

"Improving models of Pacific salmonids: drawing upon experience from non-salmonid systems"

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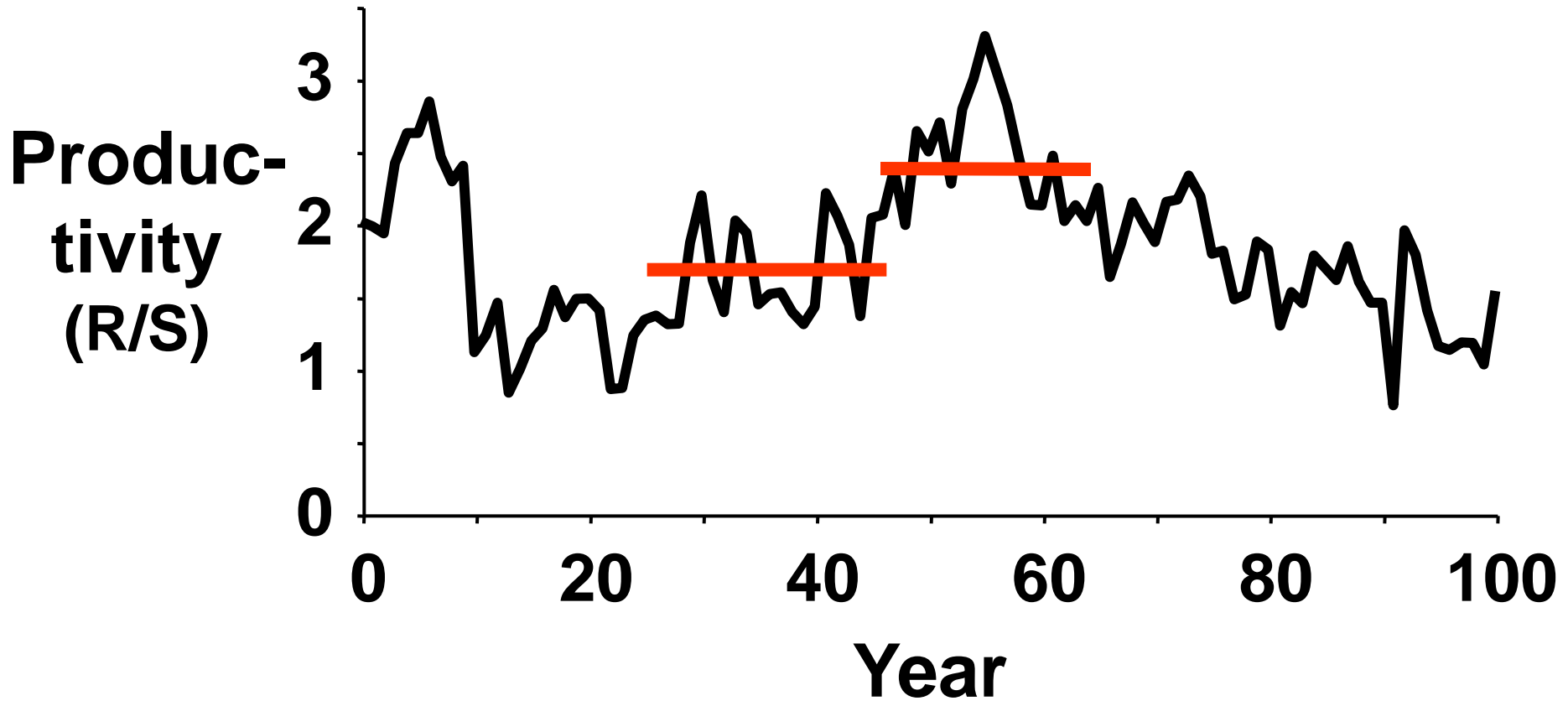


NPAFC Symposium, "International Symposium on Pacific Salmon and Steelhead Production in a Changing Climate: Past, Present, and Future",
Kobe, Japan, 17-19 May 2015

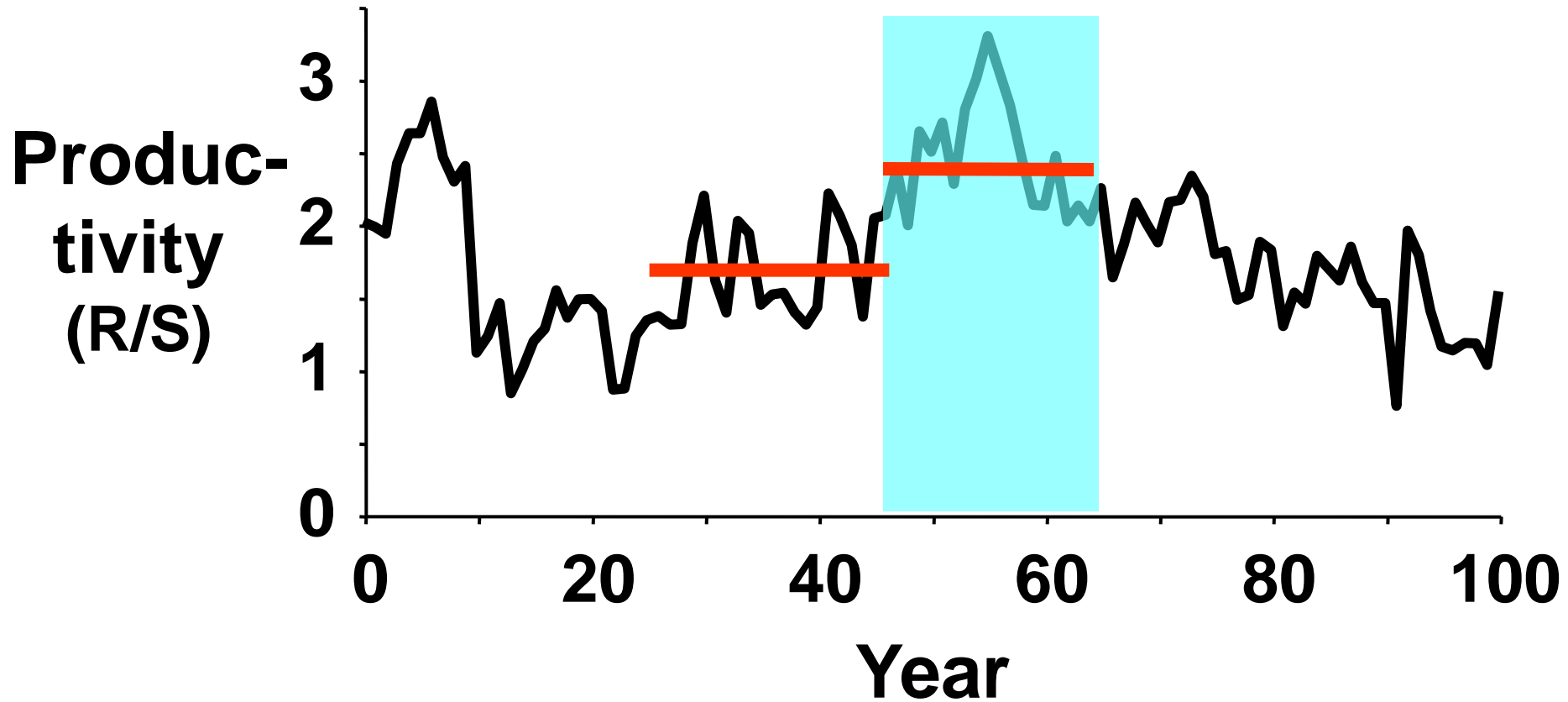
Outline

- **Importance of changing productivity of salmon**
- **How quantitative models can help**
- **Features of advanced models**

Climate-driven non-stationarity



Climate-driven non-stationarity

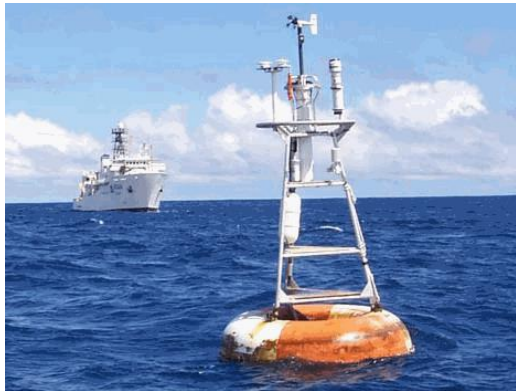
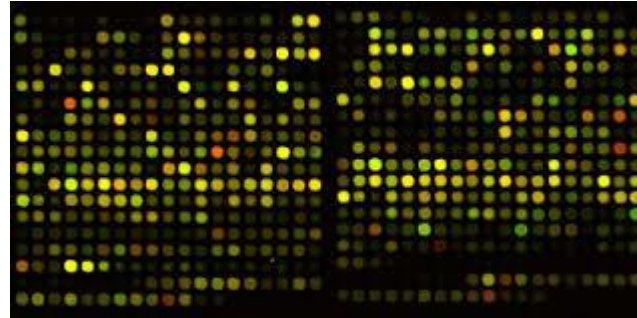
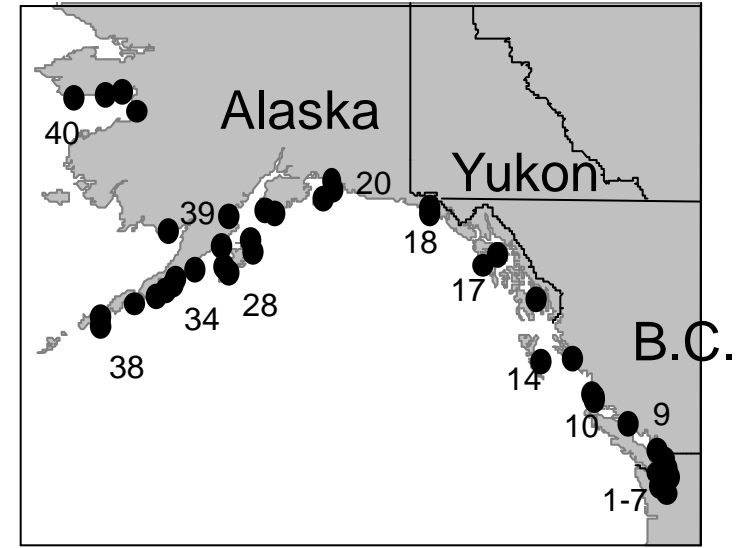
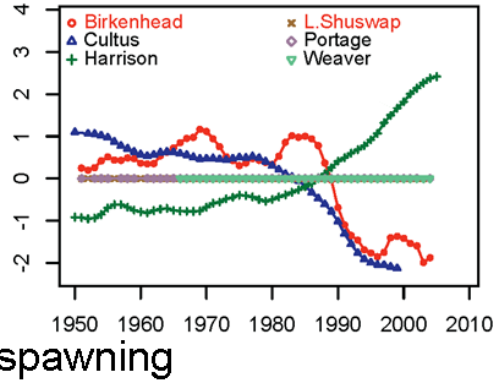
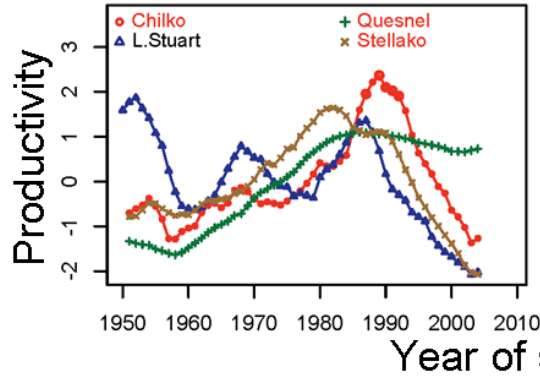


Given changing productivity, how do we choose

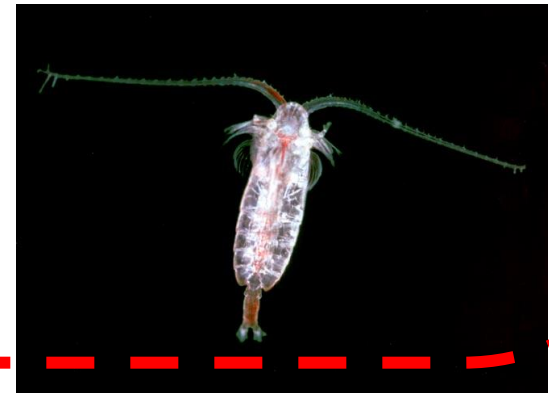
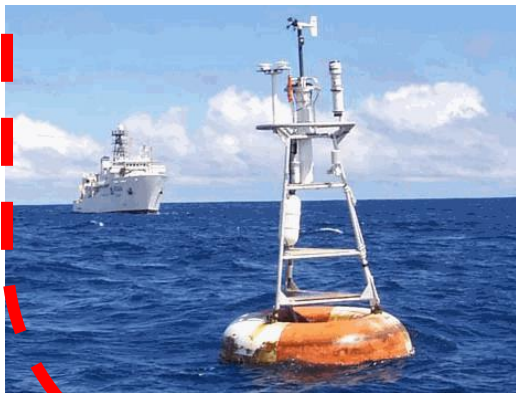
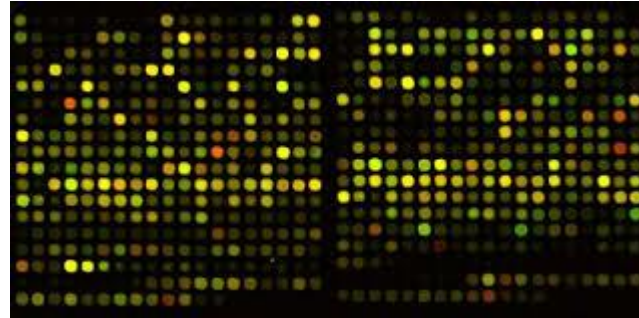
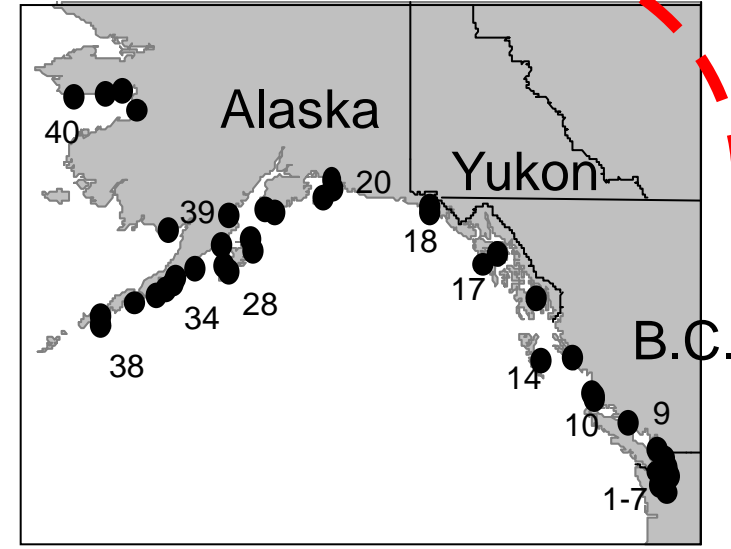
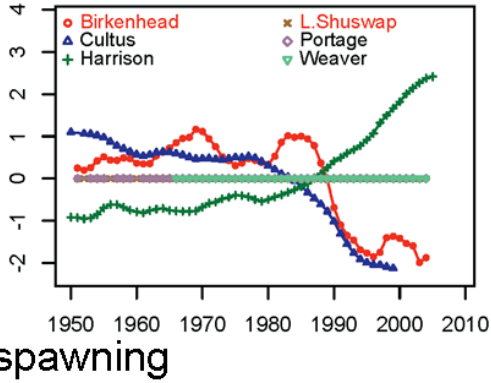
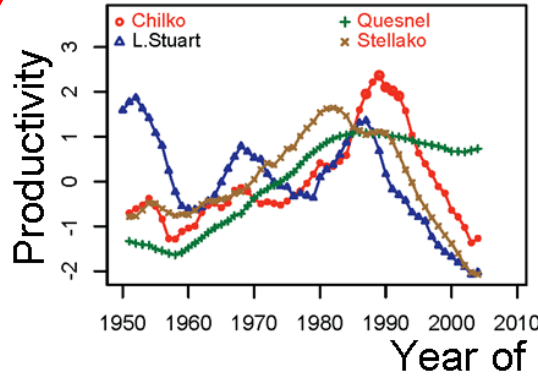
- Fishing regulations?
- Hatchery release strategies?
- Habitat protection activities?

Answers: from data and models

Salmon and related data

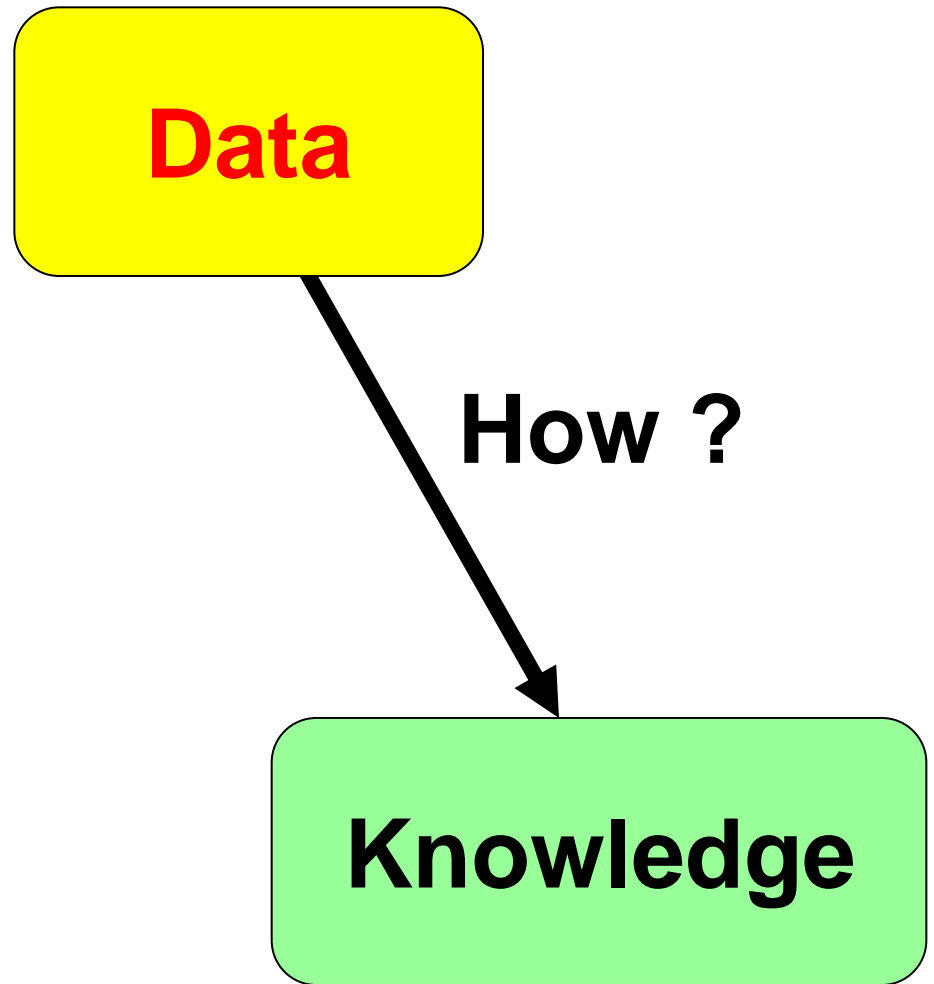


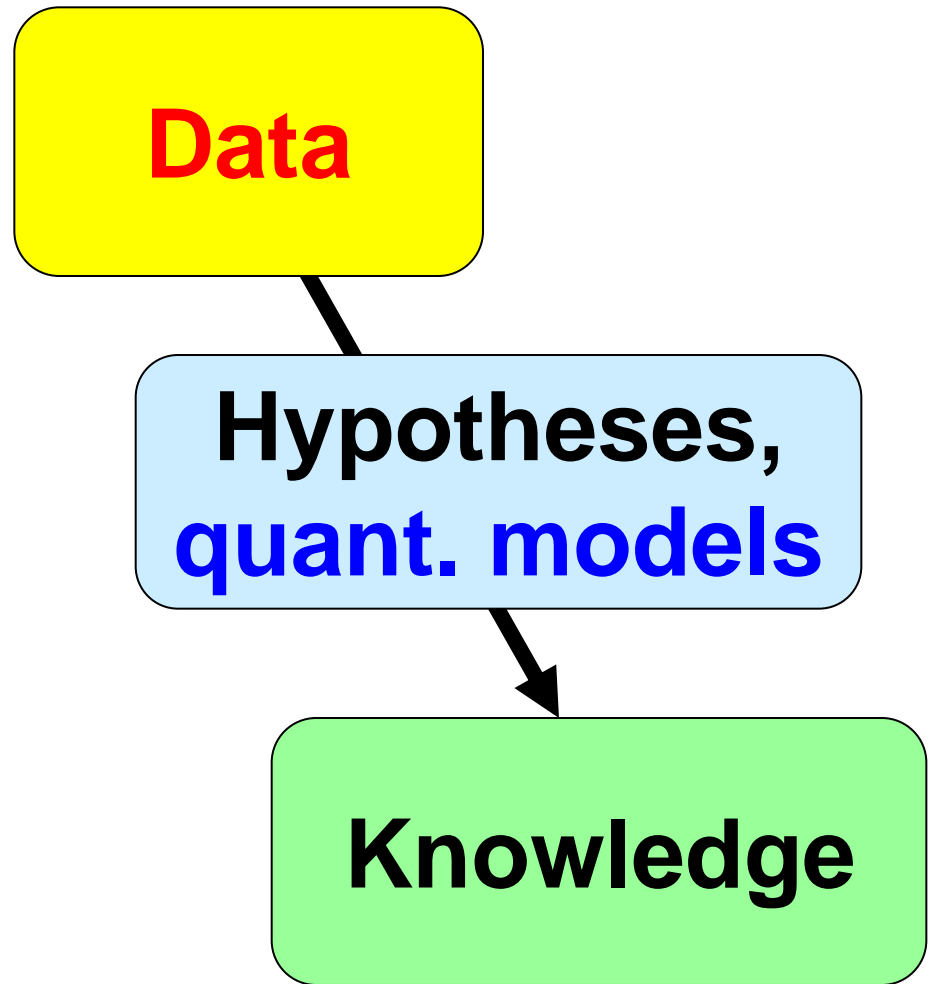
Salmon and related data





Data





How can **quantitative salmon models** help in presence of climate change?

- Improve understanding
- Forecast abundance
- Rank alternative management options

Emerging new modeling methods ...

Quantitative models

```
graph TD; A[Quantitative models] --> B[Statistical]; A --> C[Simulation];
```

Statistical

Simulation

Quantitative models

```
graph TD; QM[Quantitative models] --> S[Statistical]; QM --> Sim[Simulation]; S --> NEP[No explicit processes]; NEP --> FTSM[1. Formal time series models (auto-regressive AR1, ARIMA, ...)];
```

Statistical

Simulation

No explicit processes

1. **Formal time series models**
(auto-regressive AR1, ARIMA, ...)

Quantitative models

```
graph TD; QM[Quantitative models] --> S[Statistical]; QM --> Sim[Simulation]; S --> NEP[No explicit processes]; S --> WP[With processes];
```

Statistical

No explicit processes

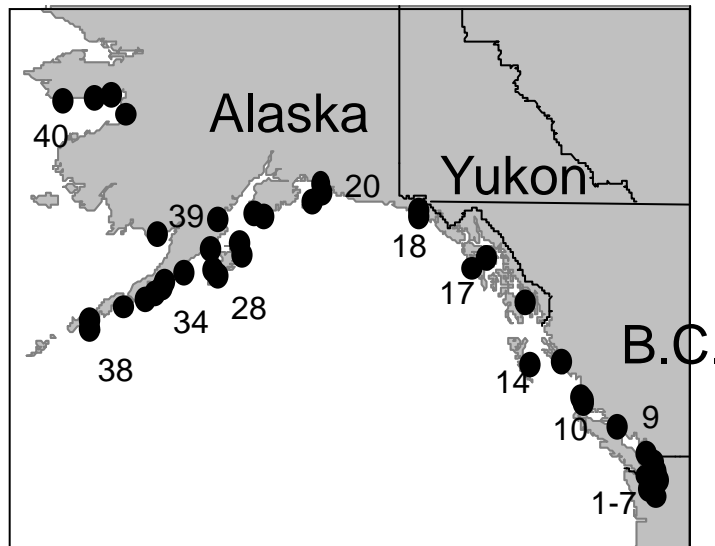
With processes

Simulation

1. Ricker models with environmental covariates
2. Hierarchical models for data across many popul.
3. State-space models (e.g., Kalman filter)
4. Dynamic Factor Analysis (DFA)

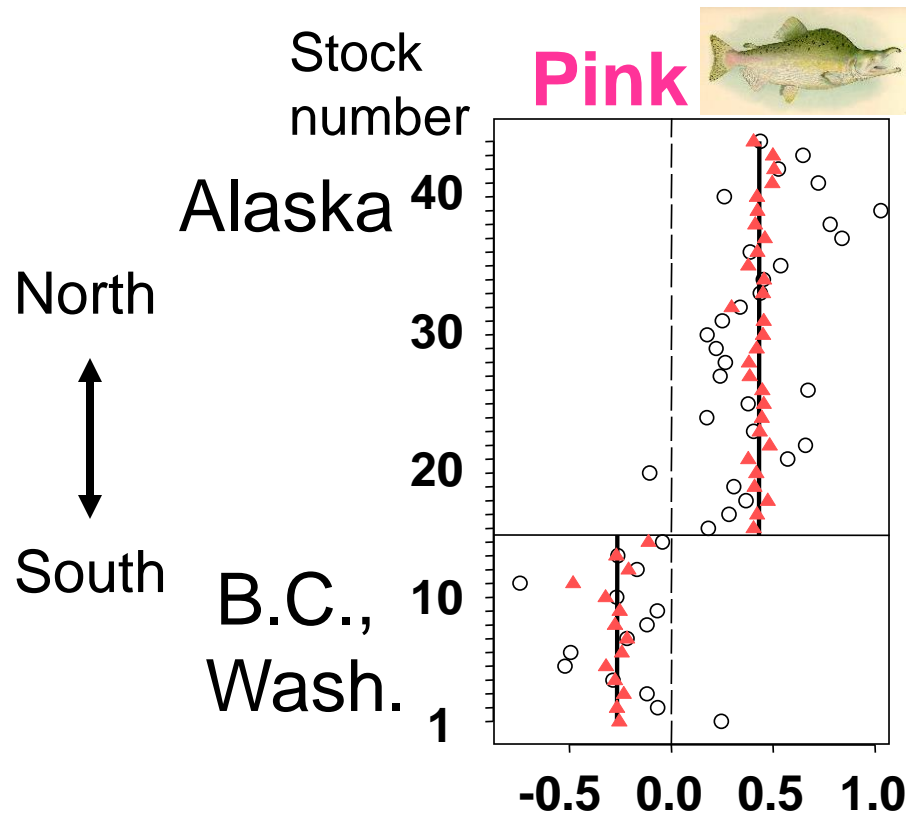
2. Hierarchical models

- Use data from multiple salmon stocks ("replicates")
- Estimate variation shared among salmon stocks



Time series
for 43 pink
salmon stocks





- **Single-stock**
- ▲ **Multi-stock, hierarchical model**

Change in salmon productivity, $\log_e(R/S)$, per $^{\circ}\text{C}$ increase in stock-specific summer SST

- **Similar results for chum and sockeye**
- **Northern populations of North American salmon may benefit as SST increases, but southern ones may lose.**

Compared to single-stock analyses,
multi-stock hierarchical models:

- Improve understanding of environmental effects (like SST) shared across salmon populations
- But stock-specific estimates will be biased (shrinkage)

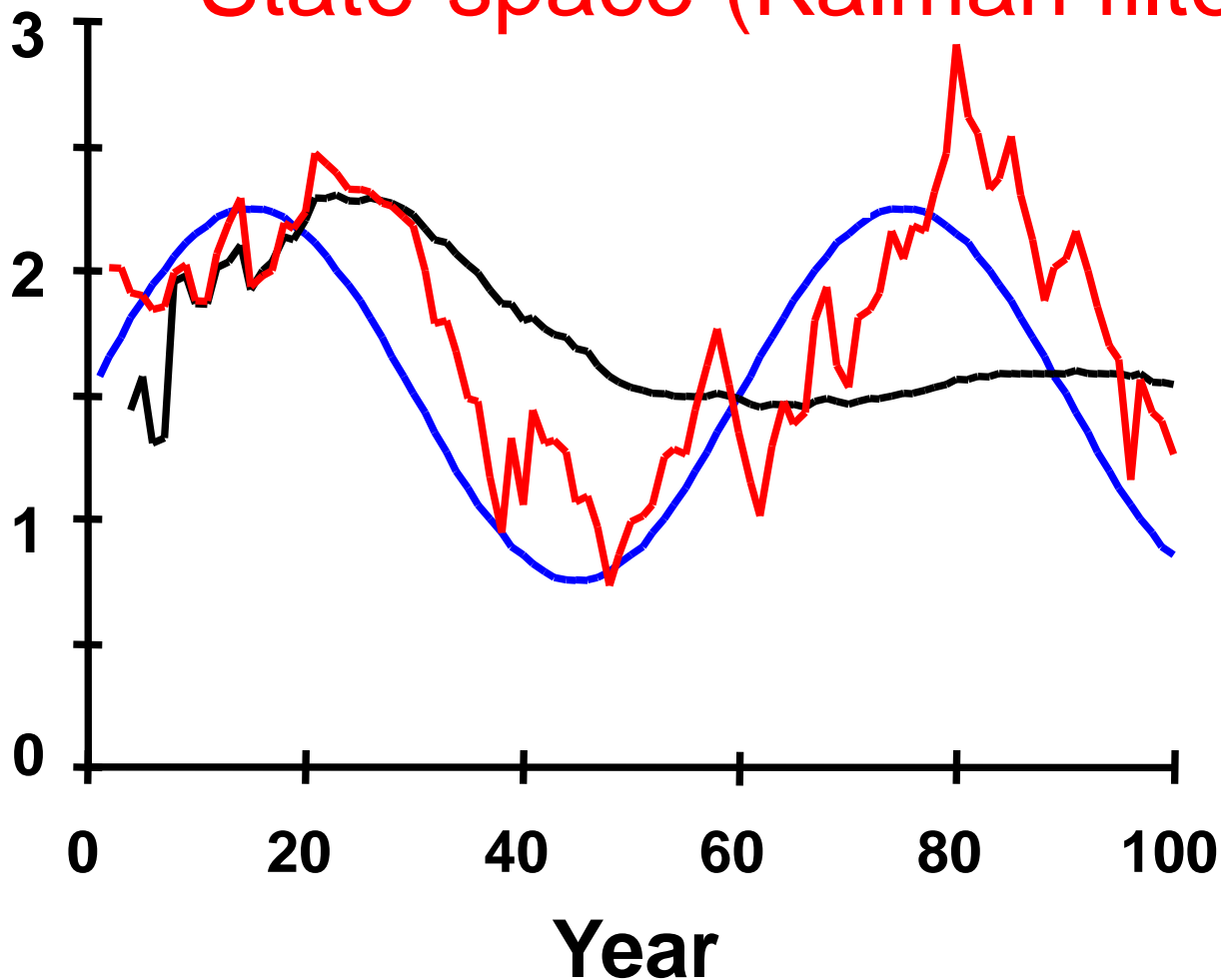
3. State-space models

- Estimate time-varying parameters (e.g. productivity)
- Take into account **natural variability and observation error**
 - Used for several non-salmonid species, both pelagic and demersal

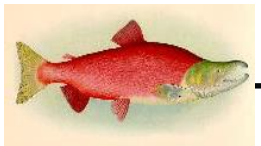
*One example
simulation trial*

- "True"
- Standard method
- State-space (Kalman filter)

Productivity
(Ricker a
parameter)

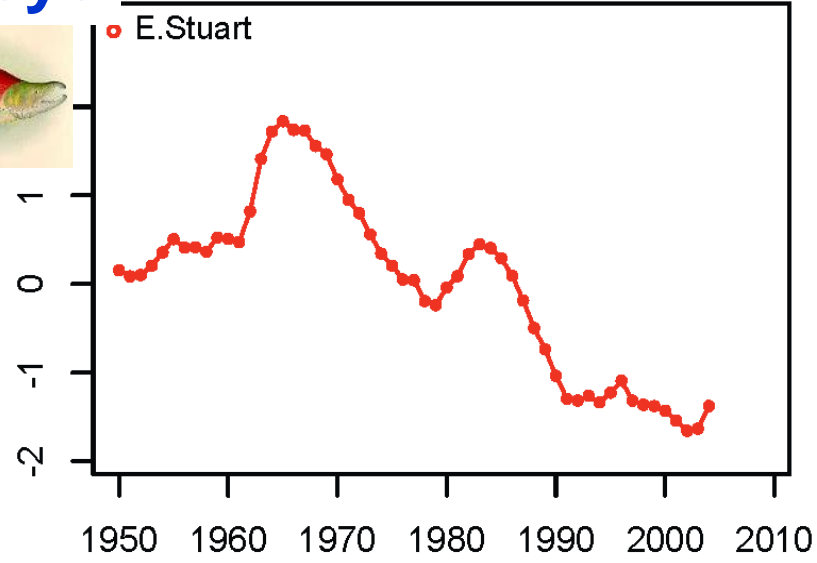


Sockeye

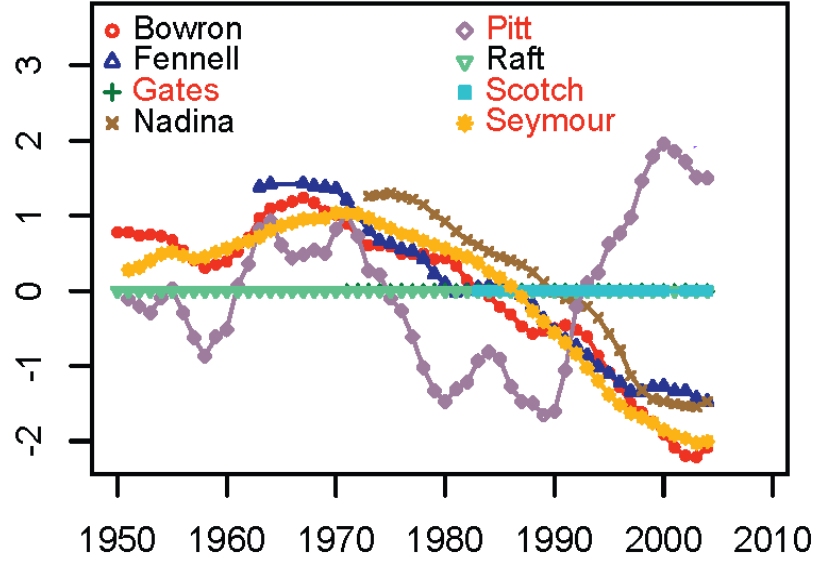


Fraser Early Stuart

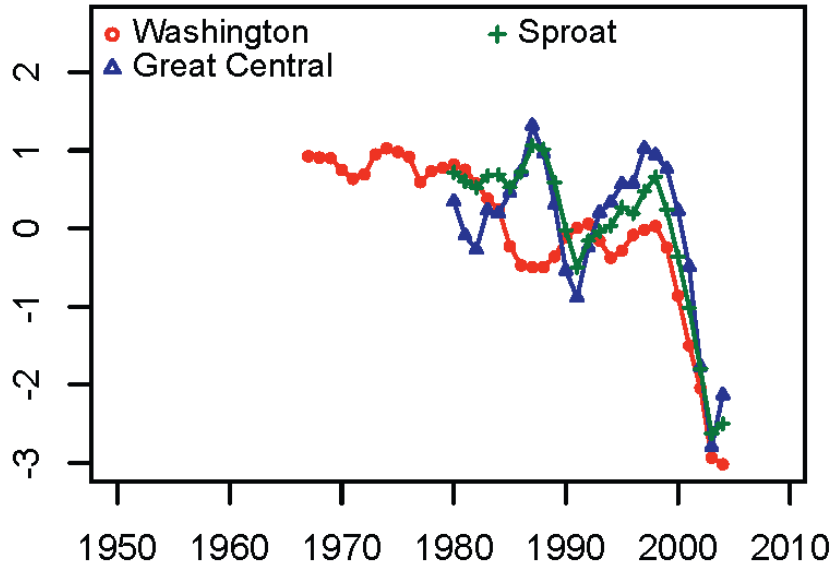
Standardized productivity



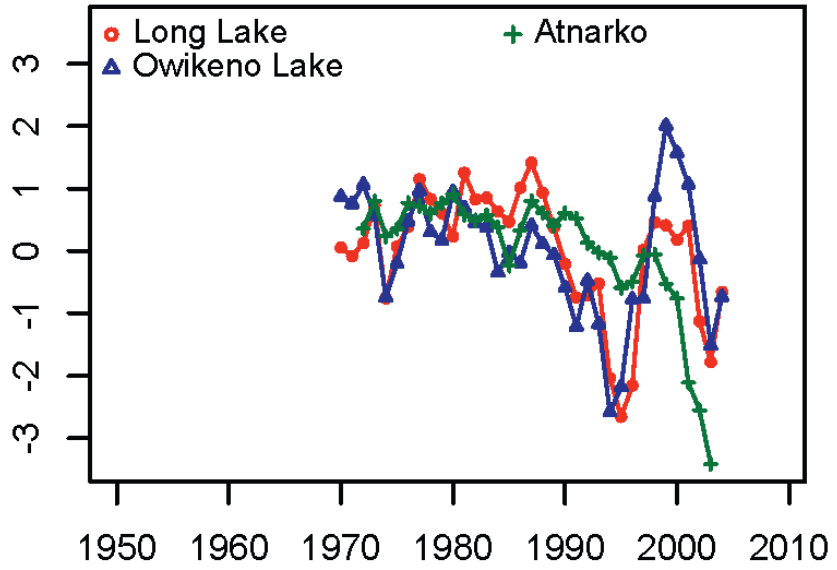
Fraser Early Summer



Wash. & Barkley Sd., B.C.



Central Coast B.C.



Year of spawning

Peterman and Dorner 2012

4. Dynamic Factor Analysis (DFA)

- Zuur et al. 2003, CJFAS
- Stachura et al. 2014, CJFAS

Many stocks share temporal patterns of variation

Dynamic Factor Analysis:

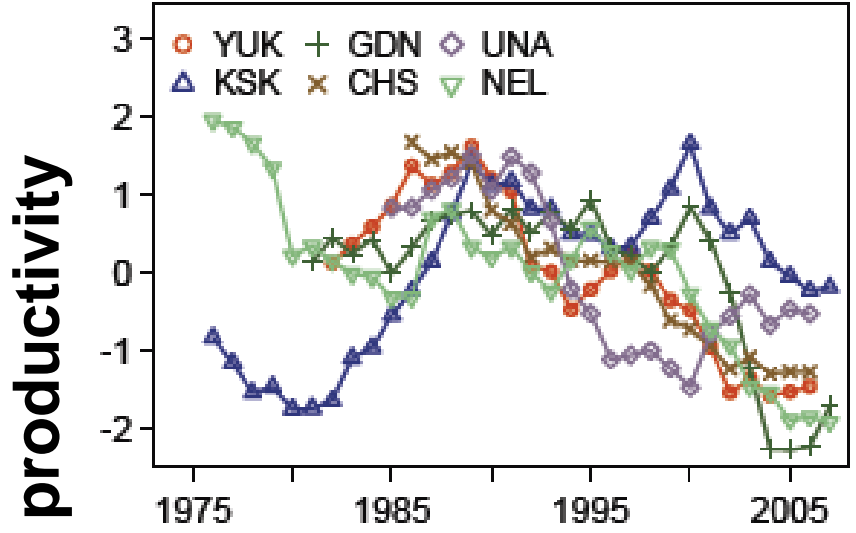
Quantifies those shared temporal patterns in productivity (finds "signal" amid "noise").

Dynamic Factor Analysis (DFA):

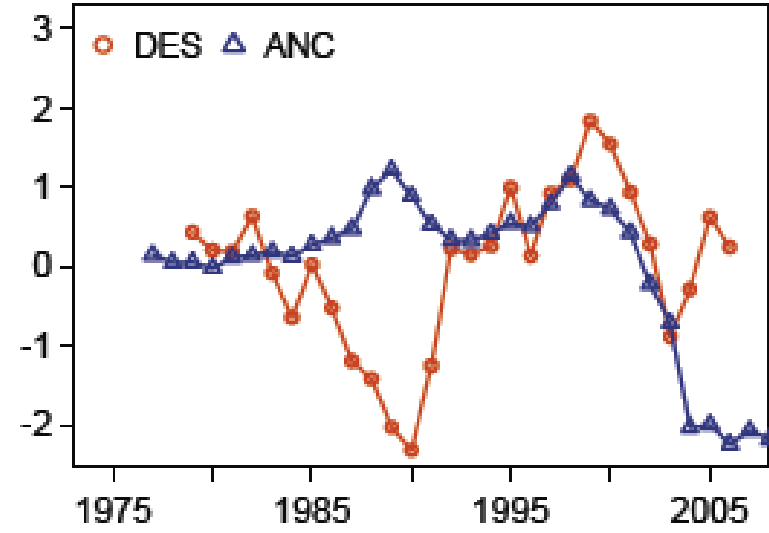
1. Is a factor analysis for time series data
2. Is a regression where independent variables (factors) are temporal patterns
3. Allows missing data, unlike PCA
4. Can evaluate importance of independent variables

Chinook (24 wild popul., Oregon through Alaska)

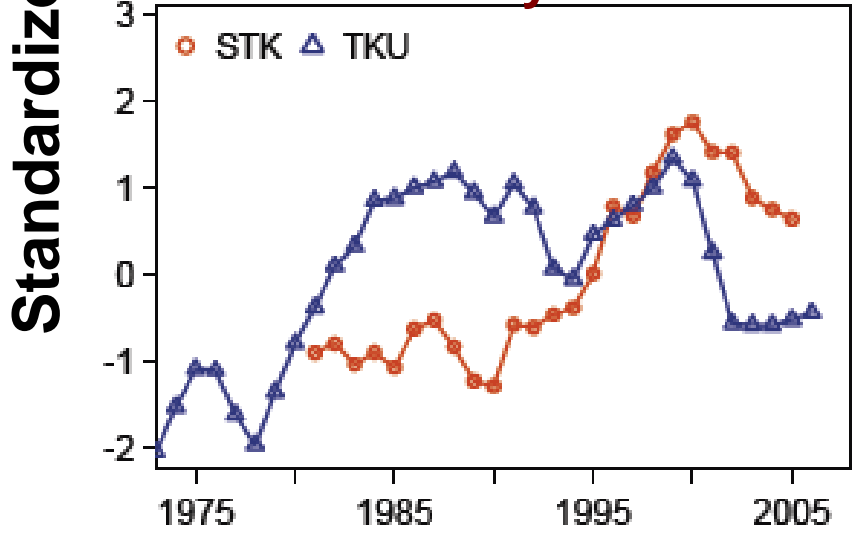
AYK, Alaska



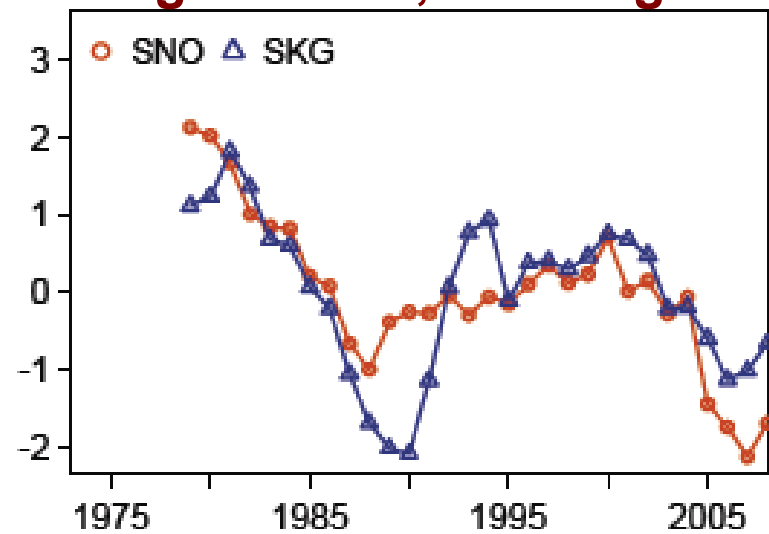
Cook Inlet, Alaska



Trans-boundary B.C-Alaska

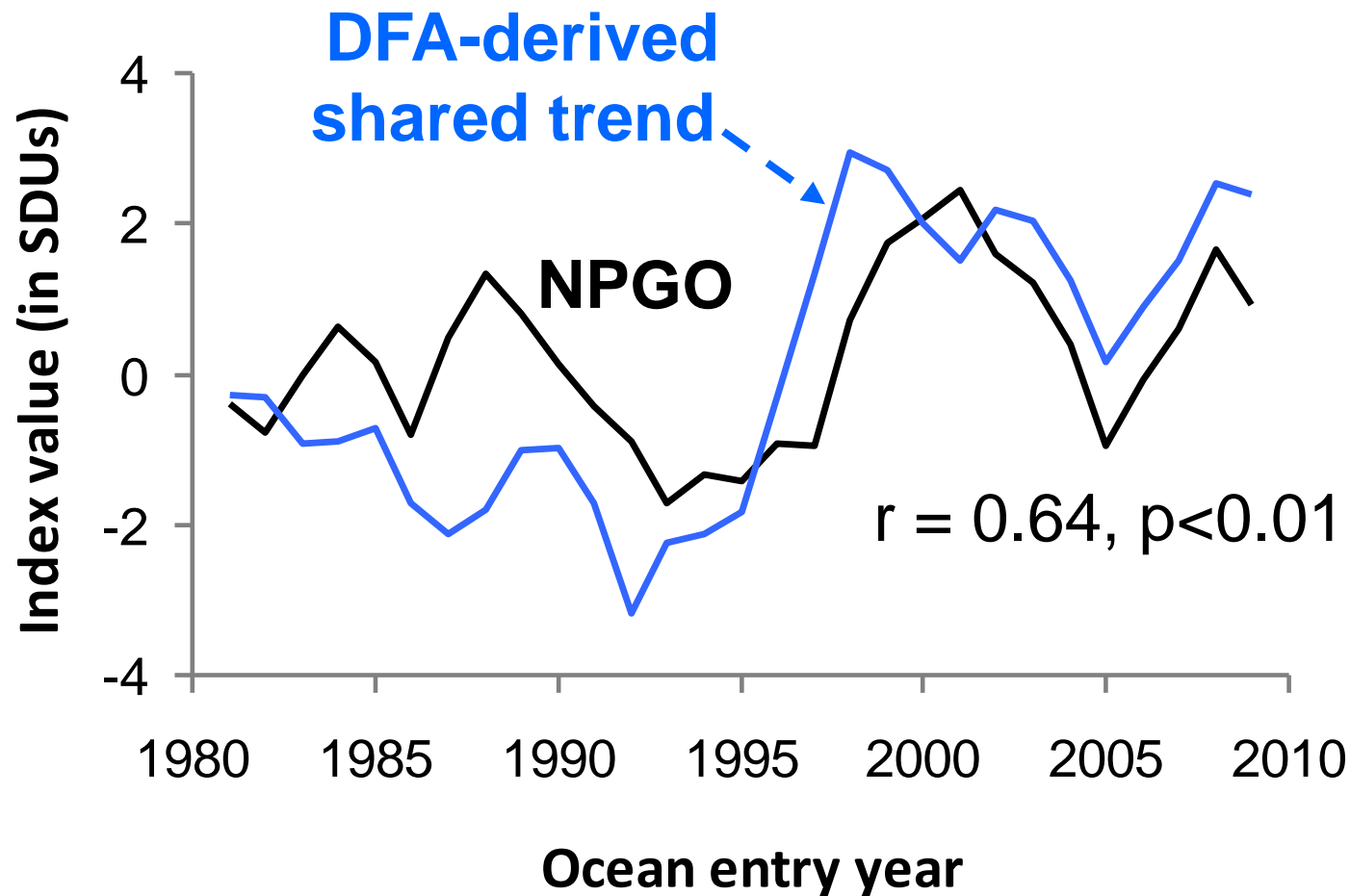


Puget Sound, Washington



Year of spawning

Chinook salmon (24 wild populs., Oregon thru Alaska)



Dorner et al. 2015, *in prep.*

Quantitative models

```
graph TD; QM[Quantitative models] --> S[Statistical]; QM --> Sim[Simulation]; S --> NEP[No explicit processes]; S --> WP[With processes]; S --> BN[Bayesian (i.e., probabilistic) networks]; BN --> B1[- Data-derived distrib. of probabilities of events]; BN --> B2[- Explicitly represent uncertainties in all processes];
```

Statistical

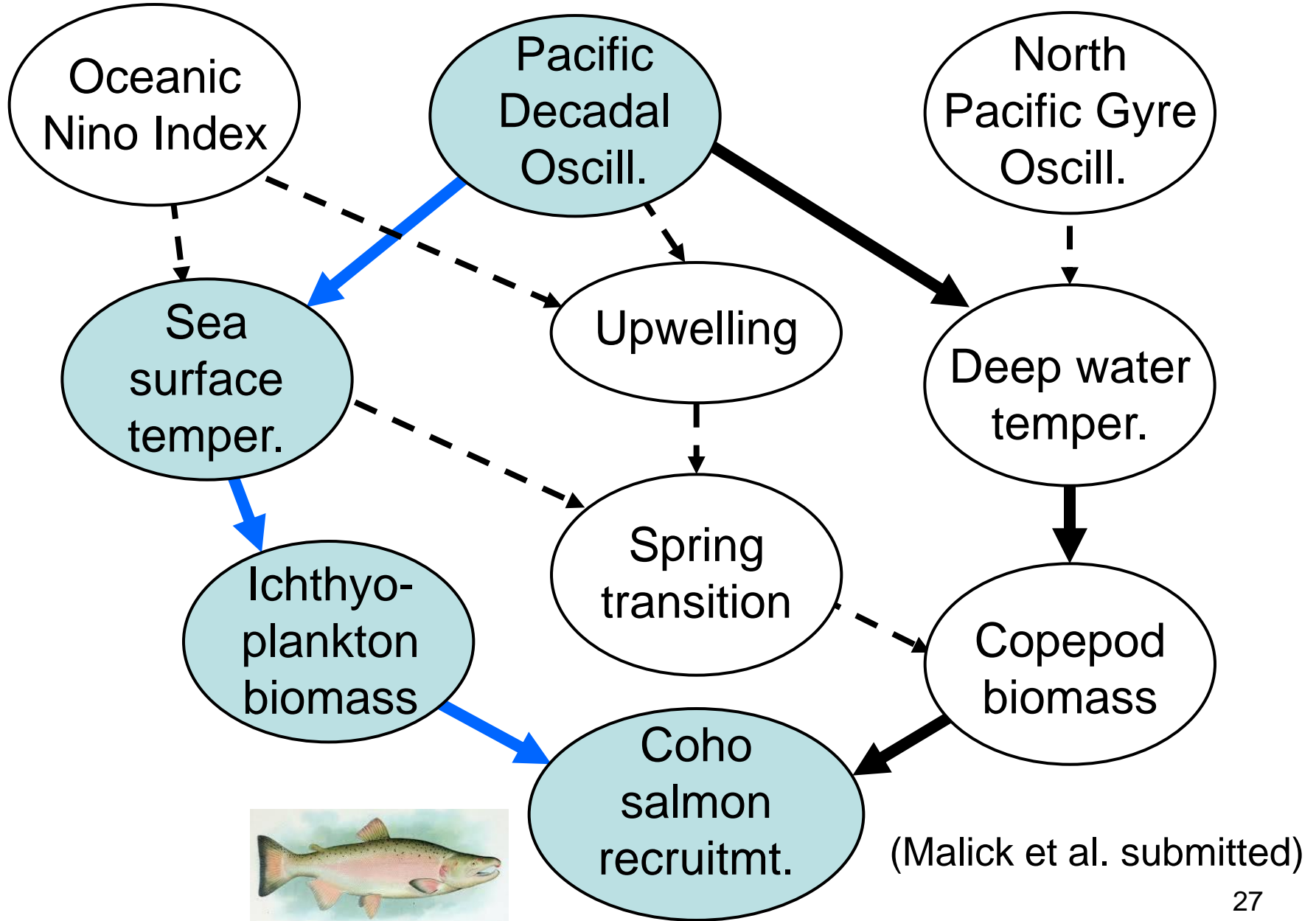
No explicit processes

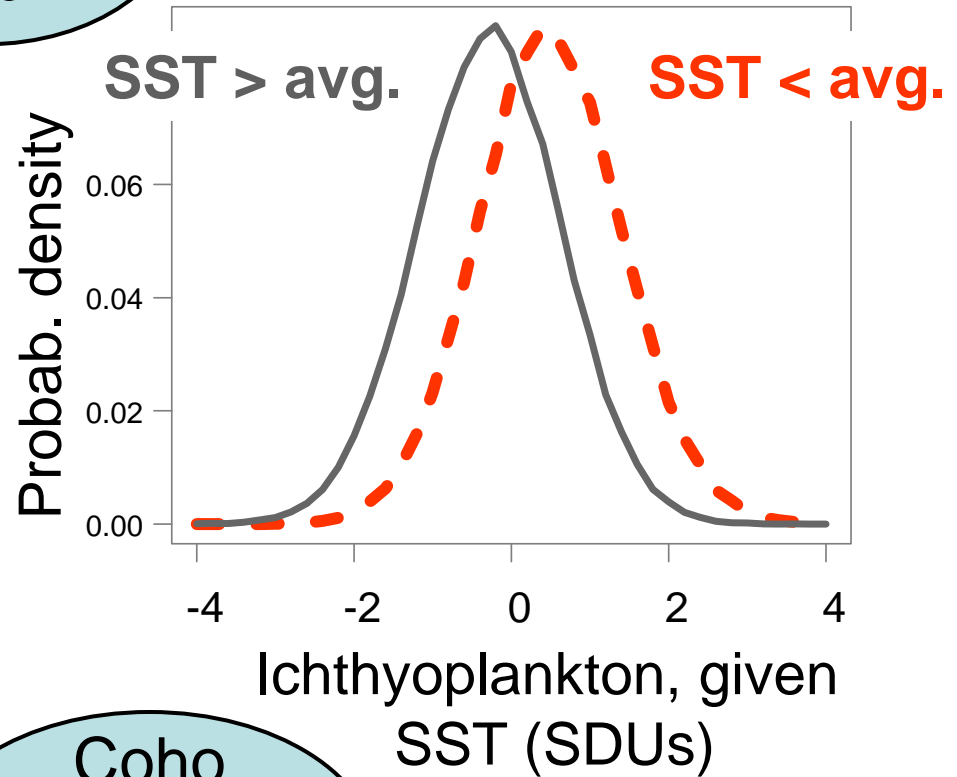
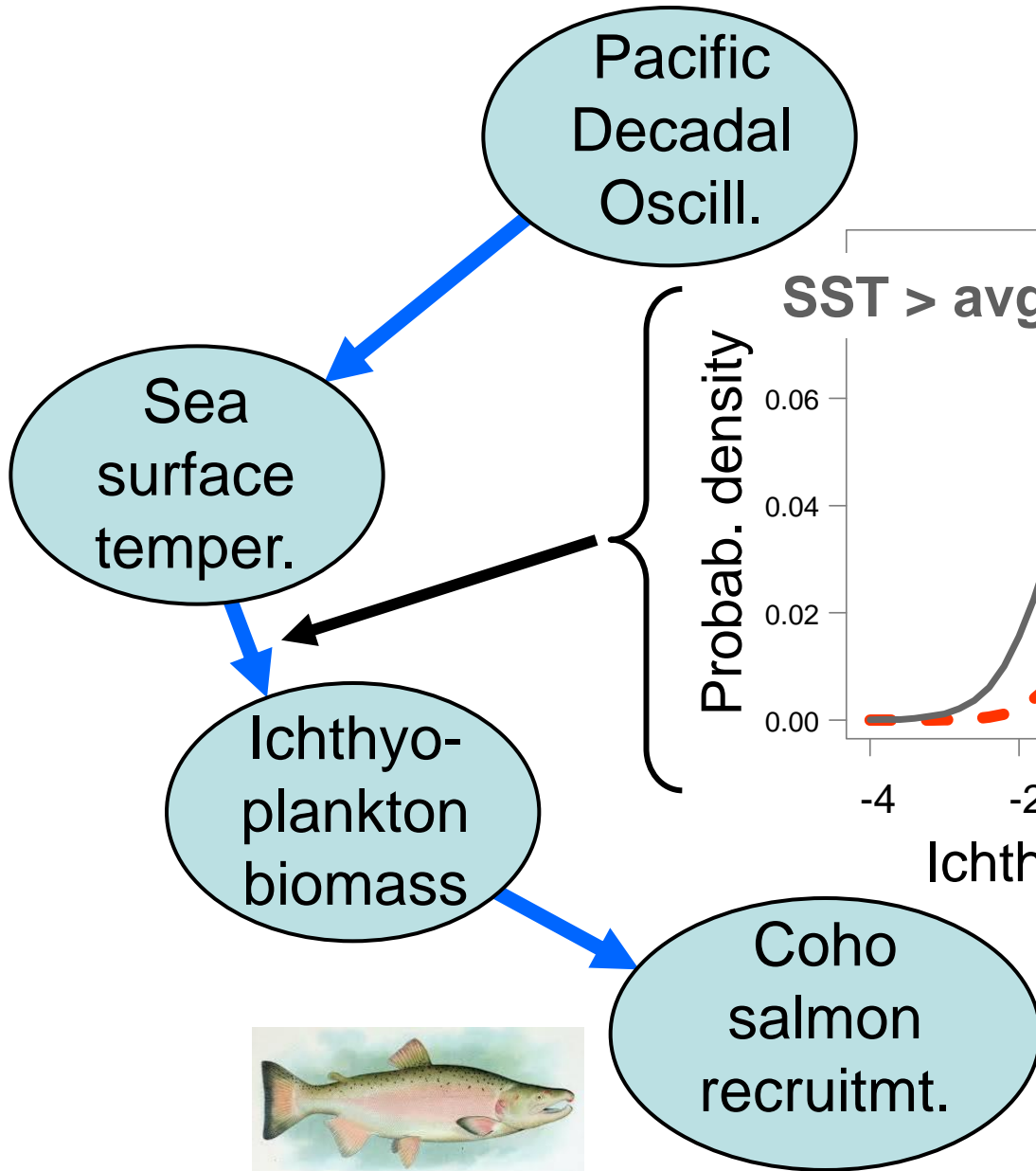
With processes

Bayesian (i.e., probabilistic) networks

- Data-derived distrib. of probabilities of events
- Explicitly represent uncertainties in all processes

Bayesian network linking climate to Oregon coho salmon





(Malick et al. submitted)

Advantages of Bayesian networks

1. Complexity is represented simply
2. Can do extensive sensitivity analyses

Quantitative models

```
graph TD; A[Quantitative models] --> B[Statistical]; A --> C[Simulation]; C --> D["- Mechanistic<br/>- Include physical and biological processes<br/>-- Salmon prey, predators, competitors<br/>- Range from simple to complex"]; style B fill:#eee,stroke:#333; style C fill:#f00,stroke:#333; style D fill:#fff,stroke:#333;
```

Statistical

Simulation

- Mechanistic
- Include physical and biological processes
 - Salmon prey, predators, competitors
- Range from simple to complex

Quantitative models

```
graph TD; A[Quantitative models] --> B[Statistical]; A --> C[Simulation]; C --> D[Simple]; D --- E["- Single-population, Ricker model with SST or ..."]
```

Statistical

Simulation

Simple

- Single-population, Ricker model with SST or ...

Quantitative models

Statistical

Simulation

Simple

Moderate complexity

- NEMURO.FISH
- **M**odels of **I**ntermediate **C**omplexity (**MIC** or **MICE**) for **non-salmonids** (Hannah et al. 2010 Prog. Oceanog.; Plaganyi et al. 2014 Fish and Fisheries)
- Multi-popul., climate-driven changes (Dorner et al. 2009)

Quantitative models

Statistical

Simulation

Simple

Moderate complexity

Very complex ("end-to-end", socio-ecological)

- Bering Sea Ecosystem (BSIERP)
- ATLANTIS for California Current system

Quantitative models

```
graph TD; QM[Quantitative models] --> S[Statistical]; QM --> Sim[Simulation]; Sim --> Simple[Simple]; Sim --> MC[Moderate complexity]; Sim --> VC[Very complex ("end-to-end", socio-ecological)]; Sim --> SD[State-dependent dynamics of decision making];
```

Statistical

Simulation

Simple

Moderate complexity

Very complex ("end-to-end", socio-ecological)

State-dependent dynamics of decision making

- In-season management models (Bristol Bay, ...)

- **Management Strategy Evaluation (MSE) models**

Management Strategy Evaluation (MSE) models

Take into account uncertainties

- Include multiple alternative models (structural uncert.)
- Change focus away from finding the best model to finding **management procedures** that are the most **robust to uncertainties**

Management Strategy Evaluation (MSE) models

Take into account uncertainties

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Management procedures

1. Data collection
2. Analysis of data
3. Decision rules based on data on state of system

CLIM2 simulation model (intermediate complexity)

Climatic effects:
4 temporal patterns

15 sockeye salmon populations:
- Uncertain parameters
- 3 levels of spatial covariation

Harvesting:
3 types of outcome uncert.

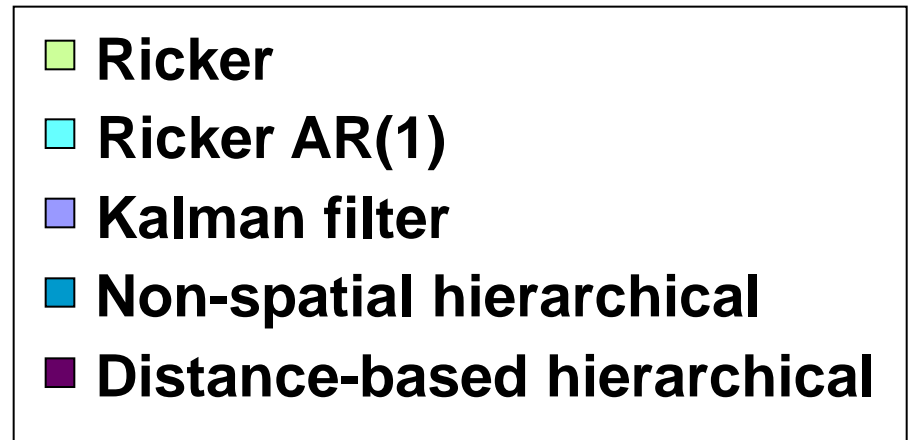
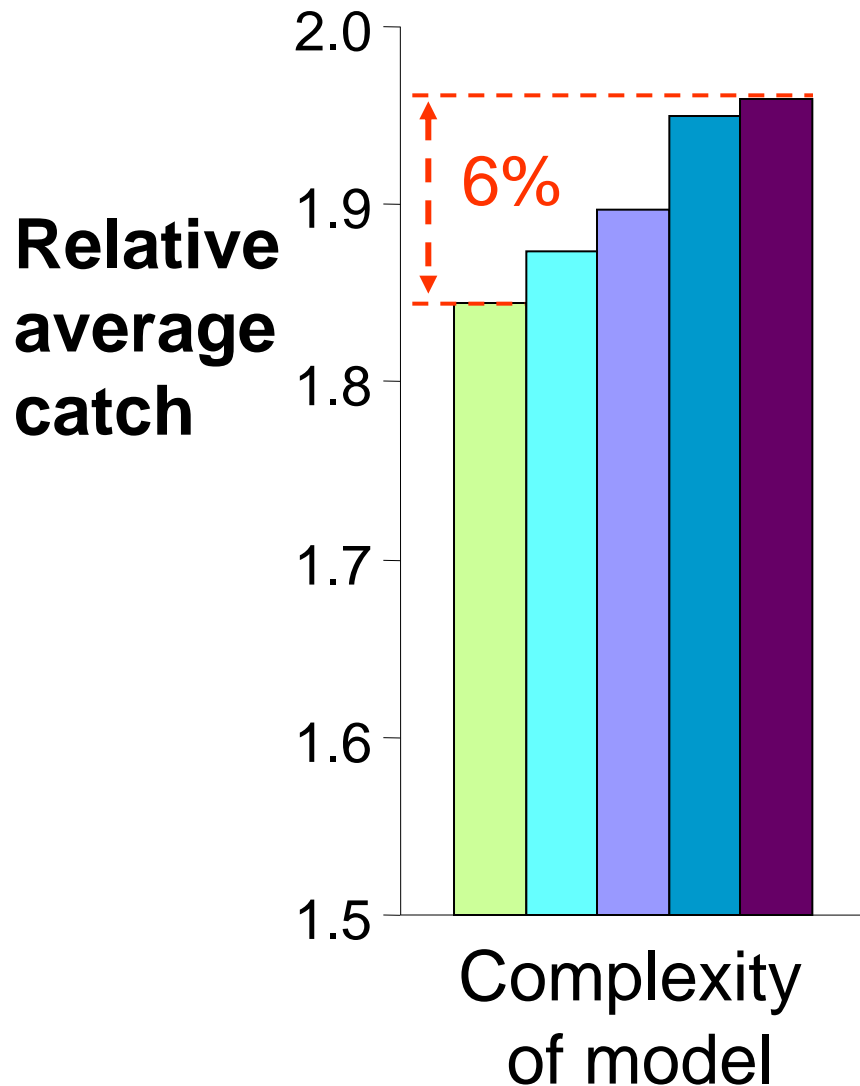
Repeat over
100 years,
500 Monte
Carlo trials

Data collection:
Imperfect

Stock assessment:
7 different models

Management:
3 safety margins

Management procedure



**More complex models
are not much better**

Similar result for "Index of conservation concern"

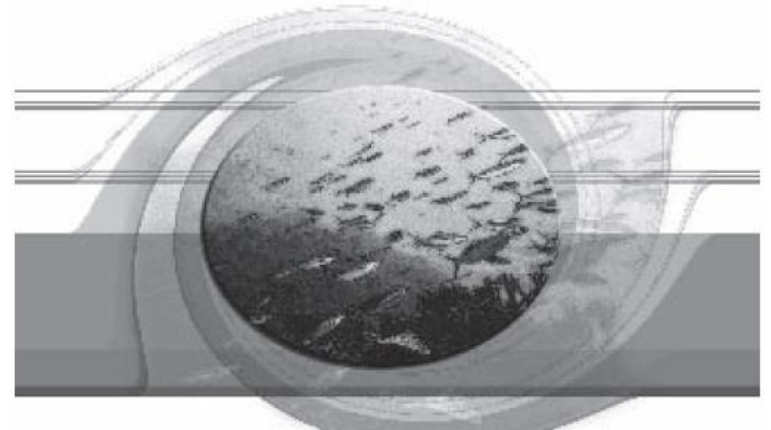
FAO's 2008 "Best practices" for evaluating models of aquatic ecosystems

- Multiple models
- Management Strategy Evaluation is best approach

FISHERIES MANAGEMENT

2. The ecosystem approach to fisheries

2.1 Best practices in ecosystem modelling for informing an ecosystem approach to fisheries



Features of advanced models ...

Explicitly include uncertainties in ...

- Natural variability
- Observation error
- Model structure
- Parameter values
- Outcome uncertainty (difference between actual outcomes and planned outcomes)

Incorporate knowledge from social as well as natural sciences

- Economics
- Management decision-making processes
- Fishing behaviour (e.g., discarding)

...

Recommendations for climate-salmon modellers

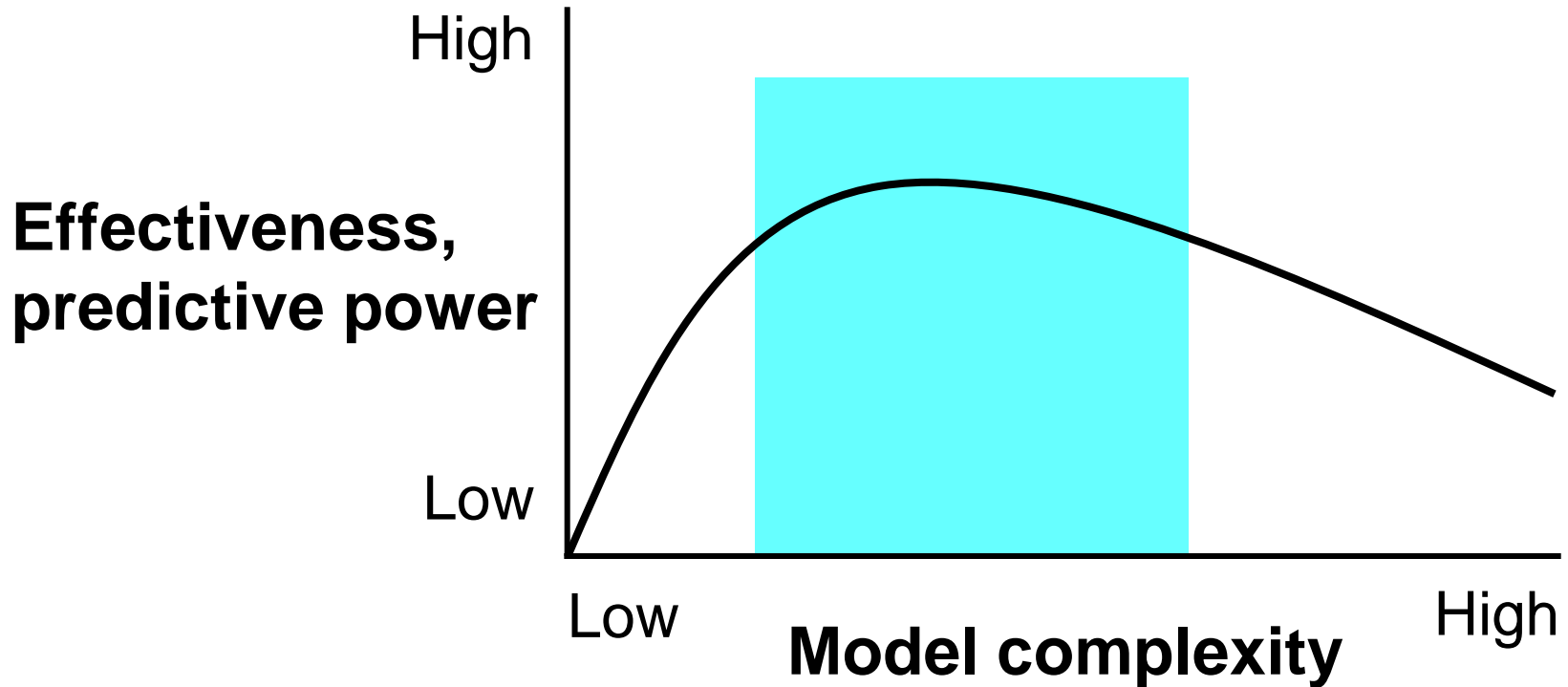
1. Explore multiple models (i.e., multiple hypotheses) instead of aiming to develop THE single "best" model.
 - a. Such models that include uncertainties will better reflect uncertainties about climatic effects on salmon.
 - b. Cover range of complexity:



Models of intermediate complexity (MICE);
frequently used in non-salmonid fisheries

Recommendations for climate-salmon modellers ...

2. Models that are too complex may be unreliable.

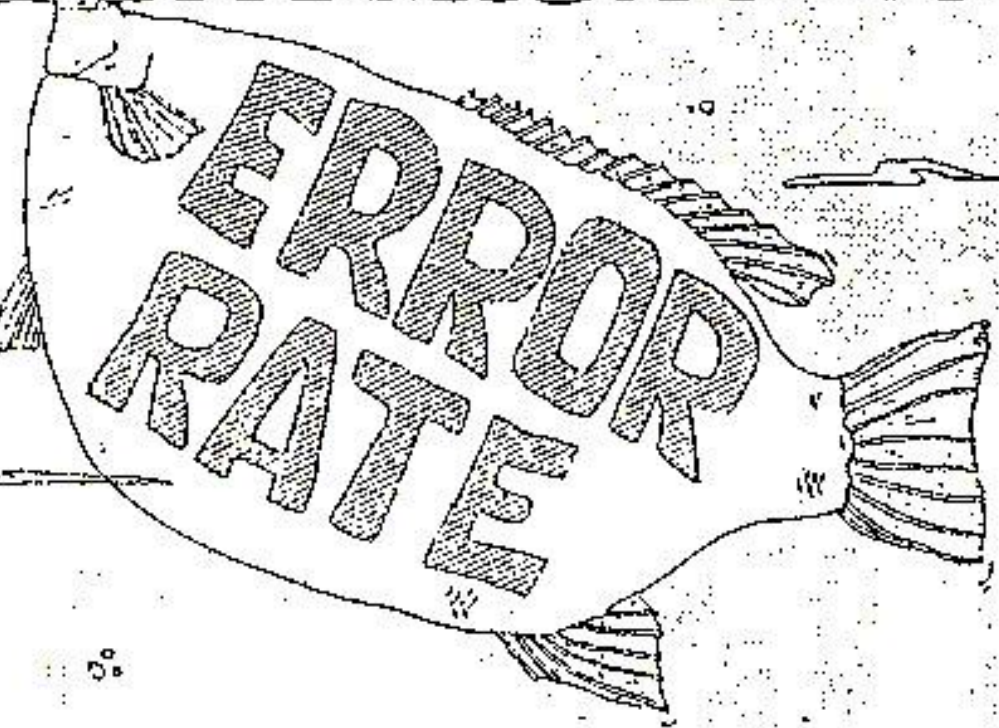
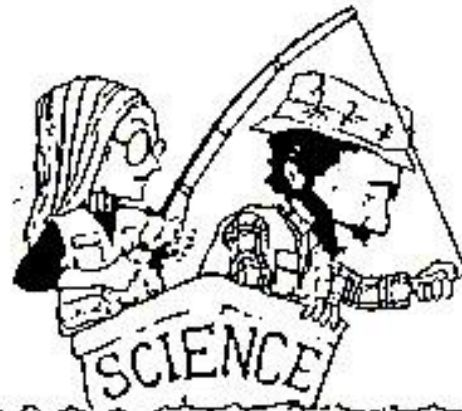


Fulton et al. (2003), others

Recommendations for climate-salmon modellers ...

3. Explore newer methods such as:
 - **Hierarchical models** (for data on multiple locations)
 - **State-space models** (for time-varying parameters)
 - **Dynamic Factor Analysis** (for both multiple locations and time-varying parameters)
 - **Bayesian networks**
 - **Management Strategy Evaluation (MSE)**

EH, IT'S JUST A LITTLE ONE. NO PROBLEM!



Stelly ©85