

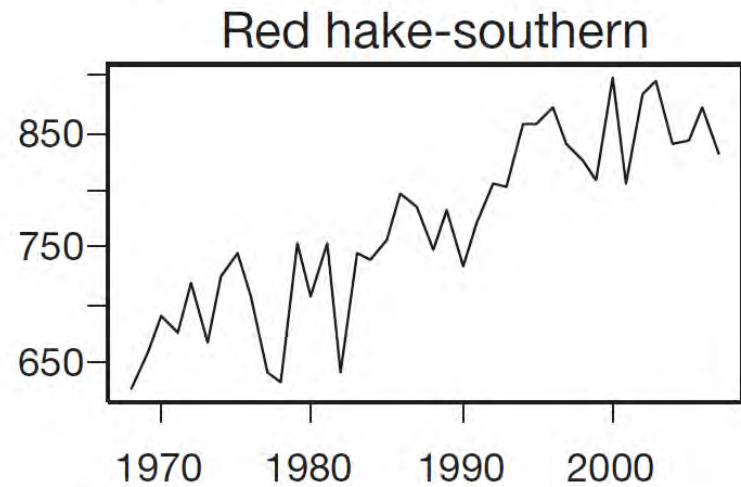
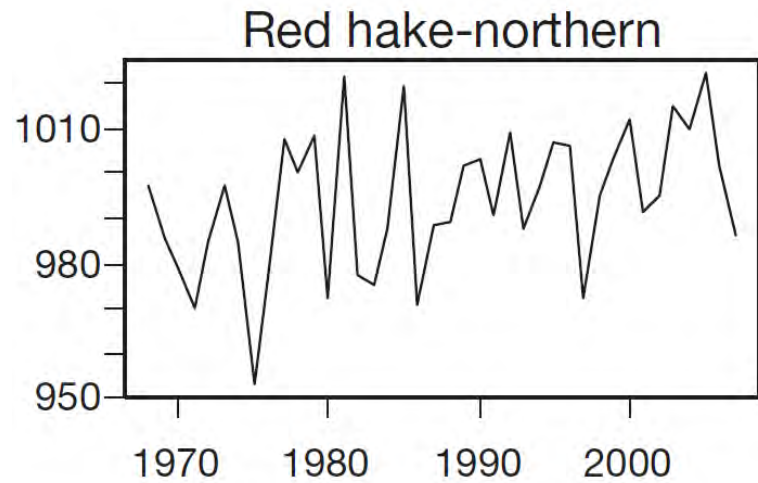
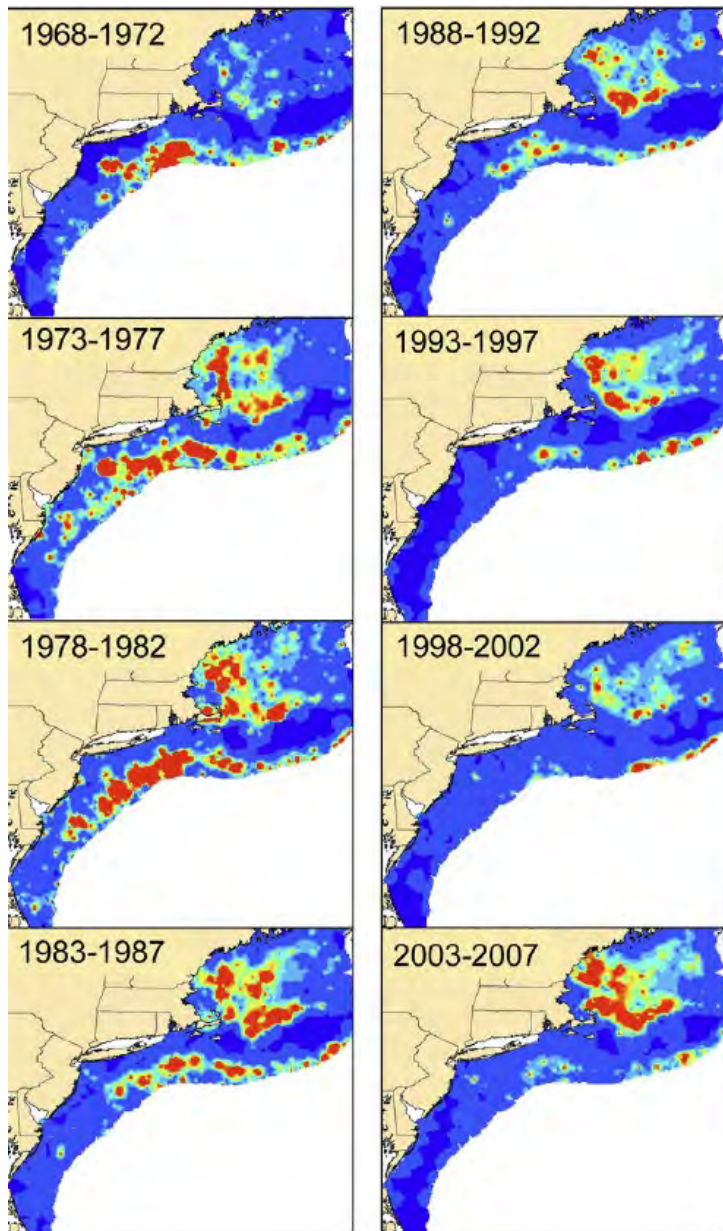
Predicting fish movement and migration in response to a changing climate

Kenneth Rose
University of Maryland Center
for Environmental Science
Horn Point Laboratory
Cambridge, Maryland



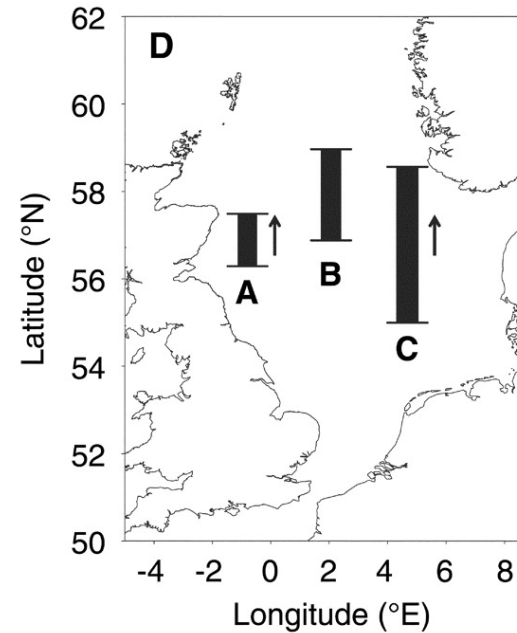
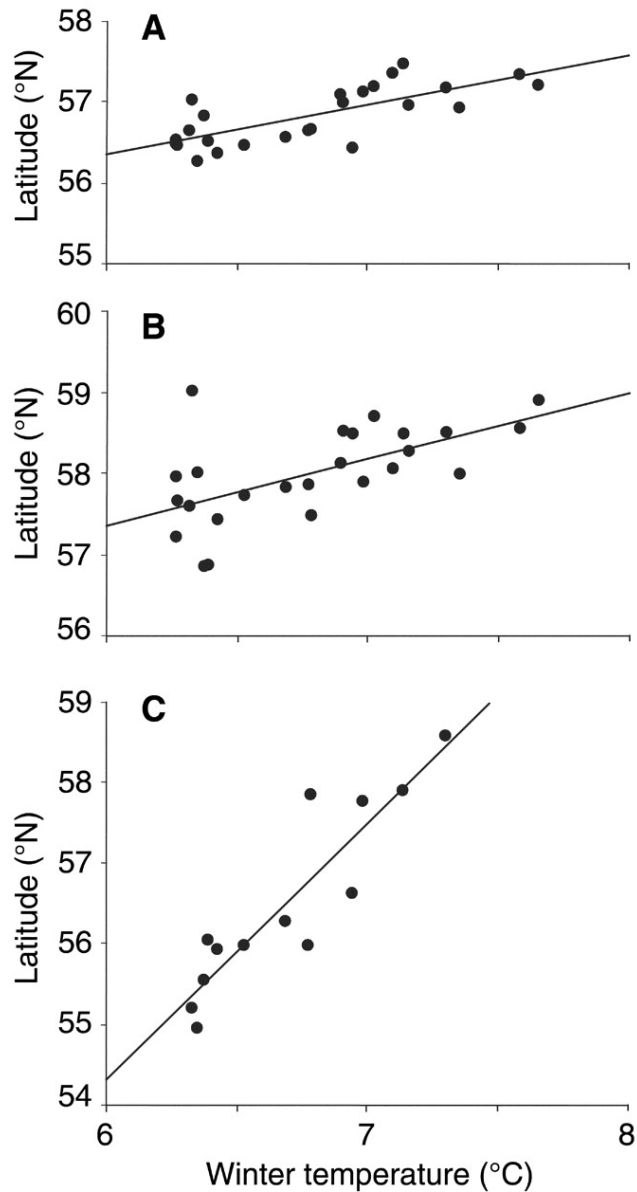
Today

- Organisms will move in response to climate change
- Information informing management is simplified
- Progress on movement
 - Observations
 - Modeling
 - Examples
- Needs and opportunities

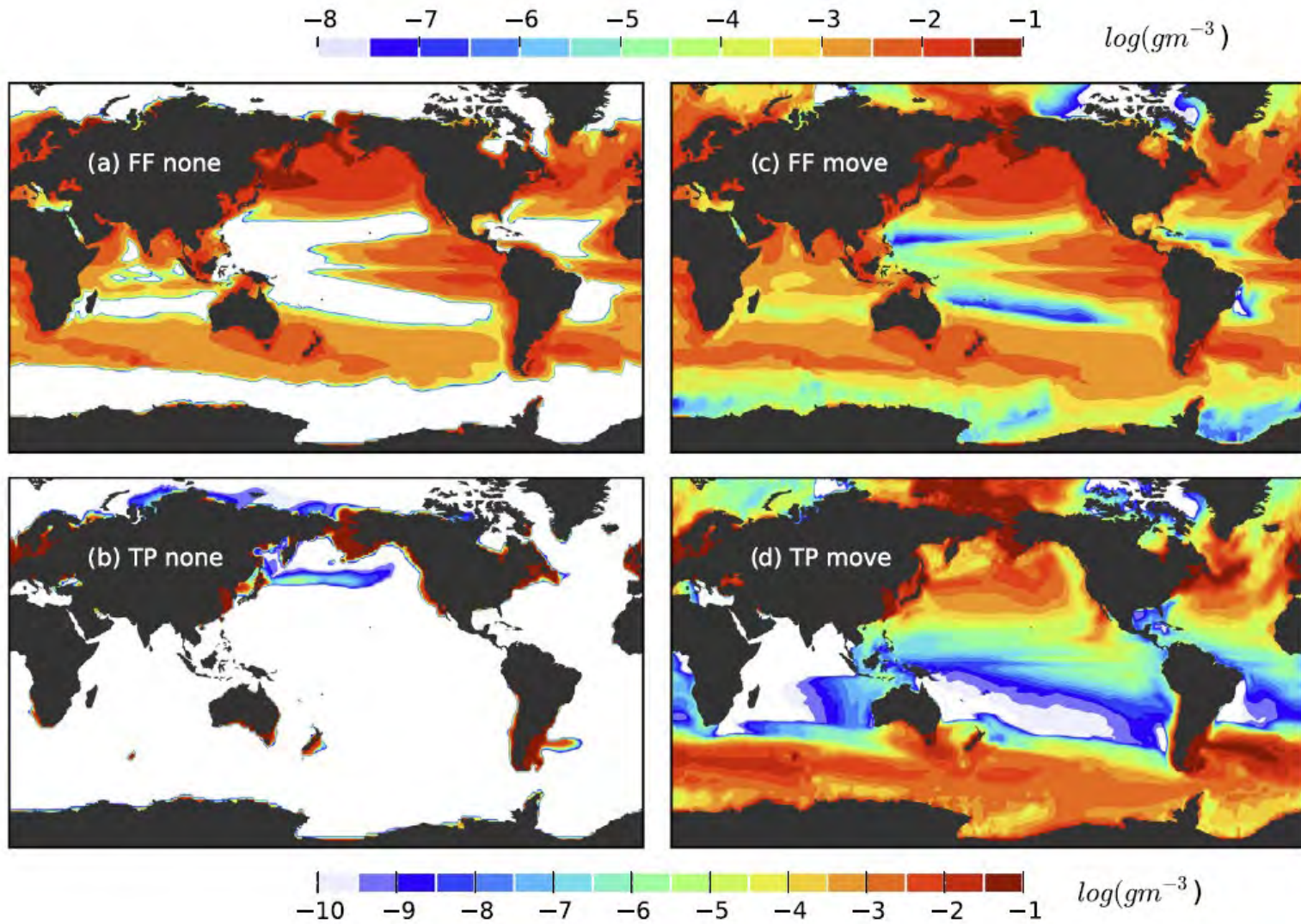


Nye et al. 2009. Marine Ecology Progress Series 393: 111-129.

Fig. 1. Examples of North Sea fish distributions that have shifted north with climatic warming.



Allison L. Perry et al. *Science* 2005;308:1912-1915



Watson et al. 2015. Progress in Oceanography 138: 521-532.

Progress: Movement Data

REVIEW SUMMARY

ECOLOGY

Aquatic animal telemetry: A panoramic window into the underwater world

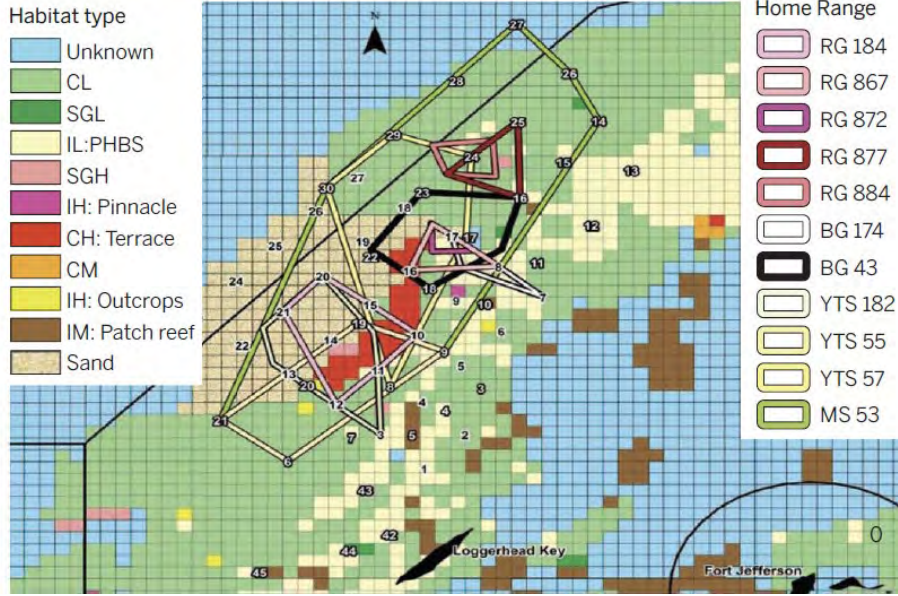
Nigel E. Hussey, Steven T. Kessel, Kim Aarestrup, Steven J. Cooke, Paul D. Cowley, Aaron T. Fisk, Robert G. Harcourt, Kim N. Holland, Sara J. Iverson,* John F. Kocik, Joanna E. Mills Flemming, Fred G. Whoriskey

Science 348: 1255642, 2015



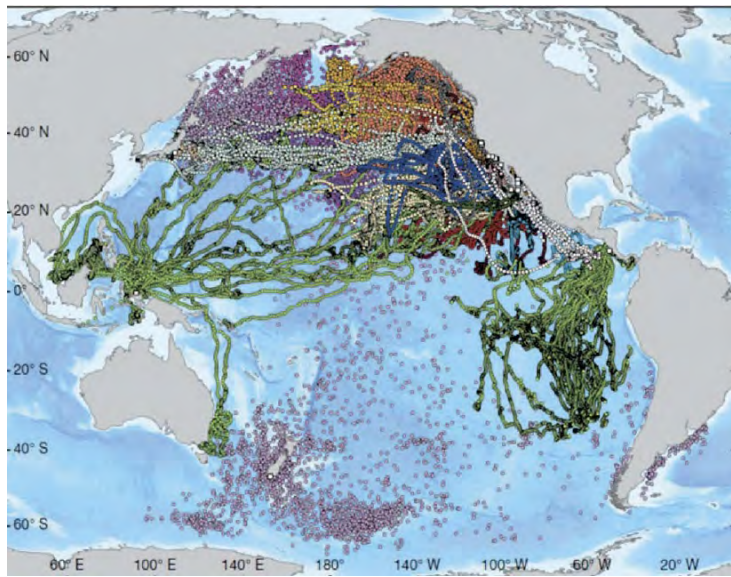
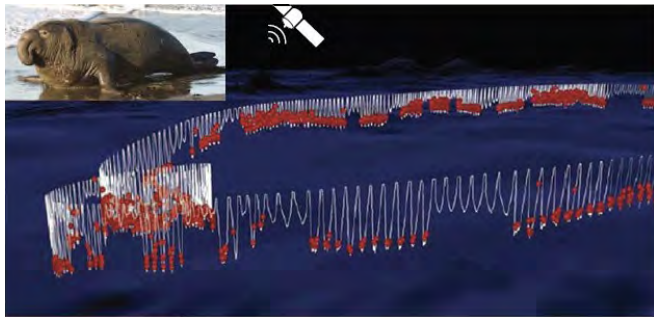
Habitat type

- Unknown
- CL
- SGL
- IL:PHBS
- SGH
- IH: Pinnacle
- CH: Terrace
- CM
- IH: Outcrops
- IM: Patch reef
- Sand

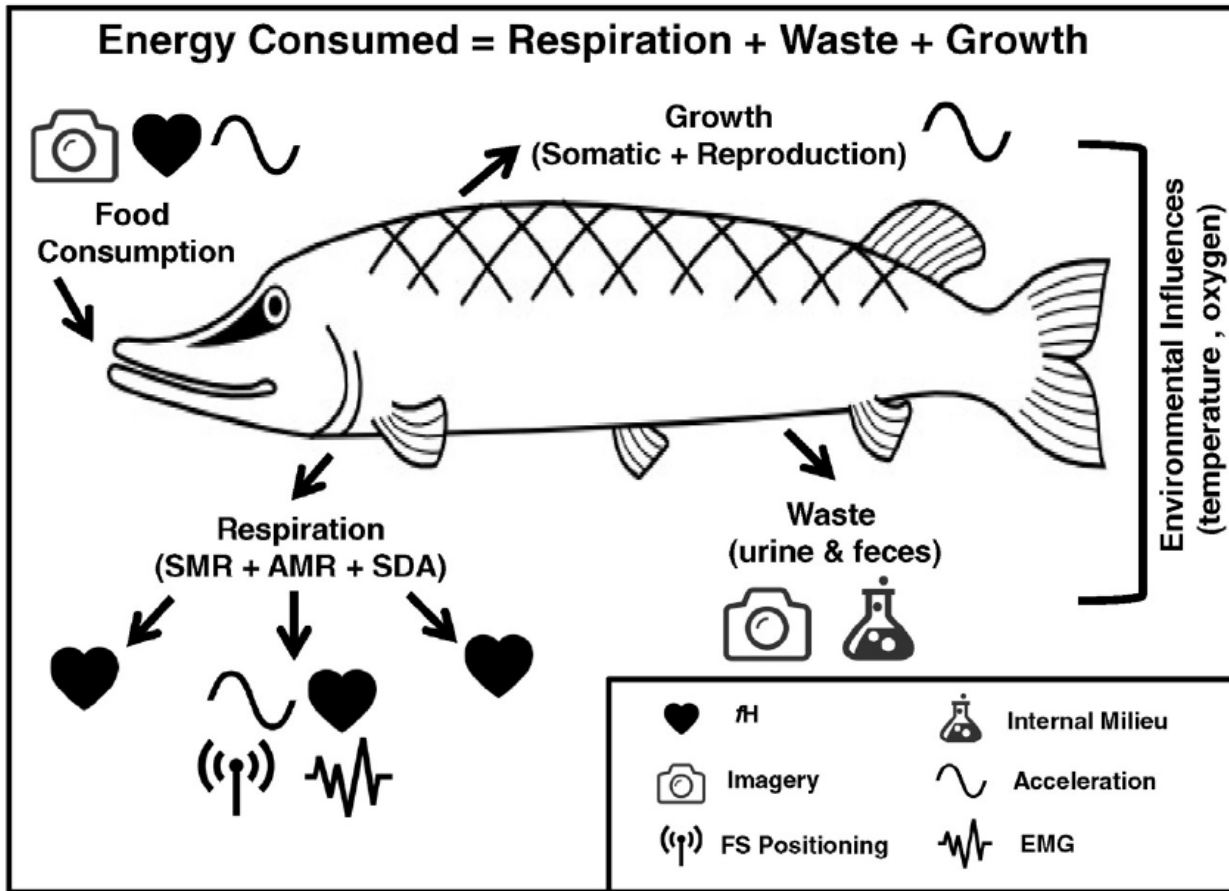


Home Range

- RG 184
- RG 867
- RG 872
- RG 877
- RG 884
- BG 174
- BG 43
- YTS 182
- YTS 55
- YTS 57
- MS 53



- Humpback whale
- Fin whale
- Sperm whale
- Sooty shearwater
- California sea lion
- Northern fur seal
- Blue whale
- Northern elephant seal
- Thresher shark
- Yellowfin tuna
- Albacore tuna
- Blue shark
- Mako shark
- White shark
- Loggerhead turtle
- Mola mola*
- Pacific bluefin tuna
- Leatherback turtle
- Salmon shark
- Laysan albatross
- Black-footed albatross
- Humboldt squid



Contents lists available at ScienceDirect

Comparative Biochemistry and Physiology, Part A

journal homepage: www.elsevier.com/locate/cbpa



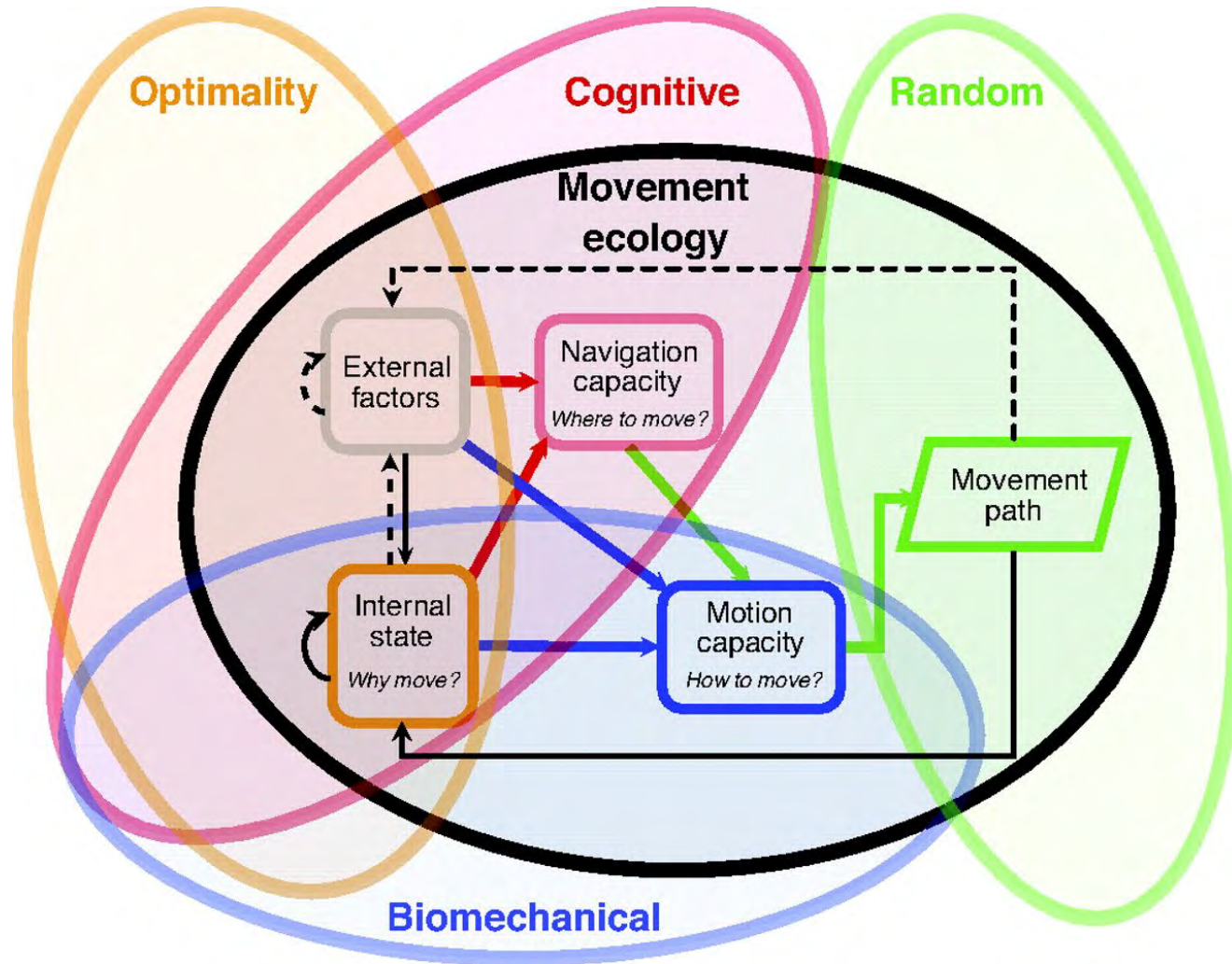
Remote bioenergetics measurements in wild fish: Opportunities and challenges☆

Steven J. Cooke^{a,*}, Jacob W. Brownscombe^a, Graham D. Raby^b, Franziska Broell^c, Scott G. Hinch^d, Timothy D. Clark^e, Jayson M. Semmens^f

Progress: Movement Modeling

- Many approaches have been proposed
 - $X(t+1) = X(t) + V_x(t)$
 - $Y(t+1) = Y(t) + V_y(t)$
 - $Z(t+1) = Z(t) + V_z(t)$
 - Determine the cell
- Quite confusing because of non-standard descriptions and terminology for V_x , V_y , and V_z
 - Random walk
 - Run and tumble
 - Event-based
 - Restricted-area
 - Kinesis
 - ANN





Ran Nathan et al. J Exp Biol 2012;215:986-996

PHILOSOPHICAL TRANSACTIONS OF THE ROYAL SOCIETY B

BIOLOGICAL SCIENCES

Collective movement ecology

Theme issue compiled and edited by Andrew M. Berdahl, Dora Biro, Peter A.H. Westley and Colin J. Torney



THE
ROYAL
SOCIETY
PUBLISHING



MOVEMENT ECOLOGY GROUP



biomove symposium 2018

Integrating Biodiversity Research with Movement Ecology



Movement Ecology of Animals

Gordon Research Conference

Major Issue

- If we are to use these methods to simulate management actions and climate change, then the methods must predict responses to changes in cue(s)

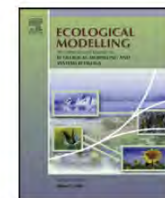
Ecological Modelling 250 (2013) 214–234



Contents lists available at SciVerse ScienceDirect

Ecological Modelling

journal homepage: www.elsevier.com/locate/ecolmodel

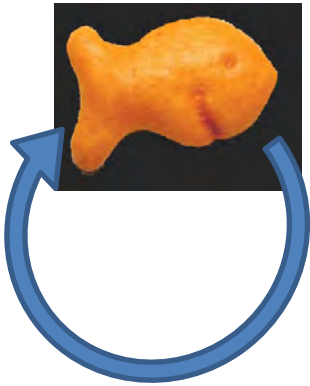


Evaluating the performance of individual-based animal movement models in novel environments

Katherine Shepard Watkins*, Kenneth A. Rose

Model Structure

**Simplified
Hypothetical
Species**



Scale

Grid: 540 x 540 cells

Cells: 5 m²

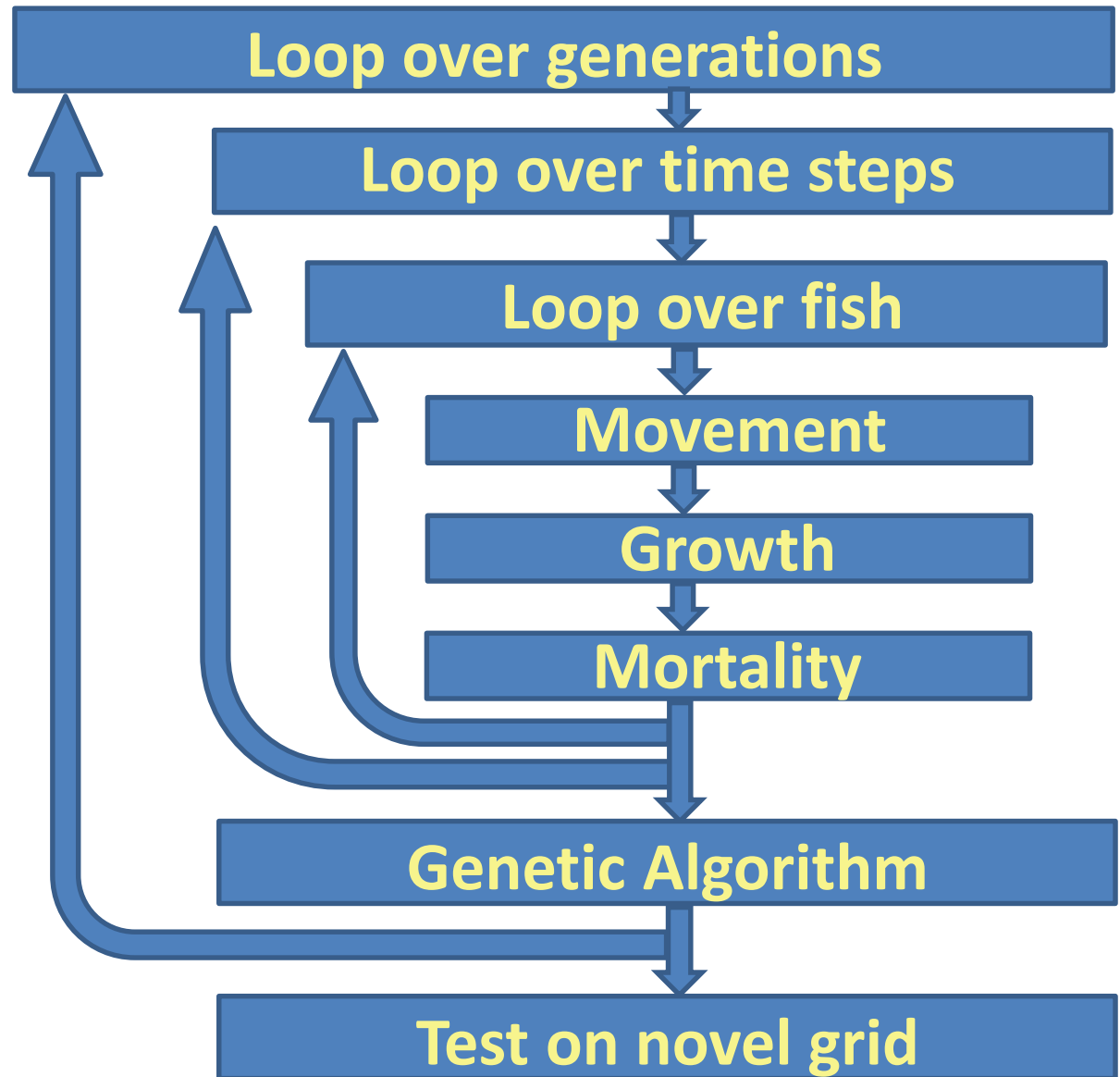
Time step: 5 minute

Generation: 30 days

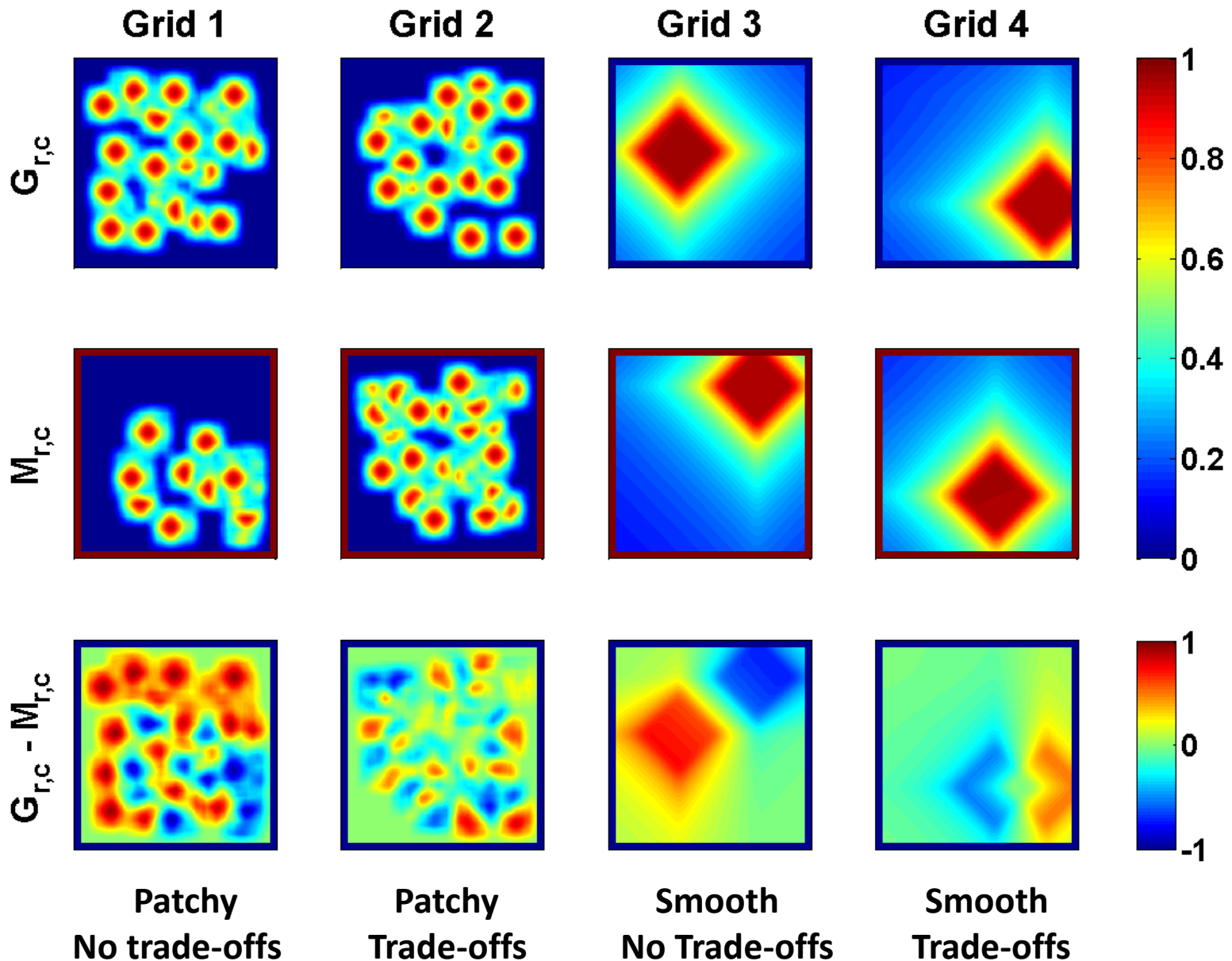
Initial size = 73.3 mm

Initial worth = 100 fish

3000 super-individuals



Environmental Gradients



Model Processes

Growth (mm 5-min⁻¹)

$$G = G_{\max} * G_{r,c}$$

$$L(t+1) = L(t) + G$$

$$W(t+1) = a * L(t+1)^b$$

Mortality (5-min)⁻¹

$$M = M_{\max} * M_{r,c} * M_L$$

$$S(t+1) = S(t) * e^{-M}$$

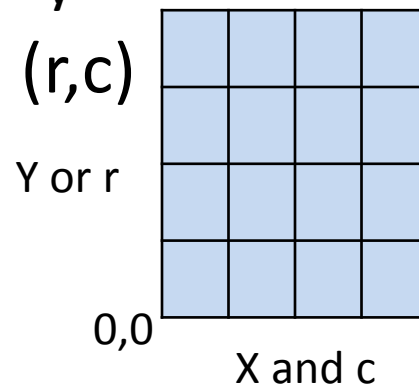
$$M_L = 1 - \frac{L_i^{-73.3}}{L_{\max}^{-73.3}}$$

Movement

$$X(t+1) = X(t) + V_x(t)$$

$$Y(t+1) = Y(t) + V_y(t)$$

cell location (r,c)



Reproduction

$$E = 55 \cdot S(30) \cdot (421.84 \cdot W(30) + 304.79)$$

GA Calibration

- 3000 strategy vectors of parameter values
 - Start with random values for everyone
- Every 30-day generation, select 3000 individuals:
 - $P(\text{selection}) = E_i / \Sigma E$
 - Mutate each vector: 6% of parameters, ± 0.25
- Use these 1000 vectors for the next generation
- Continue until egg production levels off
- Parameter values should have converged

Restricted Area Search

- Rank cells in a D_{hood} cell radius by habitat quality ($Q_{c,r}$)

$$Q_{c,r} = (1 - \delta) * (G_{c,r} + n) - \delta * (M_{c,r} * M_L + n)$$

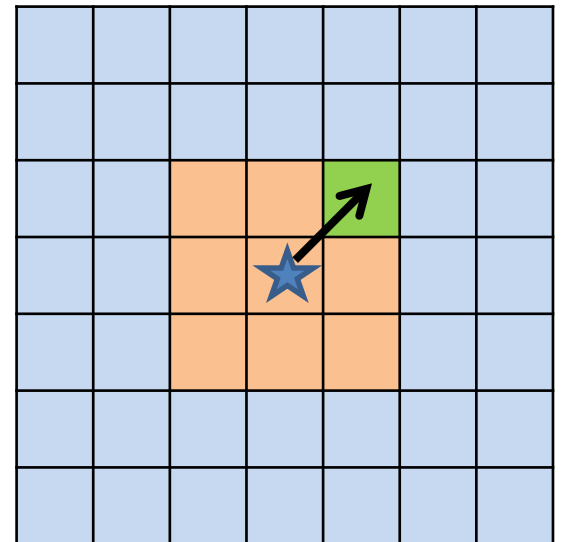
- $$n = \left(1 - \frac{1.42}{\sqrt{(c-x_{cell})^2 + (r-y_{cell})^2}} \right)$$

- Compute Θ = toward the cell with the highest $Q_{c,r}$

$$V_x(t) = (SS + RV_1 \cdot R_{dist}) \cdot \cos(\theta + RV_2 \cdot R_\theta)$$

$$V_y(t) = (SS + RV_1 \cdot R_{dist}) \cdot \sin(\theta + RV_2 \cdot R_\theta)$$

- GA evolves: δ , R_θ , R_{dist} , D_{hood}



Kinesis

- Velocities are the sum of inertial (f) and random (g)

- Compute random swim speed: $\varepsilon_x = N(\sqrt{1.0/2}, 0.5)$

- Compute habitat quality: $Q_{c,r} = (1 - \delta) * G_{c,r} - \delta * M_{c,r} * M_L$

○

- Compute f and g weighted by how close habitat quality ($Q_{c,r}$) is to the optimal habitat (Q_{opt})

$$f_x = \text{Vel}_x(t-1) \cdot H_1 \cdot e^{-0.5 \left(\frac{Q_{c,r} - Q_{opt}}{\sigma_Q} \right)^2}$$

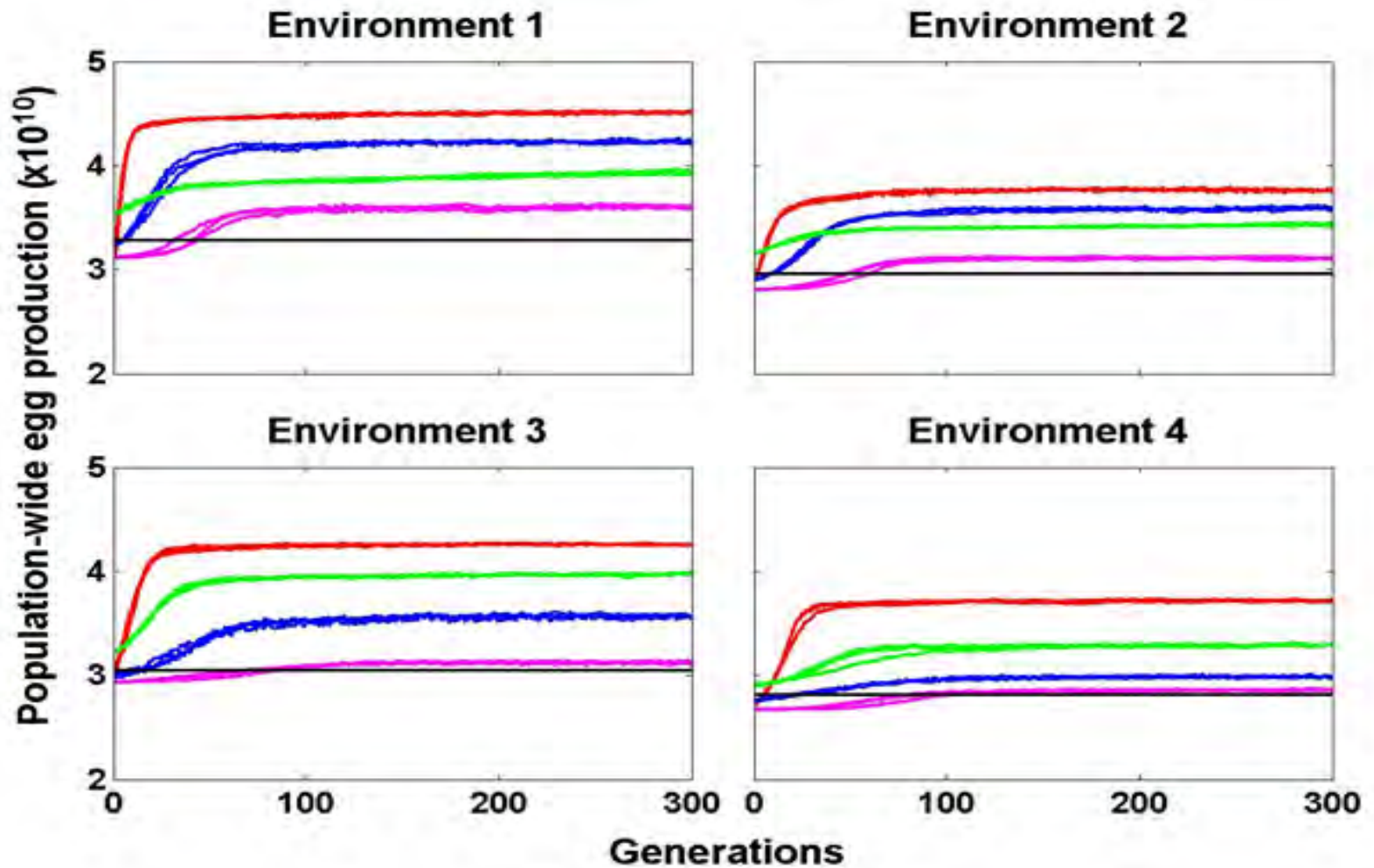
$$g_x = \varepsilon_x \cdot \left(1 - H_2 \cdot e^{-0.5 \left(\frac{Q_{c,r} - Q_{opt}}{\sigma_Q} \right)^2} \right)$$

$$V_x(t) = f_x + g_x$$

$$V_y(t) = f_y + g_y$$

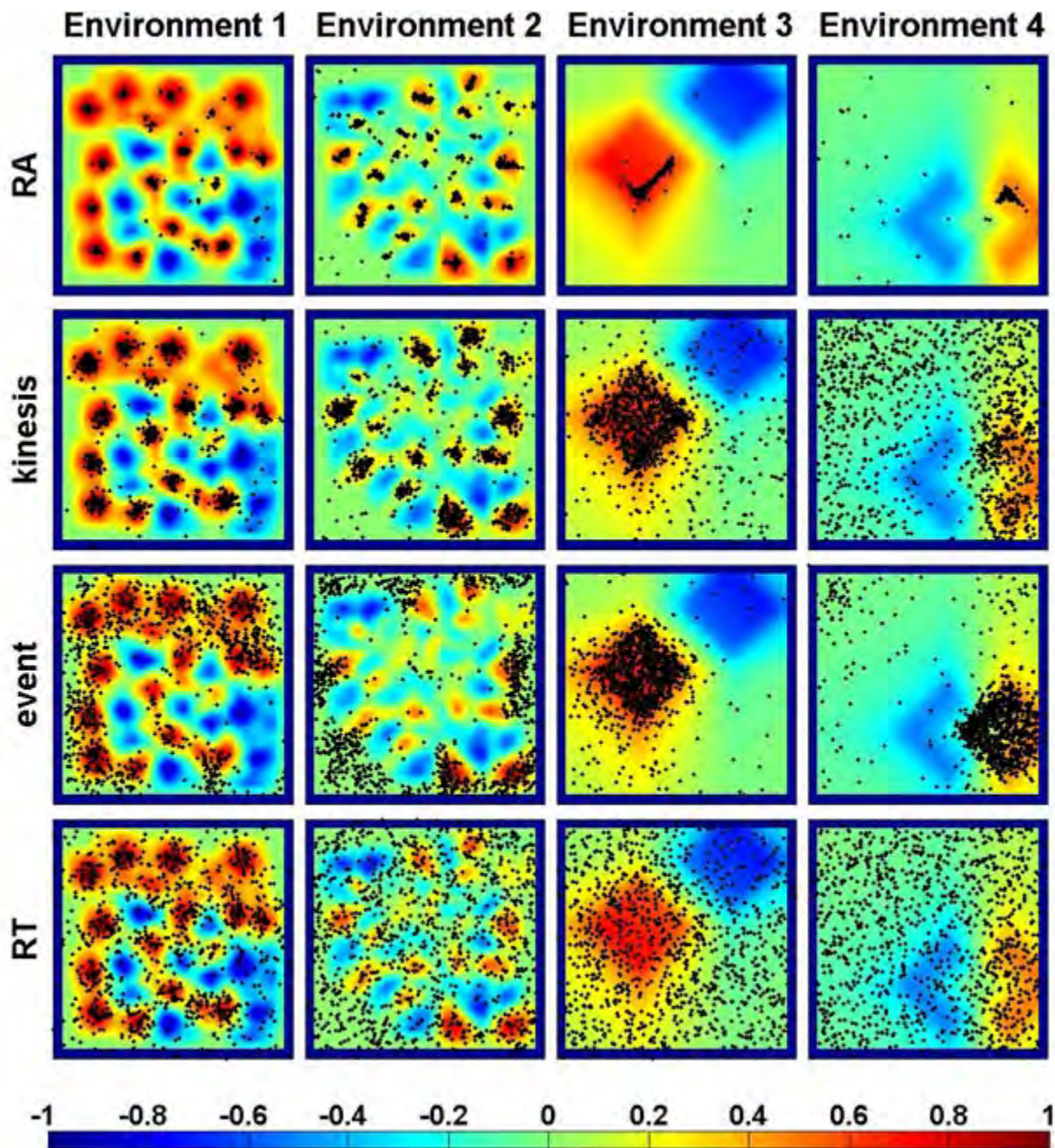
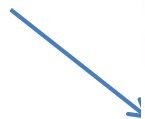
- GA evolves Q_{opt} , σ , H_1 , H_2 , δ

Calibration – Fitness Convergence

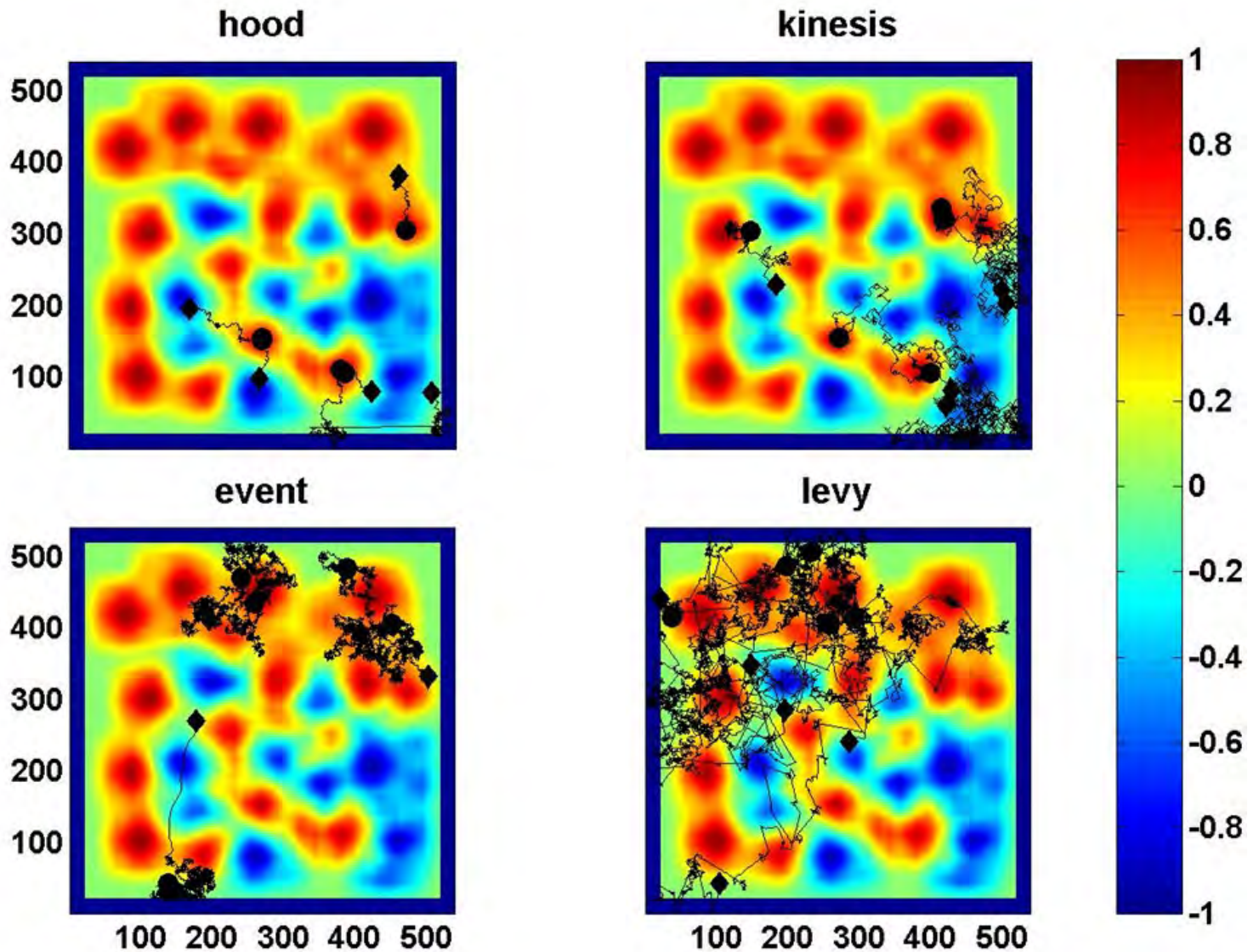


Restricted area, Kinesis, Event-based, Run-tumble

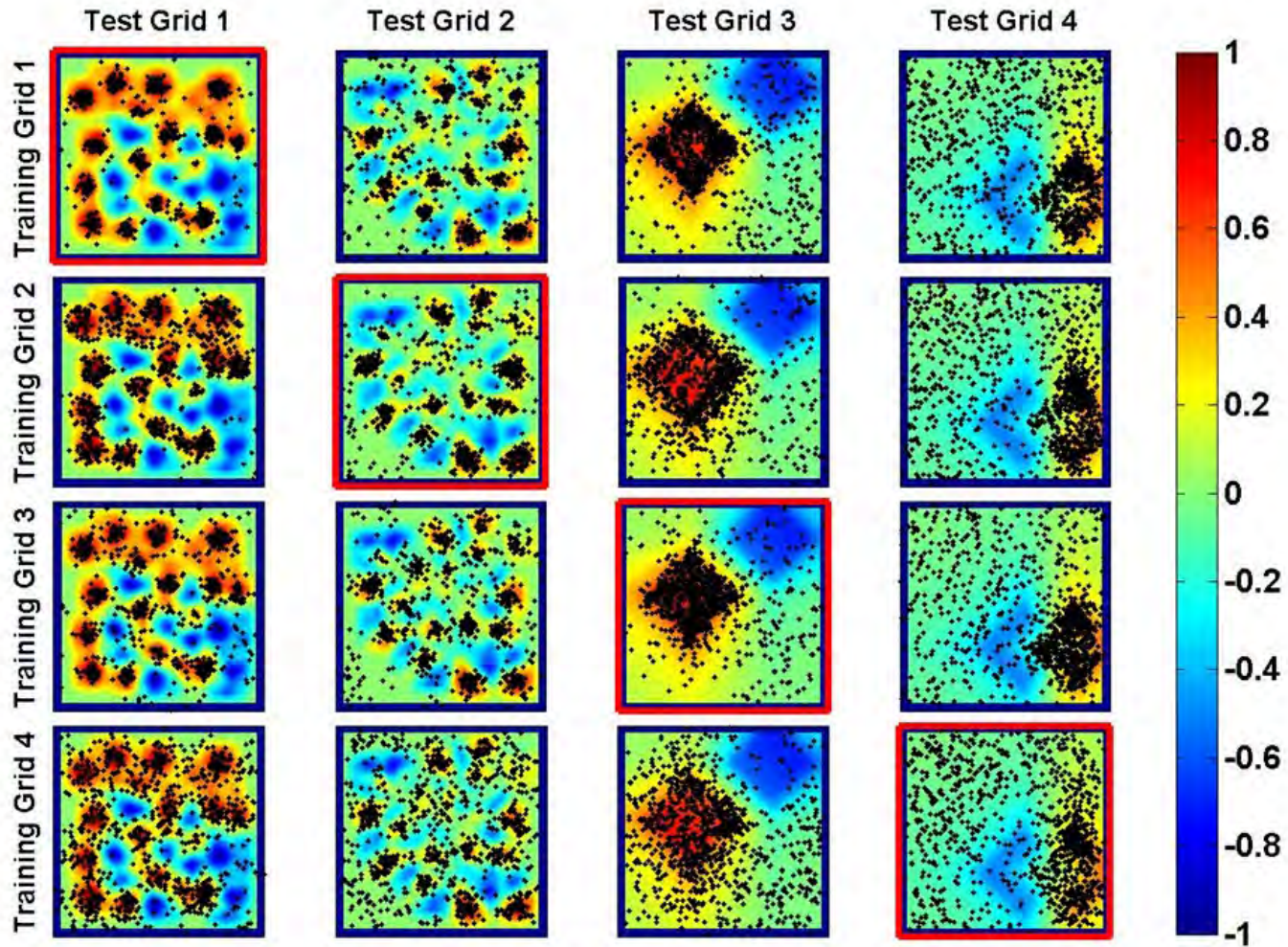
Last day
of 300th
generation



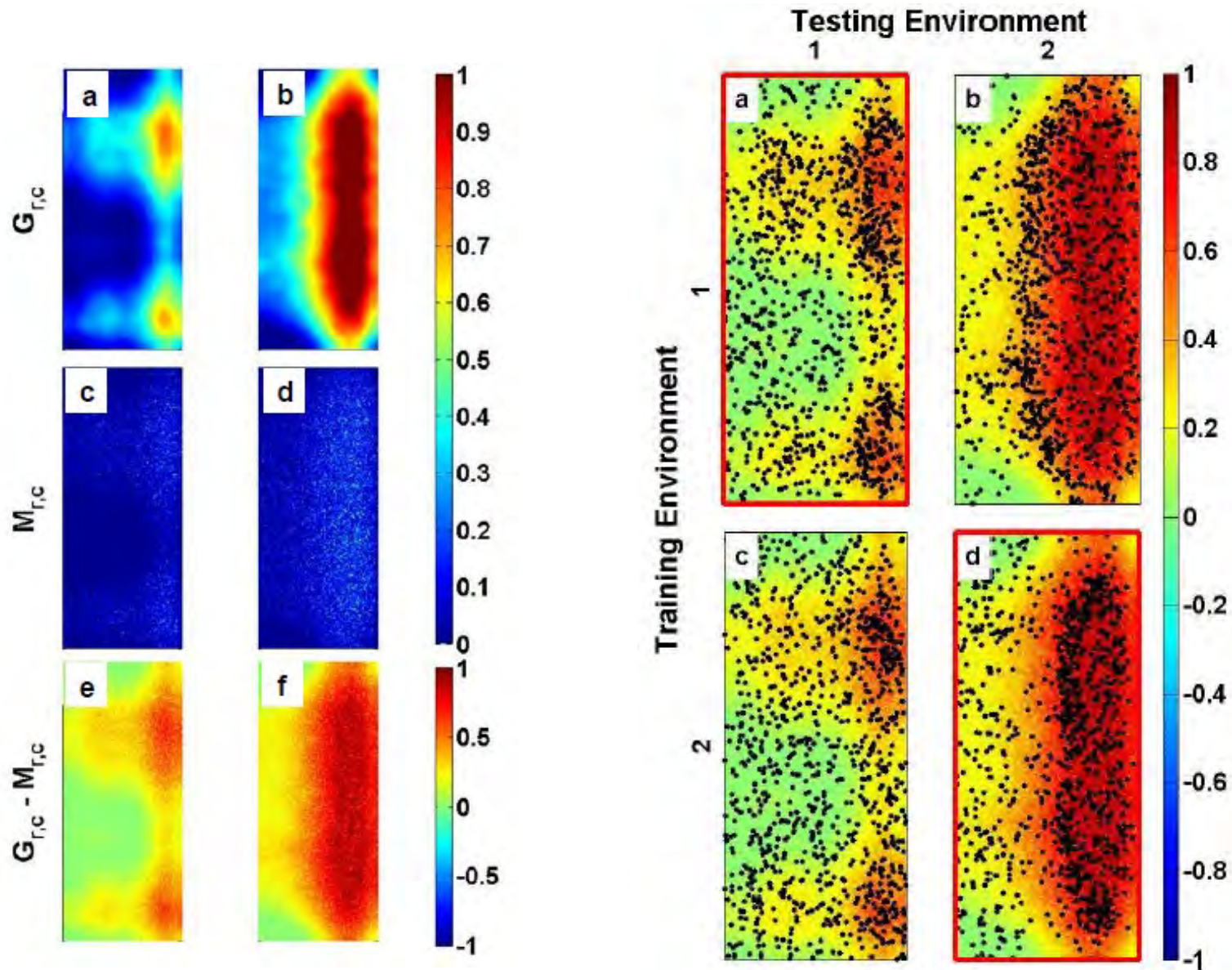
10 Individuals



Kinesis - Testing



Regional Ocean Grid



Enrique N. Curchitser
Rutgers University

Jerome Fiechter
University of California – Santa Cruz

Kate Hedstrom
Institute of Marine Science - University of Alaska

Miguel Bernal
FAO – Rome

Sean Creekmore
Louisiana State University

Alan Haynie
Alaska Fisheries Science Center - NOAA

Shin-ichi Ito
University of Tokyo

Bernard Megrey
Alaska Fisheries Science Center - NOAA

Chris Edwards
University of California – Santa Cruz

Dave Checkley
Scripps Institute of Oceanography

Tony Koslow
Scripps Institute – CALCOFI

Sam McClatchie
Southwest Fisheries Science Center - NOAA

Francisco Werner
Southwest Fisheries Science Center - NOAA

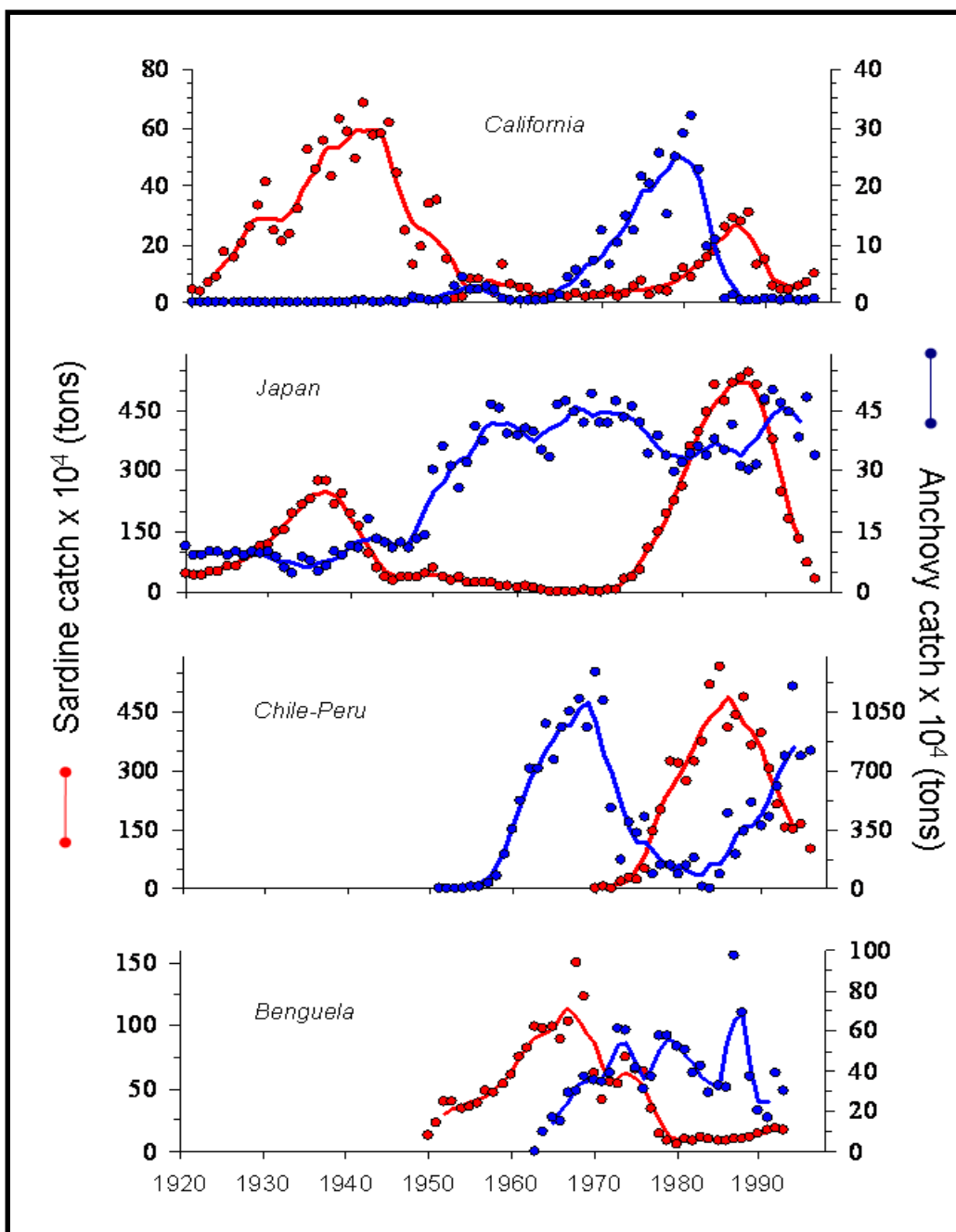
Alec MacCall
Southwest Fisheries Science Center - NOAA

Vera Agostini
Nature Conservancy

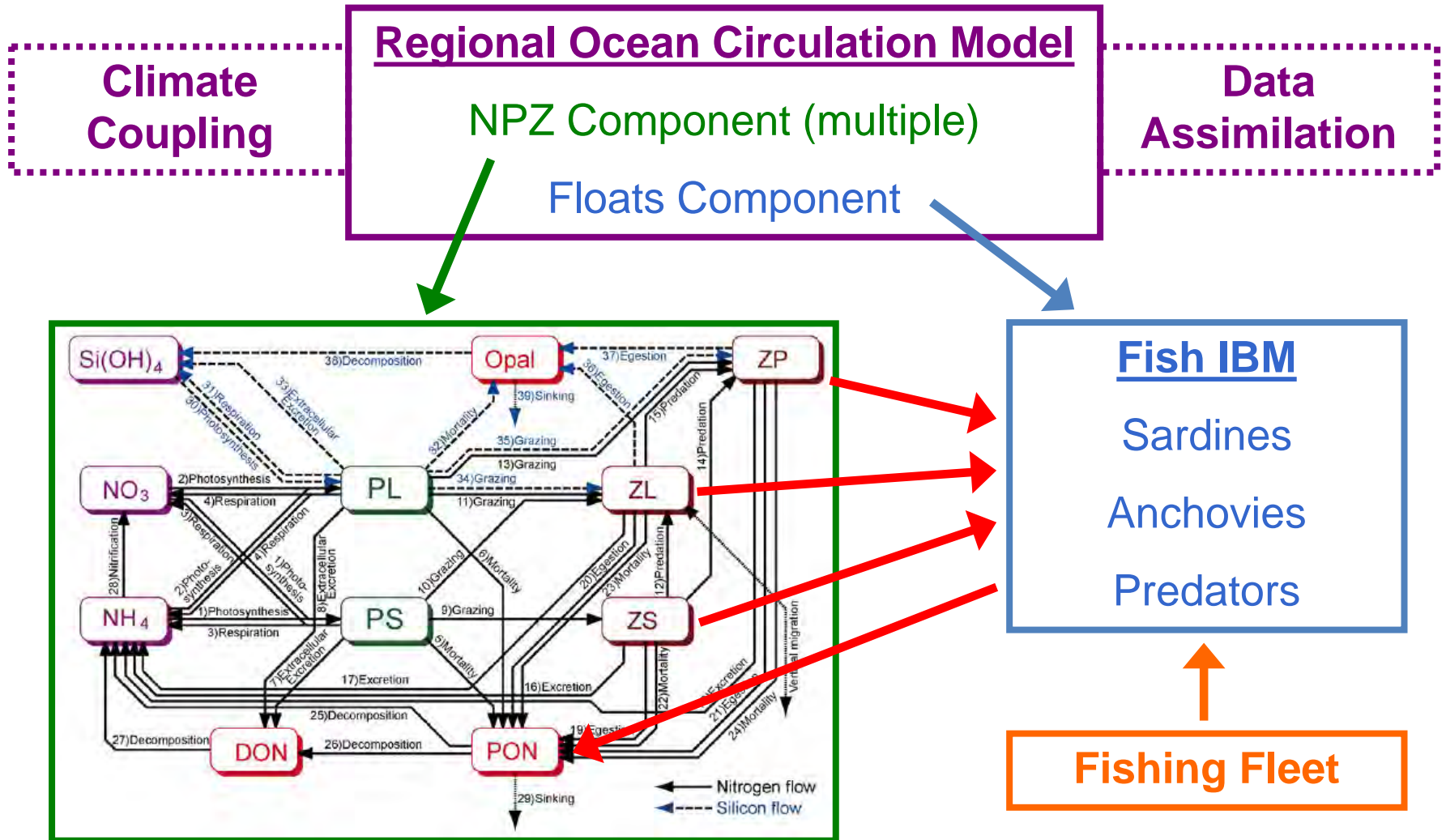
Rose et al. 2015. Demonstration of a fully-coupled end-to-end model for small pelagic fish using sardine and anchovy in the California Current. *Progress in Oceanography* 138: 348-380.

Fiechter et al. 2015. The role of environmental controls in determining sardine and anchovy population cycles in the California Current: Analysis of an end-to-end model. *Progress in Oceanography* 138: 381-398.





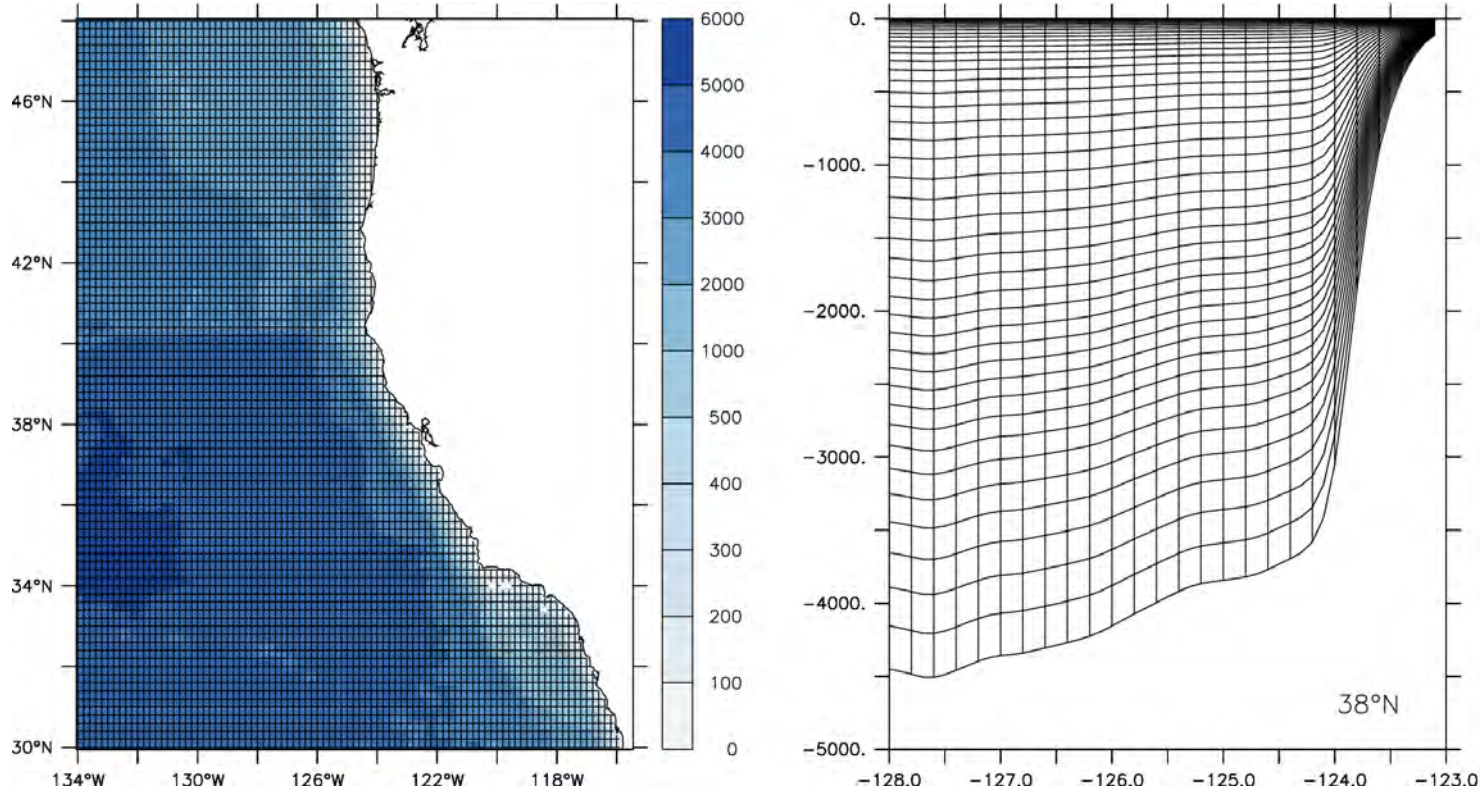
Fully-Coupled Model Within ROMS



Model 1: ROMS

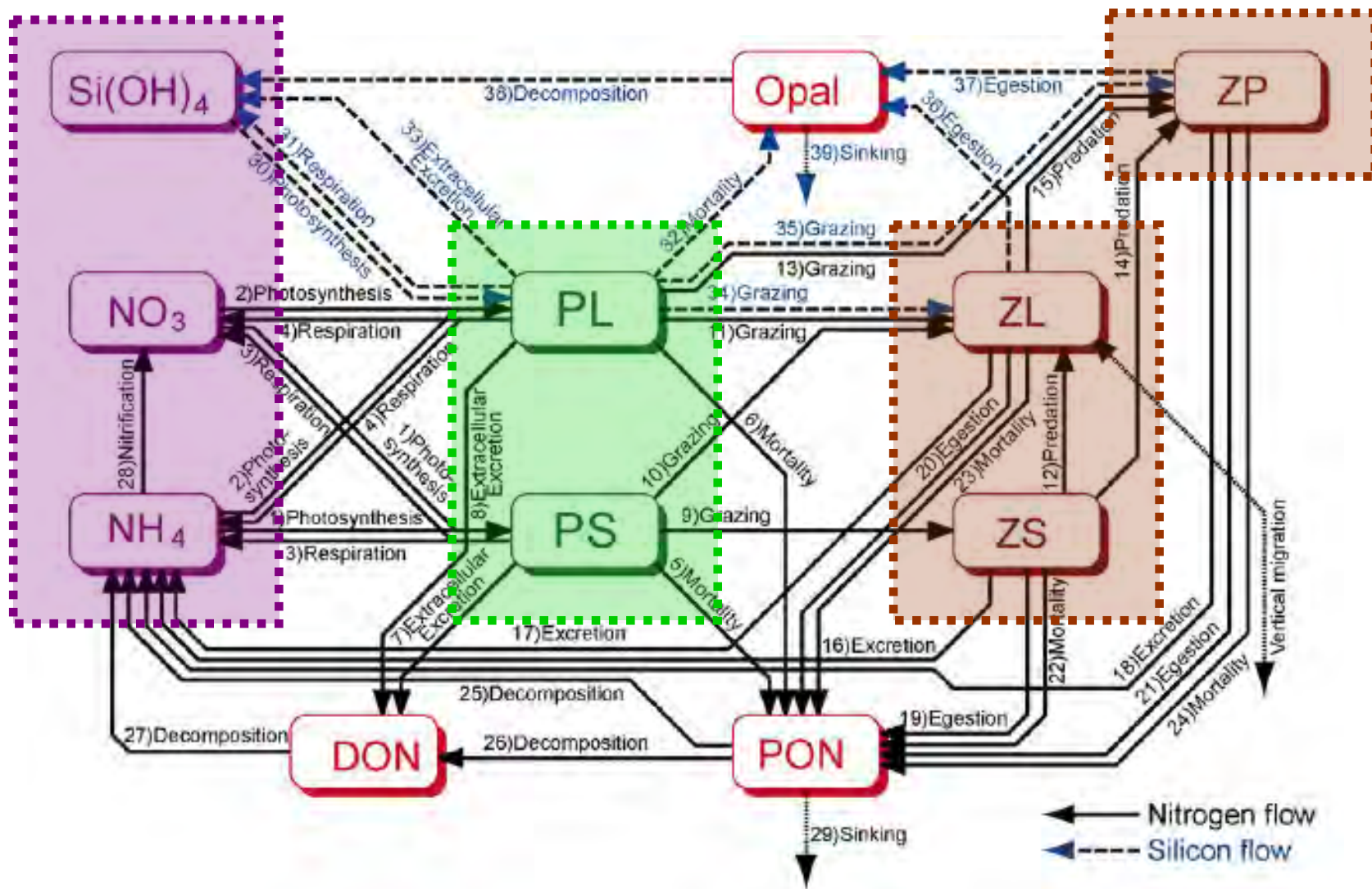
- Grid:
10 km
42 levels

- 900 s



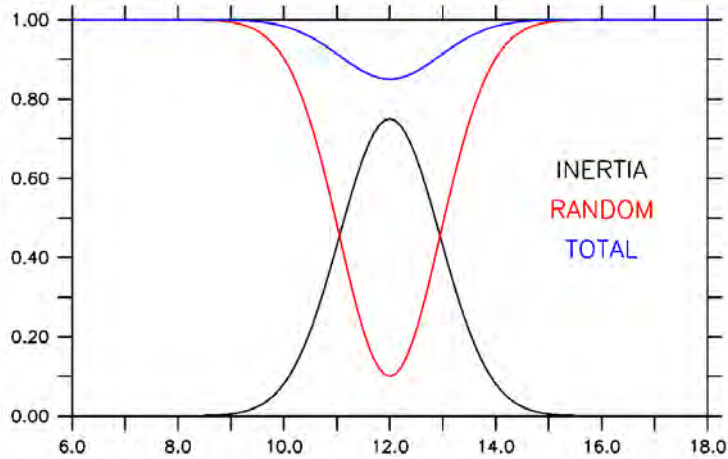
- Run duration: 50 years (1959-2009)

Model 2: NEMURO

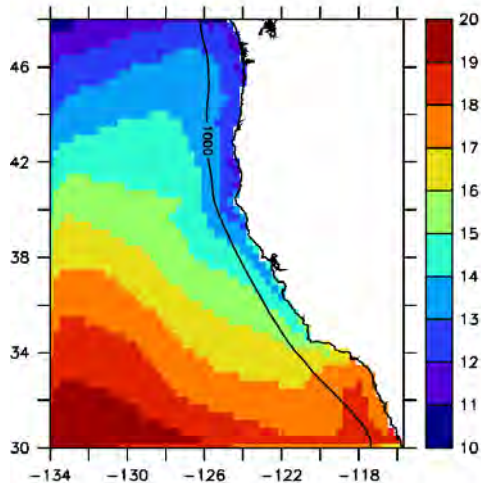
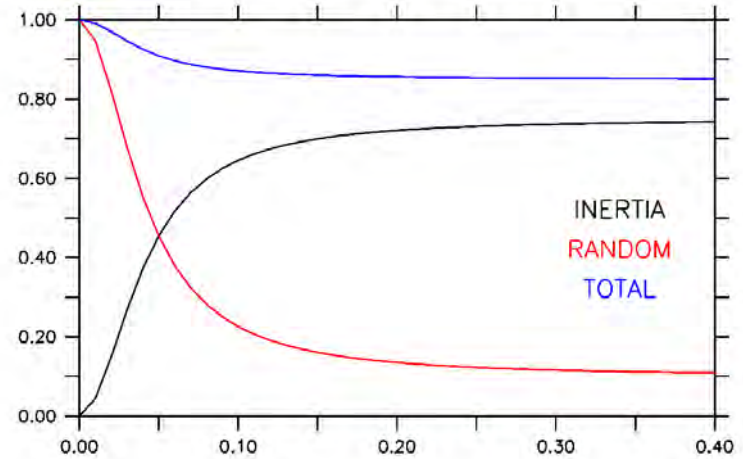


Environmental Cues for Movement (Kinesis)

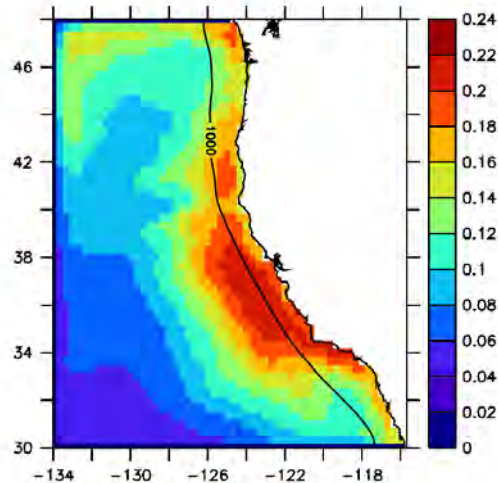
Temperature (°C) - Gaussian



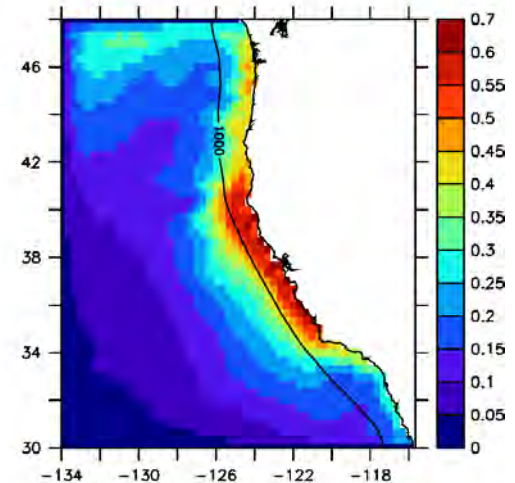
P-Value (C/C_{max}) - Holling Type III



Mean SST (1985-2005)



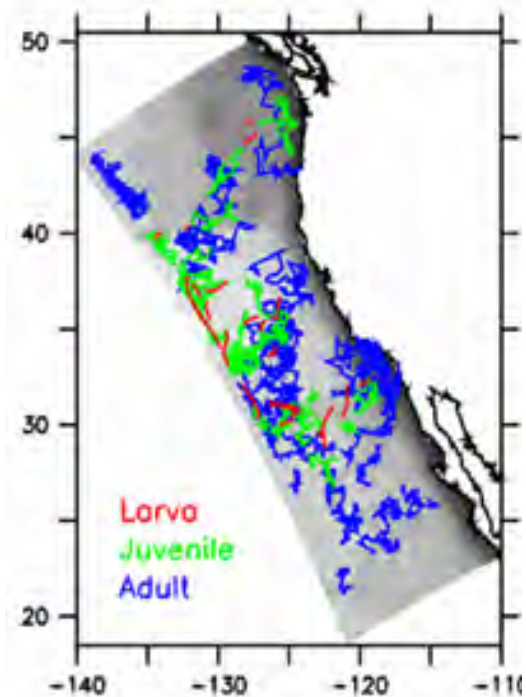
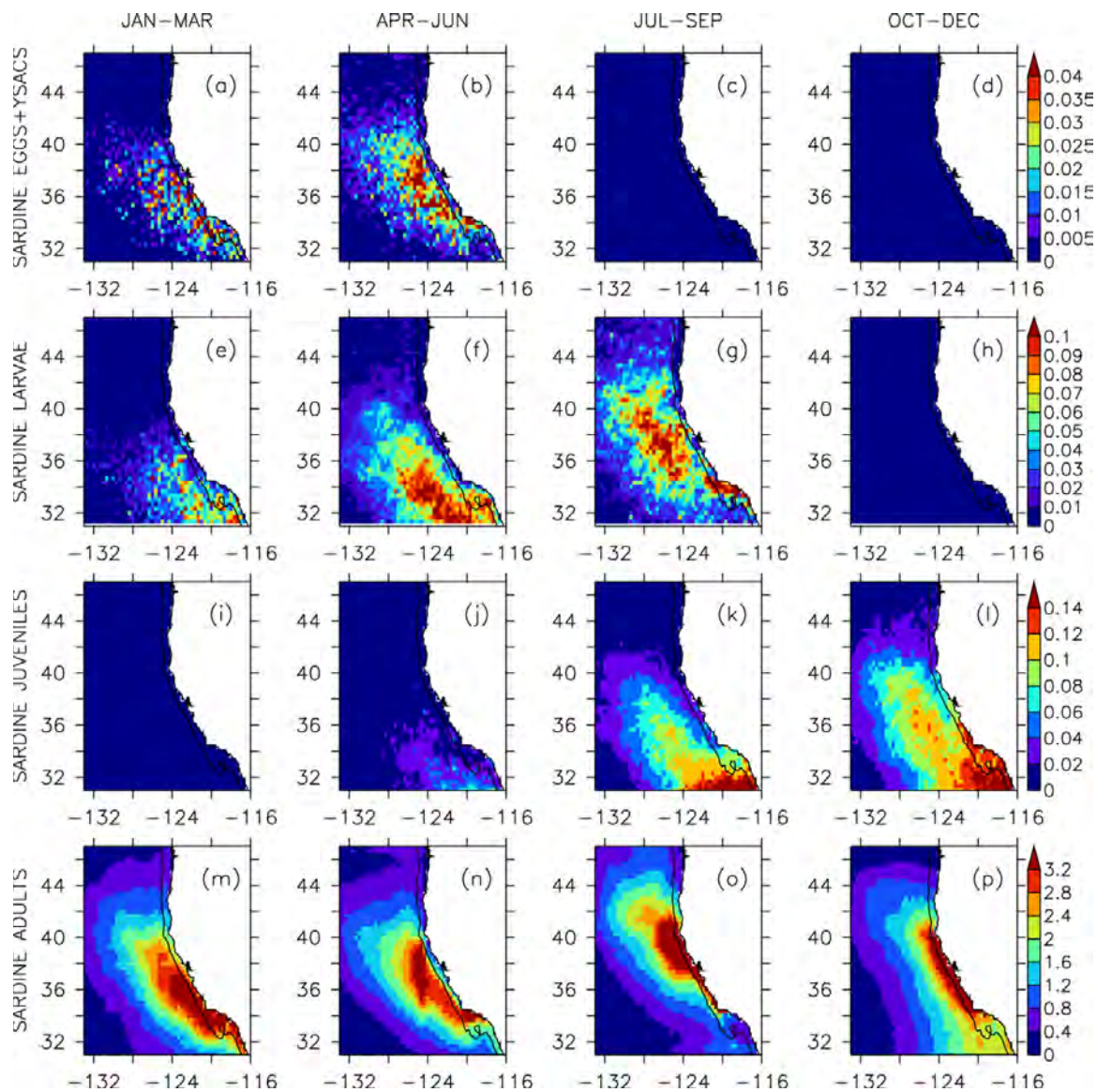
Mean ZS (1985-2005)



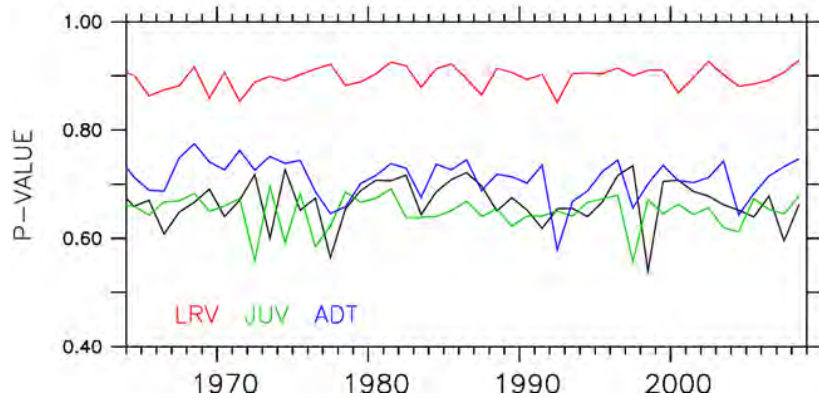
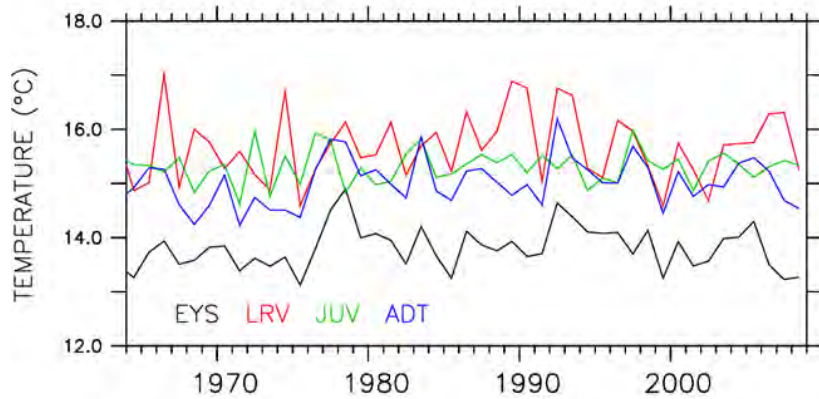
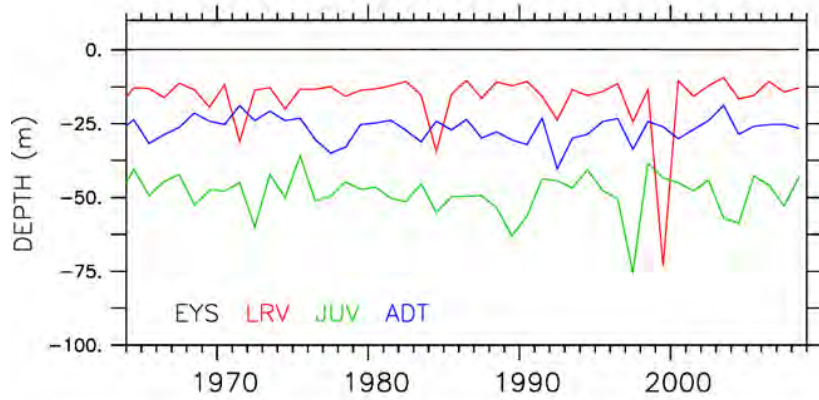
Mean ZL (1985-2005)

Sardine Spatial

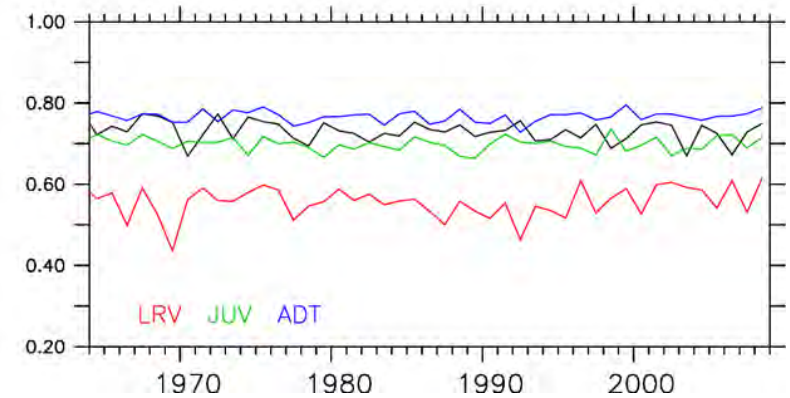
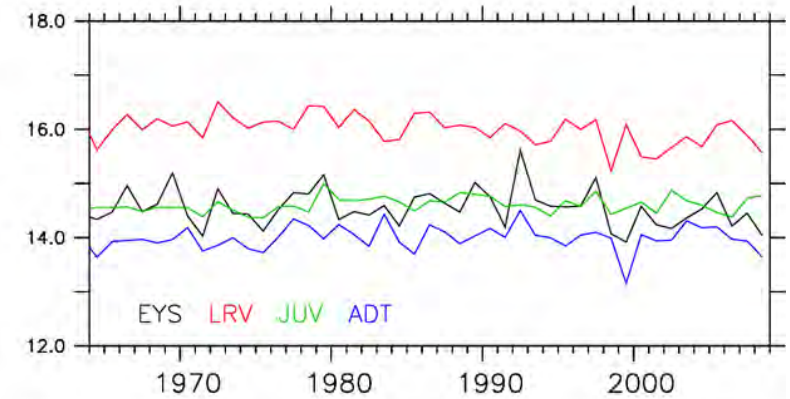
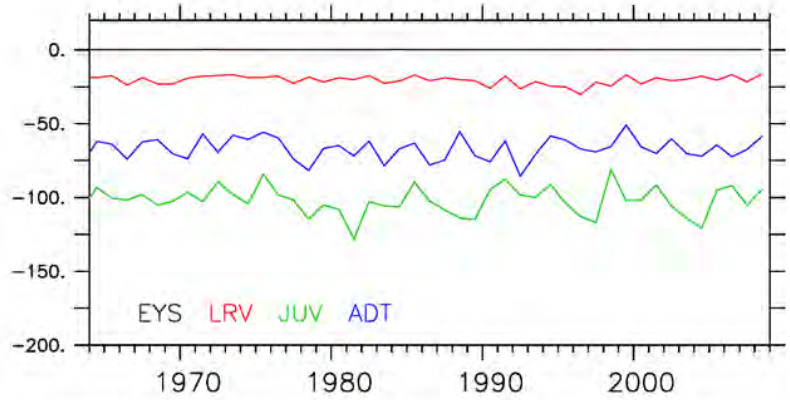
(E&YS – 10^{12} ; 1000 MT)

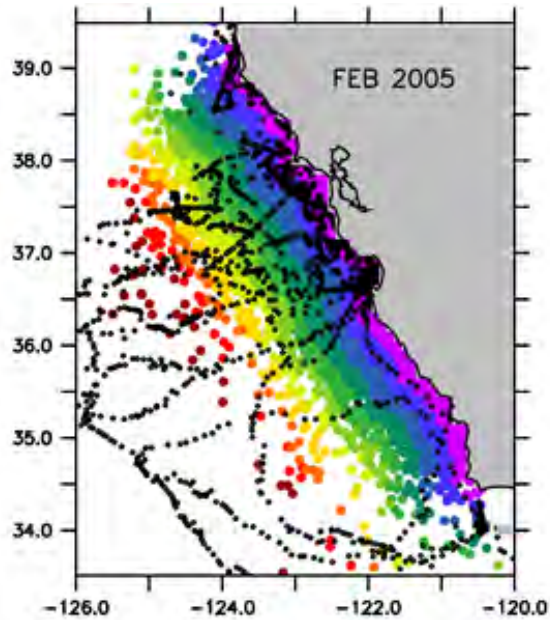


ANCHOVY

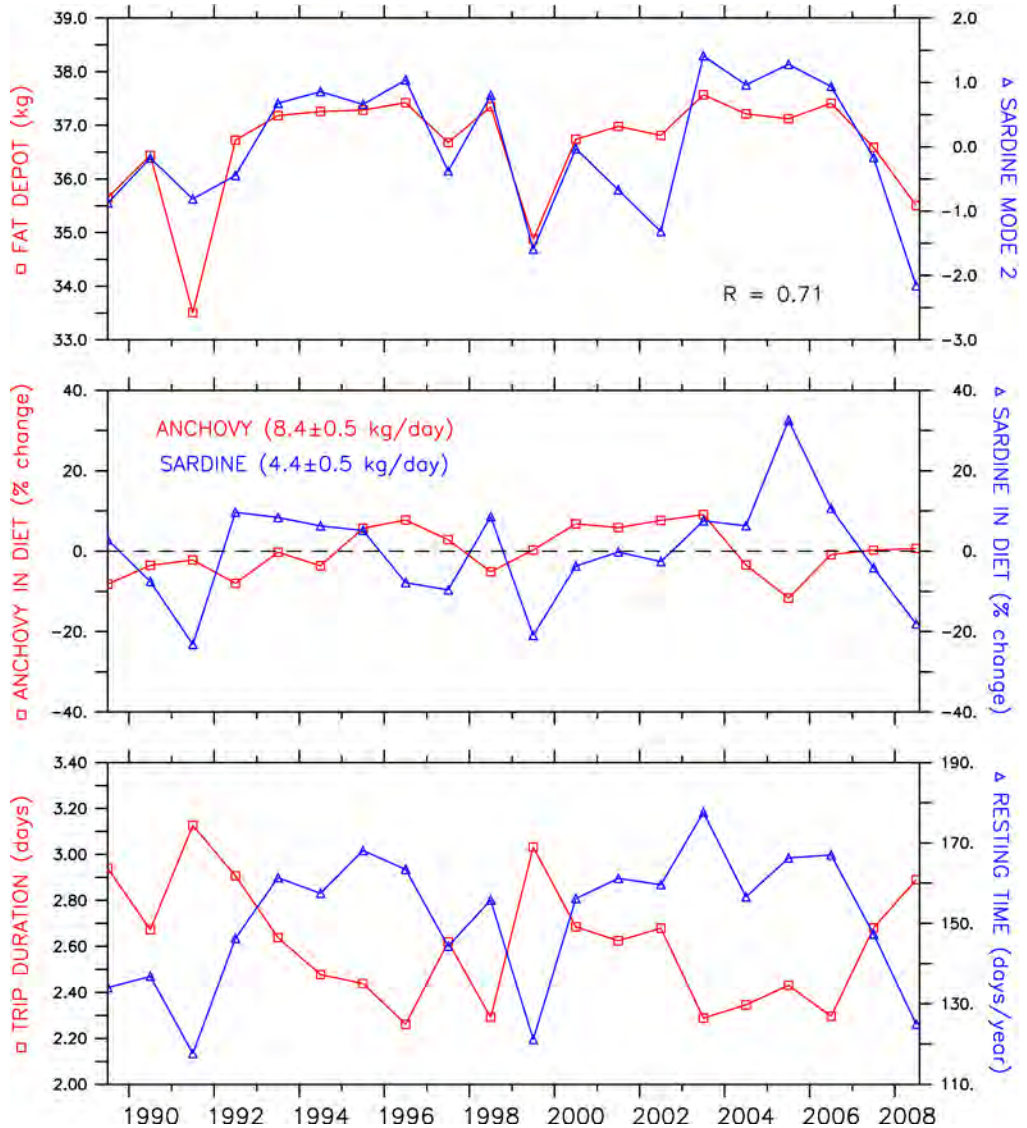


SARDINE





Fiechter et al. 2016. Marine Ecology Progress Series 556: 273-285.



Needs and Opportunities

- Merging of traditional and new data with modeling movement
- Standardization of movement algorithms
 - Description
 - Validation and testing: “out-of-sample”
- Best practices guidance

Needs and Opportunities

- Confidence in projected and ecological cues
- Two-way link to bioenergetics
- Richness of changes in spatial distributions into fisheries management