

Community-level modelling for global biodiversity monitoring – challenges and opportunities

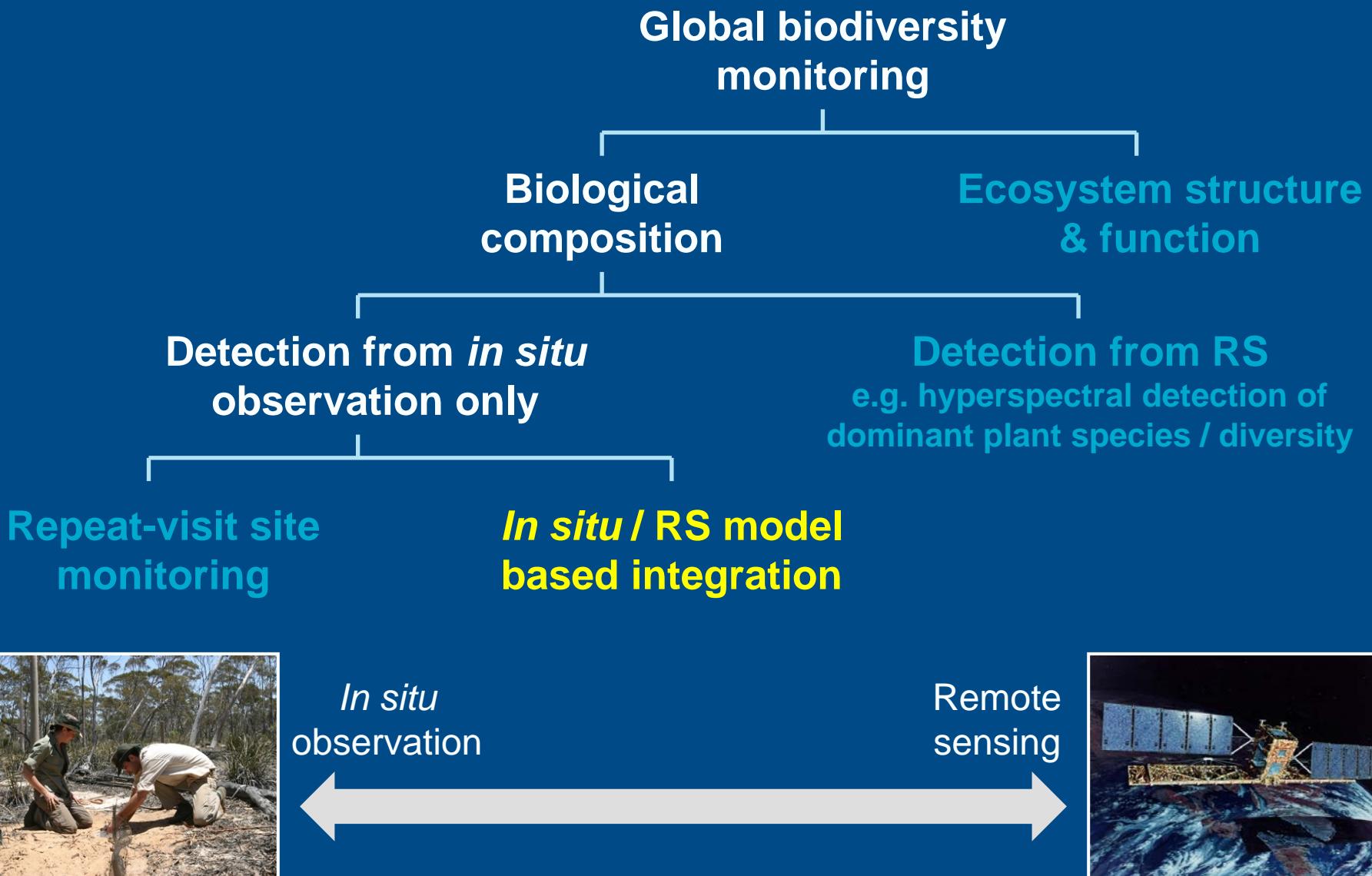
Simon Ferrier

Tom Harwood, Kristen Williams, Andrew Hoskins, Justin Perry

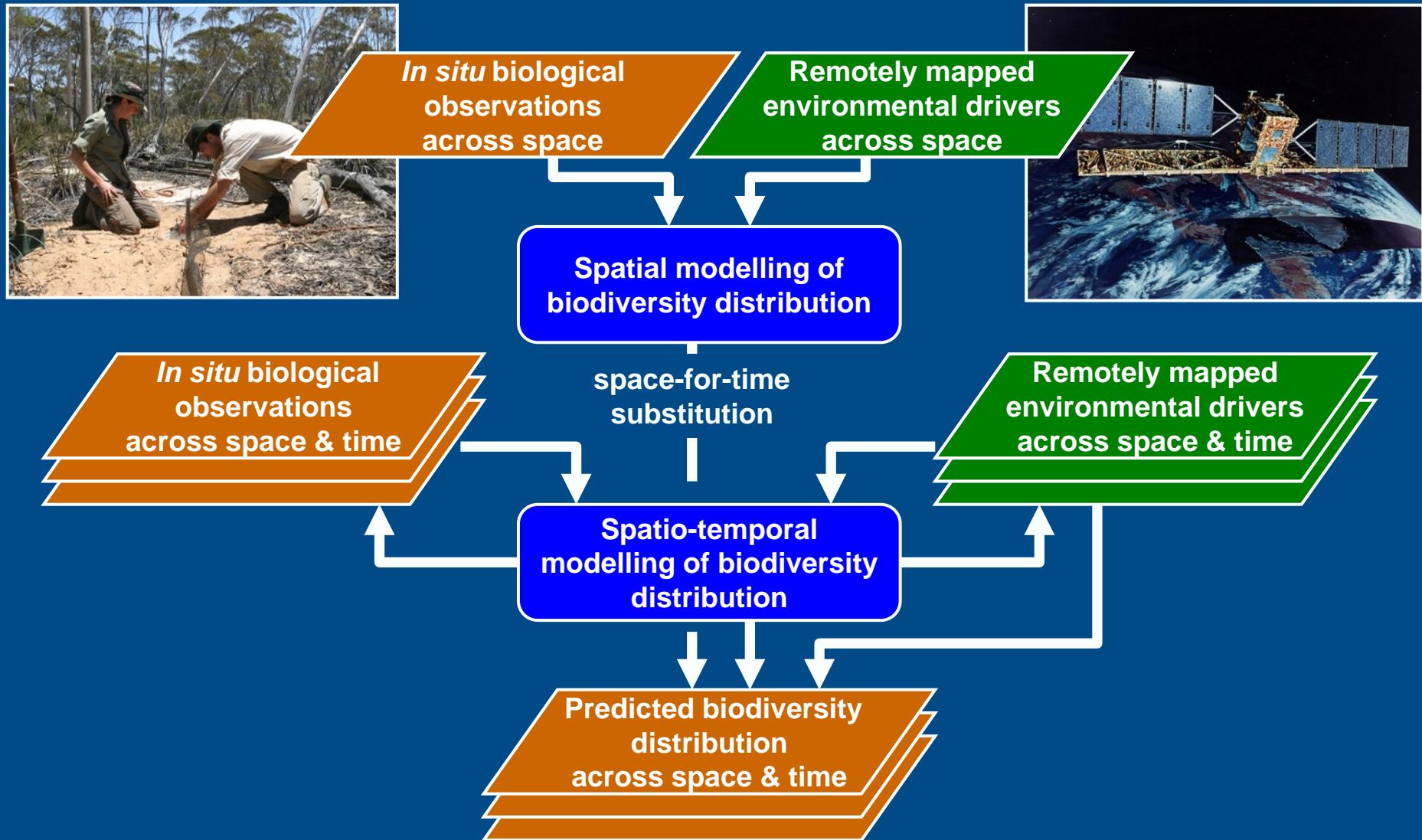
CSIRO LAND & WATER FLAGSHIP
www.csiro.au



Global biodiversity monitoring – multiple pathways to addressing multiple needs

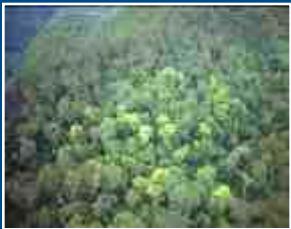


Mapping biodiversity change through model-based integration of *in situ* and remote sensing observations



Spectrum of distributional modelling strategies

Ferrier & Guisan (2006) *Journal of Applied Ecology*



- interested in individual species of particular concern
- reasonable number of records per species

Individual species distribution
(niche) modelling

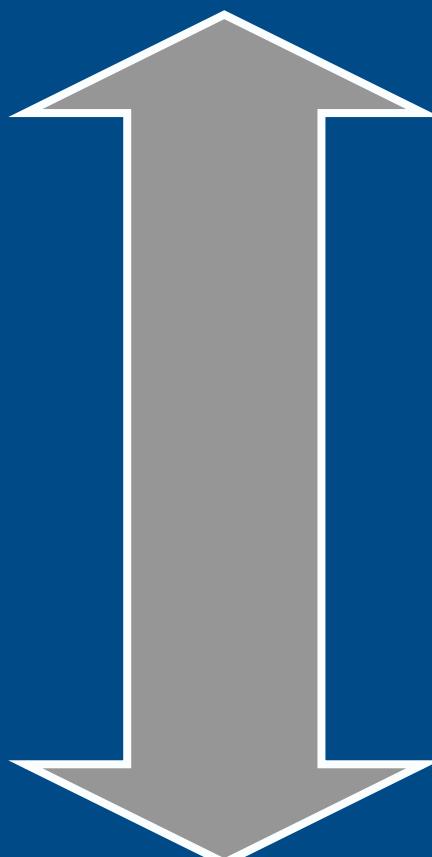
“Predict first, assemble later”
techniques

Simultaneous multi-response
modelling of multiple species

“Assemble first, predict later”
techniques

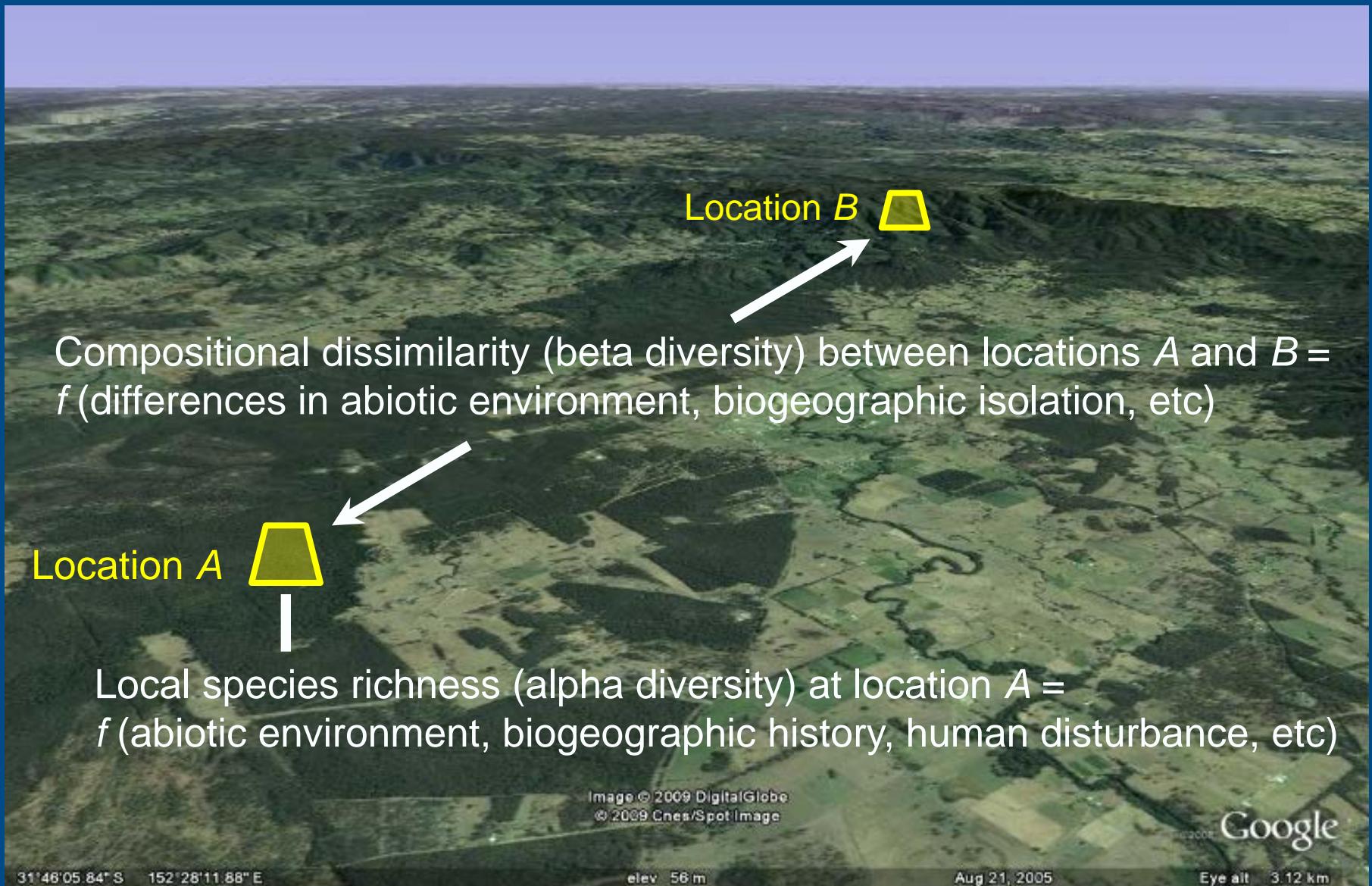
Macroecological modelling of
collective biodiversity properties
(richness, compositional turnover)

- interested in biodiversity as a whole
- huge number of species, each with few (or no) records



} community-level
approaches

Macroecological modelling of collective biodiversity properties – richness and compositional turnover

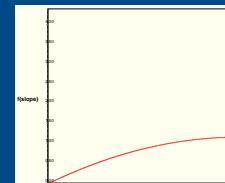
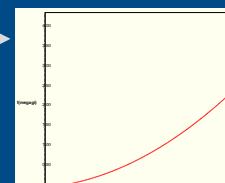
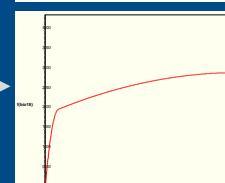
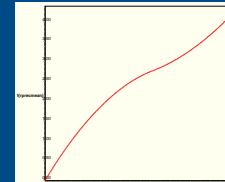


Modelling compositional turnover using generalised dissimilarity modelling (GDM)

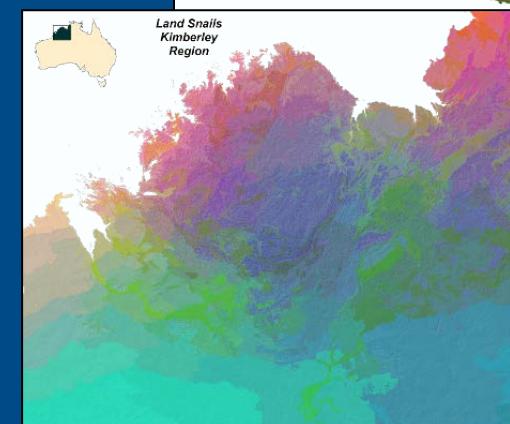
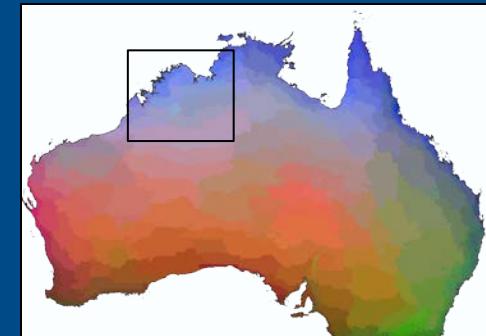
77,000 records of 2,700 land-snail species



Generalised
dissimilarity
modelling (GDM)

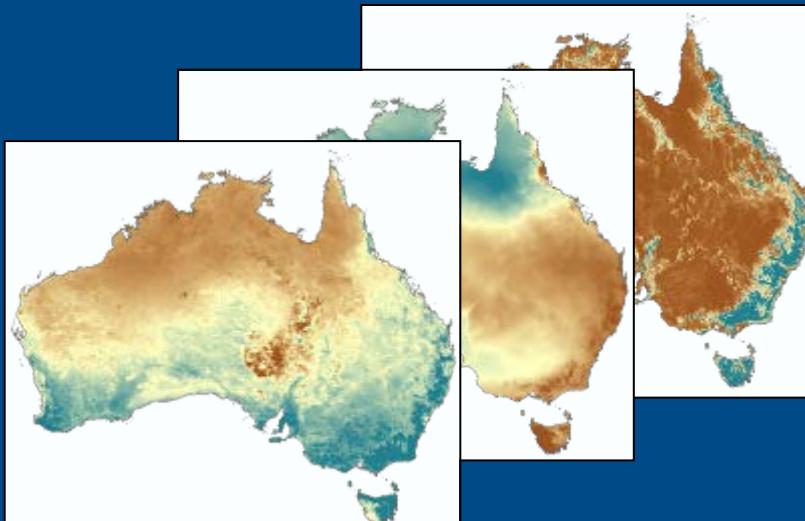


Spatial pattern in
compositional turnover



etc ...

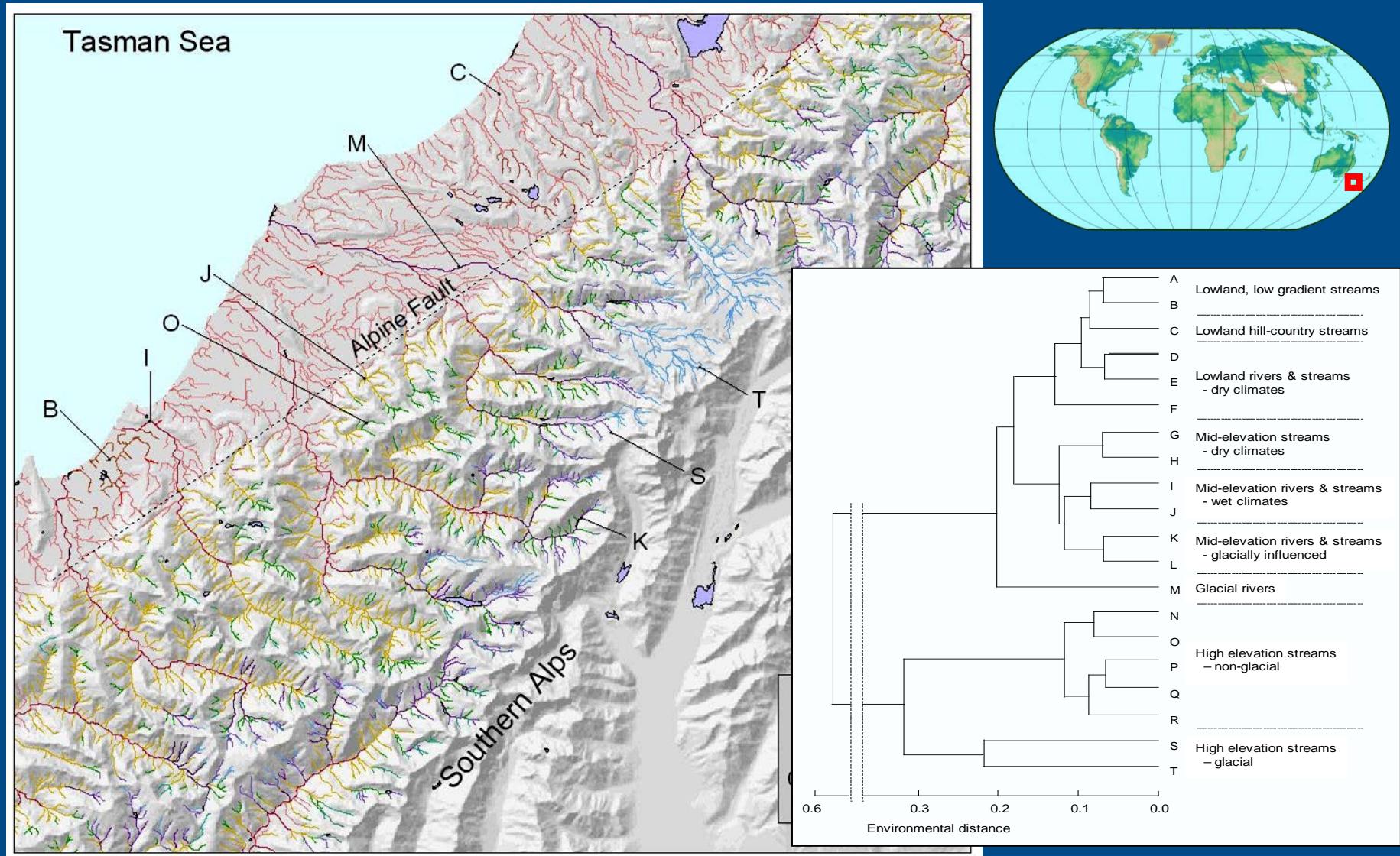
Remotely derived environmental variables:
climate, terrain, soils, geographic isolation etc



etc ...

Funded by Aust. Department of Environment

Applicable across a wide range of scales, from landscape scale ...

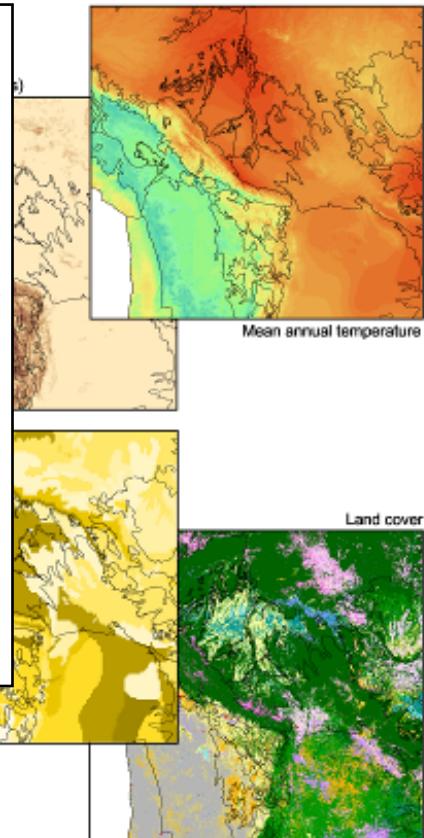
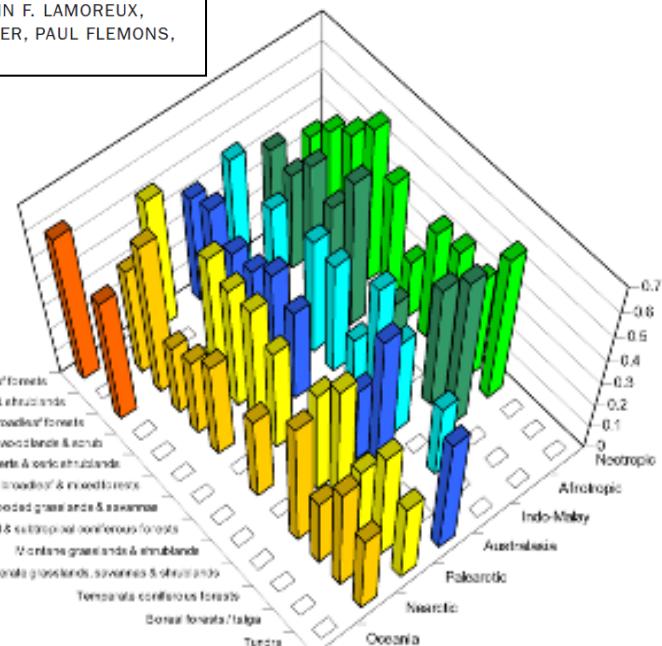
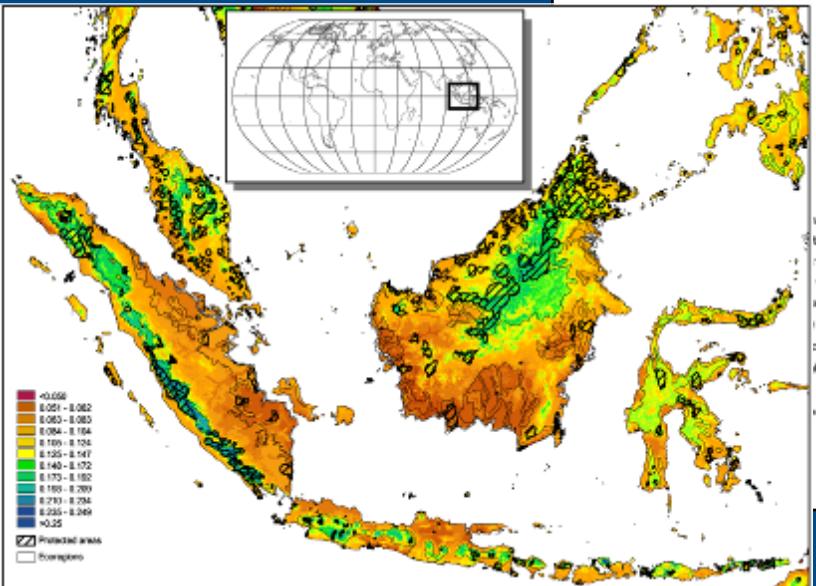


... to global scale

Articles

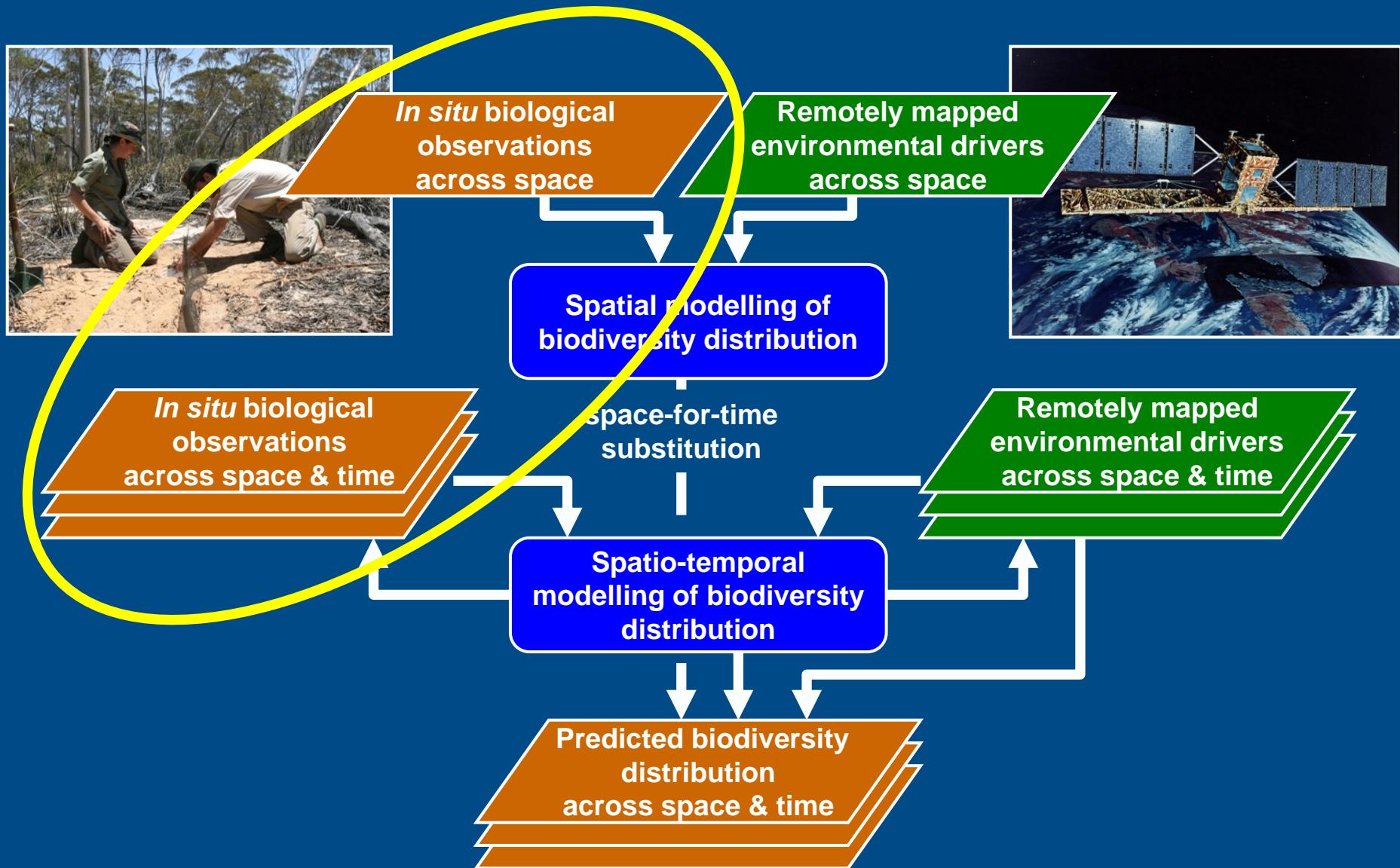
Mapping More of Terrestrial Biodiversity for Global Conservation Assessment

SIMON FERRIER, GEORGE V. N. POWELL, KAREN S. RICHARDSON, GLENN MANION, JAKE M. OVERTON, THOMAS F. ALLNUTT, SUSAN E. CAMERON, KELLIE MANTLE, NEIL D. BURGESS, DANIEL P. FAITH, JOHN F. LAMOREUX, GEROLD KIER, ROBERT J. HIJMANNS, VICKI A. FUNK, Gerasimos A. CASSIS, BRIAN L. FISHER, PAUL FLEMONS, DAVID LEES, JON C. LOVETT, AND RENAAT S. A. R. VAN ROMPAEY



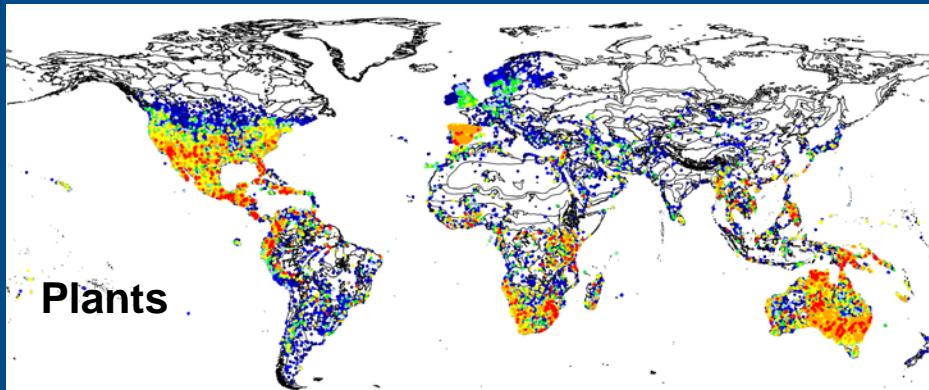
Ferrier et al (2004) *BioScience*

Challenges & opportunities in applying this approach to global biodiversity monitoring: **biological data**

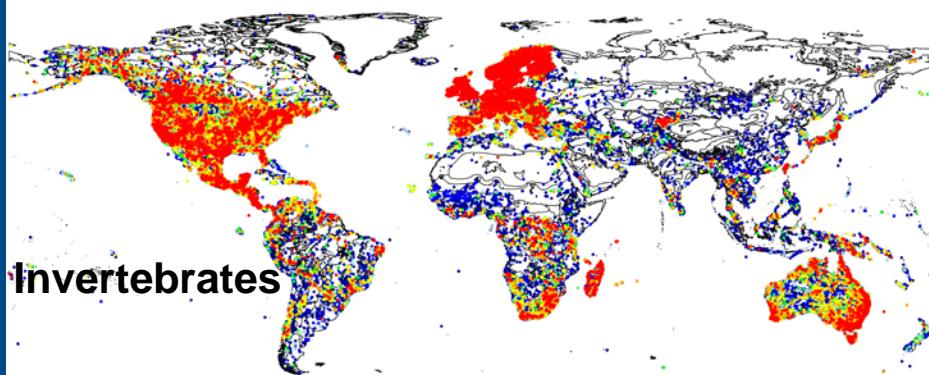


Making effective use of technological advances in biodiversity informatics, citizen science, metagenomics etc

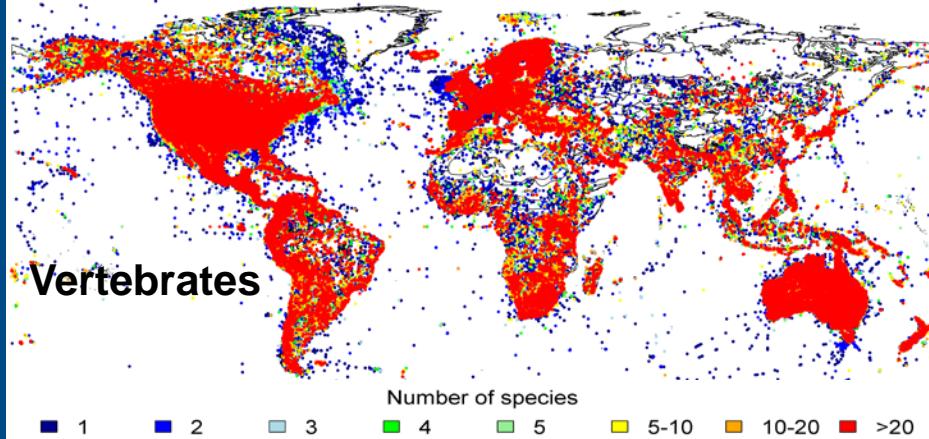
52.5 million records of 254,000 species



13.2 million records of 133,000 species



286.4 million records of 24,000 species



All vascular plants

Ants, Bees,
Beetles, Bugs,
Butterflies,
Centipedes,
Dragonflies, Flies,
Grasshoppers,
Millipedes, Snails,
Moths, Spiders,
Termites, Wasps

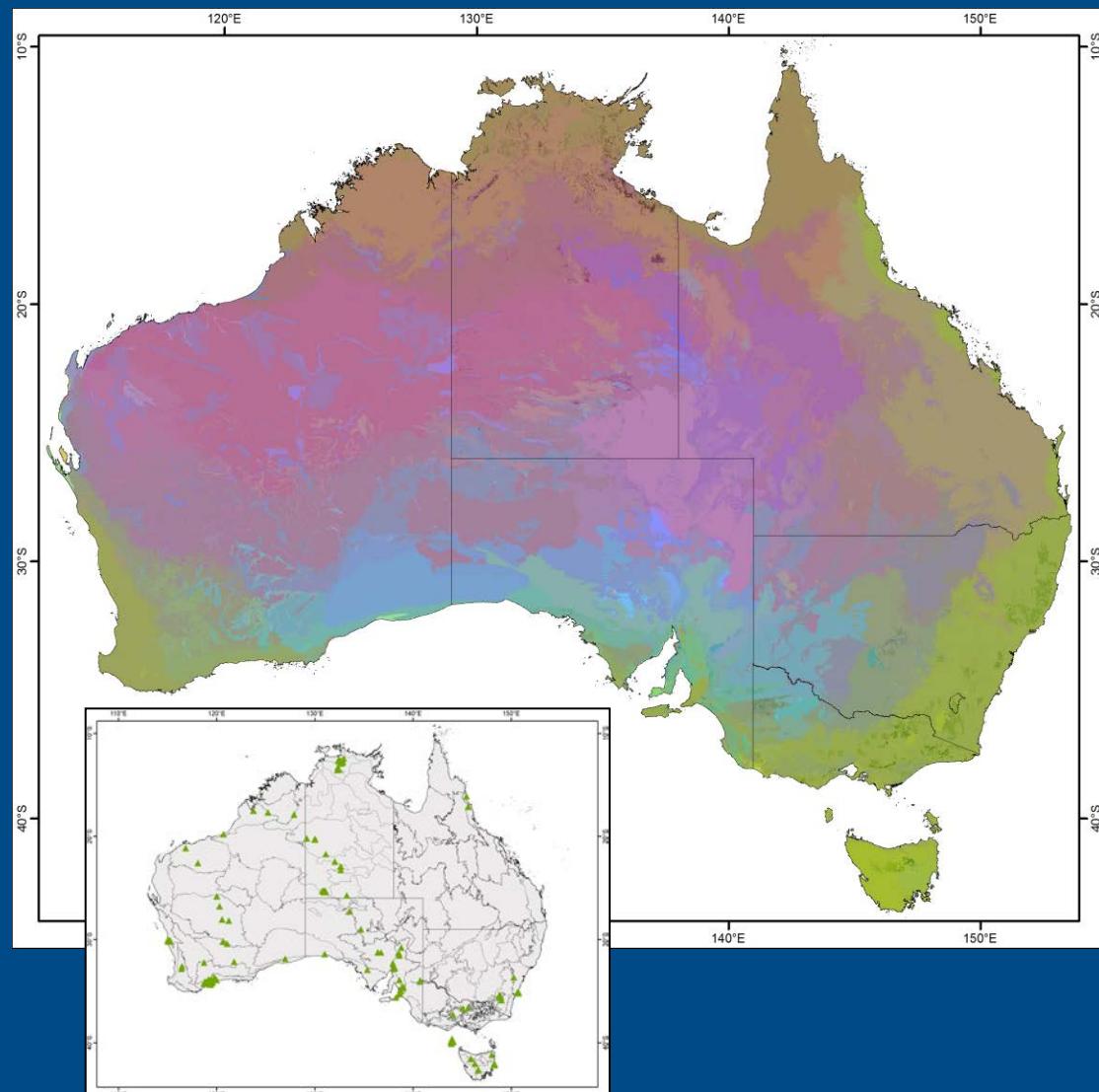
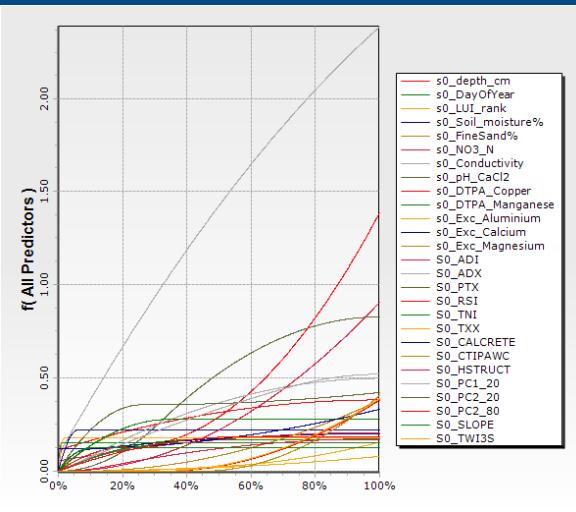
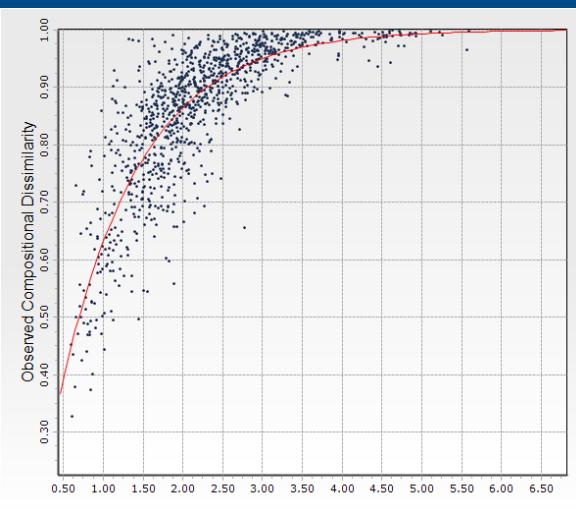


Reptiles

Amphibians
Birds
Mammals



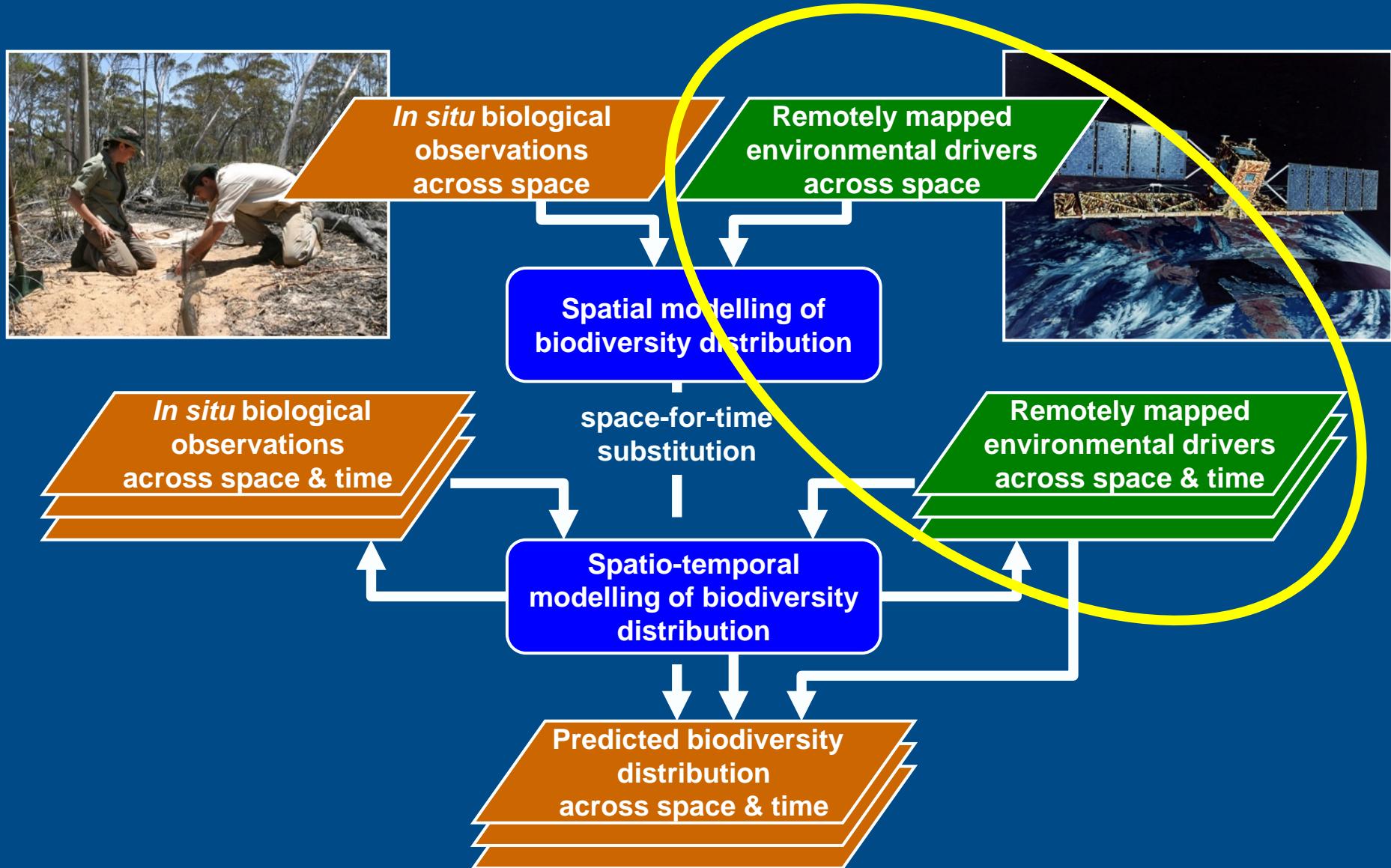
Making effective use of technological advances in biodiversity informatics, citizen science, metagenomics etc



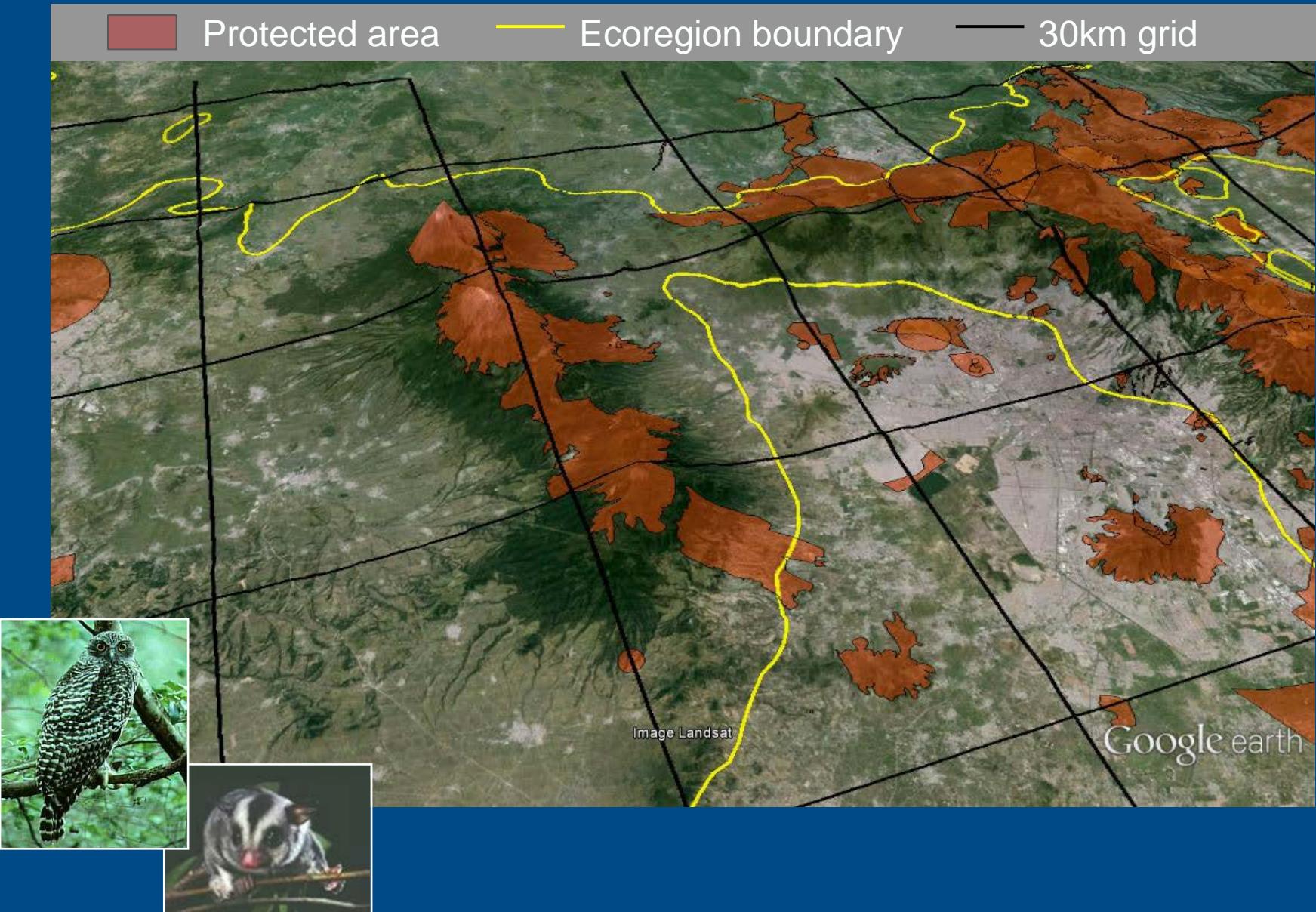
GDM model run 9
Spatial environmental variables (landscape context) only
All soil depths combined
Geographic Distance predictor included
Site-level sampling covariates excluded
Classification:
20,000 samples
300 groups
BGR colouring by compositional similarity

Model Reference:
...IR159705_SIEFsoilworking
GDM Alvaro bck05 GD bck05

Challenges & opportunities in applying this approach to global biodiversity monitoring: environmental data



Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions

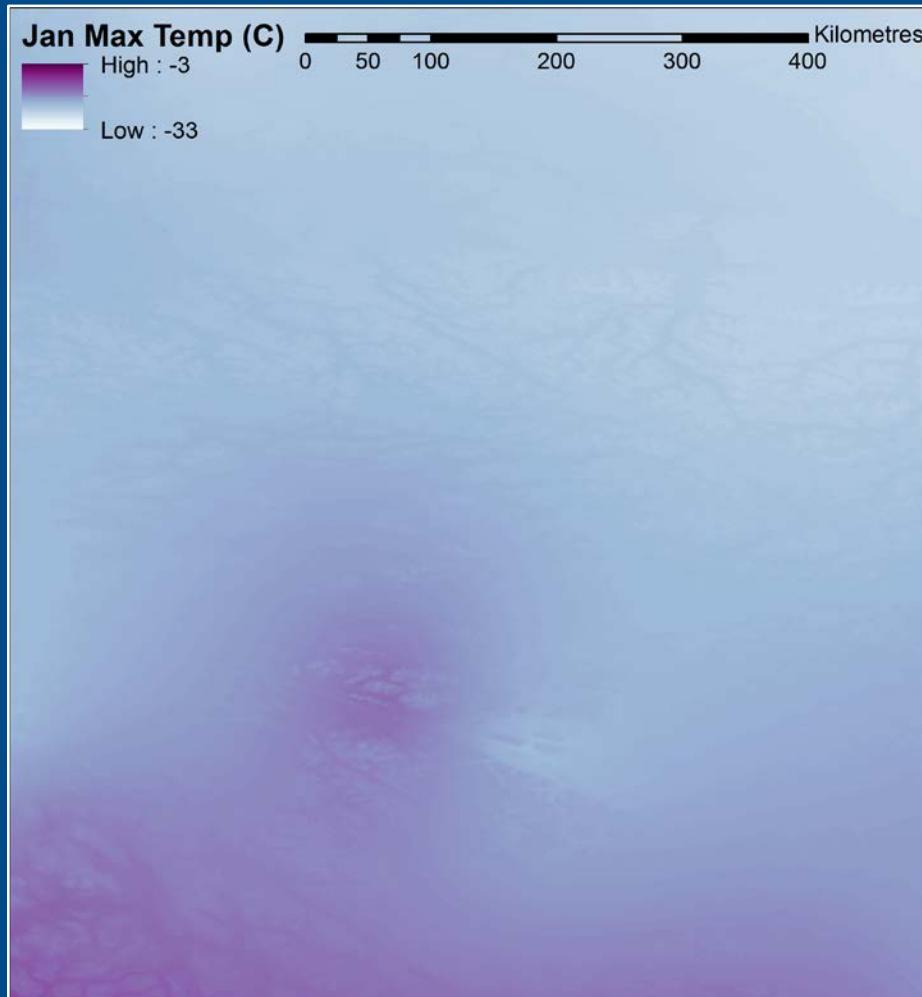


Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions



Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions

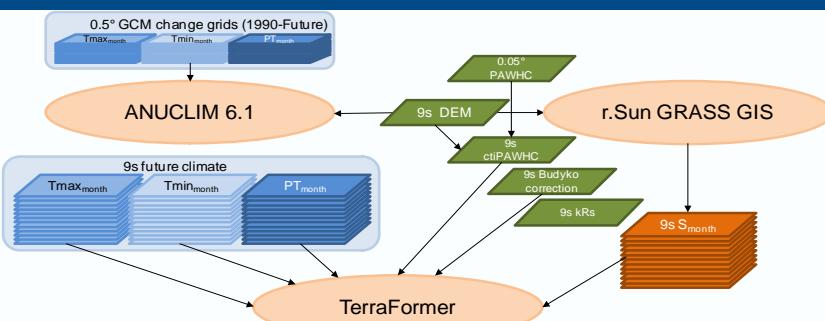
WorldClim: Jan Max Temp



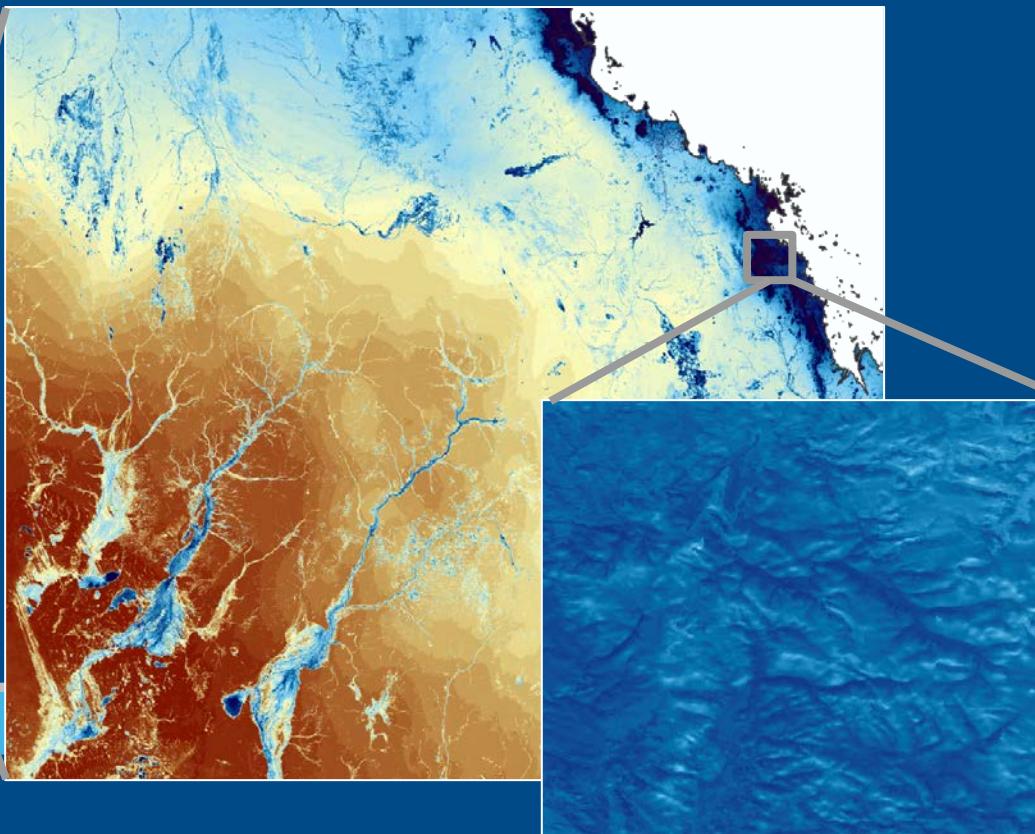
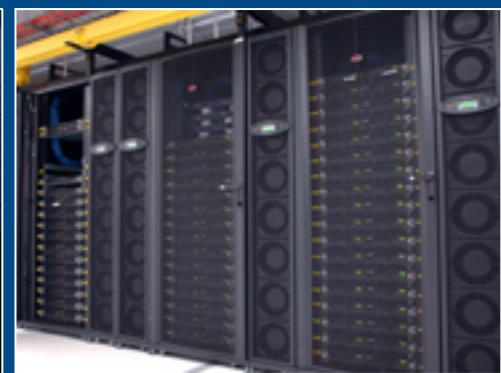
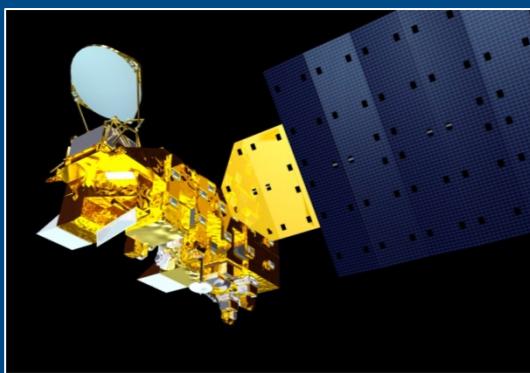
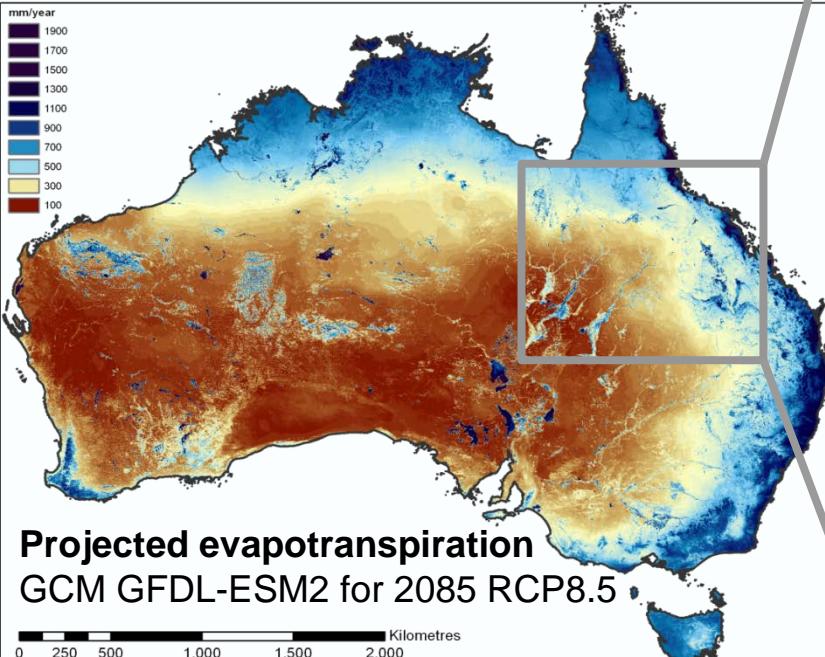
Adjustment for effects of terrain



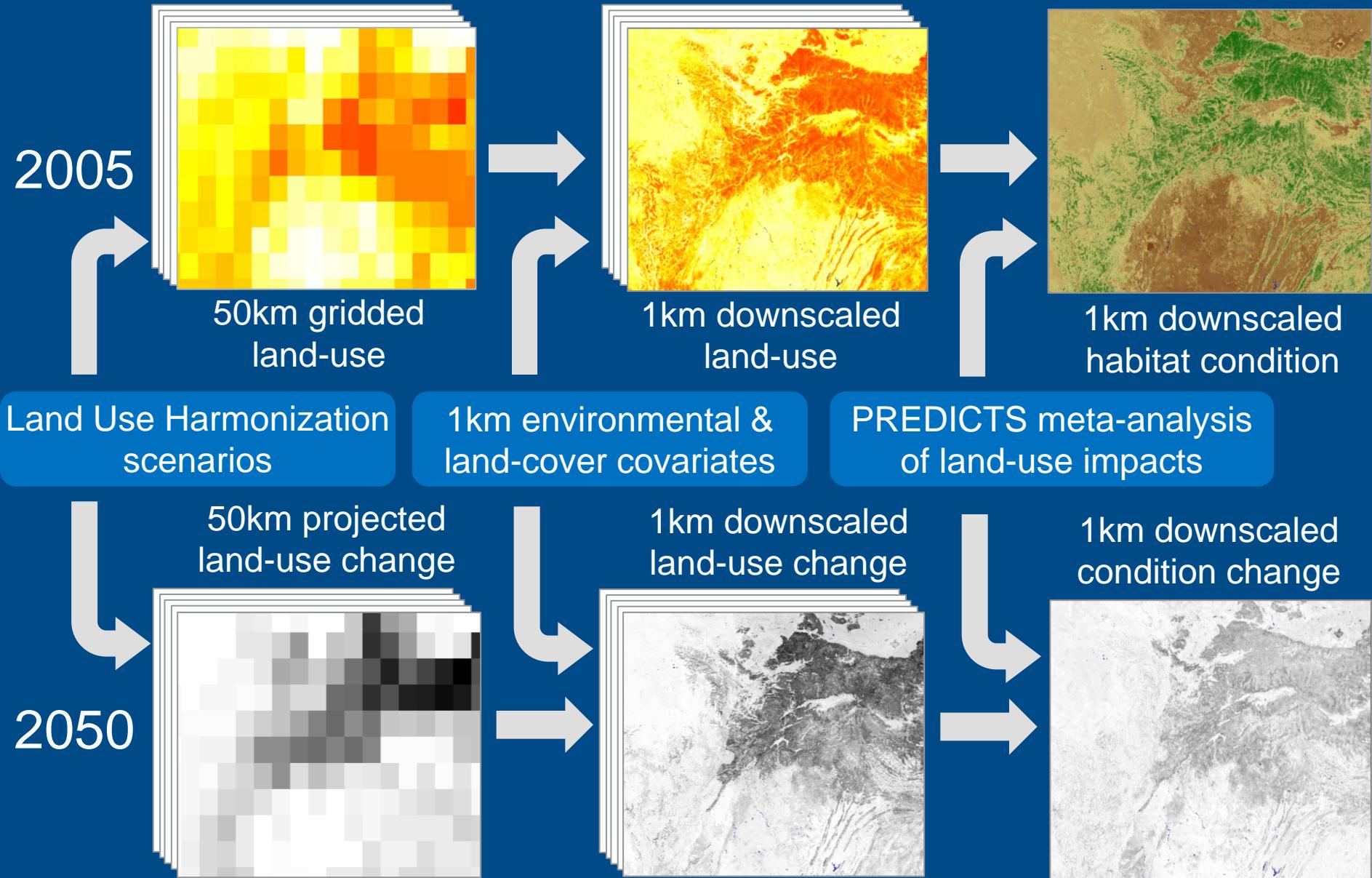
Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions



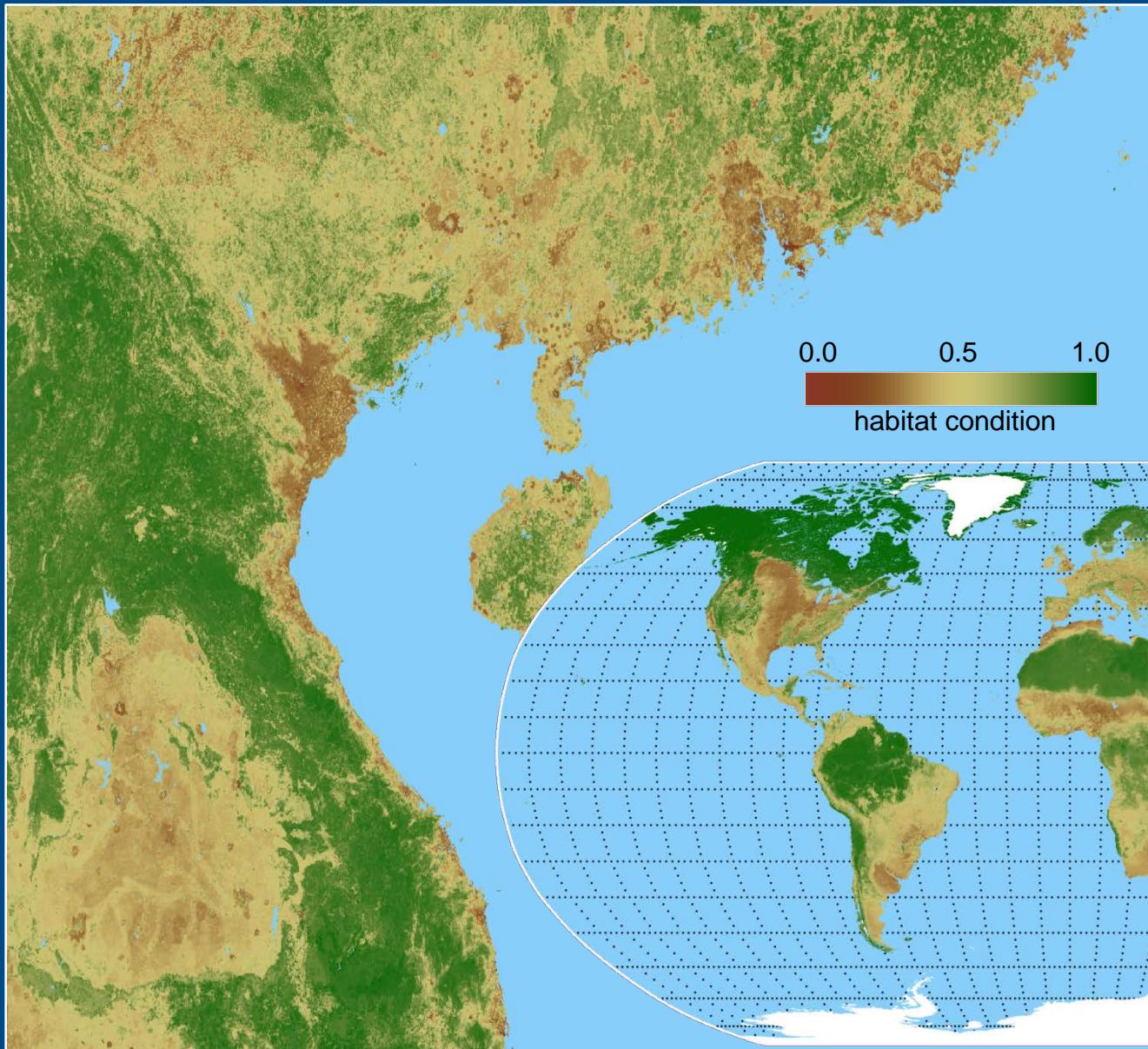
Tmax _{adj}	Tmin	RS _{adj}	PT	EP _{adj}	EA _{adj}	WD _{adj}
Maximum Temperature TXM: Annual Mean	Minimum Temperature TNM: Annual Mean	Shortwave radiation RSM: Annual Mean	Precipitation PTS: Annual Total PTX: Annual Max	Potential Evaporation EPA: Annual Total EPX: Annual Max	Actual Evaporation EAs: Annual Total (simulated)	Water Deficit WDA: Annual Total WDX: Annual Max
TXX: Annual Min	TNI: Annual Min	RSX: Annual Max	PTI: Annual Min	EPI: Annual Min	EAA: Annual Total (topocorrected EAA)	WDI: Annual Min
TXR: Max monthly change	TNR: Max monthly change	TSR: Max monthly change	PTR: Max monthly change	EPR: Min monthly change	WDR: Max monthly change	WDR: Min monthly change
TXR: Min monthly change	TNR: Min monthly change	TSR: Min monthly change	PTS1: Seasonality summer/winter	EPR: Min monthly change		



Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions

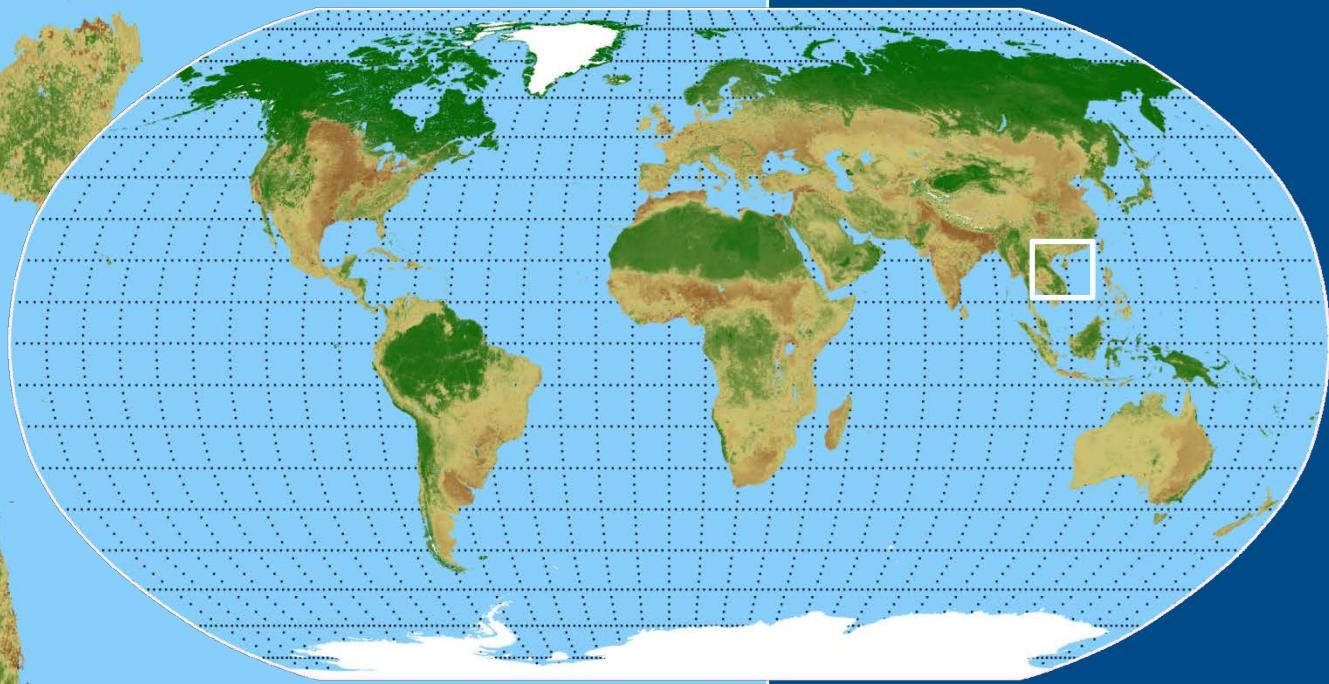


Making effective use of technological advances in mapping biologically-relevant variables at appropriate resolutions

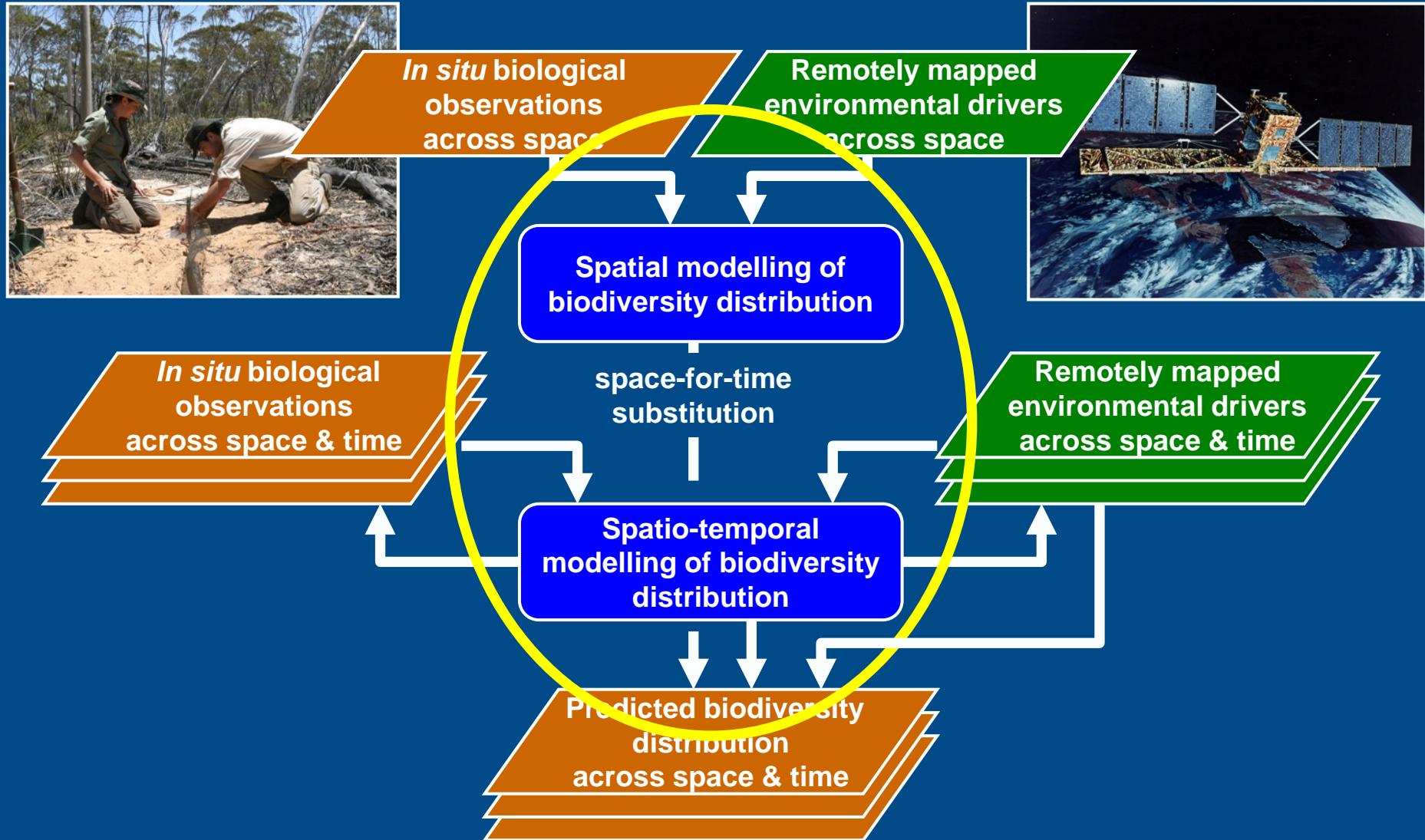


2050 projected
habitat condition

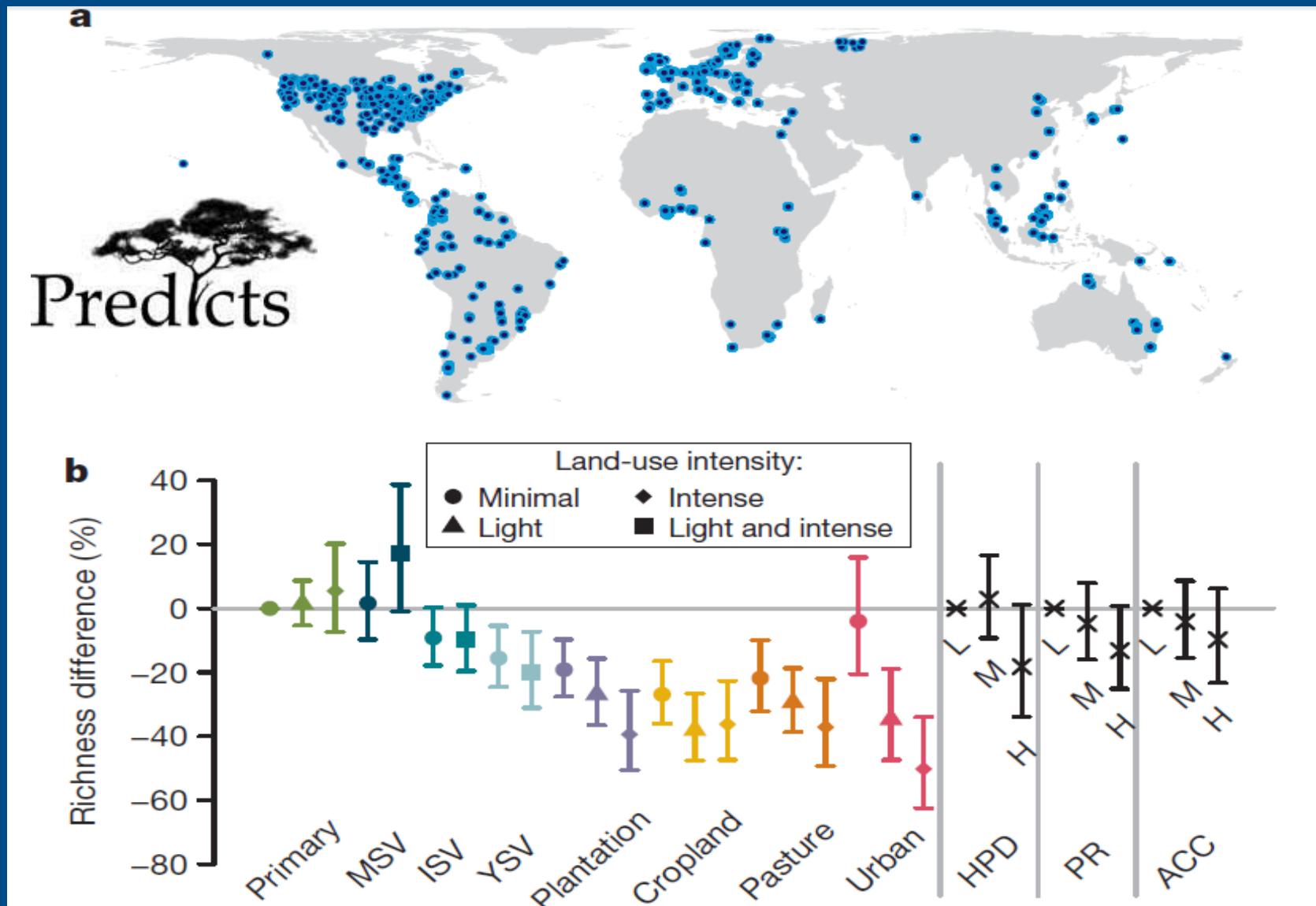
based on downscaled
RCP8.5-MESSAGE
land-use scenario



Challenges & opportunities in applying this approach to global biodiversity monitoring: **modelling techniques**

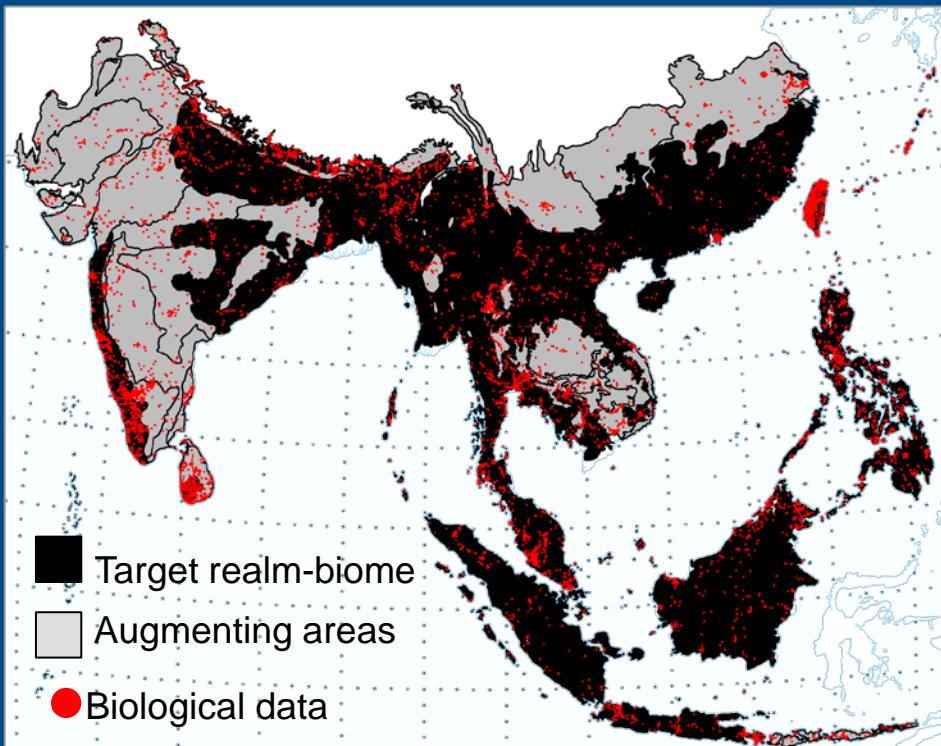


Adapting modelling techniques to extract maximum value from best-available biological & environmental data

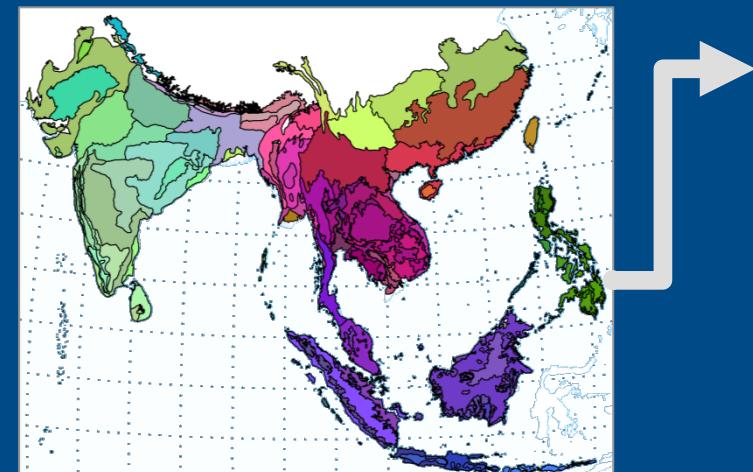
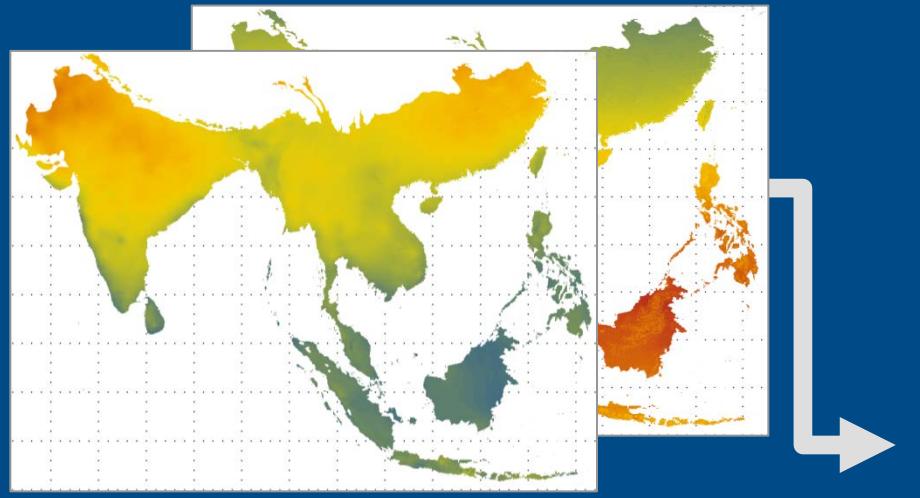


Adapting modelling techniques to extract maximum value from best-available biological & environmental data

Biological data for targeted realm-biome & augmenting realm-biome combinations

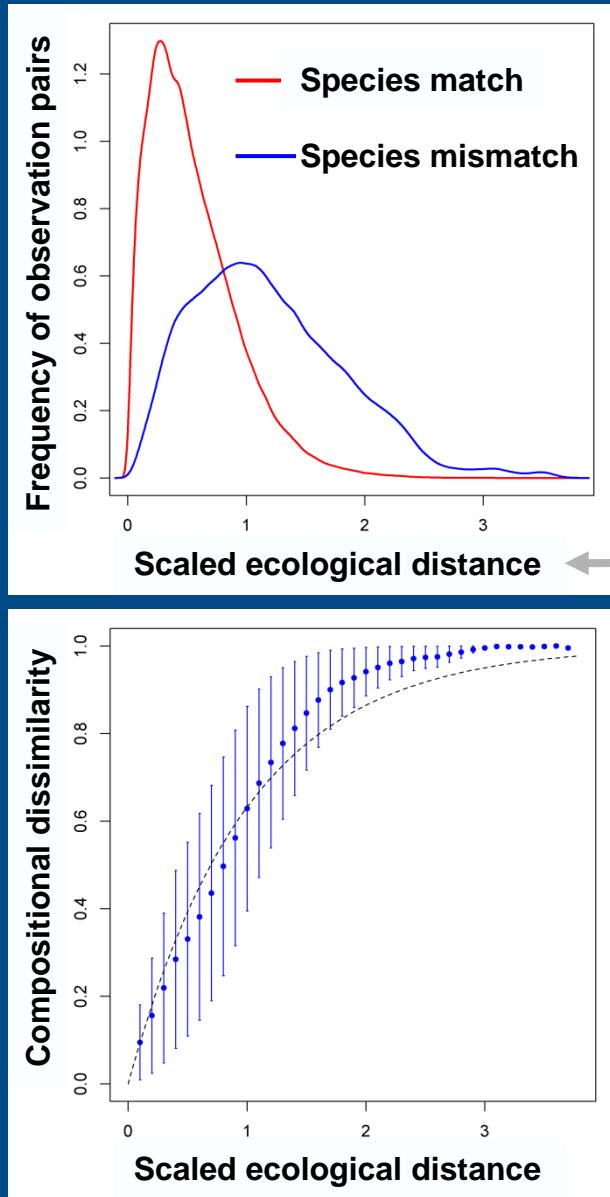


Continuous environmental surfaces

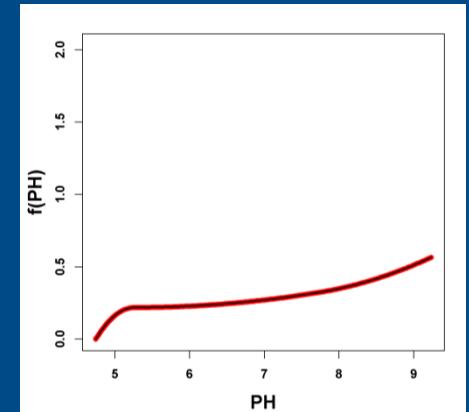
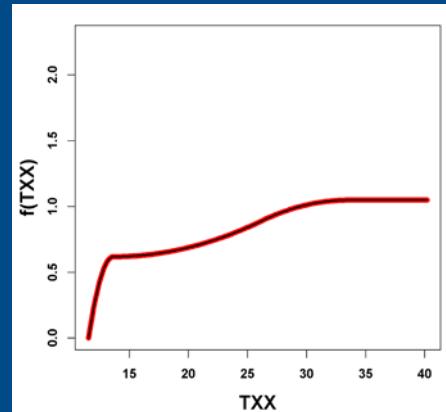
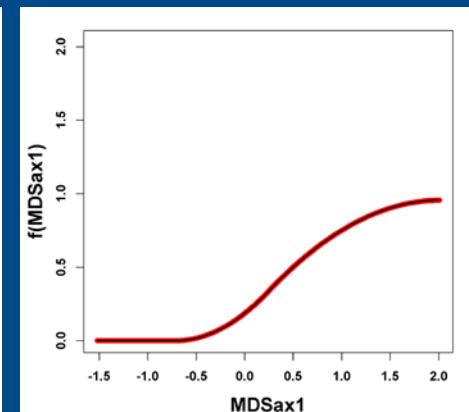
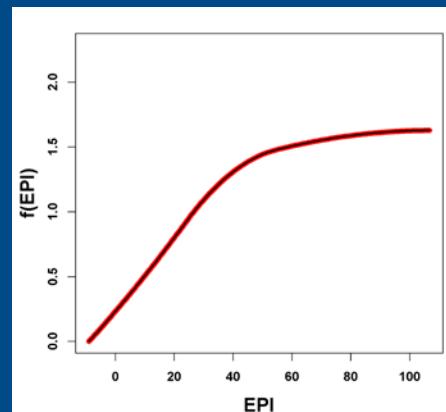


Inter-ecoregional MDS axes

Adapting modelling techniques to extract maximum value from best-available biological & environmental data



Biologically scaled
environmental gradients



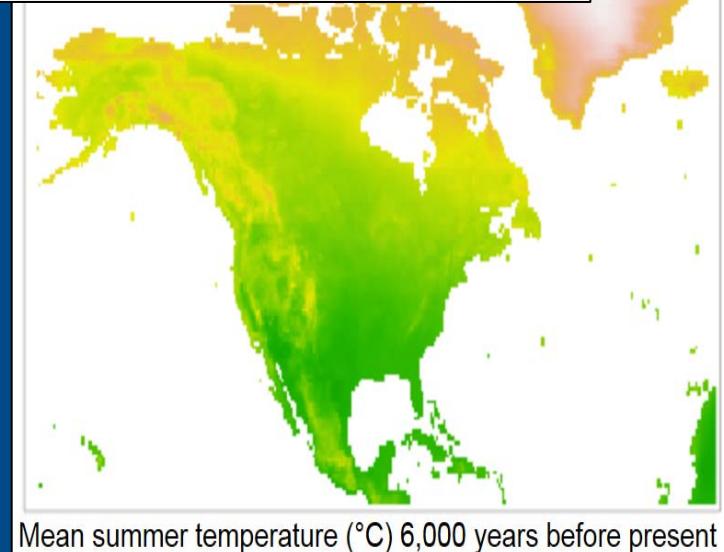
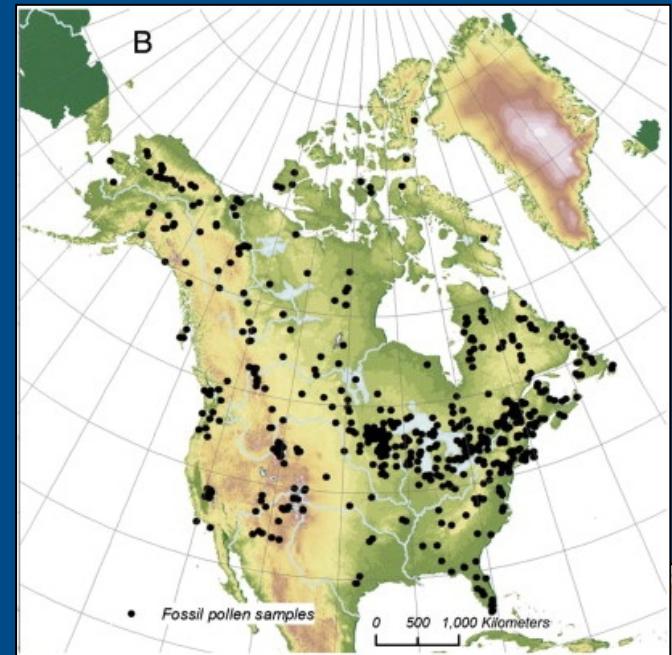
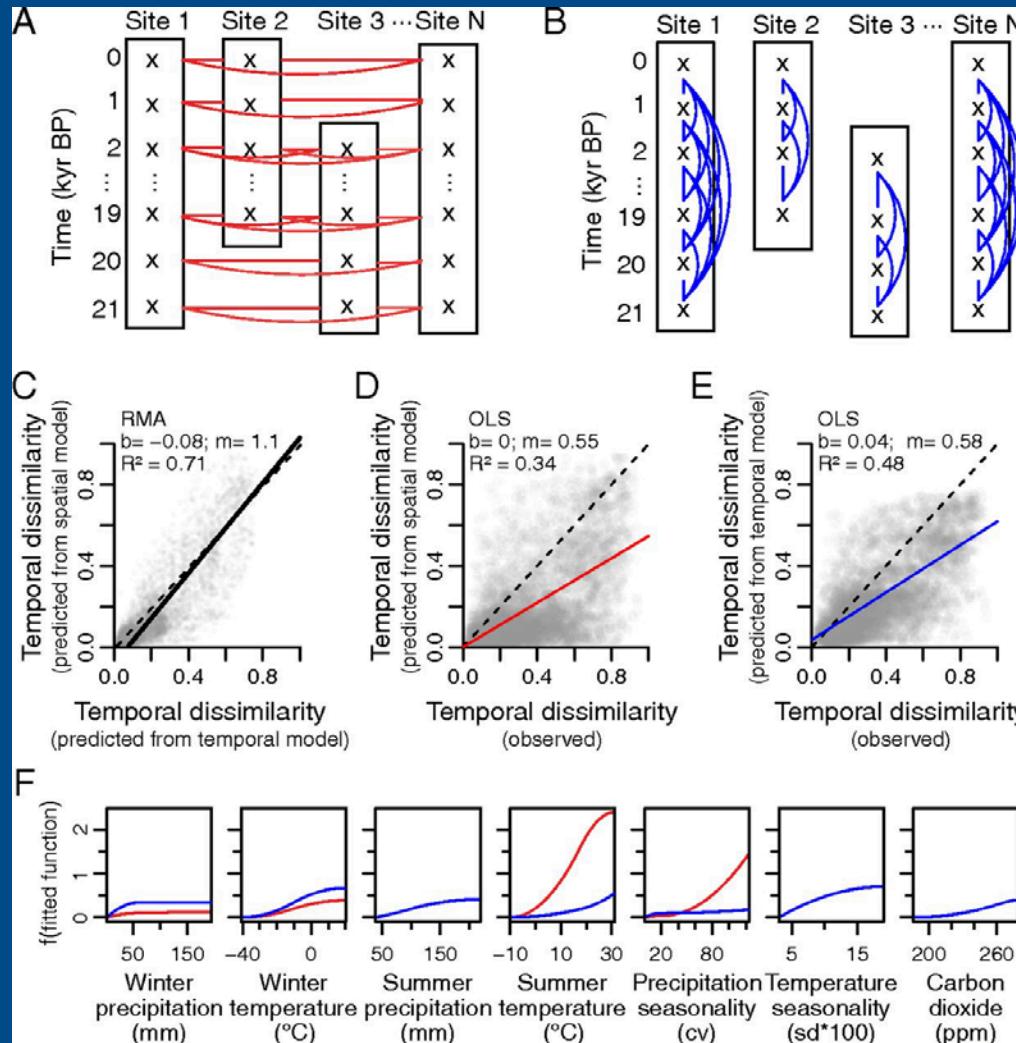
etc ...

Adapting modelling techniques to extract maximum value from best-available biological & environmental data

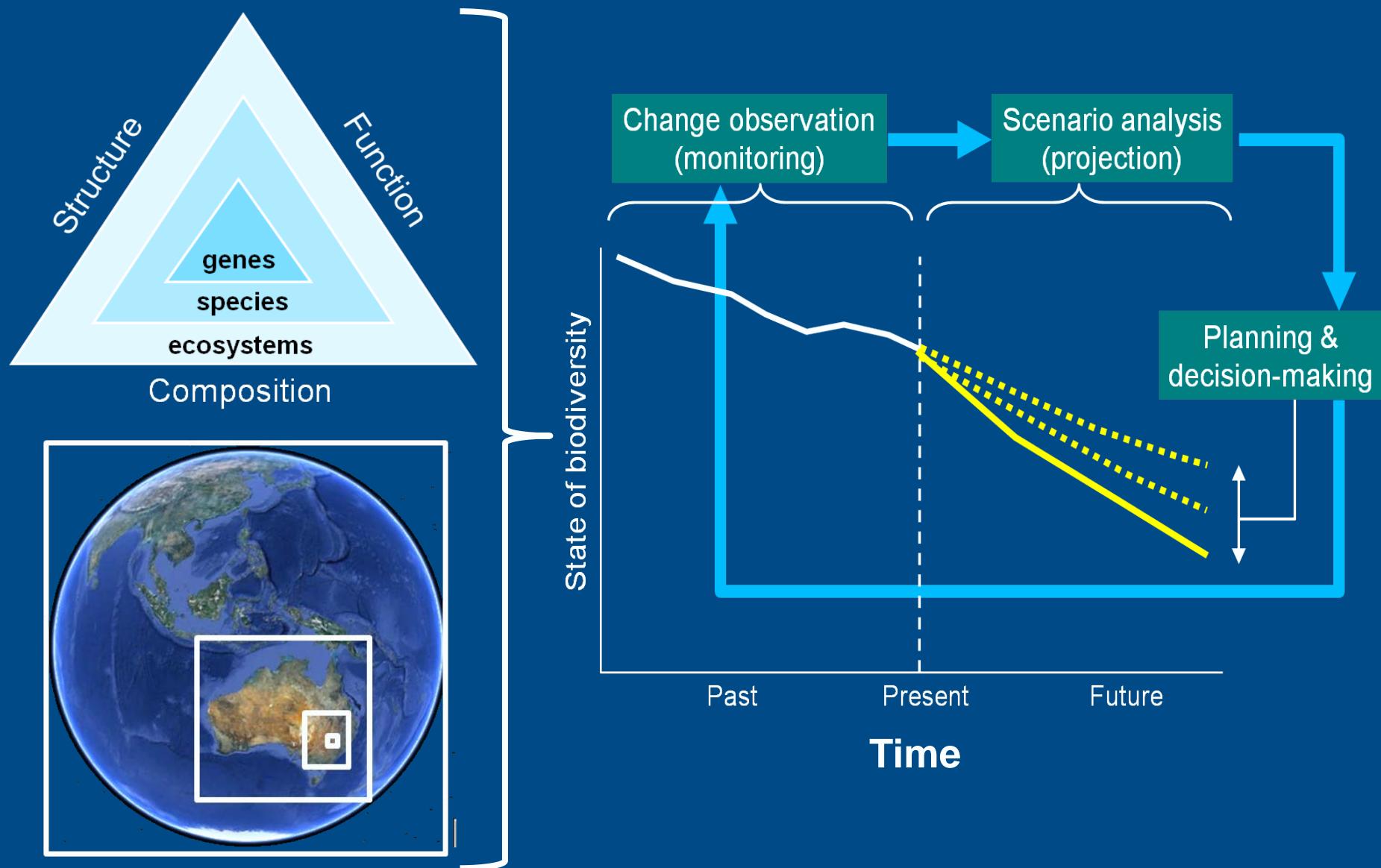
PNAS

Space can substitute for time in predicting climate-change effects on biodiversity

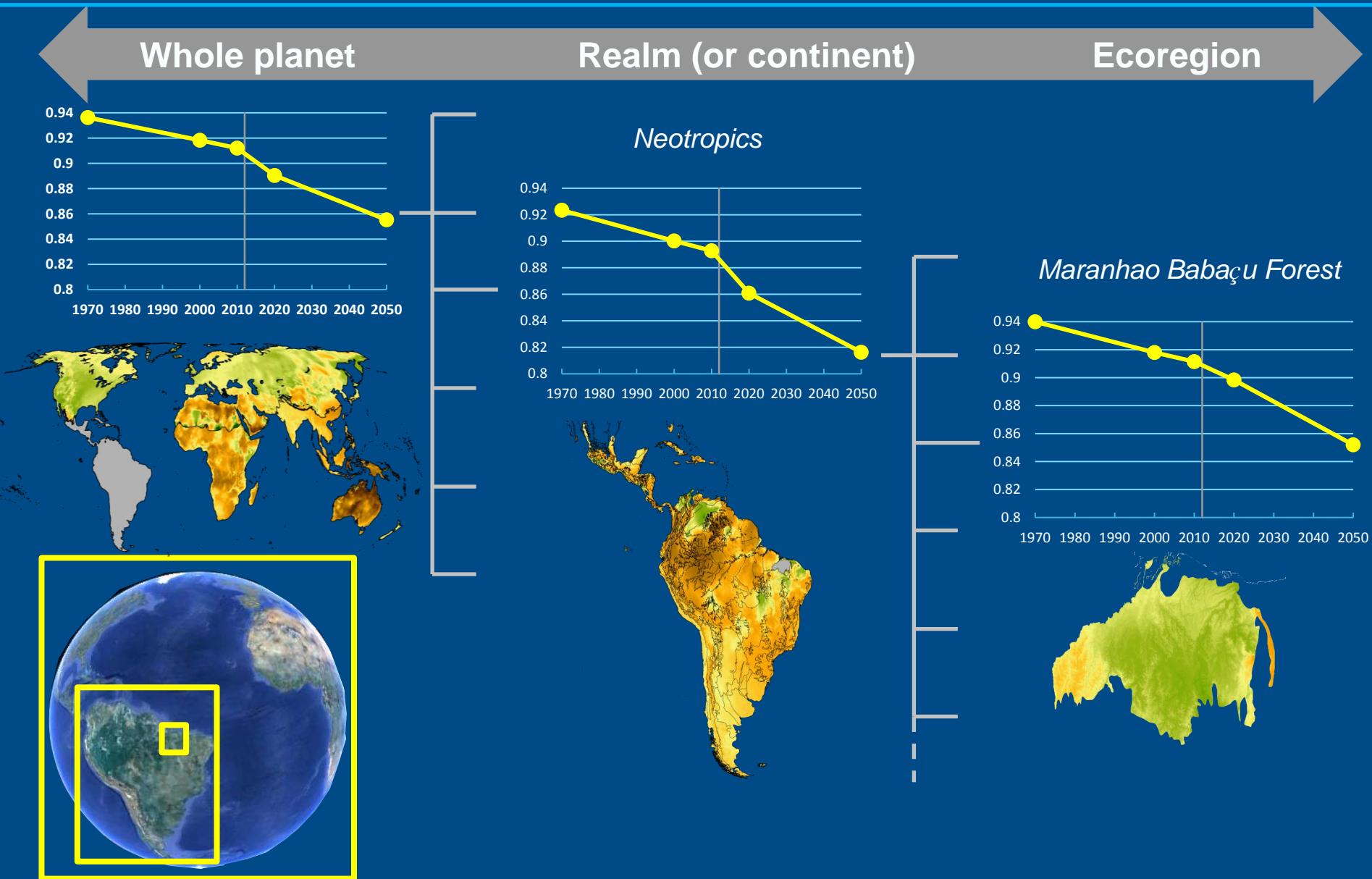
Jessica L. Blois^{a,1}, John W. Williams^a, Matthew C. Fitzpatrick^b, Stephen T. Jackson^c, and Simon Ferrier^d



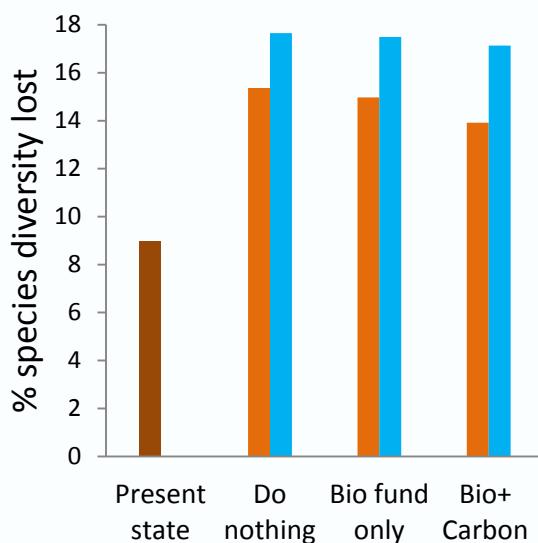
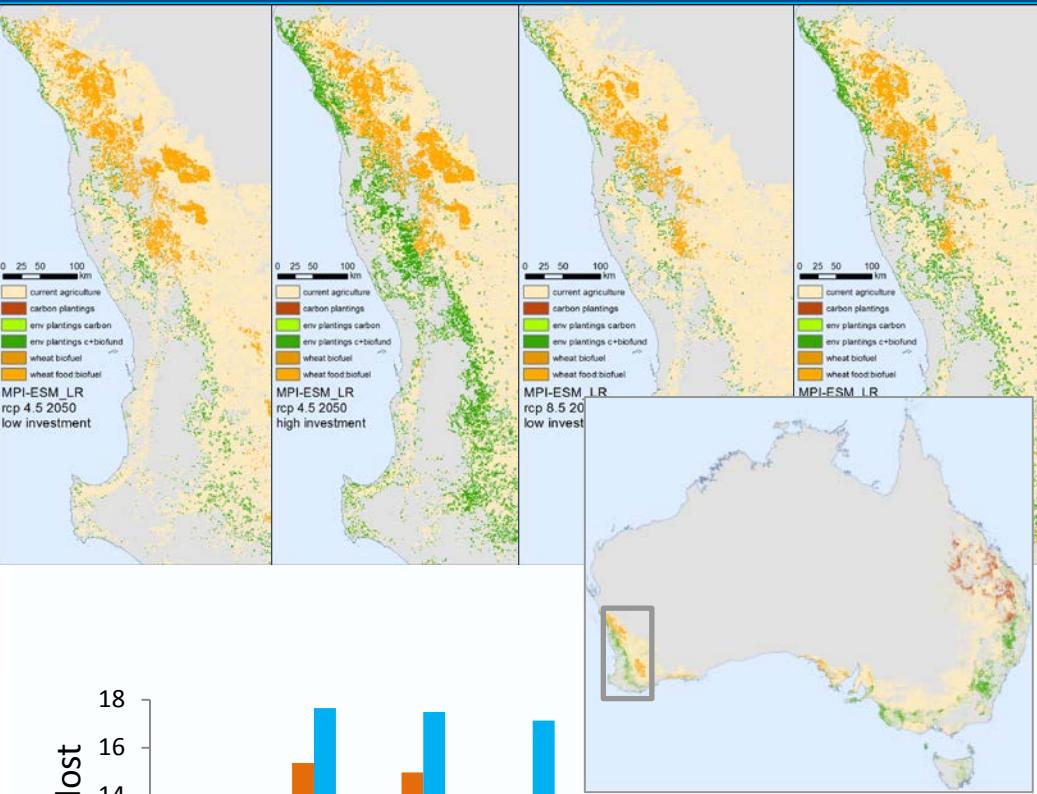
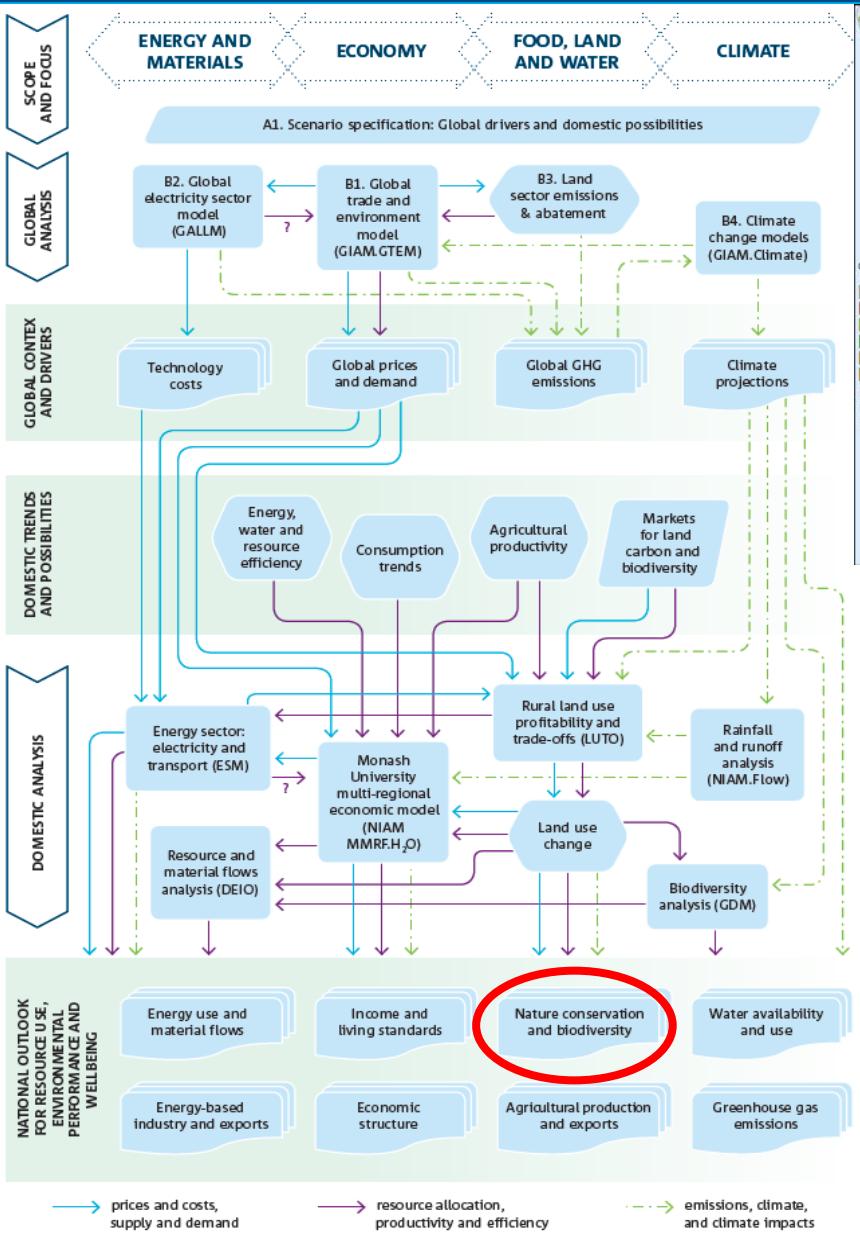
Challenges & opportunities in applying this approach to global biodiversity monitoring: broader integration



Effectively linking global biodiversity monitoring to future projection, scenario analysis and decision support

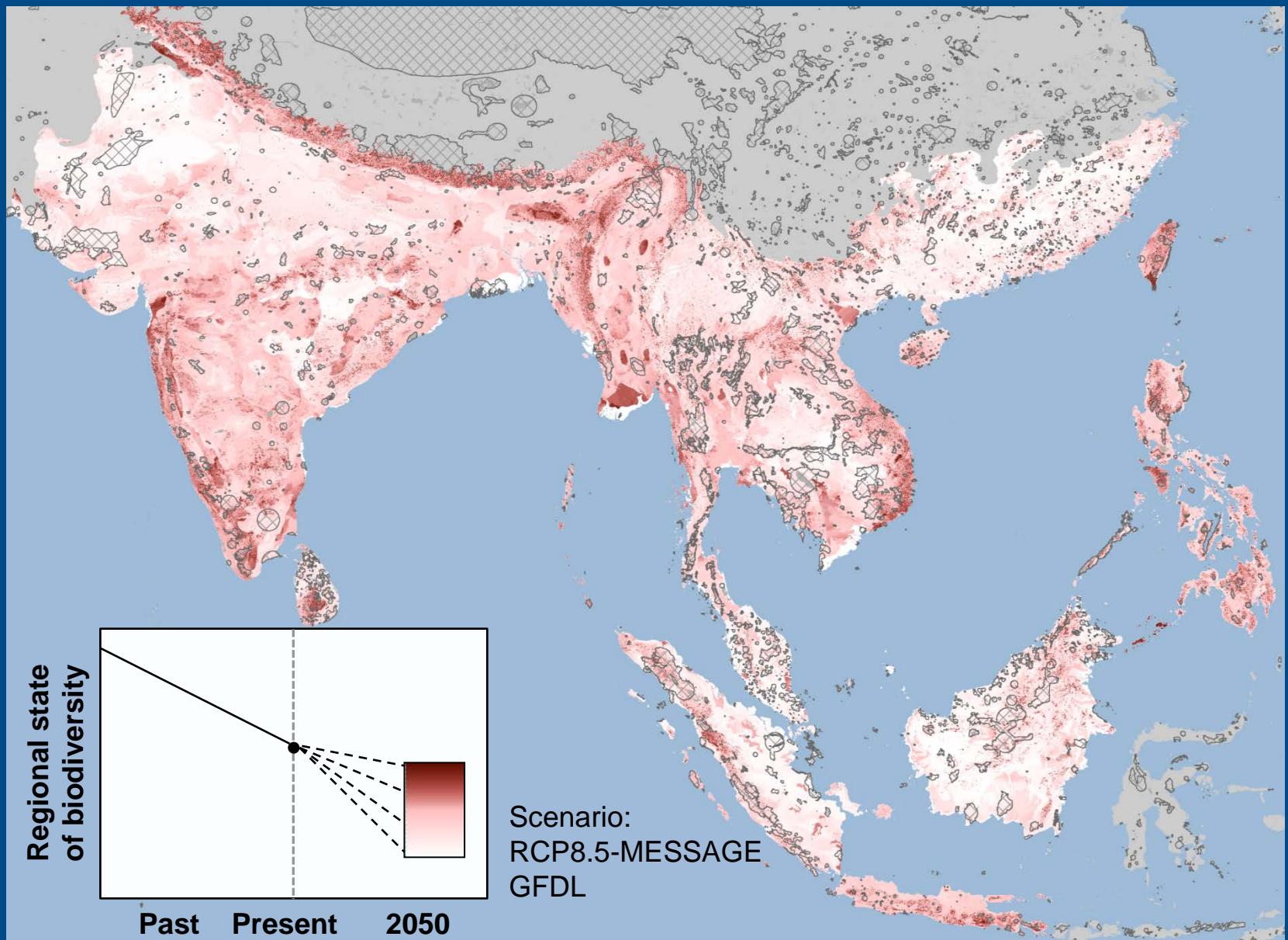


Effectively linking global biodiversity monitoring to future projection, scenario analysis and decision support



■ RCP45
■ RCP85

Effectively linking global biodiversity monitoring to future projection, scenario analysis and decision support



Effectively linking global biodiversity monitoring to future projection, scenario analysis and decision support

