

# Comparative Retrospective Assessment of Biogeochemical Model Outputs for Fish and Foodweb Modelling in the Baltic Sea

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Workshop on Uncertainties of  
Scenario Simulations  
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# ECOSUPPORT



-will combine different models and outputs to enable modelling of entire Baltic foodweb

-to be used for scenario simulations of how Baltic Sea foodweb will respond to changes in forcings such as:

-climate

-nutrient loading (eutrophication, oligotrophication)

-fishing

-will present some ideas and concepts re. methodology and very preliminary results – work still in progress

# Foodweb-Fish related Workpackage

WP3: Impact on the foodweb (including fish – cod, herring, sprat)

3.1 Process validation of foodweb models

3.2 Scenario simulations of the food web

3.3 Quantification of uncertainty of future food web projections

# Scenario Simulations of Biology

Some considerations:

- the biology is dependent in various ways on the physics
  - e. g., rates proportional to temperature, salinity, nutrients,  $O_2$
- we have several oceanographic and NPZD models which will make projections of variables (e. g., temperature,  $O_2$ ) which affect fish biology

# Scenario Simulations of Biology

Some considerations:

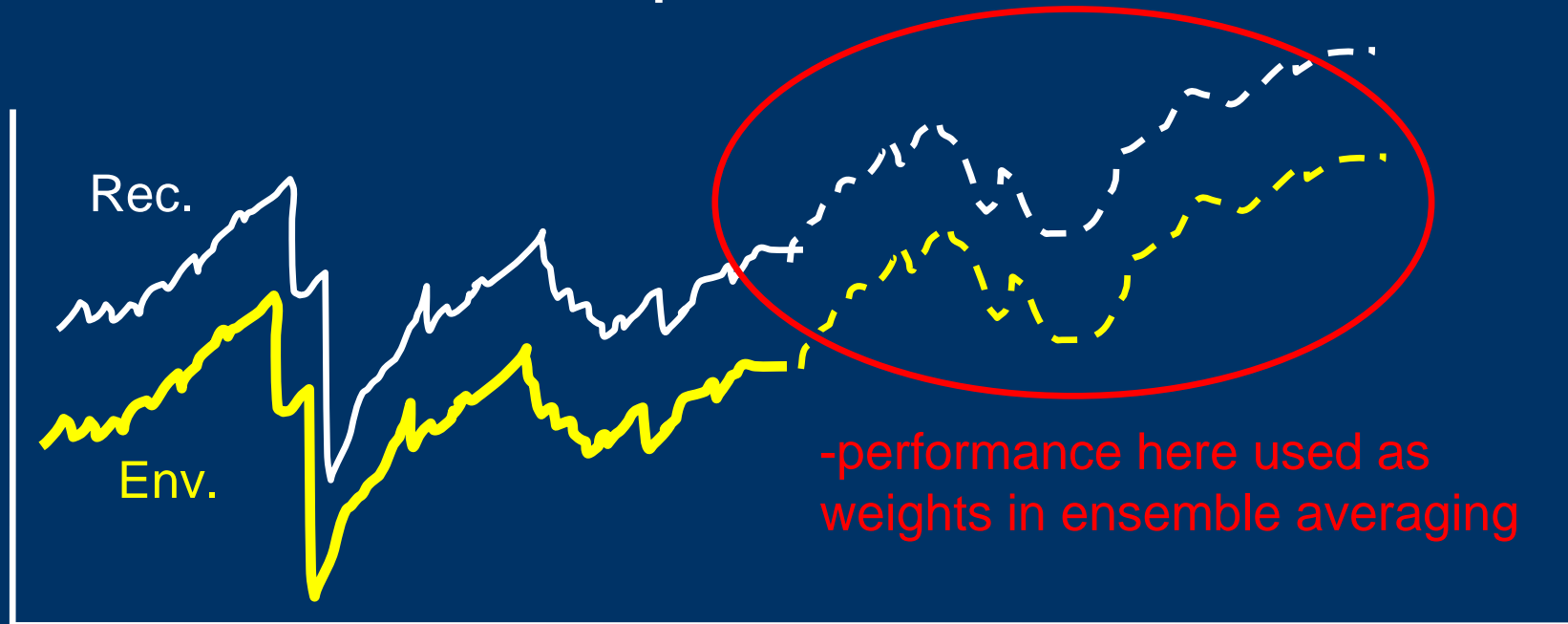
- the biology is dependent in various ways on the physics
  - e. g., rates proportional to temperature, salinity, nutrients,  $O_2$

- we have several oceanographic and NPZD models which will make projections of variables (e. g., temperature,  $O_2$ ) which affect fish biology

**-need to quantify how well the models reproduce *independent* “history” so we know how much confidence we can have in projections**

**-if models can't reproduce well the past, they likely won't give us good projections about the future.**

# Future Projections with Validated Models of the Independent Past



Fitting Period

Independent past data  
for performance testing

# Scenario Simulations of Biology

Model validation is important pre-requisite for future projections.

1. quantifies uncertainty of model outputs
2. can help choice of models to use, or how to combine model outputs, if there are *multiple* models available for the same response
  - e. g., whether and how to construct ensemble averages

# Combining Model Outputs

Within ECOSUPPORT, models are combined in two ways:

1. We have several models linked in sequential fashion for a given set of Scenarios or forcings (CO<sub>2</sub>, nutrients, fishing).

-how can those be combined?

2. We have several chains of linked models, for a given set of scenarios or forcings (CO<sub>2</sub>, nutrients, fishing).

-how can those be combined?



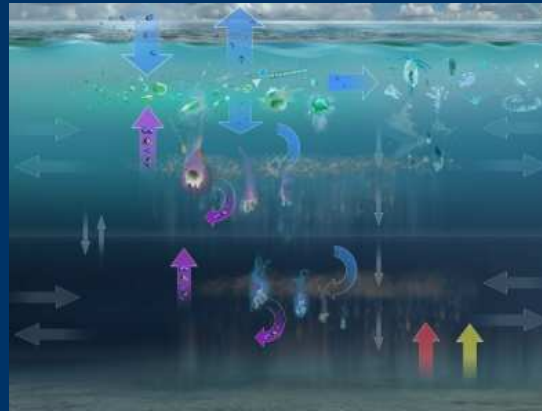
# Model Outputs Linked Sequentially

Atmos.-Ocean



T, S, ...

NPZD



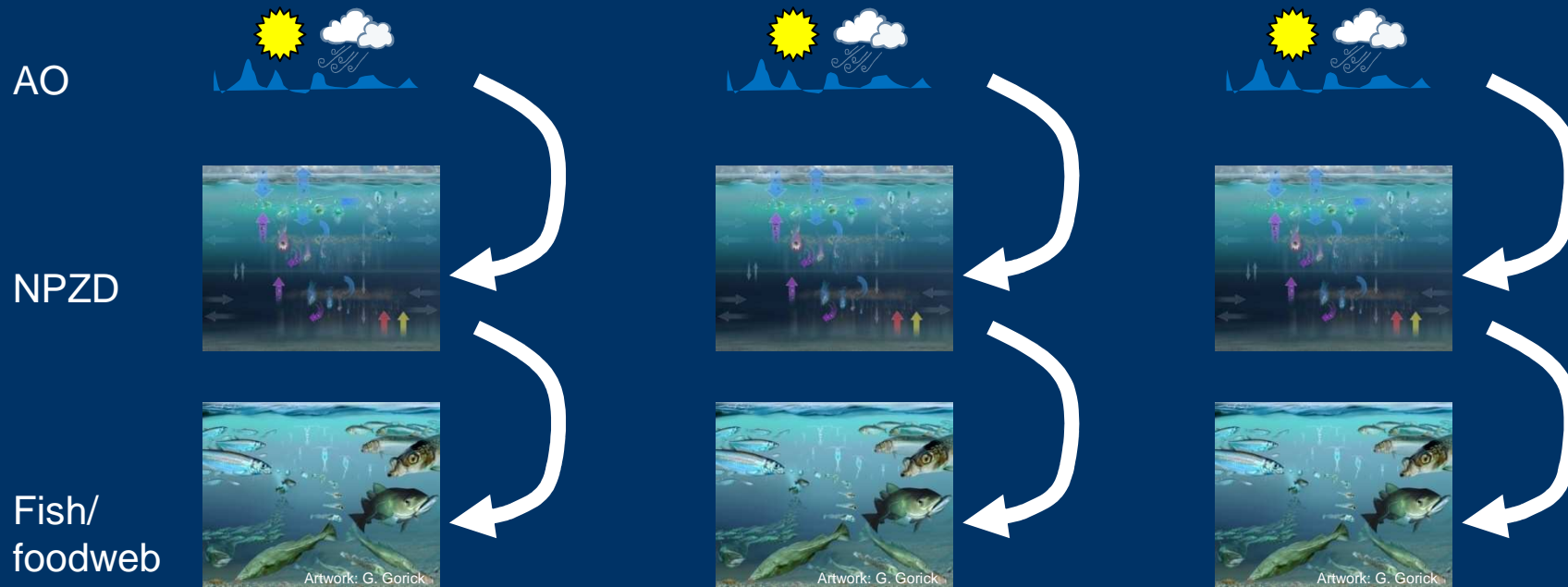
Habitat (RV),  
zooplankton, ...

Fish/foodweb



Artwork: G. Gorick

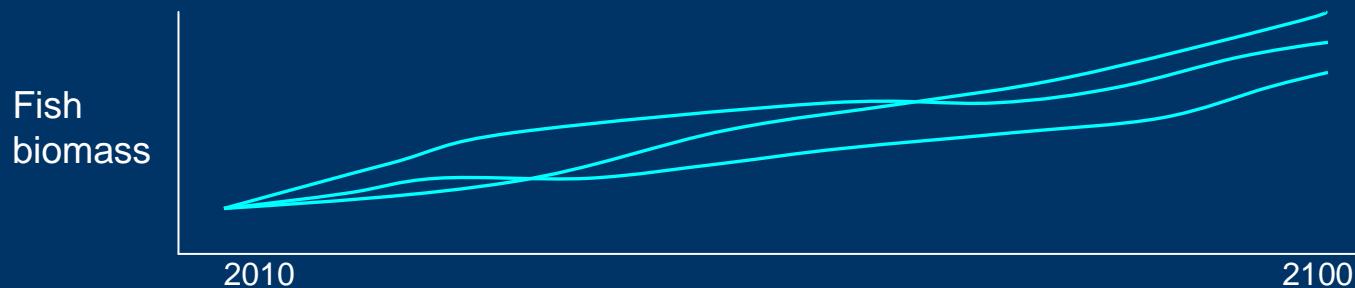
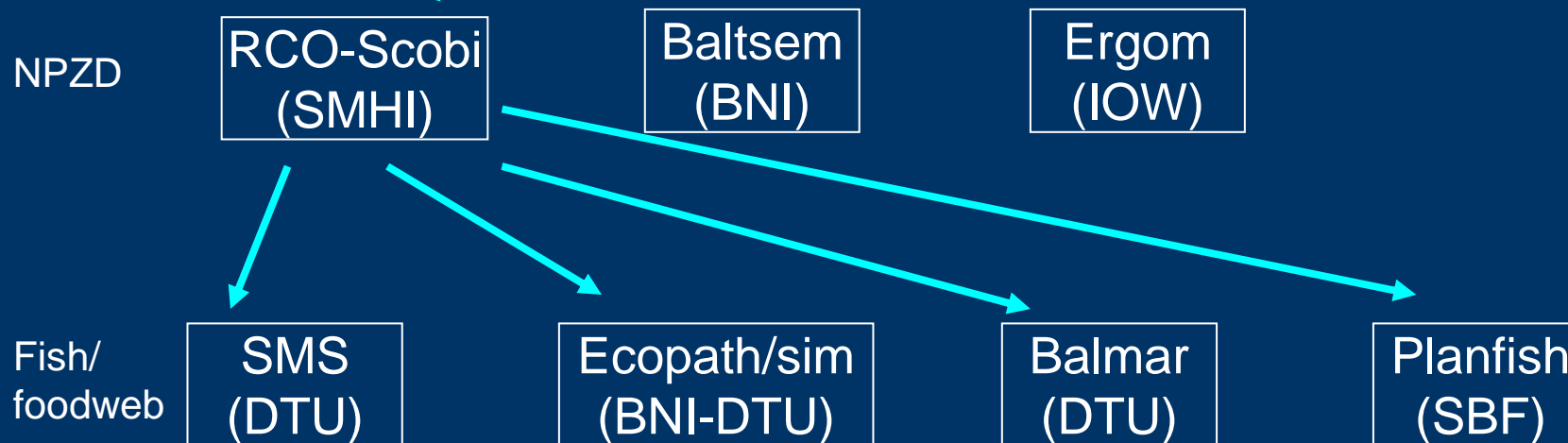
# Outputs Combined Across Models



# Inter-Linking Models within ECOSUPPORT

Climate models and CO<sub>2</sub> emissions  
for given nutrient scenario:

RCAO	
ECHAM5	
/A1B	



# Inter-Linking Models within Eco-Support

Climate models and CO<sub>2</sub> emissions

RCAO	
ECHAM5	
/A1B	

NPZD

RCO-Scobi  
(SMHI)

Baltsem  
(BNI)

Ergom  
(IOW)

Fish/  
foodweb

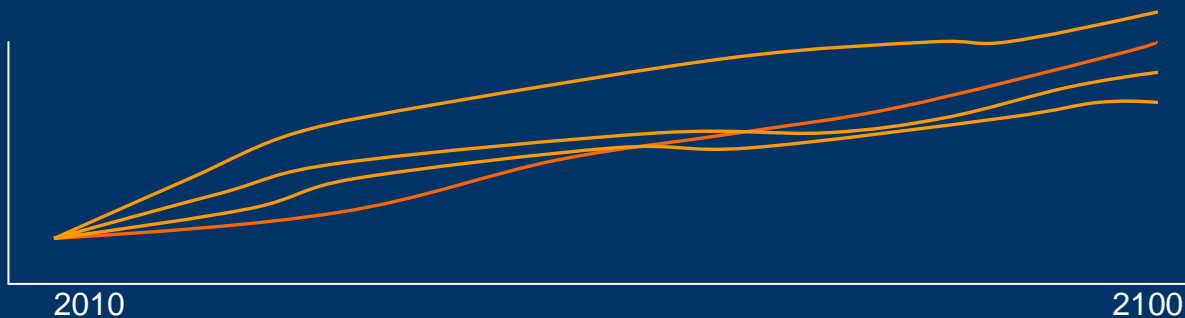
SMS  
(DTU)

Ecopath/sim  
(BNI-DTU)

Balmar  
(DTU)

Planfish  
(SBF)

Fish  
biomass



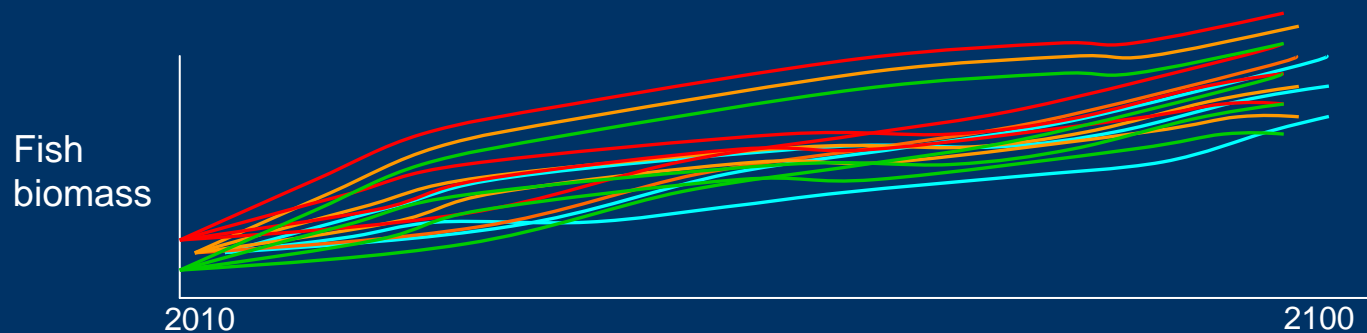
# Dealing with Uncertainties

Will use approach of “ensemble averaging” across model outputs  
-same approach applied in climatology, IPCC, etc.

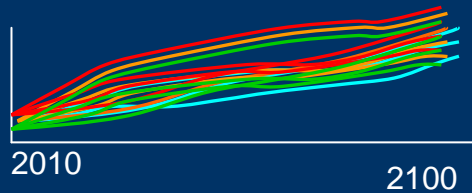
-calculate average and uncertainty (variability) for same set of forcings  
but with different models

-for a given CO<sub>2</sub>, nutrient and fishing scenario, have following time series:

2 climate models x 3 NPZD models x 4 foodweb/fish models = 24 time series



# Questions



How should model outputs be combined into 1 time series?

Some options:

1. simple unweighted average across models
2. account for past performance :  
-select only the model with best fit to observed data, or
3. weigh outputs according to past performance with observed data, and calculate weighted average

-we will attempt option 3.

# Case Studies for Learning and Methodological Development

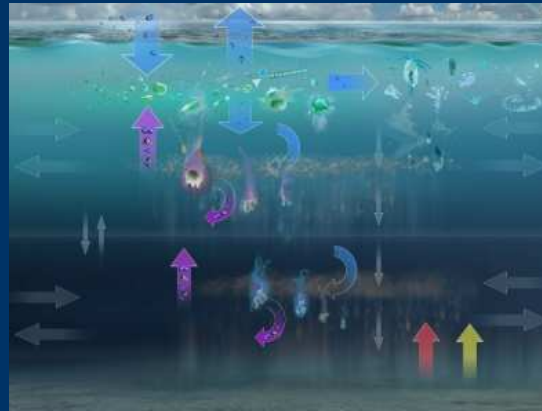
- temperature (e.g ., for sprat recruitment)
- cod reproductive volume

# Model Outputs Linked Sequentially

Atmos.-Ocean



NPZD



Fish/foodweb



Artwork: G. Gorick

Temperature

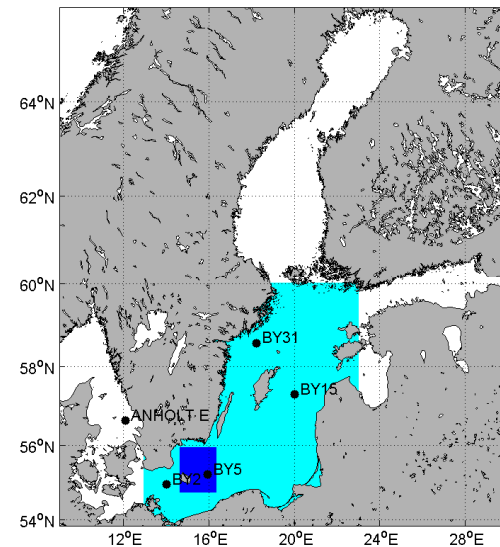
Temperature effects on sprat recruitment



# Modelled Data Available

Monthly temperature data near halocline (45-65 m) for Bornholm Basin, 1970-2007 from SCOBI and ERGOM

Monthly temperature data in surface (0-10 m) for Baltic Proper, 1970-2007, from SCOBI and ERGOM

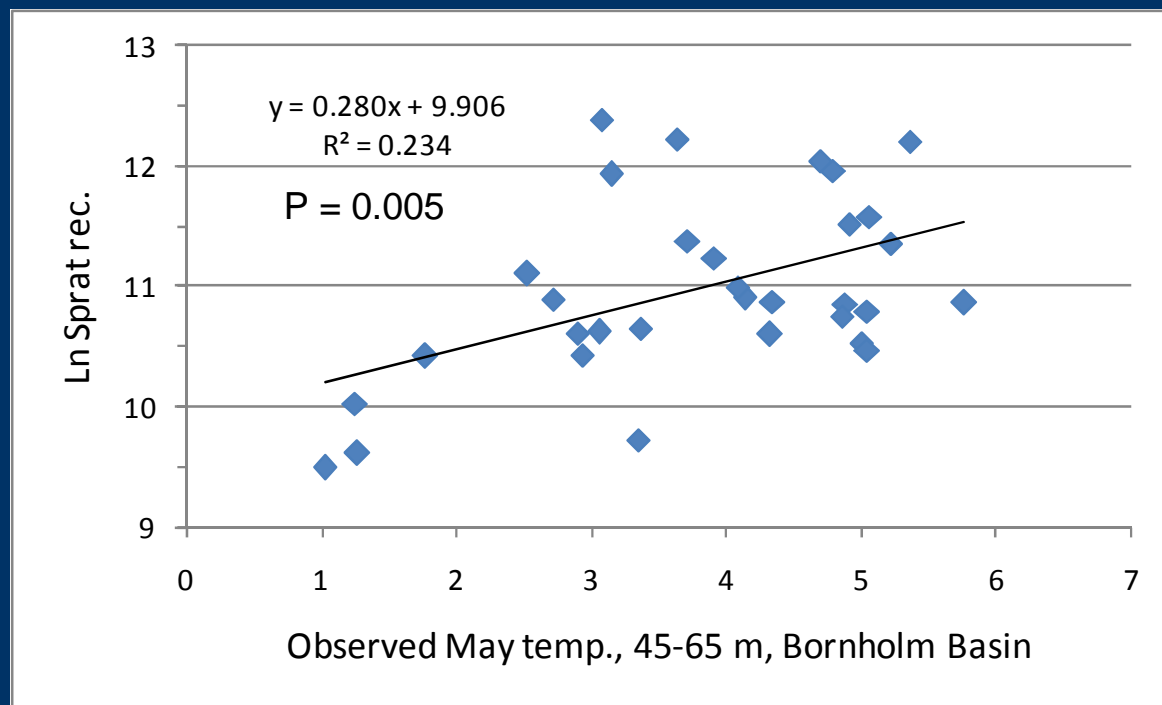


# Temperature Validations and Comparisons

-temperature affects processes such as sprat recruitment

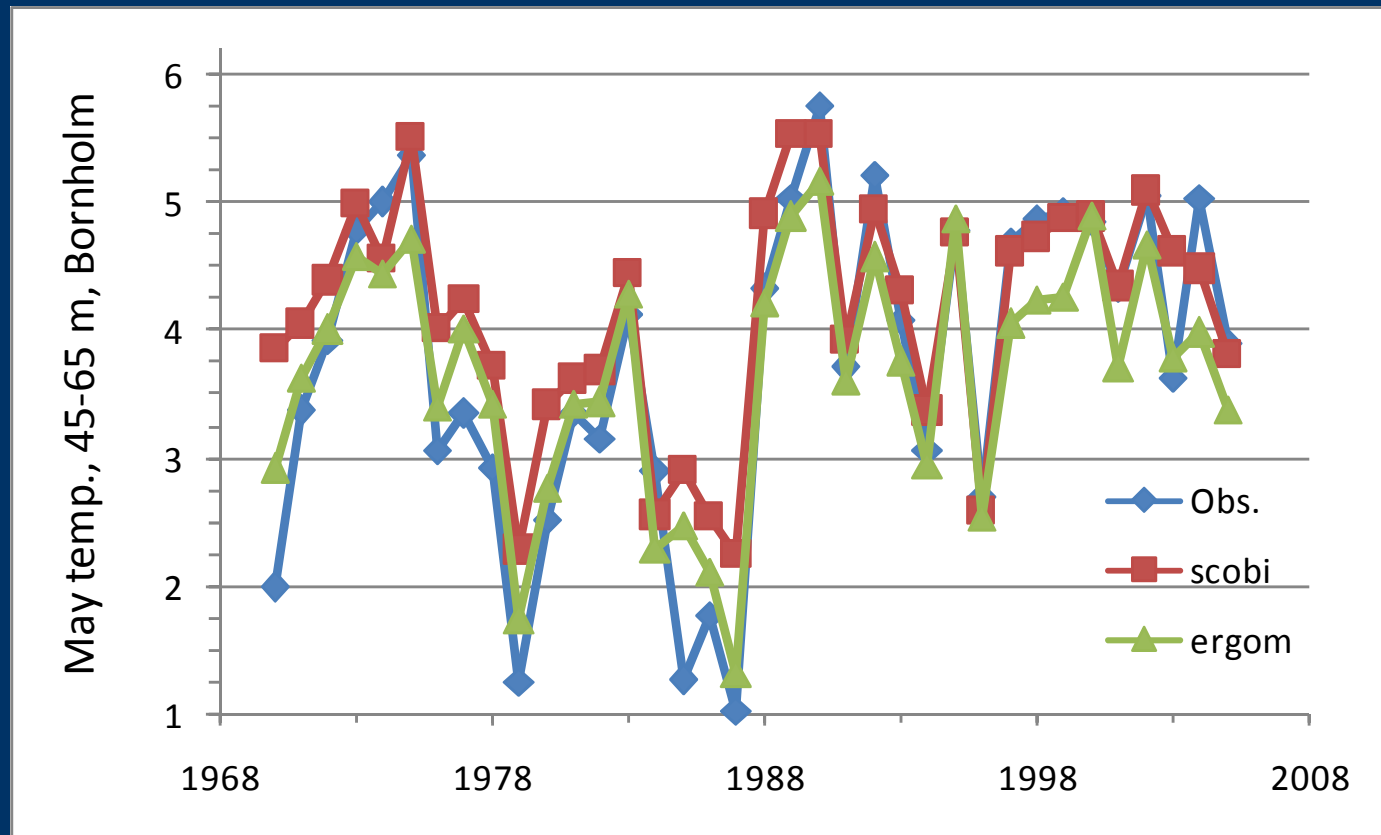
-yearclasses 1974-2005; sprat data from ICES WGBFAS 2010

-temperature data from MacKenzie et al. 2008 CJFAS



# Temperature Validations and Comparisons

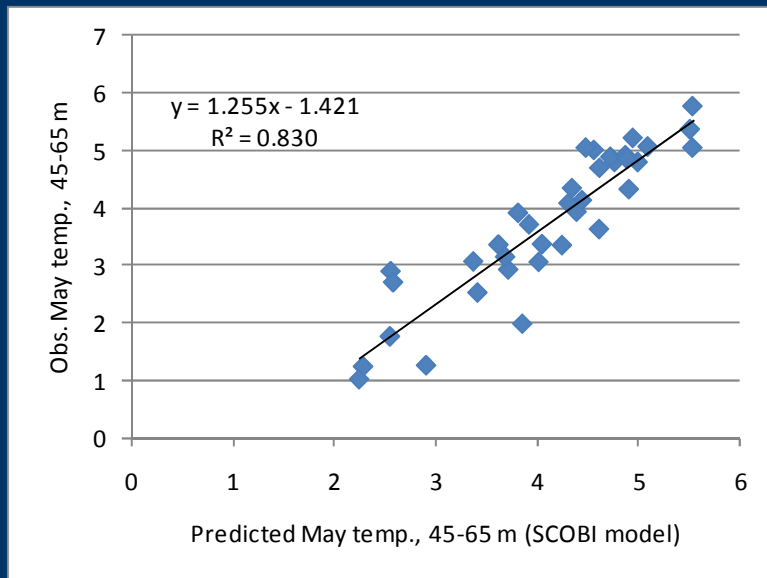
May temperature at 45-65 m, Bornholm Basin



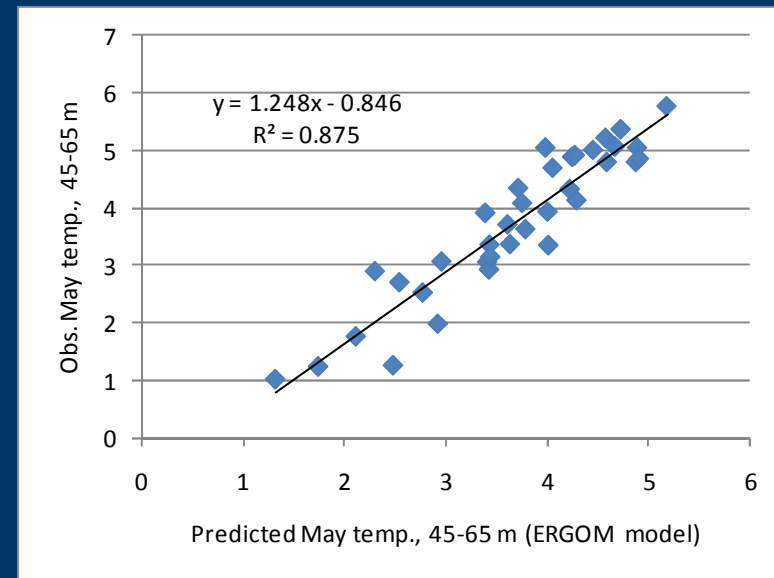
# Temperature Validations and Comparisons

May temperature at 45-65 m, Bornholm Basin

RCO-SCOBI



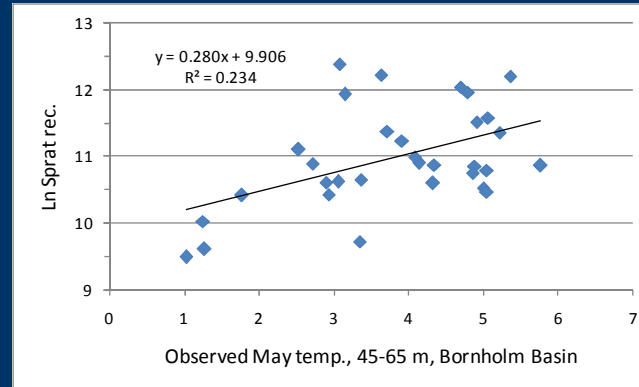
ERGOM



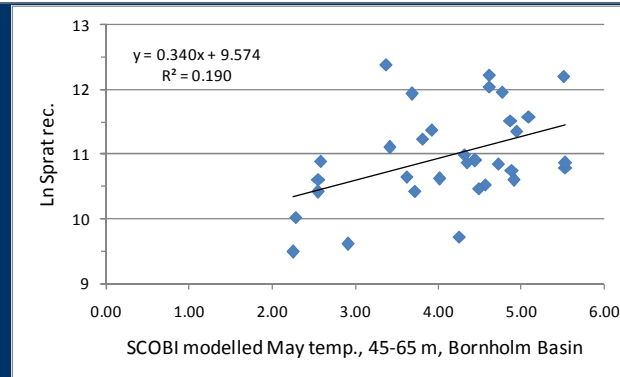
- very good correspondance between observed and modelled spring temperatures!
- are same temperatures which affect sprat recruitment...

# Do Modelled Temperatures also Explain Sprat Recruitment Variability?

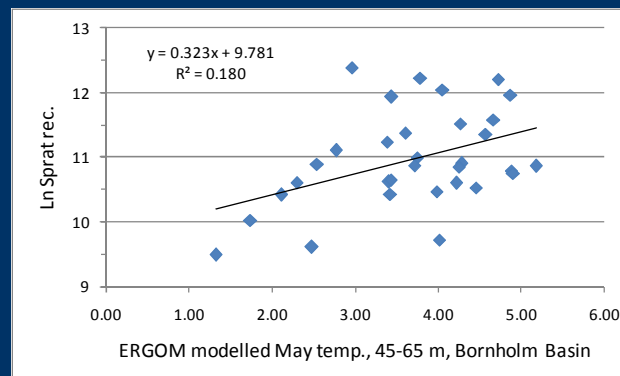
1974-2005



Observed temp.  
 $R^2 = 0.23$ ;  $P = 0.005$



RCO-SCOBImodelled temp.  
 $R^2 = 0.19$ ;  $P = 0.012$



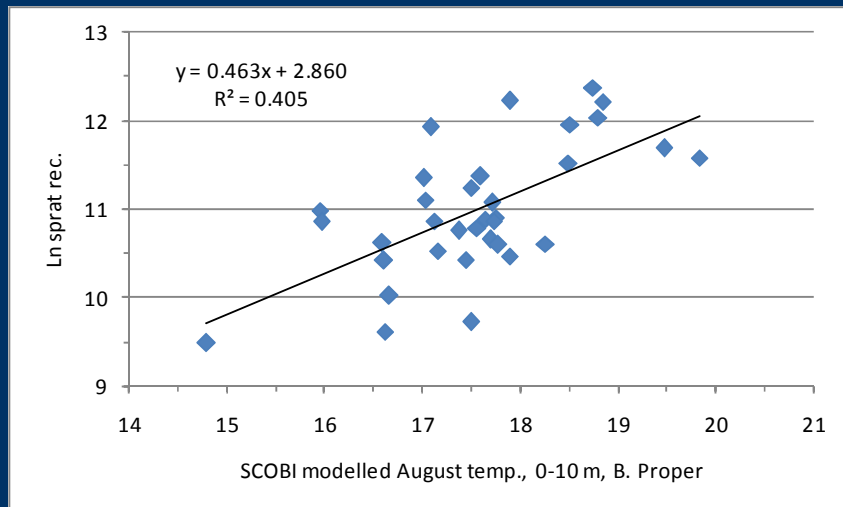
ERGOMmodelled temp.  
 $R^2 = 0.18$ ;  $P = 0.015$

# Modelled Summer Temperatures and Sprat Recruitment 1974-2007

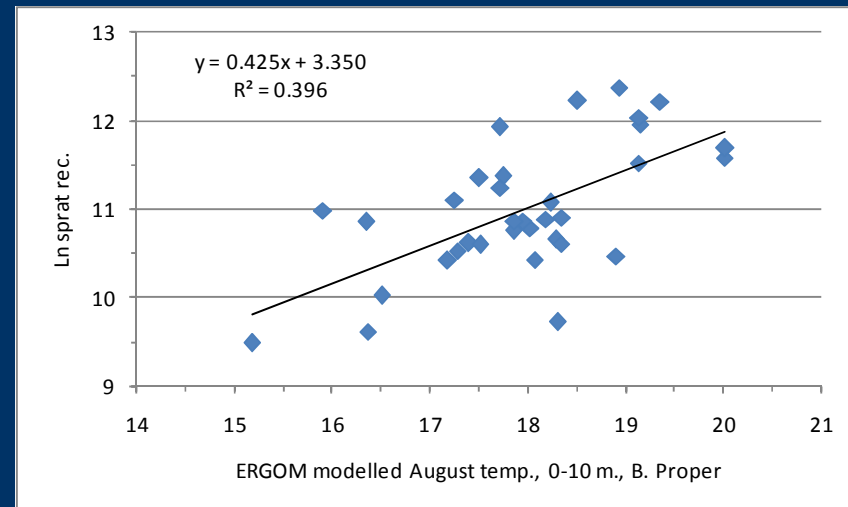
Summer surface temperature explains more variability in recruitment than spring deep temperatures (Baumann et al. 2006)

-check whether modelled summer temperatures also explain sprat rec. var.

RCO-SCOBI temp.;  $P < 0.0001$



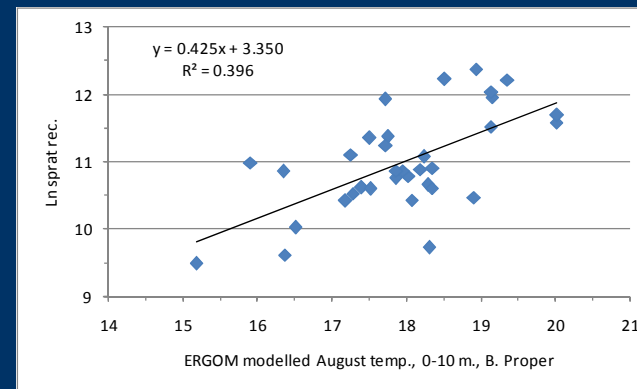
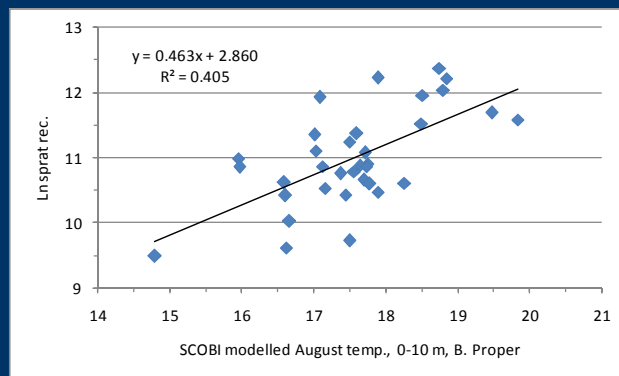
ERGOM temp.;  $P < 0.0001$



# Combining Model Outputs

In this specific case, both models perform nearly equally well

-weightings would be nearly equal



-is *not* (!) the case with other variables – e. g., cod reprod. volume

# Conclusions

- AO and NPZD models give good representation of some key hydrographic and biological variables that affect sprat and cod recruitment
- some of those variables themselves explain similar levels of variability in recruitment as observed data
- need to continue and expand analyses (Baltsem model to be included)
- some ways to proceed with ensemble averaging are possible
- very promising possibilities to use AO and NPZD models for projections of sprat and cod recruitment

-thank you





# Case Studies for Learning and Methodological Development

-temperature (e.g ., for sprat recruitment)

-cod reproductive volume – a habitat indicator for cod spawning and reproduction (based on salinity and oxygen concentration)

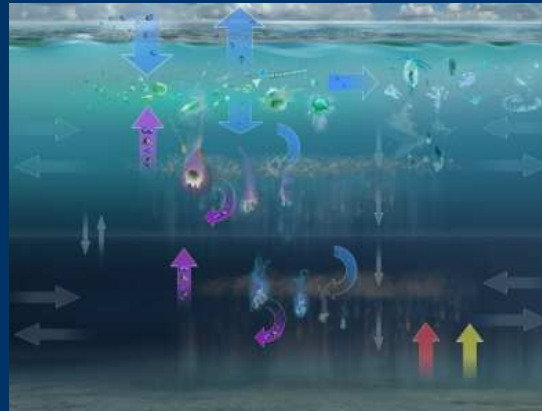
# Model Outputs Linked Sequentially

Atmos.-Ocean



T, S, ...

NPZD



Cod reproductive volume

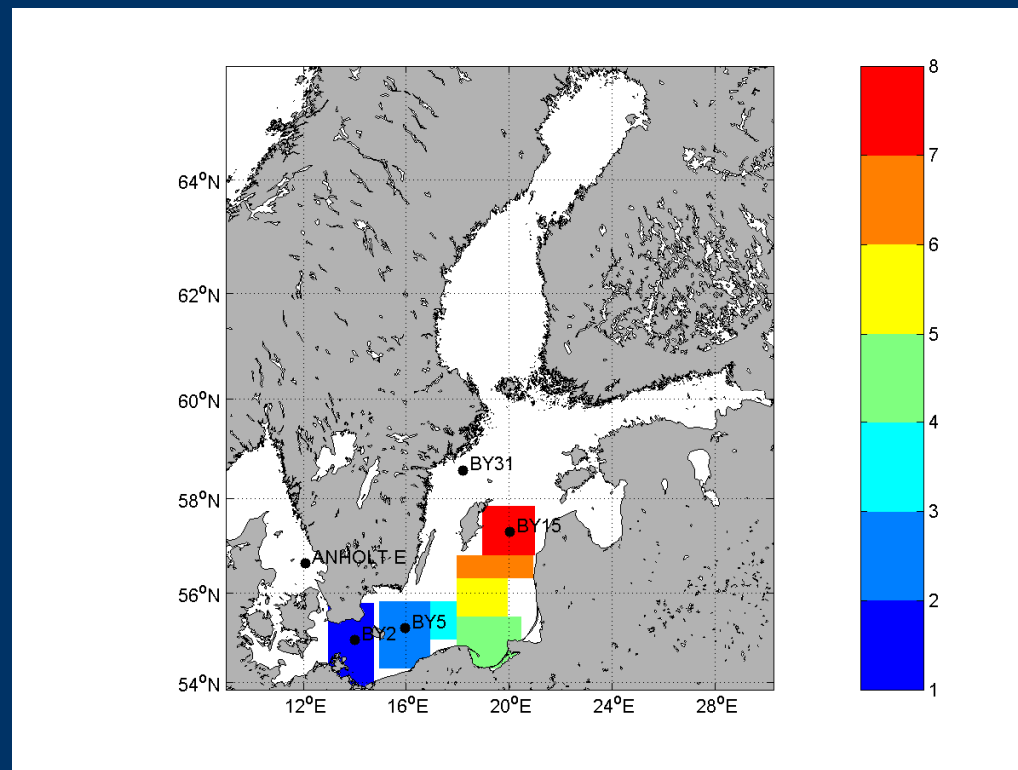
Fish/foodweb



Artwork: G. Gorick

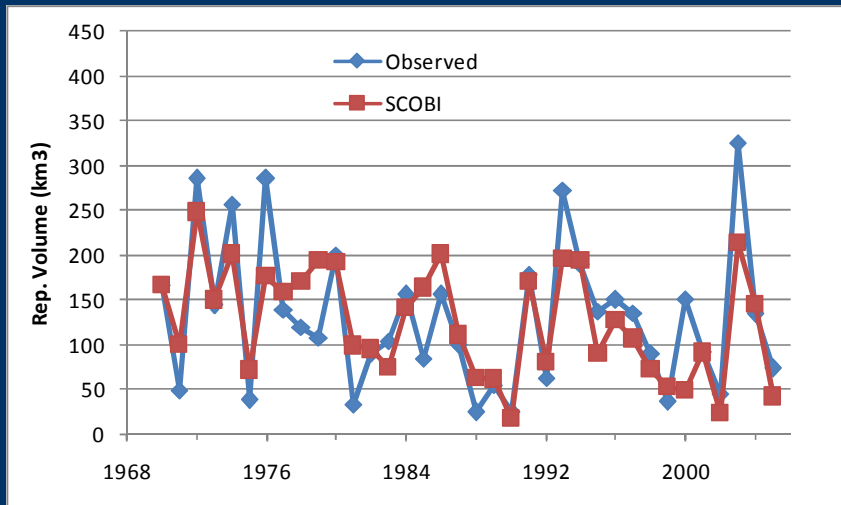
# Modelled Data Available

Cod reproductive volumes by basin and month, 1970-2005  
from SCOBİ and ERGOM

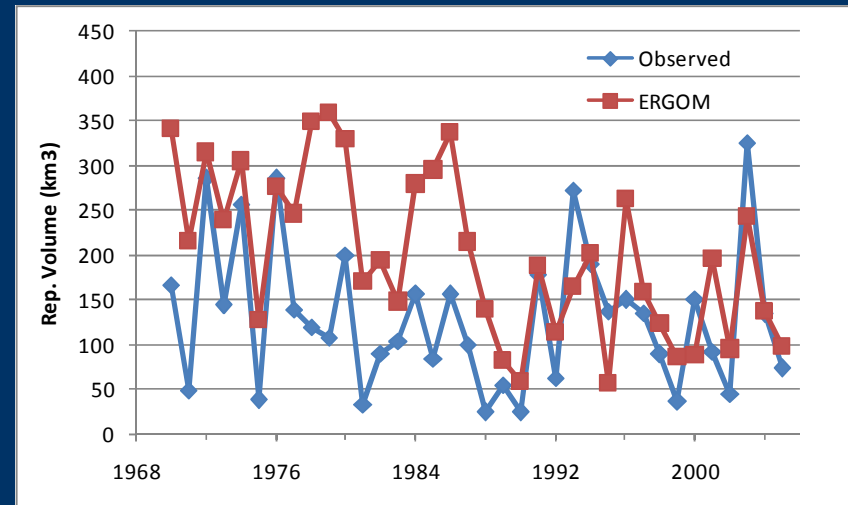


# Model Fitting

## RCO-SCOBI



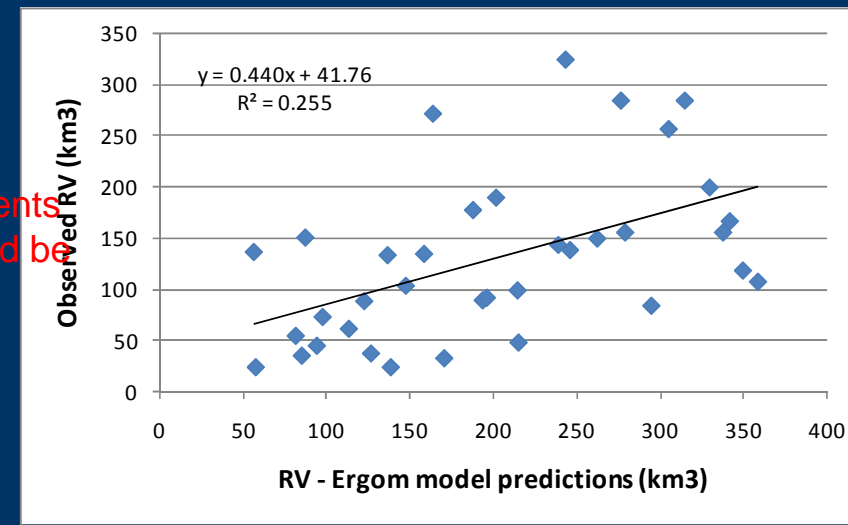
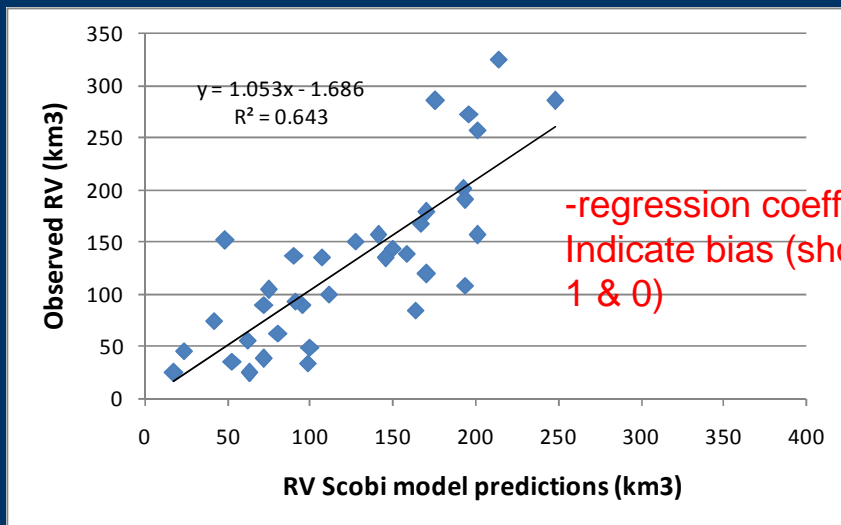
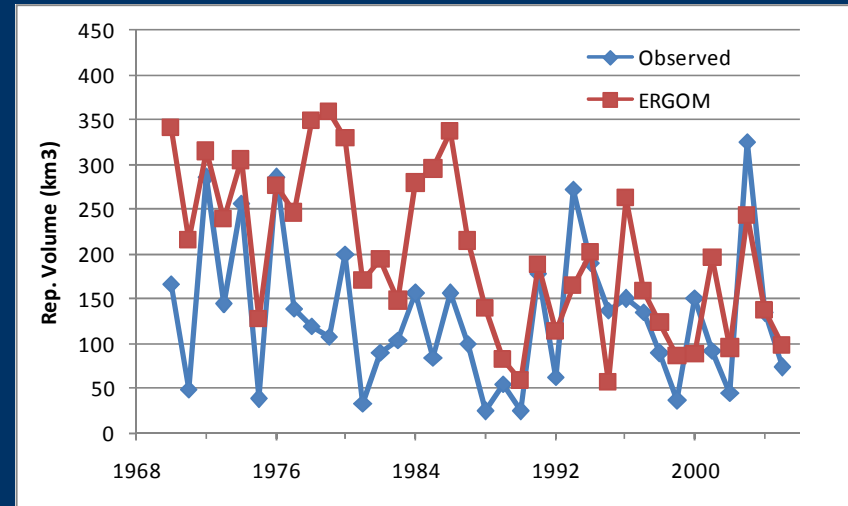
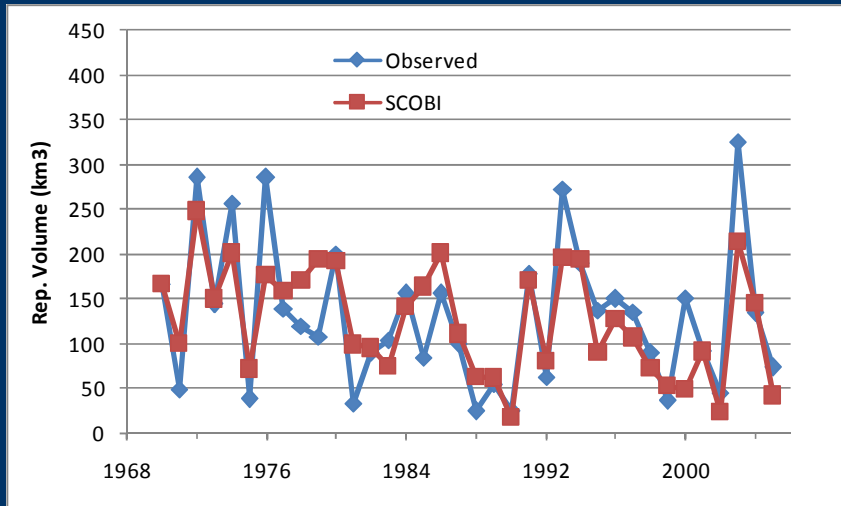
## ERGOM



# Model Fitting 1970-2005

## RCO-SCOBI

## ERGOM



# Model Comparisons of Forecasts

- both models track the past and “forecasted” data quite well
- one model explains more variation and has less biased results

# Model Outputs Linked Sequentially

Atmos.-Ocean

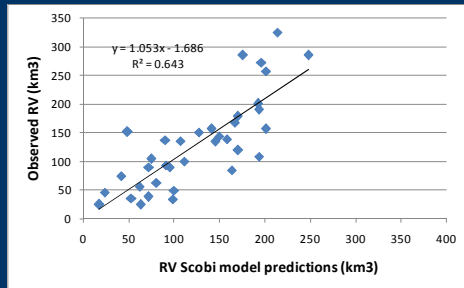


T, S, ...



Habitat (RV),  
zooplankton, ...

NPZD



-RV uncertainty from  
NPZD models is  
quantifiable  
-can be implemented in  
cod rec. models

Fish/  
foodweb

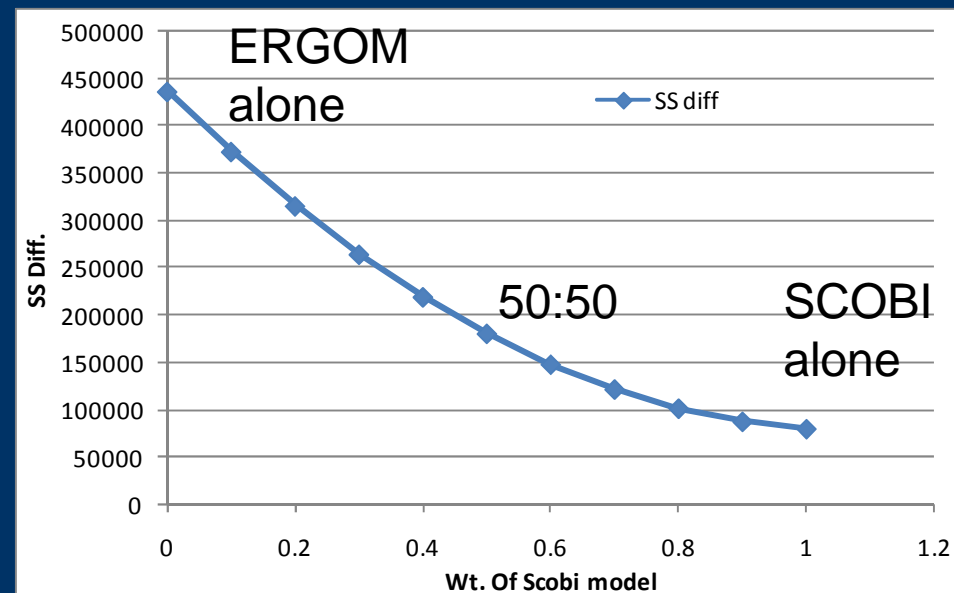


# Sensitivity to Weighting in Ensemble Averages

August	Model 1 SCOB1	Model 2 ERGOM	wt-scobi	wt-ergom	wt. Pred.	Observed Plikshs	(o-p) <sup>2</sup>
Bornholm_ 1970	167	342	1	0	167	167	0.1
Bornholm_ 1971	100	215	1	0	100	48.6	2614.4
Bornholm_ 1972	248	315	1	0	248	285	1380.9
Bornholm_ 1973	150	239	1	0	150	144	34.7

-calc. sum of squared differences (= SSE, SSRes)

-compare SS Diff. for different weightings of the two models.



-best model always penalized by weighting with less effective model



# Conclusions

- AO and NPZD models give good representation of some key hydrographic and biological variables that affect cod and sprat recruitment (!)
- some of those variables themselves explain similar levels of variability in recruitment as observed data (!)
- need to continue and expand analyses (3<sup>rd</sup> model included)
- some ways to proceed with ensemble averaging are possible
- very promising possibilities to use AO and NPZD models for projections of cod and sprat recruitment

-thank you





# Questions

How is uncertainty and variability in an AO model output (for a given scenario) passed forward into the NPZD or fish models?

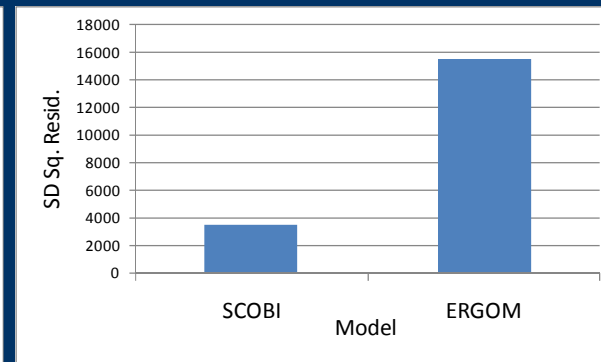
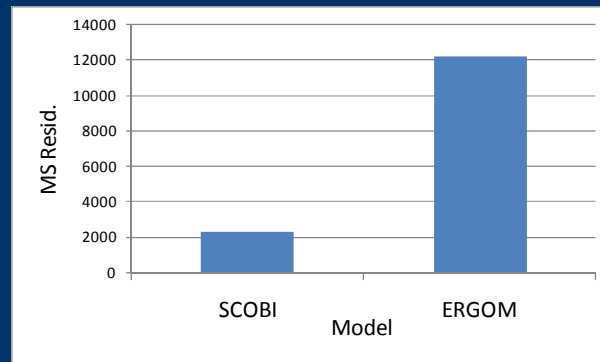
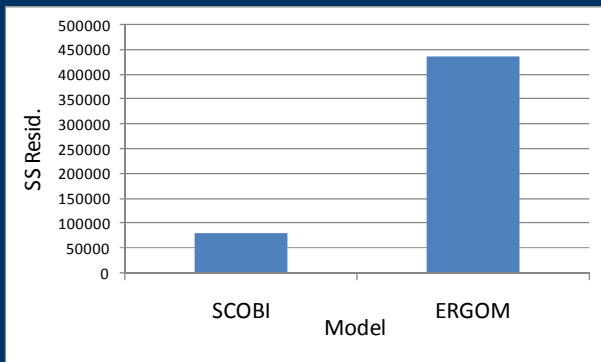
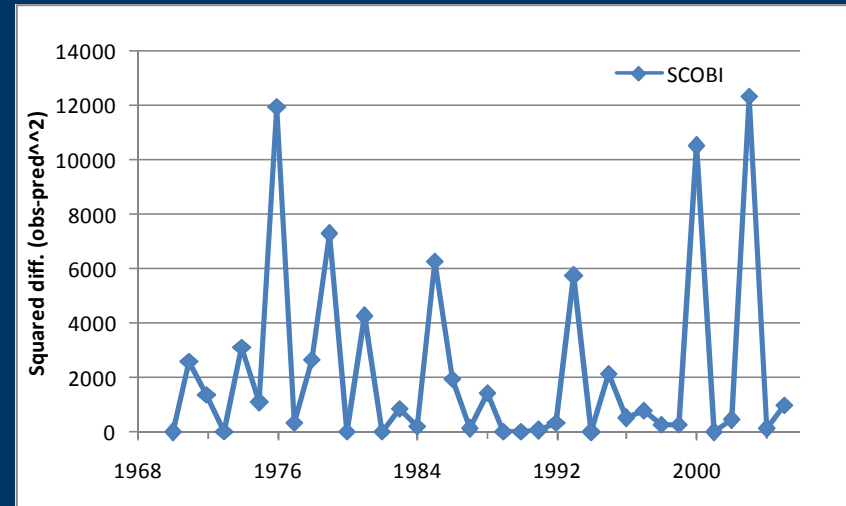
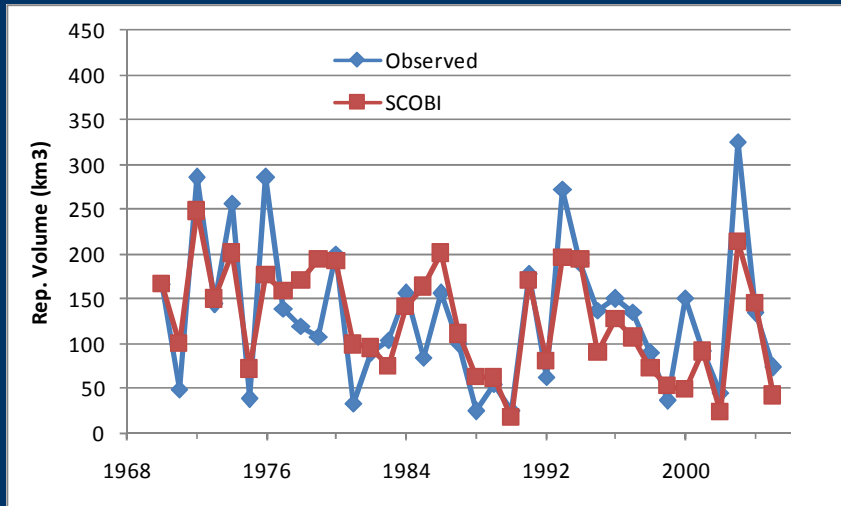
Is this possible to do?

# Modelling the Future

Confidence in estimating past variations

Applications for future projections

# Residual Diagnostics and Uncertainties



# Questions

How is uncertainty and variability in an AO model output (for a given Scenario) passed forward into the NPZD or Fish models?

Is this possible to do?

How should model outputs be combined into 1 time series?

Some options:

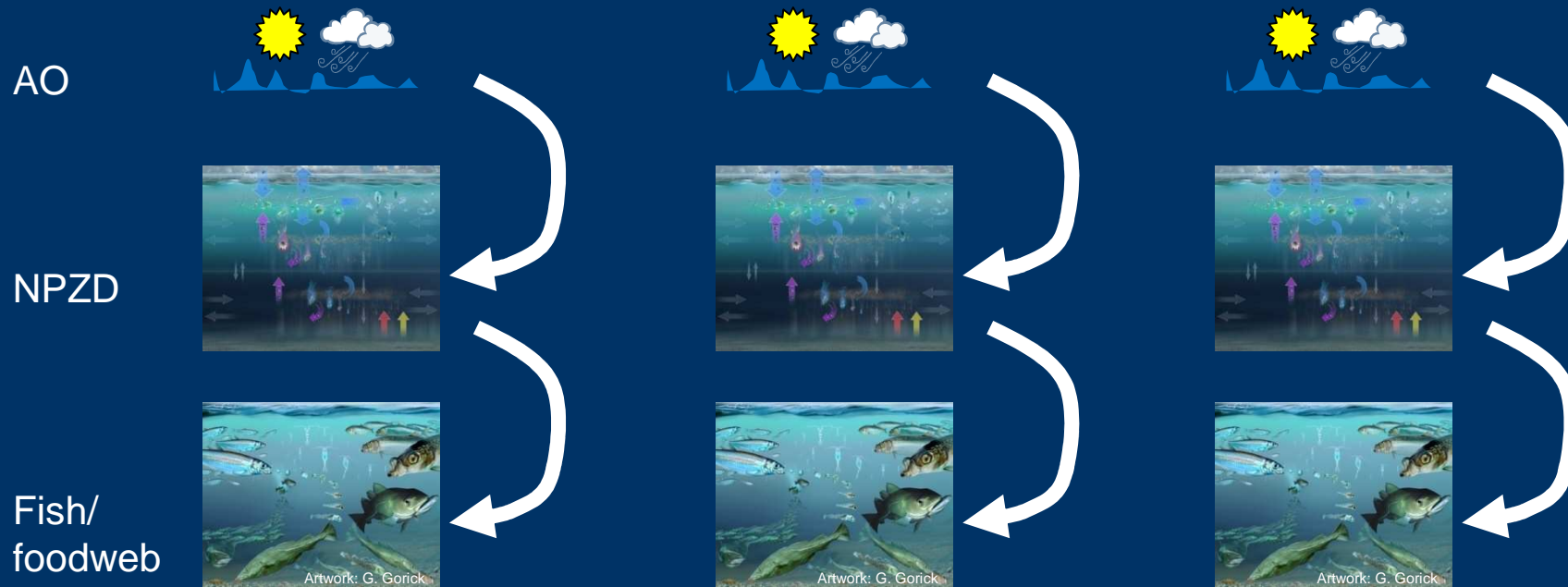
- simple unweighted average across models

Or account for past performance :

- select only the model with best fit to observed data

- weigh outputs according to past performance with observed data, and calculate weighted average

# Outputs Combined Across Models



# Future Projections with Validated Models of the Past

