

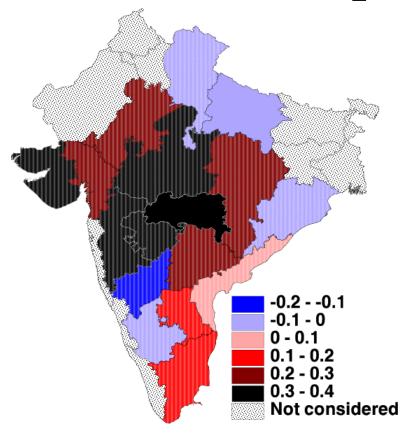
Combined crop-climate forecasting

Find spatial scale of weather-crop relationships

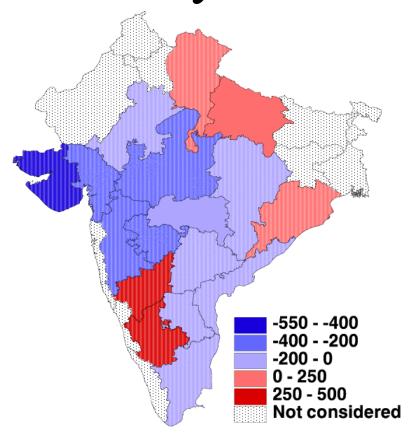


Hindcasts with observed weather data and reanalysis

The spatial scale of yield-weather relationships: EOF analysis



First EOF of sub-divisional yield (23.4% of variance)

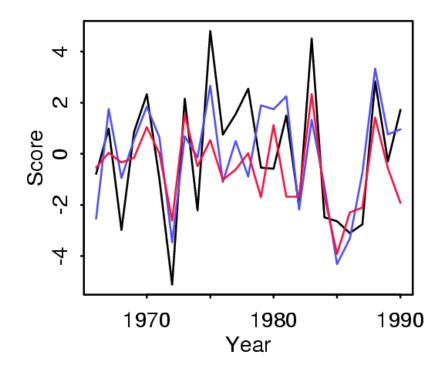


Yield anomaly for 1985

Challinor et. al. (2004)

Principle component time series

 Coherent associated large-scale patterns of seasonal rainfall and groundnut yield in India



First principal component of

—— rainfall

— yield

and PC3 of

—— 850hPa circulation

 $r^2(ppn,yield)=0.53$

 r^2 (circ, yield)=0.45

Combined crop-climate forecasting

 \approx scale of the climate model

Find spatial scale of weather-crop relationships



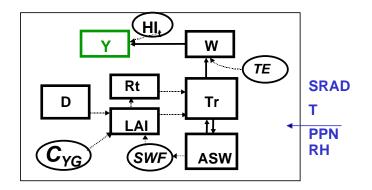
Crop modelling at appropriate spatial scale

Hindcasts with observed weather data and reanalysis

Crop modelling methods circa 2000

- Empirical and semi-empirical methods
 - +Low input data requirement
 - + Can be valid over large areas
 - May not be valid as climate, crop or management change
- Process-based
 - + Simulates nonlinearities and interactions
 - Extensive calibration is often needed
 - skill is highest at plot-level

General Large Area Model for Annual Crops (GLAM)

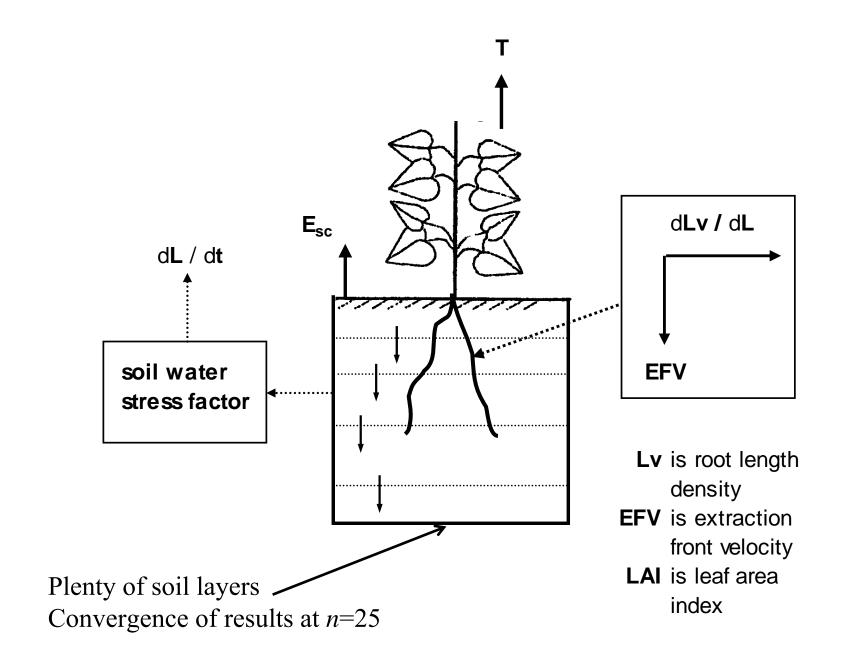


Challinor et. al. (2004)

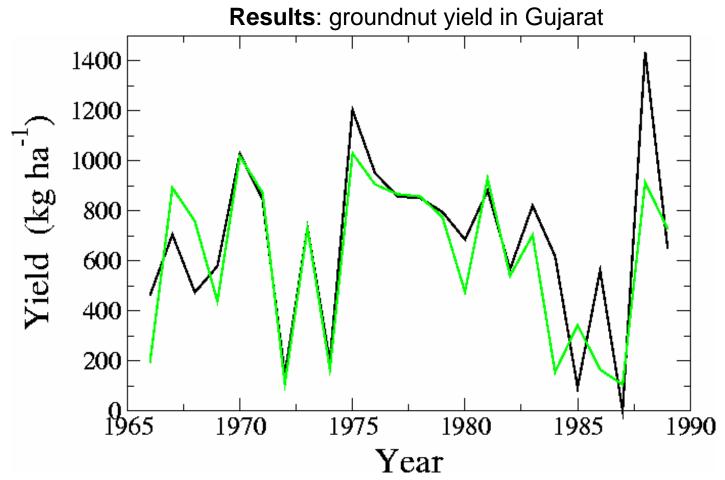
- Combines:
 - the benefits of more empirical approaches (low input data requirements, validity over large spatial scales)

with

- the benefits of the process-based approach (e.g. the potential to capture intra-seasonal variability, and so cope with changing climates)
- Yield Gap Parameter to account for the impact of differing nutrient levels, pests, diseases, non-optimal management etc.

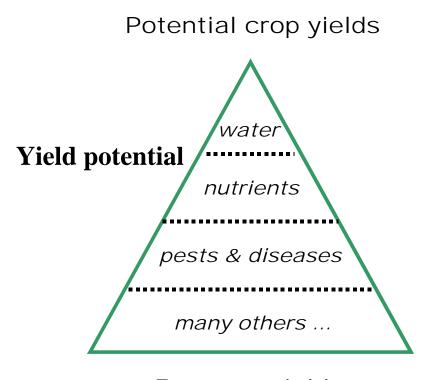


General Large-Area Model for annual crops



Challinor, A. J., T. R. Wheeler, J. M. Slingo, P. Q. Craufurd and D. I. F. Grimes (2004). Design and optimisation of a large-area process-based model for annual crops. Agricultural and Forest Meteorology, 124, (1-2) 99-120.

The yield gap parameter



Farmer's yields

GLAM uses a timeindependent site-specific yield gap parameter

GLAM can be used to simulate yield potential, but:

- Less useful
- Validation data harder to find



To what extent are mean yields determined by YGP?

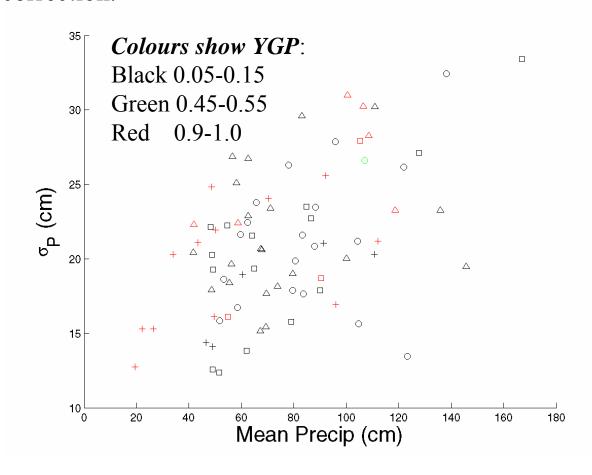
- YGP values differ when calibrated on different input data; Hence it contains an element of bias-correction.
- However, does this mean that the mean yields are tuned to correct values by varying YGP?

Symbols show accuracy in simulation of mean yield:

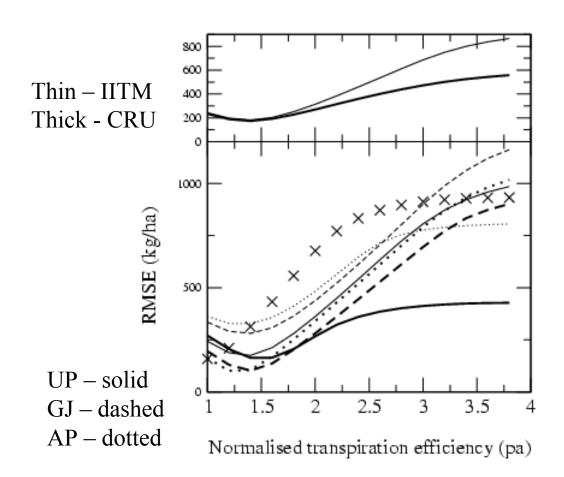
Circles: within 5% Squares: within 10%

Triangles: within 25%

+: within 50% X: within 100%



Optimisation of global parameters for groundnut across India



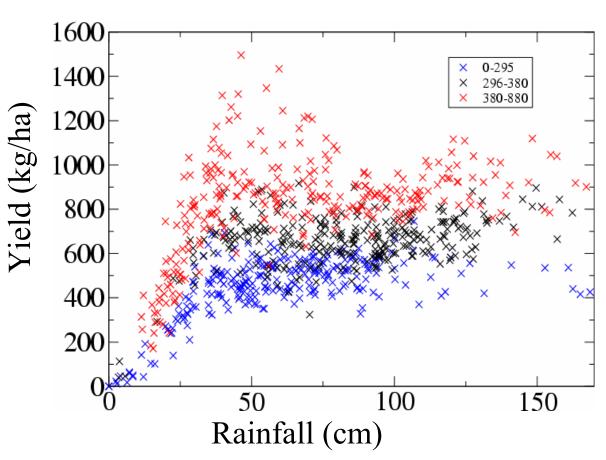
Optimal values are within literature range (~ 1.3 - 4 Pa)

Optimal values are stable over space and input dataset provided C_{YG} is calibrated

C_{YG} can correct for (some) data input bias

X – IITM-calibrated UP CRU run

Model response to rainfall and radiation



Irrigated GLAM RUE = 0.71, 0.99, 1.00 g/MJ

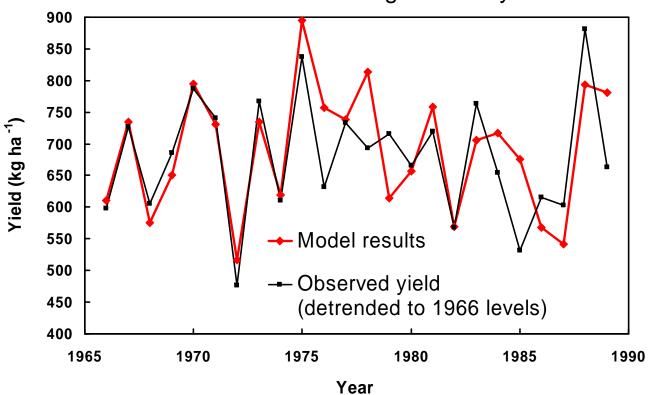
Rainfed GLAM RUE = 0.83, 0.63, 0.40 g/MJ

Observed values e.g. 0.74 (Azim-Ali, 1998), 1.00 (Hammer et. al. 1995) g/MJ

UP, AP, GJ

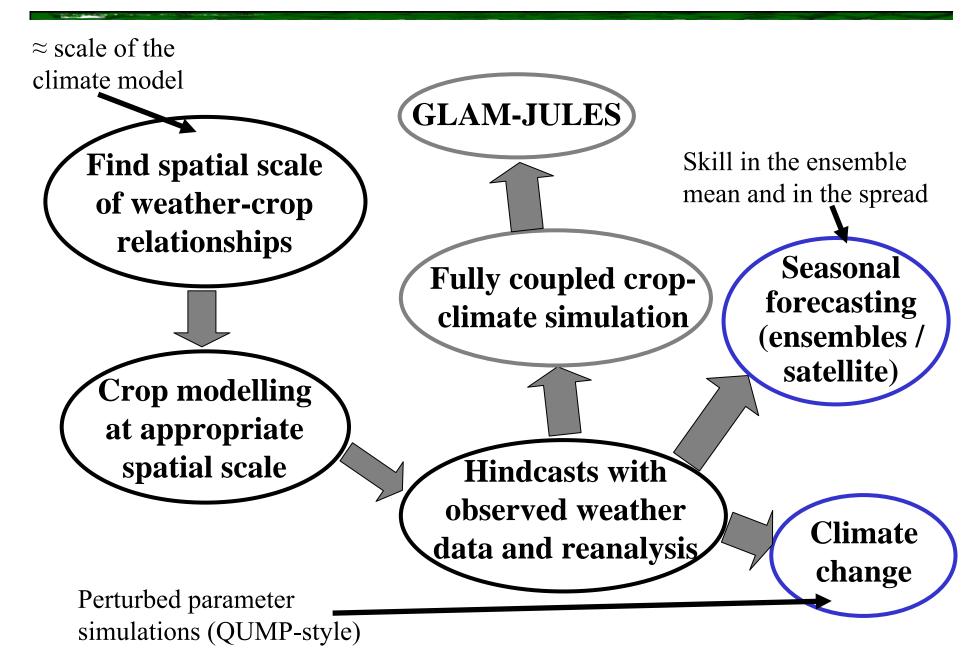
General Large-Area Model for annual crops

Results: all-India groundnut yield



Challinor, A. J., T. R. Wheeler, J. M. Slingo, P. Q. Craufurd and D. I. F. Grimes (2004). Design and optimisation of a large-area process-based model for annual crops. Agricultural and Forest Meteorology, 124, (1-2) 99-120.

Combined crop-climate forecasting



Conclusions

Land surface and boundary-layer modelling based on a sound understanding of underlying physics and biology (and chemistry) enables

- Prediction using robust process-based models
- 'Large area' modelling, which is especially useful as computer power and resolution increase
- Quantification of uncertainty due to both parameter values and model formulation
- Study of interactions between processes

General Large Area Model for Annual Crops (GLAM)

