meinke, H., Asseng, S., Robertson, M., Huth, N., Keating, B. and Probert, M. (2006) The new APSIM-Wheat Model - performance and future improvements. Proceedings of the 13th Australian Agronomy Conference, Australian Society of Agronomy, 10-15 September, 2006, Perth.