

Inter-model variability in wheat yield responses to changes in climate in the IRS1 model experiment

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Session C4 “CropM Uncertainty and Risk analysis”

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Outline

- Introduction and aims
- Material and methods: sensitivity analysis with ensemble of wheat models
- Results I: Ensemble averages and ranges (published in Clim Res paper)
- Results II: Classification of response (preliminary results)
- Conclusions and outlook

Aims

- Crop modelling experiment in MACSUR/CropM/WP4

Aims:

- To study crop model sensitivity to changes in precipitation and temperature using a large ensemble of crop models across a transect
- To quantify differences in winter and spring wheat yield responses to changed climate across models
- By plotting results of the sensitivity analysis as impact response surfaces (IRSs)

MATERIAL AND METHODS

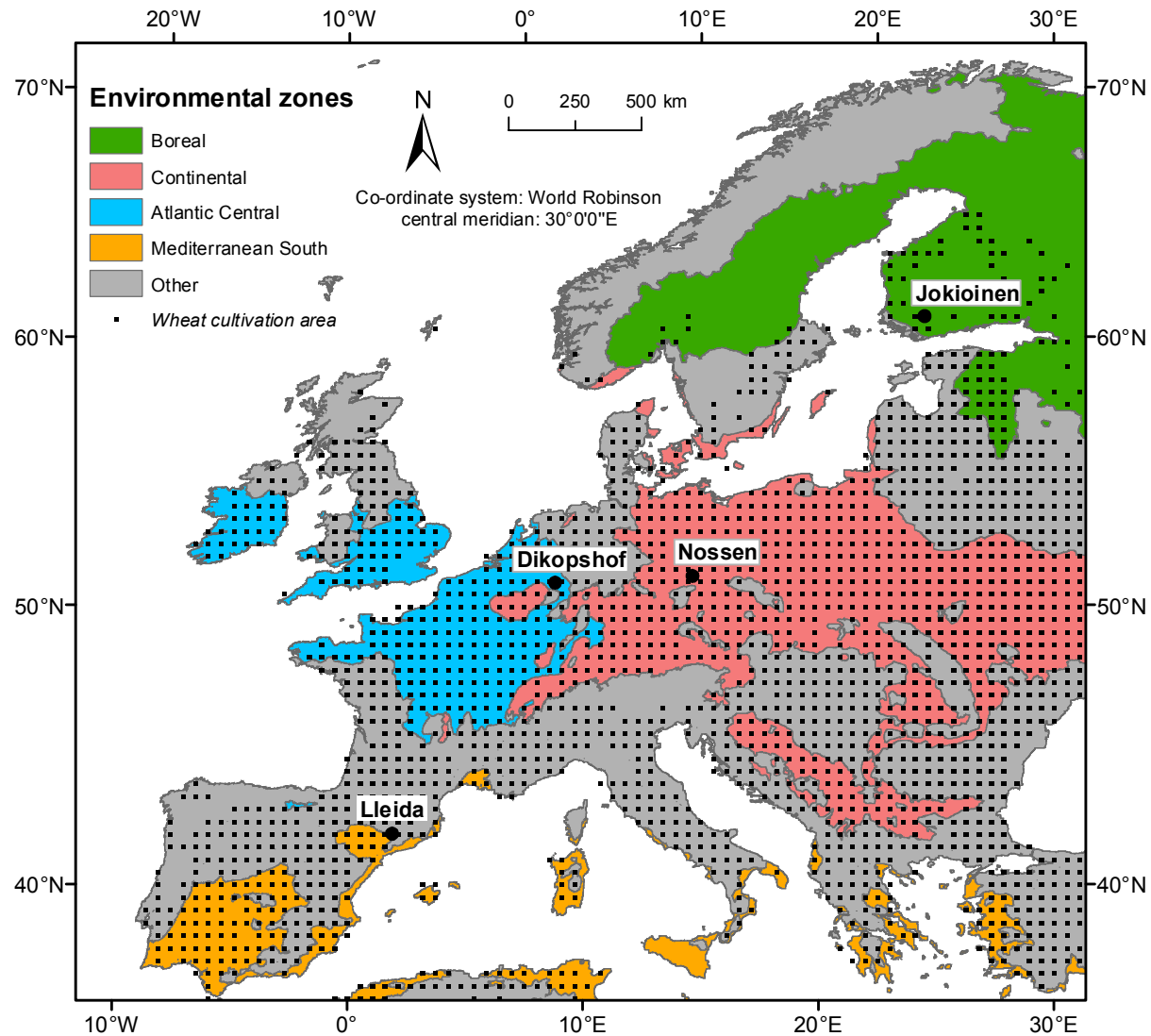
Ensemble of 26 wheat models

Model	Modelling groups		
	Contact person(s)	Institute	Country
AFRCWHEAT2	Manuel Montesino	University of Copenhagen	Denmark
APSIM-Nwheat	Senthold Asseng, Davide Cammarano	University of Florida	USA
APSIM-Wheat	Enli Wang	CSIRO Land and Water	Australia
AquaCrop	Ignacio Lorite	IFAPA Junta de Andalucia	Spain
ARMOSA	Alessia Perego	University of Milan	Italy
CARAIB Crop	Julien Minet	Université de Liège	Belgium
CERES-wheat DSSAT v.4.6	Mirek Trnka, Petr Hlavinka	Mendel University in Brno	Czech Republic
CERES-wheat DSSAT v.4.5	Margarita Ruiz-Ramos	Universidad Politecnica de Madrid	Spain
CERES-wheat DSSAT v.4.5	Paola Deligios	University of Sassari	Italy
CropSyst	Marco Moriondo, Roberto Ferrise, Marco Bindi	CNR-IBIMET University of Florence	Italy Italy
DNDC	Cezary Slawinski; Piotr Baranowski	Polish Academy of Sciences	Poland
Fasset	Isk Öztürk	Aarhus University	Denmark
HERMES	Chris Kollas, Christian Kersebaum	Leibniz Centre for Agric. Landscape Research (ZALF)	Germany
Lintul4	Iwan Supit	Wageningen University	Netherlands
LPJ-GUESS	Per Bodin	Lund University	Sweden
LPJml	Christoph Müller	Potsdam Institute for Climate Impact Research	Germany
MCWLA	Fulu Tao	Luke Natural Resources Institute Finland	Finland
MONICA V1.2	Claas Nendel	Leibniz Centre for Agric. Landscape Research (ZALF)	Germany
SALUS	Bruno Basso	Michigan State University	USA
SIMPLACE<Lintul2, Slim>	Holger Hoffmann, Thomas Gaiser, Frank Ewert	University of Bonn	Germany
Sirius 2010	Mikhail Semenov, Pierre Stratonovitch	Rothamsted Research	UK
Sirius Quality	Roberto Ferrise, Marco Bindi	University of Florence	Italy
SPACSYS	Lianhai Wu	Rothamsted Research	UK
STICS	Benjamin Dumont, Françoise Ruget, Samuel Buis	Université de Liège & INRA EMMAH	Belgium & France
WOFOST 7.1	Cezary Slawinski; Jaromir Krzyszczak	Polish Academy of Sciences	Poland
WOFOST 7.1	Taru Palosuo, Reimund Rötter	Luke Natural Resources Institute Finland	Finland



Study sites across a European transect

Locations of weather stations used in this study and environmental zones of Metzger et al. (2005)



Mainly
temperature
limited

High
current
suitability

Mainly
precipitation
limited

Simulation set-up (1/2)

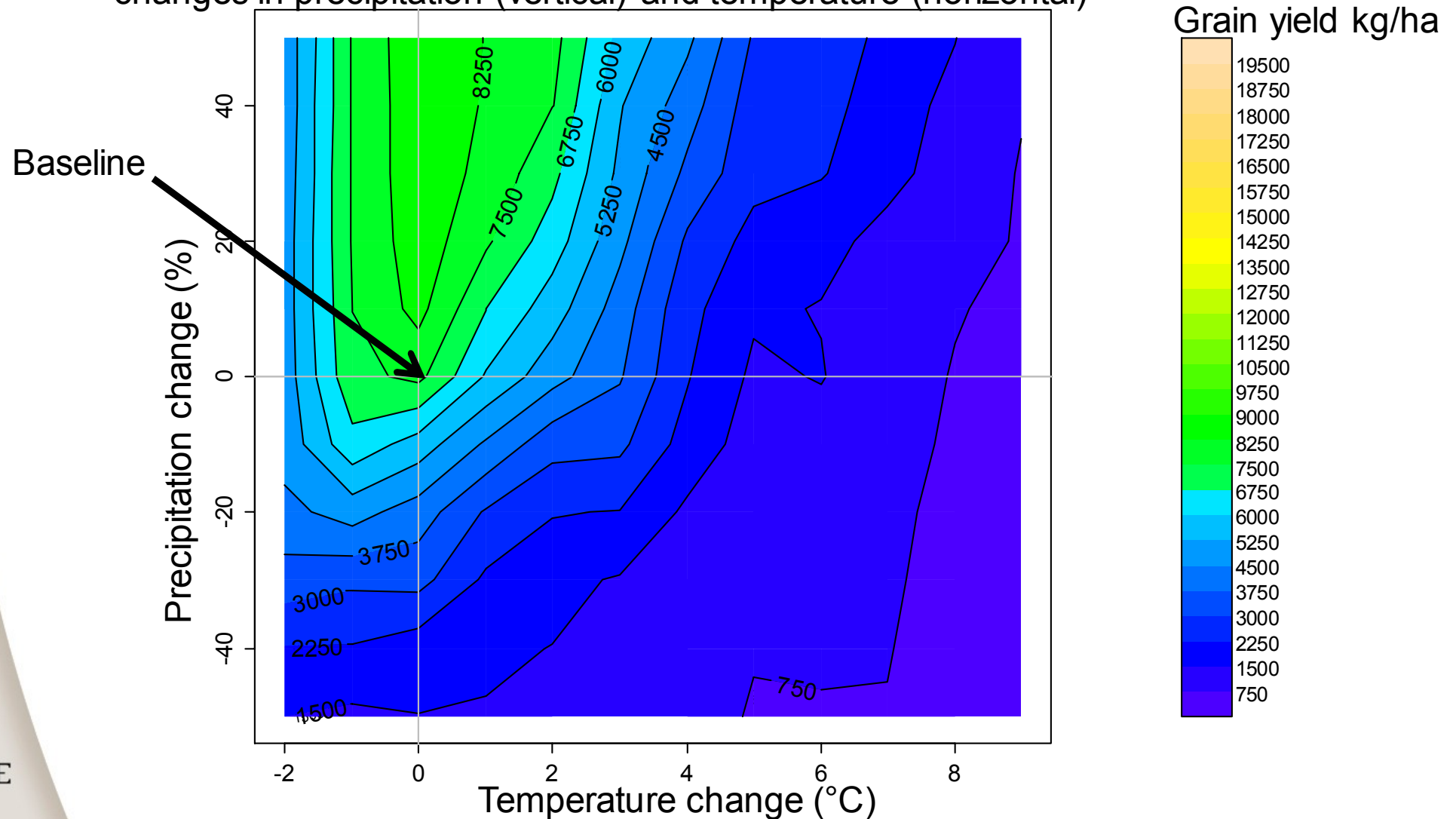
- Each group calibrated their model independently
- Limited data for calibration was provided on:
 - crop phenology and yield
 - soil conditions
 - fertilisation, tillage and irrigation (Spain) where available
- Model simulations were performed
 - on a daily time-step, for water-limited yields
 - assuming optimal nutrients
 - as a succession of independent years (no carry-over effects)
 - for modelled harvest dates up to a local "harvest cutoff"
- Error checking and model iteration
- Several output variables: annual grain yield, biomass, phenology, cumulated water use, nitrogen content of yield

Simulation set-up (2/2)

Sites	Country	Location	N		
	Finland	Jokioinen			
	Germany	Dikopshof (winter wheat), Nossen (spring wh.)	3		
	Spain	Lleida			
Crops	Crop /Cultivar type	Cultivar			
	Spring wheat	Different cultivar for each location			
	Winter wheat	Different cultivar for each location	2		
Baseline	Harvest years	1981-2010	30		
Perturbations	Variable	Min	Max	Interval	
	Precipitation (%)	- 50	+ 50	10	11
	Temperature (°C)	- 2	+ 9	1	12
CO₂ level	360 ppm (Year 1995)				1
Soils	Clay loam				1
Management	Fixed sowing date	Location specific (observed)			1
Total number of simulations	Sites x crops x years x P-changes x T-changes				23760

Impact response surface (IRS) of a single crop model for spring wheat yield, Germany, 2008

IRSs represent the sensitivity of modelled crop yield to incremental changes in precipitation (vertical) and temperature (horizontal)





Temperature and precipitation effects on wheat yield across a European transect: a crop model ensemble analysis using impact response surfaces

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P. Stratonovitch²⁵, I. Supit²⁶, K. Waha^{21,27}, E. Wang²⁸, L. Wu²⁹, Z. Zhao^{28,30}, R. P. Rötter⁴

RESULTS I: FOCUS ON ENSEMBLE AVERAGES AND RANGES

Baseline 1981-2010 yield levels

O: regional statistics

E50: ensemble median

1-26: individual models

Site	Crop	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	E50	O
FI	S												N/A											N/A					30
FI	W																												30
DE	S												N/A																20
DE	W																												29
ES	S												N/A																30
ES	W																												30



Sites – Finland (FI), Germany (DE) and Spain (ES)

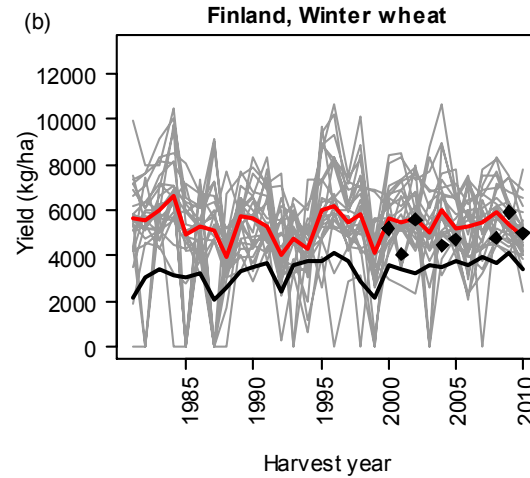
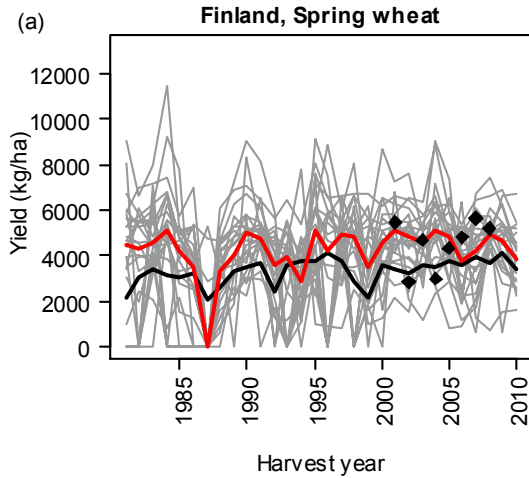
Crops – spring (S) and winter (W) wheat

Values for Observations (O) indicate the number of years for which observed crop yield data were available

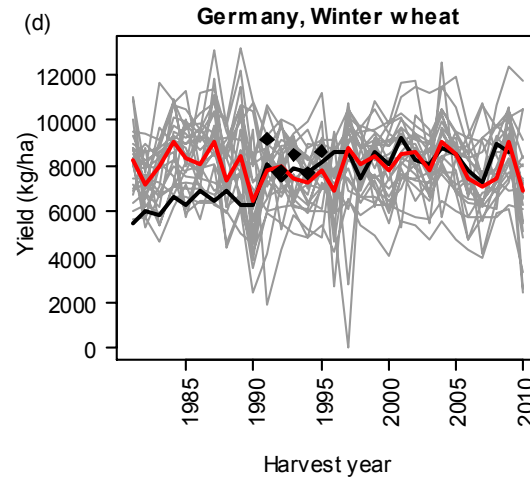
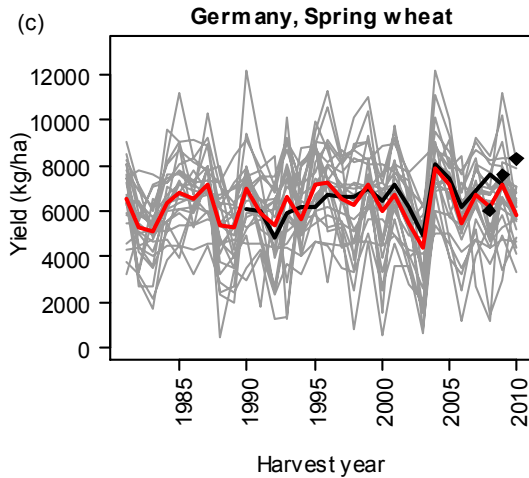
Models for which no results for a specific site or crop were provided are marked with N/A.

Simulated yields for the baseline 1981-2010

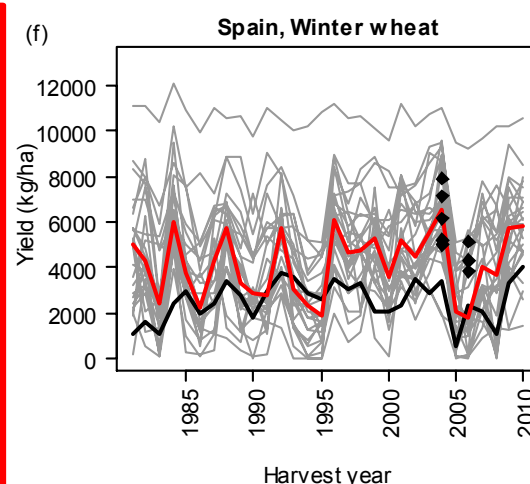
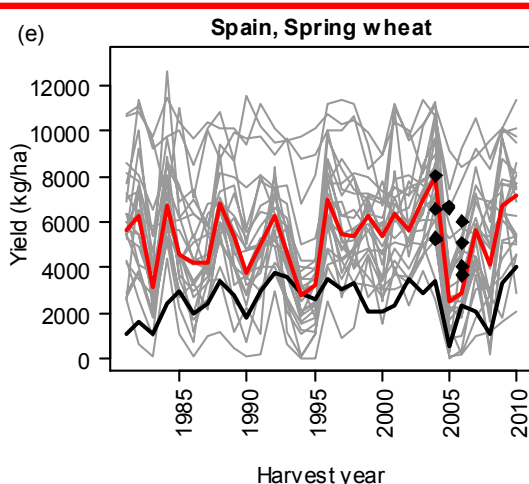
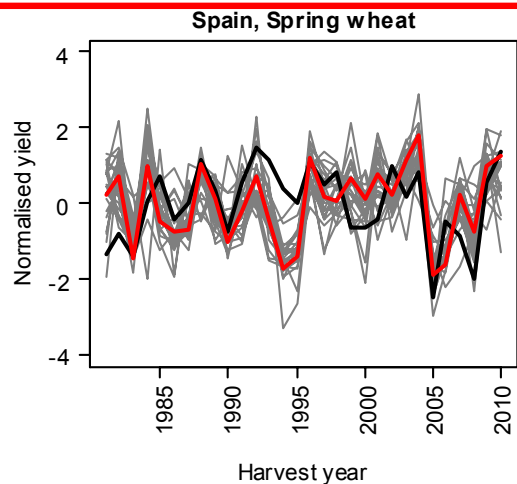
- Individual model results
- Ensemble median
- Historical yields of wheat
Finland: FAO Country level statistics
Germany: Eurostat regional statistics
Spain: provincial statistics for northern Spain, Spanish Ministry of Agriculture
- Calibration data



Finland

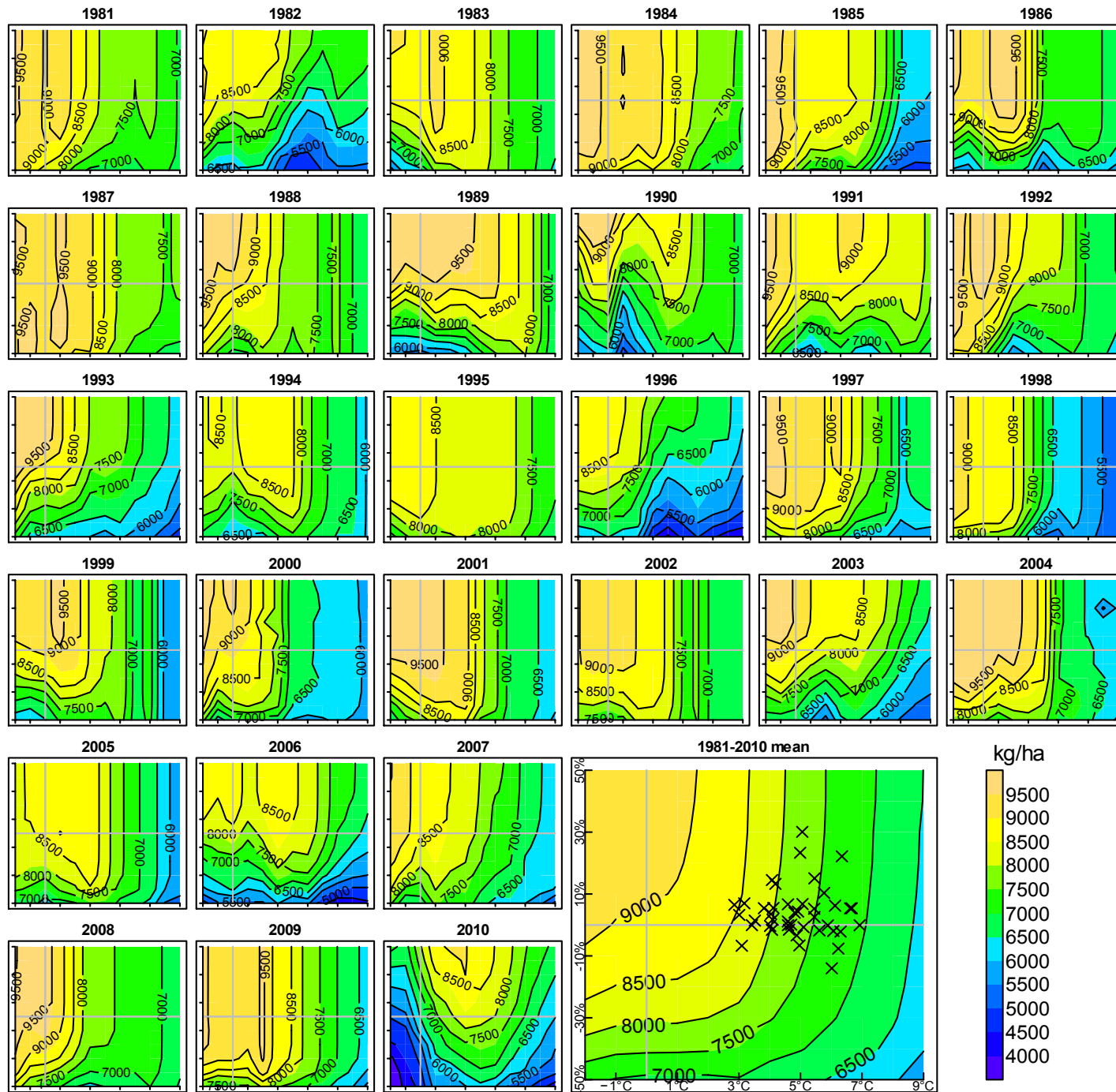


Germany



Spain

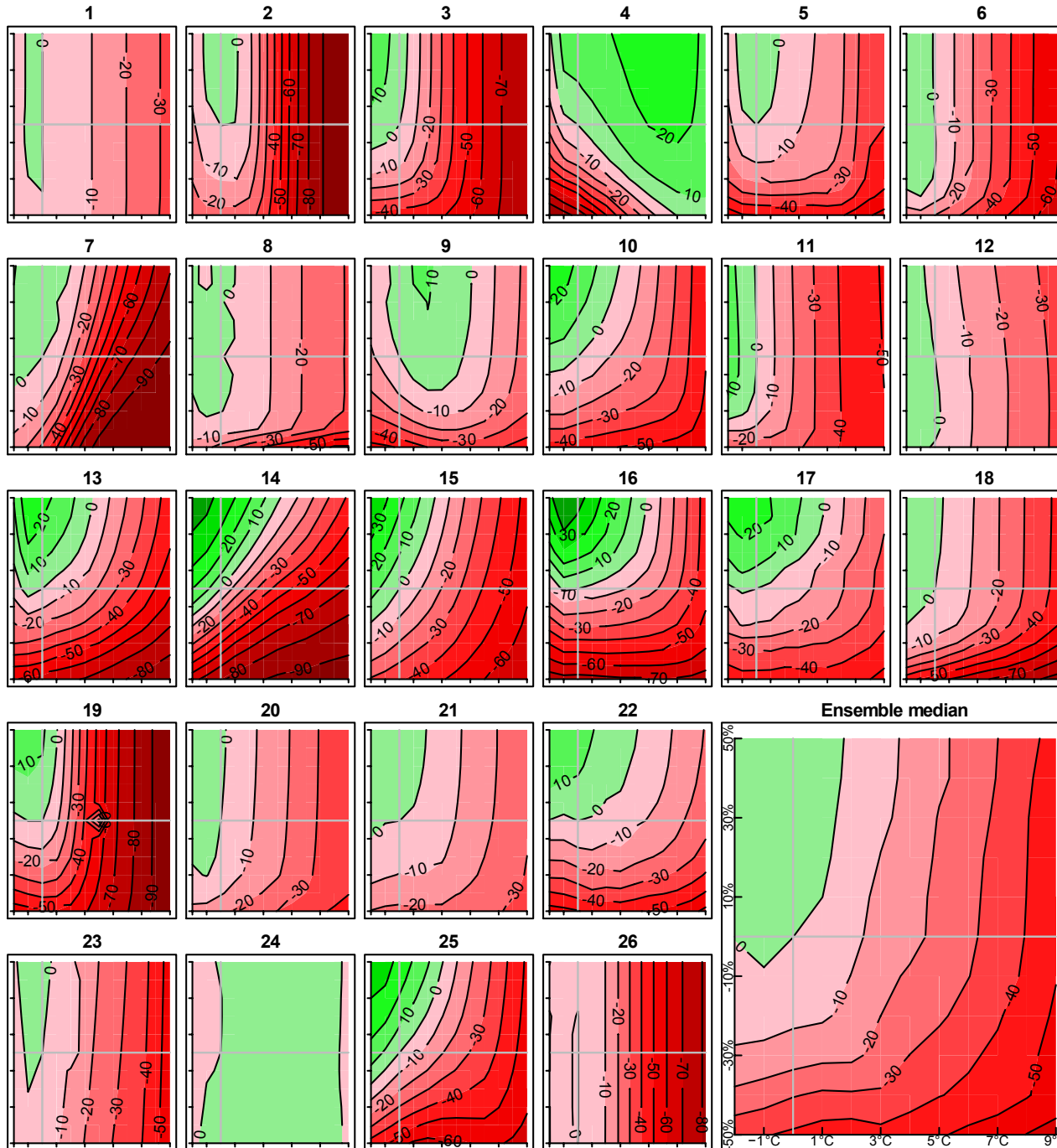
Winter wheat DM grain yields, Germany



One crop model, individual years 1981-2010 (small sub-plots) and 30-year mean (larger sub-plot)

Crosses in the 30-year mean plot: changes in annual temperature and precipitation projected by the CMIP5 ensemble of 36 global climate models for RCP8.5 over central Europe by 2070-2099 relative to 1981-2010.

Yield changes relative to unperturbed baseline

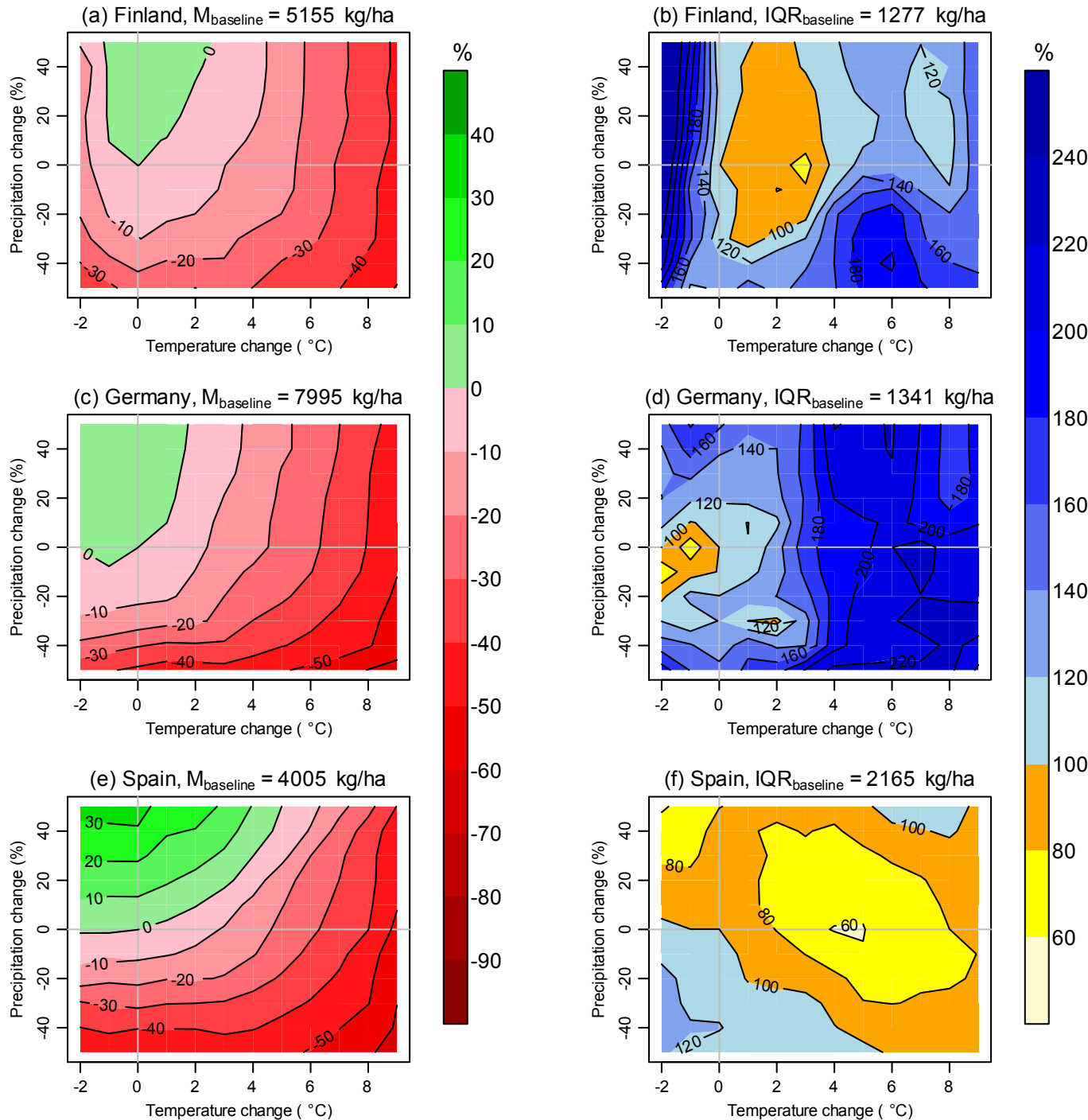


30-year average change in **winter wheat DM** yields relative to baseline climate (1981-2010) in **Germany**

26 models (small sub-plots) and ensemble median (larger sub-plot)

By definition, the yield change is 0% for the baseline climate at the intersection of the grey lines.

Ensemble medians and IQR of yield changes



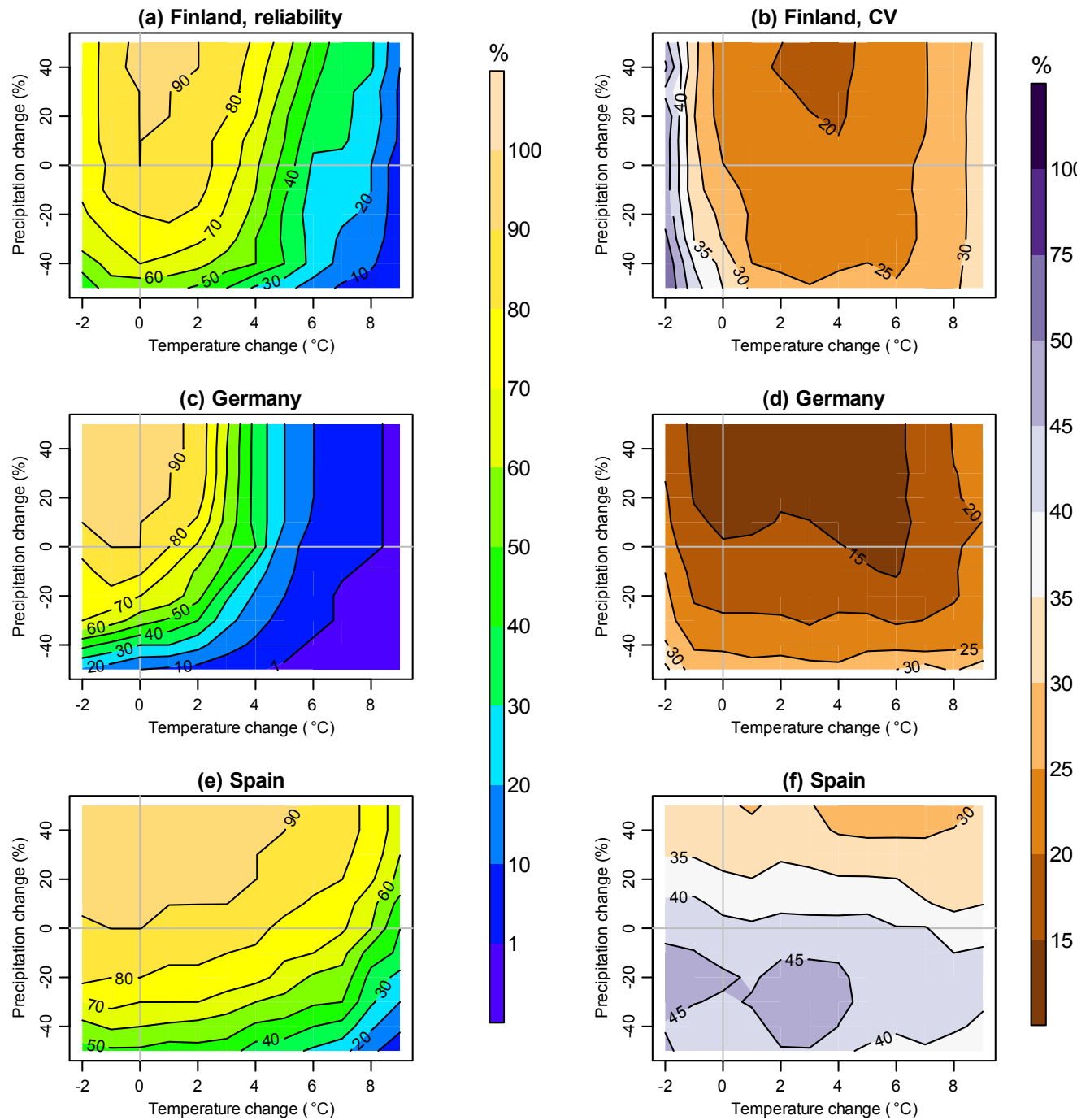
Winter wheat

Left: Median of yield changes by 26 crop models

Right: Inter-quartile range (IQR) of relative responses scaled to 100% at baseline

The ensemble median (M_{baseline}) and ensemble inter-quartile range (IQR_{baseline}) of absolute yields for the baseline are listed above each plot.

Ensemble medians of inter-annual variability



Winter wheat

Left:

Yield reliability = % of years when yield is above the 10th %-tile of the baseline yield

Right:

Coefficient of variation (CV) of annual yields

Ensemble medians of 26 crop models

RESULTS II: CLASSIFICATION OF MODEL RESPONSES

Classification of responses

- Different models and different conditions show different behaviour
- ⇒ Can we group models according to their different sensitivities to climate change?
 - Two approaches for grouping IRSs:
 - Clustering algorithm
 - Rules defined by expert judgment
- ⇒ Can we find explanations for different model responses?

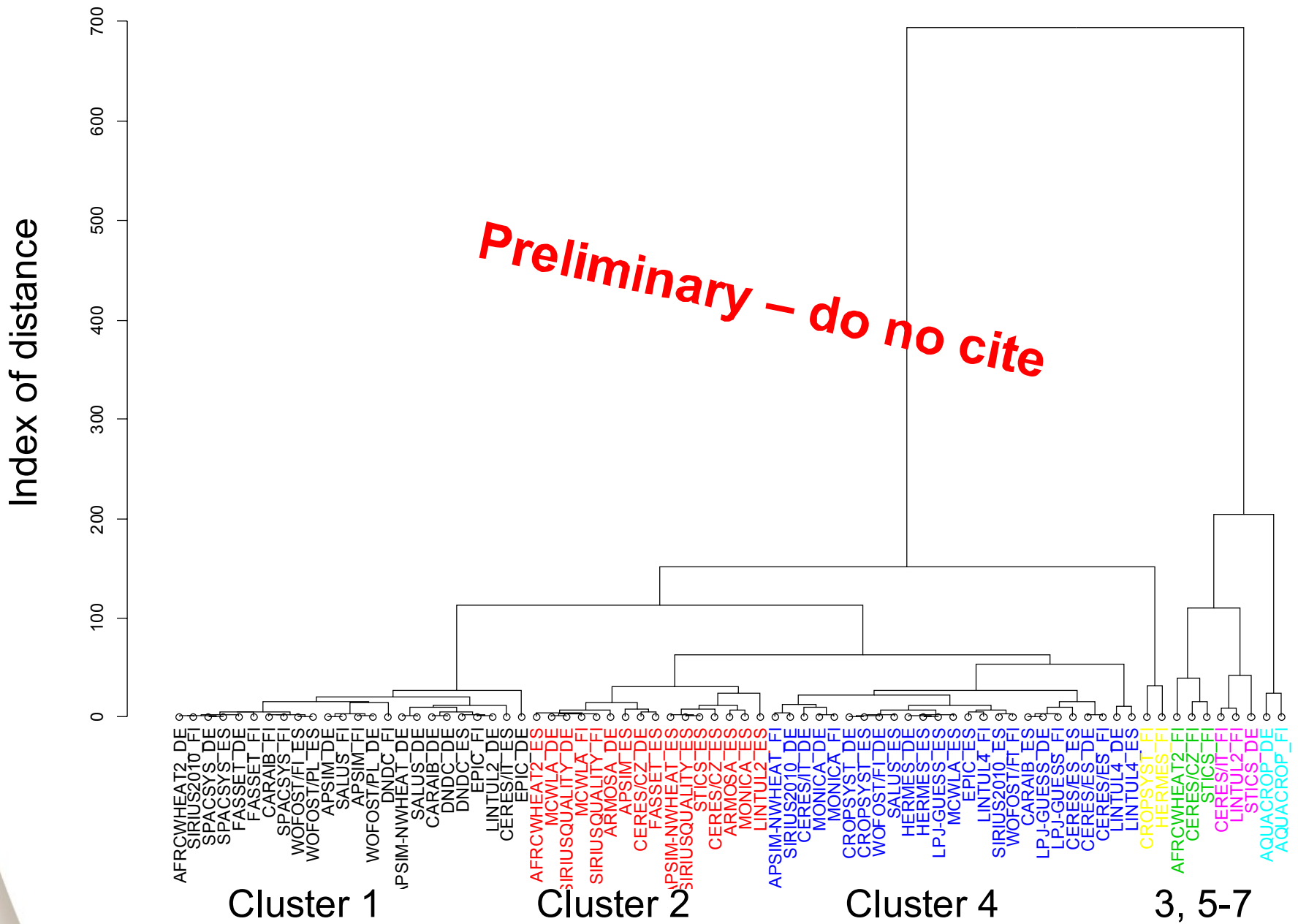
Some preliminary results and ideas

Clustering of IRSs based on correlation and Euclidian distance

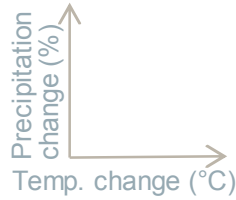
- Distances between two IRSs are defined based on their pattern and magnitude:
 - Pearson correlation coefficient $r^{*(-1) + 1}$
 - Euclidian distance over all points of the IRScombined by taking the product of the two
- IRSs are clustered (per crop, for 3 locations) by hierarchical clustering that minimize the distances between members of each cluster:
 - *agnes* (agglomerative nesting) algorithm in R (Kaufman & Rousseeuw 1990), using the average method to determine clusters
 - The number of clusters was set to 7 (according to a rule of thumb = $\sqrt{n/2}$), but after removing "outlying" IRSs that were in a separate cluster

Winter wheat 30-yr mean change in yield IRSs

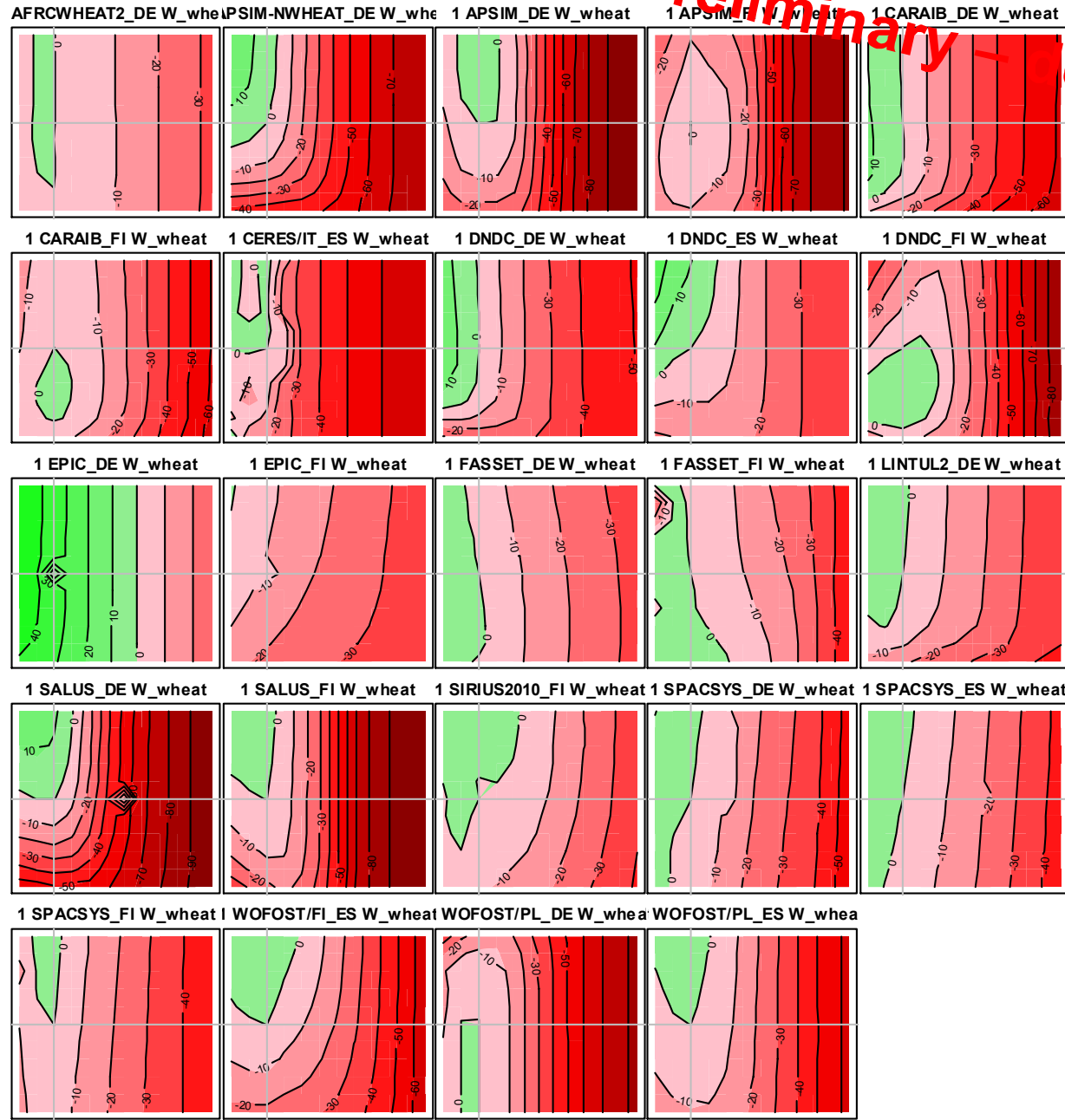
W_wheat/product, Agglomerative Coefficient = 0.99



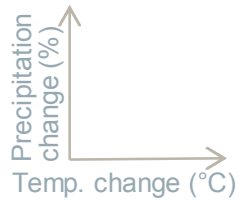
Cluster 1: strong temperature-sensitivity, yield decreases with warming, no precip-sens for high warming



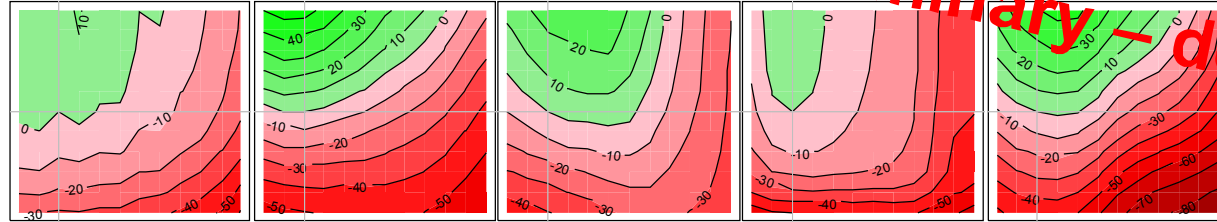
Preliminary - no cite



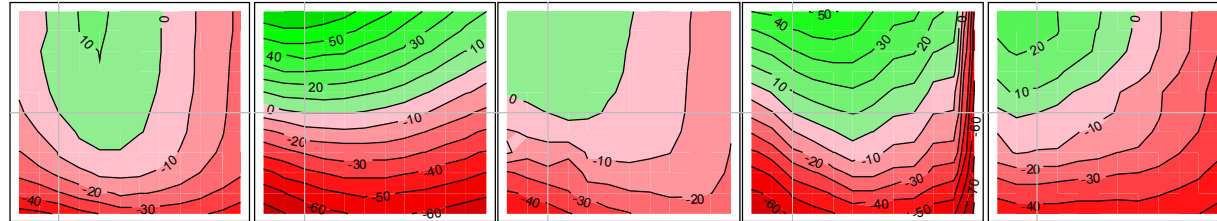
Cluster 2: strong precip-sensitivity, except for large warming



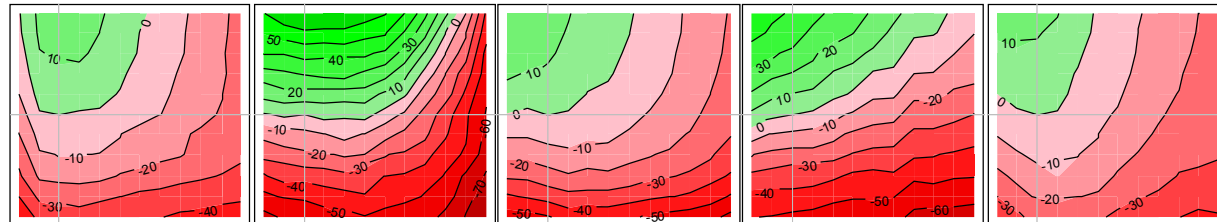
AFRCWHEAT2_ES W_wheat PSIM-NWHEAT_ES W_wheat 2 APSIM_ES W_wheat 2 ARMOSA_DE W_wheat 2 ARMOSA_ES W_wheat



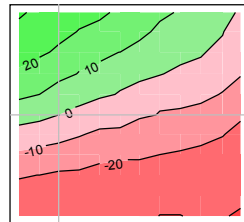
2 CERES/CZ_DE W_wheat 2 CERES/CZ_ES W_wheat 2 FASSET_ES W_wheat 2 LINTUL2_ES W_wheat 2 MCWLA_DE W_wheat



2 MCWLA_FI W_wheat 2 MONICA_ES W_wheat IRIUSQUALITY_DE W_wheat IRIUSQUALITY_ES W_wheat IRIUSQUALITY_FI W_wheat



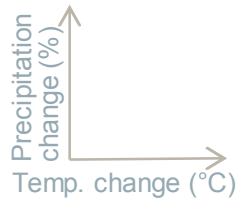
2 STICS_ES W_wheat



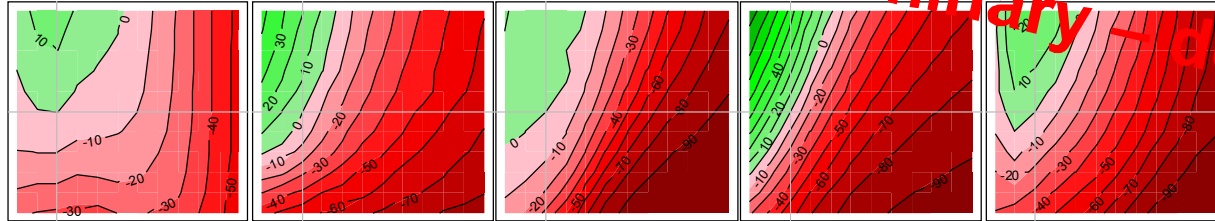
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Cluster 4: yield decrease with warming and drying

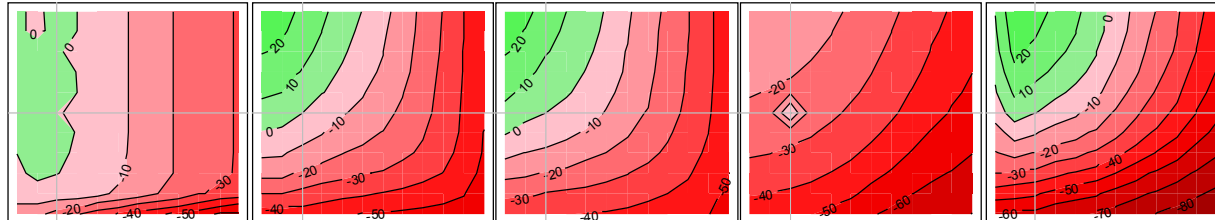
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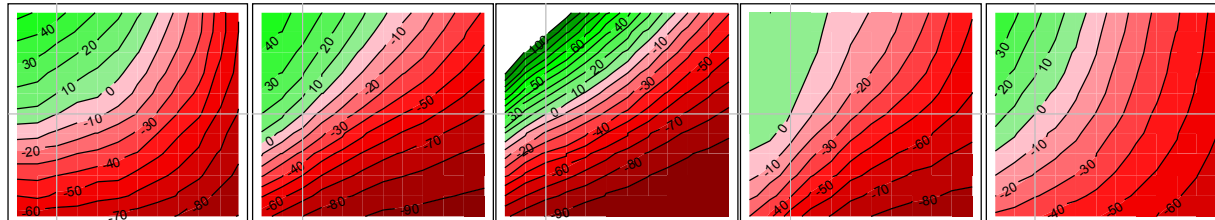
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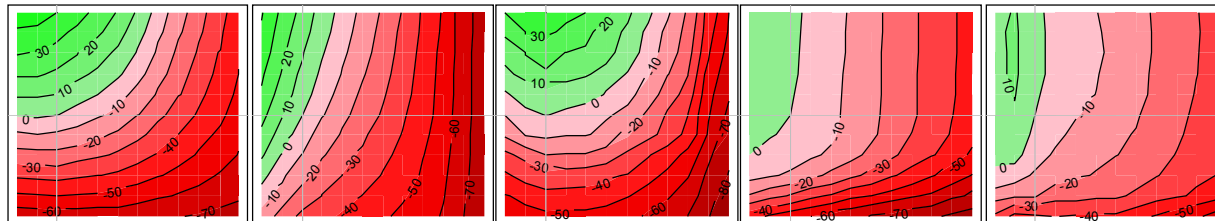
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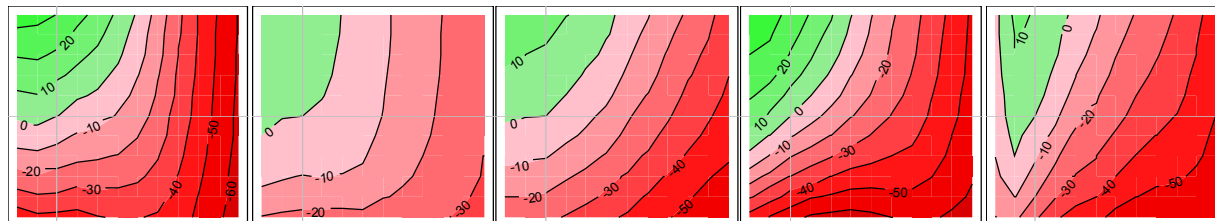
4 HERMES_ES W_wheat 4 LINTUL4_DE W_wheat 4 LINTUL4_ES W_wheat 4 LINTUL4_FI W_wheat 4 LPJ-GUESS_DE W_wheat



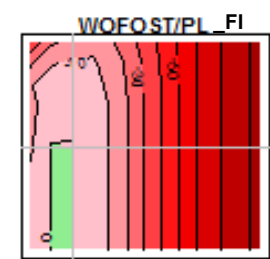
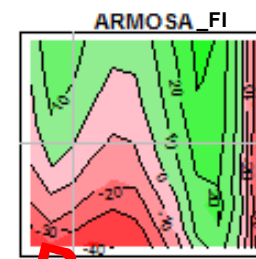
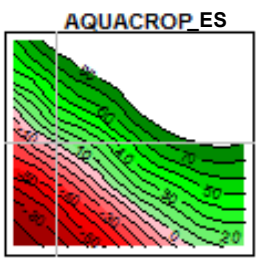
4 LPJ-GUESS_ES W_wheat 4 LPJ-GUESS_FI W_wheat 4 MCWLA_ES W_wheat 4 MONICA_DE W_wheat 4 MONICA_FI W_wheat



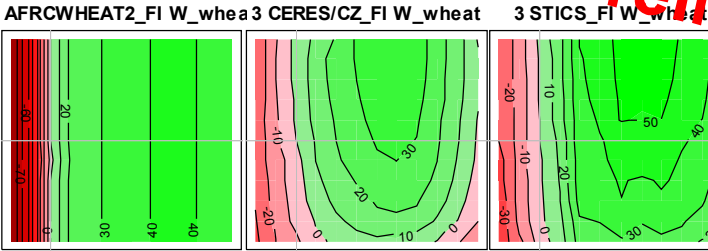
4 SALUS_ES W_wheat 4 SIRIUS2010_DE W_wheat 4 SIRIUS2010_ES W_wheat 4 WOFOST/FI_DE W_wheat 4 WOFOST/FI_FI W_wheat



Separate clusters

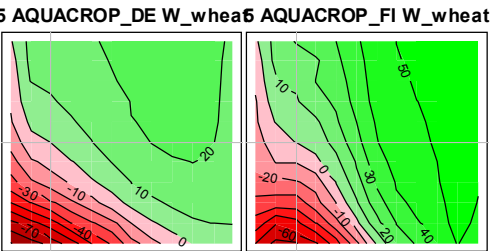


Cluster 3: strong temp-sensitivity, except for large warming

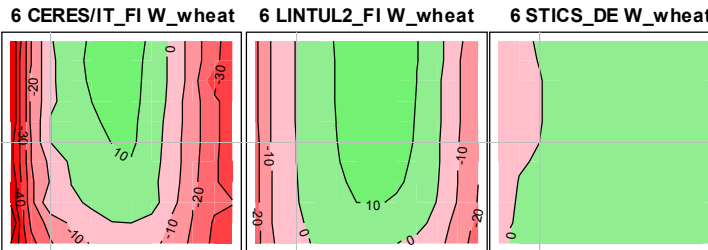


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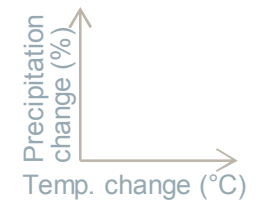
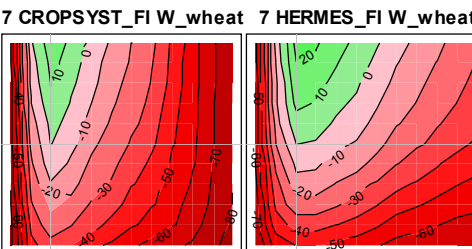
Cluster 5: yield increases with warming and precip increases



Cluster 6: yield increases with warming and precip increases, but decreases for high warming

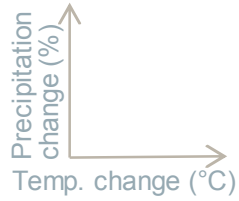


Cluster 7: V-shape with centred around 0-change

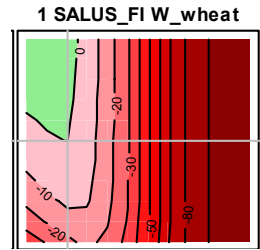


Central examples from each cluster

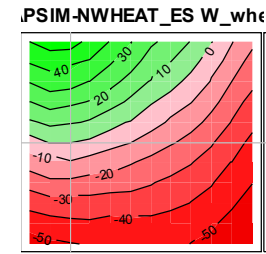
Preliminary – do no cite



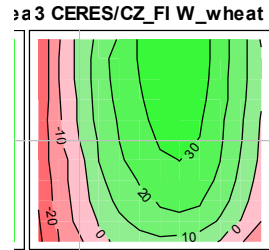
Cluster 1: n=24



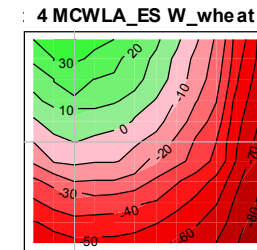
Cluster 2: n=16



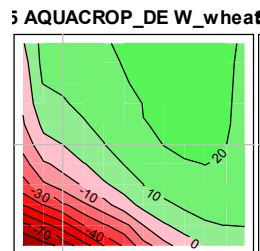
Cluster 3: n=3



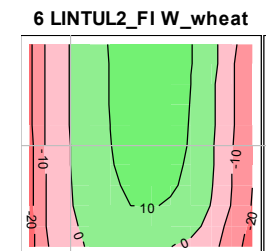
Cluster 4: n=25



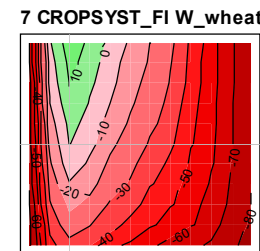
Cluster 5: n=2



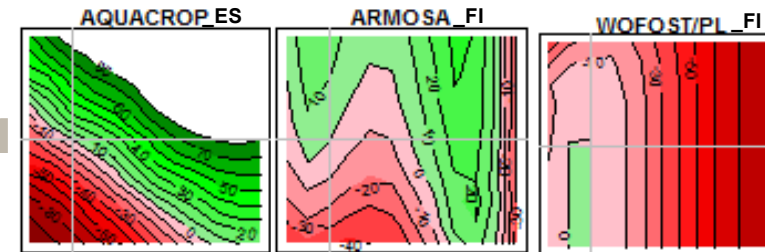
Cluster 6: n=3



Cluster 7: n=2



Separate: n=1



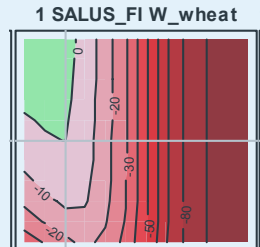
Labels and cluster groups: winter wheat

Yield decrease (Y-) with temperature increase (T+)

Preliminary – do no cite

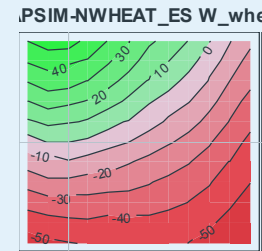
T-dominant (Y-: T+)

Cluster 1: n=24



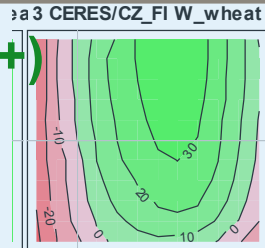
P-dominant (Y+: P+)

Cluster 2: n=16

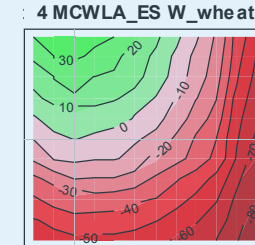


T-dominant (Y+: T+)

Cluster 3: n=3

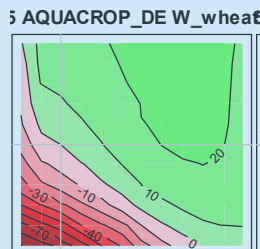


Cluster 4: n=25



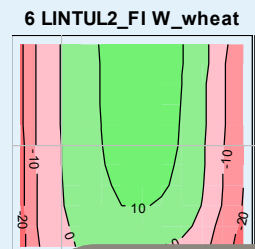
Y+: T+ & P+

Cluster 5: n=2



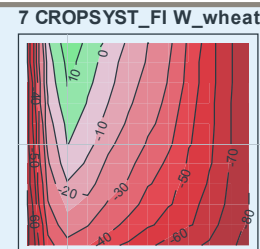
Y+: T+, Y-: T++

Cluster 6: n=3

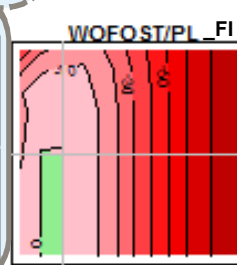
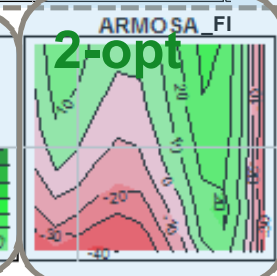
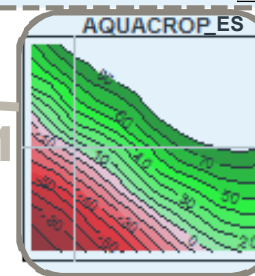


V-shape

Cluster 7: n=2



Separate: n=1



Clustering of IRS for 30-year-mean change in winter wheat yields

Preliminary – do not cite

T-dominant (Y-: T+)

P-dominant (Y+: P+)

T-dominant (Y+: T+)

Y+: T+ & P+

Y+: T+, Y-: T++

V-shape

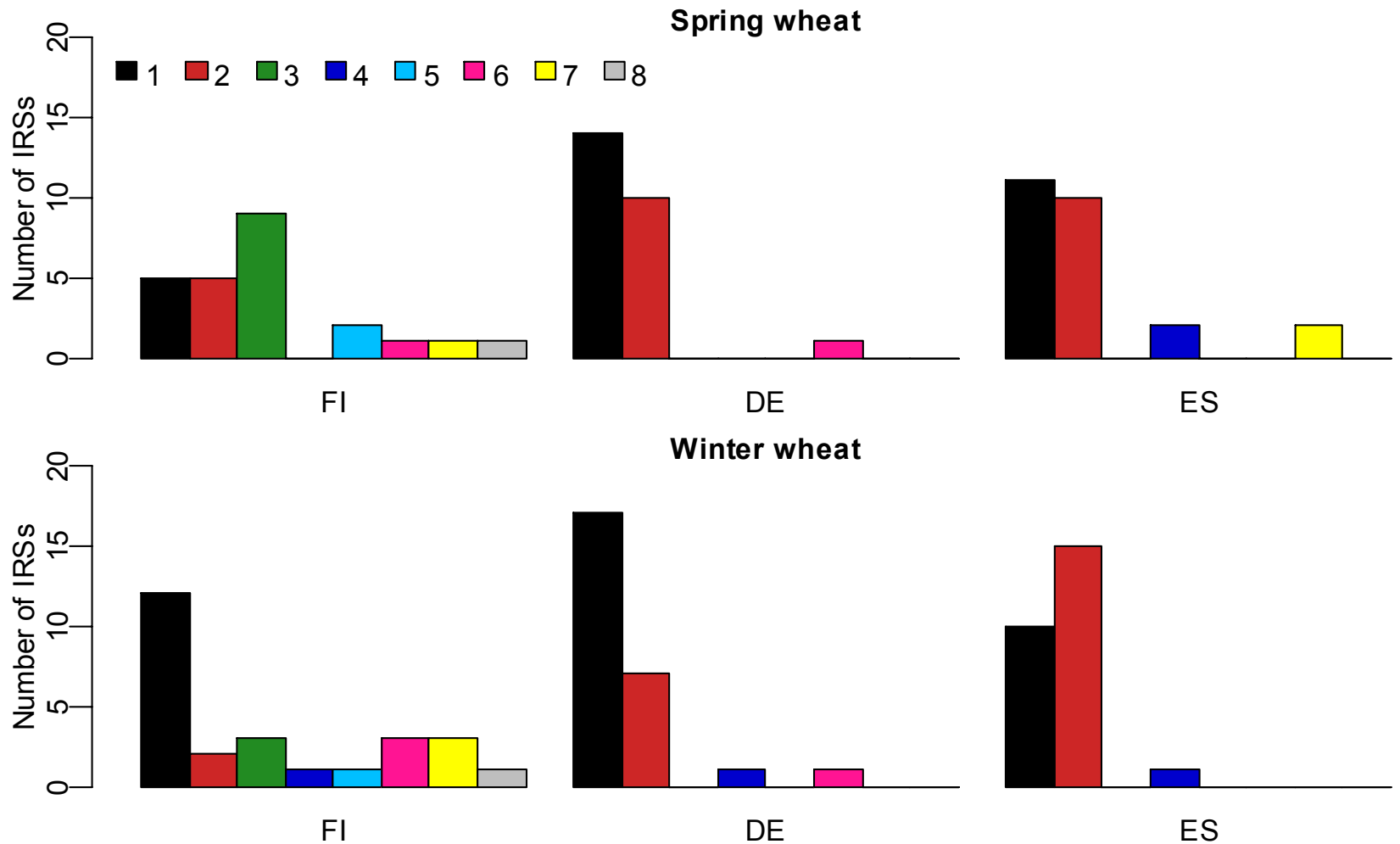
2-opt

	FI	ES	DE	# of different clusters
AFRCWHEAT2	Y+:T+ 3	Y+:P+ 2	Y-:T+ 1	3
APSIM	Y-:T+ 1	Y+:P+ 2	Y-:T+ 1	2
APSIM-NWHEAT	Y+:P+ 4	Y+:P+ 2	Y-:T+ 1	3
AQUACROP	Y+:T+,P+ 5	Y+:T+,P+ 5	Y+:T+,P+ 5	1
ARMOSA	2-T-opt 8	Y+:P+ 2	Y+:P+ 2	2
CARAIB	Y-:T+ 1	Y+:P+ 4	Y-:T+ 1	2
CERES/CZ	Y+:T+ 3	Y+:P+ 2	Y+:P+ 2	2
CERES/ES	Y+:P+ 4	Y+:P+ 4	Y+:P+ 4	1
CERES/IT	Y+:T+,Y-:T++ 6	Y-:T+ 1	Y+:P+ 4	3
CROPSYST	V-shape 7	Y+:P+ 4	Y+:P+ 4	2
DNDC	Y-:T+ 1	Y-:T+ 1	Y-:T+ 1	1
EPIC	Y-:T+ 1	Y+:P+ 4	Y-:T+ 1	2
FASSET	Y-:T+ 1	Y+:P+ 2	Y-:T+ 1	2
HERMES	V-shape 7	Y+:P+ 4	Y+:P+ 4	2
LINTUL2	Y+:T+,Y-:T++ 6	Y+:P+ 2	Y-:T+ 1	3
LINTUL4	Y+:P+ 4	Y+:P+ 4	Y+:P+ 4	1
LPJ-GUESS	Y+:P+ 4	Y+:P+ 4	Y+:P+ 4	1
MCWLA	Y+:P+ 2	Y+:P+ 4	Y+:P+ 2	2
MONICA	Y+:P+ 4	Y+:P+ 2	Y+:P+ 4	2
SALUS	Y-:T+ 1	Y+:P+ 4	Y-:T+ 1	2
SIRIUS2010	Y-:T+ 1	Y+:P+ 4	Y+:P+ 4	2
SIRIUSQUALITY	Y+:P+ 2	Y+:P+ 2	Y+:P+ 2	1
SPACSYS	Y-:T+ 1	Y-:T+ 1	Y-:T+ 1	1
STICS	Y+:T+ 3	Y+:P+ 2	Y+:T+,Y-:T++ 6	3
WOFOST/FI	Y+:P+ 4	Y-:T+ 1	Y+:P+ 4	2
WOFOST/PL	?	Y-:T+ 1	Y-:T+ 1	1



Number of IRSs per cluster for spring and winter wheat at the Finnish (FI), German (DE) and Spanish (ES) sites

Preliminary – do no cite

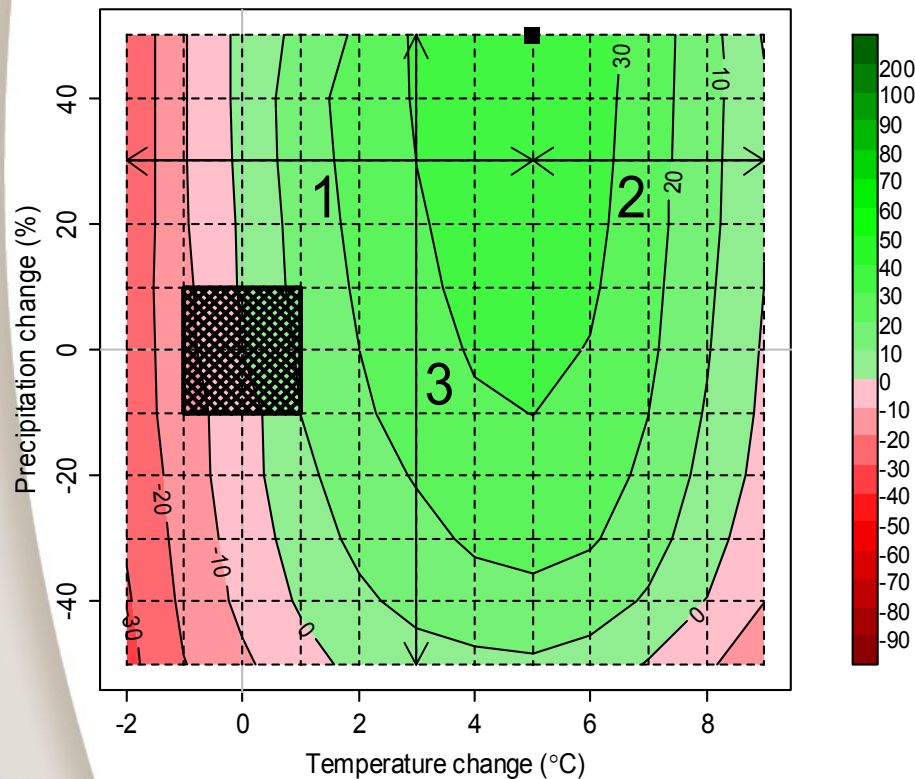


Grouping of IRSs with expert judgement

$\underline{I}_{B+P} \underline{D}$

T-dominant response; optimal yield at baseline T; strong decline with rising T; baseline P-deficit

Preliminary – do no cite



Symbol	Variants
<u>I</u>	Temperature response dominates
<u>P</u>	Precipitation response dominates
B	Optimum yield close to baseline climate
C	Optimum yield cooler than baseline T
W	Optimum yield warmer than baseline T
D	Precipitation deficit limits baseline yield
+	Strong response with large increase relative to baseline
-	Strong response with large decrease relative to baseline
±	Strong response with large increase and large decrease relative to baseline

Results – Clustering of IRSs

Winter wheat

”Subjective” method

applies expert judgement to describe the climatic conditions relative to the baseline and the relative influence of temperature and precipitation on yields away from the baseline

Yield response behaviour

- 1 T-dominates; Insensitive to P
- 2 T-dominates; P has less effect
- 3 T and P have comparable effect
- 4 P-dominates; T has less effect
- 5 P-dominates; Insensitive to T
- 6 Unclassified

Model	FI	ES	DE
AFRCWHEAT2	1	4	1
APSIM-Nwheat	3	4	2
APSIM	2	4	2
AquaCrop	2	4	4
ARMOSA	6	4	2
CARAIB	2	2	2
CERES-wheat DSSAT v.4.6/CZ	2	5	4
CERES-wheat DSSAT v.4.5/ES	2	2	2
CERES-wheat DSSAT v.4.5/IT	2	2	2
CropSyst	2	2	3
DNDC	2	2	2
EPIC	6	6	6
Fasset	1	4	1
HERMES	3	3	4
Lintul2	2	4	2
Lintul4	2	4	4
LPJ-GUESS	2	3	2
MCWLA	4	3	4
MONICA V1.2	2	4	3
SALUS	2	3	2
Sirius 2010	2	2	2
Sirius Quality	4	4	4
SPACSYS	2	1	1
STICS	2	4	1
WOFOST 7.1/FI	2	2	2
WOFOST 7.1/PL	2	2	2

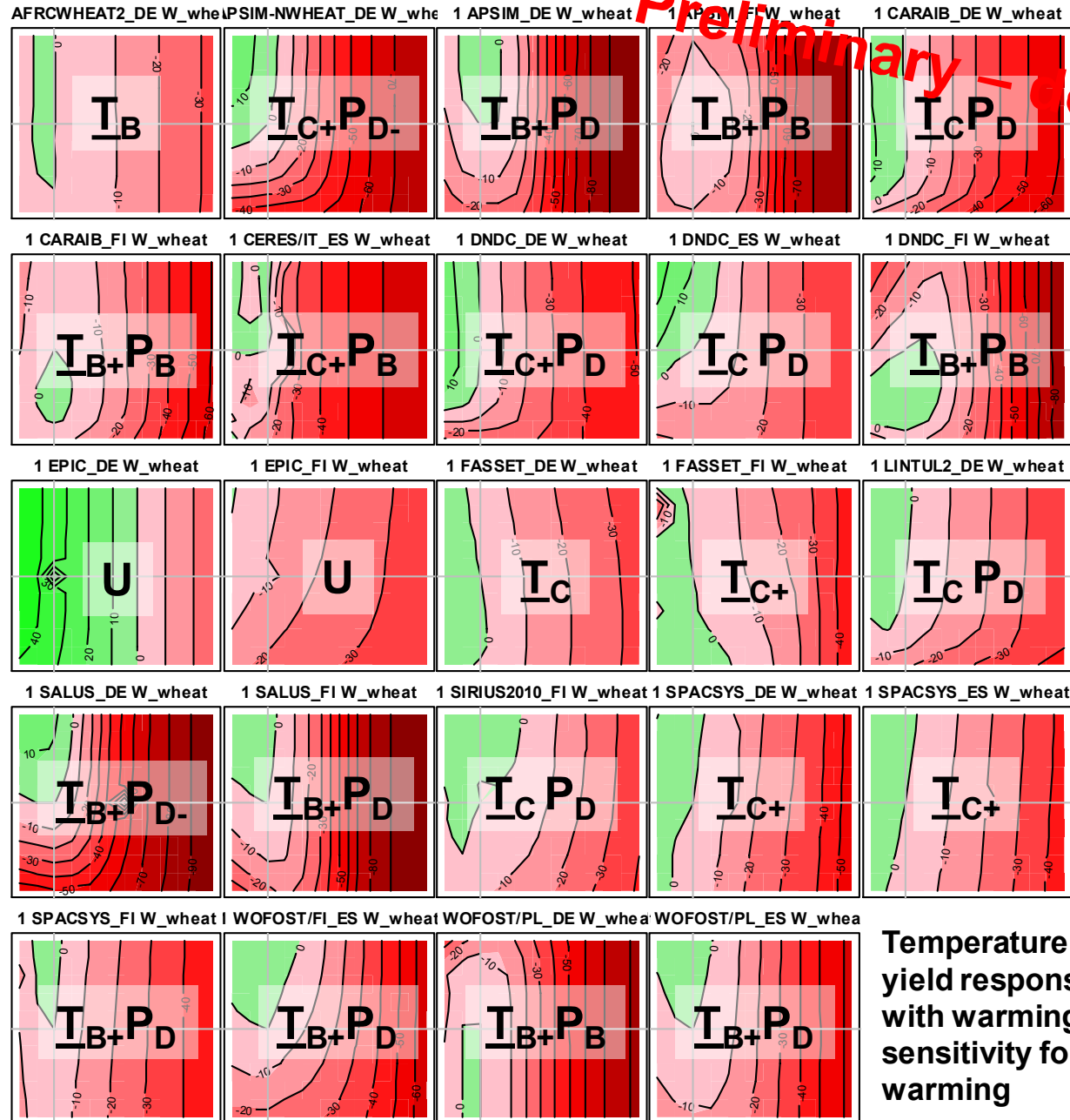
Preliminary - do not cite

Results – Clustering of IRSs

”Objective” method, Cluster 1

”Subjective” method

	Basic class
1	I_B
	I_C
2	I_W
	$I_B P_B$
	$I_B P_D$
	$I_C P_B$
	$I_C P_D$
3	$I_W P_D$
	$I_B P_D$
	$I_C P_D$
4	$T_B P_D$
	$T_C P_D$
	$T_W P_D$
5	P_D
6	U



Preliminary – do not cite

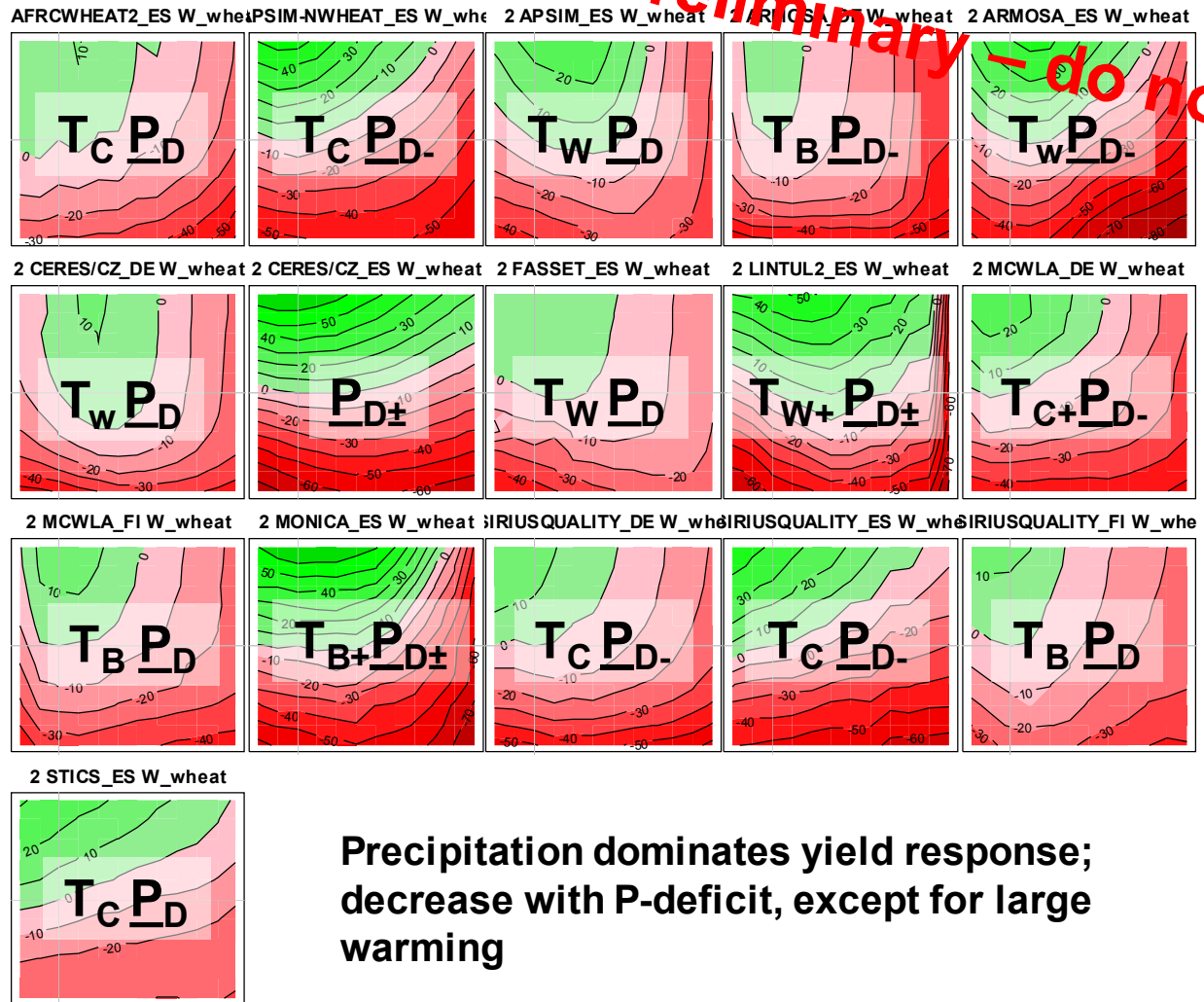
Temperature dominates yield response; decrease with warming; low P-sensitivity for strong warming

Results – Clustering of IRSs

”Subjective” method

	Basic class
1	T_B
	T_C
	T_W
2	$T_B P_B$
	$T_B P_D$
	$T_C P_B$
	$T_C P_D$
	$T_W P_D$
3	$T_B P_D$
	$T_C P_D$
4	$T_B P_D$
	$T_C P_D$
	$T_W P_D$
5	P_D
6	U

”Objective” method, Cluster 2



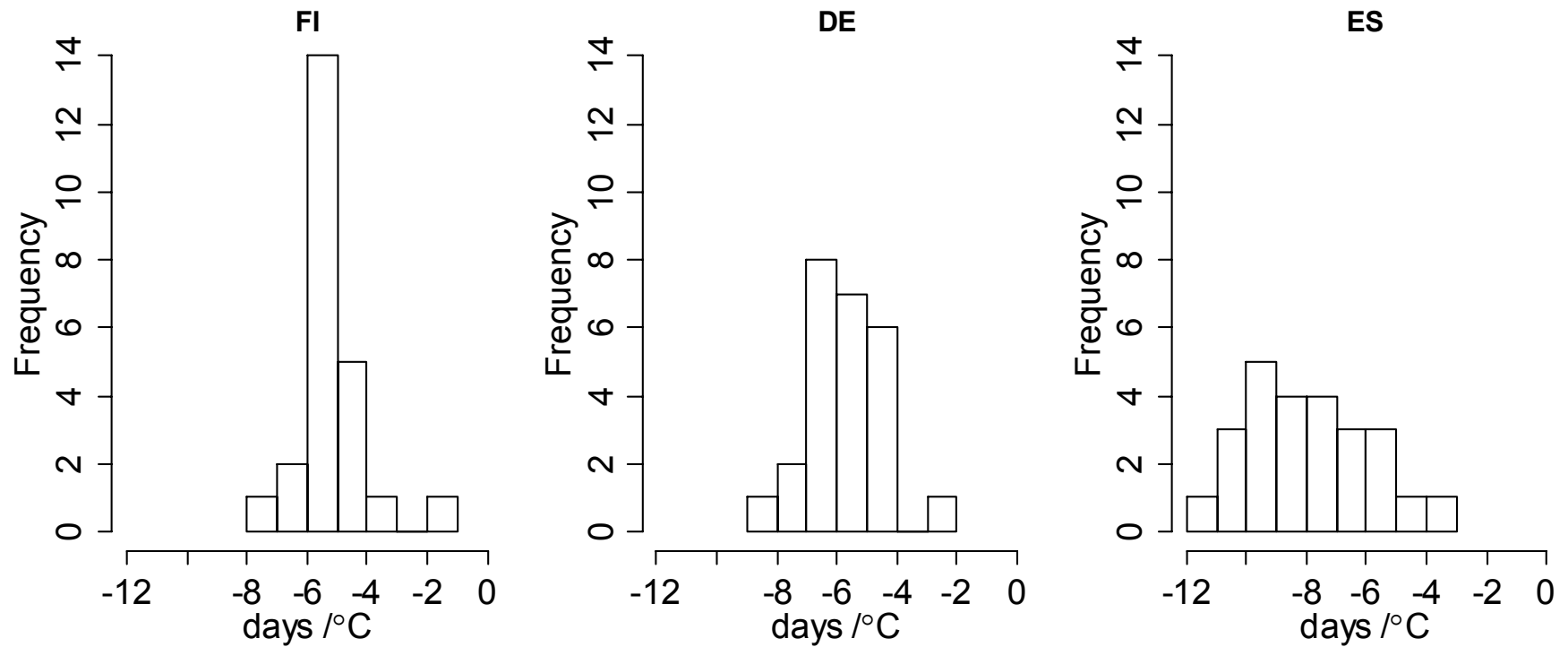
Precipitation dominates yield response; decrease with P-deficit, except for large warming

Ideas for attempting to explain model response

- Comparing clusters of yields responses to
 - Simulated harvest index and length of growing period
 - model characteristics

Ensemble distribution of simulated 30-year averaged responses in the rate of change of growing period length for spring wheat (sowing to maturity)

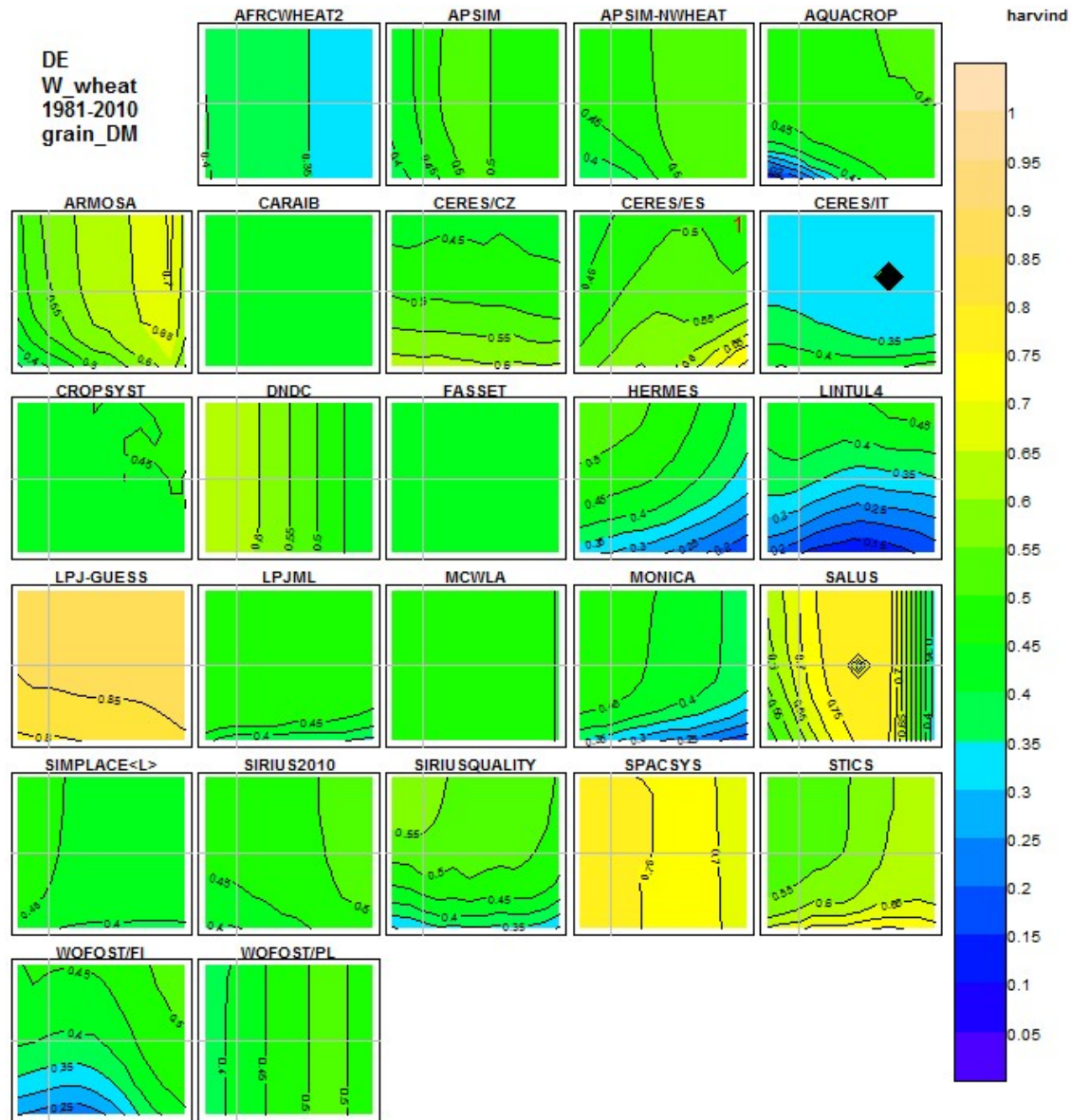
Preliminary – do not cite



Harvest index for winter wheat in Germany

Ratio of grain to above-ground dry matter at harvest

Preliminary – do no cite



Harvest index for winter wheat in Germany

Preliminary – do not cite

Number of models in four range classes of the harvest index (HI; ratio of grain to above-ground dry matter at harvest) for spring and winter wheat in Finland, Germany and Spain for the baseline climate, for a large warming (T+9; temperature change = +9°C, precipitation at baseline) and large drying (P-50; temperature at baseline, precipitation change = -50%). Thresholds for the HI ranges are based on experimental data presented by Hay (1995) and Foulkes et al. (2011). The colours indicate if the number of models remains the same as for the baseline (grey), decreases (blue) or increases (red).

HI class (range)	Finland			Germany			Spain		
	Baseline	T+9	P-50	Baseline	T+9	P-50	Baseline	T+9	P-50
Spring wheat									
Low (<0.31)	2	3	2	1	2	3	2	3	7
Normal (0.31-0.50)	11	13	14	19	17	17	18	13	11
High (0.51 - 0.64)	9	5	7	3	4	3	3	7	6
Implausibly high (>0.64)	2	3	1	2	2	2	2	2	1
Winter wheat									
Low (<0.43)	10	13	12	6	9	15	16	17	15
Normal (0.43-0.53)	12	9	10	15	13	4	8	6	9
High (0.54-0.64)	2	1	2	2	2	4	0	0	1
Implausibly high (>0.64)	2	3	2	3	2	3	2	3	1



Comparison of clusters to model characteristics

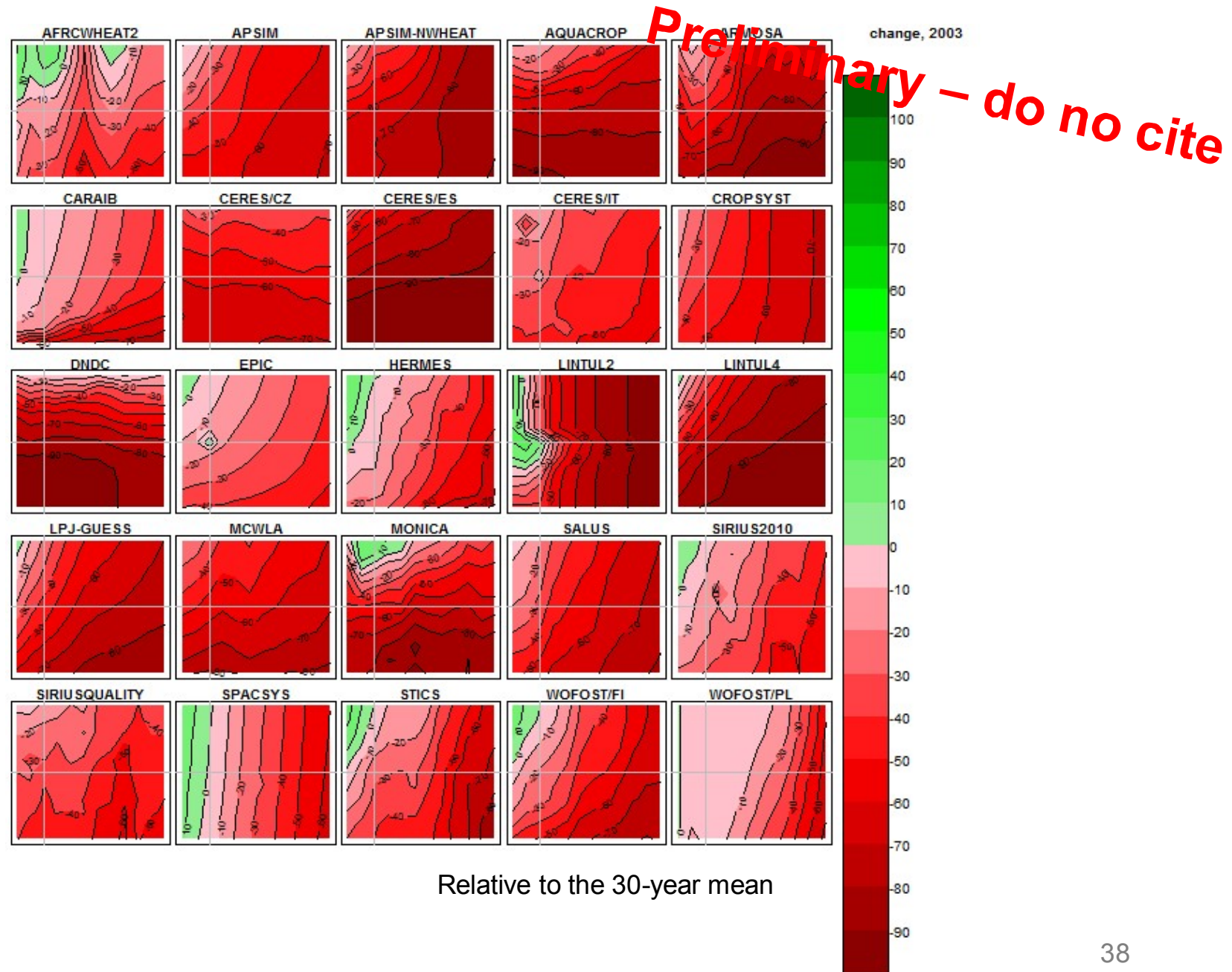
Source: Pirttioja et al. 2015

Table S1. Characteristics of the models applied in the study (based on Asseng et al. 2013; modified). Key references of the models are listed below the table.

ID	Model	Leaf area / light interception (a)	Light utilization (b)	Yield formation (c)	Phenology (d)	Root distribution over depth (e)	Type of water stress (f)	Type of heat stress (g)	Water dynamics (h)	Evapo-transpiration (i)	Process modified by elevated CO ₂ (j)	Climate input variables (k)	Model relative (l)	Model type (m)
1	AFRCWHEAT2	D	P-R	Prt/Gn/B	T/DL/V	EXP	E/S	P/R	C	PM	RUE	R/Tx/Tn/Ta/Td/Rd/e/W	I	P
2	APSIM-Nwheat	S	RUE	Prt	T/DL/V	EXP	S	V	C	PT	RUE/TE	R/Tx/Tn/Rd	C	P
3	APSIM-Wheat	S	RUE	Prt/Gn/B	T/DL/V/O	O	E	V	C/R	PT	RUE/TE/CLN	R/Tx/Tn/Rd	C	P
4	AquaCrop	S	TE	HI_mw/B	T	O	E/S	R	C	PM	TE	R/Tx/Tn,Ta,Rd,RH,W	none	P
5	ARMOSA	D	P-R	Prt/B	T/DL/V	O	E/S	V/P	C	PT	RUE/TE	R/Tx/Tn/Rd	S	P/G
6	CARAIB	D	P-R	HI/B	T	O	S	-	C/R	P	F	Cl/R/Tx/Tn/RH/W	none	G
7&8	CERES-wheat DSSAT	S	RUE	B/Gn	T/DL/V	EXP	E/S	-	C	PT	RUE/TE	R/Tx/Tn/Rd/RH/W	C	P
9	CERES-wheat DSSAT	S	RUE	B/Gn/Prt/HiMw	T/DL/V	EXP	E/S	-	C	PT	RUE/TE	R/Tx/Tn/Rd	C	P
10	CropSyst	S	TE/RUE	HI/B	T/DL/V	EXP	E	R	C/R	PM	TE/RUE	R/Tx/Tn/Rd/RH/W	none	P
11	DNDC	S	RUE	HI/B	T/DL	EXP	E/S	P	C	P	RUE	R/Tx/Tn/Rd/RH/W	none	P/G
12	FASSET	D	RUE	HI/B	T/DL	EXP	E/S	-	C	MAK	RUE	R/Tx/Tn/Rd	none	P
13	HERMES	D	P-R	Prt	T/DL/V/O	EXP	E/S	-	C	PM	TE/F	R/Tx/Tn/Ta/Rd/RH/W	S/C	P
14	LINTUL-4	D	RUE	Prt/B	T/DL	LIN	E	-	C	P	RUE/TE	R/Tx/Tn/Rd/e/W	L	P
15	LPJ-GUESS	S	P-R	HI_mw/B	T/DL/V	LIN	E/S	P	C	PT	F	R/Ta/Rd	E	G
16	LPJmL	S	P-R	HI_mw/B	T/DL/V	EXP	E	-	C	PT	F	R/Ta/Rd	E	G
17	MCWLA-Wheat	S	P-R	HI/B	T/DL/V	EXP	E	V/ R	R	PM	F	R/Tx/Tn/Rd/e/W	none	G
18	MONICA	S	P-R	Prt	T/DL/V/O	EXP	E	V	C	PM	F	R/Tx/Tn/Rd/RH/W	H	P
19	SALUS	S	RUE	Prt/Hi	T/DL/V	EXP	E	V	C	PT	RUE	R/Tx/Tn/Rd	C	P
20	SIMPLACE <Lintul2, Slim>	S	RUE	Prt/B	T/DL/V	LIN	E	P	C	P	not considered	R/Tx/Tn/Rd/W	L	P
21	Sirius	D	RUE	B/Prt	T/DL/V	LIN	E	-	C	PT	RUE	R/Tx/Tn/Rd/e/W	none	P
22	SiriusQuality	D	RUE	B/Prt	T/DL/V	EXP	S	-	C	P	RUE	R/Tx/Tn/Rd/e/W	I	P
23	SPACSYS	S	P-R	Prt	T/DL/V	EXP/Call	E	-	R	PM	RUE/TE	R/Tx/Tn/Rd/RH/W	none	P
24	STICS	S	RUE	HI/Gn	T/DL/V/O	O	E/S	R/V/P	C	SW	RUE/Tr	R/Tx/Tn/Rd/e/W	none	P
25&26	WOFOST	D	P-R	Prt/B	T/DL	LIN	E/S	-	C	P	RUE/TE	R/Tx/Tn/Rd/e/W	S	G

Results – analysis of extreme years

Example of a particularly dry year at Nossen (DE) (spring wheat), year 2003



Relative to the 30-year mean

CONCLUSIONS

Conclusions 1/2

From Climate Research paper (1st part of this presentation):

- Demonstration of using Impact Response Surfaces (IRSs) for a systematic intercomparison of crop model behaviour under conditions of changing climate
- Ensemble average yields decline with higher temperatures (3–7% per 1°C) and decreased precipitation (3–9% per 10% decrease), but benefit from increased precipitation (0–8% per 10% increase)
- Yields are more sensitive to temperature than precipitation changes at the Finnish site while sensitivities are mixed at the German and Spanish sites
- Inter-model variability is highest for baseline climate at the Spanish site, but relatively insensitive to changed climate; modelled responses diverge most at the Finnish and German sites for winter wheat under temperature change
- Optimal temperatures for present-day cultivars are close to the baseline under Finnish conditions but below the baseline at the German and Spanish sites

Conclusions 2/2

- We have shown that clustering methods can be used to analyse patterns of IRSs for
- Next step is to start diagnosing the reasons for the different behaviours, the approach for this was outlined using information about the harvest index and phenology of simulations
- Clustering will also be tested for extreme years