



Opportunities, challenges and limitations of Species Distribution Models (SDM):

Implications for decision-making in support of
biodiversity conservation and sustainable land use

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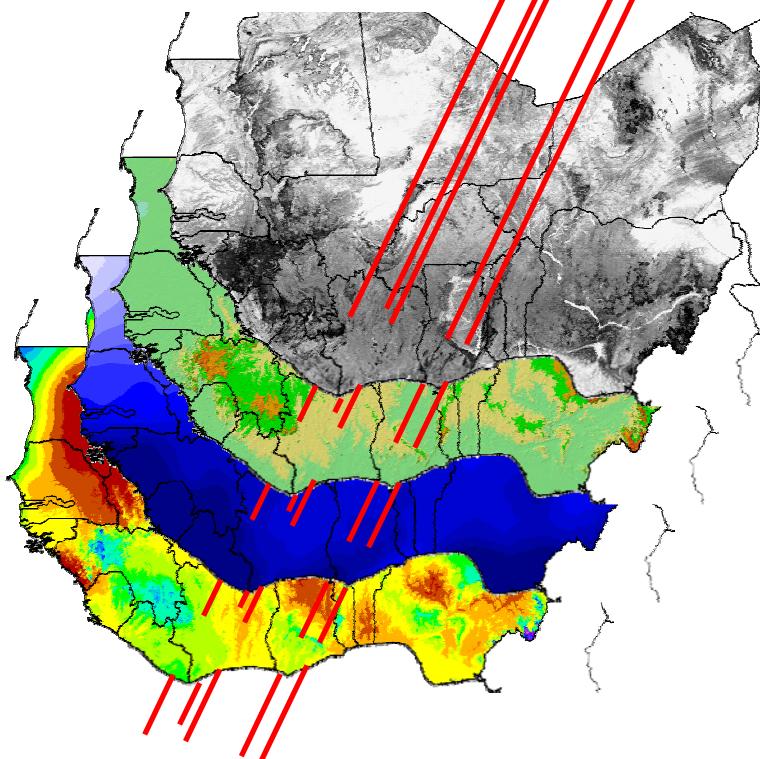
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German Remote Sensing Data Center (DFD), Germany

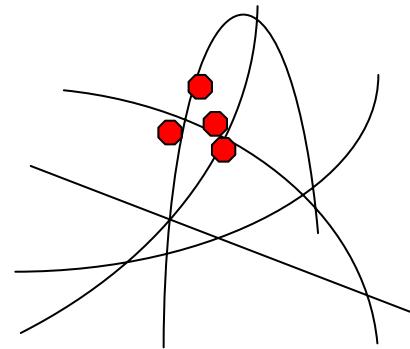
What is an SDM?

species localities

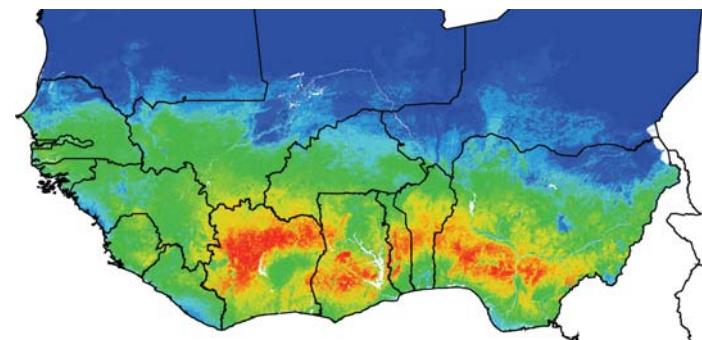
geographic space



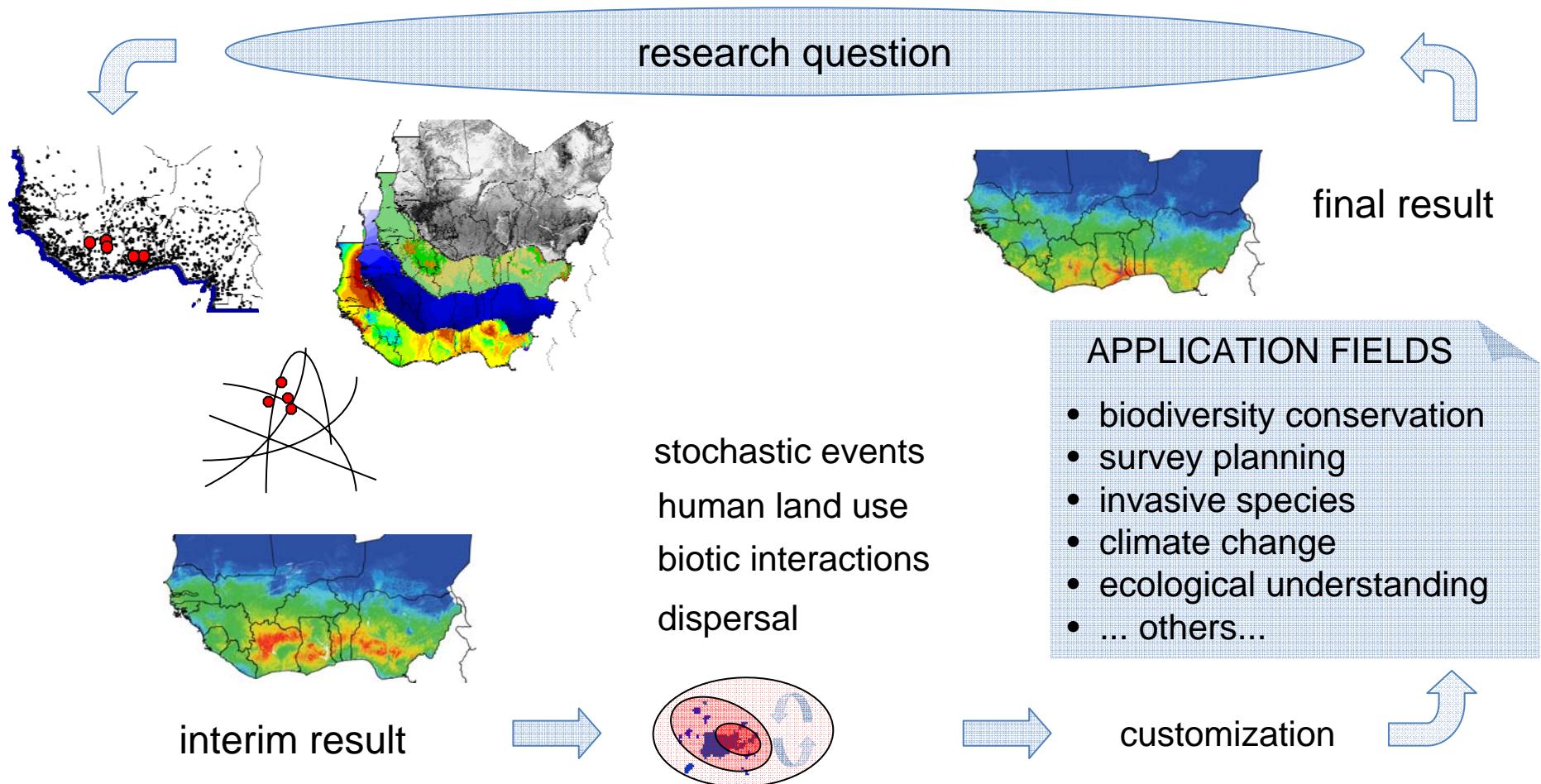
n-dimensional ecological space



prediction into geographic space



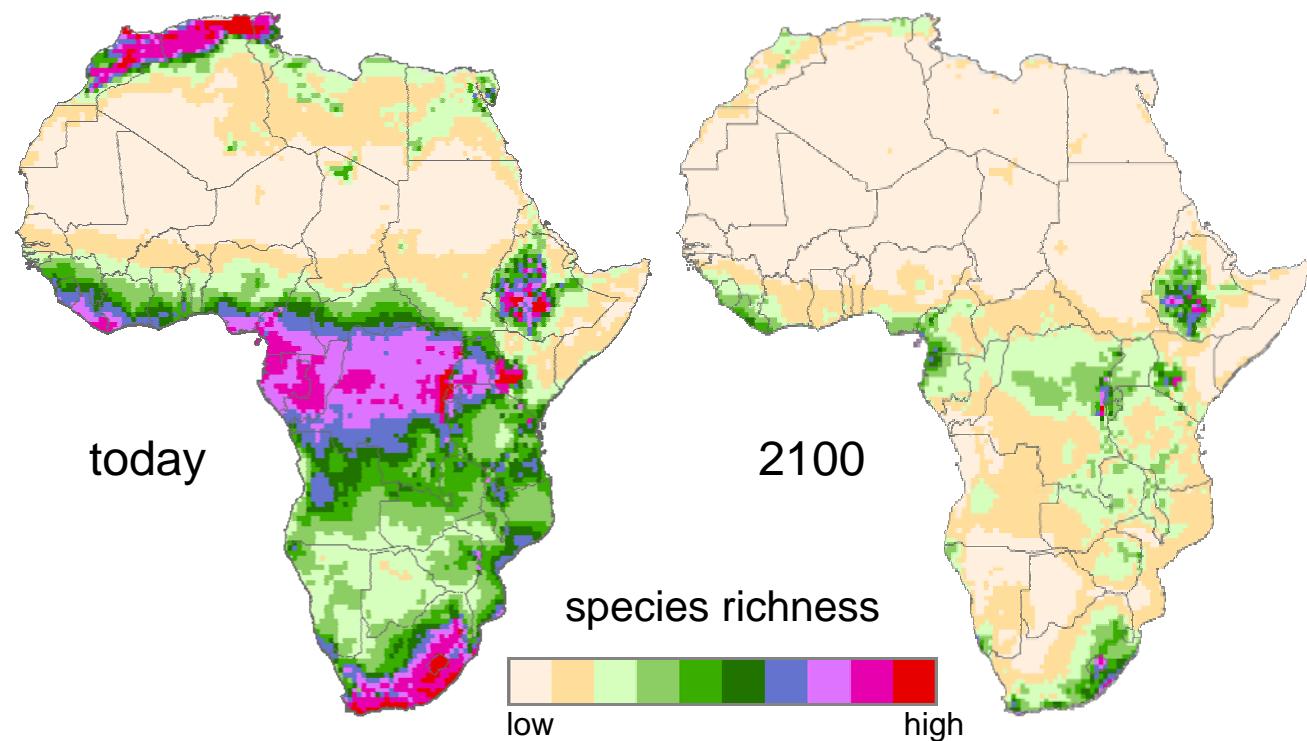
The complete workflow



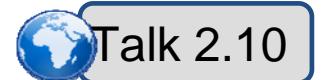
What are SDM good for?

Example 1: prepare for climate change

[BIOTA West / Sommer et al.]



HadCM3 climate model, IPCC A1FI Scenario, +4°C



What are SDM good for?

Example 2: identify biodiversity hotspots

also see  Talk 4.1

Species richness in West Africa at 1 km²

[BIOTA West; Penner et al.; Fahr et al.]

FIGURE REMOVED

FIGURE REMOVED

Amphibians

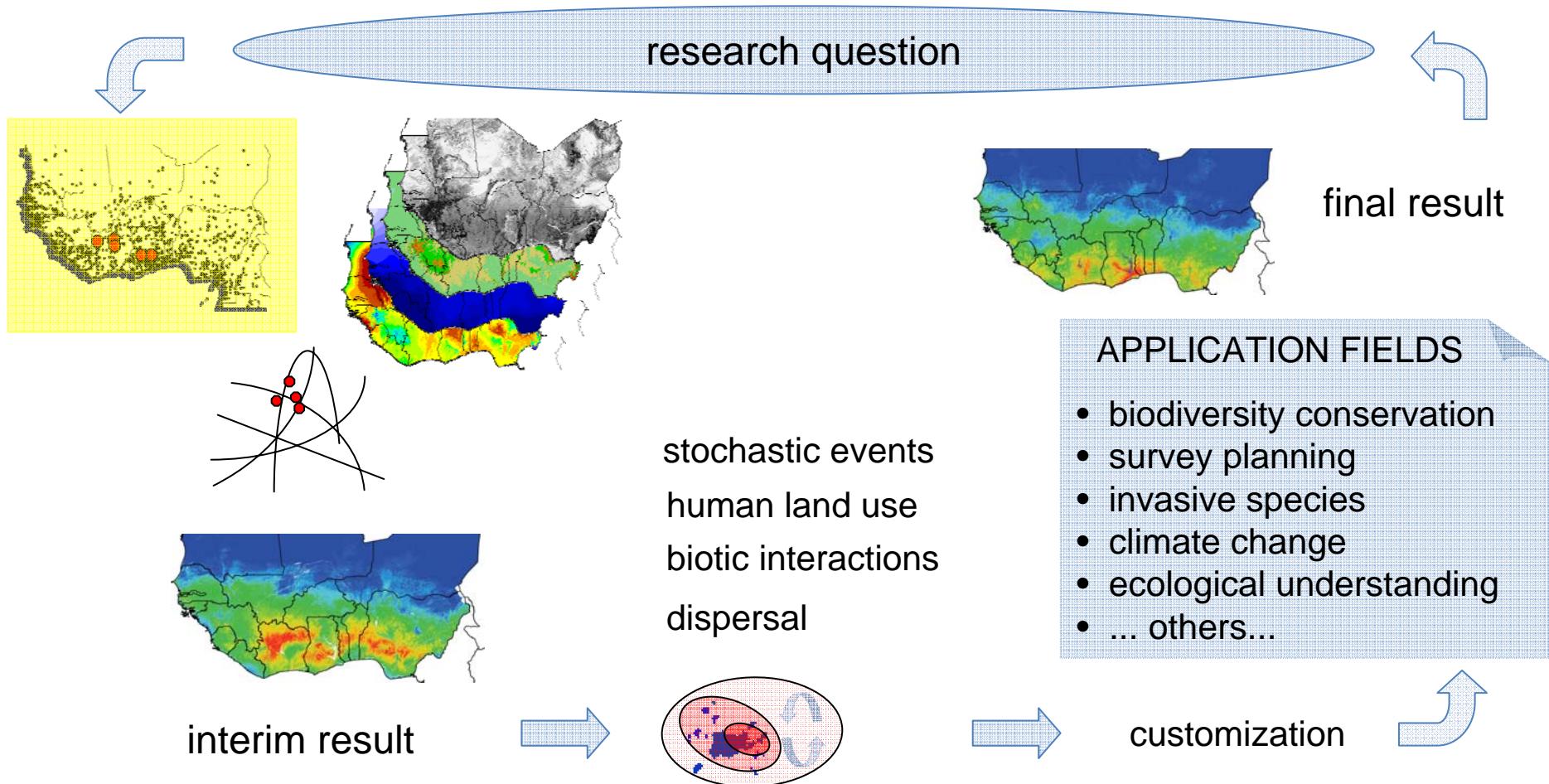
Bats

For another example (Conservation) see  Talk 3.2



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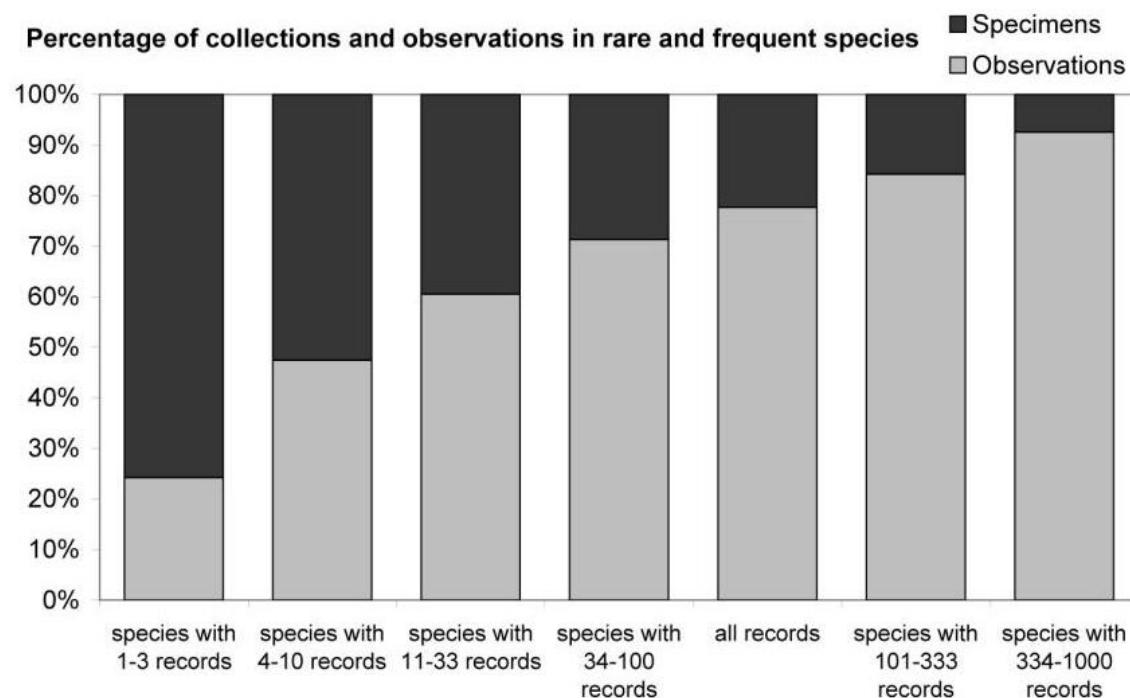
Selected challenges and solutions



Species occurrence data

all good?

Observations: an untapped resource for SDM?



[BIOTA West / Schmidt et al.]

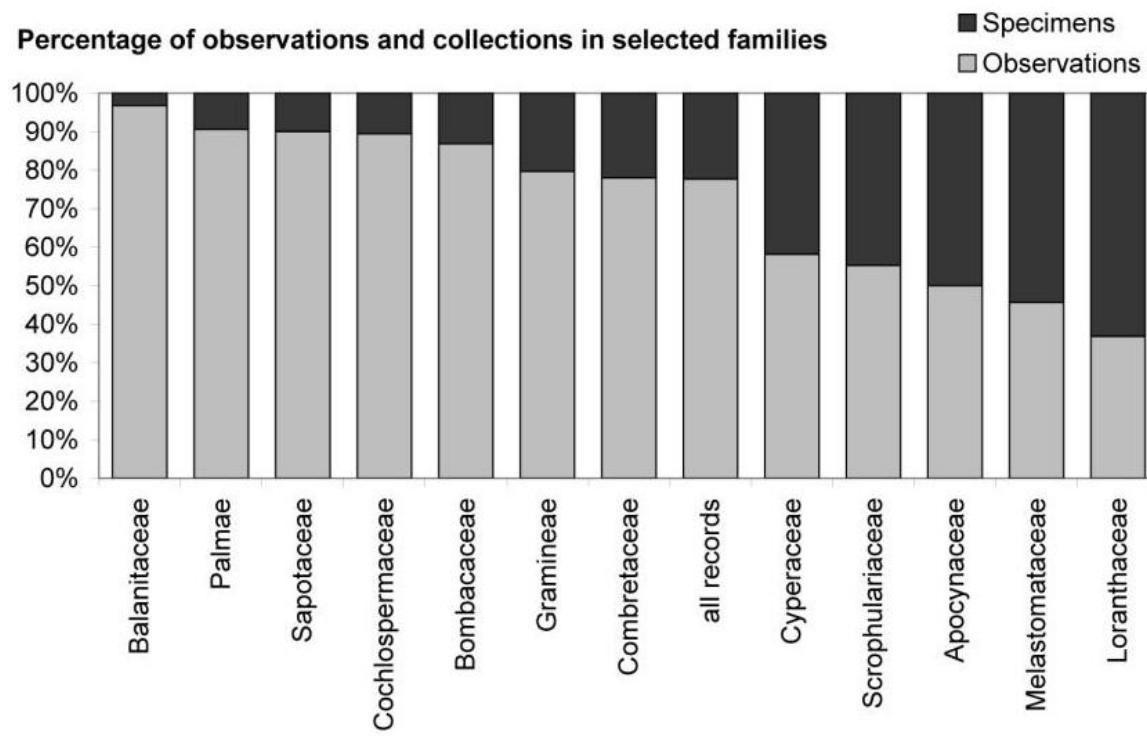
- Availability
- Sample Size
- Sampling Bias
 - Spatial autocorrelation
- Accuracy
 - geographic
 - taxonomic

supports validation
(and possibly calibration)

Species occurrence data

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Observations: an untapped resource for SDM?



[BIOTA West / Schmidt et al.]

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Effect of targeted field survey on predicted distribution

FIGURE REMOVED

- Availability
- Sample size
- Sampling Bias

→ Spatial
autocorrelation

- Accuracy
 - geographic
 - taxonomic

Species richness map of West African bats

BEFORE

predicted distribution changes
in under-sampled regions

[BIOTA West / Weber & Fahr]



Species occurrence data

all good?

Effect of targeted field survey on predicted distribution

FIGURE REMOVED

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- Sample size
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→ Spatial
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- Accuracy
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Species richness map of West African bats

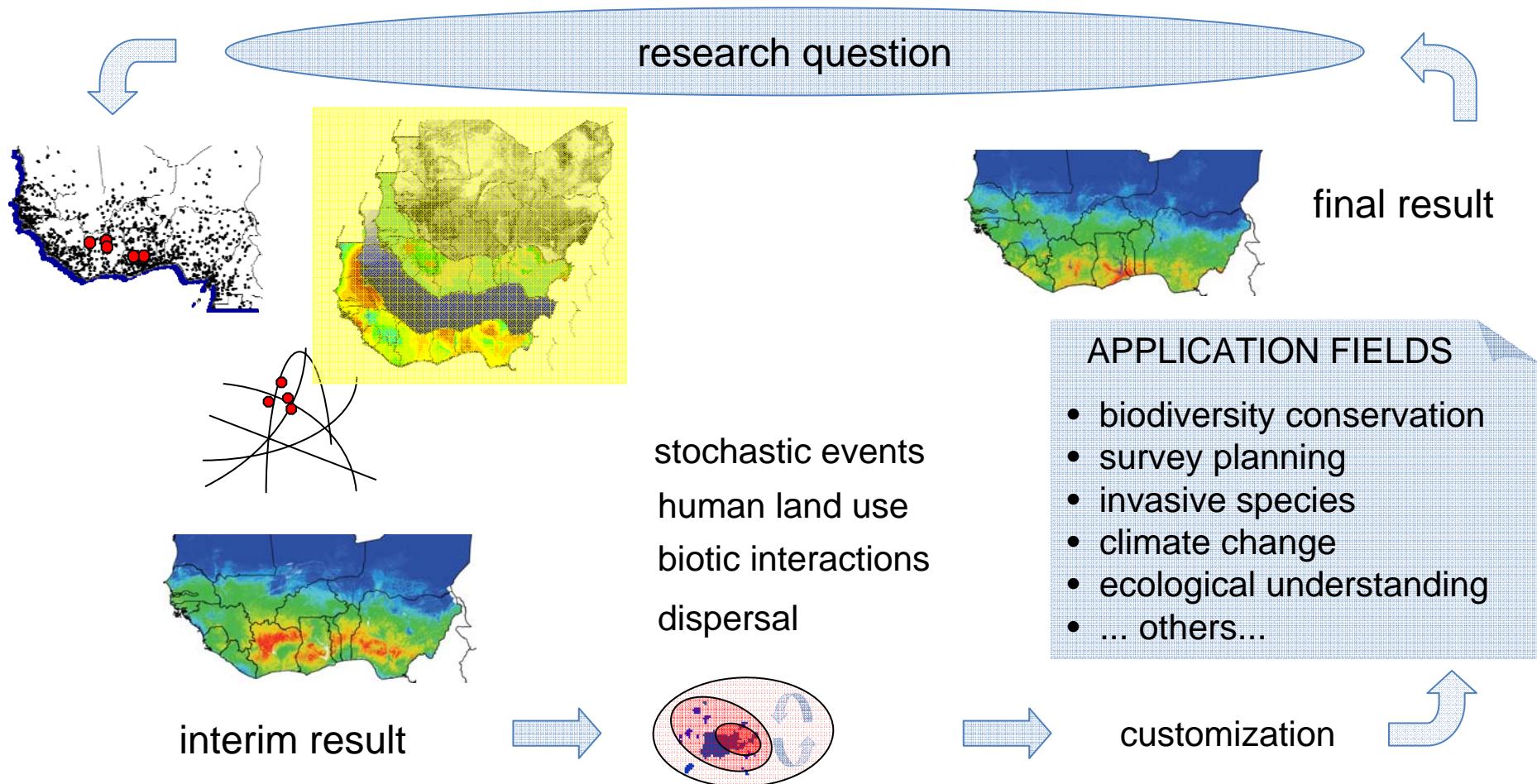
AFTER

predicted distribution changes
in under-sampled regions

[BIOTA West / Weber & Fahr]



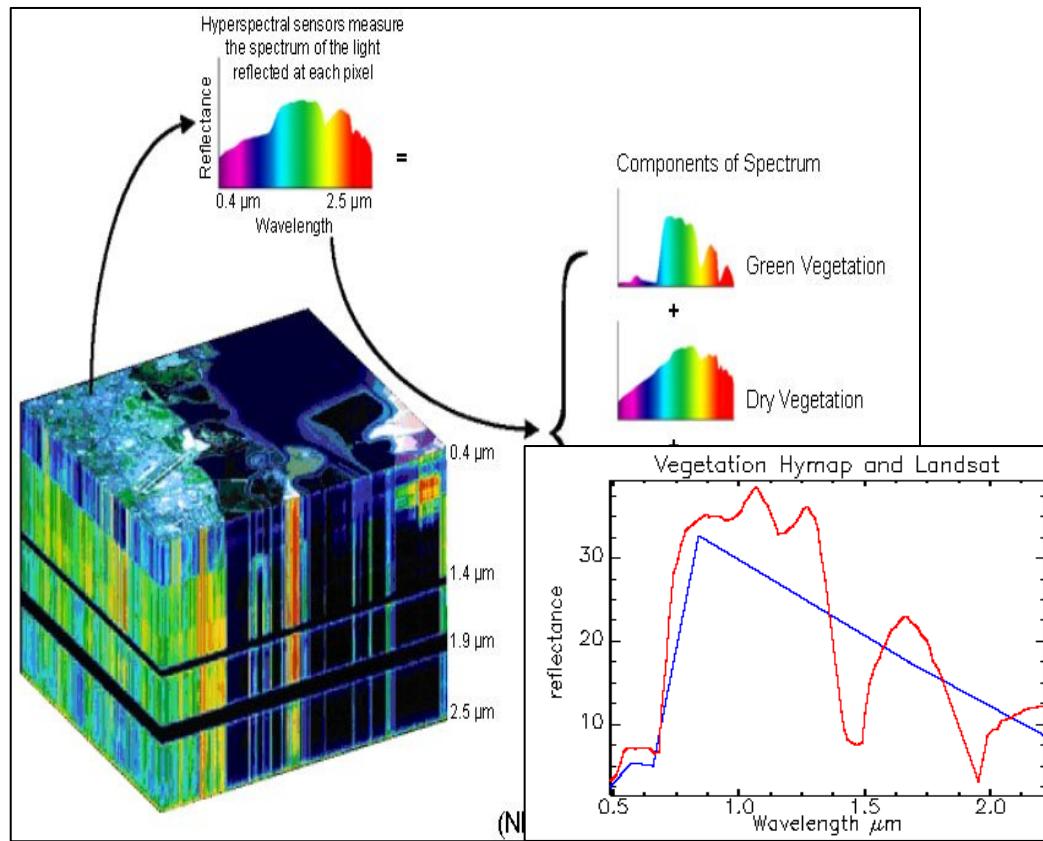
Selected challenges and solutions



Environmental variables

spoilt for choice

Seeing the world through the eyes of a ...

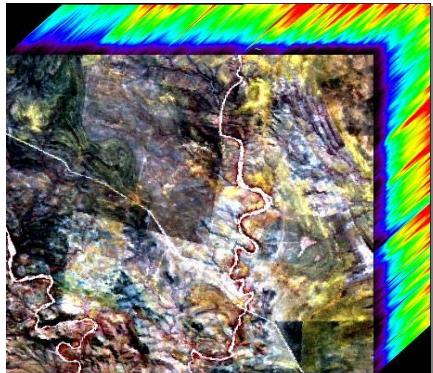


- Availability
 - finer spatial res.
 - time series
 - hyperspectral
- Selection
 - multicollinearity
- Transformation
- Consistency

Environmental variables

spoilt for choice

Seeing the world through the eyes of a ...



e.g. CHRIS-Proba satellite data
32 / 64 channels
32 m / 17 m spat. res.
however: swath width only 14 km

extract spectral profiles
of surface features and indices,
e.g. NDVI

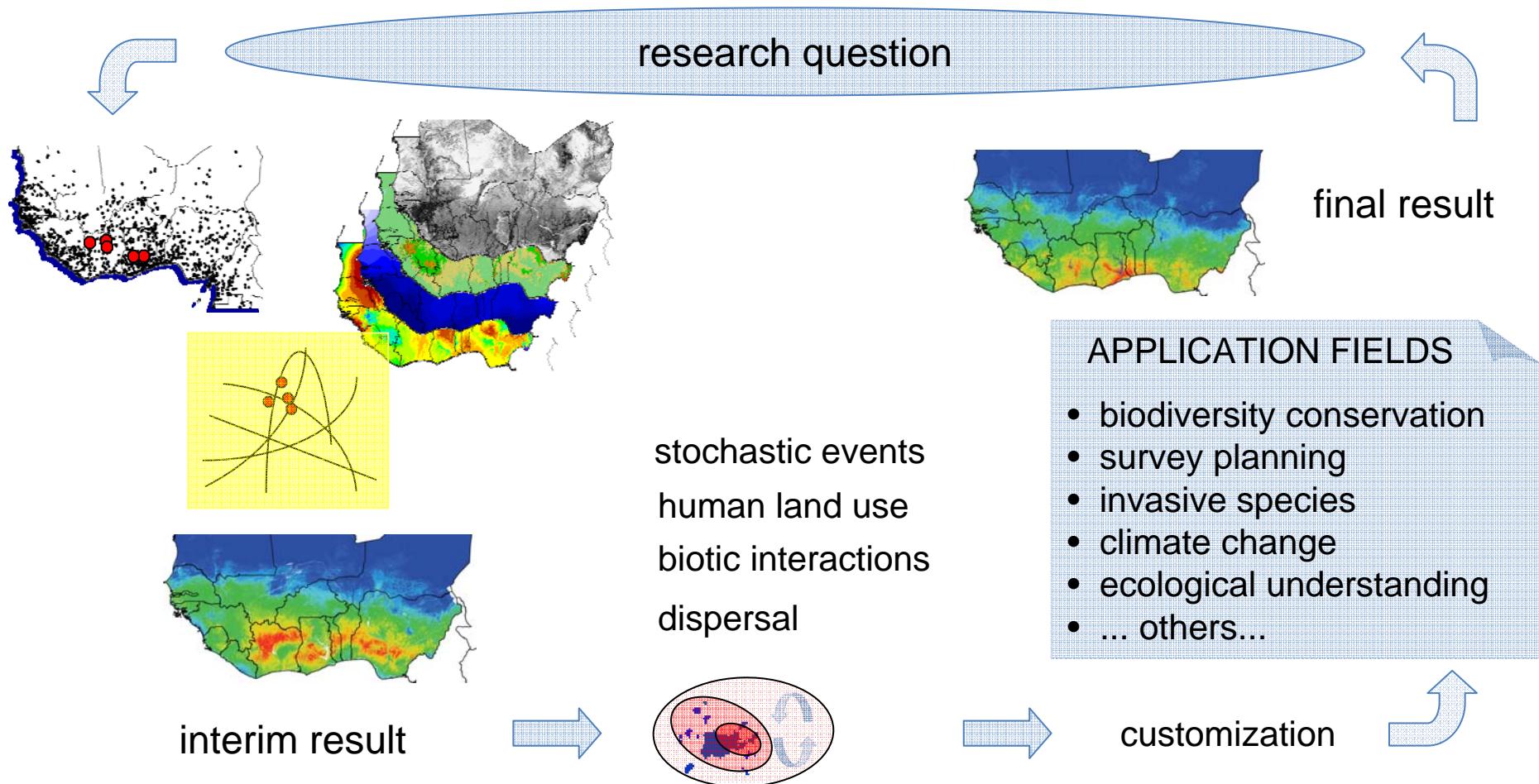
Inferring biodiversity from spectral heterogeneity
(spectral variation hypothesis) → direct use

Inferring biodiversity from aggregated SDM: create
new environmental variables from hyperspectral data
→ indirect use

- Availability
 - finer spatial res.
 - time series
 - hyperspectral
- Selection
 - multicollinearity
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[BIOTA Süd / Oldeland et al in prep.]

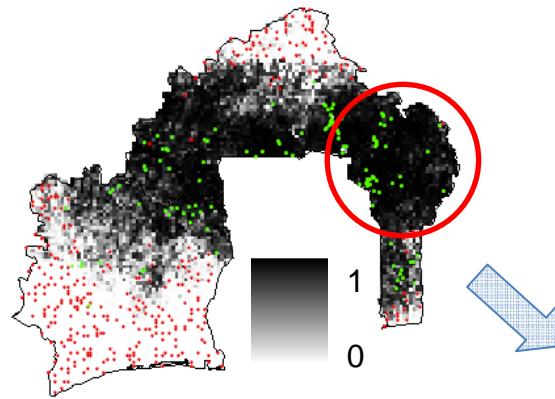
Selected challenges and solutions



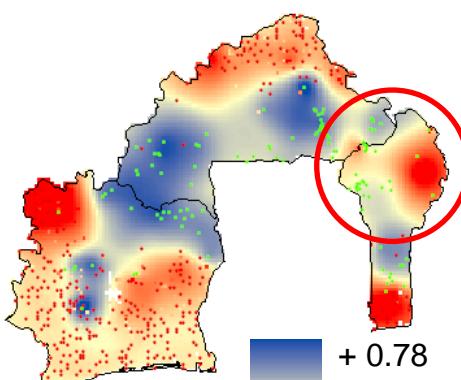
The statistical model

magic or math?

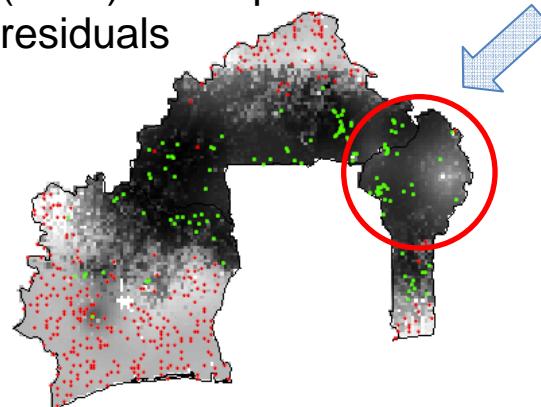
predicted probability (GLM)



Kriging interpolation
of residuals



SDM prediction
(GLM) + interpolated
residuals



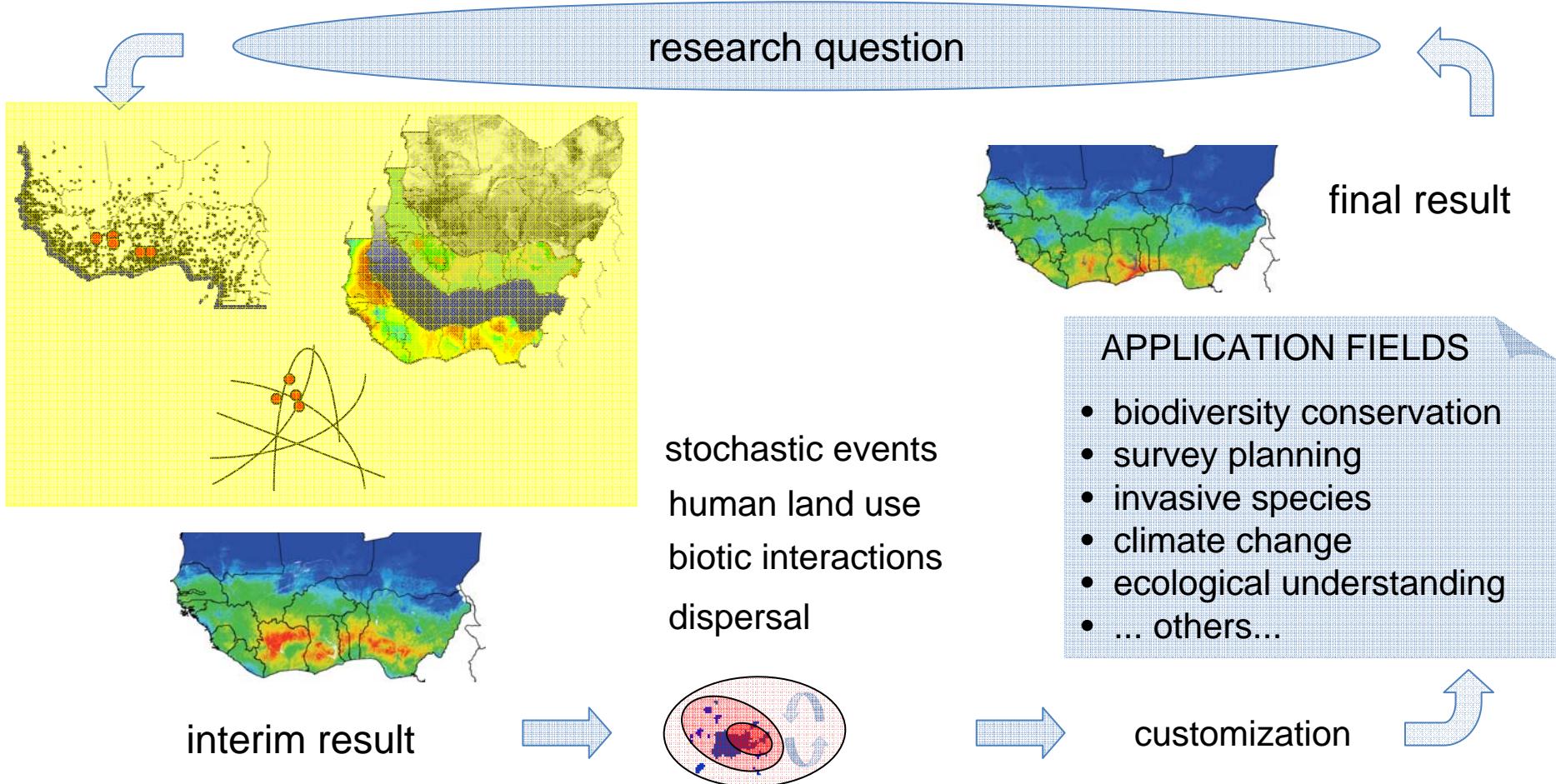
- Data assumptions
- Choosing model type
- Evaluating model quality
 - calibration / fit
 - discriminatory power
 - validation options

- Quantifying uncertainty

added value for decision-makers

[BIOTA West / García Marquez]

Selected challenges and solutions



An adequate framework

why bother?

FIGURE REMOVED

Species richness of bats; cell size ~ 1 km²

FIGURE REMOVED

Species richness of bats; cell size 0.5° (~ 3,000 km²)

- Spatial extent
- Spatial resolution
- Temporal extent
- Temporal resolution

Expert knowledge required to capture a species' ecological niche at relevant scale

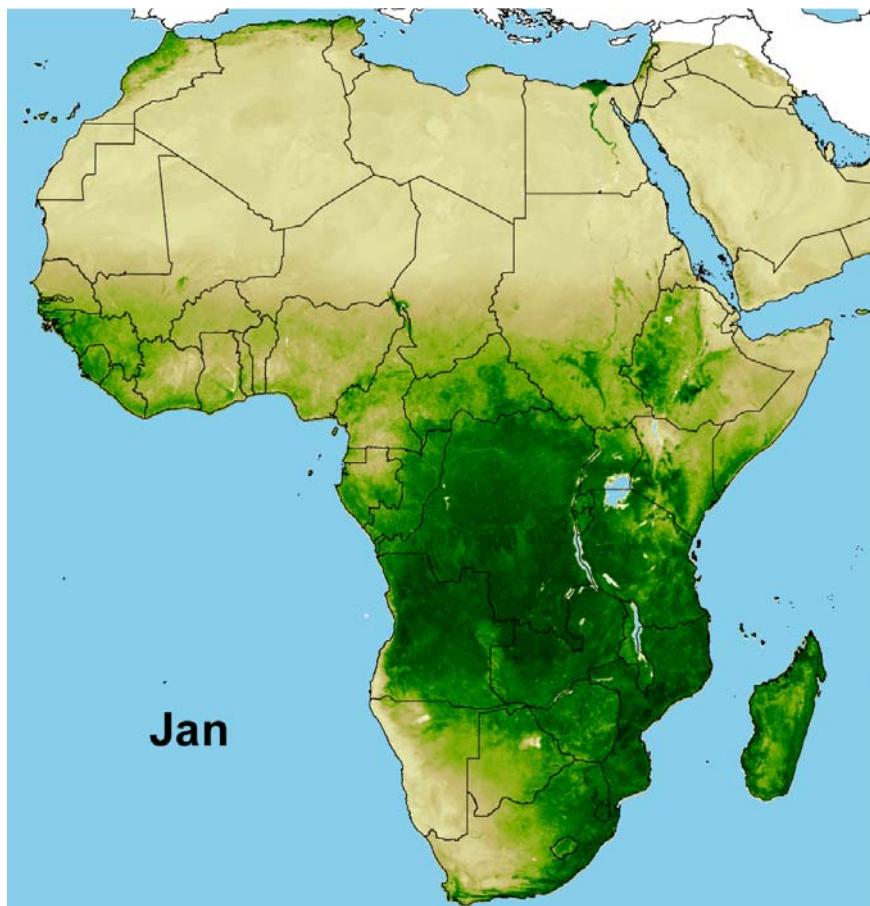
prevents decision-makers from using maps with false precision or over-prediction

[BIOTA West / Fahr et al.]



An adequate framework

why bother?



[ADDS]

- Spatial extent
- Spatial resolution
- Temporal extent
- Temporal resolution

Do mobile species respond to seasonal greening of vegetation?

(AVHRR NDVI, decadal, monthly averages 1981 – 2005)

An adequate framework

why bother?

FIGURE REMOVED

- Spatial extent
- Spatial resolution
- Temporal extent
- Temporal resolution

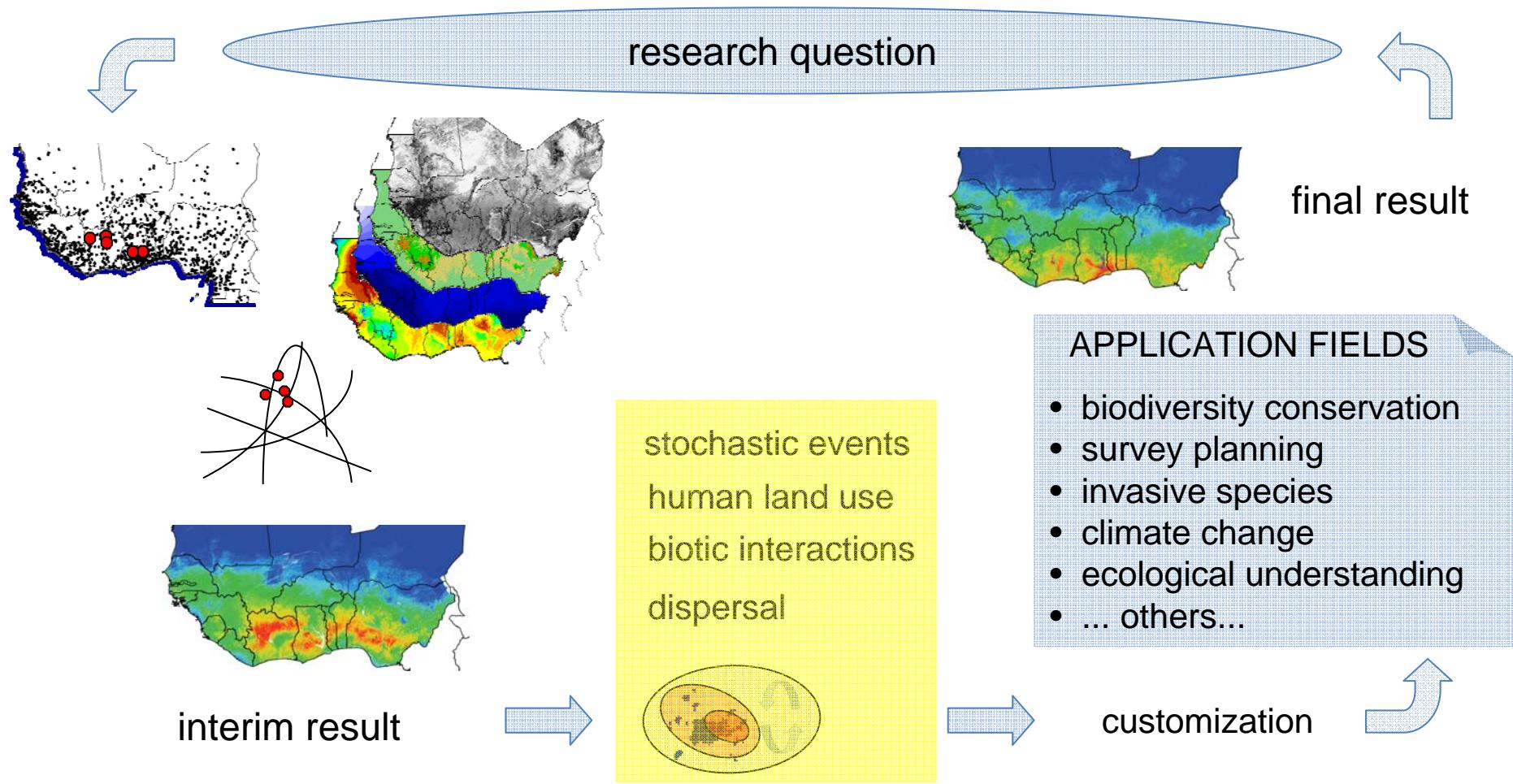
Monthly distribution models of *Eidolon helvum*; Maxent; 8 environmental variables; n = 537

SDM provide evidence that some mobile species depend on seasonally shifting resources → important for decision-makers

[BIOTA West, Fahr et al.]



Selected challenges and solutions

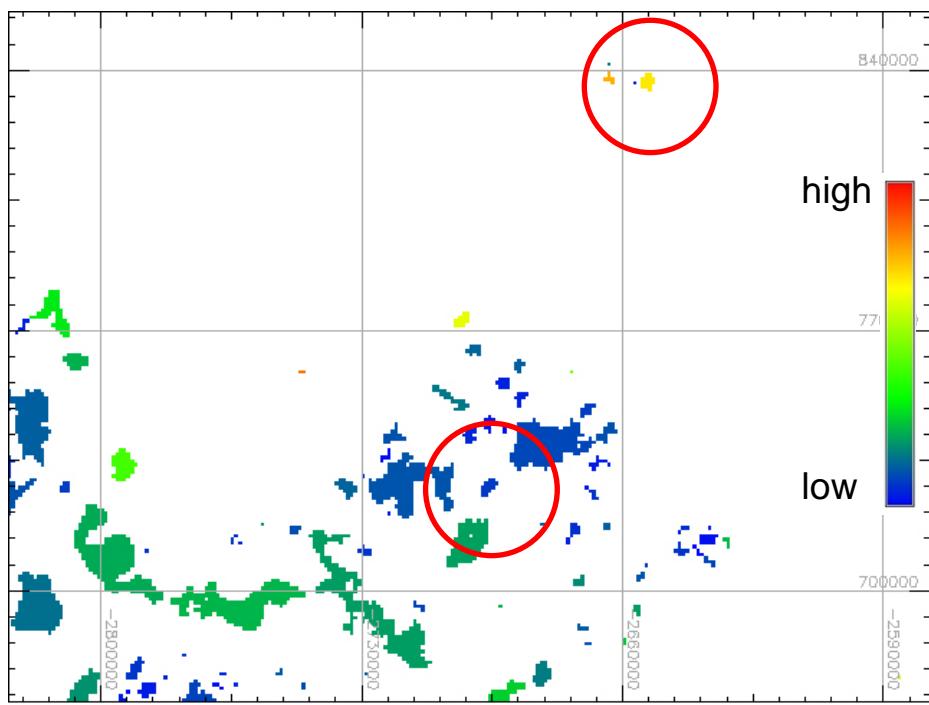


Enhancing results

let's get real



Predicted habitat patches evaluated with an „omnidirectional connectivity“ index

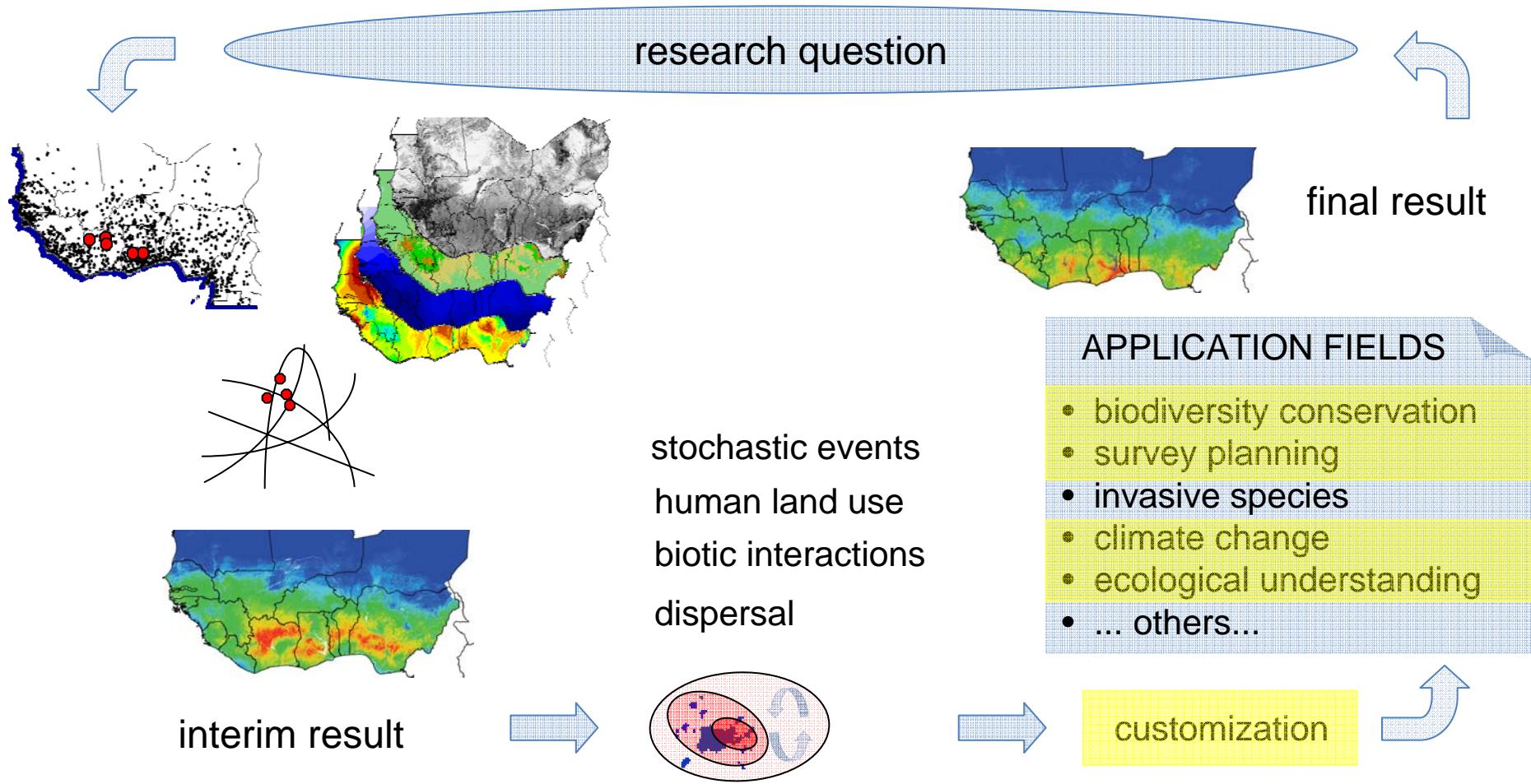


[BIOTA West, Wegmann]

- Dispersal
 - biogeography
 - metapopulation / habitat fragmentation
- Human land use
- Stochastic events
- Biotic interactions

ensures that predicted distribution maps exclude unsustainable patches

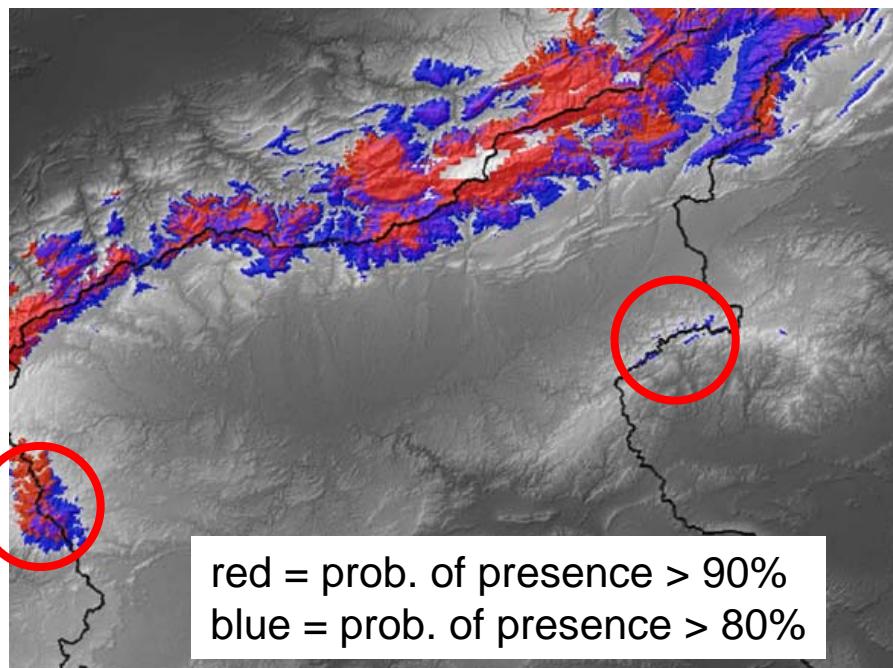
Selected challenges and solutions



The application field

menu du jour ou à la carte?

Change of predicted presence area of
Alyssum spinosum if temperature + 2.5° C



- Ecological understanding
- Direction of future surveys
- Climate change
- Invasive species
- Biodiversity conservation

Climate change SDM should be based on abiotic variables only, as long as biotic interaction of target species are unknown - as done in this BIOTA project.

12 Climatic Variables
ECHAM-REMO-LM-FOOT3DK
60 Geological / Soil Variabels
6 Topographical Variabels / DEM 30m

Variables actually used in tree construction:
ID367TMIN0 <= 6 °C
FROSTDAYS1 <= 140 days of frost
ID367TMAX0 <= 20.4 °C

[BIOTA-Morocco]

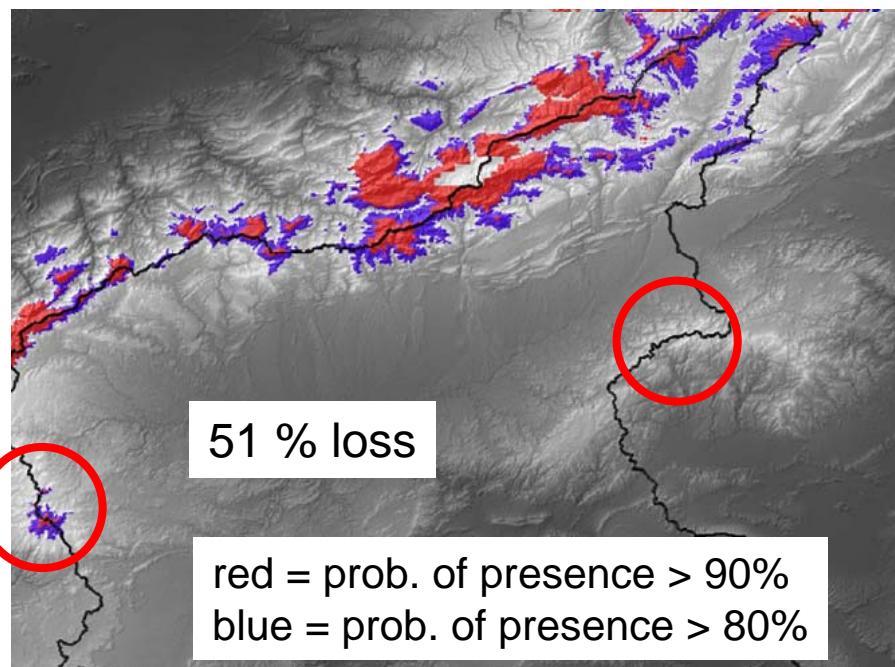


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Application Field

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Ordination analysis of SDM output (non-metric multidimensional scaling) for West African Red List Amphibians

SDM help understand drivers of species richness, ecosystem functions and services

[BIOTA West, Penner]





The application field

menu du jour ou à la carte?

Conversion of SDM output from continuous probability values to binary prediction map

- Ecological understanding
- Direction of future surveys
- Climate change
- Invasive species
- Biodiversity conservation

FIGURE REMOVED

- Presence *Rhinolophus maclaudi*
- Core habitat (predicted)
- Peripheral habitat (predicted)

Threshold setting must (and can) be done depending on application field

[BIOTA-West, Fahr et al.]

Conclusions & Implications

- SDM create important new knowledge
- SDM stand to benefit from rapidly improving data
- SDM can be exploited particularly well by researchers with biological expert knowledge
- Understanding of model reliability and data quality issues is critical
- Always consider intended application of SDM
- Decision-makers should (and can) be supplied with robust information on uncertainties of SDM

Outlook

- Need to improve integration of **biotic interactions** into SDM
- Improving „streaming“ data availability will enable SDM to support not only analysis but **also observation** tasks
- SDM would be a valuable contribution to any **early warning system** designed to detect threats to biodiversity (e.g. as in Scholes et al. 2008), especially if
 - a **dynamic link** was established between model predictions and remotely sensed land-cover change,
 - the user given the option to change the **scale** of analysis
 - if processing of interim results was done in response to the **management objective**



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