

Ecosystem and bioclimatic modelling in a global change perspective

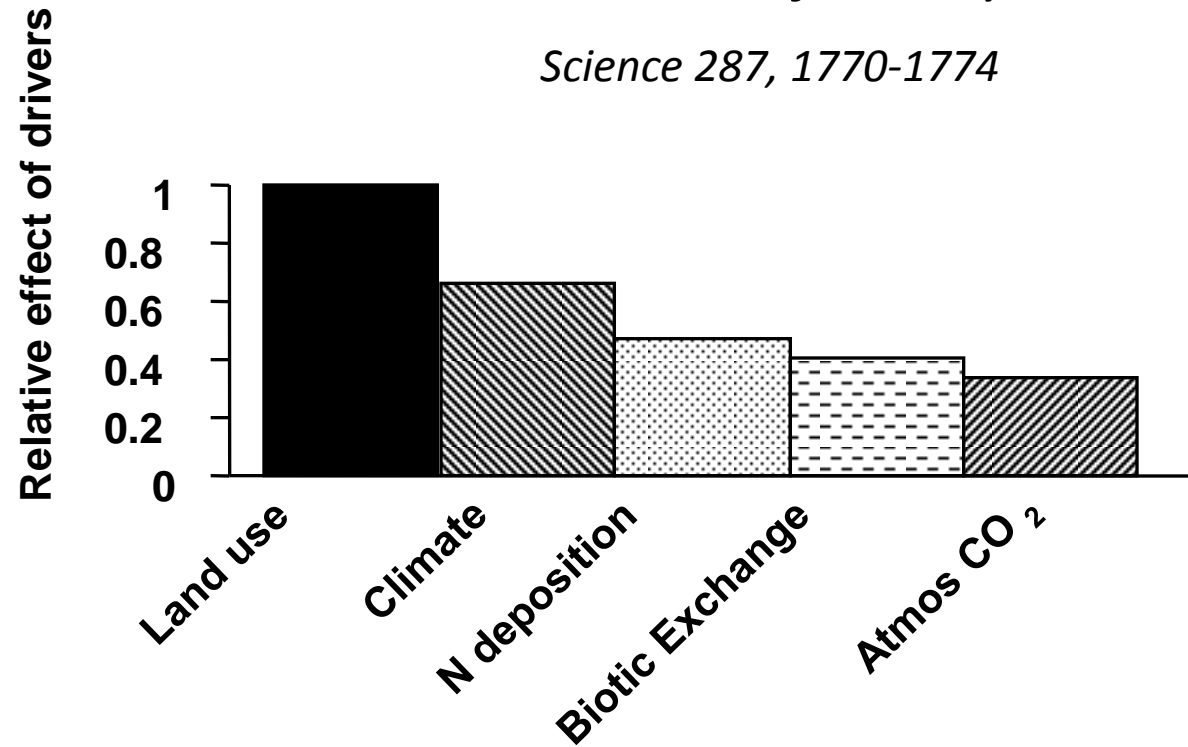
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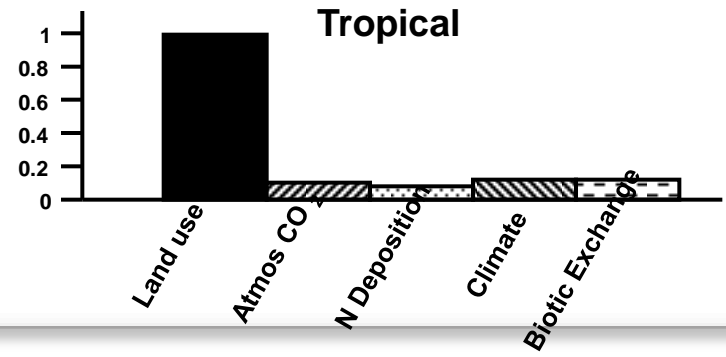
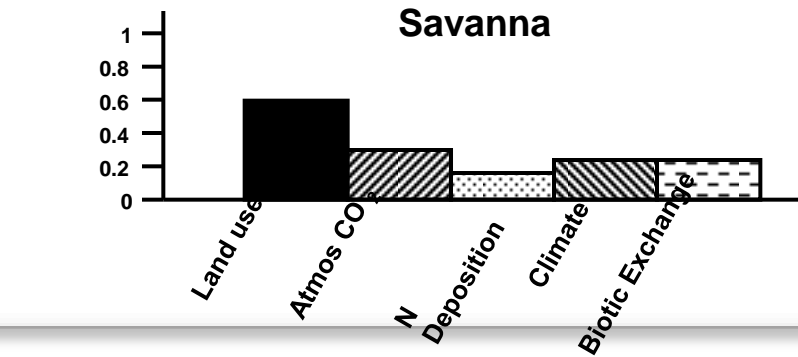
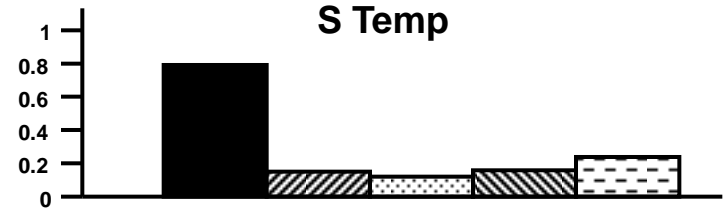
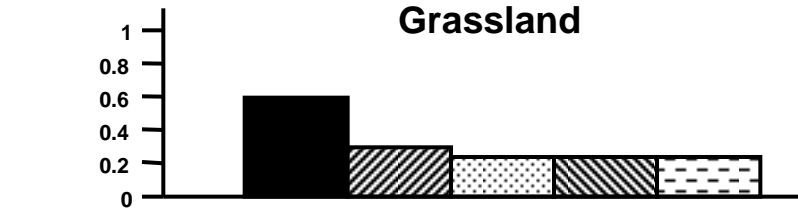
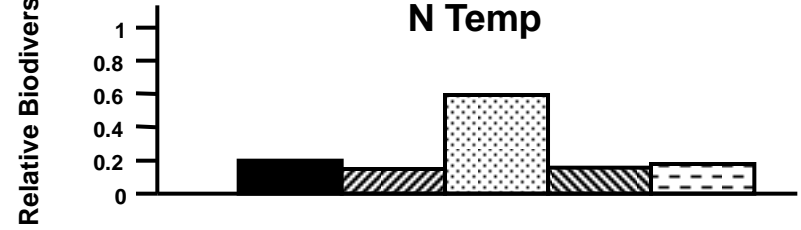
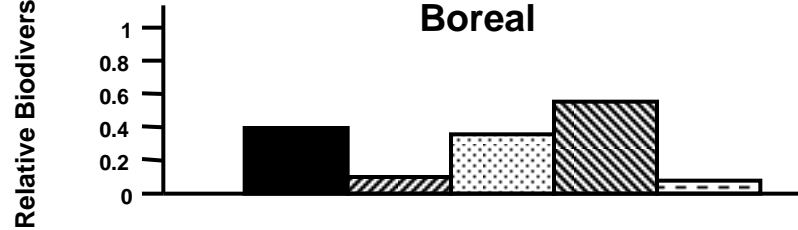
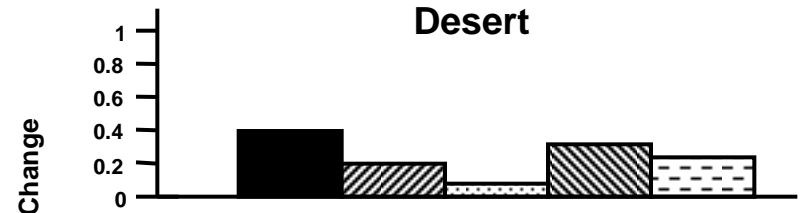
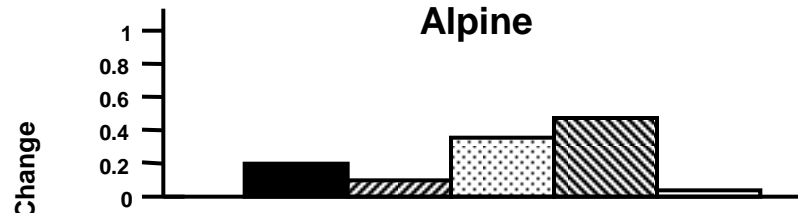
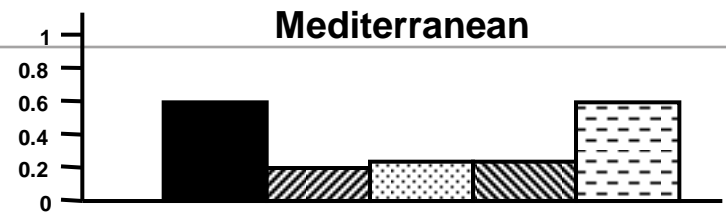
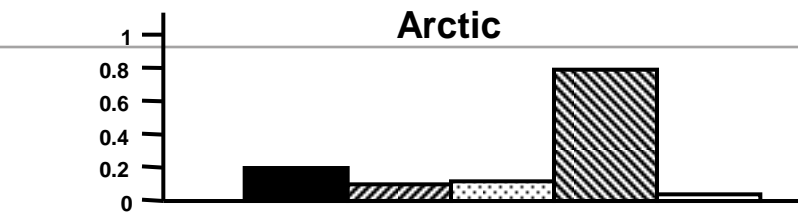
- Climate change and scenarios
- Bioclimatic modelling – statistical and process-based and examples
- Ecosystem dynamics – the role of CO₂
- Modelling ecosystem dynamics
- Examples

Contents

*Sala et al. 2000 Global biodiversity
scenarios for the year 2100*

Science 287, 1770-1774





Så kan vi få det i Europa om 100 år

Framtidens svenskar får leva i ett klimat som liknar dagens i norra Tyskland om prognoserna slår in. De kan odla paprikor och vindruvor ute. Samtidigt blir vädret hårdare för sydeuropeerna med fler svåra torrperioder och sämre odlingsmöjligheter.

NORRA EUROPA

- Varmare väder
- Mer nederbörd
- Längre tillväxtperiod

NORRA SVERIGE

- Mindre sträng kyla och mer nederbörd på vintern kan gynna skidturismen

SÖDRA SVERIGE

- Klimatet varmare, nya grödor, mer lövskog

ALPERNA

- Snögränsen flyttas högre upp. Mer än hälften av skidanläggningarna får stänga
- Tidigare snösmältning, vårflo den blir vinterflod

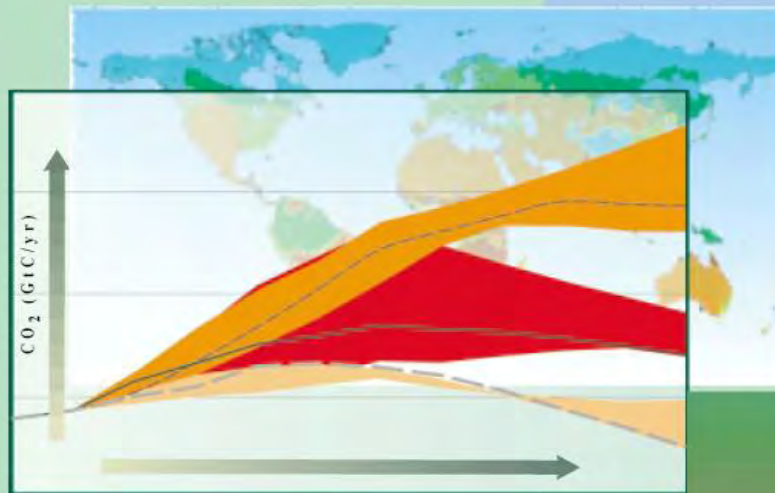
MEDELHAVSOMRÅDET

- Torka, hetta
- Flera växter utrotningshotade
- Skogsbränder
- Mindre nederbörd
- Svår vattenbrist
- Kortare tillväxtperiod

Greenhouse gas emissions



INTERGOVERNMENTAL PANEL ON CLIMATE CHANGE



IPCC SPECIAL REPORT
EMISSIONS SCENARIOS



IPCC SRES SCENARIOS

- *IPCC Special Report on Emission Scenarios*
- 4 storylines describe the relationship between the forces driving the emissions (greenhouse gases and aerosols)
- 40 Scenarios with a mix demographic, economic and technology drivers

IPCC SRES SCENARIOS

- A1 storyline: very rapid economic growth, global population peaks 2050 and then declines. New and more efficient technologies. Convergence among regions, reduction in differences in per capita income
- Includes 3 groups A1FI – fossil intensive A1T – non-fossil energy sources, A1B balanced energy

IPCC SRES SCENARIOS

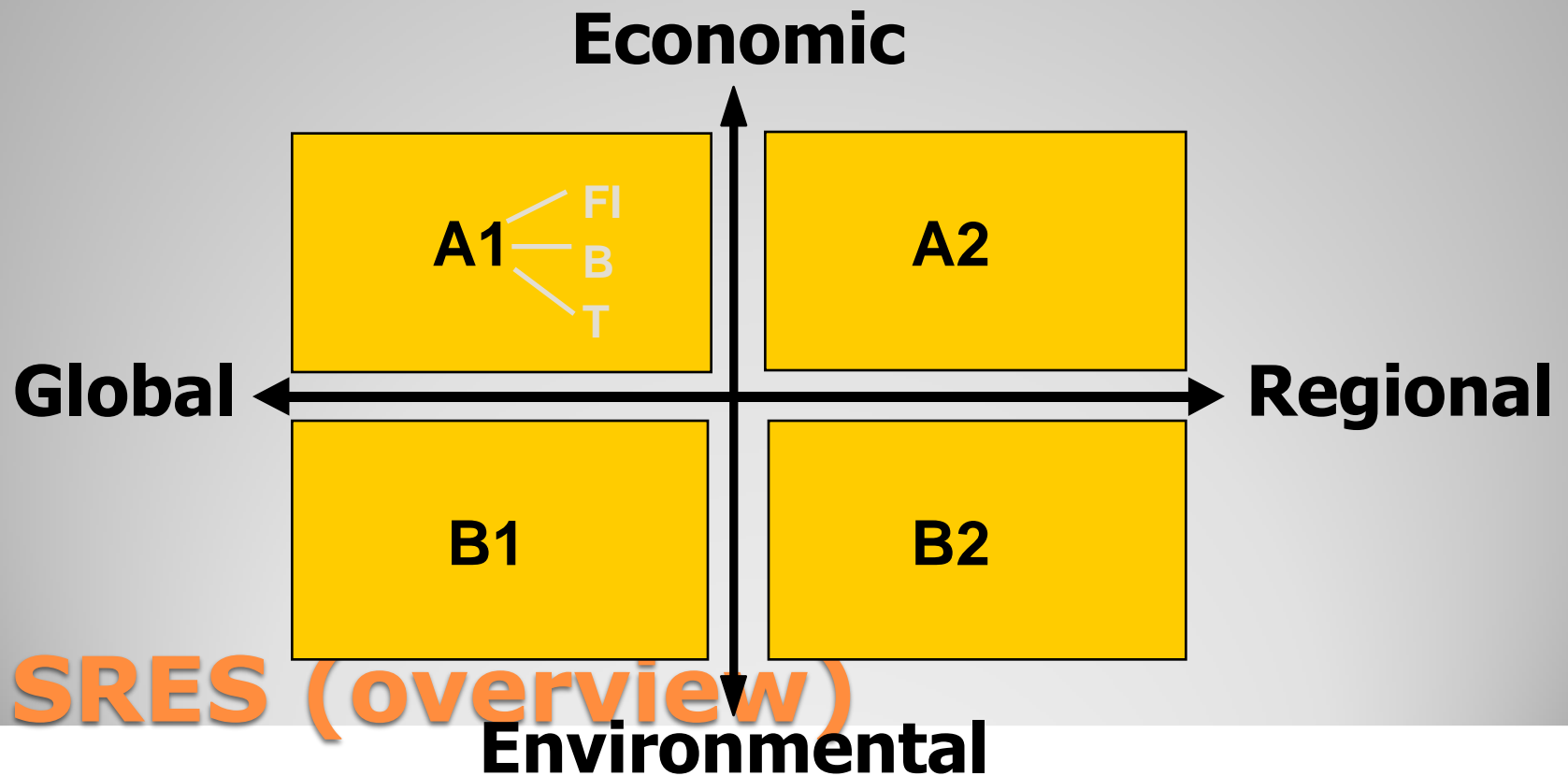
- B1 storyline – convergent world with same population peak as in A1 – but rapid changes towards a service and information economy – clean and resource efficient technologies – global solutions to economics, social and environmental sustainability

IPCC SRES SCENARIOS

- A2 storyline – a heterogeneous world, self reliance, local identities, fertility patterns converge very slowly so increasing population. Economics primarily regionally oriented, per capita growth more fragmented

IPCC SRES SCENARIOS

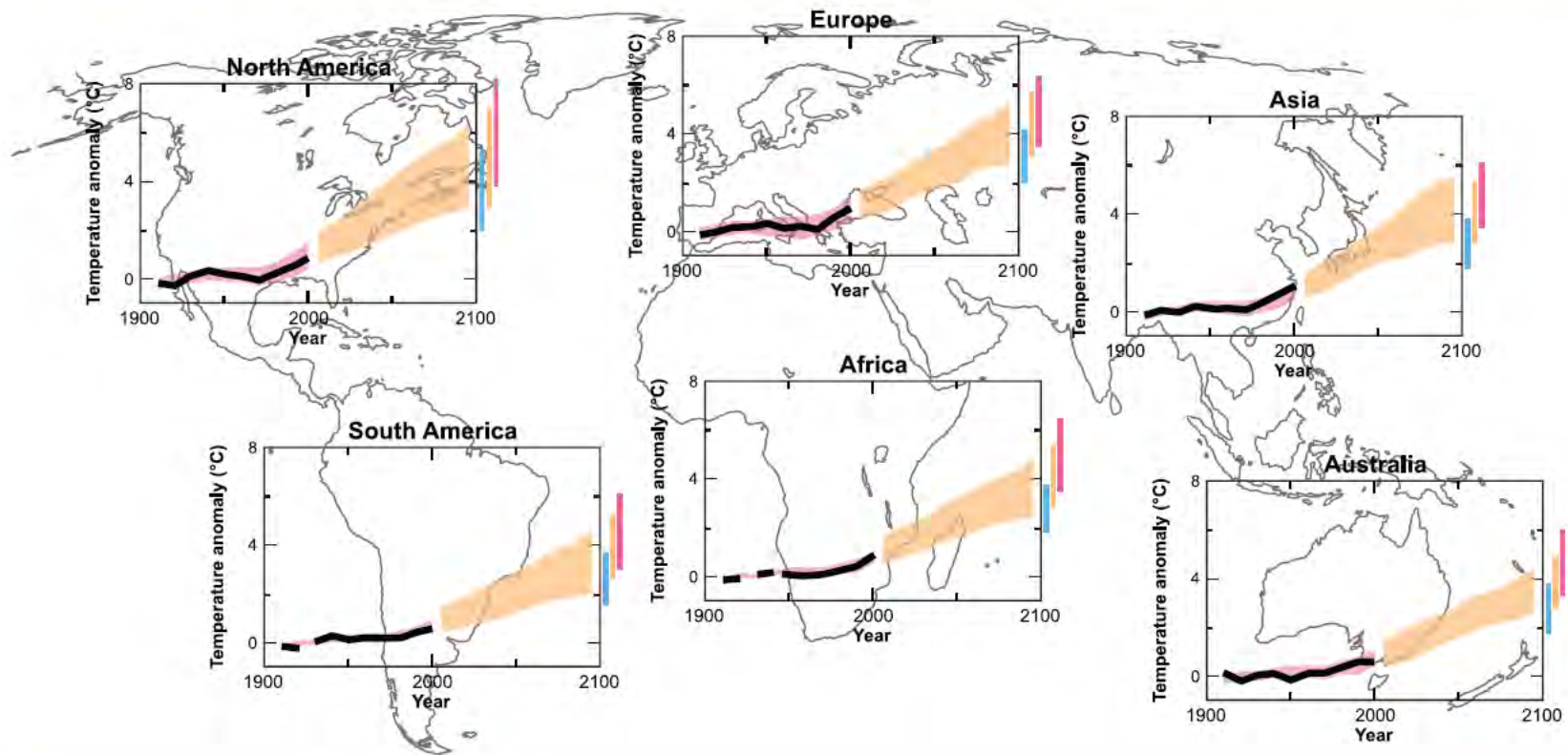
- B2 storyline – world of local solutions to economic, social and environmental sustainability. Continuously increasing population but lower than A2, intermediate levels of economic development, less rapid and more diverse technological change
- Local and regional levels



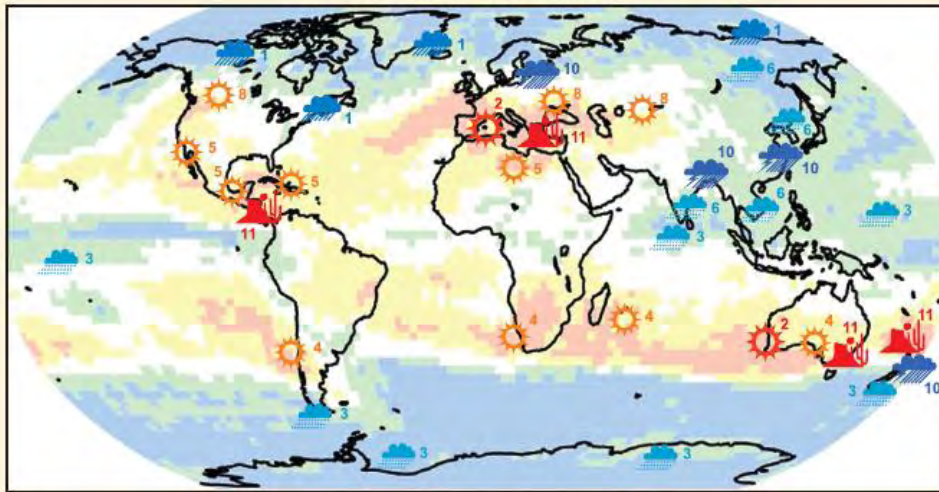
Regional Climate Projections

IPCC 2007

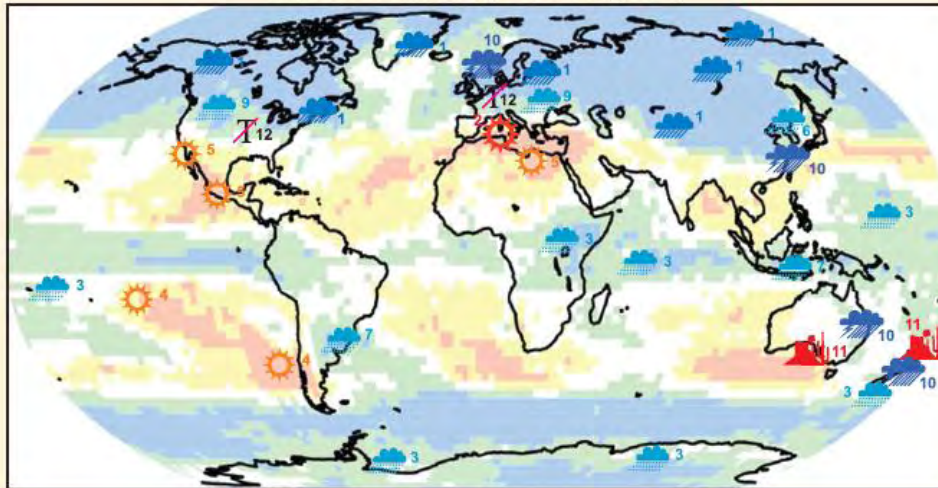
Box 11.1, Figure 1. Temperature anomalies with respect to 1901 to 1950 for six continental-scale regions for 1906 to 2005 (black line) and as simulated (red envelope) by MMD models incorporating known forcings; and as projected for 2001 to 2100 by MMD models for the A1B scenario (orange envelope). The bars at the end of the orange envelope represent the range of projected changes for 2091 to 2100 for the B1 scenario (blue), the A1B scenario (orange) and the A2 scenario (red). The black line is dashed where observations are present for less than 50% of the area in the decade concerned. More details on the construction of these figures are given in Section 11.1.2.



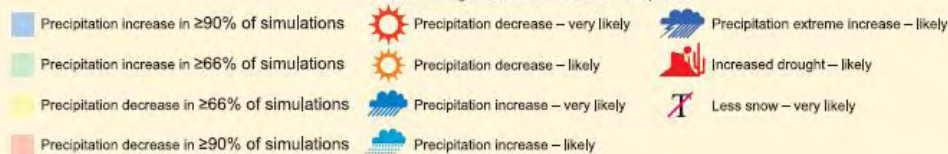
June–July–August (JJA)



December–January–February (DJF)



Based on regional studies assessed in chapter 11:



Box 11.1, Figure 2. Robust findings on regional climate change for mean and extreme precipitation, drought, and snow. This regional assessment is based upon AOGCM based studies, Regional Climate Models, statistical downscaling and process understanding. More detail on these findings may be found in the notes below, and their full description, including sources is given in the text. The background map indicates the degree of consistency between AR4 AOGCM simulations (21 simulations used) in the direction of simulated precipitation change.

Changes in precipitation (mean and extremes), drought and snow (SRES A1B).

IPCC 2007

Species respond to many aspects of the environment, but individually – they all have their own multidimensional environmental hyperspace

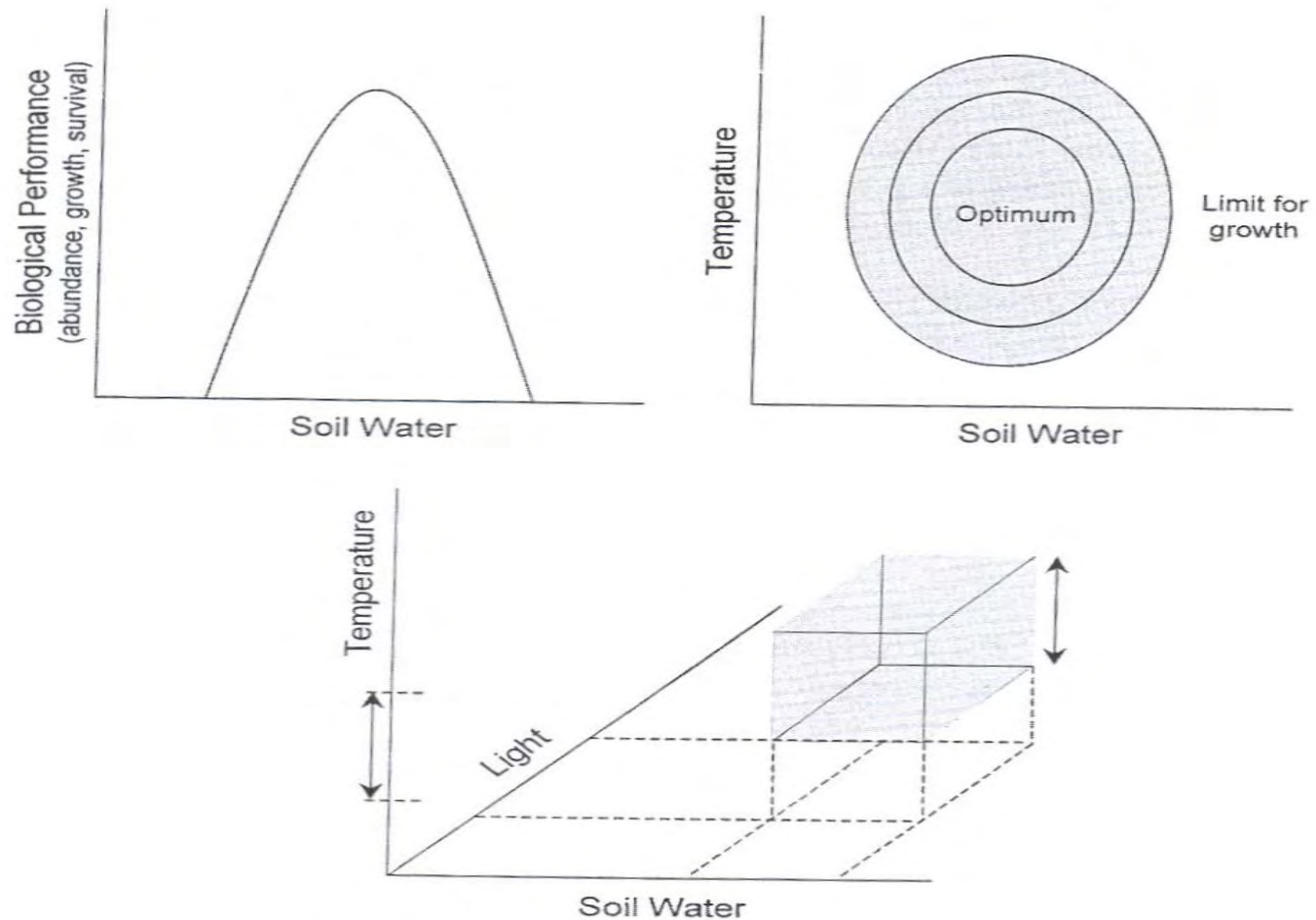


Figure 10.2. Tolerance of a species in relation to environmental conditions. Top left: Response to soil water. Top right: Response to soil water and temperature showing the range of conditions for which growth can occur and is optimal. Bottom: Range of conditions along three dimensions (soil water, temperature, light) over which growth is possible.

Risk from shifting climate space

Ohlemüller, R., Gritti, E.S., Sykes, M.T. & C.D. Thomas (2006). Towards European climate risk surfaces: the extent and distribution of analogous and non-analogous climates 1931-2100. *Global Ecology and Biogeography*, 15, 395-405.

- five biologically relevant climate variables (10' grid)
- 1945 (1931-1960) as last pre-warming reference period
- for each grid cell:
 - how much area in Europe with pre-warming analogous conditions
 - distance to these areas
 - direction to these areas

Table 1 The five primary climate variables, their units and three tolerance ranges (*TR*) investigated. Tolerance ranges define the range (\pm) within which the future climate in a grid cell was considered analogous to 1945 conditions

Code	Climate variable	Unit	Tolerance range (<i>TR</i>)		
			narrow	medium	wide
Tm	Mean annual temperature	°C	± 0.1	± 1.0	± 2.0
Tc	Mean temperature coldest month	°C	± 0.1	± 1.0	± 2.0
GDD5	Growing degree days > 5 °C	°C	± 25	± 250	± 500
Pa	Mean annual precipitation	mm	± 10	± 50	± 100
Pd	Annual water deficit	mm	± 10	± 50	± 100

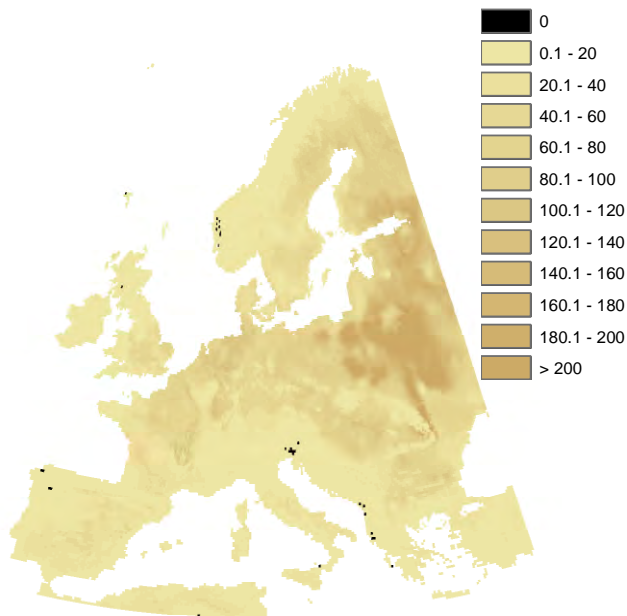
Area with pre-warming analogous conditions

1945

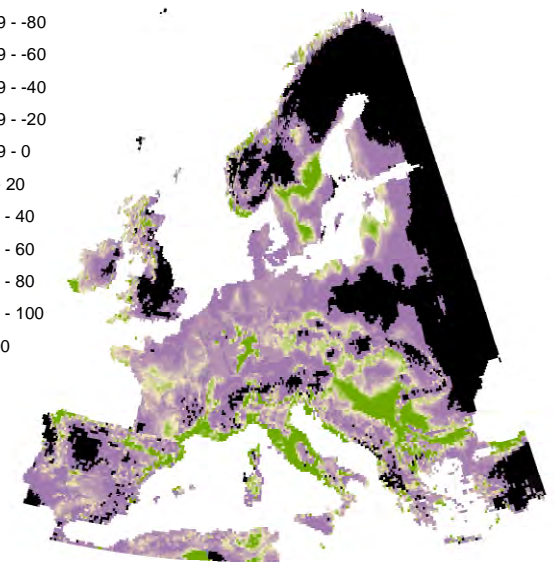
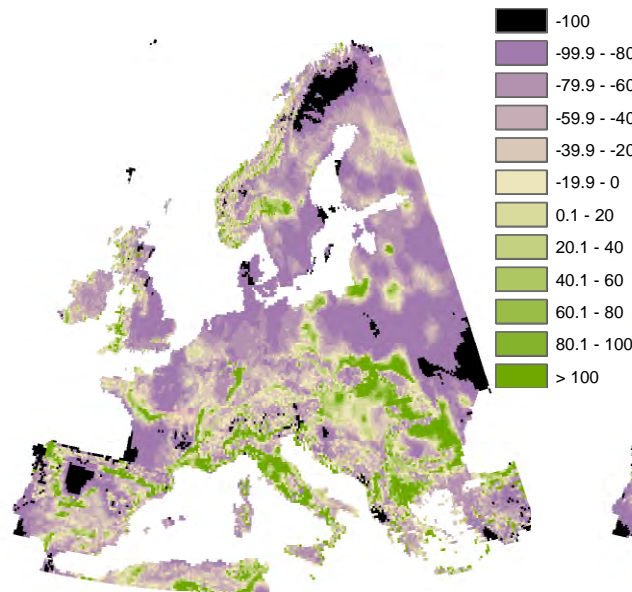
+2°C

+4°C

1945-analogous area [10^3 km^2]



Change [%] in
1945-analogous area



■ = disappearing climates

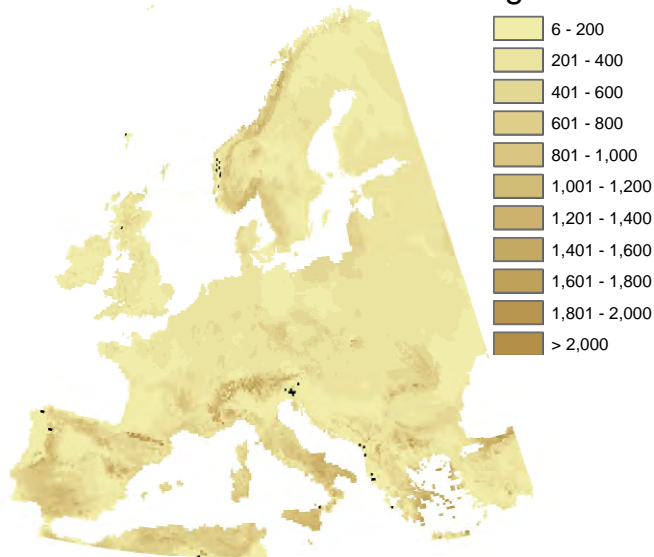
Distance to areas with analogous climates

1945

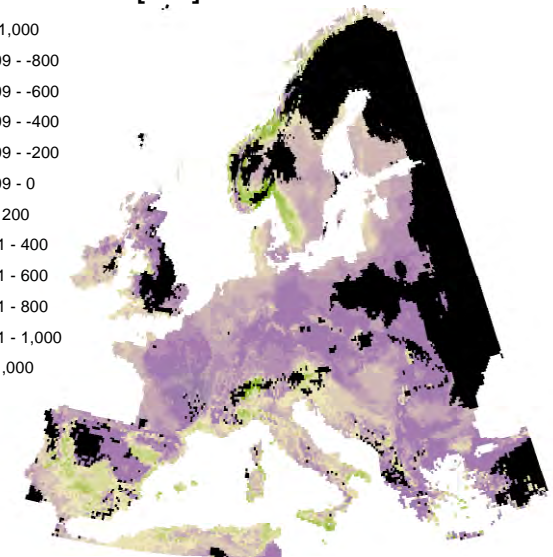
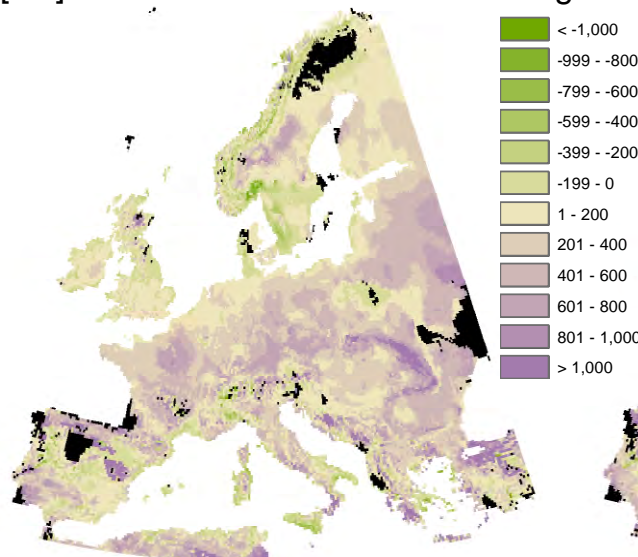
+2°C

+4°C

Mean distance to
1945-analogous cells [km]



Change in mean distance to
1945-analogous cells [km]



■ = disappearing climates

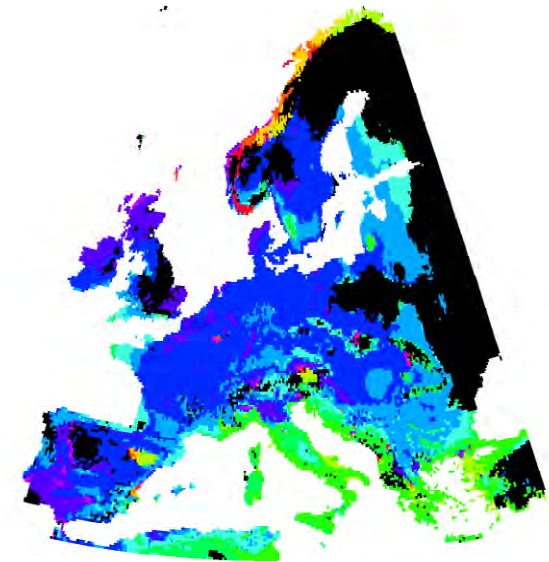
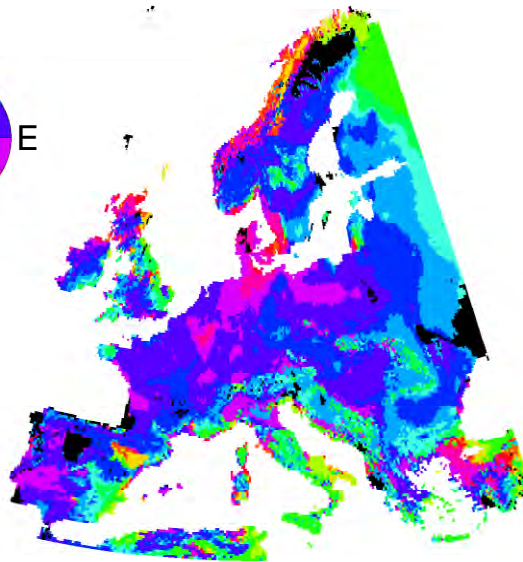
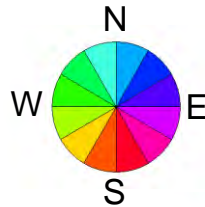
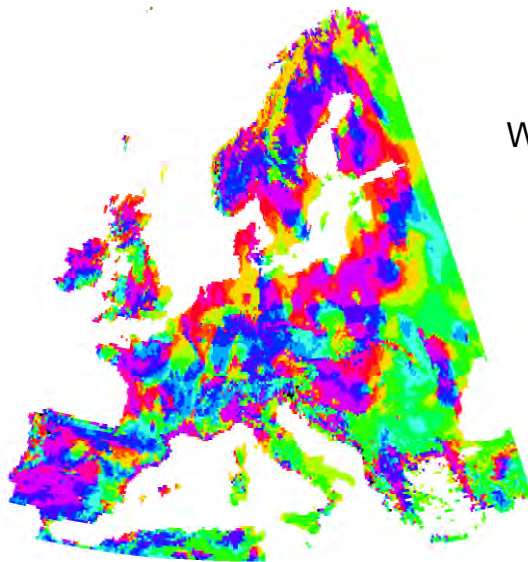
Direction to areas with analogous climates

1945

+2°C

+4°C

Mean direction to
1945-analogous cells [°N]



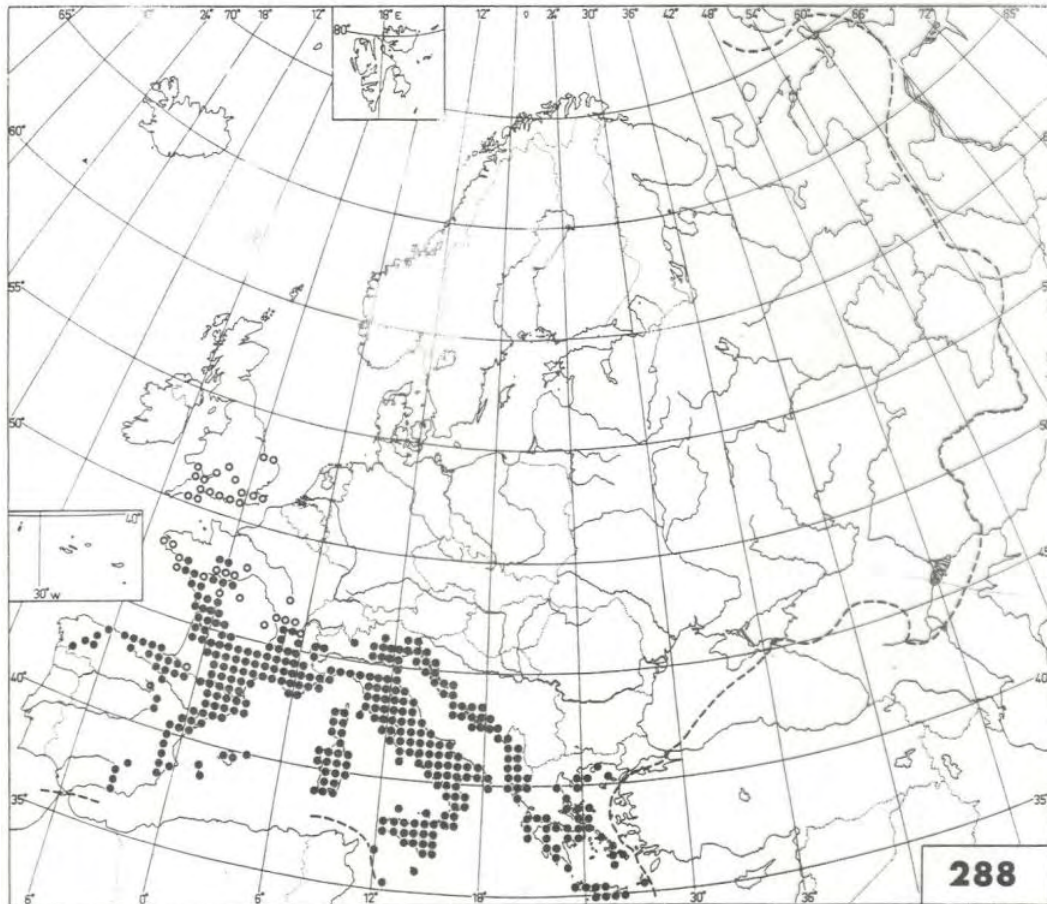
■ = disappearing climates

Statistical methods to assess species responses

Bioclimatic envelope models
Species distribution models
Habitat models
Ecological niche-based modelling

- Can be used for many species (1000s) over a wide range of species e.g. plants, animals, insects
- Assumes equilibrium with “current” climate
- Uses “current” digitised species distributions
- Uses statistics on the spatial distribution of climate variables to define a climate envelope **that best describes a species “current” range.**
- This envelope used to project a species distribution under different climate scenarios

Statistical methods



Quercus ilex

Example species
dataset:
Atlas Florae Europaeae
Jalas & Suominen 1972-
1991

A digitised version



CTA/CART- Classification and regression tree analysis

GLM – Generalised linear models

GAM – Generalised additive models

Locally weighted regressions

ANN - Artificial neural networks

GARP – Genetic algorithm for rule-set prediction

MARS – Multivariate adaptive regression splines

GBM – Generalised boosted models

MAXENT – Maximum entropy method

Random Forests Analysis

Statistical methods (2) e.g.

Table 1 Examples of the statistical techniques, and their abbreviations, applied in bioclimatic envelope modelling

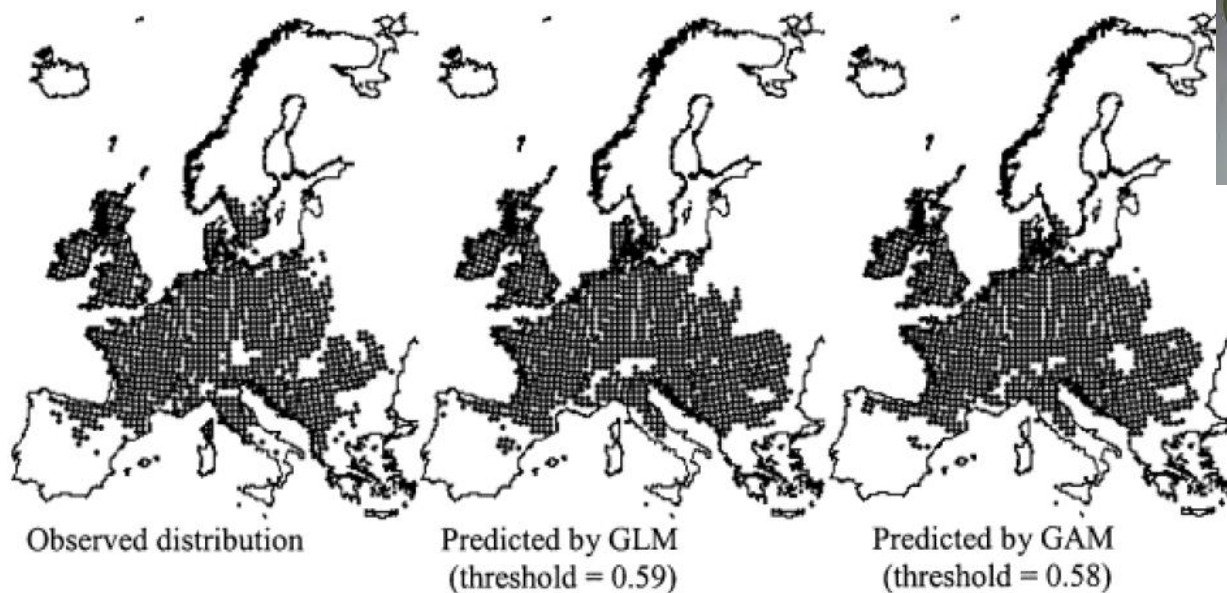
Study	Modelling methods
Brereton <i>et al.</i> , 1995; Beaumont and Hughes, 2002	BIOCLIM
Kadmon <i>et al.</i> , 2003; Meynecke, 2004; Beaumont <i>et al.</i> , 2005	
Box <i>et al.</i> , 1993; 1999; Crumpacker <i>et al.</i> , 2001	'The Florida Model'
Walker and Cocks, 1991	HABITAT
Carpenter <i>et al.</i> , 1993	DOMAIN
Baker <i>et al.</i> , 2000	CLIMEX
Skov and Svenning, 2004; Svenning and Skov, 2004	Fuzzy minimal rectilinear envelope modelling
Iverson and Prasad, 1998; 2001; 2002	Classification and regression tree analysis (CTA / CART / RTA)
Guisan and Theurillat, 2000; Price, 2000	Logistic regression/binomial GLM
Bakkenes <i>et al.</i> , 2002; Burns <i>et al.</i> , 2003	
Leathwick <i>et al.</i> , 1996; Midgley <i>et al.</i> , 2003	GAM
Araújo <i>et al.</i> , 2004; Luoto <i>et al.</i> , 2005	
Beerling <i>et al.</i> , 1995; Huntley <i>et al.</i> , 1995; 2004	Locally weighted regression (local regression/loess)
Hill <i>et al.</i> , 1999; 2002	ANN
Berry <i>et al.</i> , 2002; Pearson <i>et al.</i> , 2002; 2004	GARP
Peterson, 2001; Anderson <i>et al.</i> , 2002a; 2002b	
Peterson <i>et al.</i> , 2002a; 2002b; 2004	
Prasad and Iverson, 2000	MARS
Gavin and Hu, 2005	GM-SMAP
Thuiller, 2003; 2004; Araújo <i>et al.</i> , 2005a; 2005b	GLM, GAM, CTA, ANN
Thuiller <i>et al.</i> , 2005a; 2005b	

ANN = artificial neural networks; GAM = generalized additive models; GARP = genetic algorithm for rule-set prediction; GLM = generalized linear models; GM-SMAP = Gaussian mixture distributions and multiscale segmentation; MARS = multivariate adaptive regression splines.

Heikkinen, R.K., Luoto, M., Araújo, M., Virkkala, R., Thuiller, W., & Sykes, M.T.. 2006. Methods and uncertainties in bioclimatic modelling. *Progress in Physical Geography*, 30. 751-777

- Measure accuracy of prediction – preferably based on independent data e.g. Kappa **statistic**, Cohen's κ , AUC values of the ROC statistic, etc
- Comparison of modelling techniques
- **Approaches to account for model predictions' variability**
 - use a framework of different methods
 - use a consensus analysis (via PCA or cluster analysis) that represents the central tendency across all models considered.

Evaluating Performance



Global Change Biology (2003) 9, 1353–1362

BIOMOD – optimizing predictions of species distributions and projecting potential future shifts under global change

WILFRIED THUILLER

Centre d'Ecologie Fonctionnelle et Evolutive, Centre National de Recherche Scientifique, 1919 route de Mende, 34293 Montpellier Cedex 5, France

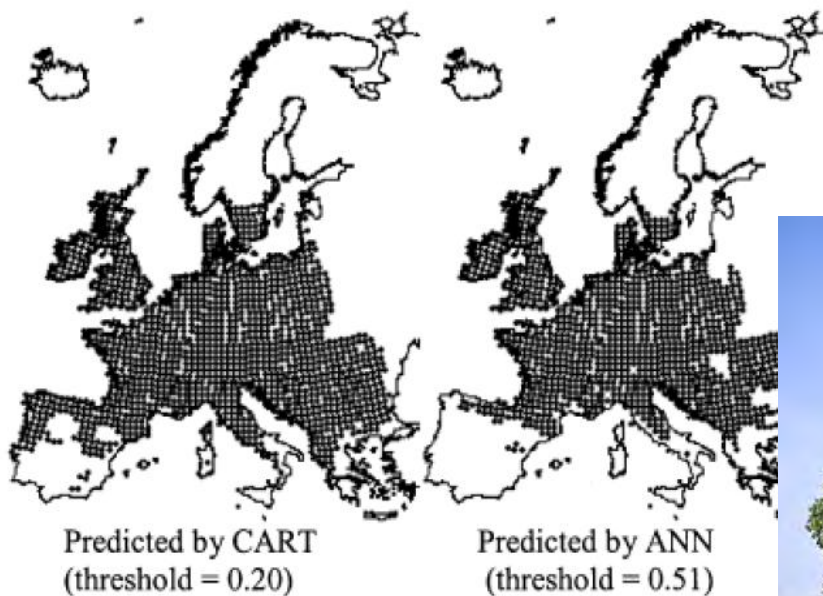
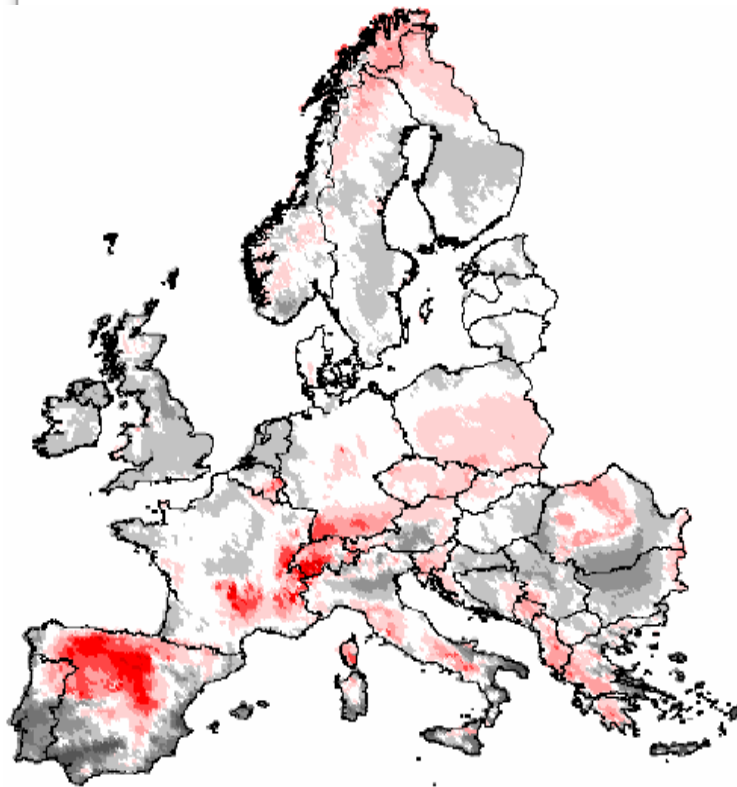


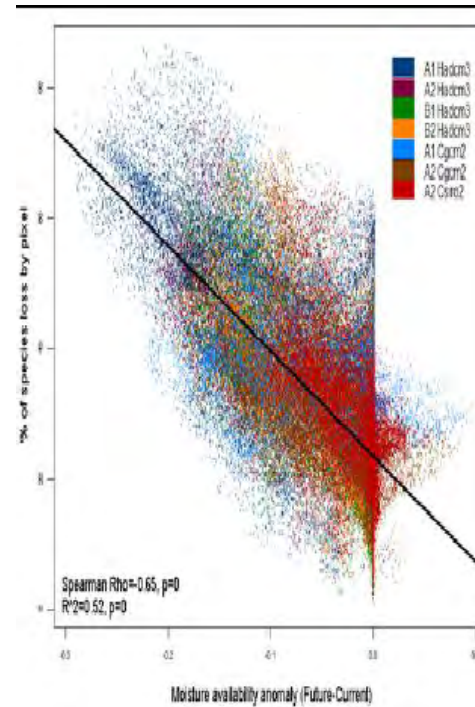
Fig. 1 Observed and predicted distribution for *Quercus petraea* according to the modelling technique used. The ROC curve and κ statistics on evaluation data are the following: GLM (AUC = 0.939; κ = 0.74), GAM (AUC = 0.945; κ = 0.77), CART (AUC = 0.927, κ = 0.76) and ANN (AUC = 0.960; κ = 0.79).



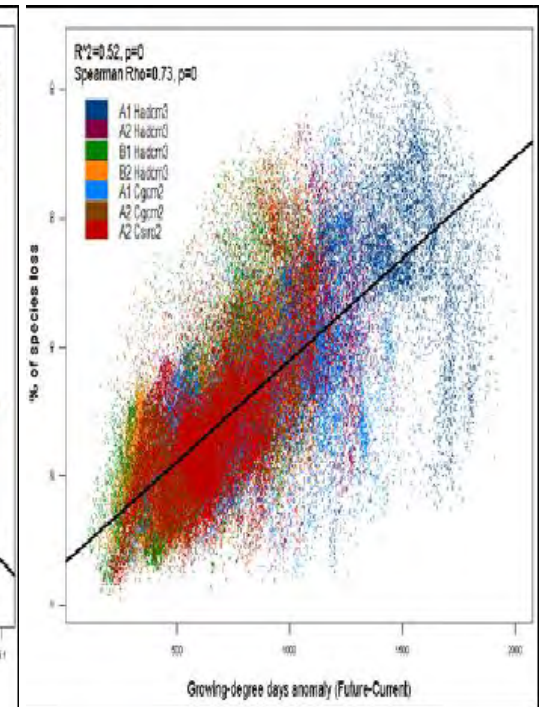
Statistical based bioclimatic envelope approaches e.g. BIOMOD



Regional projections of residuals from the multiple regression of species loss against GDD and moisture availability index - RED excess species loss, grey a deficit



Moisture availability anomaly



Growing degree days anomaly

Relationships between the % of species loss and anomalies of moisture availability & GDD – different colours different climate change scenarios

1350 plant species, climate: annual, winter, summer precip, annual & min temp. GDD, moisture avail. Index in each 50X50 grid. GLM GAM, CART, ANN % consensus PCA

Climate change threats to plant diversity in Europe

Thuiller, W., Lavorel, S., Araújo M.B., Sykes, M.T. & Prentice I.C. 2005

Proceedings National Academy of Sciences of the United States of America (PNAS) 102 8245-8250

Different measures of extinction risk associated with shifting climates by 2095

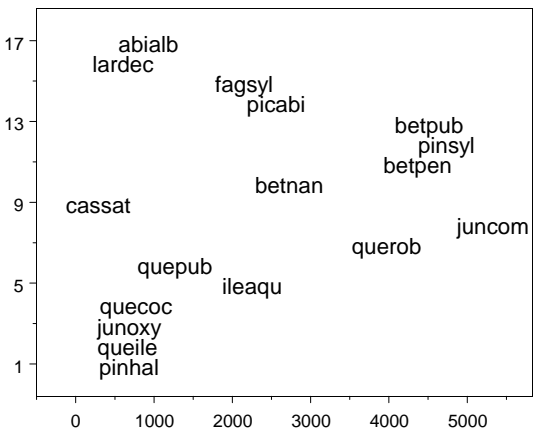
Ohlemüller, R., Gritti, E.S., Sykes, M.T., & Thomas, C.D. (2006). Quantifying components of risk for European woody species under climate change. *Global Change Biology*, 12, 1788-1799.

10'x10' climate grid, GAM, CART & GLM models - mostly good agreement between models (showing GAM) – comparing 1945 climate data and species p/a and then used to predict 1995, 2045, 2095 with HADCM3 GCM (scenarios B1/A1Fi)

Risk measure I

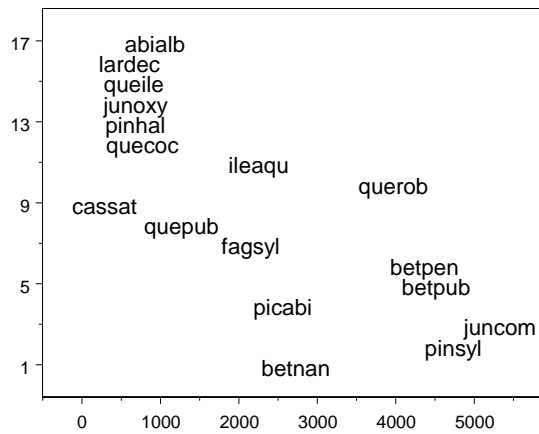
Loss in climatically suitable area

highest risk
↑
lowest risk



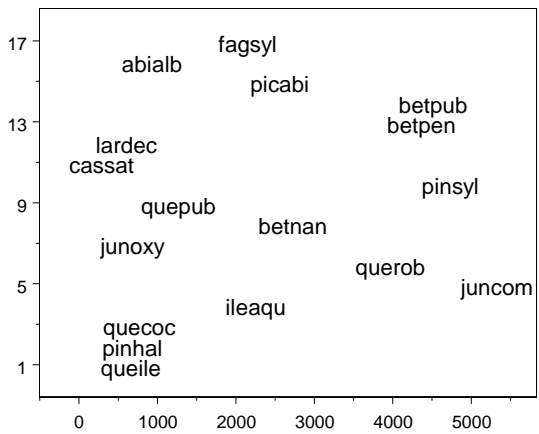
Risk measure II

Increase in distance to climatically suitable areas (2095)



Risk measure III

Decrease in climatic suitability in already occupied locations



present-day range size [10^3 km^2]



Abies alba (silver fir)

© Renzo Motta

Risk measure I

Loss in climatically
suitable area

Risk measure II

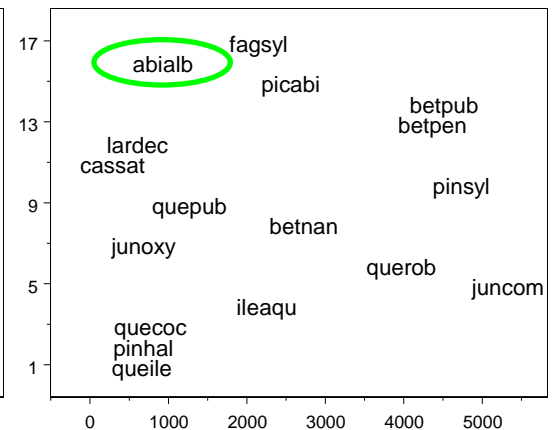
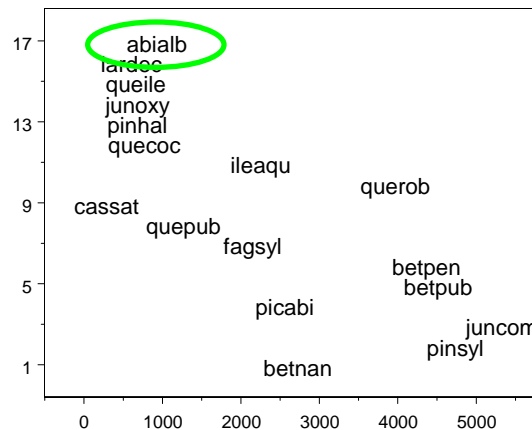
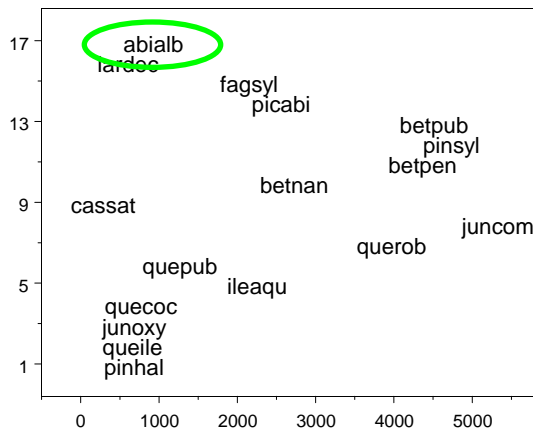
Increase in distance to
climatically suitable
areas

Risk measure III

Decrease in climatic
suitability in already
occupied locations

highest
risk

lowest
risk



present-day range size [10^3 km²]



Abies alba (silver fir)

Quercus ilex (holm oak)



© Brian Ecott

Risk measure I

Loss in climatically
suitable area

Risk measure II

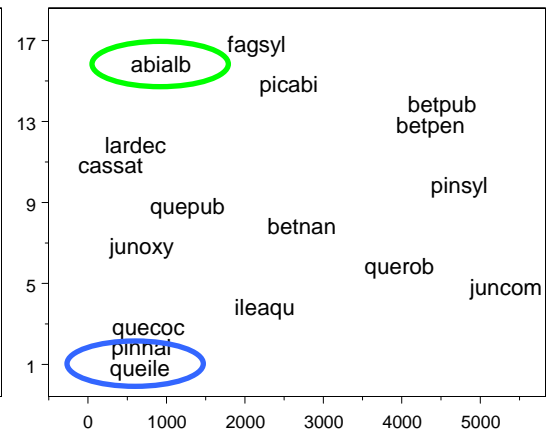
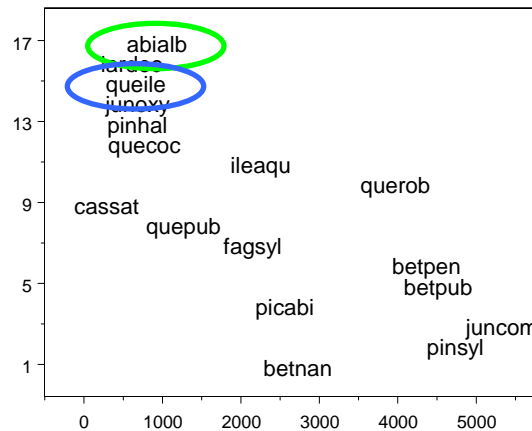
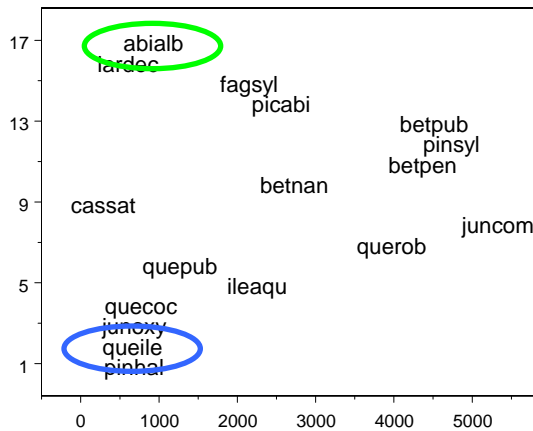
Increase in distance to
all climatically suitable
areas

Risk measure III

Decrease in climatic
suitability in already
occupied locations

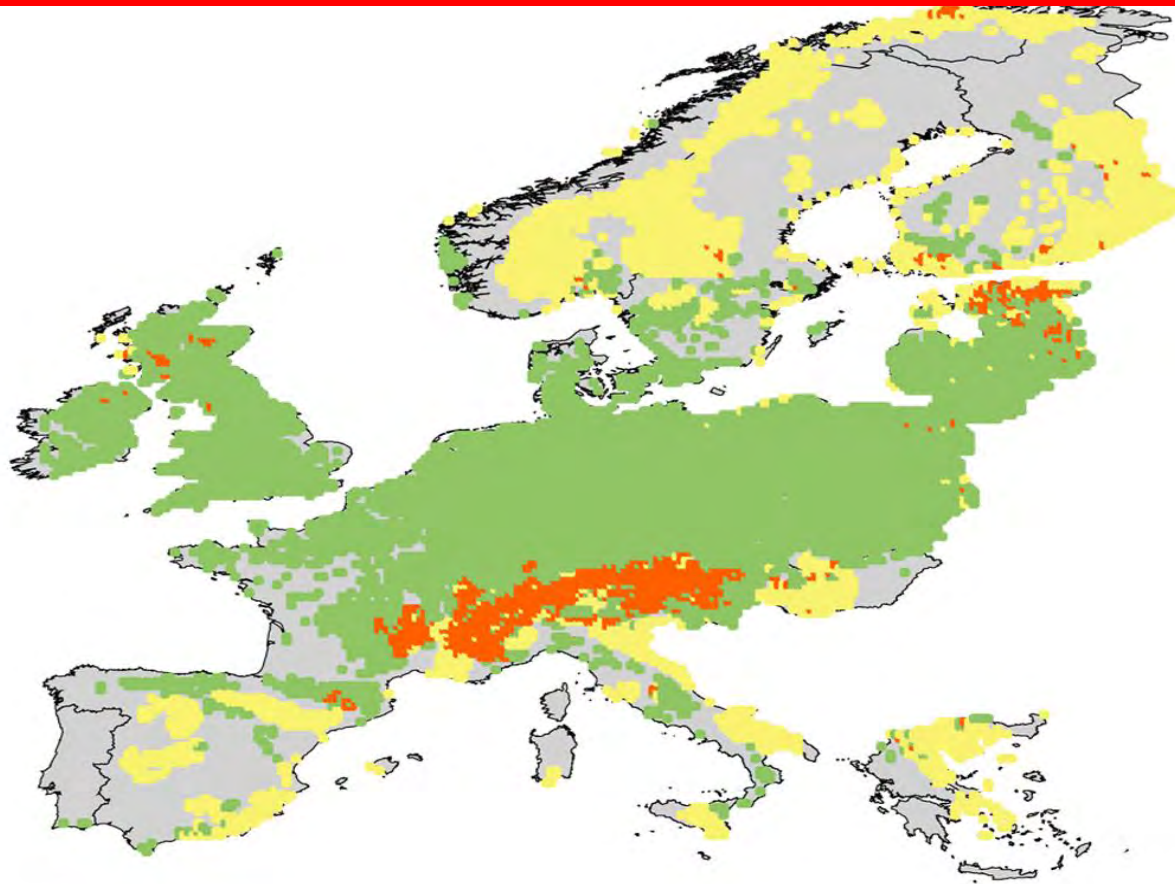
highest
risk

lowest
risk



present-day range size [10^3 km^2]

Current niche spaces of *Boloria titania* and *Polygonum bistorta*



- *P. bistorta*
- *B. titania*
- Overlap



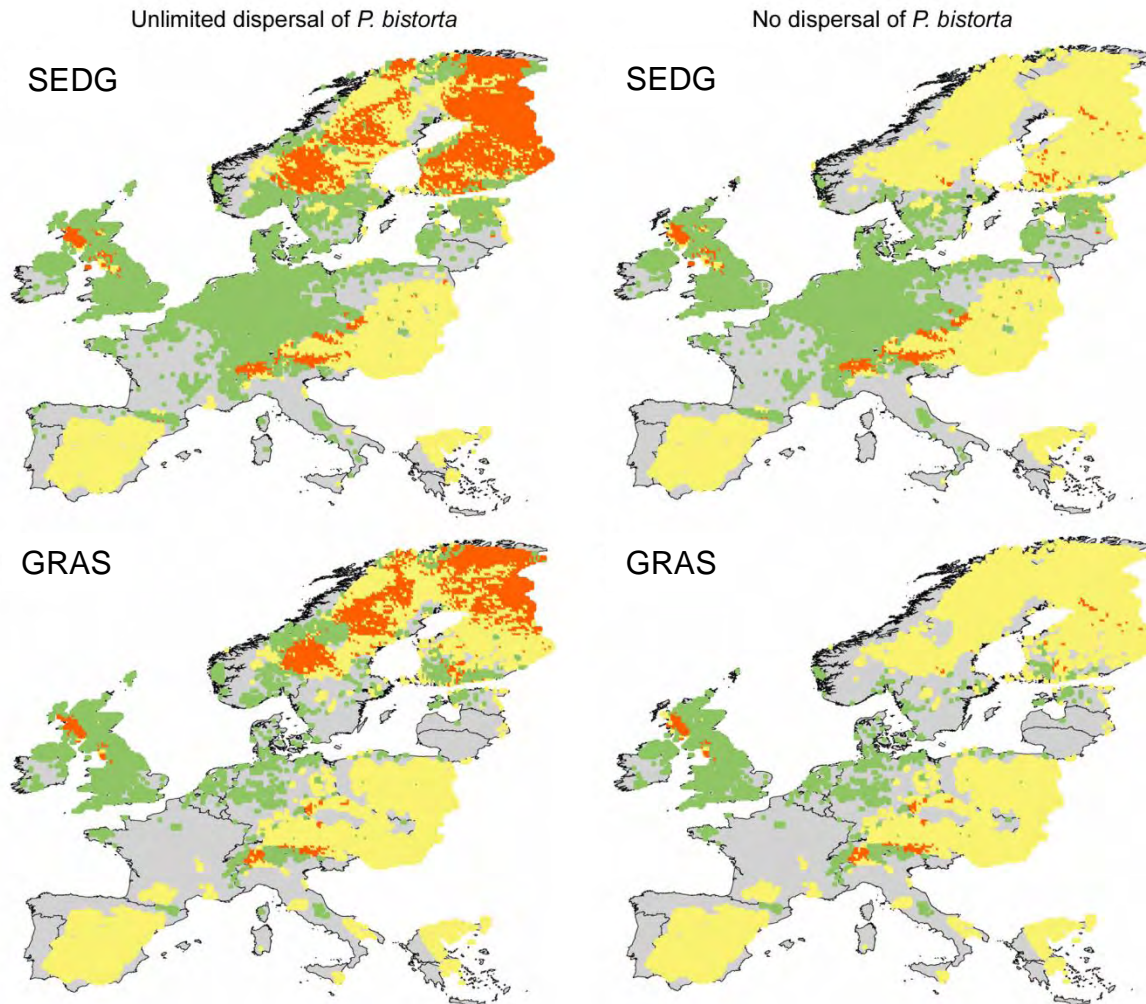
Two separate ecological niche models for the butterfly and its host plant based on climate, land use and soil characteristics.
AUC's > 0.93

➤ Niche spaces of *Boloria titania* and *Polygonum bistorta* show today only small areas of overlap.

Schweiger, O., Settele, J., Kudrna, O., Klotz, S. Kuhn, I. Climate change destabilises ecosystem functioning in press



Projected changes in niche space of *Boloria titania* and *Polygonum bistorta* for 2080



- Niche space of *B. titania* would increase
- But mismatch of both niche spaces increases, too
- Serious decrease of overlap in the butterfly's original distribution (Alps and Baltic States).
- Potentially new areas in the north could only be colonised in the unlikely case of high dispersal ability of *P. bistorta*.
- Climate change can disrupt trophic interactions.

● *P. bistorta*
● *B. titania*
● Overlap



- See Thuiller 2003, 2004 GCB, Thuiller et al. 2003 JVS, Segurado & Araújo 2004 , Heikkinen et al 2006 etc

No single method that can be used with reliability in the majority of studies – but some approaches are more reliable (e.g. GAM, ANN, MARS)

Many models are species-dependent and thus using one or a few models over many species is fraught

Variability between methods greater under future climate scenarios

Multi-**model frameworks** to find the “best” model – the consensus approach might be better – BUT?

**So what about statistical models
any good ?**

1. Model dependent results

- Are species in equilibrium with current climate ?
- Model dependent results
- Climate data/land use data uncertainties
- Sampling a species climatic range ??
- Downscaling
- Bioclimatic parameter choice
- Problems of scale
- Stochastic events, extreme events
- Non-analogue futures
- Extinction lags
- Genetic differences at range boundaries
- Fragmentation
- Dispersal/migration
- The effect of changing CO₂ (fertilisation and water balance)
- Changing competitive relationships
- Increasing nitrogen deposition
- Changing phenology
- Invasive species

BIG question – Are broad scale results relevant for Biodiversity conservation policy??

2. Many uncertainties and assumptions

- Phenology models
- Bioclimatic models /BIOME/SPECIES range shift models

Physiologically based equilibrium approaches

Phenology- timing of events through the year.

Climate driven changes:

- e.g. earlier budburst/leaf drop

- earlier flowering times

- earlier appearance of
butterflies

- earlier breeding in birds and
amphibians

- earlier arrival of migrating birds

Phenology

Sensitivity analysis of the tree distribution model PHENOFIT to climatic input characteristics: implications for climate impact assessment

XAVIER MORIN and ISABELLE CHUINE

Centre d'Ecologie Fonctionnelle et Evolutive, Equipe BIOFLUX, CNRS, 1919 route de Mende, 34293 Montpellier cedex 5, France

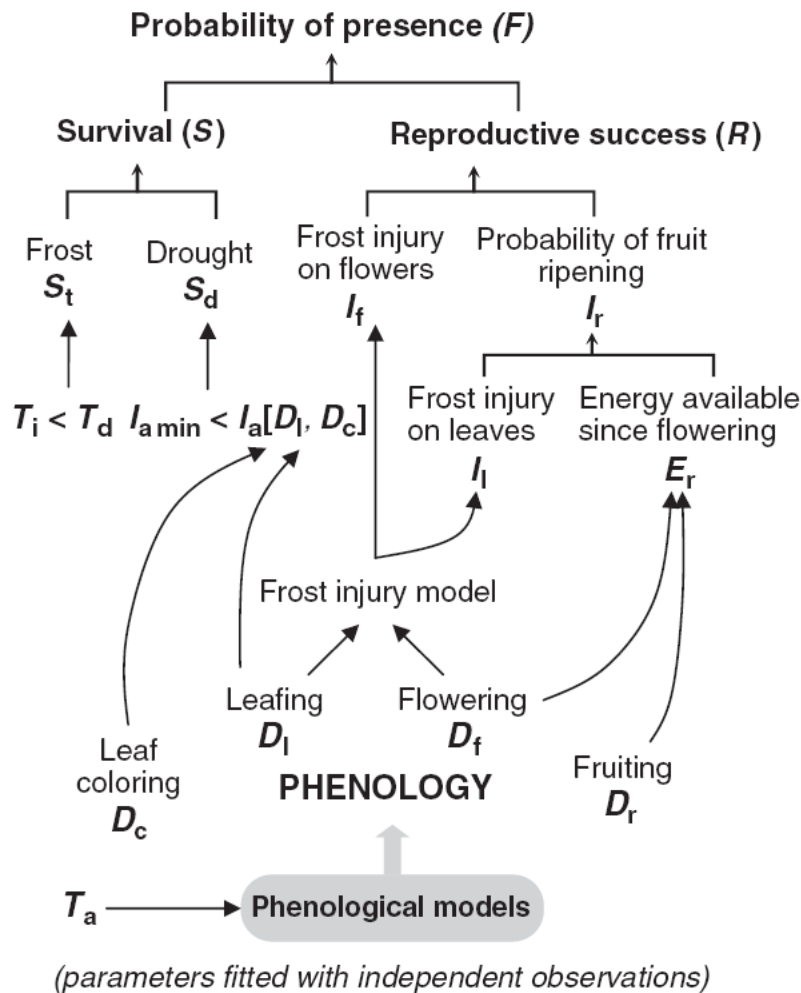


Fig. 1 Description of PHENOFIT. T_a , average daily temperature; T_i , minimum daily temperature; D_l , date of leafing; D_f , date of flowering; D_r , date of fruiting; D_c , date of leaf coloring; $I_a[D_l; D_c]$, moisture index between leafing and leaf coloring; $I_{a\min}$, minimal moisture index (species dependent); I_l , leaves frost injury index; E_r , available energy since flowering; I_f , flowers frost injury index; I_r , index of fruit maturation; S , probability to survive; S_t , probability to survive frost; S_d , probability to survive drought.

PHENOFIT – a process based model that predicts species distributions
Based on principle that adaptation of a tree species to its environment strongly depends on the synchronization of its development timing to seasonal variations in climate.

Output – a probability of presence of an adult individual after several years
Calculated as the product of its probability to survive until the next reproductive season and to be able to produce viable seeds – reproductive success)

Chuine & Beaubien 2001: Phenology is a major determinant of temperate tree range Ecol Letts 4:500-510

PROCESS-BASED MODELING OF SPECIES' DISTRIBUTIONS:
WHAT LIMITS TEMPERATE TREE SPECIES' RANGE BOUNDARIES?

XAVIER MORIN,^{1,3} CAROL AUGSPURGER,² AND ISABELLE CHUINE¹

Ecology, 88(9), 2007, pp. 2280–2291
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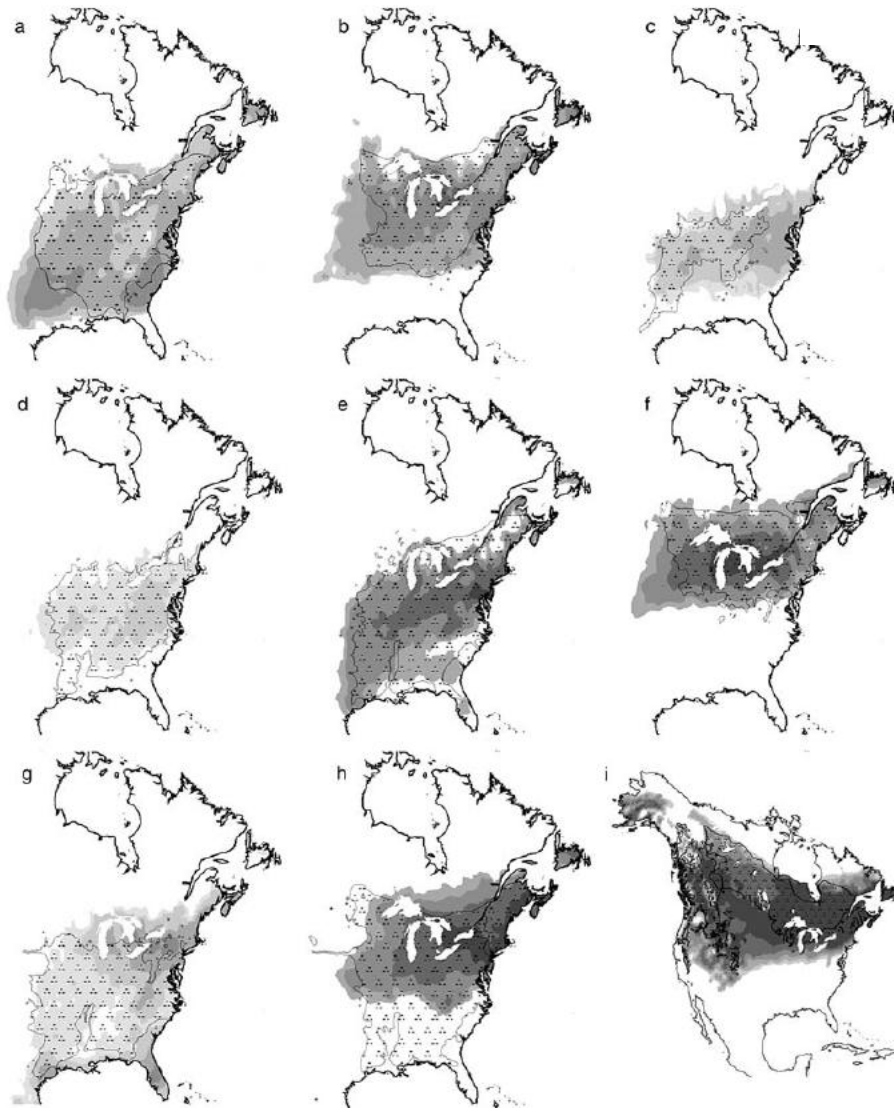


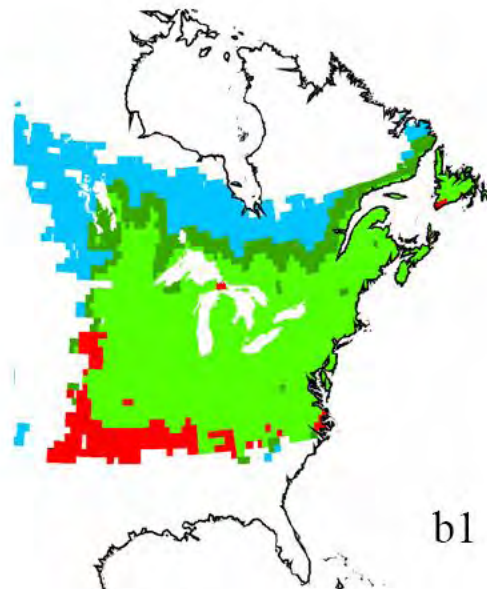
FIG. 1. Probability of presence (above SPT [specific presence threshold], see Table 1) simulated with PHENOFIT (Chuine and Beaubien 2001) for: (a) *Acer saccharinum*; (b) *Acer saccharum*; (c) *Aesculus glabra*; (d) *Carya ovata*; (e) *Fraxinus americana*; (f) *Fraxinus nigra*; (g) *Juglans nigra*; (h) *Ostrya virginiana*; (i) *Pinus contorta*; (j) *Pinus monticola*; (k) *Populus deltoides*; (l) *Populus tremuloides*; (m) *Quercus bicolor*; (n) *Quercus macrocarpa*; (o) *Salix nigra*; (p) *Sassafras albidum*; and (q) *Ulmus americana*. Note that the distribution of *Pinus banksiana* (sister species of *Pinus contorta*) is also shown in panel (i) on the eastern part of the continent. Values can range from 0 to 1. Area with triangles corresponds to the observed species' distribution.

Probability of presence of
different species simulated
with PHENOFIT
Area with Δ s observed ranges

Phenology model

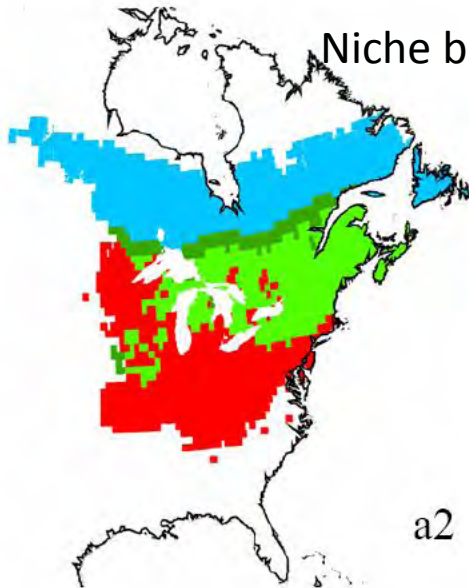


a1

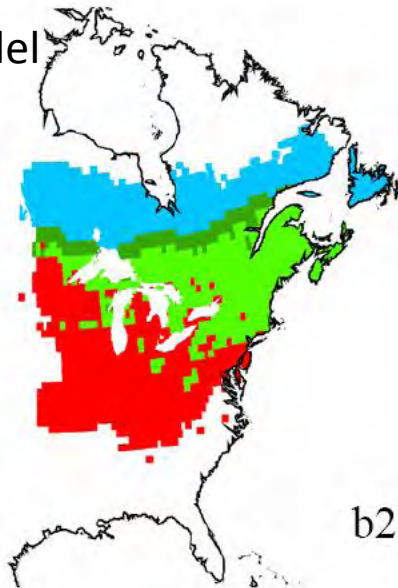


b1

Niche based model



a2



b2

- Extinction in 2100
- Presence conserved in 2100
- Realized colonizations in 2100
- Suitable zones in 2100

Comparison of species probability presence between 2100 & 2000 for e.g. *Acer saccharum*

IPCC SRES a: scenario A2
b: scenario B2

Greater possible extinctions in NBM's due to not taking phenotypic plasticity or local adaptation into account

Trees in a changing climate – shifts in species ranges under climate change – reducing predictions' uncertainty through comparison of niche and process-based modelling approaches

- BIOME (PFTs) – plant functional types – generalised grouping based around function – global maps – BIOME 1, 3, 4
- STASH species used currently plants and then currently less than 40 – regional/continental maps
- These approaches are concerned with responses to the environment at a physiological level
- Assumes a number of bioclimatic variables to be important for a species survival e.g. winter cold kills – (mean coldest month), e.g. drought reduces growth (AET/PET) etc
- Still basically correlative
- Some added features related to amount (NPP – Biome3) or likely degree of establishment at any site (STASH)

**More physiologically-based
range shift/bioclimatic
equilibrium approaches**

- One of the first bioclimatic models – a model of individual species distributions and productivity
- **Using 12 monthly values of mean temperature, precipitation and cloudiness –**
- **Downscaled to daily values to get bioclimatic variables and growth multipliers**
- **Defines grid climate and compares species bioclimatic requirements to define p/a and likely productivity**
- **Gridded output at** regional to continental scale

- Bioclimatic variables representing distinct *physiological limiting* mechanisms:
- Minimum temperature tolerance (mean coldest month temp;
- Growing season length - complete life cycle (GDD);
- Chilling requirements for budburst e.g. Beech;
- Drought tolerance AET/PET

Bioclimatic Equilibrium Modelling – STASH

(Sykes, Prentice & Cramer 1996, Sykes & Prentice 1995, Walther, Birger & Sykes 2005, Giesecke et al. 2007, Walther et al. 2007)

- A period of chilling, before budburst can take place, is required for some deciduous boreal and temperate trees e.g. *Fagus sylvatica*
- Presumed to be an adaptation to weather variability and late frosts
- Use the negative exponential relationship between GDD to budburst and length of the chilling period from Murray, Cannell & Smith 1989

STASH Chilling Requirements



- Effects of temperature & drought on assimilation & respiration thru multipliers (see Forska gap model 1993 paper) in the range 0-1 giving likely degree of establishment at any site (STASH)
- Species - a parabolic response to daily temperature between -4 & 36°C (pine & spruce) and other species between -4 & 42°C (see Larcher 1983)
- Growth rates do not decline at range limits – especially in the south – species can increase growth rates until an ON-OFF switch kills them

ON-OFF switches

Minimum mean coldest month temperature (surrogate for Abs min)

Max mean coldest month temperature (e.g. Spruce -1.5°C a surrogate for ? Snow)

Minimum values for alpha a drought index (AET/PET)

Minimum GDD

STASH – effects on growth and ON-OFF switches

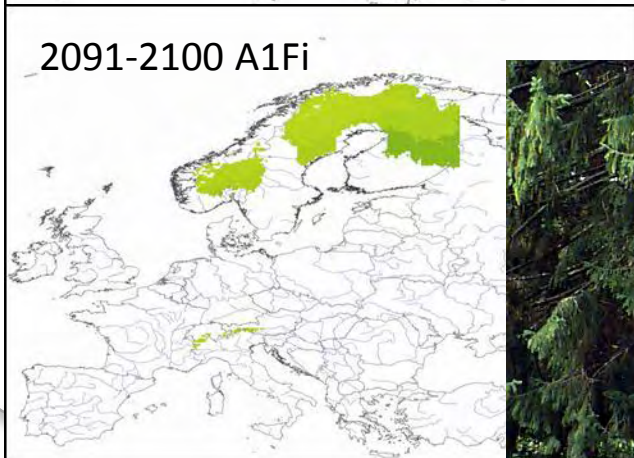
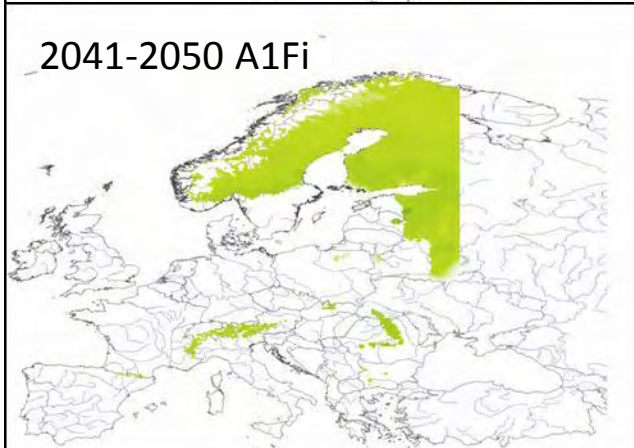
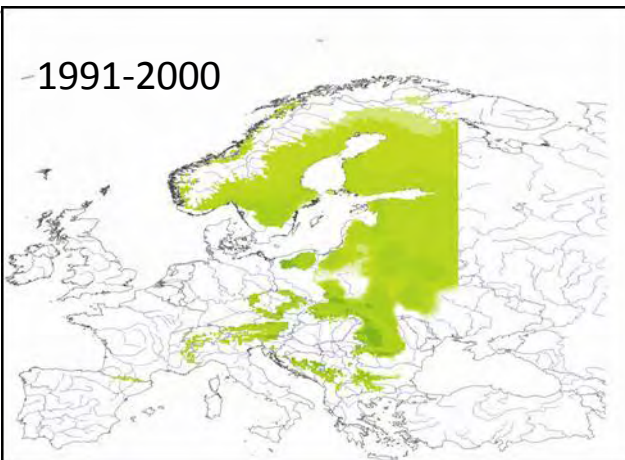
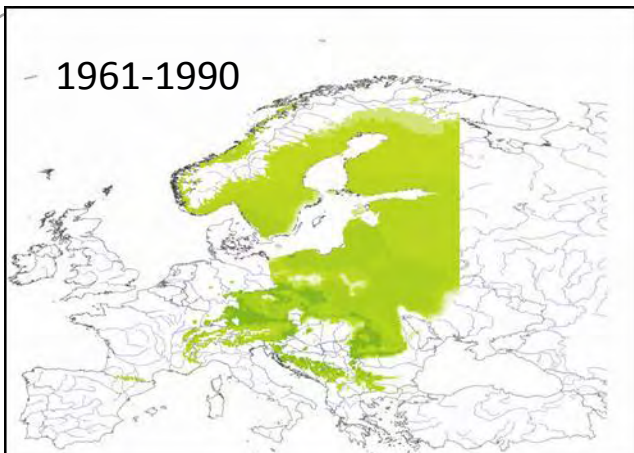


TABLE 1. Bioclimatic parameters for north European tree species.

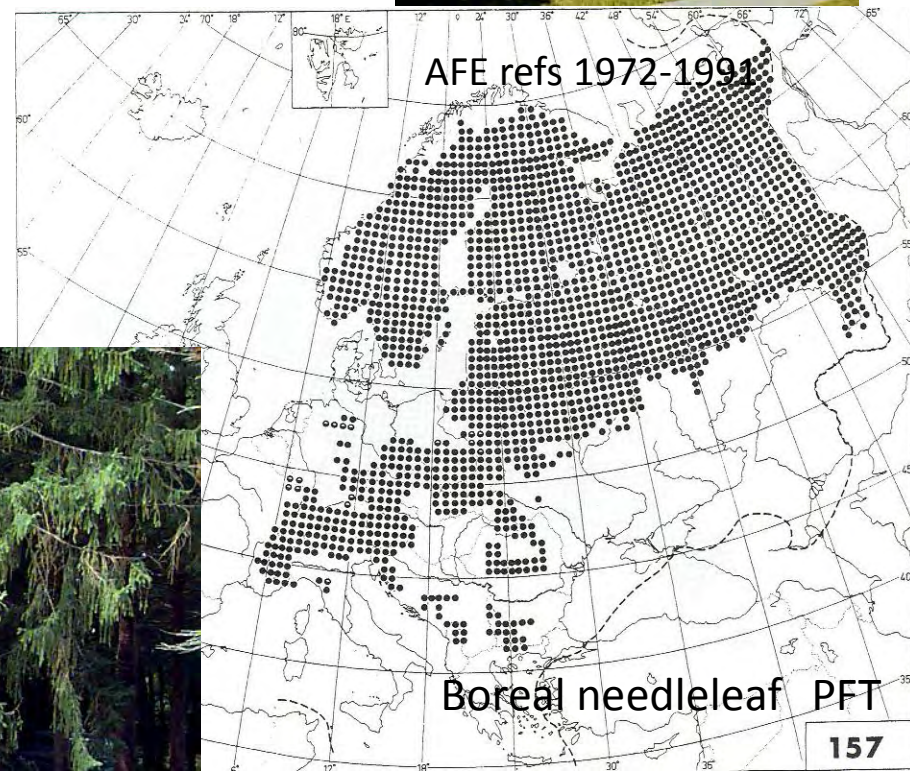
Species	min α^*	min T_c	max T_c	min GDD*	b	k
Evergreen conifers						
<i>Picea abies</i> (L.) Karsten	0.85		-1.5	600	100	0.05
<i>Pinus sylvestris</i> L.	0.70		-1	500	100	0.05
<i>Juniperus communis</i> L.	0.55			150	100	0.05
<i>Taxus baccata</i> L.	0.50	-5		1000	100	0.05
Evergreen broad-leaved						
<i>Ilex aquifolium</i> L.	0.50	-0.5		1100	100	0.05
Deciduous broad-leaved						
<i>Acer platanoides</i> L.	0.75	-16	0.5	1150	100	0.05
<i>Alnus incana</i> (L.) Moench	0.75		-2.5	430	100	0.05
<i>Betula pendula</i> Roth	0.70			700	500	0.02
<i>Betula pubescens</i> Ehrh.	0.77			150	100	0.05
<i>Carpinus betulus</i> L.	0.70	-8	5	1100	1200	0.0075
<i>Corylus avellana</i> L.	0.55	-15		800	450	0.02
<i>Fagus sylvatica</i> L.	0.65	-3.5	6	990	1150	0.0065
<i>Fraxinus excelsior</i> L.	0.65	-16	6	1100	100	0.05
<i>Populus tremula</i> L.	0.60		6	400	100	0.05
<i>Quercus petraea</i> (Mattuschka) Liebl.	0.67	-3.5	6.5	1150	100	0.05
<i>Quercus robur</i> L.	0.65	-16		1100	100	0.05
<i>Sorbus aucuparia</i> L.	0.60			300	100	0.05
<i>Tilia cordata</i> Millar	0.70	-18	5	830	1250	0.0065
<i>Ulmus glabra</i> Hudson	0.65	-15		850	100	0.05

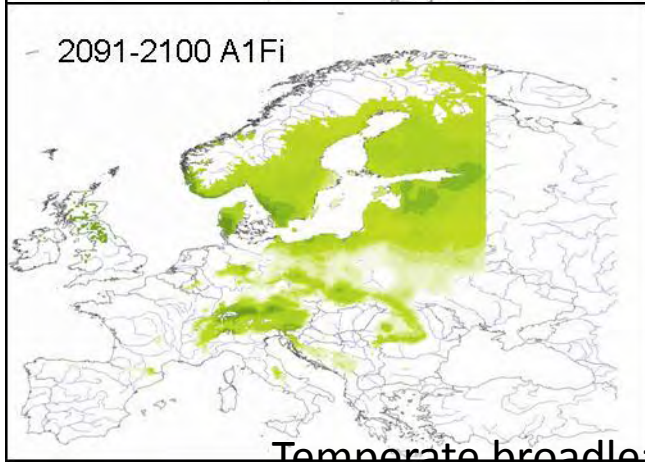
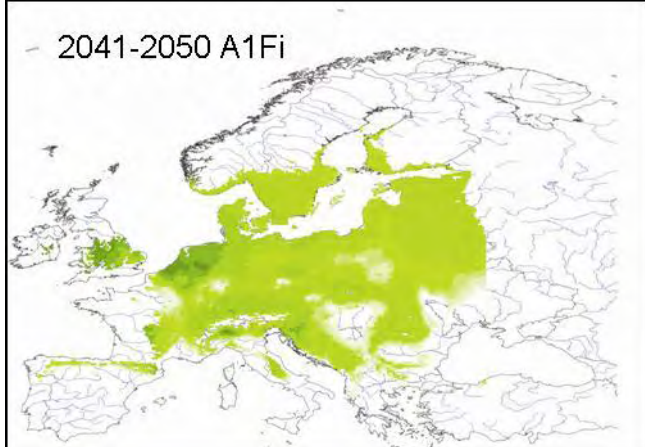
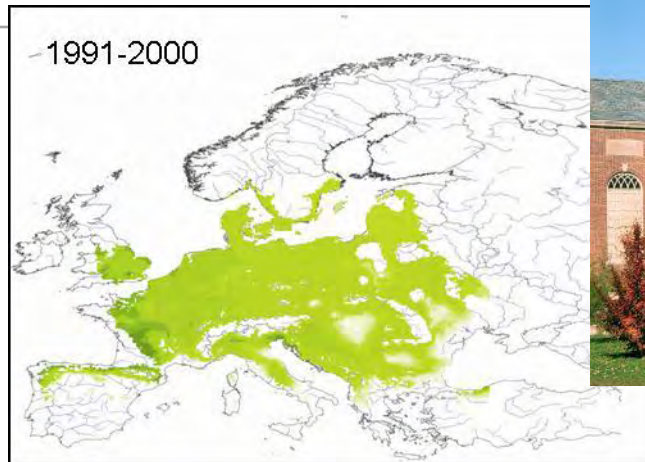
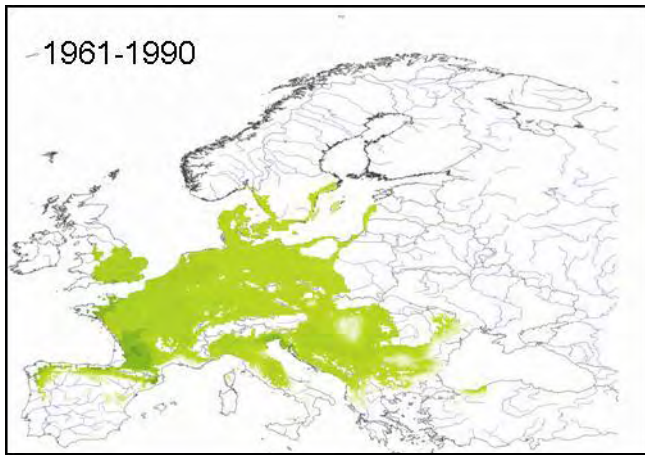
Some species parameters



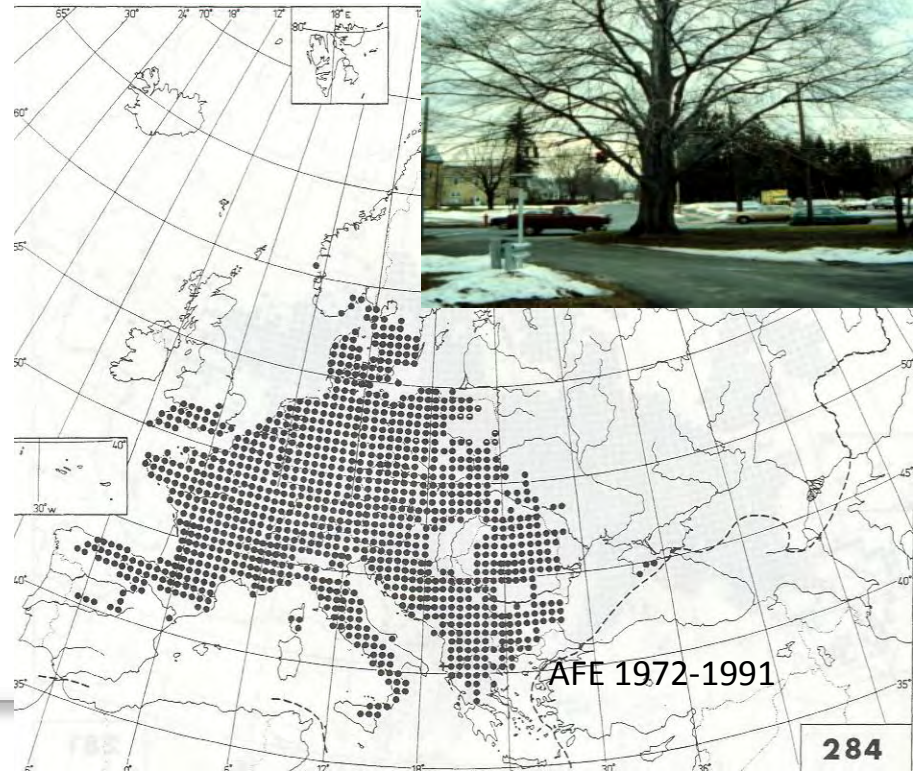


Picea abies





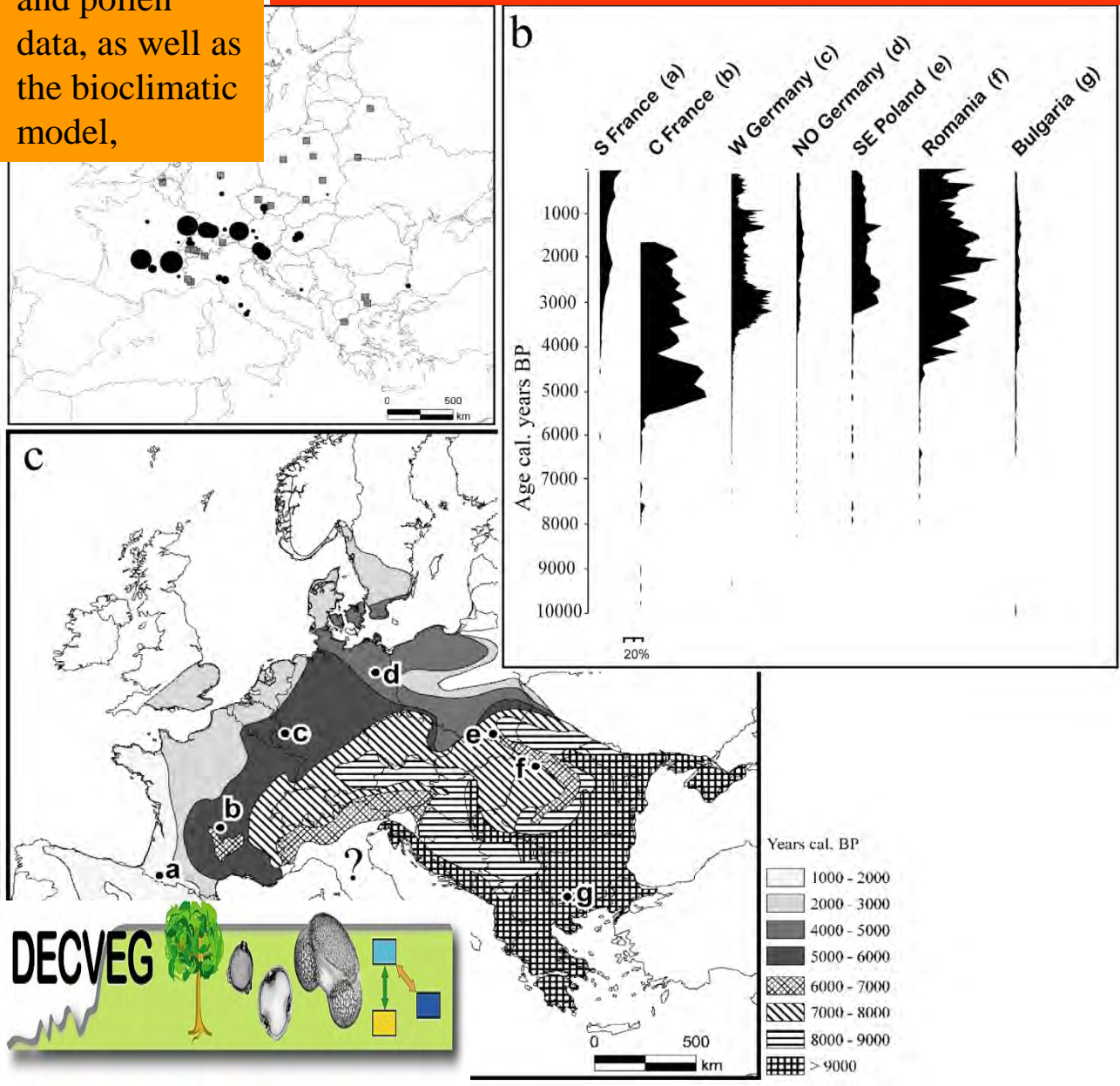
Fagus sylvatica



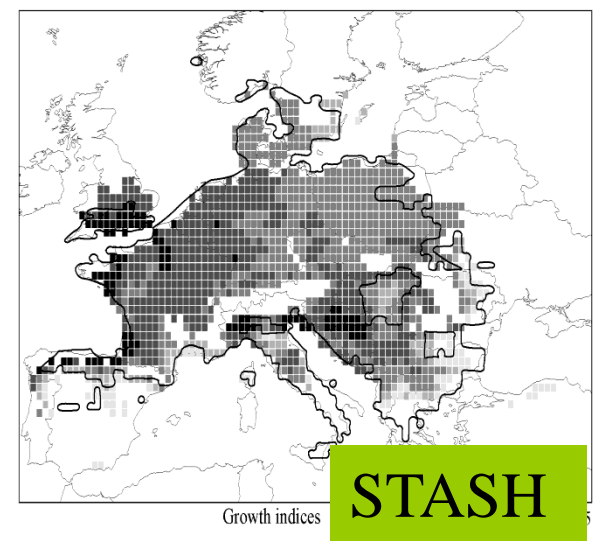
Temperate broadleaf deciduous PFT

NB:
Uncertainties
in the climate
and pollen
data, as well as
the bioclimatic
model,

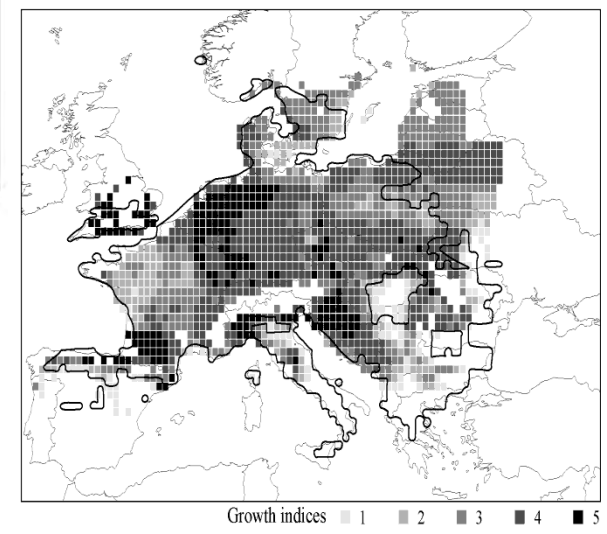
Towards an understanding of the Holocene distribution of *Fagus sylvatica* (L) *Giesecke et al.* 2007 *J. Biogeography* 34, 118-131



An equilibrium approach



6000BP GEN2



6000BP ECHAM

Ecosystem dynamics and climate change

Physical and chemical coupling between terrestrial ecosystems and climate

Drivers of change in ecosystems

CO₂ concentration increases

Climate change

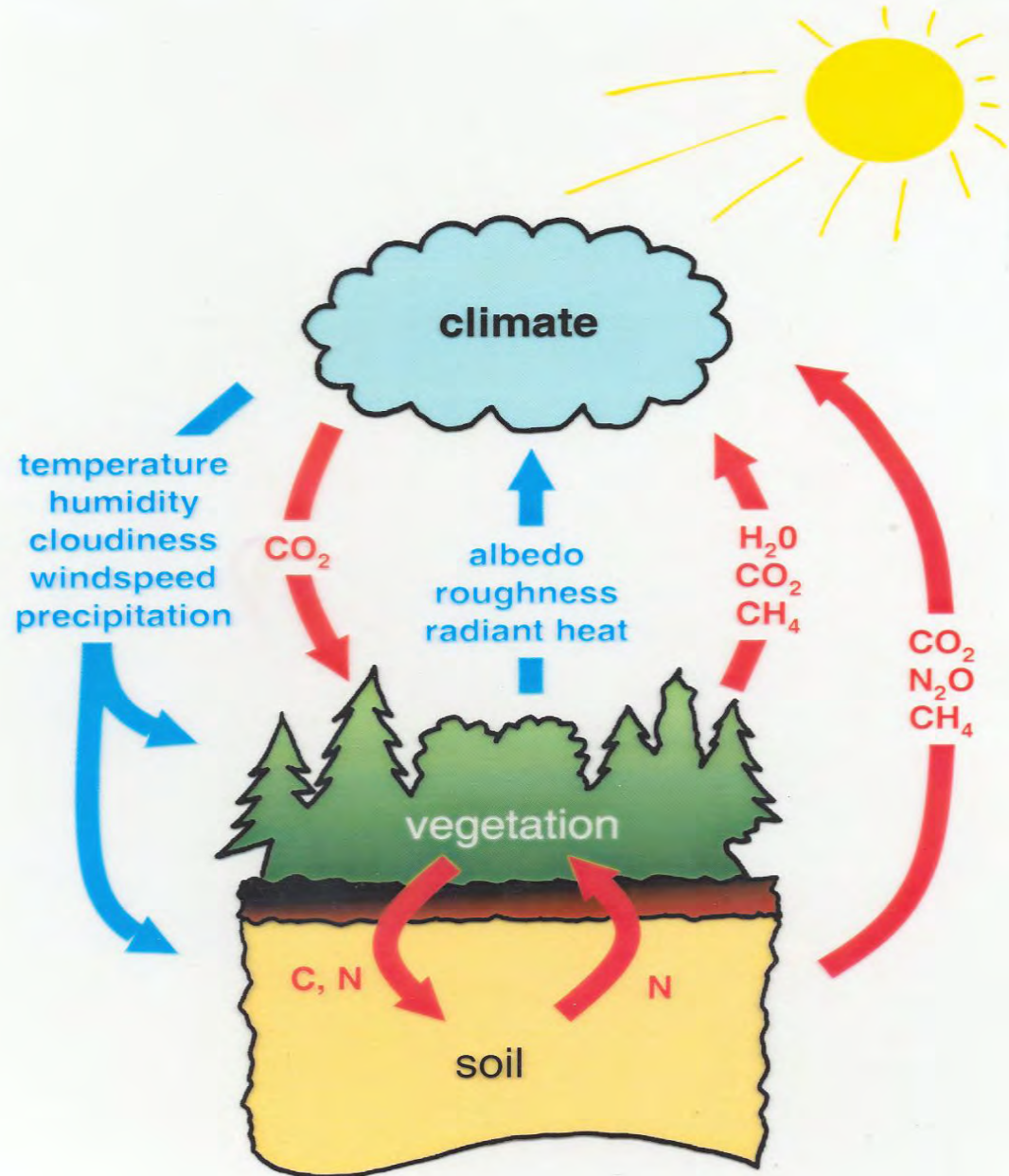
Land use change

Exotic species

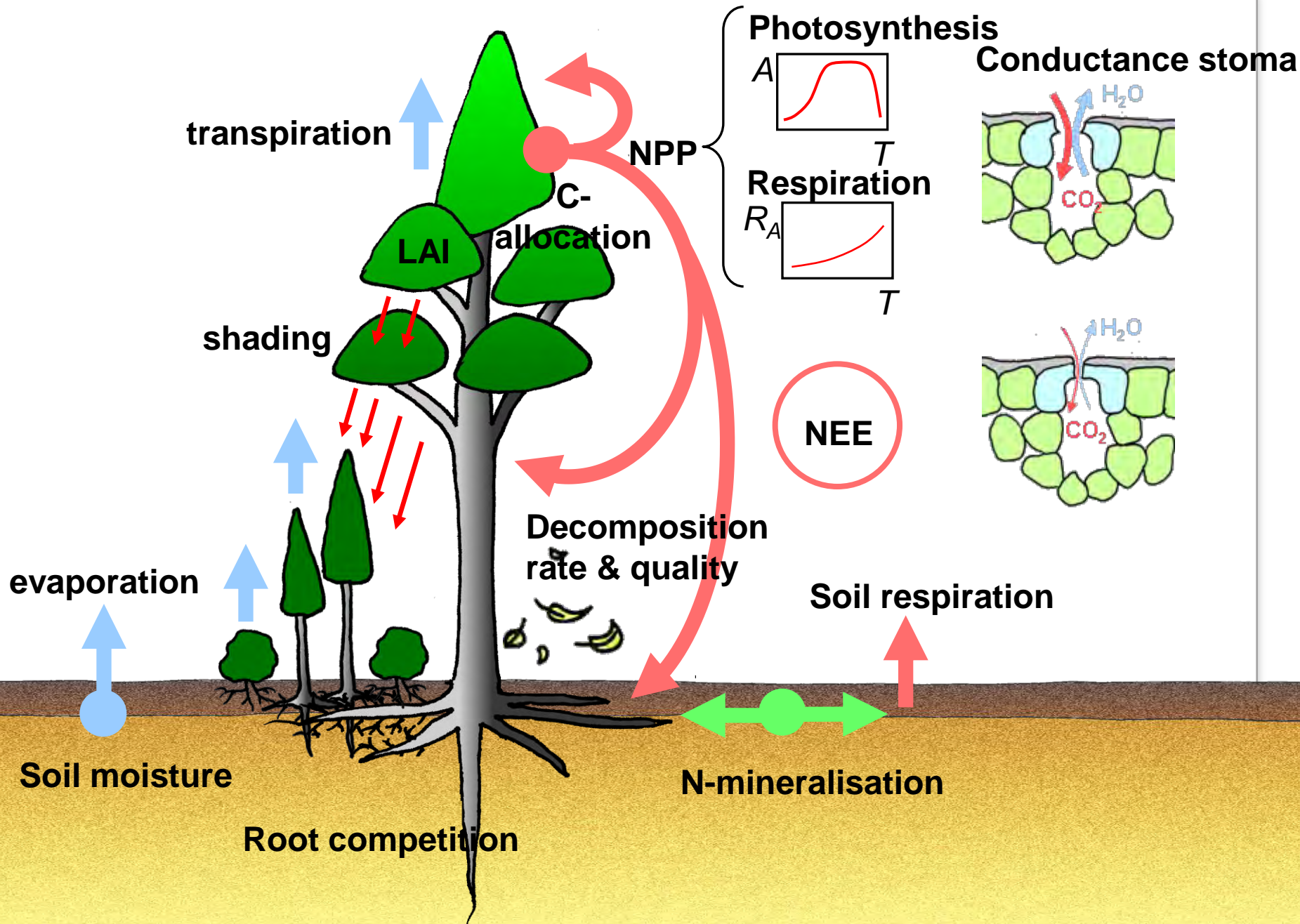
N deposition

Disturbance

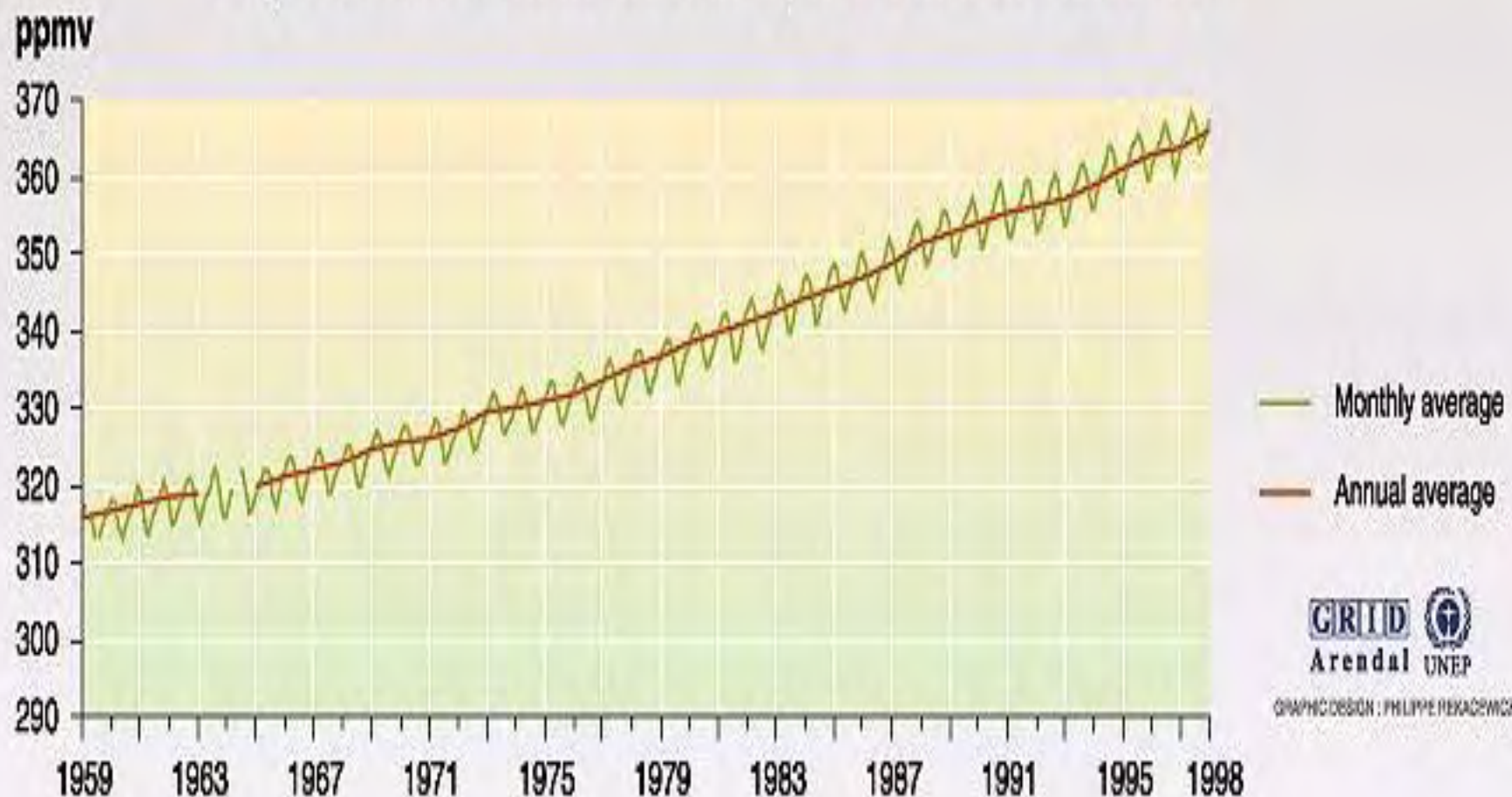
TIME



Influence of temperature on ecosystem processes

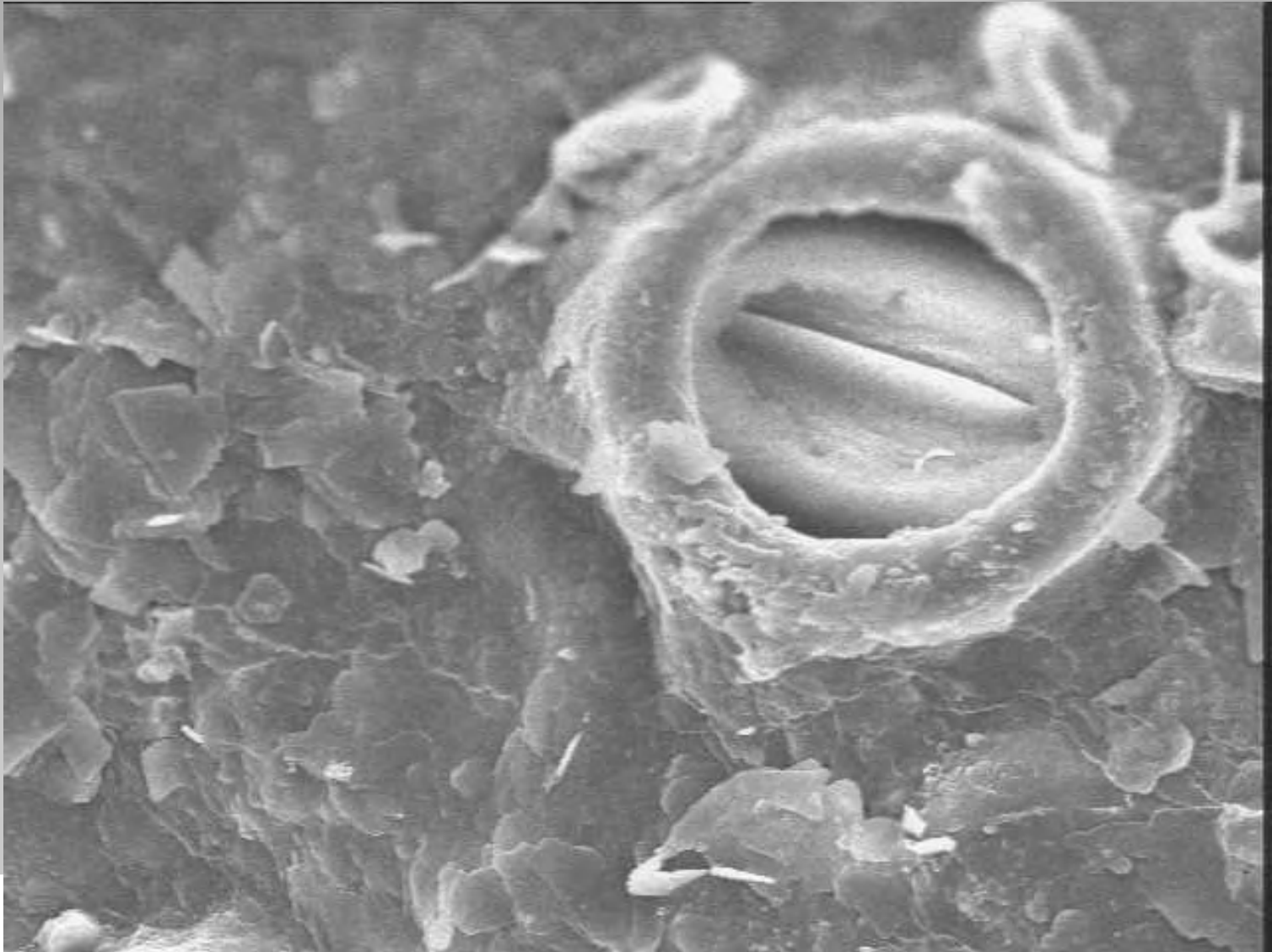


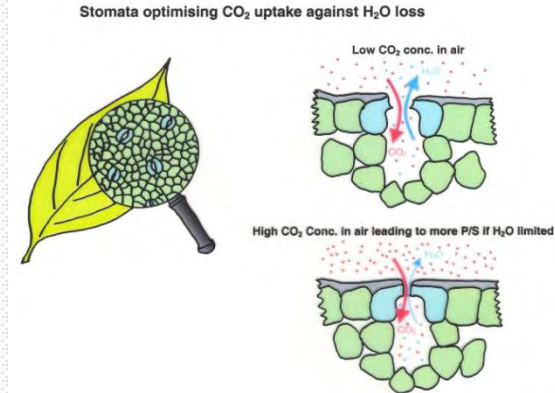
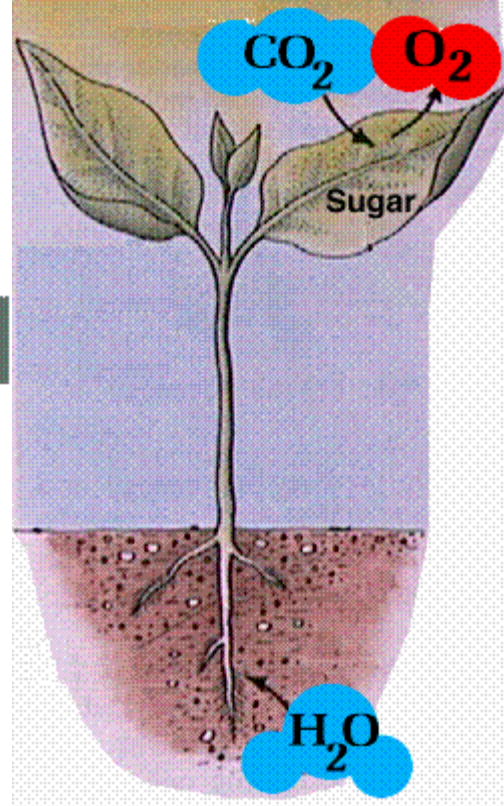
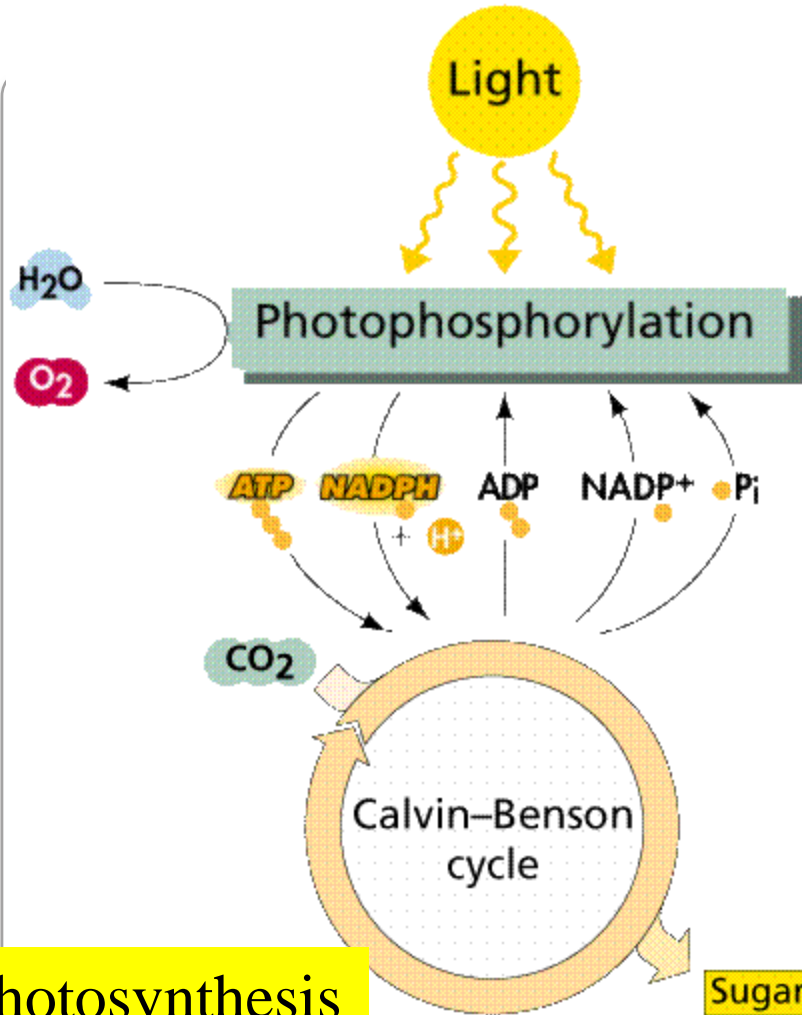
CO₂ concentration in the atmosphere: Mauna Loa curve



Source : Scripps Institution of oceanography (SIO), University of California, 1998.

CO₂ is a resource for plants



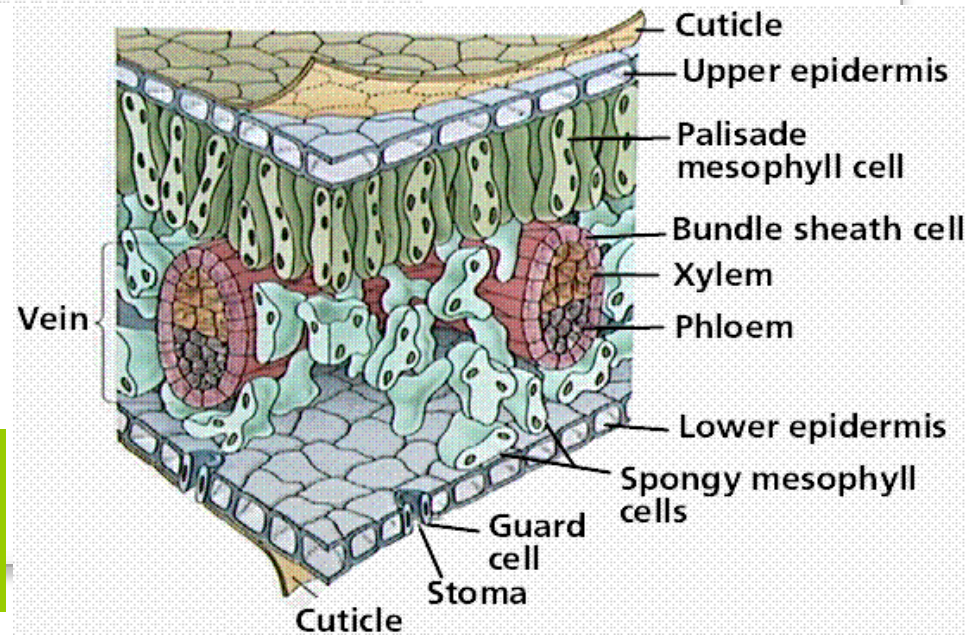


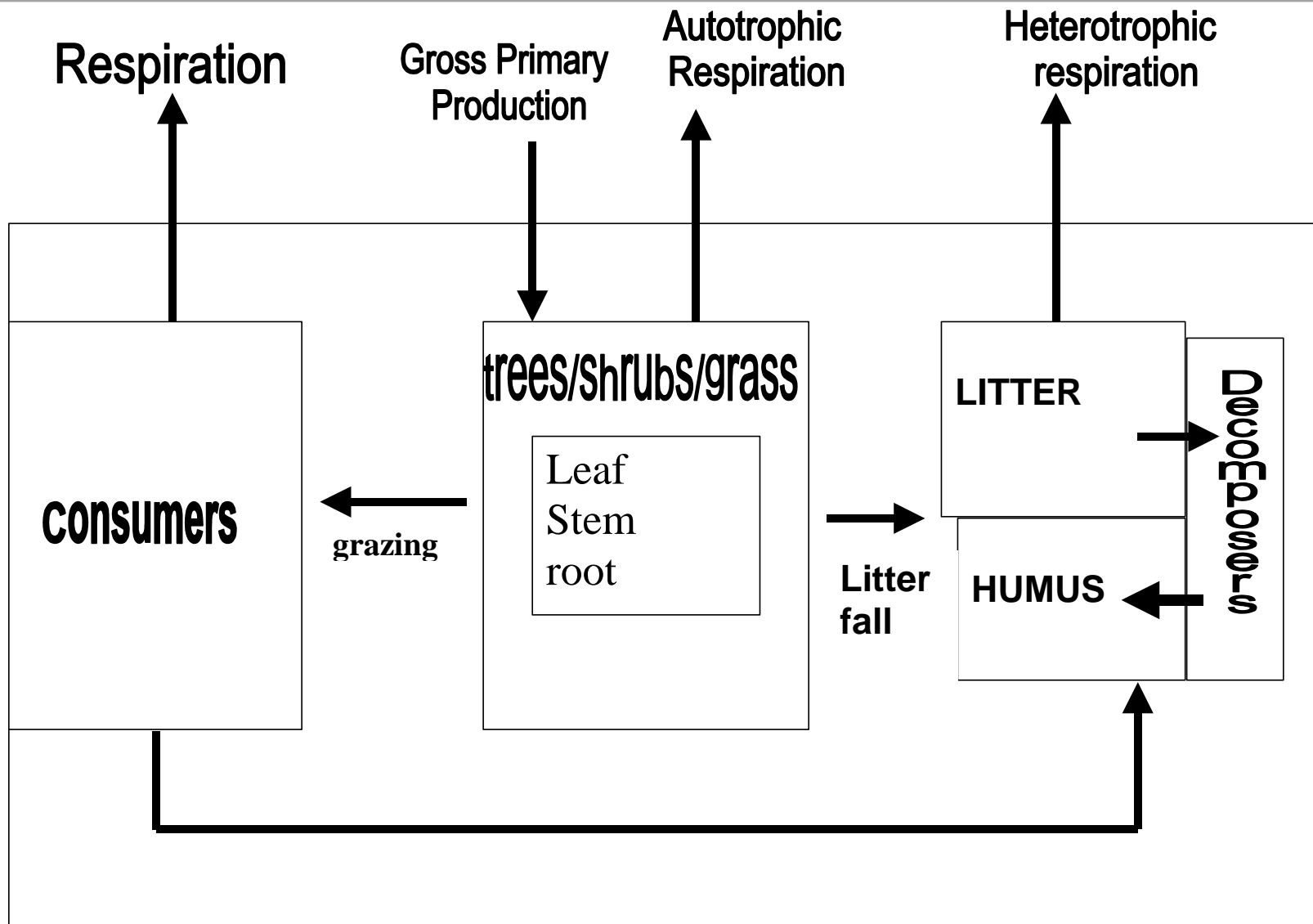
Photosynthesis

$6\text{CO}_2 + 6\text{H}_2\text{O} + \text{photons}$ gives $\text{C}_6\text{H}_{12}\text{O}_6 + 6\text{O}_2$
glucose

Respiration

$\text{C}_6\text{H}_{12}\text{O}_6 + 6\text{O}_2$ gives $6\text{CO}_2 + 6\text{H}_2\text{O} + \text{energy (kJ)}$





Flows of carbon

- CO₂ fertilisation effects
- Important in some biomes e.g. Dry grasslands, Mediterranean etc
- WUE & Increased Carbon dioxide

**Increasing Carbon Dioxide –
direct effects**

CO2 direct effects

Ainsworth & Long 2005 What have we learned from 15 years of free-air CO2 enrichment (FACE) ? A meta-analytic review of the responses of photosynthesis, canopy properties and plant production to rising CO2.

New Phytologist 351-371

Table 1 Large-scale free-air CO₂ enrichment (FACE) facilities used in this review

Site	Location	Elevated [CO ₂]	Site description reference	Ecosystem	First year of exposure (p)
Aspen FACE FACTS 2	Rhineland, WI, USA 45°36'-N, 89°42'-W	Ambient + 200	Dickson <i>et al.</i> (2000)	Aspen forest	1998
BioCON	Cedar Creek, MN, USA 45°24'-N, 93°12'-W	550	Reich <i>et al.</i> (2001)	Natural prairie grassland	1998
Cedar Creek ETH-Z FACE	Eschikon, Switzerland 47°27'-N, 8°41'-E	600	Zanetti <i>et al.</i> (1996)	Managed grassland	1993
Swiss FACE FACTS 1	Orange County, NC, USA 35°58'-N, 70°5'-W	Ambient + 200	Hendrey <i>et al.</i> (1999)	Loblolly pine forest	1996
Duke Forest Maricopa FACE	Maricopa, AZ, USA 33°4'-N, 111°59'-W	550* Ambient + 200+	Lewin <i>et al.</i> (1994)	Agronomic C ₃ and C ₄ crops	1989
Nevada Desert	Mojave Desert, NV, USA 36°49'-N, 115°55'-W	550	Jordan <i>et al.</i> (1999)	Desert ecosystem	1997
Oak Ridge	Roane County, TN, USA 35°54'-N, 84°20'-W	Ambient + 200	Norby <i>et al.</i> (2001)	Sweetgum plantation	1998
Pasture FACE	Bulls, New Zealand 40°14'-S, 175°16'-E	475	Edwards <i>et al.</i> (2001)	Managed pasture	1997
POPFACE	Viterbo, Italy 42°37'-N, 11°80'-E	Ambient + 200	Miglietta <i>et al.</i> (2001)	Poplar plantation	1999
Rapolano Mid FACE	Chianti Region, Italy 43°25'-N, 11°35'-E	560–600	Miglietta <i>et al.</i> (1997)	<i>Vitis vinifera</i> <i>Solanum tuberosum</i>	1995
Rice FACE	Shizukuishi town, Japan 39°38'-N, 140°57'-E	Ambient + 200	Okada <i>et al.</i> (2001)	<i>Oryza sativa</i>	1998
SoyFACE	Champaign, IL, USA 40°02'-N, 88°14'-W	550		<i>Glycine max</i> <i>Zea mays</i>	2000

The Brookhaven National Laboratory (BNL) injection method is described in detail by Hendrey *et al.* (1993) and Lewin *et al.* (1994). Pure C injection methods are described by Miglietta *et al.* (2001) and Okada *et al.* (2001). A detailed map of all FACE experiments, and links to individual websites, are given at the Carbon Dioxide Information Analysis Center website: <http://cdiac.esd.ornl.gov/programs/FACE/whereisface.html>.

*1989–94; †1996–2000.



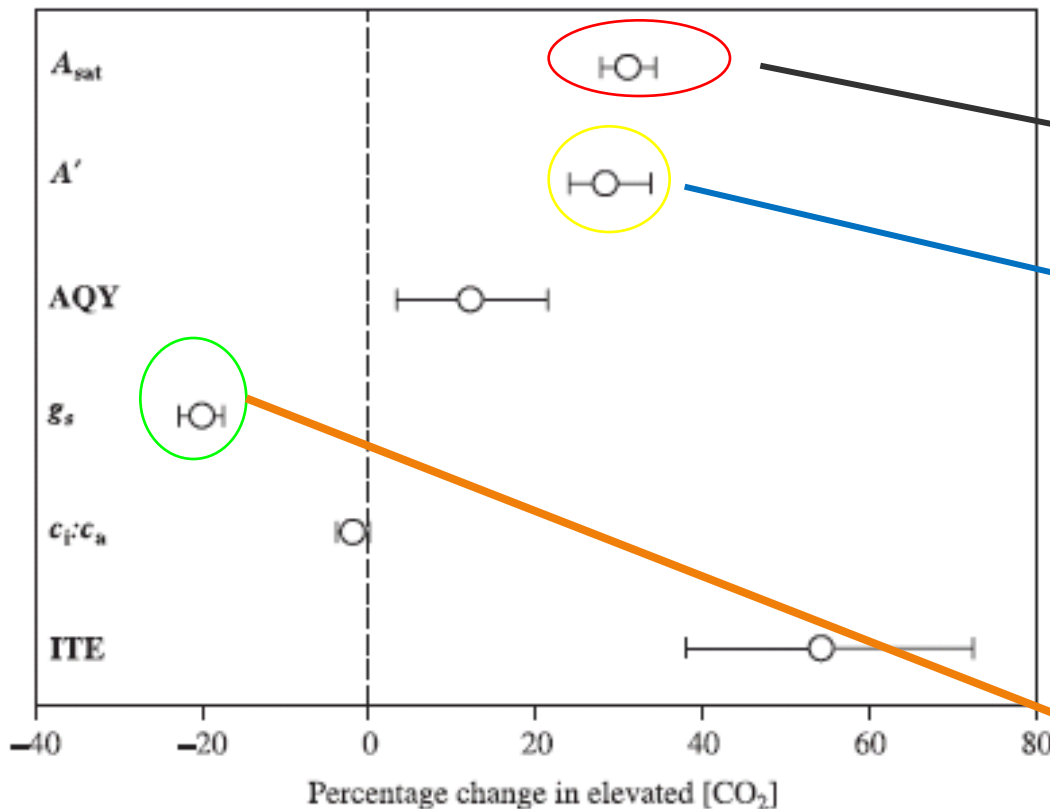


Fig. 1 Mean response to elevated [CO₂] ($\pm 95\%$ CI) of light-saturated CO₂ uptake (A_{sat}), diurnal carbon assimilation (A'), apparent quantum yield of CO₂ uptake (AQY), stomatal conductance (g_s), ratio of intercellular (c_i) to atmospheric CO₂ concentration (c_a), and instantaneous transpiration efficiency (ITE). Number of species, FACE experiments and individual observations for each response are given in Appendix 2.

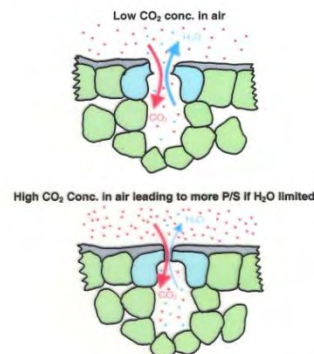
Elevated CO₂ increased light saturated photosynthetic rate by 31%

Diurnal Ps carbon assimilation increased by 28%

Across all experiments and all species

Stomatal conductance (g_s) Averaged over 40 spp declined by 20% in elevated CO₂)

Stomata optimising CO₂ uptake against H₂O loss



C3 functional groups

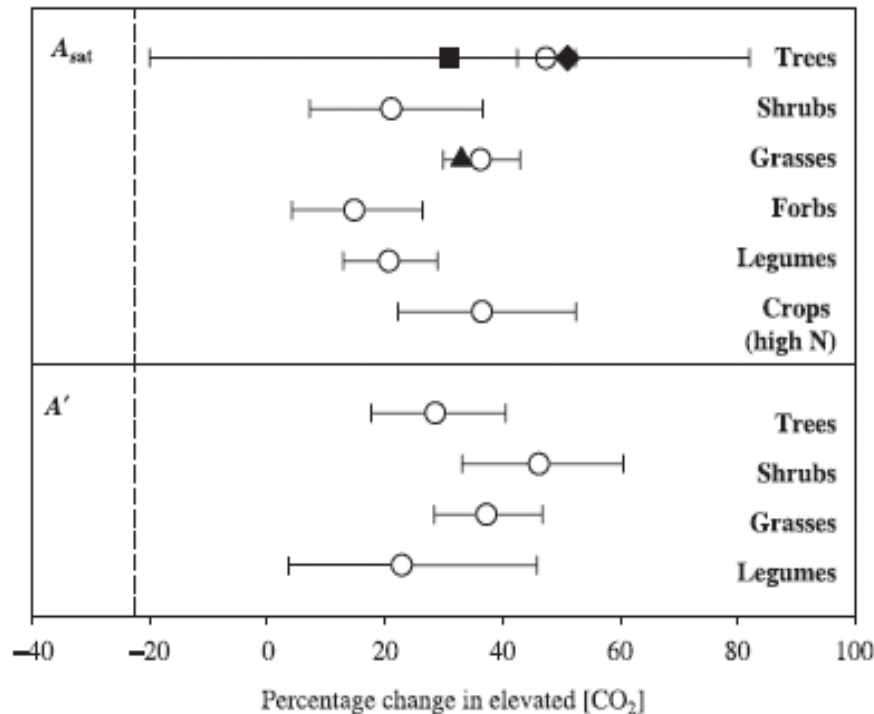


Fig. 3 Comparative photosynthetic responses of different C₃ functional groups to elevated [CO₂]. Results from: ○, this meta-analysis; ■, a meta-analysis of tree species (Curtis & Wang, 1998); ◆, a meta-analysis of European tree species (Medlyn *et al.*, 2001); ▲, a meta-analysis of C₃ grasses (Wand *et al.*, 1999). Number of species, FACE experiments and individual observations for each response in our meta-analysis are given in Appendix 2.

Trees more responsive than grass, forbs, legumes and crops

NB however FACE trees are young and rapidly growing trees

Mature trees ?

To maintain a balance in N and other resources allocated to Ps reactions species acclimate to elevated CO₂ – acclimation occurs and its different between different functional groups

Accentuated acclimation under N-limitation

BUT FACE decrease in N leads only to a marginal decrease in response to elevated CO₂

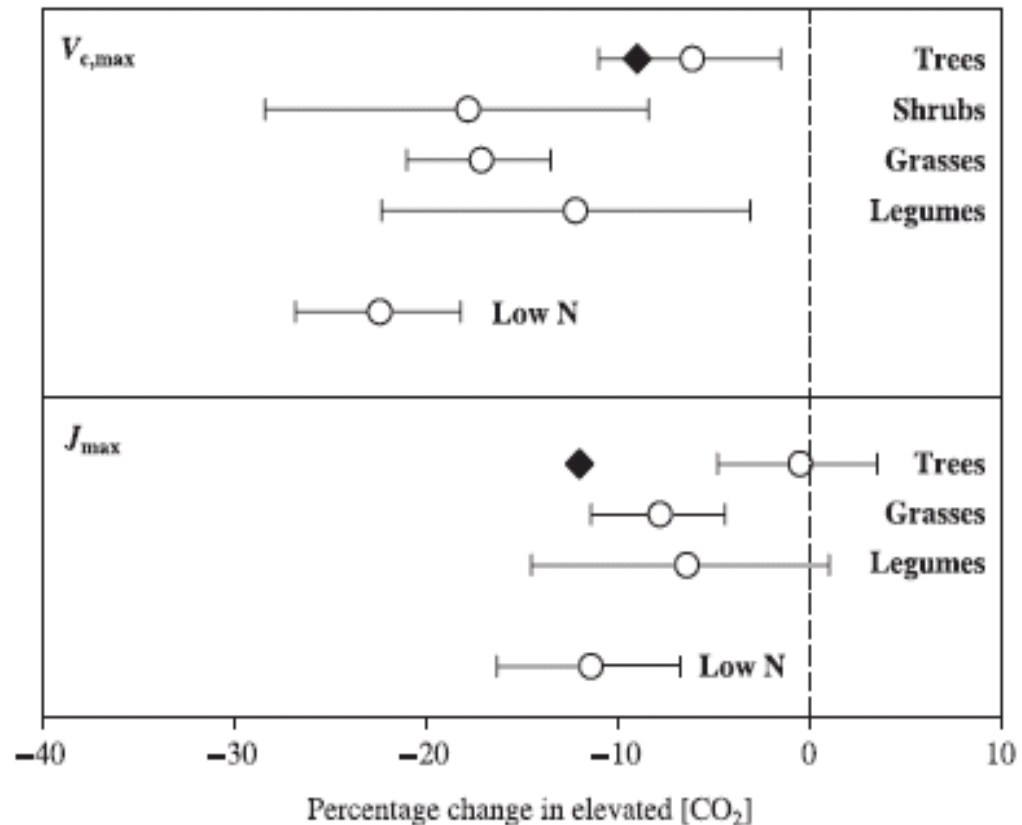
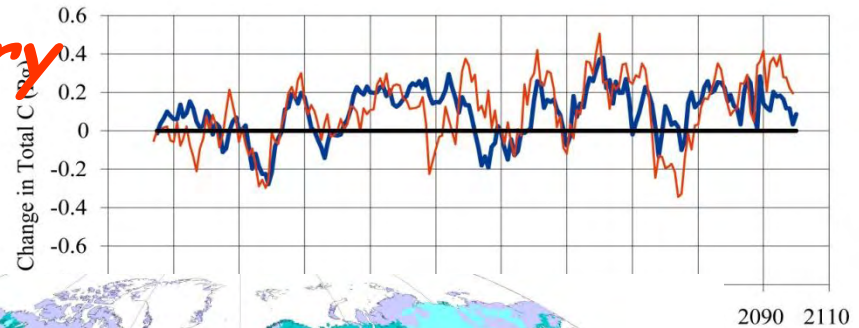


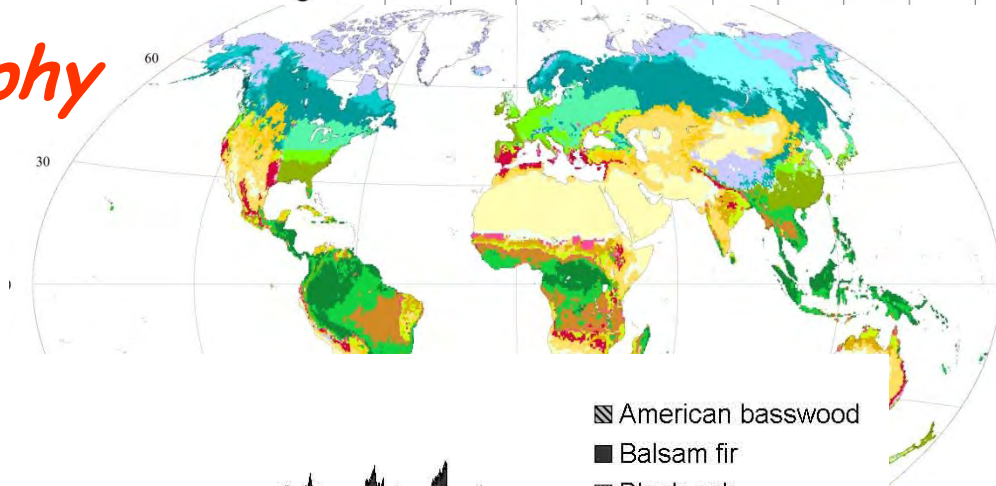
Fig. 6 Comparative acclimation responses of different C₃ functional groups to elevated [CO₂]. Results from: ○, this meta-analysis; ◆, a prior meta-analysis of European tree species (Medlyn *et al.*, 2001). Number of species, FACE experiments and individual observations for each response are given in Appendix 2.

Dynamic Ecosystem modelling

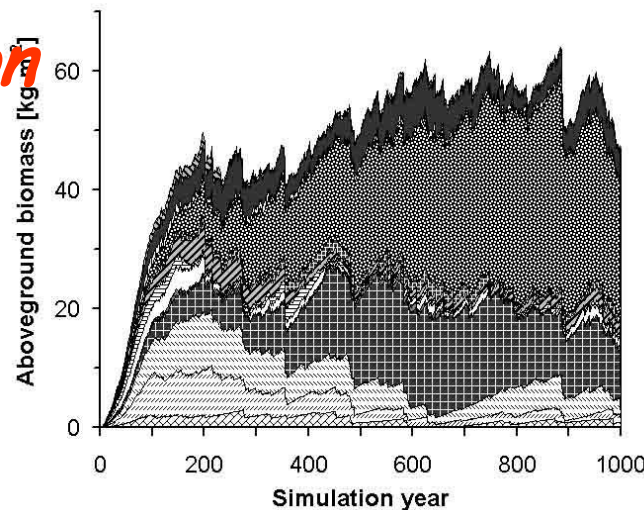
Biogeochemistry



Biogeography



Population Ecology



- American basswood
- Balsam fir
- Black ash
- Eastern hemlock
- Eastern hornbeam
- Northern white cedar
- Paper birch
- Populus spp.
- Red maple
- Sugar maple
- White pine
- White spruce
- Yellow birch

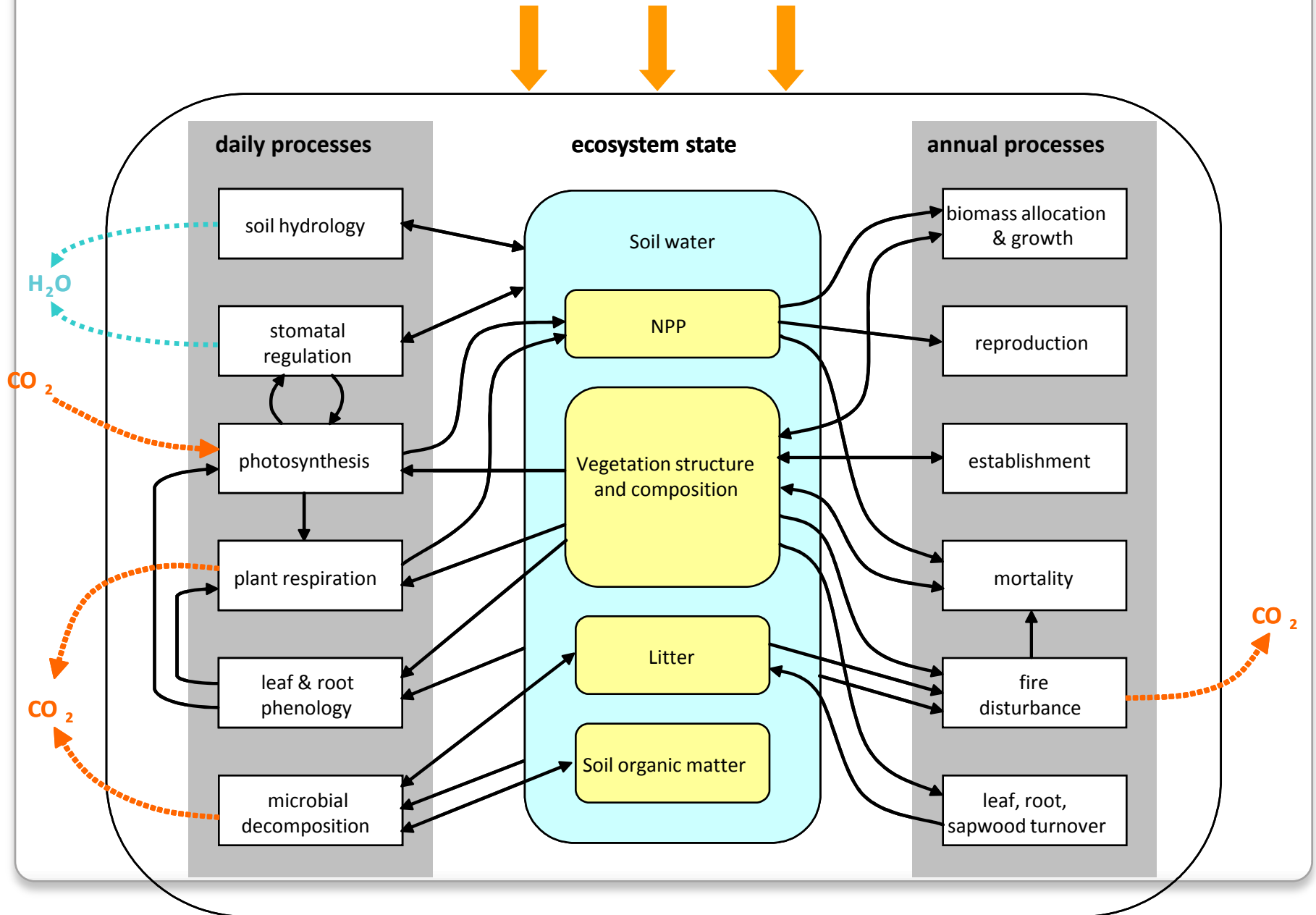
Plant Physiology



Scaling

Dynamic Vegetation Modelling

temperature, precipitation, radiation, CO₂



Modelling the biosphere

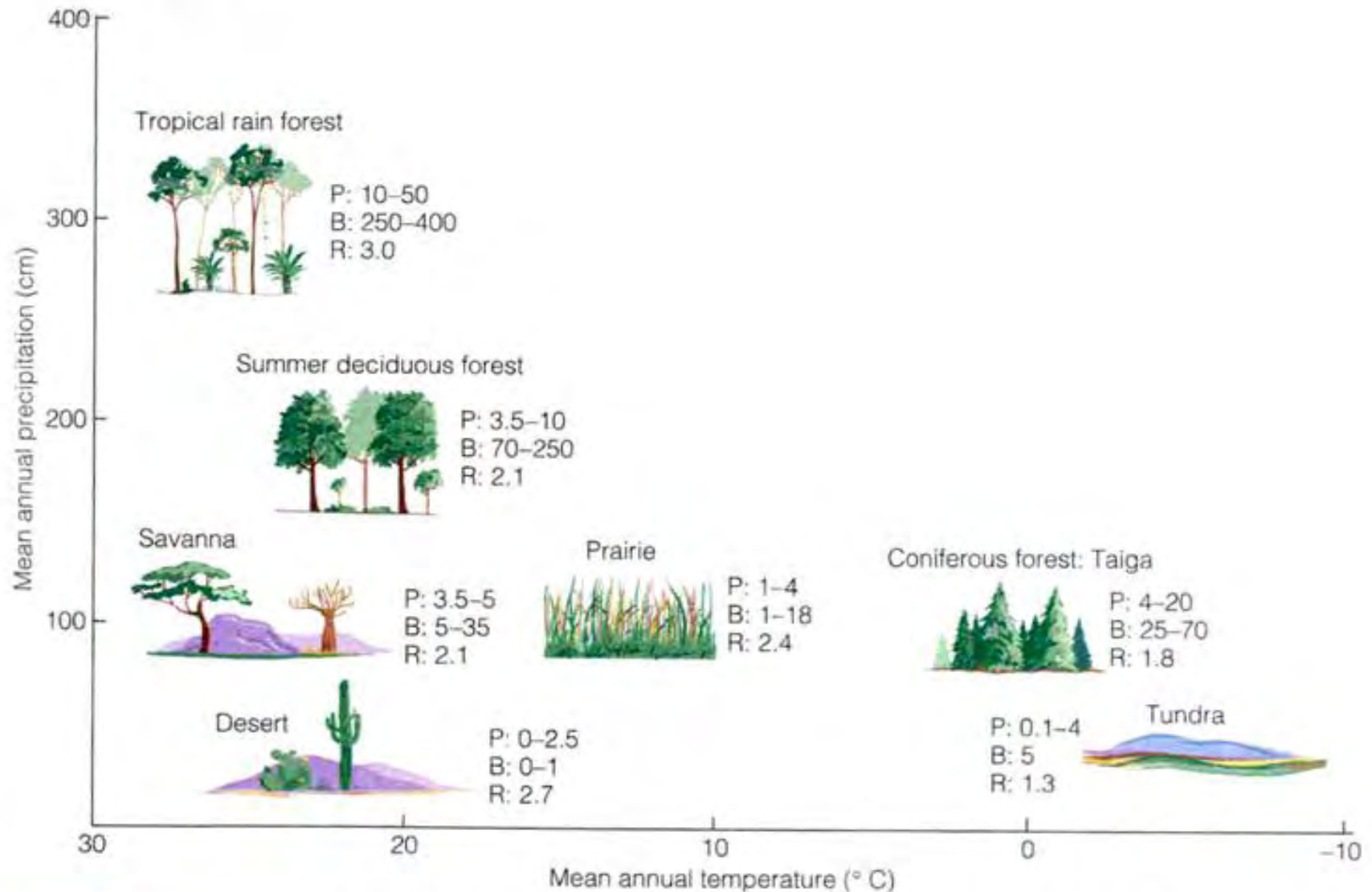


Figure 23.5 Distribution of primary production, standing biomass, and radiation input relative to rainfall and temperature. P = primary production (tn/ha); B = biomass (tn/ha); R = PAR solar radiation (kcal/m²/yr). (Adapted from J. R. Etherington, *Environmental and Plant Ecology*, 2nd ed., p. 355. New York, John Wiley, 1975. Used by permission.)

Climate inputs: Temperature, Precipitation, Net shortwave radiation

Vegetation: PFT/Species

Light 

Water 

Carbon & water fluxes 

CO₂ 

Photosynthesis

Respiration

Allocation

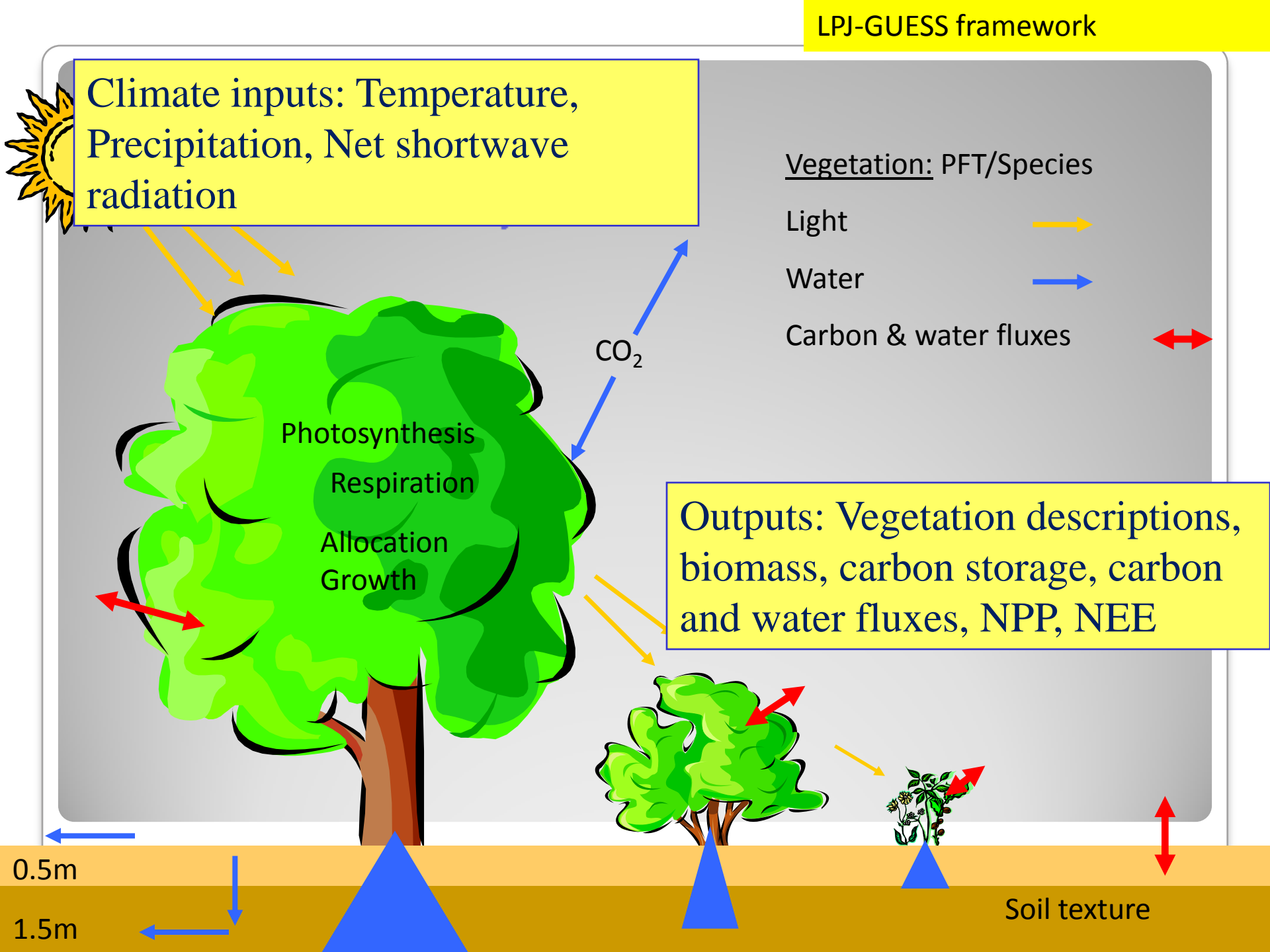
Growth

Outputs: Vegetation descriptions, biomass, carbon storage, carbon and water fluxes, NPP, NEE

0.5m

1.5m

Soil texture 



Sitch et al. GCB 2003 etc

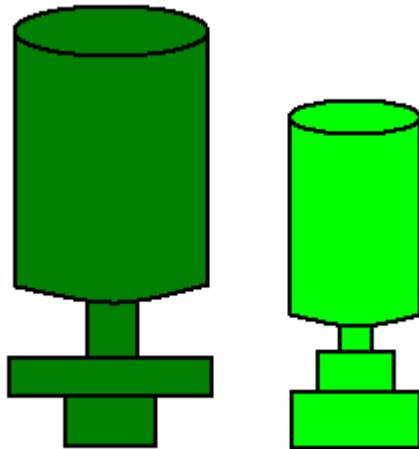
Global to Continental scale

Smith et al. 2001, Hickler et al. 2004 etc etc

Patch to landscape to regional scale

LPJ-DGVM population mode

- Average individuals
- No age structure



PFT 1

PFT 2

GUESS cohort mode

- Age structure SPECIES/PFTS

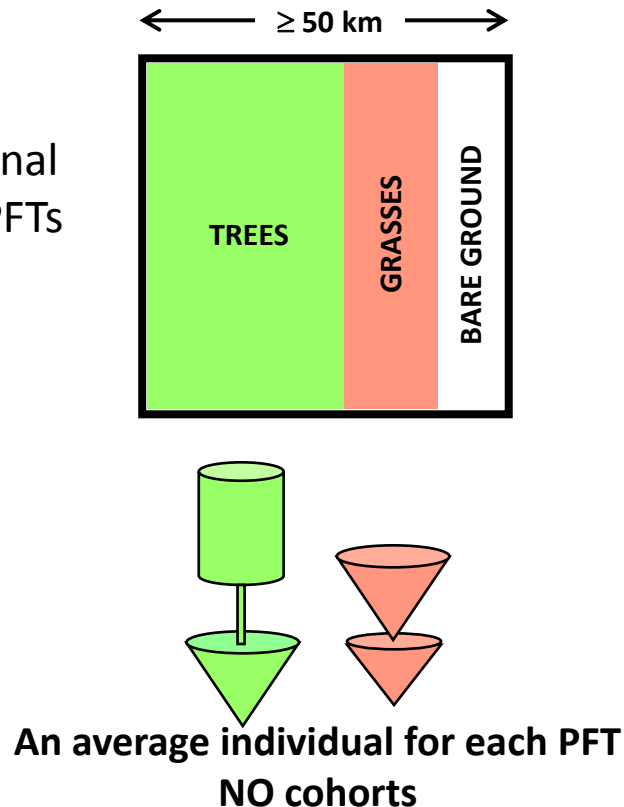


LPJ-GUESS DYNAMIC VEGETATION MODEL

AREA-BASED MODELS POPULATION MODE (LPJ-DGVM)

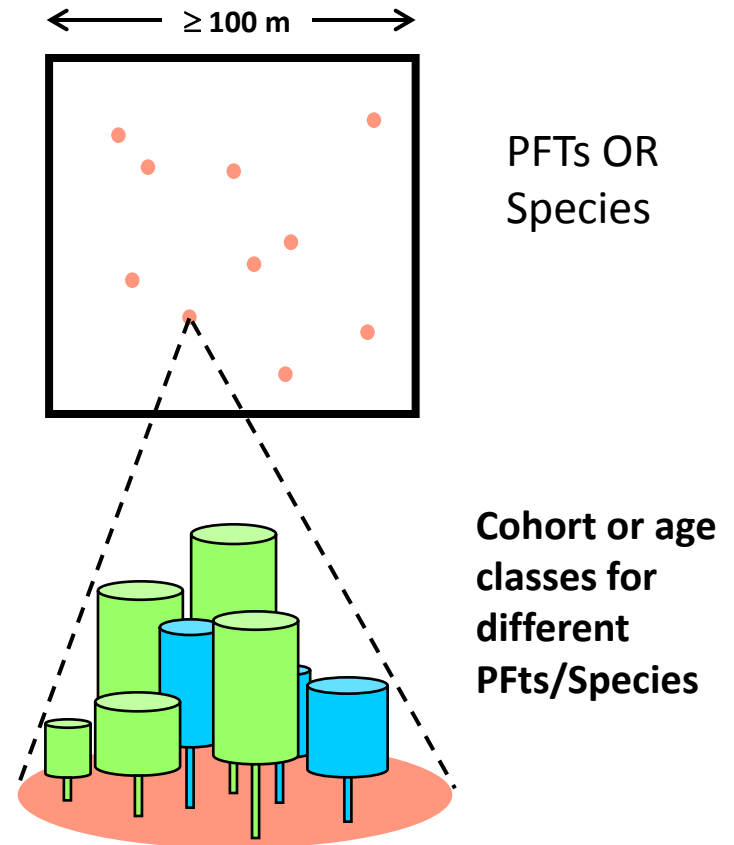
- computationally efficient
- area-averaging assumption valid at regional but not finer scales
- simplistic vegetation dynamics

Plant
functional
types PFTs



INDIVIDUAL-BASED MODELS COHORT MODE (GUESS)

- computationally demanding
- valid at local-landscape scale and upscaleable to region to continental
- mechanistic vegetation dynamics



IF NOT SPECIES THEN PLANT FUNCTIONAL TYPES (PFTs)

Physignomy (tree/grass)

Phenology (evergreen/deciduous)

Bioclimatic Limits (cold and heat tolerance)

Physiology (e.g. C_3/C_4 photosynthesis)

e.g. "Temperate summergreen broadleaved tree"

(e.g. *Fagus sylvatica*)

PLANT FUNCTIONAL TYPES

Compete for dominance in the 'carbon market'

PARAMETER

max establishment rate

light requirement for establishment

leaf phenology class

allocation to stems

fraction of roots in upper/lower soil layer

Needleleaved
evergreen



shade-intolerant
broadleaved
summergreen



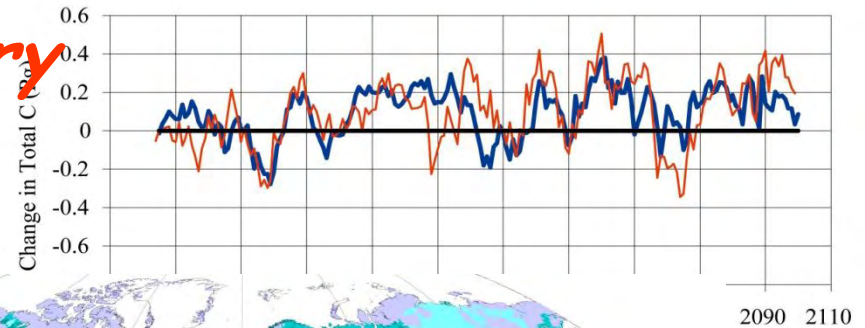
shade-tolerant
broadleaved
summergreen



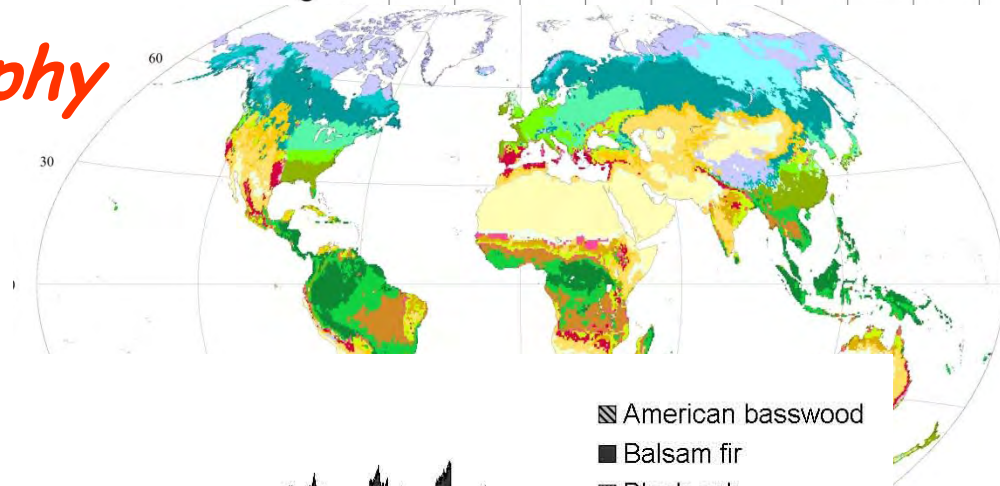
grass



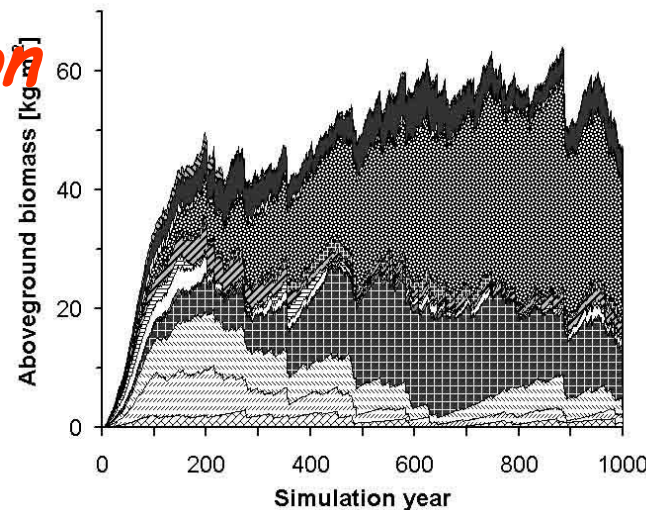
Biogeochemistry



Biogeography



Population Ecology



- American basswood
- Balsam fir
- Black ash
- Eastern hemlock
- Eastern hornbeam
- Northern white cedar
- Paper birch
- Populus spp.
- Red maple
- Sugar maple
- White pine
- White spruce
- Yellow birch

Plant Physiology



Scaling

Vegetation dynamics



herbaceous vegetation

light-demanding pioneer trees

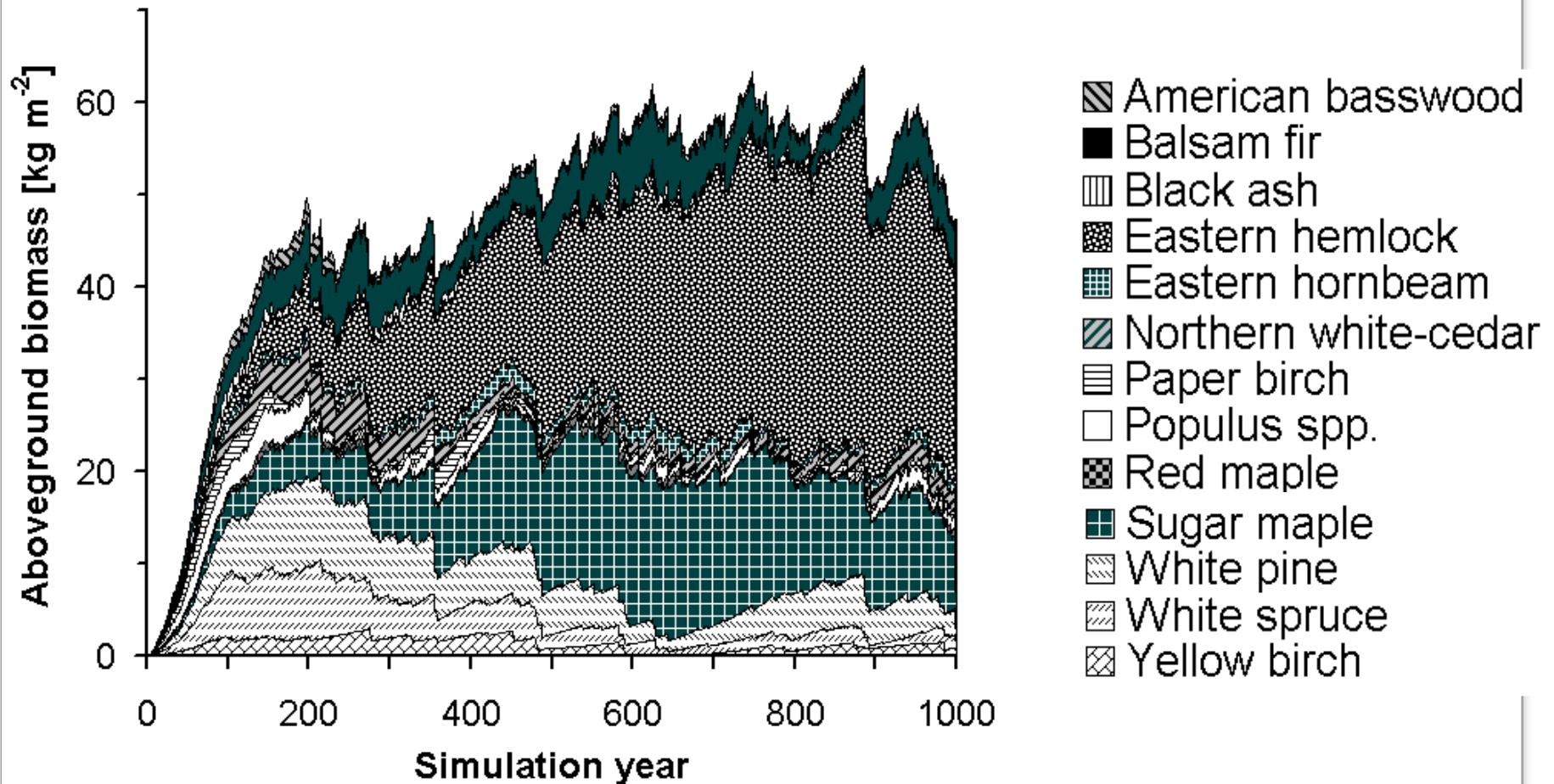
initial cohort of shade-tolerant trees



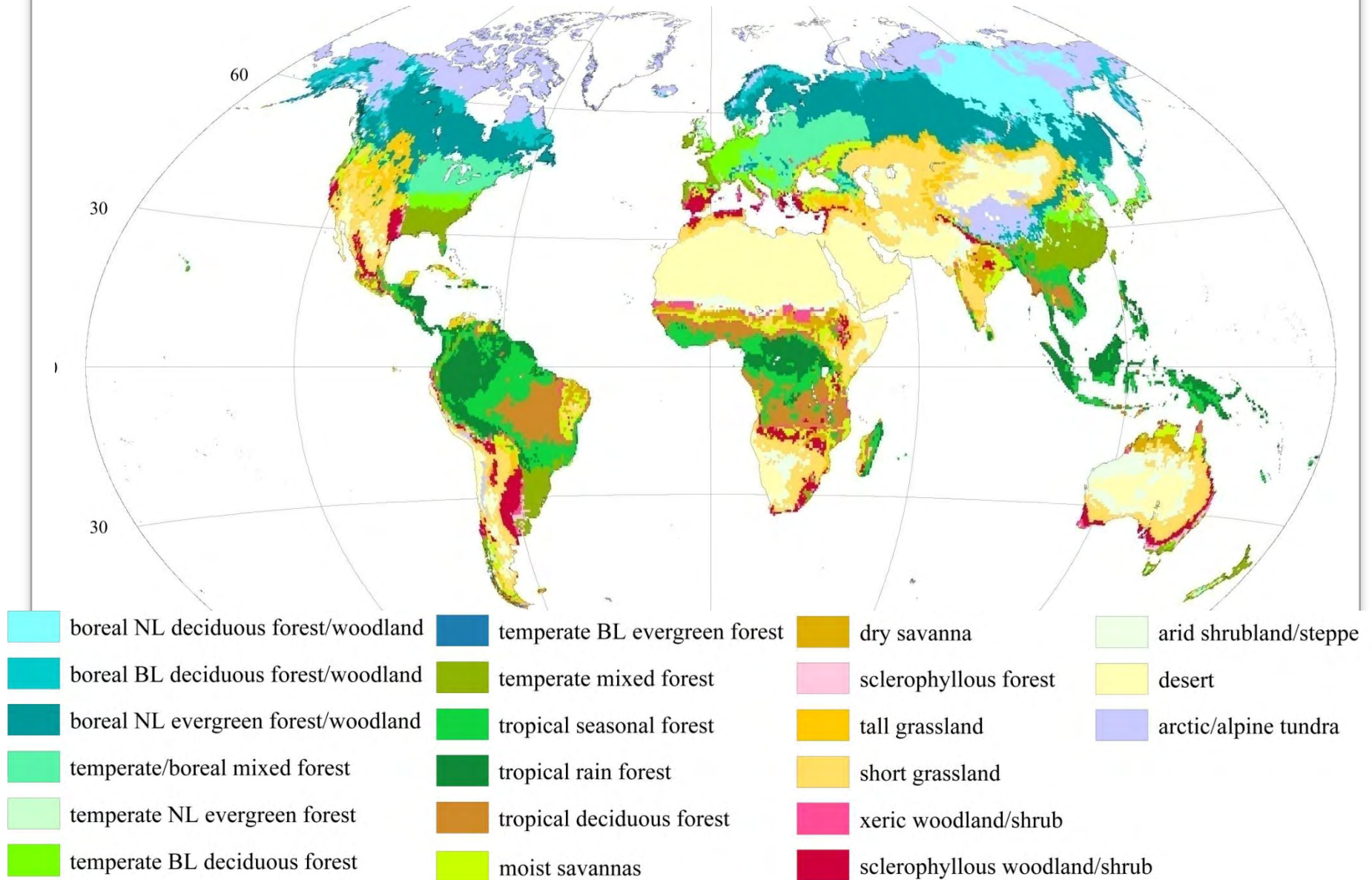
mixed-species, multi-aged and
of shade-tolerant trees

regeneration in treefall gap

Vegetation dynamics



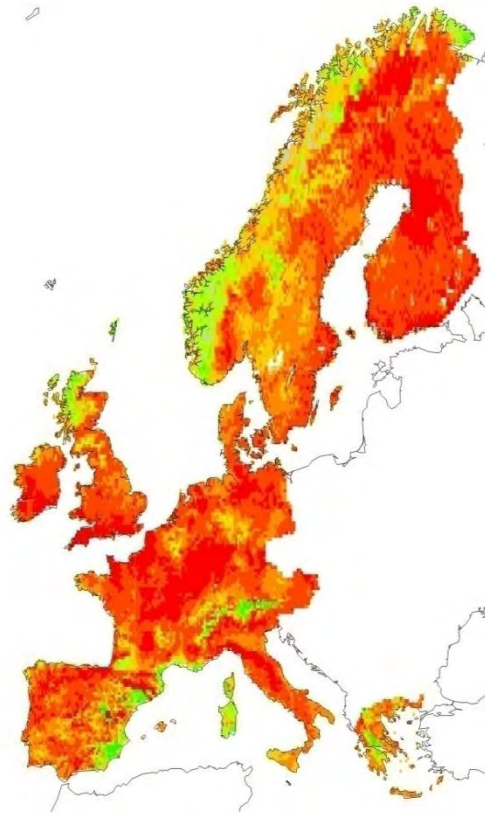
Biogeography



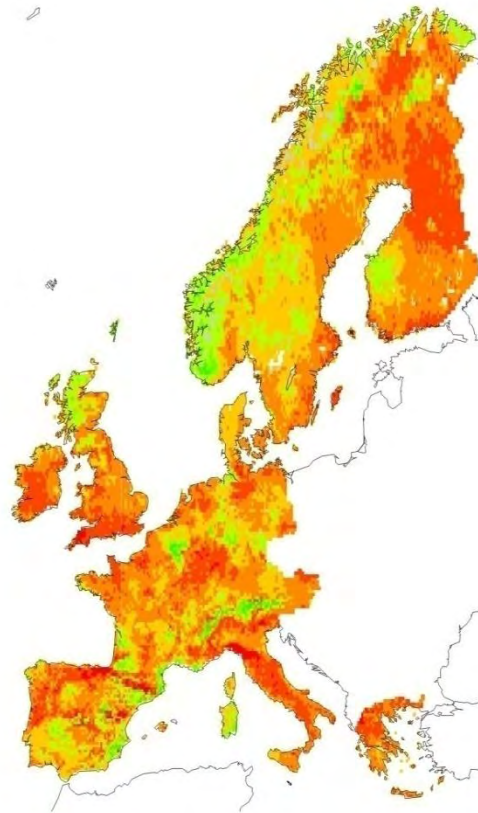
NL = needle-leaved; BL = broadleaved

Biogeochemistry

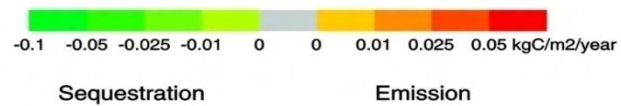
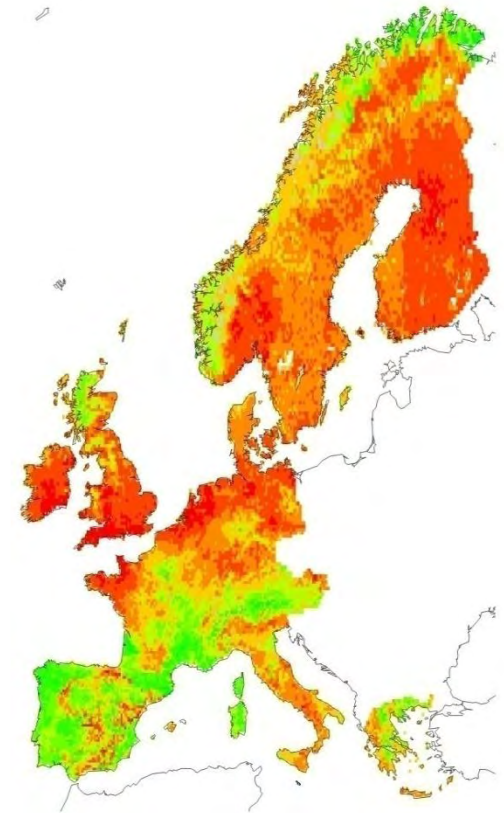
HadCM3 A1fi



HadCM3 B2

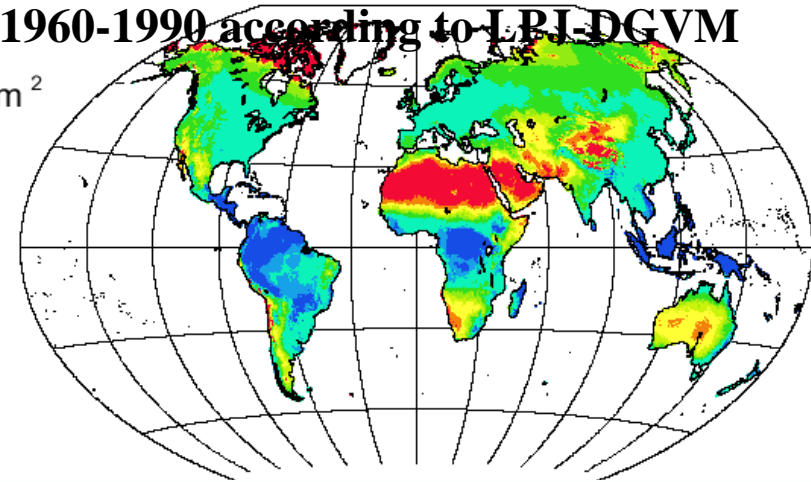
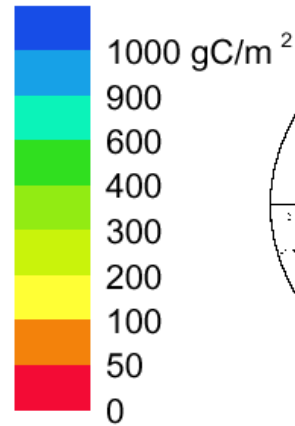


PCM2 A2

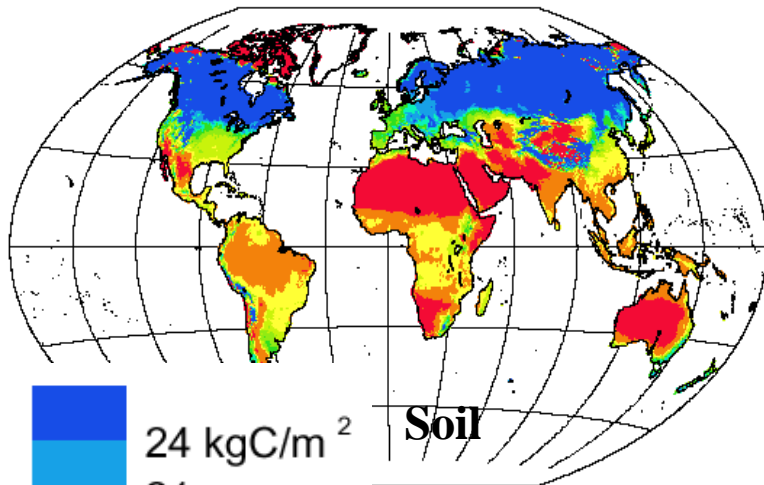


**Applications using an
ecosystem model to
address impacts of
climate change**

Global NPP 1960-1990 according to LPJ-DGVM



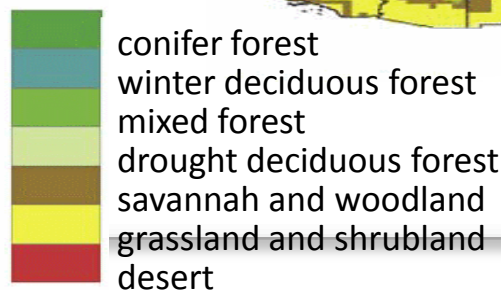
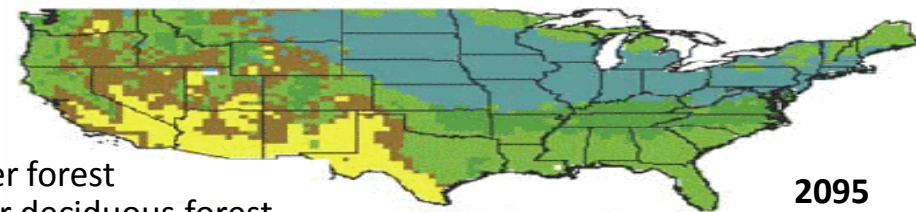
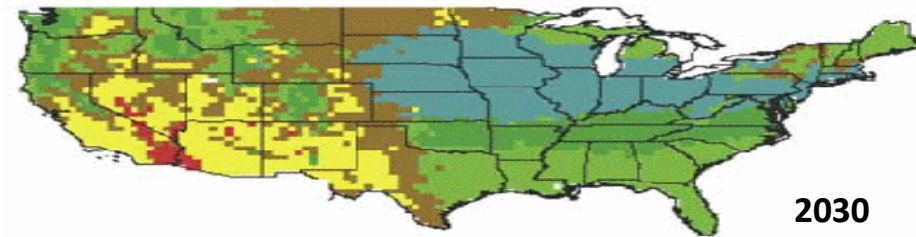
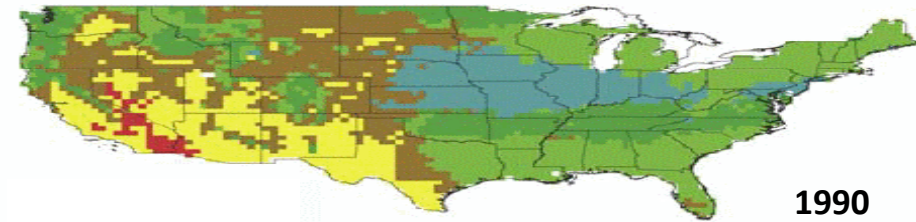
Vegetation



Soil

**Global C stocks 1960-1990
 according to LPJ-DGVM**

**Sitch et al. 2003.
 Global Change Biology 9:
 161-185**



**Bachelet et al. 2003.
 Global Biogeochem. Cycles
 17: 1045-1065**

Using LPJ-GUESS (POPULATION MODE) & Regional Climate Model output driven by data from one or more GCMs.

Land use was simulated using observed present day land use (CORINE/PELCOM)

- RCMs: SMHI Rossby centre – RCAO,
- DMI – HIRHAM RCM, GKSS – CLM
- UK Hadley HadRM3H, MPI – REMO
- GCMs Hadley –HadAM3H & ECHAM4/OPYC3

Morales et al. 2007 *Changes in European ecosystem productivity and carbon balance driven by regional climate model output. Global Change Biology* 13: 108-122

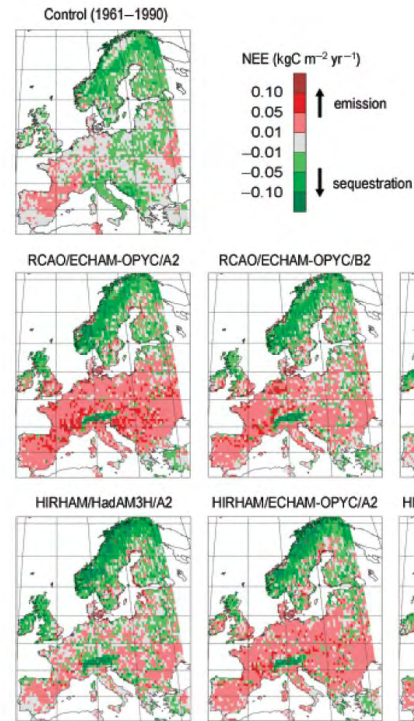


Fig. 5 Net ecosystem carbon exchange (NEE) for the scenario period 2071–2100 simulated by LPJ-GUESS under the 10 regional climate model-generated climate scenarios. Results for the control period 1961–1990 are included for comparison. Negative values represent an uptake of carbon by terrestrial ecosystems.

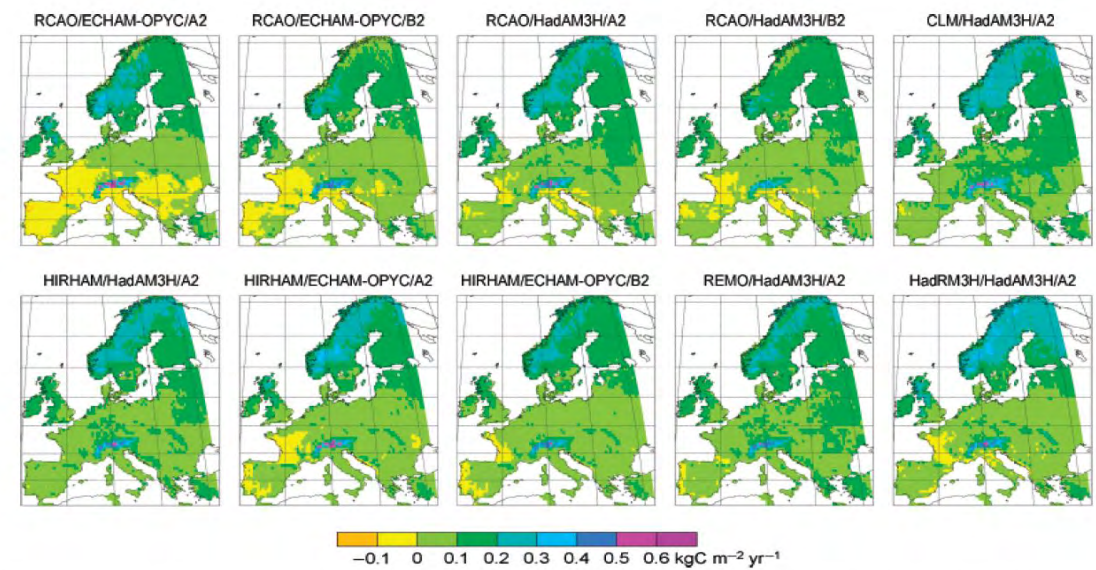


Fig. 3 Net primary production changes by 2071–2100 compared with the control period (1961–1990) simulated by LPJ-GUESS under the 10 regional climate model-generated climate scenarios.

Different impacts between north and south Europe

S. Europe growing season water deficits

N Europe – changes in growing season length

Choice of GCM boundary conditions more important than RCM

Between 1991-2100 modelled impacts on carbon balance range from a sink of 11.6GtC to a source of 3.3 Gtc

The Sahel has become greener

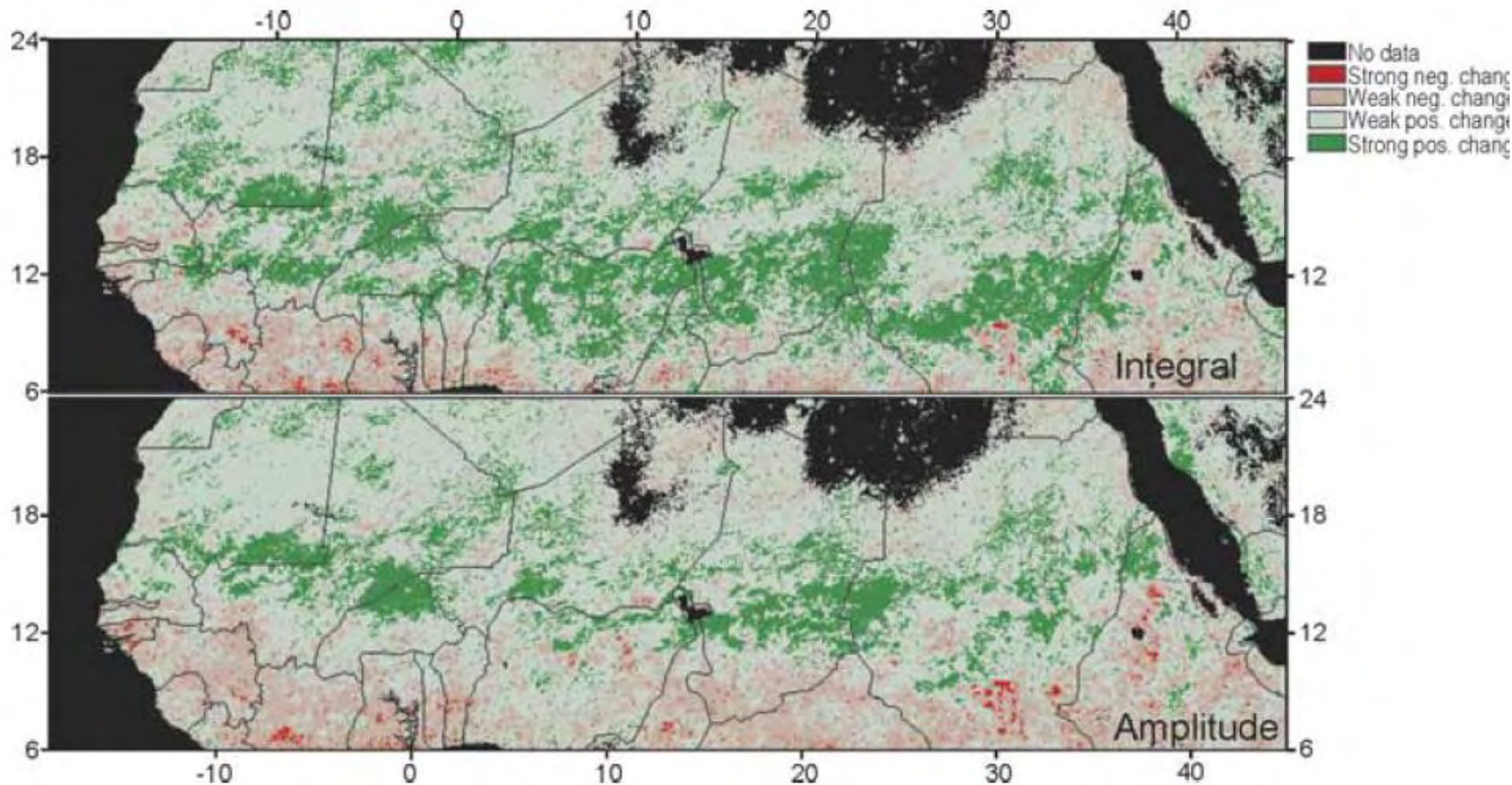


Figure 4. Trend in integrated NDVI (top) and amplitude (bottom) 1982–1999.

Precipitation controls Sahel greening trend.

Hickler et al. 2005

Geophysical Research letters 32 L21415

Using:

NDVI dataset from NOAA/AVHRR – peak
NDVI values for growing seasons 1982-1988

LPJ-DGVM driven by monthly data (CRU05)

Factorial experiments changing different
driving variables – temp, precip, sunshine,
CO₂

**Results – model produced overall trends in
greenness – Fig 1**

**Precipitation almost alone explains
changes (CO₂ minor) – Fig 3**

**Therefore anthropogenic forcing not
required to explain greening trend**

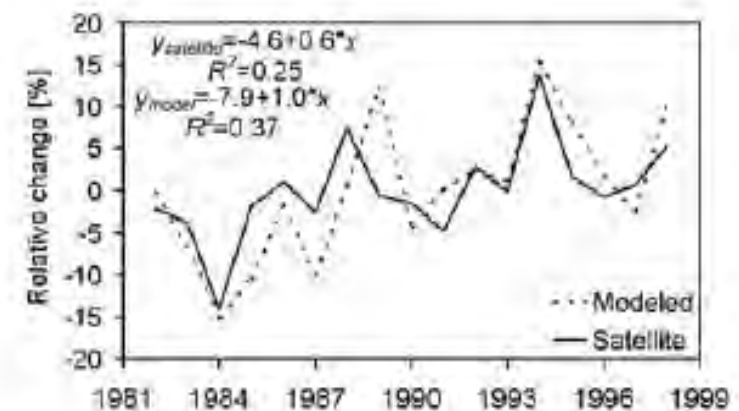


Figure 1. Relative changes (compared to the 1982–1998 mean) of satellite-derived peak-season NDVI and modeled peak-season leaf area index (LAI), averaged over the Sahel.

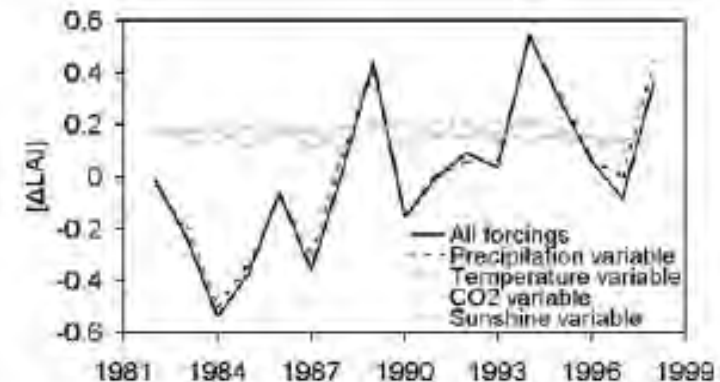
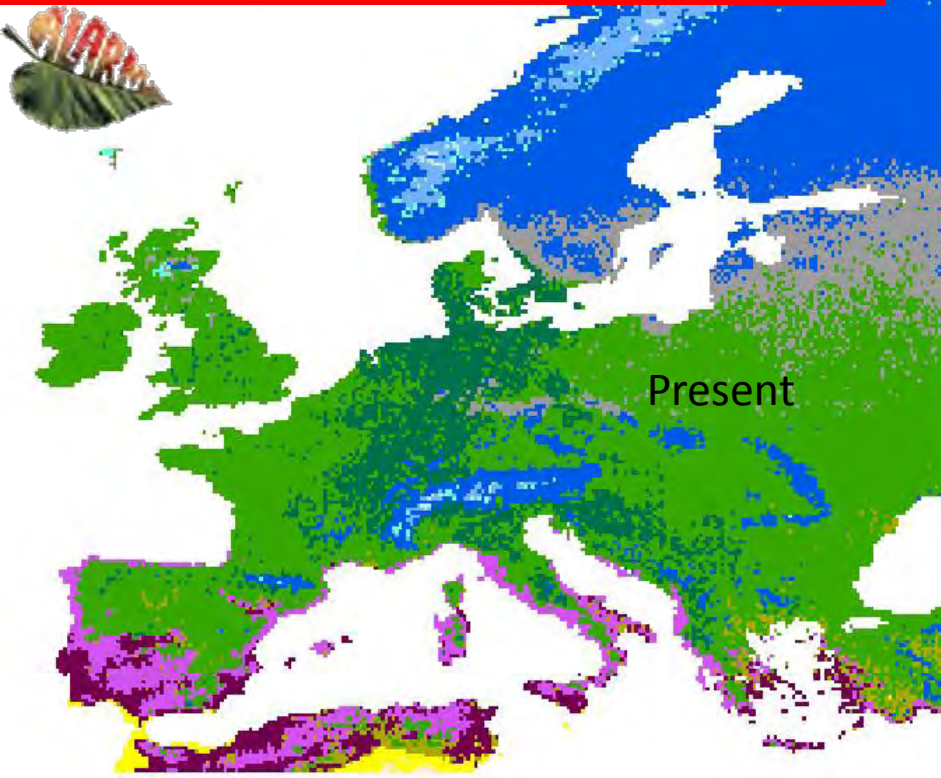


Figure 3. Modeled LAI anomalies, relative to the mean value for 1982–1998, averaged over the Sahel region. In separate simulations, the model was run either with variation in all climatic inputs (all forcings, i.e., changing temperature, precipitation, relative sunshine and atmospheric CO₂ concentration), or varying only one input variable at a time, keeping the remaining input variables constant at 1982–1998 averages.

Present and future bird envelopes, and changing habitats

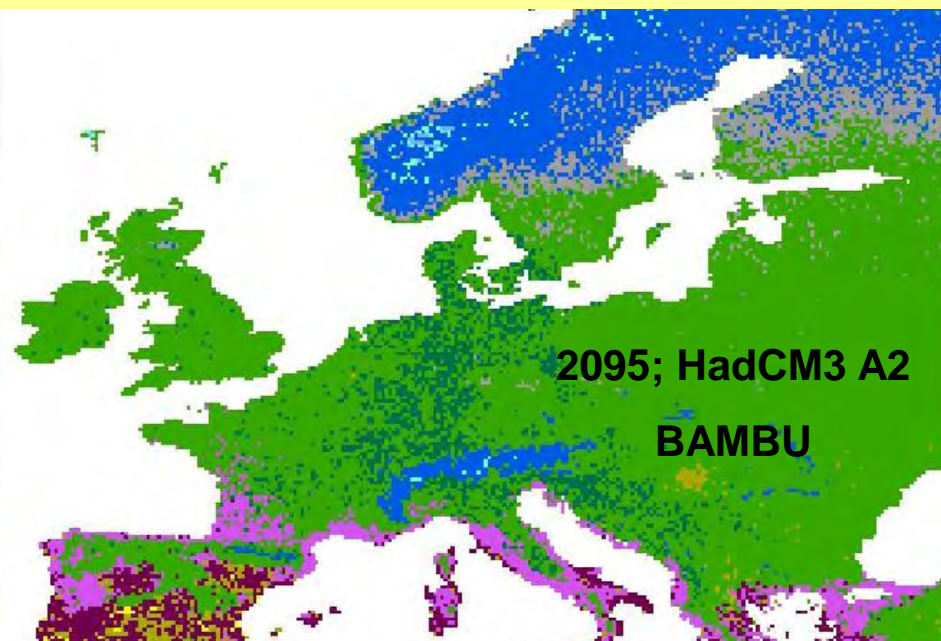


- | | |
|-------------------------------------|--|
| arctic/alpine desert | broad-leaved temperate forest |
| arctic/alpine tundra | evergreen Mediterranean forest |
| boreal/mountain open birch woodland | sclerophyllous Mediterranean scrubland |
| boreal forest | forest steppe |
| hemiboreal forest | steppe |
| beech-dominated temperate forest | desert |

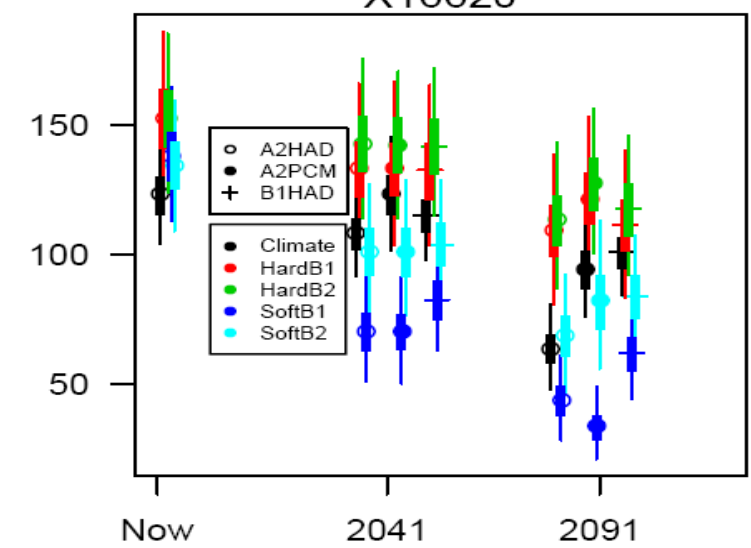


Twite

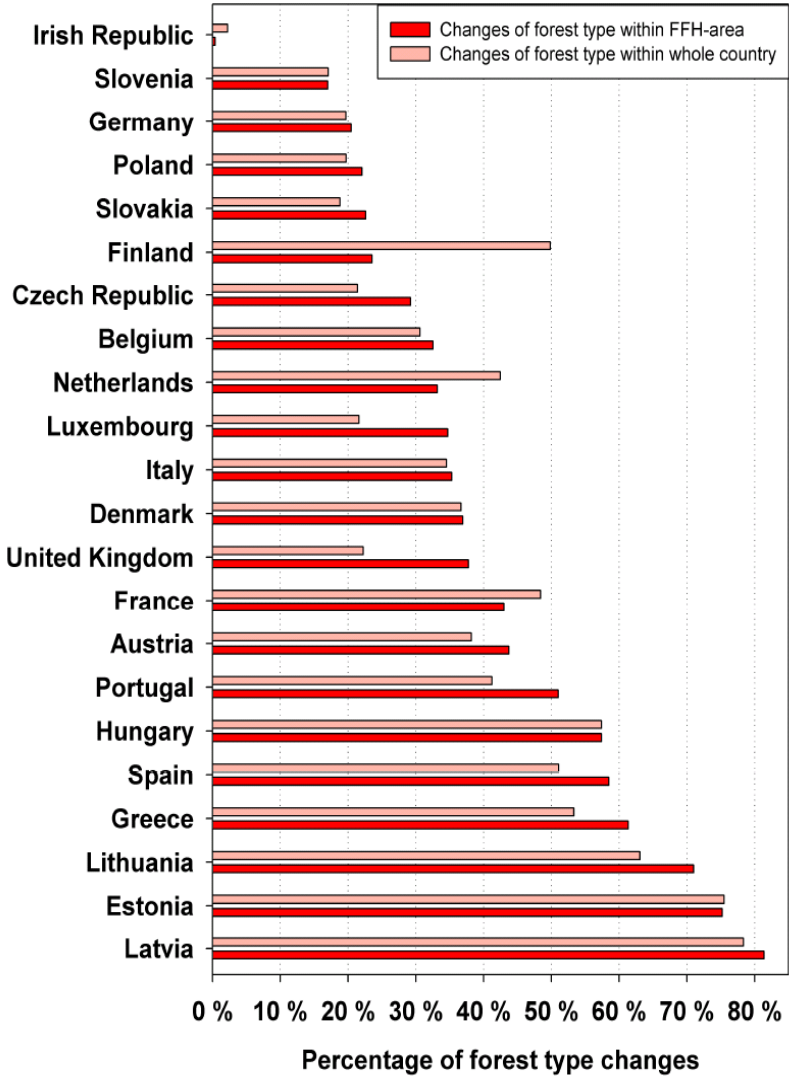
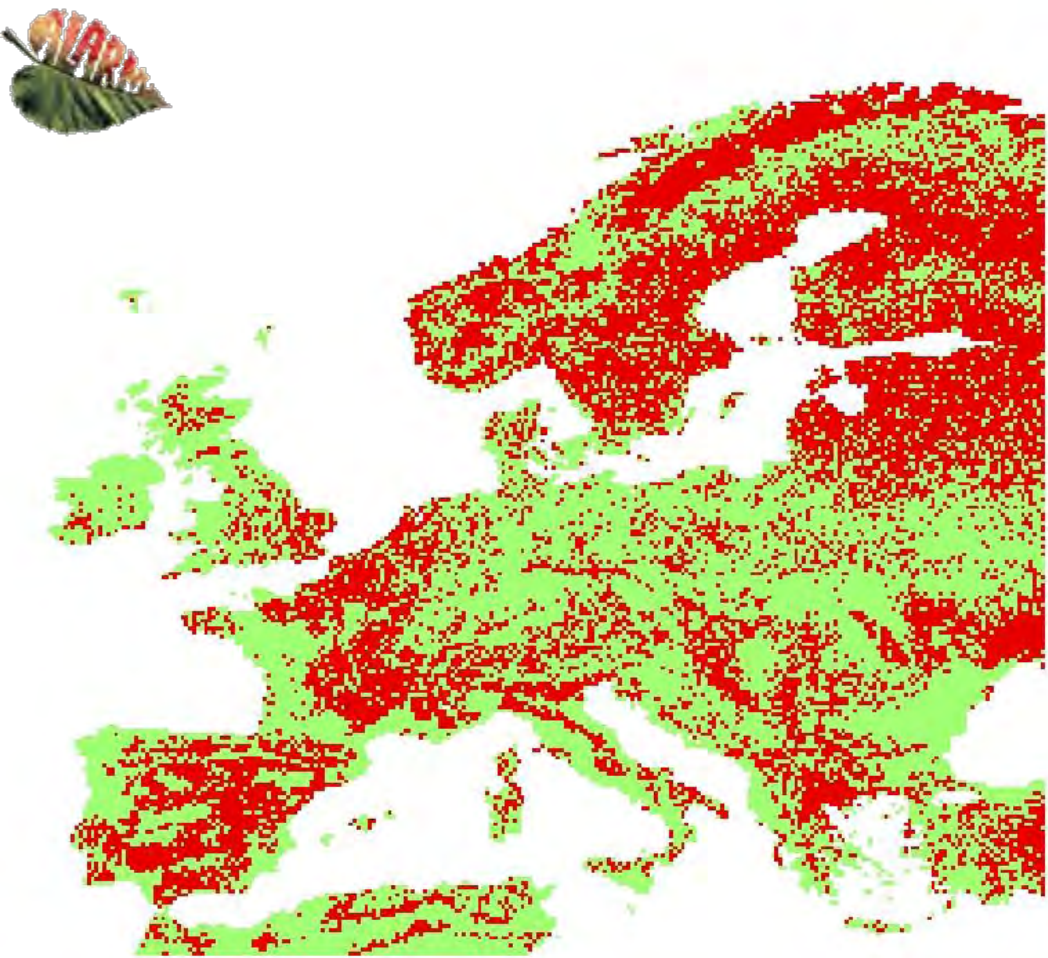
Vegetation/habitats (PNV) simulated by dynamic ecosystem LPJ-GUESS (species level) – time series



X16620



Predicted future vegetation changes in Europe. Grid cells in red change under A2.



Hickler, T. Miller, P., Smith, B., Sykes, M.T. (Lund), Vohland, K. (Potsdam), Kuhn I., (UFZ), Gisecke (Lpool), Franzek, S., Carter, T. (SYKE)

Percentage of forest type changes by country



Vegetation change in the ffh areas HAD B1

Vegetation
Changes in
NATURA 2000
sites (using area of
change in each
country)

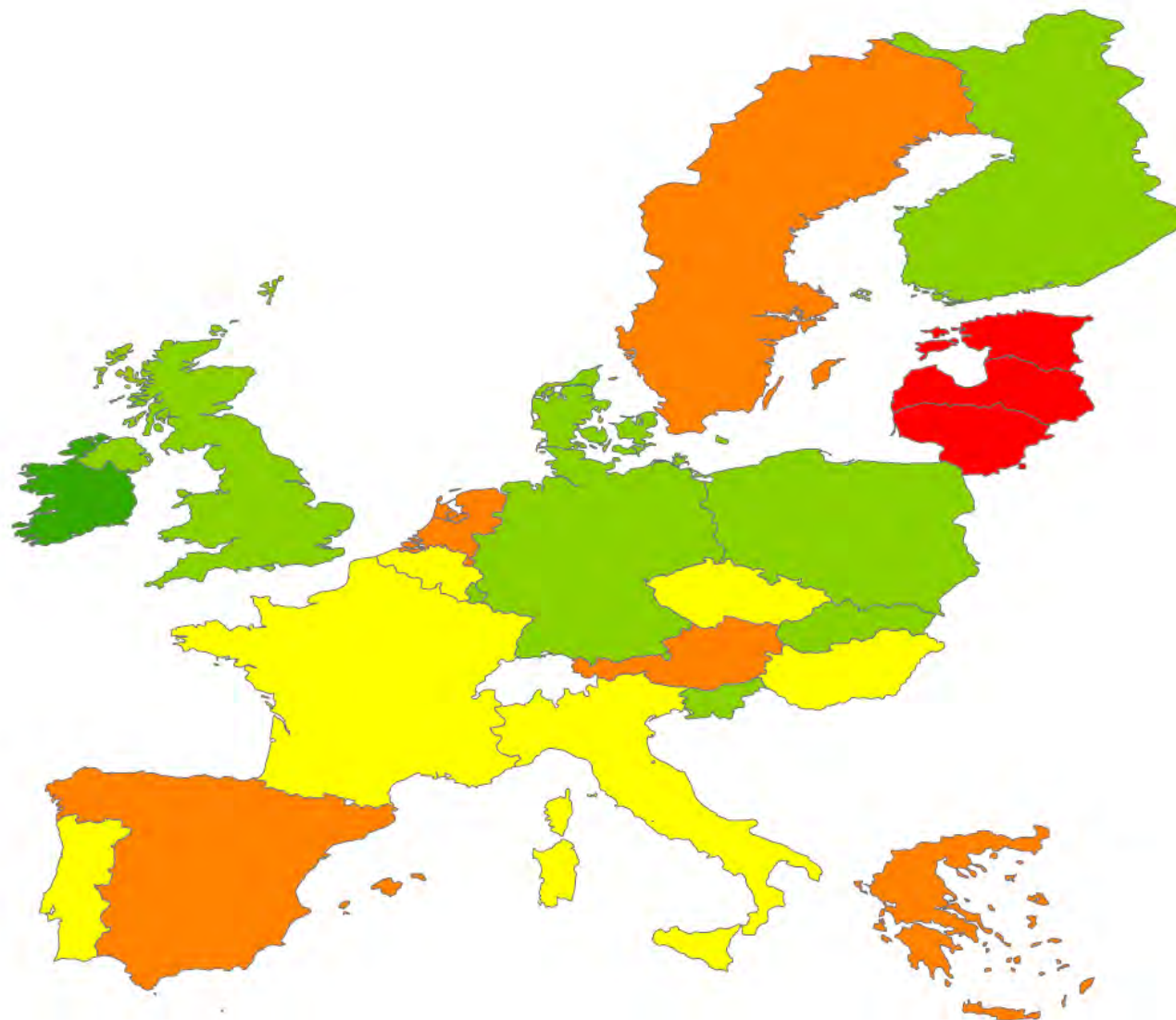


SEDG (B1) climate
only scenario

Legend

Change_ffh

- <15%
- 15% - 30%
- 30% - 45%
- 45% - 60%
- >60%



0 500 1.000 Kilometers

Exporting the ecological effects of climate change

— developed and developing countries will suffer the consequences of climate change, but differ in both their responsibility and how badly it will affect their ecosystems

Chris. D. Thomas, Ralf Ohlemüller, Barbara Anderson, Thomas Hickler, Paula A. Miller, Martin T. Sykes & John W. Williams 2008
EMBO reports vol 9 Special issue 51-58

EMBO
reports

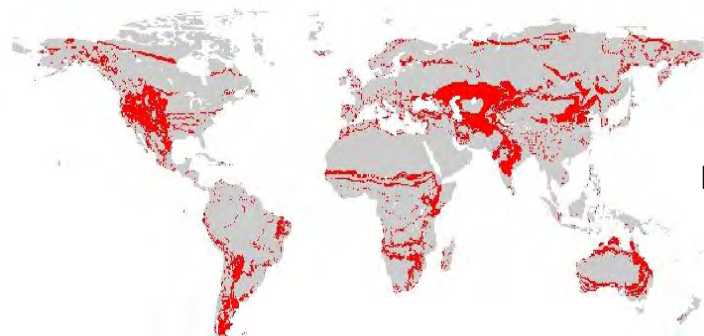
Used LPJ-DGVM in a moderate warming scenario (B1) to model the changes in Biomes between 1931-1960 “1945” and 2041-2050 “2045”

science & society
science & society

The global distribution of risk by 2050

risk = changes to natural systems

1. Biome change yes (red)/no (grey)

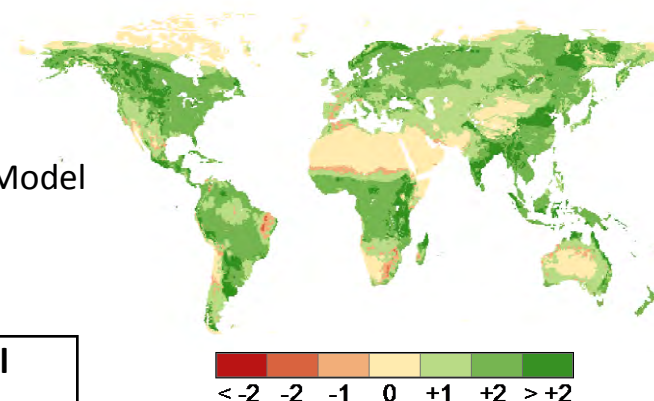


LPJ-GUESS (Lund)

Dynamic Global Vegetation Model

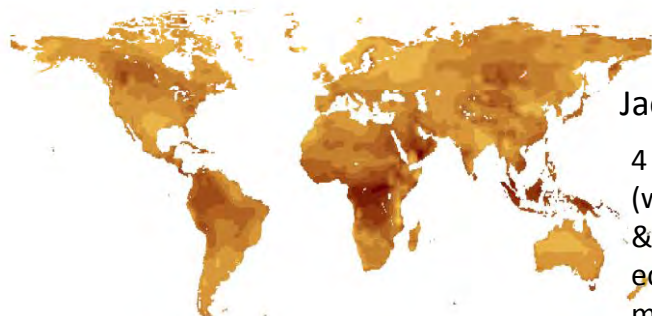
**HadCM3 climate model
B1 emission scenario**

2. Change in woody LAI



<-2 -2 -1 0 +1 +2 >+2

3. Climate outside recent range 1931-1960

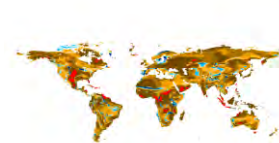


Jack Williams (US)

4 climate variables
(winter & summer temp
& precip). Weighted
equally to produce a
measure of change –
greatest change in tropics

1 2 3 4 5 6 7 8 9

4. Change in analogous area



-100 -80 -60 -40 -20 0 +20 +40 +60 +80 +100 >100

Distribution of shrinking (brown/red) and expanding (blue)

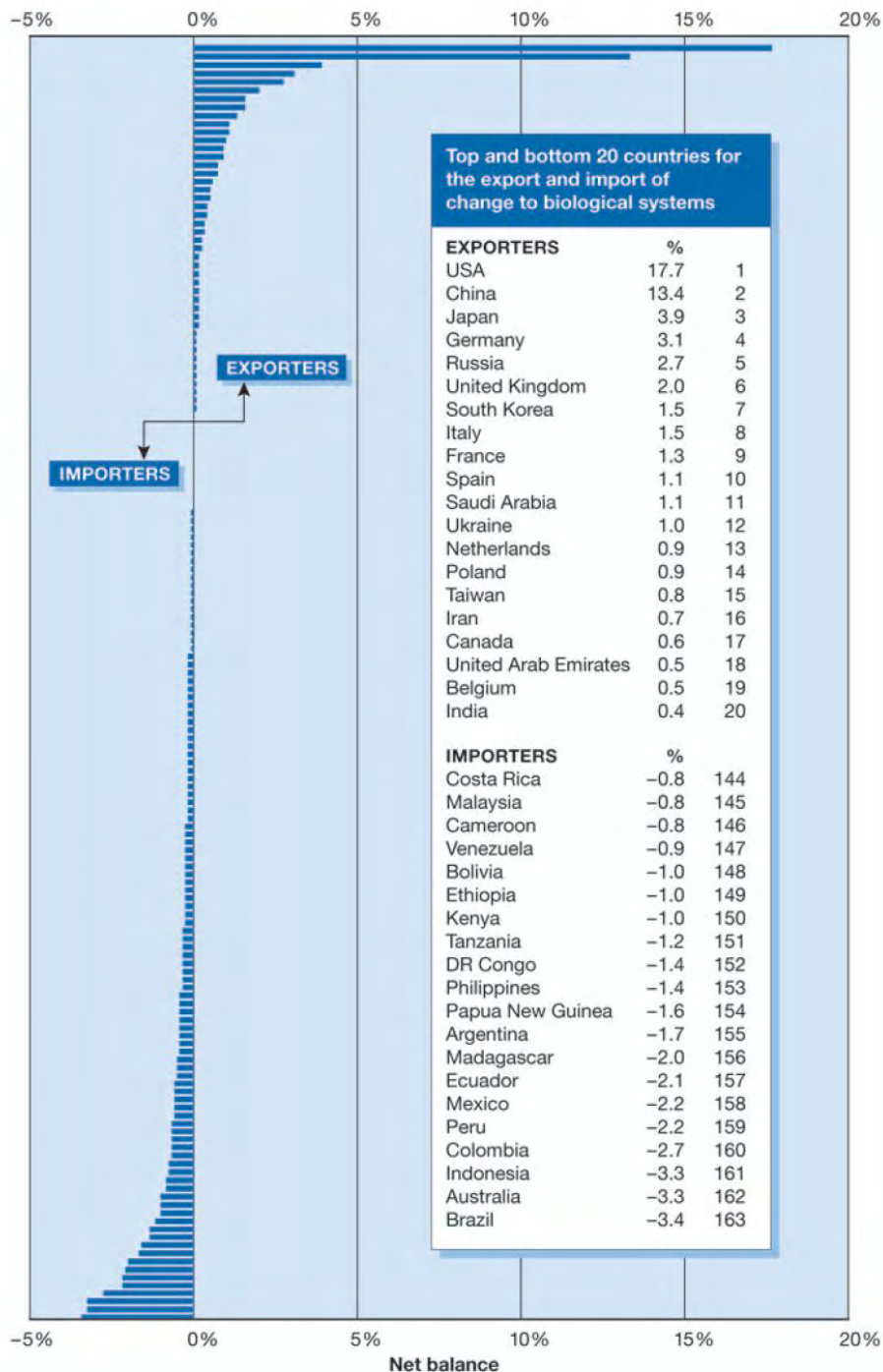


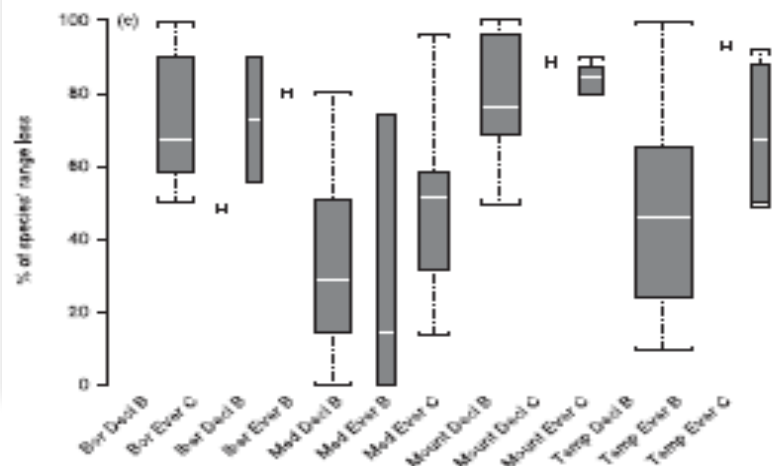
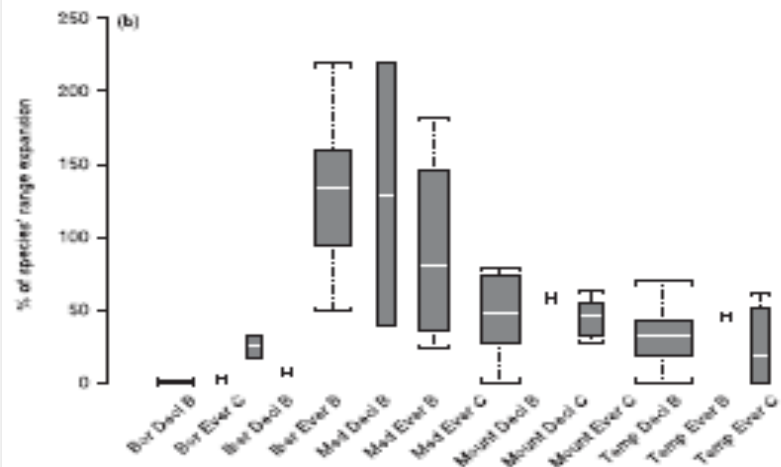
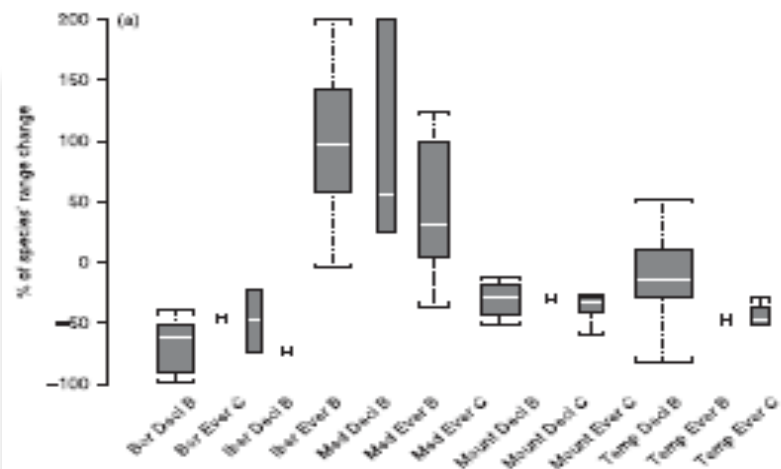
Fig 2 | Global distribution of the export and import of projected changes to natural biological systems through fossil fuel-based CO₂ emissions. Positive values indicate that a country exports global change to ecosystems through emissions above any change within the country, and negative values indicate that a country would be a net recipient of ecosystem/biodiversity change. Each bar represents one of 163 countries (excluding Antarctica and Greenland, most small island nations and some others that occupied less than 50% of any 0.5° global grid cell, and a few others for which full data were not available)

All significant relationships indicate that climate change will cause more severe alterations in ecological systems in high biodiversity than in low-biodiversity countries

1. Greater changes in future woody cover and climate space in countries with low per capita income and GDP AND lower CO₂ emissions.
2. Countries least responsible for climate change and do not have economic means to develop adaptive strategies will experience greatest changes – with severe effects on biodiversity
3. Some countries are exporters of the biological effects of climate change

Ecosystem and bioclimatic modelling in a global change perspective

Martin T. Sykes
Department Physical Geography & Ecosystems Analysis,
Geobiosphere Science Centre,
Lund University
Sweden
Martin.Sykes@nateko.lu.se



Using niche-based modelling to assess the impact of climate change on tree functional diversity in Europe

Wilfried Thuiller^{1,2,3,4*}, Sandra Lavorel^{1,4}, Martin T. Sykes⁵ and Miguel B. Araújo^{1,3,6,7}

Diversity and Distributions, (Diversity Distrib.) (2006) 12, 49–60

122 trees and tall shrubs using AFE for comparison – then classified into Functional groups (PFTs) and phytogeographic classes e.g. Boreal evergreen coniferous versus mediterranean evergreen coniferous

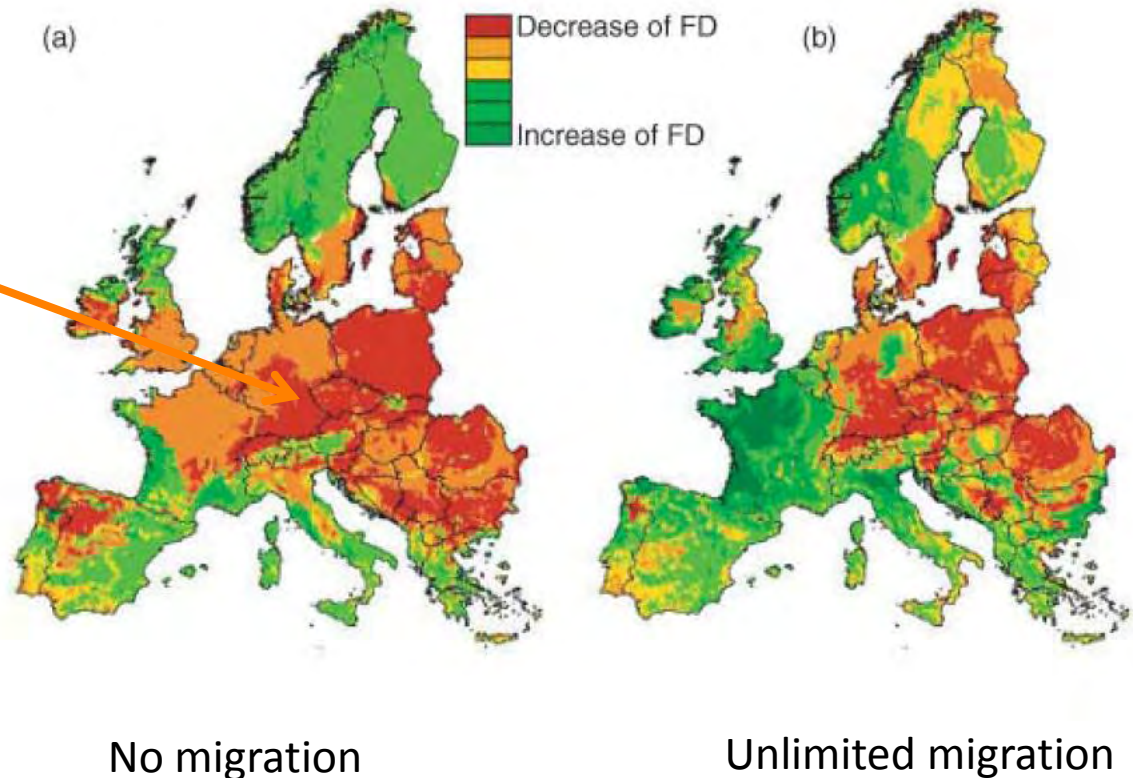
7 bioclimatic variables = mean annual temperature, mean temperature of coldest month, mean annual precipitation, mean winter precip, mean summer precip, GDD and AET/PET

On a random sample of data for each species GLM, GAM, CTA, ANN models were calibrated and then evaluated against the rest of the data

Simpson's diversity index variation
between current and 2080
Functional diversity with and without
unlimited migration

Temperate areas
lose FD due to loss
of broadleaved
deciduous and
coniferous trees

Figure 3 Variability in tree functional diversity under null (a) and unlimited (b) migration assumptions. FD means functional diversity.



The balance

