

US EPA ARCHIVE DOCUMENT

Individual-based Modeling for Salmonid Management

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<http://math.humboldt.edu/~ecomodel/>

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Credits

- Research collaborators:
 - Steve Railsback, Lang, Railsback, and Associates
 - Bret Harvey, USFS Redwood Sciences Lab
 - Software: Steve Jackson, Jackson Scientific Computing

Funding: EPA STAR Grant

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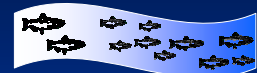
Presentation Objectives

- inSTREAM our individual-based trout model
- Advantages of IBMs for modeling fish population response to stressors
- Example applications of our stream trout IBM to management research and decision-making

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What is an IBM?

- A model of the environment +
- Models of individual animals
 - The mechanisms by which the environment affects an individual
 - The mechanisms by which individuals interact
 - The behaviors individuals use to adapt to their environment and each other
- Population responses that emerge from individual behaviors



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What is an IBM? Demo



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Why an Individual-based Model?

- IBM's resolve the two fundamental dilemmas of modeling:
 - Models usually assume many individual organisms can be described by a single variable like population size or biomass. **IBM's provide for individuals and their differences.**
 - Most models don't distinguish between organisms' locations. **IBM's provide for distinctive interactions with neighboring individuals and the local environment.**

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Advantages of IBMs for Modeling Fish Population Response to Stressors

- Complex, cumulative effects can be simulated:
 - Base flow
 - High and low flows: timing and magnitude
 - Temperature
 - Turbidity
 - Losses of individuals (angler harvest, diversion entrainment)
 - Food production
 - Reproduction, recruitment
 - Species interactions: competition, predation
 - ...

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Advantages of IBMs for Modeling Response to Stressors

- Complex, cumulative effects can be simulated:
- These complex interactions *emerge* from individual-level mechanisms
 - instead of having to be foreseen and built into a model
 - you just have to model how stressors affect *individuals*

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Advantages of IBMs

- IBMs are testable in many ways
 - They can produce many kinds of predictions that can be tested with many kinds of data
 - Habitat selection patterns over space, time, flow ...
 - Statistical properties of population (size, abundance)
 - Trends in abundance with environmental factors
 - etc.

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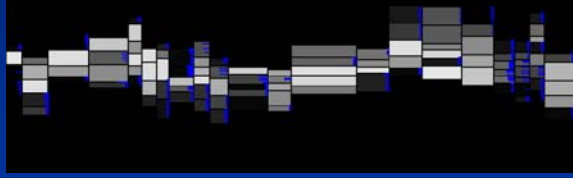
Advantages of IBMs

- IBMs provide a way out of the complexity - uncertainty dilemma:
 - A well-designed IBM is a collection of *simple* submodels for separate processes at the individual level
 - Each submodel can be parameterized and tested with all the information available for its process
 - Yet IBMs can simulate complex population level responses

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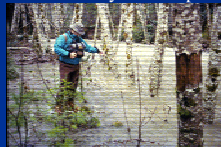
Stream Trout Model

- Habitat is modeled as rectangular cells
- External hydraulic model simulates how depth, velocity vary with flow



Stream Trout Model

- Habitat:
 - Water depths and velocities
 - Temperature, turbidity
 - Food availability
 - Daily time step
- Fish:
 - Habitat selection (choosing the best cell)
 - Feeding and growth
 - Mortality
 - Spawning & incubation



Feeding Model

- Drift feeding strategy
 - Food intake per fish:

$$\text{Food concentration} \times \text{velocity} \times \text{capture area.}$$

Capture area:

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Feeding Model

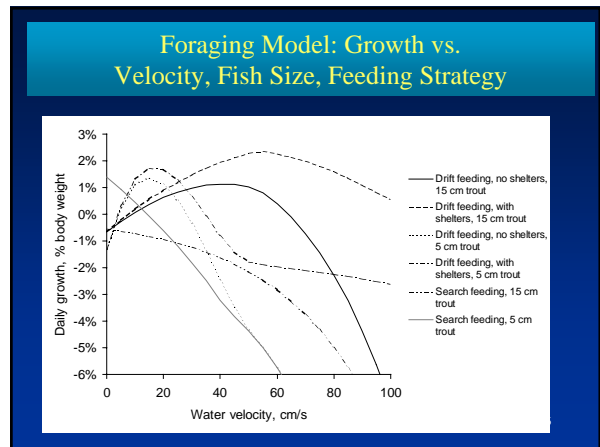
- Food intake varies between drift and search feeding strategies
 - Relative advantages depend on flow, fish size, habitat
- Food intake can be limited by competition (food consumed by bigger fish)
- Each fish picks the feeding strategy offering highest growth
 - Preferred strategy can vary among cells

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Growth Model (bioenergetics)

- Growth = Food intake - metabolic costs
 - Metabolic costs:
 - increase with swimming speed
 - increase with temperature

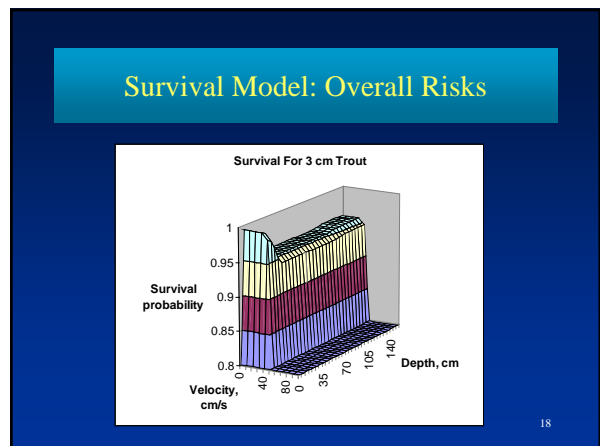
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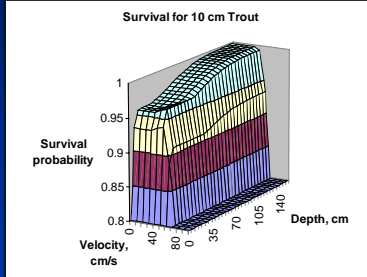
Survival Model

- Survival probabilities:
 - Vary with habitat
 - Depend on fish size, condition
 - Include:
 - Poor condition (starvation)
 - Terrestrial predation
 - Aquatic predation
 - High temperature
 - High velocity (exhaustion)
 - Stranding (low depth)

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Survival Model: Overall Risks



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Habitat Selection: Overview

- Habitat selection is critical:
 - Moving is the primary way fish adapt to changing conditions
- Our approach assumes fish use behaviors that evolved to maximize fitness

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Habitat Selection Rules

- Move to the cell that offers highest potential “fitness”
 - (within the radius that fish are assumed to be familiar with)
- Railsback, S. F., R. H. Lamberson, B. C. Harvey and W. E. Duffy (1999). Movement rules for spatially explicit individual-based models of stream fish. *Ecological Modelling* 123: 73-89.

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Habitat Selection: Fitness Measure

- Fish move to cell offering highest fitness
- Key elements of fitness are:
 - Future survival
 - Attaining reproductive size

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Habitat Selection: Summary *How a Fish Rates A Potential Destination Cell*

- Considers:
 - Potential growth in cell (function of habitat, competition)
 - Mortality risks in cell (function of habitat)
 - Its own size and condition
- Probability of surviving for 90 days in the cell?
 - Assuming today’s conditions persist for the 90 days
- How close to reproductive size after 90 d in the cell?
- Rating = Survival probability × fraction of reproductive size

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Habitat Selection

- Many realistic behaviors emerge:
 - Normal conditions: territory-like spacing
 - Short-term risk: fish ignore food and avoid the risk
 - Hungry fish take more chances to get food (and often get eaten)
 - Conditions like temperature, food availability, fish density affect habitat choice

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Analyzing Individual-based Models

- The “pattern-oriented” analysis approach:
 - Test specific processes of an IBM by whether it reproduces a wide range of behaviors that emerge from the process
 - Test a complete IBM by whether it reproduces a wide range of observed population-level patterns
- Railsback, S. F. (2001). Getting “results”: the pattern-oriented approach to analyzing natural systems with individual-based models. *Natural Resource Modeling* 14: 465-474.

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Pattern-Oriented Analysis of inStream

- Validation:
 - Individual level
 - Railsback, S. F. and B. C. Harvey (2002). Analysis of habitat selection rules using an individual-based model. *Ecology* 83: 1817-1830.
 - Population level
 - Railsback, S. F., B. C. Harvey, R. H. Lamberson, D. E. Lee, N. J. Claasen and S. Yoshihara (2002). Population-level analysis and validation of an individual-based cutthroat trout model. *Natural Resource Modeling* 15: 83-110.

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Validation of Habitat Selection Rules: Six Patterns (a)

- Feeding hierarchies
- Movement to channel margin during high flow
- Juveniles respond to competing species by using less optimal habitat (higher velocities)

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Validation of Habitat Selection Rules: Six Patterns (b)

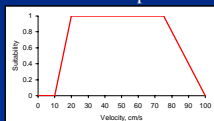
- Juveniles respond to predatory fish by using shallower, faster habitat
- Use of higher velocities in warmer seasons
- Habitat shift in response to reduced food

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Expected Reproductive Maturity vs. Habitat Suitability Criteria as Indicators of Habitat Quality

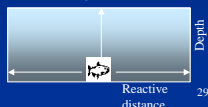
- PHABSIM habitat suitability criteria (HSC)

– Basis: Empirical



- Expected Reproductive Maturity (EM)

– Basis: Mechanistic models of feeding, mortality risks, fitness



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EM vs. HSC Indicators of Habitat Quality

- HSC

Habitat rating varies only with fish life stage:

fry, juvenile, adult, spawning

(occasionally: season)

- EM

Habitat rating varies with:

- ↗ Fish size
- ↗ Fish condition
- ↗ Temperature & season
- ↗ Food availability
- ↗ Cover for hiding, feeding
- ↗ Other factors affecting growth or survival

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EM as an Indicator of Habitat Quality

- Adult trout
 - drift feeding
 - using velocity shelter

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EM as an Indicator of Habitat Quality: With vs. Without Velocity Shelters for Drift Feeding

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EM as an Indicator of Habitat Quality: Without vs. With Hiding Cover

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EM as an Indicator of Habitat Quality: 15° vs. 5° Temperature

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EM as an Indicator of Habitat Quality: Low vs. High Turbidity

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Example Use of IBM for Management Research: Effect of Habitat Complexity on Population Dynamics

- Observed pattern: When deep pools are eliminated, a lower abundance of large trout results:
 - Bisson & Sedell (1984) observed fewer pools & fewer large trout in clearcuts
- Simulation experiment:
 - Simulate populations over 5 years with, without pool habitat in the model
 - Railsback, S. F., B. C. Harvey, R. H. Lamberson, D. E. Lee, N. J. Claassen and S. Yoshihara (2002), Population-level analysis and validation of an individual-based cutthroat trout model. *Natural Resource Modeling* 15: 83-110.

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Effect of Habitat Complexity on Population Dynamics

- Simulation results (1):
 - Abundance of all age classes was lower when pools were removed
 - Impact was greatest on oldest age class
 - Terrestrial predation caused the lower abundance - pools provide shelter from terrestrial predators

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Effect of Habitat Complexity on Population Dynamics

- Simulation results (2):
 - Size of age 0 and 1 trout *increased* when pools were removed -
 - Why??

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Effect of Habitat Complexity on Population Dynamics

- Simulation results (2):
 - Size of age 0 and 1 trout *increased* when pools were removed -
 - Abundance decreased, so there was less competition for food
 - Age 1 trout were forced to use faster, shallower habitat where predation risk is higher BUT food intake and growth is higher

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Example IBM Application: Effects of Instream Flow Magnitude & Variability

- How does the amount *and timing* of flow affect trout *abundance* and *growth*?
- Site: Little Jones Creek (3rd order coastal stream in N. California)
- Scenarios: hypothetical hydropower reservoir
 - Constant flow vs. Natural monthly mean flow
- Simulations: 10 years, 5 replicates per scenario

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Example IBM Application: Effects of Instream Flow Magnitude & Variability

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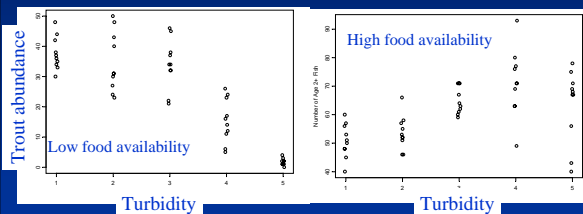
Example Application: Effects of Turbidity

- Turbidity decreases feeding ability, but decreases predation risk
What are the population-level consequences?
- Site: Little Jones Creek
- Five turbidity scenarios:
 - Turbidity = $x Q$
 - Five values of x : very clear to very turbid streams

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Example Application: Effects of Turbidity

- Result: Interactions between turbidity and food availability are strong



Example Use of IBM for Management Research: Habitat Selection vs. Habitat Quality

- Theory to be tested: The habitat that animals use most often is the best habitat
 - This assumption is the basis for many management models
 - It is widely questioned but very difficult to test in the field
- "Relations between habitat quality and habitat selection in a virtual trout population." Railsback, S. F., H. B. Stauffer, and B. C. Harvey. (to appear in *Ecological Applications*.)

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Habitat Selection vs. Habitat Quality

- "Habitat Selection" = the observed choice of habitat
- DEN is evaluated as observed animal density

$$DEN = (\# \text{ animals using a habitat type}) / (\text{area of habitat type})$$

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Habitat Selection vs. Habitat Quality

- "Habitat Quality" or Fitness Potential (FP) = the fitness provided to an animal by a habitat type, in the absence of competition
 - "Preference"; the habitat a fish selects in absence of competitors
- In our IBM:
 - We know the FP of each habitat cell because we programmed it
 - FP varies among habitat cells with water depth, velocity, feeding shelter, hiding shelter

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Habitat Selection vs. Habitat Quality

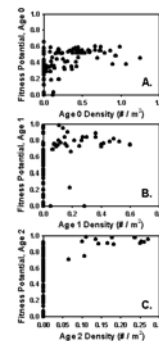
- The experiment:
 - Observe DEN (fish density) in each habitat cell (snapshot)
 - Calculate FP for each cell
 - Examine: How well does DEN predict FP? (What can you learn about the quality of habitat by observing the habitat that animals use?)
 - Three ages of trout examined separately

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What Does Habitat Selection Tell You about Habitat Quality??

Not much!

- Cells with high density usually are fairly high quality
- Many high quality cells have zero fish
- There is no predictive relationship between observed fish density and habitat quality



Management Research with the IBM:
 Why is There So Little Relation Between
 Habitat Selection and Habitat Quality?

- (1) Competition:
 - Smaller trout don't use the habitat that is best for them because they are excluded by larger fish

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Why is There So Little Relation Between
 Habitat Selection and Habitat Quality?

- (2) Unused and unknown habitat:
 - Good habitat for large trout may be vacant because there are not enough trout to use it all
 - Trout may not use the best available habitat because it is too far away to know about

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Why is There So Little Relation Between
 Habitat Selection and Habitat Quality?

- (3) Cells where food is plentiful but hard to catch can support more fish at lower fitness:
 - Example: Cells with high velocity
 - Each fish can catch less food than optimal
 - Because each fish gets less of the food, more fish can share the cell



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Why is There So Little Relation Between
 Habitat Selection and Habitat Quality?

- (4) Cells where food is plentiful but mortality risks are high can support more fish at lower fitness:
 - Density is high because there is plenty of food *but*
 - Fitness is low because mortality risk are high

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Habitat Selection vs. Habitat Quality

- Conclusions:
 - Observed patterns of habitat selection by animals tell us little about how good the habitat is
 - But does this mean models based on habitat selection are worthless??

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Are There Problems with Models Based on
 Habitat Selection?

- A second simulation experiment:
 - A *good* habitat selection model can be a useful predictor of population response over short times
 - *When* habitat modifications are small
 - *And* it is a dominant species or life stage
 - BUT:
 - Habitat selection models have fundamental problems (mainly: neglecting that habitat selection varies over time)

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Conclusions:
Key advantages of IBMs for assessing impacts of multiple stressors on fish

- IBMs can be used to address more questions that are difficult to address with other modeling approaches
- IBMs can be more credible than alternatives
 - More testable
 - Able to simulate complex responses to many stressors without high parameter uncertainty

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Conclusions:
Potential Limitations of IBMs

- Computation: There is a limit to how many fish / how much habitat we can simulate (overcome with bigger computers, clusters?)
- Models for new groups of fish can be expensive to build
- Expertise: Few biologists are familiar with IBMs (or the mechanistic, individual-based view of ecology)
- Acceptance by managers: IBMs are unfamiliar, not as simplistic as alternative approaches
- We haven't done anadromy yet (but have put a lot of work into concepts and software)


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Conclusions: Our Status

- Continued evolution, application of the trout model
 - Diel shifts in habitat & activity: feeding vs. hiding
 - Sub-daily time steps and fluctuating flows
- Interest in new applications of our salmonid IBMs
 - Instream flow assessment
 - Assessment of restoration activities ...
 - Regional stressor-response applications
- Development of new models (juvenile Colorado pikeminnow)
- Development & publication of theory & software

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