

Adaptation response surfaces for managing wheat under perturbed climate and CO₂ in a Mediterranean environment

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ABSTRACT

Adaptation of crops to climate change has to be addressed locally due to the variability of soil, climate and the specific socio-economic settings influencing farm management decisions. Adaptation of rainfed cropping systems in the Mediterranean is especially challenging due to the projected decline in precipitation in the coming decades, which will increase the risk of droughts. Methods that can help explore uncertainties in climate projections and crop modelling, such as impact response surfaces (IRSSs) and ensemble modelling, can then be valuable for identifying effective adaptations. Here, an ensemble of 17 crop models was used to simulate a total of 54 adaptation options for rainfed winter wheat (*Triticum aestivum*) at Lleida (NE Spain). To support the ensemble building, an *ex post* quality check of model simulations based on several criteria was performed. Those criteria were based on the “According to Our Current Knowledge” (AOCK) concept, which has been formalized here. Adaptations were based on changes in cultivars and management regarding phenology, vernalization, sowing date and irrigation. The effects of adaptation options under changed precipitation (P), temperature (T), [CO₂] and soil type were analysed by constructing response surfaces, which we termed, in accordance with their specific purpose, adaptation response surfaces (ARSs). These were created to assess the effect of adaptations through a range of plausible P, T and [CO₂] perturbations. The results indicated that impacts of altered climate were predominantly negative. No single adaptation was capable of overcoming the detrimental effect of the complex interactions imposed by the P, T and [CO₂] perturbations except for supplementary irrigation (SI), which reduced the potential impacts under most of the perturbations. Yet, a combination of adaptations for dealing with climate change demonstrated that effective adaptation is possible at Lleida. Combinations based on a cultivar without vernalization requirements showed good and wide adaptation potential. Few combined adaptation options

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performed well under rainfed conditions. However, a single sI was sufficient to develop a high adaptation potential, including options mainly based on spring wheat, current cycle duration and early sowing date. Depending on local environment (e.g. soil type), many of these adaptations can maintain current yield levels under moderate changes in T and P, and some also under strong changes. We conclude that ARSs can offer a useful tool for supporting planning of field level adaptation under conditions of high uncertainty.

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1. Introduction

The Mediterranean basin has been identified as one of the most prominent climate change hotspots due to ongoing and projected changes in both means and variability of temperature and precipitation (Diffenbaugh and Giorgi, 2012). Observed climate showed a trend towards increasing temperature and declining rainfall over recent decades (Hartmann et al., 2013, Figs. 2.22 and 2.29). This trend is estimated to continue over the 21st century, with multi-model median projections showing reduced annual precipitation (ca. –15% of projected mean change, with high uncertainty) and increased mean annual temperature (ca. +5 °C) by the end of the century under the highest scenario of radiative forcing of the atmosphere (8.5 W m^{-2}) simulated by climate models (RCP8.5; IPCC, 2013). Associated with these changes is a projected increase in the frequency and intensity of extreme daily maximum temperatures (Collins et al., 2013; Sánchez et al., 2004, for the Iberian Peninsula). In turn, climate change is estimated to generate severe impacts on agriculture, constituting an important risk to food production (Asseng et al., 2015). These impacts are mainly driven by drought and heat stress. Specifically for wheat (*Triticum aestivum*), the majority of the impacts are driven by temperature (Tack et al., 2015) rather than precipitation, resulting in significant yield reduction with temperature increases (Asseng et al., 2015). However, high-temperature episodes might have larger negative effects under water stress conditions, which are very frequent in the Mediterranean region (e.g. Moriondo et al., 2010). The positive effects of increasing $[\text{CO}_2]$ do not completely counteract this trend (Ferrise et al., 2011). The wheat yield stagnation detected in several regions during recent decades (Grassini et al., 2013) could be the first evidence of these effects.

Specifically for winter wheat in Spain, Olesen et al. (2007) projected a mean yield decrease of 21% by the end of the 21st century, and Mínguez et al. (2007) revealed a northward shift of winter wheat failure due to disturbed vernalization (i.e. when cold requirements for normal flowering induction are not met due to higher temperatures). Mínguez et al. (2007) also projected a variable yield response in northern Spain, with yield decreases linked to low elevation areas and increases to higher elevation. Also for Spain, Iglesias et al. (2010) concluded that decreases in cereal yields would be less severe in northern parts of Spain than in southern parts. Taken together, this evidence for substantial impacts of climate change on winter wheat cultivated under current practices in the Mediterranean basin, points to a clear need to examine options for adaptation.

To identify optimal adaptation strategies, the quantification of complex crop–climate–soil interactions and phenology analysis is essential (Rötter et al., 2011; Wang et al., 2015). Current research is exploring strategies to deal with heat and water stress by modifying crop varieties and management according to the environment, optimizing the Genotype \times Management \times Environment (G \times M \times E) interaction (Montesino-San Martín et al., 2015; Rötter et al., 2015).

The development of new heat-tolerant cultivars has been recommended (Olesen et al., 2011; Semenov et al., 2014; Tanaka et al., 2015) to deal with the projected changes in both mean climate conditions and occurrence of extreme events during sensitive periods of the crop cycle (Moriondo et al., 2011; Ruiz-Ramos et al., 2011). Other recommendations include changes in cycle duration, seeking a better match of phenology with future local climate (Semenov et al., 2014). Specifically for Spain and winter wheat, the shortening of the crop

cycle has been suggested as an adaptation option, in particular when considering the effect of heat stress (Moriondo et al., 2011). Also, where higher temperatures trigger an increase of development rate, longer cycles could show adaptation potential (Giannakopoulos et al., 2009). Other potential cultivar characteristics, desirable for both current and future cultivars, could include increased water use efficiency and the ability to exploit high $[\text{CO}_2]$ (O'Leary et al., 2015). Removal of vernalization requirements is also a promising strategy, as indicated by the positive impacts reported for spring wheat due to milder temperatures in Spain under future conditions. This effect was more evident in northern Spain (the region where the winter wheat currently meets vernalization requirements), as current winter temperatures limit crop growth (Mínguez et al., 2007).

Regarding management, researchers are proposing the adoption of early sowing dates for many European regions (Olesen et al., 2011) and specifically for Spain and winter wheat (Giannakopoulos et al., 2009; Moriondo et al., 2010). Also, water management optimization under rainfed (Olesen et al., 2011) or irrigated cultivation has been suggested, as well as the expansion of irrigation infrastructure (Tanaka et al., 2015), and an increased intensity of both the cropping system and the irrigation system (Mínguez et al., 2007; Moriondo et al. (2010), although the latter might be precluded in Spain due to future competition for water resources among agriculture and other sectors. The ultimate goal of these techniques is not only to increase but also to stabilize yields.

A single adaptation option might not be enough to cope with climate change impacts in a Mediterranean climate. Results for Spain indicate that combined adaptation measures ameliorate yield declines better than single ones, and in some cases are capable of maintaining current yields (e.g. Rey et al., 2011). For example, Gabaldón-Leal et al. (2015) simulated yield responses to a combination of adjustments in sowing date, growth duration, phyllochron, grain filling rate and water use efficiency for irrigated maize in Spain.

In practice, there are trade-offs between the outcomes of individual adaptations, which can complicate the implementation of potential adaptation measures. For example, under dry conditions early maturity helps a crop to avoid excessive water stress but also results in lower yields due to a shorter crop cycle (Semenov et al., 2014). In turn, these trade-offs will depend on site-specific conditions and the particular combination of adaptations selected. In this complex context, researchers, technicians and farmers require the development of locally tailored adaptation strategies (Reidsma et al., 2015) according to the magnitude and features of the projected regional climate change, soil type, and economic analyses of favourable management options (e.g. for different water prices, Rey et al., 2011).

Besides practical implementation, the modelling of adaptation is also challenging because of multiple sources of uncertainty (Iglesias et al., 2010). There can be large discrepancies between simulated and measured results (O'Leary et al., 2015), and wheat model sensitivities to climate can also differ substantially (Pirttioja et al., 2015; Ruane et al., 2016). Thus, in spite of the recent improvements in modelling, there are still challenges in simulating processes affecting responses in phenology, water use efficiency or harvest index, even under current climate conditions. These uncertainties can be exacerbated under higher temperatures outside the range of historical experience (Asseng et al., 2015).

Crop models, usually based on ecophysiological knowledge and drawing, to a varied extent, on empirical relationships, have been proven to be useful tools to simulate G \times M \times E interactions for both deepening our

understanding of crop behaviour and also facilitating application in studies of crop response (Montesino-San Martín et al., 2015; Rötter et al., 2015). Moreover, crop model ensembles can provide a powerful tool for combining the responses of various crop simulation models. This can provide useful insights on inter-model uncertainties in yield responses. It can also help to identify robust behaviour across models, both in terms of average responses (e.g. see the multi-model sensitivity study by Pirttioja et al. (2015) that utilizes impact response surfaces) as well as sensitivity to interannual climate variability (Ruane et al., 2016). No model reproduces observations perfectly under different environments (e.g. Palosuo et al., 2011) and some authors claim that multi-model ensemble-average estimates are more accurate than estimates from any single model in simulating yield responses to climate change (Asseng et al., 2015), with improved accuracy as ensemble size increases up to a certain number of ensemble members beyond which improvements become negligible (Martre et al., 2015).

The objective of this paper is to analyse yield response to changes in mean climatic conditions, as well as to explore options for effective local adaptation of wheat in the Mediterranean under highly uncertain future. For this purpose, we used an ensemble of 17 crop simulation models to calculate wheat yield responses in Lleida, Spain, through a range of possible climate perturbations with various adaptation options. Impact response surfaces (IRs) were used in the analysis, and we have developed the concept of adaptation response surfaces (ARs) to support the analysis of adaptation.

2. Material and methods

2.1. Study site, experimental data and climate data

The study was conducted at Lleida in northeastern Spain, one location representative of the wheat cultivation area in the “Mediterranean

South” environmental zone (Metzger et al., 2005, Fig. 1a). The principal characteristics and agro-climatic conditions of Lleida are summarized in Fig. 1b, c, d.

Modellers were provided with phenological observations (flowering and maturity dates), total above-ground biomass, average grain weights and yields for the winter wheat variety Soissons. These field data were taken from Abeledo et al. (2008) and Cartelle et al. (2006) and were basically the same as those used by Pirttioja et al. (2015). The data included three irrigated treatments and one for rainfed, four N management treatments and six sowing dates (spanning from November to February), and comprised two locations close to each other in Lleida province (Agramunt and Gimenezells) and two growing seasons (2003–2004 and 2005–2006). Statistical data for Lleida province means from Spanish Ministry of Agriculture were used to check interannual variability for the period 1980–2010. Two representative actual soil profiles from these locations, exhibiting contrasting soil depth and texture and therefore water holding capacity were considered in the study (Fig. 1b).

Weather data from micrometeorological weather stations located in the field experiments were considered for model calibration. For the simulation phase (described in Section 2.2.3), the baseline climate was taken from the AEMET weather station at Lleida during the period 1980–2010. These baseline data were then perturbed to conduct a sensitivity analysis of the crop models to changes in climate. All datasets were at a daily time step for global solar radiation, minimum and maximum temperature (Tmin, Tmax), precipitation (P), wind speed and relative humidity. Missing values were derived from ERA-Interim reanalysis as described in Pirttioja et al. (2015). To create the perturbed dataset, P and temperature (T) baseline values were systematically modified using a “change factor” approach in combination with a seasonal pattern of the T and P changes (Fronzek et al., 2010). Observed daily Tmax and Tmin were modified between -1 °C and $+7$ °C at 1 °C intervals. Daily P was modified between -40% and $+30\%$ at 10%

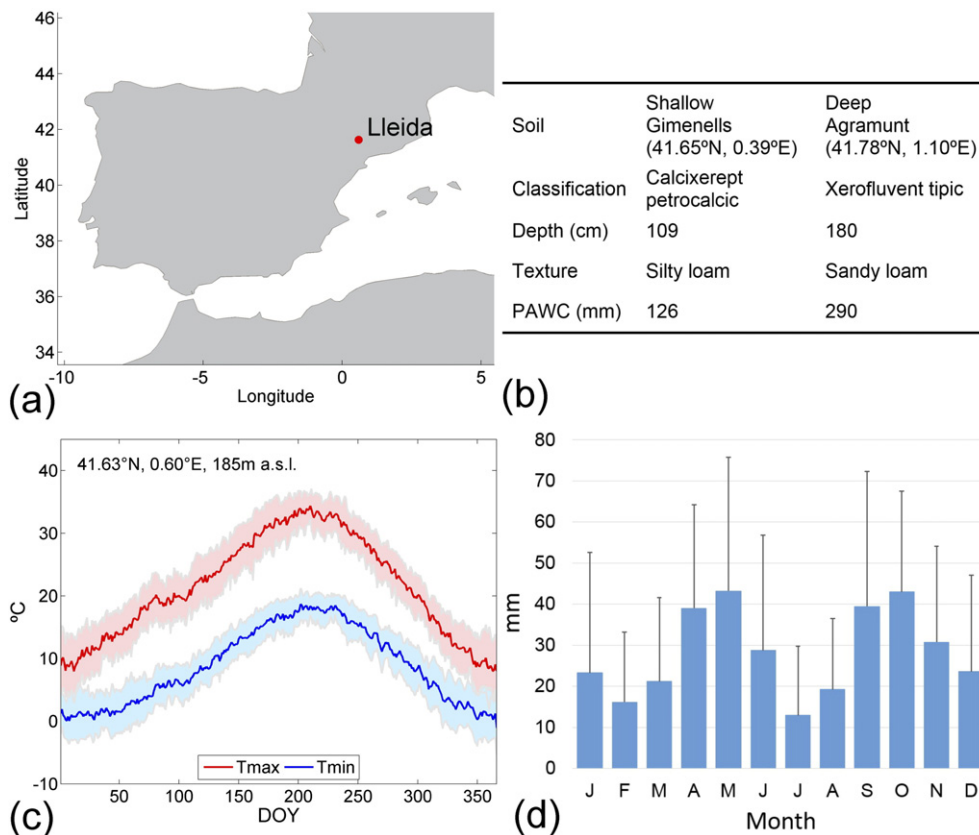


Fig. 1. Location of the study site (a). Main features of the soils considered in the simulation phase (b). Long-term (1981–2010) mean daily minimum (Tmin) and maximum (Tmax) temperature with standard deviation (SD) as shadow areas (c) and monthly precipitation with SD as error bars (d) at the Lleida weather station.

intervals. These ranges were set to encompass uncertainties in projected climate changes at Lleida by the middle of the 21st century for the medium SRES A1B emissions scenario. These uncertainties combine information from observations and from climate models, specifically multi-model ensemble results from the Coupled Model Intercomparison Project phase 3 (CMIP3) and perturbed physics ensembles (Harris et al., 2010), and are also consistent with regional projections for the highest implied emissions (RCP8.5) from the multi-model ensemble CMIP5 data sets (Collins et al., 2013). Thus, each year of the baseline was modified according to 72 combinations of perturbed T and P. To account for differences in the treatment of air humidity by crop models, vapour pressure and dew point temperature were corrected for temperature changes by assuming relative humidity to remain unchanged. The introduction of the seasonal pattern consisted of applying seasonal weights to the baseline climate instead of constant T and P changes over all days of the year, while retaining the annual mean changes (Fronzek et al., 2010). The seasonal weights were calculated from the ensemble mean of the Hadley Centre probabilistic climate change projection based on the A1B emissions scenario (Harris et al., 2010).

2.2. Crop modelling

2.2.1. Crop models and calibration guidelines

An ensemble of 17 members made up of 14 wheat models and 17 modelling groups was applied in this study (listed in Table S1 in Suppl. mat.). All the models used in this study have been applied successfully over a range of regions and applications (references in Rosenzweig et al., 2013; examples of ensemble modelling Asseng et al., 2015; Martre et al., 2015; Pirttioja et al., 2015; Palosuo et al., 2011). The same model/version was contributed separately by five modelling groups for two models (2 × CERES-wheat and 3 × WOFOST) and they were considered as separate models following Pirttioja et al. (2015). An overview of all crop models and their differences in terms of main process representation is provided in Pirttioja et al. (2015). Most of the models were developed for the field scale, except for CARAIB, LPJ and MCWLA, which were developed for larger area modelling. All models operate on a daily time step.

Modellers were asked to calibrate their model using the experimental data provided that included a diversity of treatments to represent the crop cultivation in the study area (see Section 2.1) and assuming [CO₂] to be 360 ppm. The calibration was done first for parameters affecting phenology and thereafter for parameters affecting crop growth and yield. The specific formulation of these parameters varied among models. The modellers were left free to use the calibration method they preferred (e.g. trial and error or automatic), but they were asked to check that calibration performance was within the “good” category according to Jamieson et al. (1991). This corresponds to a normalized root mean square error between simulated and observed values lower than 20%.

2.2.2. Preliminary study to identify potential adaptations

To save the workload of modellers, two crop models, CERES-wheat (Jones et al., 2003) DSSATv4.5 and SiriusQuality2 (SQ2) (Martre et al., 2006) were selected to explore and narrow the wide range of possible adaptations, and to identify the most promising ones to be simulated by all the participants. This was necessary due to the huge number of possible adaptation options that result just from combining some of the most important options, such as changes in crop phenology, sowing dates and water management.

Thus, two types of phenological change were tested: changes in the vernalization requirements (Mínguez et al., 2007) and changes in the length of the phenological phases. In the study region spring wheat spans the same sowing dates as winter wheat; sowing date in turn depends on autumn rain, so switching between spring- and winter-type cultivars is already possible. Both shorter and longer growth cycles

were proposed; the shorter cycles could provide beneficial if they reduce the risk of encountering adverse weather events (Trnka et al., 2014) while longer cycles would benefit from a prolonged period of photosynthesis and grain filling (Olesen et al., 2007; Trnka et al., 2014). Both reduced and extended growth cycles are regarded as feasible options by agronomy experts (Olesen et al., 2011).

Changes in sowing dates were also explored: advancing sowing date increases crop growth cycle length, moves the cycle towards a cooler part of year and helps the crop to avoid the stressful conditions at the end of the cycle (Ferrise et al., 2011; Olesen et al., 2007). However such advancement is constrained by the end of the dry season, projected to last longer on the Iberian Peninsula in the future (Trnka et al., 2014). A delayed sowing date moves the crop cycle towards the most effective period for fulfilling the vernalization requirements (Moriondo et al., 2011) and closer to the winter precipitation (Ruiz-Ramos and Mínguez, 2010).

Supplementary irrigation (sl) at sensitive crop growth stages is a means to stabilize yields, and is already being explored for many crops in Spain and in other Mediterranean countries (Lorite et al., 2012). Table 1 summarizes the options (single and combined) tested for identification of adaptations.

2.2.3. Simulation phase

A first set of model runs consisted of simulations without consideration of adaptation measures, i.e. standard simulations. We define standard simulations as those with the winter wheat standard cultivar; crop cycle and stages defined as in the calibration process (Section 2.2.1) and standard management, i.e. sowing date set at day of year (DOY) 302, no irrigation but no other limiting factors. A second set of model runs consisting of a limited set of promising single and combined adaptation options were selected in the preliminary phase and simulated by all participants. A full irrigation (fl) scenario, simulating crop growth without any water stress, also served as a reference for identifying yield ceilings and associated water requirements.

The sensitivity analysis of crop response to climate change was performed by following the general procedure described in, for example, Martre et al. (2015) and Pirttioja et al. (2015). The period 1981–2010 was used as the baseline. Models were applied under the 72 combinations of perturbed weather described in Section 2.1 and with two soil profiles (Fig. 1b). Simulations were performed as a series of independent growing seasons with initial soil water content set at 75% of field capacity at the beginning of the season. Two levels of [CO₂] representing two 20-year time slices for periods centred on 2030 and 2050 according to A1B projections from the BernCC model (Nakicenovic and Swart, 2000) were considered. Additionally, the standard simulations were performed with [CO₂] at 360 ppm as a baseline level.

2.3. Analysis of results

IRs are plotted surfaces that show the response of an impact variable to changes in two explanatory variables (here, P and T). They are constructed by plotting the results of the sensitivity analysis of model outputs as contour lines along the axes of P and T changes (Fronzek et al., 2010; Pirttioja et al., 2015). Contour lines were computed with the statistical software environment R (<http://www.R-project.org/>). Individual IRs of the 30-year mean yield were generated for each crop model and standard management and cultivar or adaptation option. Yield estimates across multiple models were averaged using the ensemble median to be consistent with previous studies (e.g. Asseng et al., 2015; Pirttioja et al., 2015). The utility of IRs analyses for displaying how crop models differ in their sensitivity to systematic changes in input variables such as T and P has been demonstrated in earlier studies (Ferrise et al., 2011; Pirttioja et al., 2015).

In order to analyse the effect of an adaptation option under combinations of T and P change, IRs were constructed both with and without the given adaptation option and for the same [CO₂] and then the

Table 1
Adaptation options and climate/[CO₂] perturbations tested in the preliminary phase (normal font) and selected for the simulation phase (bold font). Simulations of the preliminary phase were run with CERES-wheat and SiriusQuality2 models.

Adaptation	Tested range	N preliminary phase	N simulation phase
Vernalization (V)	Yes (WW, standard), No (SW)	2	2
Phenology (Ph)	−10%, 0%, +10%, +20%	4	3
Sowing date (SD)	−15, 0, +15, +30, +45	5	3
Irrigation (I)	Supplementary (40 mm at flowering), Full (only CERES), rainfed	3	3
Combinations Perturbations (P, T, CO ₂)	V × Ph × SD × I	2 × 4 × 5 × 3 = 120 (0%, 0 °C, 360 ppm), (−15%, +2 °C, 447 ppm), (−30%, +4 °C, 521 ppm)	54 From −40 to +30 at 1 °C intervals, from −1 to +7 °C at 10% intervals, for 360, 447 and 521 ppm

Abbreviations: winter wheat (WW), spring wheat (SW), temperature (T), precipitation (P).

difference between the two was calculated. We refer to this new plot as an Adaptation Response Surface (ARS – illustrated in Fig. 2). In this context, the isolines of the ARS can be expressed as an absolute value or as a percentage (as in Fig. 2). Baseline 30-year mean yields (i.e. of the unperturbed simulation) were considered together with the ARSs to evaluate the ability of different adaptation options to maintain current yield levels, labelled here as the “recovery response”. This was done by superimposing a mask on the Adapted IRSs and the corresponding ARSs using the current yield as threshold (colour mask in Fig. 2). Current yield was calculated as the multi-model median of 30-year mean yields at unperturbed P and T under 360 ppm and with no adaptation for each soil type. It is important to stress that the ARS isolines refer to the adaptation response as defined above, and not to the recovery response.

Inter-annual variability of yield was expressed as the ensemble median of the coefficients of variation (CVs) across the 30 years for each model. The spread in model responses was evaluated by the inter-quartile range of the 30-year mean (IQR, from the 25th to the 75th percentile).

It has been suggested that multi-member ensembles of crop simulation models generate more robust results in projecting impacts of climate change (Asseng et al., 2015; Martre et al., 2015; Palosuo et al., 2011). In our study, slightly different ensembles were used for the various adaptation options. This was because the construction of the ensembles was co-determined by the fact that some models did not simulate all the options, summarized in Table S1. Besides, we excluded all the simulations with technical errors and those with very evidently unrealistic model response or absolute values of simulated variables after one-round of discussion with the modellers and based on the

authors' expert judgment and referred to here as “According to Our Current Knowledge” (AOCK). In that way, the simulations were checked for some highly implausible ecophysiological responses. Thus, for model simulations assuming standard management and cultivar, if one or more of the AOCK-defined criteria were met by the 30-year mean of an output variable, the simulation was excluded. For these criteria, thresholds were set supported by the literature and/or AOCK. Given the subjective nature of this exercise the thresholds were set sufficiently wide to ensure that plausible results were not excluded from the ensembles. The criteria were as follows:

- yields were higher than 10.5 t ha^{−1} under rainfed conditions for unperturbed P and T, which is ca. 3 t (40%) higher than the maximum yield observed from irrigated treatments (Abeledo et al., 2008).
- yields were higher for the shallow soil than for the deep soil when all other settings were the same for unperturbed P and T, which is highly implausible in a water stressed environment, given that deep soils can hold more water.
- yields were the same for rainfed and irrigated simulations for unperturbed P and T, whereas field data show clear benefits from irrigation.
- yields were higher for rainfed compared to irrigated at unperturbed P and increased T, when the simulated length of the cycle and all other conditions were the same. This is very unlikely as, given all other conditions are identical, rainfed yields limited by water availability should always be lower or maximum the same as yields of full irrigated crops that do not experience water stress.
- the yield change was independent of P under rainfed conditions, i.e. less than 5% of change in yield from −40% to +40% of P change, whereas

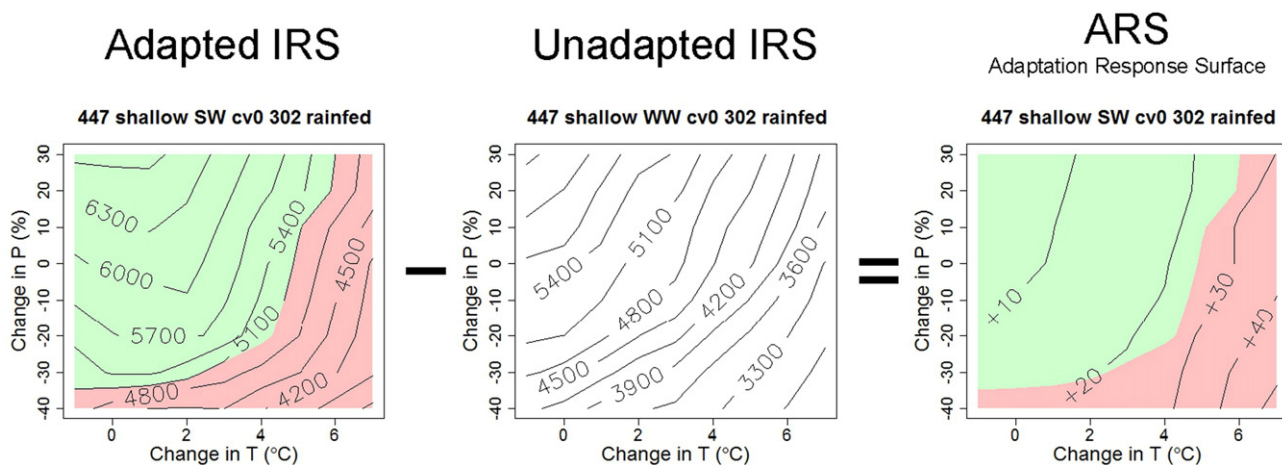


Fig. 2. Example of adaptation response surface (ARS) construction. An ARS results from subtracting two impact response surfaces (IRSs): one considering the adaptation to be evaluated (here using spring wheat), and the other the standard, unadapted option. In this case, the isolines of yield in the IRSs are in kg ha^{−1}, while the results in the ARS are expressed as % of change from the unadapted option. Both IRSs correspond to the same [CO₂] (here 447 ppm) and the same soil (here the shallow soil); therefore, only the effect of the adaptation is evaluated. The green and red mask separates respective yields above and below a reference yield, defined as the baseline 30-year mean yield at unperturbed P and T under 360 ppm for a given soil type without any adaptation (here 5100 kg ha^{−1}).

field data show a clear sensitivity to precipitation.

- the simulated crop cycle was too short for both standard and adapted cultivars. Warming is known to accelerate phenological development of crops, but it is unrealistic for crop cycle to be excessively reduced. Threshold for earliest end of cycle at maturity defined here was 1st of April when checked up to +2 °C. This is 2 months earlier than current observed dates in the South of Spain with mean annual temperature ca. 2 °C warmer than Lleida - data from www.aemet.es.

3. Results

3.1. Crop model calibration

Normalized RMSEs for the simulated anthesis and maturity dates for the ensemble of models were 3.8% and 3.2% respectively (corresponding to a mean error of 5.8 and 5.8 days respectively). Normalized RMSEs for the simulated yield and biomass for the ensemble of models were 14.6% and 9.6%. Normalized RMSE for all these variables for each individual model was always below 20% for biomass and yield and up to 5% for anthesis and maturity dates (see Table S2 in Suppl. mat. for individual model results). Although the observed data set is referred only to two years this can be compensated by the wide variability of treatments considered. Besides, we checked that the ensemble members and the ensemble median were generally able to reproduce the interannual variability for the 1980–2010 period (Fig. S1). Therefore, according to Jamieson et al. (1991), we considered all the models acceptable for this specific application at this location.

3.2. Preliminary phase

Simulated yields for the adaptations tested in the preliminary phase are reported in Table 2. Table 2 illustrates the actual situation in crop modelling, when models show opposite and still feasible answers that could be explained by differences between model sensitivity to several factors (e.g. soil, water availability, temperature) and their interactions. In our case this created an uncertainty in the selection of adaptation options. To deal with this uncertainty, the adaptation options for which at least one of the two models projected better yield performance for both soils were considered as the most promising set of adaptations to be modelled by the ensemble. The selected options (in bold font in Table 1) were as follows: 1) removing vernalization requirements, hereafter referred to as a spring wheat cultivar (SW), 2) considering a moderate change in duration of phenological phases, i.e. from –10 to +10%, 3) considering a moderate advance of the sowing date by –15 days as

well as a delay of the sowing by 30 days, and 4) application of 40 mm of sl at flowering. These options are identified hereafter as follows: 10% shorter cultivar (WW-cv1 or SW-cv1), standard cultivar duration (WW-cv0 or SW-cv0), 10% longer cultivar (WW-cv2 or SW-cv2), advance of the sowing date by –15 days (287, DOY), standard sowing date (302), delay of the sowing by 30 days (332), application of 40 mm at flowering (sl).

3.3. Simulation phase

3.3.1. Impacts and single adaptation options

The median of the 30-year mean winter wheat yield ranged between ca. 3000 and ca. 7000 kg ha⁻¹ under perturbed P and T depending on soil and CO₂ level, as shown by the IRSs for standard management (Fig. 3). The ensemble median yield was projected to increase with P and decrease with T, as expected. The shape of the yield isolines indicates that T became the limiting factor at high P increments (isolines tended to be vertical), while P became the limiting factor at low T increments (isolines tended to be horizontal). Yield was lower for shallow than for deep soil, and increased with [CO₂]. Depending on the soil and CO₂ level, current yields were maintained up to 2–4 °C of T increase at highest P increases. When P decreased up to 20%, baseline yields were maintained only at high [CO₂] (522 ppm) together with small T increments of up to 2 °C (Fig. 3, green mask).

In general, for all the adaptations, the ARSs isolines did not change with elevated [CO₂], but soil depth increased yield sensitivity to precipitation. Accordingly, the perturbations for which the current yield was recovered depended on soil depth and, to a lesser extent, [CO₂]. Hereafter, results for shallow soil and 447 ppm of CO₂ concentration are shown (Figs. 4 to 8), as this would be the most unfavourable combination, while the other combinations are illustrated in suppl. Mat. (Figs. S2 to S22).

The first single adaptation tested was removal of the vernalization requirements (Fig. 4e). An adaptation response between 10% and 40% was found depending on the T and P perturbation. This adaptation allowed current yield (green mask) to be recovered or exceeded for T perturbations up to +5 °C if P strongly increased, and for T perturbations up to +3 °C when P decreased up to –30%.

Other single adaptations were changes in cycle duration, sowing date and application of sl, all of them for winter wheat. Changing cycle duration or sowing date alone did not show any potential to overcome the negative impact (Figs. S2, S5, S11 and S17 in suppl. mat.). Applying sl at flowering resulted in an increased yield of potentially more than 30% (Fig. 5e). Current yield was recovered even for the greatest decreases of P and up to ca. 2 °C of T increase (green mask).

Table 2

Simulated yield (kg ha⁻¹) for some combinations of adaptation options tested in the preliminary phase for $\Delta P = -15\%$, $\Delta T = +2\text{ }^\circ\text{C}$ and [CO₂] at 447 ppm. Simulated yields with no adaptations and under baseline climate are reported as reference. Simulations were run with the CERES-wheat and SiriusQuality2 (SQ2) models, and adaptation options comprised: removing vernalization requirements (SWcv0), shortening crop cycle by 10% (WWcv1-10), lengthening crop cycle by 10% and 20% (WWcv2 + 10 and WWcv2 + 20, respectively), and supplementary irrigation (sl). These were combined with three sowing dates: 15 days before the standard sowing date (–15), 15 and 30 days after the standard sowing date (+15 and +30, respectively). The standard sowing date was set to day of the year 302. Results are shown for shallow and deep soils.

Sowing date	No adaptation (360 ppm)		SWcv0		WWcv1-10		WWcv2 + 10		WWcv2 + 20		sl	
	CERES	SQ2	CERES	SQ2	CERES	SQ2	CERES	SQ2	CERES	SQ2	CERES	SQ2
<i>Shallow soil</i>												
0	3825	2820										
–15			3953	3482	3934	2103	3917	2256	3944	2301	3957	2761
+15			3244	2714	3152	2238	3267	2556	3241	2577	3318	3111
+30			2985	2943	2832	2388	3031	2590	2997	2591	3067	3062
<i>Deep soil</i>												
0	6654	5640										
–15			6927	6040	6608	5422	6375	5292	6164	5335	6875	5658
+15			6204	5715	5688	5585	5694	5620	5430	5557	6246	5856
+30			5837	5542	5202	5503	5508	5547	5201	5476	5940	5833

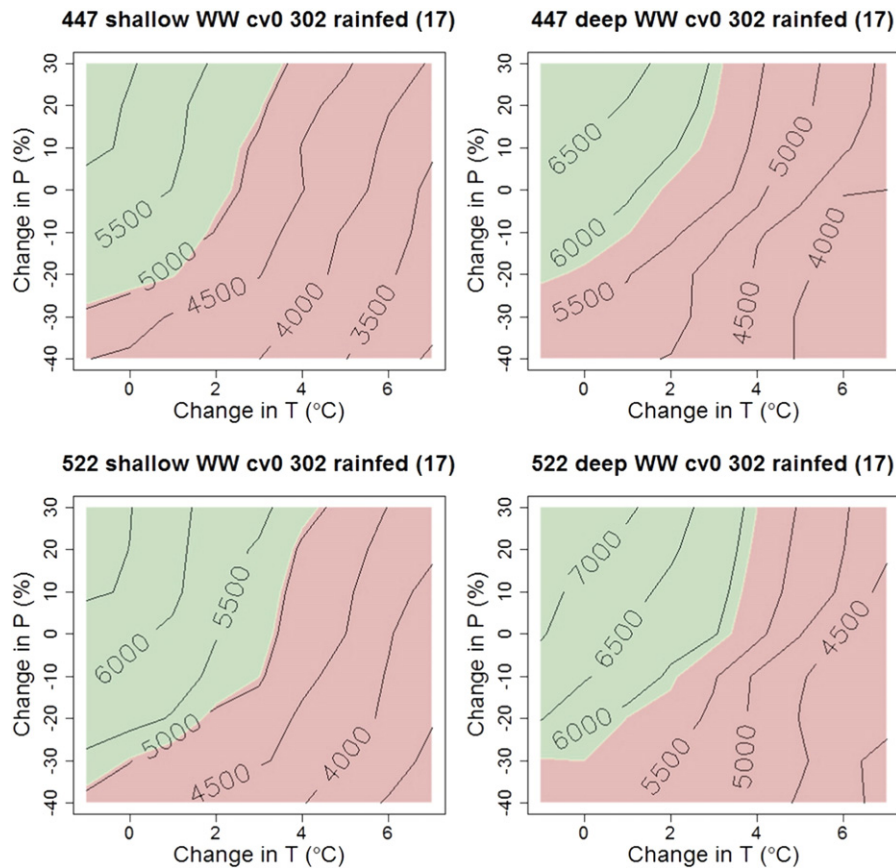


Fig. 3. Impact response surfaces (IRSs) of yield (kg ha^{-1}) for the standard cultivar (WW-cv0) under standard management, i.e. rainfed conditions and sowing date set to 302 (DOY). Results are shown for both shallow (left plots) and deep (right plots) soils, and for 447 ppm (upper plots) and 522 ppm (bottom plots) of $[\text{CO}_2]$. The number of members in the ensemble is indicated in brackets. Green (red) mask indicates absolute yield values greater (lower) than the reference yield (i.e. 30-year mean yield without adaptation under 360 ppm).

3.3.2. Combined adaptation options

The first set of combinations tested was based on rainfed SW, (i.e. with no vernalization requirements) (Fig. 4). The maximum adaptation response for rainfed SW was +60% (Fig. 4d). In general, the earlier the sowing date, the higher the adaptation response. The adaptation response to changes of cycle length was more variable. As an example, the shorter cultivar (SW-cv1) was more useful for compensating for the T effect, while the longer cultivar (SW-cv2) was more useful for dealing with the P effect, especially for the earlier sowing date (e.g. Fig. 4a vs. Fig. 4g). When P decreased, only a few combined adaptation options allowed recovery of current yield and only for small increases in T: generally not higher than 2 °C. At higher T increases, maintaining current yield (recovery response) was only possible for the highest P increases.

Then, when sl at flowering was considered, maximum adaptation response for SW was +80% (Fig. 6d). Also in this case, the earlier the sowing date the higher the adaptation response. The longer cultivar cycle (SW-cv2) provided a better adaptation response than the shorter one (SW-cv1). Although the standard cycle duration (SW-cv0) showed a strong adaptation response, only the longer cycle (SW-cv2) allowed recovery of current yields whatever the P and T perturbation (green mask, Fig. 6). The sl also revealed some adaptation potential for winter wheat (WW) (Fig. 5). Maximum adaptation response for the combinations based on sl and winter wheat was up to +30%. The standard cultivar (WW-cv0) showed an adaptation response whatever the sowing date, while the shorter cultivar (WW-cv1) only provided an adaptation response with earlier and standard sowing dates. For these cases, the higher the P decrease, the higher the adaptation response. Under these conditions the adaptation allowed maintenance of current yields

(recovery response) up to 3 °C of T increase. The longer cultivar (WW-cv2) did not present a useful adaptation response.

To explore the full potential of adaptations, and to provide a reference for identifying yield ceilings, the adaptations were simulated under full irrigation (fl) (Fig. 7). Maximum adaptation response for the combinations based on fl and SW was +130%. Both SW-cv0 and SW-cv2 obtained very high adaptation response whatever the sowing date. SW-cv1 was the only one unable to recover current yield when T increased more than ca. 4 °C. Generally, the adaptation effect was similar whatever the T increase (isolines became almost horizontal) for the highest P decreases. Vertical shape of the green to red mask transition, indicates that simulated yield responses under full irrigation are to temperature alone (as expected).

3.3.3. Recommended adaptation options

Figs. 4 to 7 have been shown to illustrate the response of the ensembles of crop models to different adaptation options (complete set of plots in Suppl. mat.). To identify the options with high adaptation potential (high increment on crop yield when the adaptation was applied), all possible combinations for both 447 ppm and 522 ppm of $[\text{CO}_2]$ for both shallow and deep soils, were analysed. We focused our attention on perturbations including P decrease. Thus, five perturbation regions were defined: 1) ΔP from 0 up to -20% and ΔT up to +1.5 °C, 2) ΔP from 0 up to -20% and ΔT from 1.5 to +3 °C, 3) ΔP from -20% up to -40% and ΔT up to +1.5 °C, 4) ΔP from -20% up to -40% and ΔT from 1.5 to +3 °C, and 5) ΔP from 0 up to -40% and ΔT higher than 3 °C. Recommendations refer to these perturbation regions, and they are summarized in Table 3.

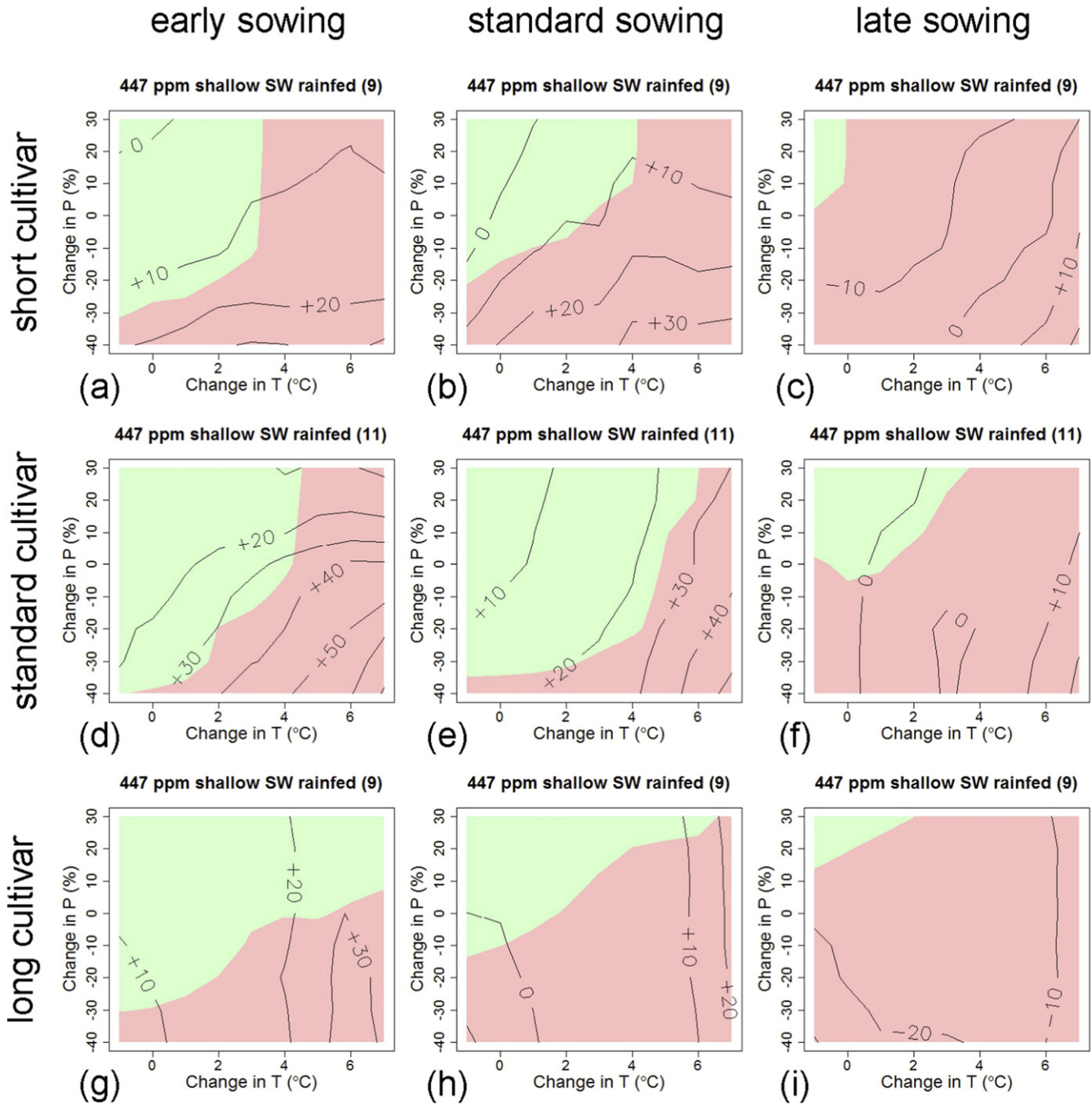


Fig. 4. Adaptation response surfaces (ARSs) of yield changes (%) from yields without adaptation (i.e. same soil and [CO₂], but for the standard cultivar and management). All ARSs are for spring wheat (SW)-based adaptation options under rainfed conditions in shallow soil and for 447 ppm of [CO₂], combined with short (SW-cv1; a, b, c), standard (SW-cv0; d, e, f) and long (SW-cv2; g, h, i) cultivars, and with early (287, DOY, a, d, g), standard (302, b, e, h) and late (332, c, f, i) sowing dates. Therefore, a total of 9 combined adaptations is shown. Number of members in the ensemble is in brackets. Green (red) mask indicates absolute yield values (kg ha⁻¹) of the evaluated adaptation greater (lower) than the reference yield) (i.e. 30-year mean yield without adaptation, WW-cv0, sown at 302, rainfed, under 360 ppm).

Rainfed WW-based combinations should be avoided in both soils as adaptation options produced recovery responses for only few of the perturbations. Winter wheat is only recommended when combined with sl. Under this management, the standard cultivar is recommended with a wide range of sowing dates, while shorter cycle cultivars should be sown earlier. These options would show adaptation response up to severe P, T perturbations (for all T perturbations under P decline), while recovery response would be possible up to severe P perturbations and moderate T changes (regions 1 and 3). Longer cultivars should be avoided in both soils, even with sl.

Spring wheat was successful under a wide variety of conditions. Earlier sowing is recommended for both soils, and later sowing dates should be avoided. Standard sowing date can also be used when sl is applied, and with the standard cultivar under rainfed management. All cultivars can be used in the shallow soil, while the shorter cultivar is not recommended in the deep soil. Most of these options show adaptation response up to severe P, T perturbations (for all T perturbations under P decline), especially for the shallow soil, while recovery response was mainly shown by standard and longer cultivars.

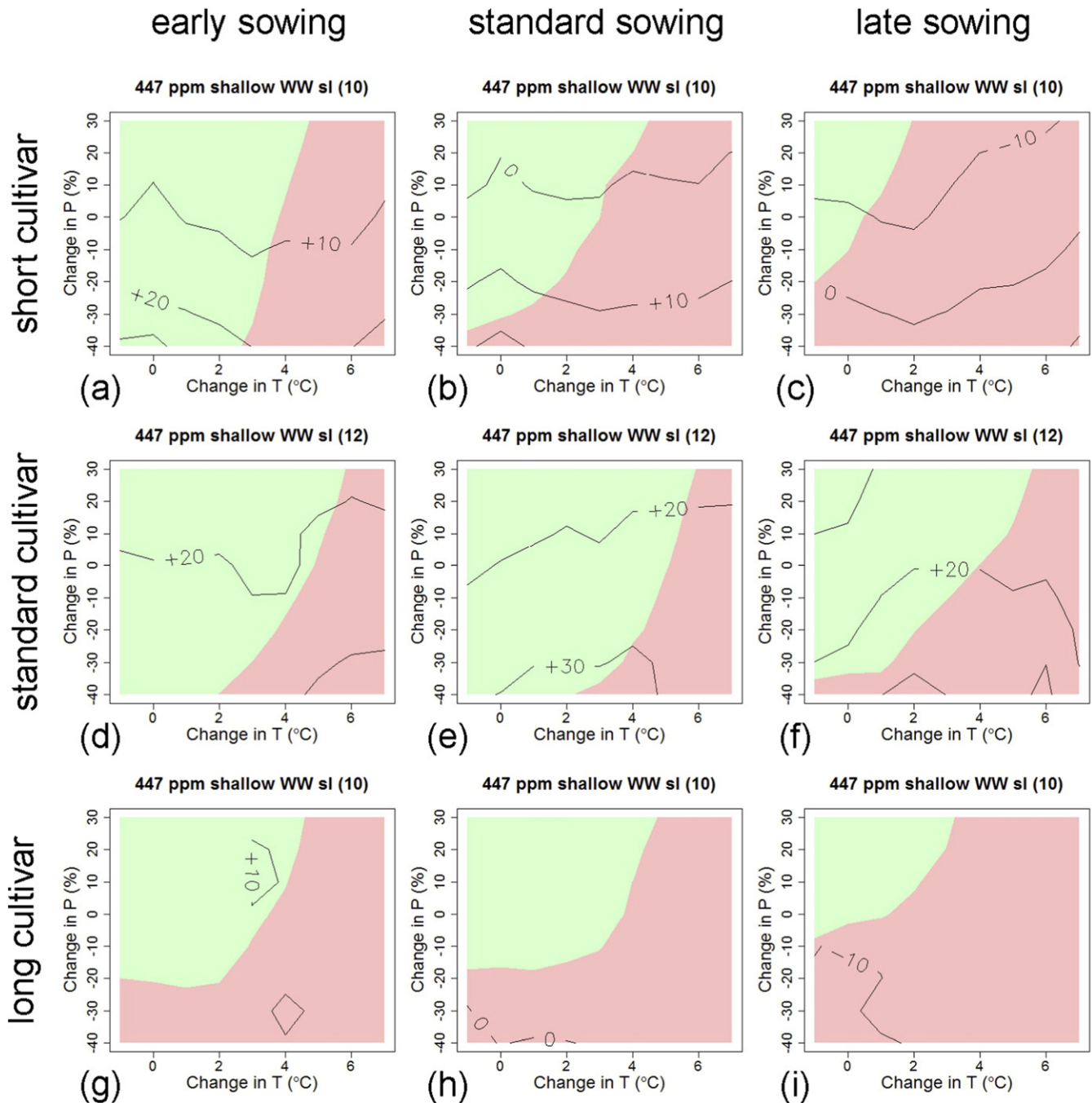


Fig. 5. As Fig. 4 but for winter wheat (WW) and supplementary irrigation (sl).

The recommended options described above did not depend on $[\text{CO}_2]$.

3.3.4. Interannual and intermodel variability

This analysis was aimed at checking the interannual and intermodel variability values of the recommended options against former studies, so only the main trends are commented on here. To illustrate these checks, three of the recommended adaptation options are shown with colour masks representing the interannual CV and the IQR of model spread (Fig. 8). CV (no more than 40% in these examples) was in line with that shown in Pirttioja et al. (2015) for Lleida. Concerning intermodel variability, there is no clear pattern of the IQR of absolute yields with respect to changes in P and T. This could be related to the complexity of single model responses.

4. Discussion

In this paper we used, for the first time to our knowledge, an ensemble of crop models for investigating potential adaptation strategies for wheat. For that purpose, we introduced the AOCC and the ARS concepts for ensemble building and analysis respectively. Below, we discuss all these aspects in depth.

4.1. Calibration

The calibration process allowed the modellers to identify one of the possible set of crop model parameters values matching a limited set of field observations. This combination consists of several parameters that could compensate for approximations elsewhere (Wallach et al., 2014); therefore, we could have several combinations of parameters

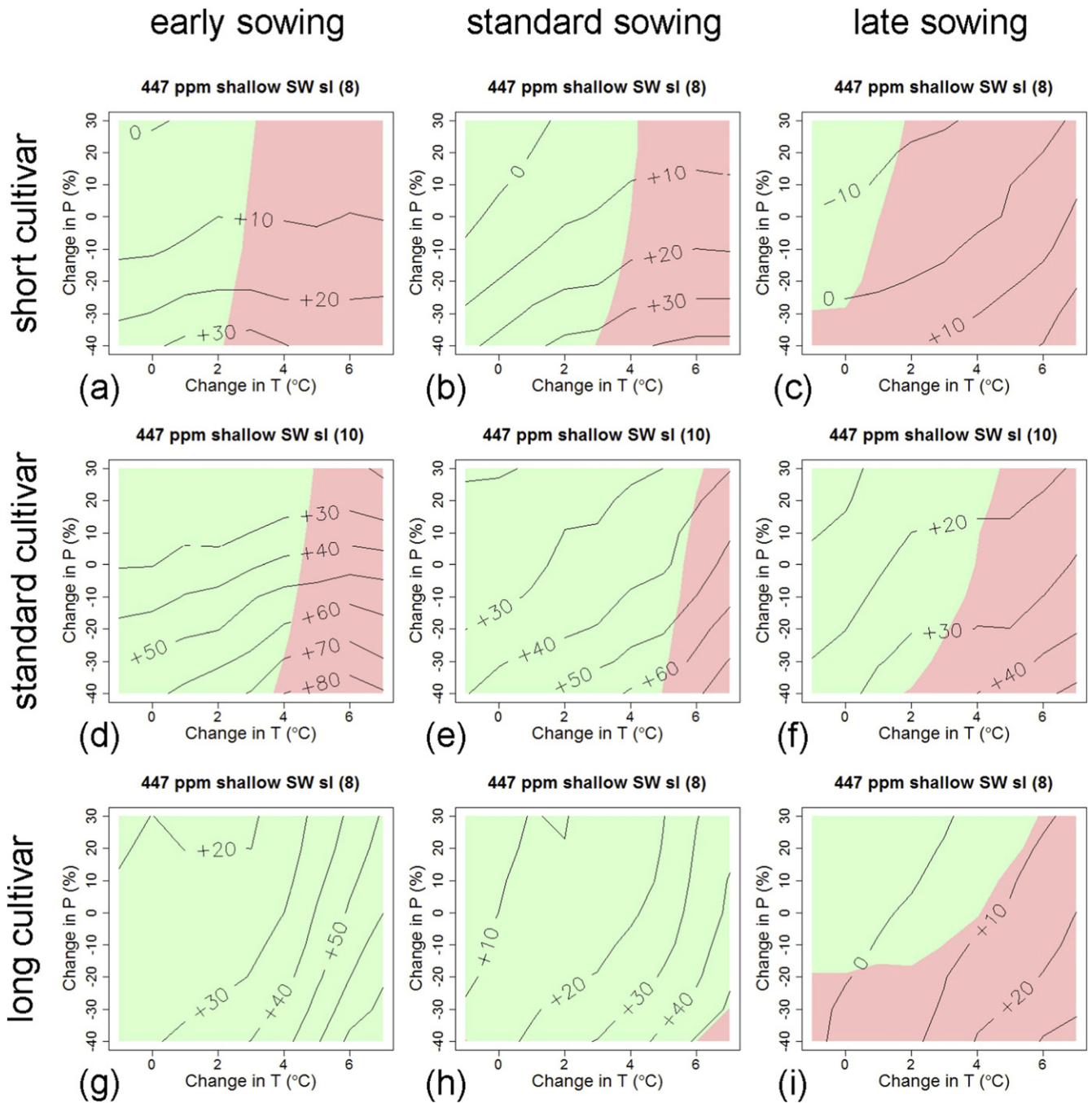


Fig. 6. As Fig. 4 but for supplementary irrigation (sl).

matching observations, particularly when calibration data set is limited (equifinality). Due to general overparameterization of the crop simulation models (Wallach et al., 2014) and uncertainties related to experimental data, in some cases, relatively good calibration performance still did not guarantee reasonable response simulations. This could in our case be due to the limited set of field data used for calibration, mainly based on yield, flowering and maturity dates from two years, which is a typical level of information available for crop model applications. However, adaptation is needed beyond locations where we have detailed, high quality data sets, and this study aims to deal with situation while new data can be generated or accessed. To overcome these limitations several recommendations should be considered before model application: 1) more than one variable (and in-season variables if possible) should be taken into account in the calibration evaluation (Challinor et al., 2014; Kersebaum et al., 2015; Martre et al., 2015), 2)

an ensemble of models should be used, because doing so could compensate for individual models' deficiencies if the ensemble is not too small (Martre et al., 2015), and 3) the AOCC concept is also needed for checking vector quality and biological coherence. In this study we have adhered to all of these recommendations.

4.2. The adaptations

Our analysis demonstrated that “business-as-usual”, i.e. WW with standard cultivar and management, should be avoided in the future. The adaptations tested here constituted measures that can already be applied by farmers, as the main purpose of this study was to support and provide guidelines for implementing adaptation. Our preliminary phase and the existing literature allowed us to establish that a wide

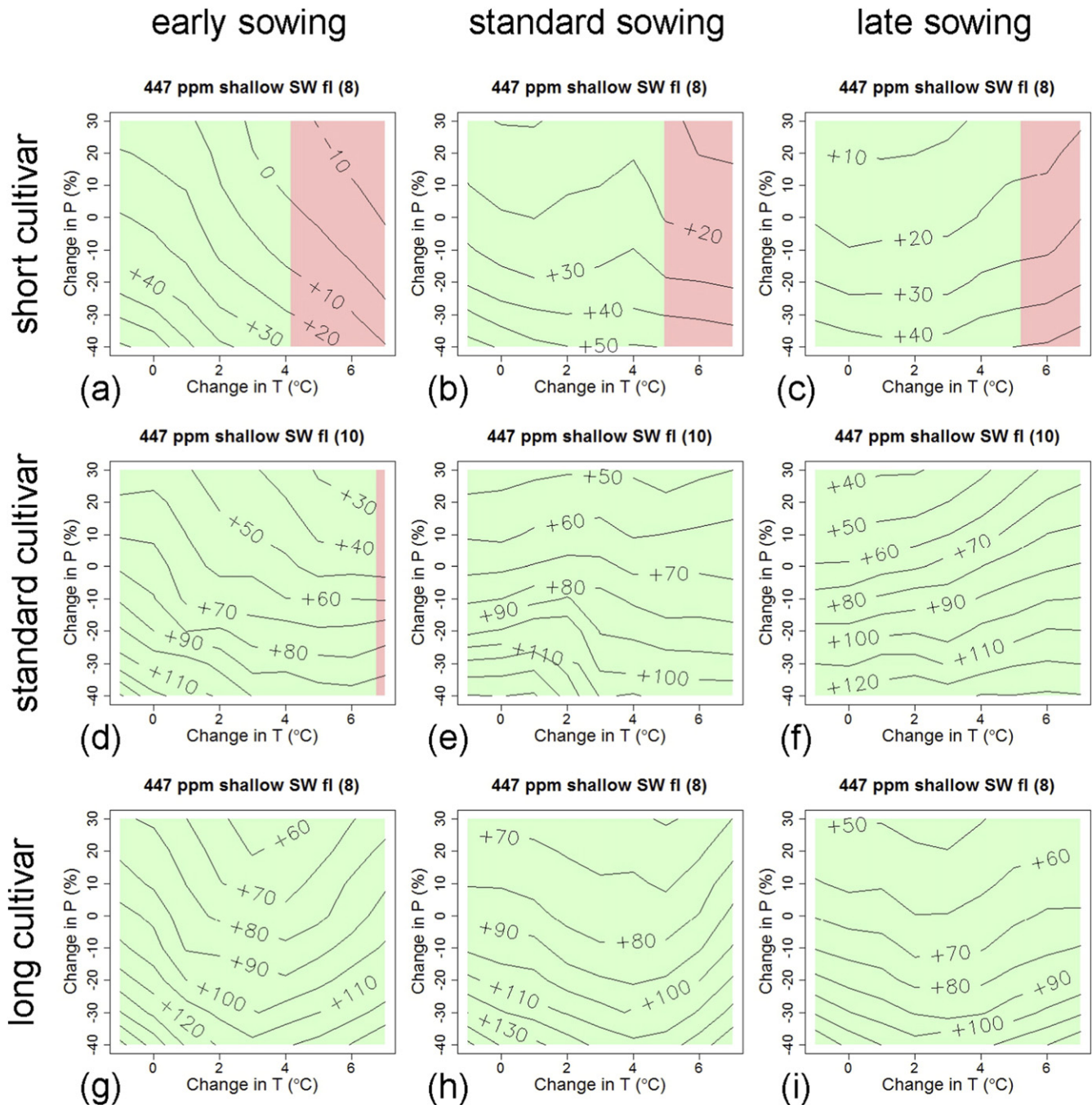


Fig. 7. As Fig. 4 but for full irrigation (fl).

range of phenology x sowing date combinations should be the starting point of any adaptation study in this Mediterranean environment.

Concerning single adaptation options, no single factor was able to overcome the detrimental effect of the complex interactions imposed by the P, T and [CO₂] perturbations. The sole exception was sl, which proved to be useful even under the most severe perturbations, probably due to the high dependency of Mediterranean rainfed cropping systems on precipitation (Oweis et al., 1998). This could be also true for current conditions. The analysed location, Lleida, is located in northeastern Spain and WW is currently grown; however, our results indicated that the latitudinal threshold of WW suitability would move towards higher latitudes under a warmer climate, in agreement with Mínguez et al. (2007). The T increases affected WW vernalization, giving an advantage to SW. Concerning the less clear response of yield to changes in crop cycle duration, we hypothesized that the main cause was the

uncertainty related to the modelling of phenology. This uncertainty comes mainly from the different descriptions and simulations of crop phenology across crop models. For example, some models consider a simple accumulation of thermal time, while others take into account daylength and vernalization. Other models consider the number of leaves produced and the phyllochron or include the effect of water and nutrient stress. To test this hypothesis, we conducted a deeper analysis of the implementation of the phenological options by the ensemble members, finding that the spread of the model response (the ensemble spread of the simulated phenological dates in days) partially masked the perturbations of the cycle duration imposed. This spread was comparable to the number of days that corresponded to a 10% change of the crop cycle. We would need to simulate a larger phenological perturbation to overcome this effect. Such a simulation should be considered in future multi-model ensemble modelling exercises.

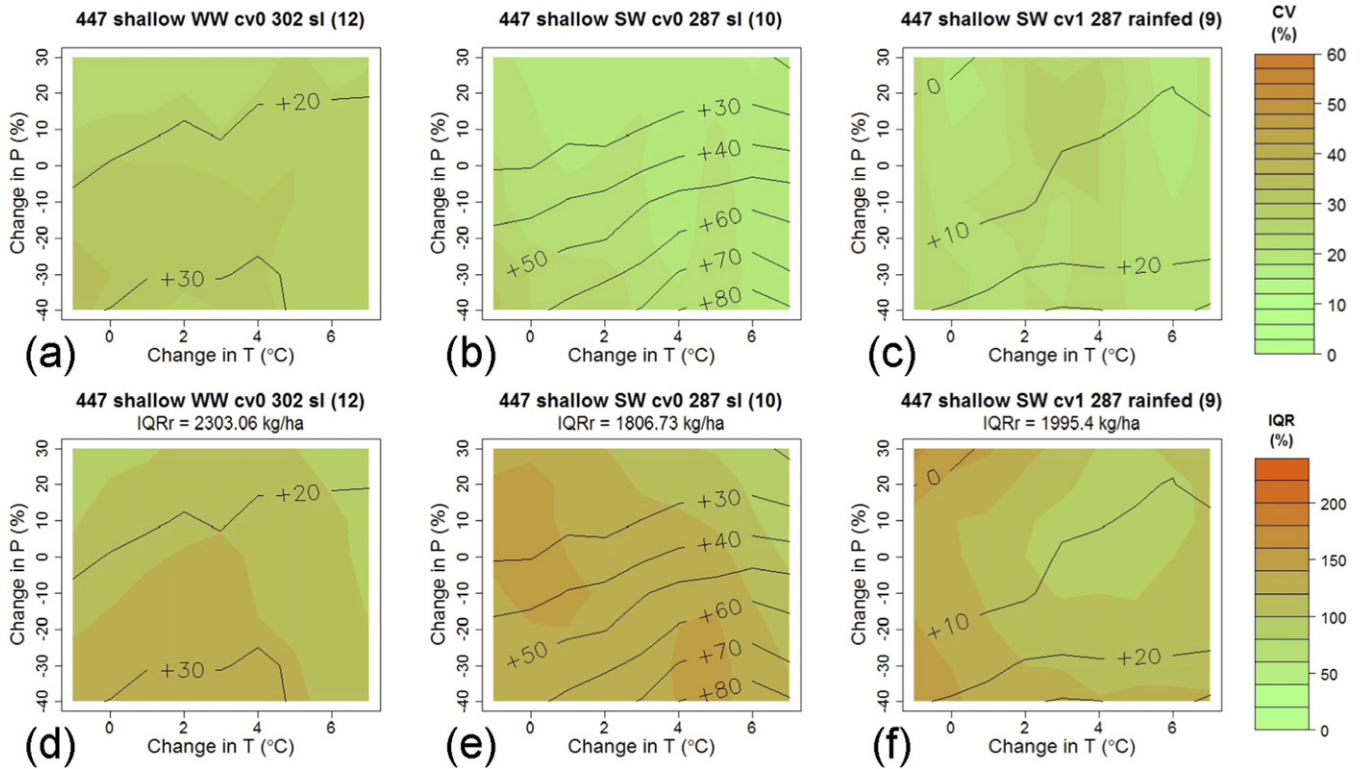


Fig. 8. Adaptation response surfaces (ARs) of yield change (%) for three adaptation options, comprising winter (WW) and spring (SW) wheat, shorter (cv1) and standard (cv0) cultivars, earlier (DOY = 287) and standard (DOY = 302) sowing dates, and rainfed and supplementary irrigation (sl) conditions for shallow soil and for 447 ppm of CO₂. Number of members in the ensemble is in brackets. Green to orange mask indicates: ensemble median of the 30-year interannual coefficient of variation (CV, %) (a, b, c); and ensemble interquartile range (IQR) (d, e, f) of absolute yields of the evaluated adaptation. IQR is expressed as percentage of the IQR for same soil, CO₂ level, standard cultivar and management for no T, P perturbations (IQRr), so values lower than 100% mean a reduction in the IQR.

When combined adaptation options were considered, our results demonstrated that effective adaptation is possible at Lleida. Moreover, a wide scope of adaptation is opened up. Both WW and SW provided some adaptation response under certain combinations of cycle duration and sowing date. Early sowing emerged as a good option when combined with other options, confirming the findings by [Giannakopoulos et al. \(2009\)](#) and [Moriondo et al. \(2010\)](#), who tested the individual effect of this adaptation. This response could be influenced by the initial soil water content (in this exercise, set to 75% of field capacity). So prudence is needed when implementing this recommendation, as for lower initial soil water contents (e.g. for years with an extended dry period lasting into autumn, frequent under the weather conditions of southern Spain) a lower or even a worst response could be obtained by the earlier sowing.

In spite of the uncertainty on the implementation of phenology-based adaptations reported above, some conclusions about the duration of the growth cycle could nevertheless be extracted. Even when in principle adaptations based on longer-duration cultivars may appear advantageous with increasing temperatures ([Giannakopoulos et al., 2009](#)), our results showed that this might not always be the case for Lleida. This is probably because the expected yield gains due to the longer cycle would be offset by grain filling happening in a warmer period. Additionally, extreme high temperatures could impose an absolute constraint on wheat viability at the end of the cycle if longer-duration cultivars were to be considered as an adaptation option ([Trnka et al., 2014](#); [Pirttioja et al., 2015](#)). Supplementary irrigation failed to overcome these effects for winter wheat. For SW, soil type made a difference for the adaptation potential of longer-duration cultivars, enabling the longer-cycle crop to take advantage of the deep soil because of its higher water holding capacity.

Supplementary irrigation at flowering was found to present potential for successful adaptation under many combinations of phenology

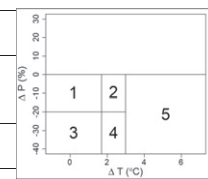
and sowing date. The use of sl at key phenological stages is in line with the current evolution already observed in the Mediterranean: e.g. in Andalusia the area under irrigation has been extended in recent decades, with no increase in total irrigation water used resulting from the promotion of deficit irrigation strategies ([Lorite et al., 2012](#)). Likewise, in southern Italy irrigated areas are increasing as a result of the use of innovative irrigation systems ([Acutis and Ventrella, 2015](#)). Supplementary irrigation was not just useful for raising crop yields but in some cases also stabilized the effect of the adaptation when T increased (e.g. [Fig. 6d](#)), as previously also reported [Oweis et al. \(1998\)](#). When sl is applied, the choice of cycle duration can be used to stabilize adaptation effect against changing P (as cv2, [Fig. 6g](#)) or T (as cv1, [Fig. 6a](#)). It is important to emphasize that several rainfed options showed adaptation potential for specific combinations of sowing date and phenology (see examples in [Table 3](#)). Full irrigation (fl) showed the greatest potential for supporting successful adaptation when water requirements were fulfilled, as all the T increases considered could be coped with by the fl option; this is in agreement with previous studies ([Mínguez et al., 2007](#); [Ruiz-Ramos et al., 2011](#)).

In general, inter-annual variability was not affected by the adaptations, and median values of CV were consistent with those found for Lleida in [Pirttioja et al. \(2015\)](#). This is because, by using the delta change approach for the sensitivity analysis combined with a seasonal pattern, we changed the intra-annual variability of climate but not the inter-annual variability. Therefore, the results may not necessarily hold true if the future interannual variability of the climatic variables are greatly different from those employed. A separate study would be necessary to analyse this more comprehensively.

The effect of adaptation options on ensemble inter-model variability was evaluated by comparing the IQR for the adapted conditions to that for the unadapted baseline. Lleida cropping systems have been reported to be very sensitive to precipitation ([Ruiz-Ramos and Mínguez, 2010](#)),

Table 3
 Perturbation regions for which specific options show positive adaptation and/or recovery responses, and derived recommendations. Perturbation regions are defined as follows (see embedded scheme in the table): region 1 is defined as ΔP from 0 up to -20% and ΔT up to $+1.5^\circ\text{C}$, region 2 is defined as ΔP from 0 up to -20% and ΔT from 1.5 to $+3^\circ\text{C}$, region 3 is defined as ΔP from -20% up to -40% and ΔT up to $+1.5^\circ\text{C}$, region 4 is defined as ΔP from -20% up to -40% and ΔT from 1.5 to $+3^\circ\text{C}$, and region 5 is defined as ΔP from 0 up to -40% and ΔT higher than 3°C .

	Water mgnt	Cultivar	Sowing date	Shallow		Deep		Recommendations
				Adaptation	Recovery	Adaptation	Recovery	
WW	R	None	None	None	–	None	–	To be avoided
	sl	cv1	Earlier	1,2,3,4,5	1,2,3,4	1,2,3,4,5	1,2,3,4	Only under sl and -Standard cultivar whatever the sowing date -Shorter cultivars own earlier -Same recommendations for both soils
		cv0	All	1,2,3,4,5	1	1,2,3,4,5	1,2,3	
		cv2	None	None	–	None	–	
SW	R	cv1	Earlier	1,2,3,4,5	1	3,4	–	Avoiding latest sowing Advancing sowing whatever the cultivar Standard sowing with sl, and with standard cultivar under rainfed conditions Shorter cultivar not recommended on deep soil
			Standard	2,3,4,5	None	2,4,5	None	
		cv0	Standard & earlier	1,2,3,4,5	1,2	1,2,3,4,5	1	
		cv2	Earlier	1,2,3,4,5	1	1,2,3,4,5	1,2,3	
	sl	cv1	Earlier	1,2,3,4,5	1,3	None	–	
			Standard	2,3,4,5	2,3	Not recommended	–	
		cv0	Standard & earlier	1,2,3,4,5	1,2,3,4	1,2,3,4,5	1,2,3,4	
		cv2	Standard & earlier	1,2,3,4,5	1,2,3,4,5	1,2,3,4,5	1,2,3,4,5	



Abbreviations: rainfed (R), supplementary irrigation (sl), shorter, standard and longer crop cycle duration (cv1, cv0 and cv2 respectively), cultivar without vernalization requirements (SW), winter wheat (WW), management (mgnt), not considered because of lack of adaptation response (–).

so it is not surprising that the uncertainty response from IQR analysis proved to be very complex. A separate study would be necessary to analyse this more comprehensively.

Finally, adaptation has to be tailored to local conditions (Reidsma et al., 2015). Our results demonstrated that, even for a single location, implementing an effective adaptation strategy depends on expected perturbation intensity and local environment (e.g. soil type), which is in agreement with Iglesias et al. (2010). As a consequence of this, we cannot rely on just one single or combined adaptation option for a whole region or for all periods into the future in order to be flexible to changing conditions over time; moreover, the resultant recommendations for adapting wheat might not be valid for other environments or periods. However, many of the results found here could still be valid for regions with decreased precipitation under climate change.

4.3. Ensemble building

Concerning ensemble building, some factors must be considered. First, we are aware that the ensemble size was different for several adaptations. Based on this, in drawing our conclusions we have avoided comparing options modelled from different ensembles; for such cases, the adaptations were evaluated separately. However, the number of members of the ensemble was always between 8 and 14, which matches the range of minimum ensemble sizes recommended in recent international crop model intercomparison exercises (e.g. Martre et al., 2015). Nevertheless, the robustness of an ensemble can be severely compromised if it includes model outcomes that are obviously implausible (see e.g. Rötter et al., 2012). Therefore (and secondly), for the decision on which models belong to the final ensemble, we have formalized here the concept of AOCK. This we have applied as an *ex post* plausibility check, using it to screen out crop responses judged to be unrealistic based on our current knowledge, rather than just

checking whether technical errors may have occurred. Similar criteria of plausibility were defined using IRSs for an ensemble of models of climate change impact on permafrost by Fronzek et al. (2011). We are perfectly aware that empirical models should not be used outside the boundary conditions for which they were developed. Similarly, mechanistic models might provide unreliable outputs when applied in conditions that largely exceed the limits for which the model formalisms were developed; also, model calibrations are limited in their spatial and temporal coverage. A generalization of the AOCK concept or other equivalent procedures should be introduced for comprehensively interpreting model results in all kinds of modelling studies.

The AOCK concept is needed when our understanding of the processes involved is significantly lacking, or when high levels of uncertainty are inherent in a particular system analysis or simulation. In these situations, a combination of sound knowledge in comparable conditions, observed data and common sense can be used to check whether simulation results appear to be reasonable. For this particular study, the concept was applied through several criteria based on the following general principles: 1) plausible absolute values, 2) plausible ecophysiological responses to inputs (weather, soil and management), and 3) thresholds set sufficiently wide to include uncertainty in empirical databases, model parameters and structure, while excluding evidently implausible responses. These principles can be transferred to other modelling studies and for other crops and environmental conditions. It should be borne in mind that the AOCK concept always includes some degree of arbitrariness, in particular in the selection of the specific criteria to be met and the threshold values selected. The consequence of an excess of arbitrariness or inappropriate criteria definition (e.g. by setting too narrow thresholds) could be the exclusion of plausible but not very likely responses that could become more likely under unknown conditions (e.g. extreme divergent future conditions). To reduce this arbitrariness, the criteria used should be well-argued and transparently

documented to allow further refinement, and sensitivity of results to AOCK application and to the specific criteria definition should be analysed.

5. Conclusions

Three novelties were presented in this paper: 1) the use of an ensemble of crop models for supporting adaptation; 2) ARSs, which have proved to be a useful tool for planning adaptations under highly uncertain conditions; and 3) the formalization of the AOCK concept, which should be a driver for comprehensively interpreting model results in all modelling studies. Based on these, we conclude that effective adaptation at Lleida is possible. When locally addressed, the scope for adaptation widens and several solutions might arise depending on the availability of irrigation. Thus, according to our results, there are few adaptation options that could work under rainfed conditions in Lleida. However, one single sI event was enough to create adaptation potential for many adaptation options based on simple changes of phenology and management. The study of recommendations for adaptation prioritized combined options based on SW, standard (current) and shorter cycle duration and early sowing date with sI. Many of these adaptations allowed maintenance of current yields for moderate T and P perturbations, some of them for severe perturbations as well.

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Appendix A. Supplementary data

Supplementary data to this article can be found online at <http://dx.doi.org/10.1016/j.agry.2017.01.009>.

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