

# Applications of Dynamic Linear Models

Mark Scheuerell

*FISH 507 – Applied Time Series Analysis*

2 February 2017

# Why use a DLM?

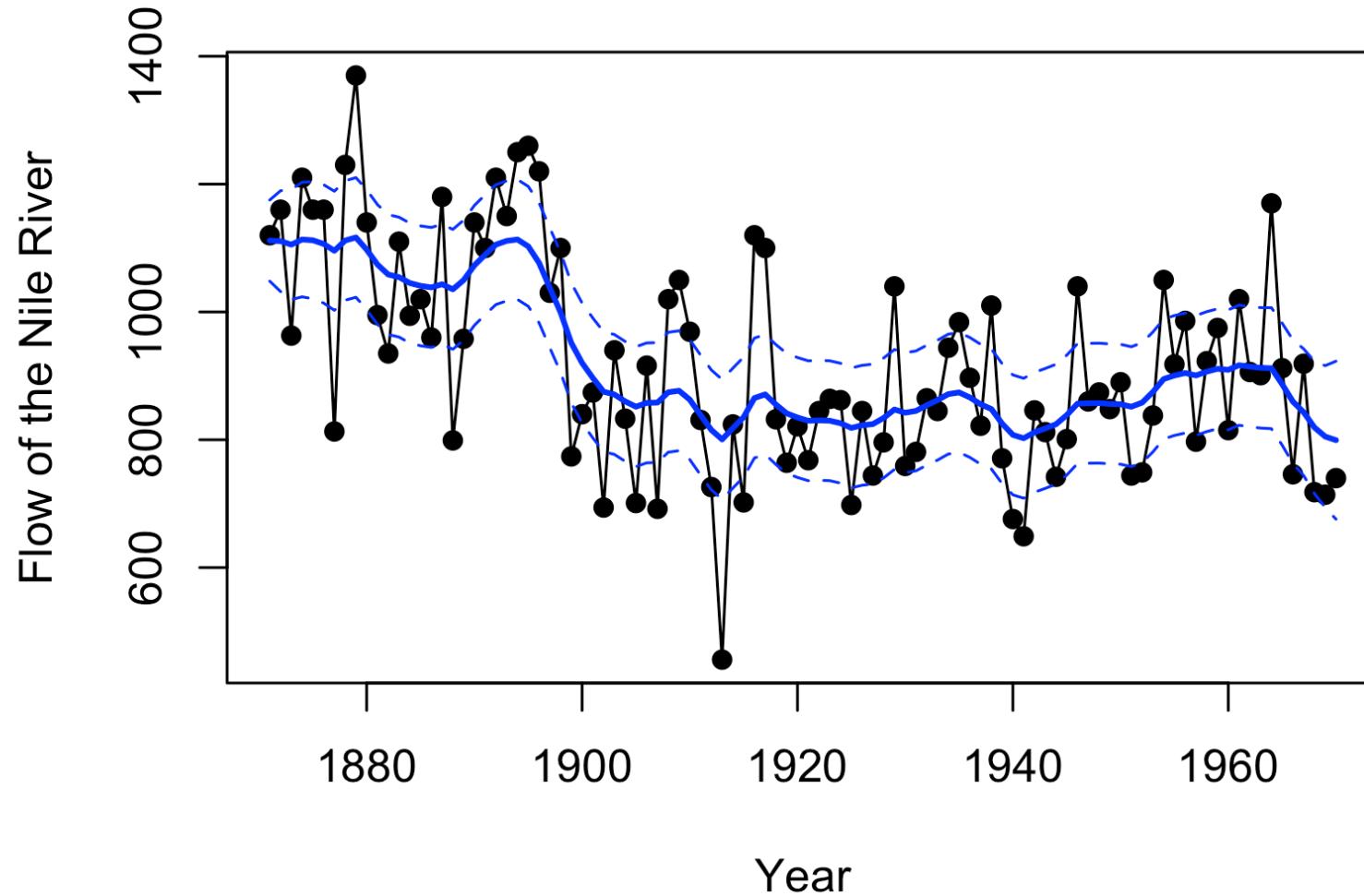
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DLMs are useful if/when:

- The underlying level (intercept) changes over time  
(eg, flow of the Nile R)

# DLM for changing level (intercept)

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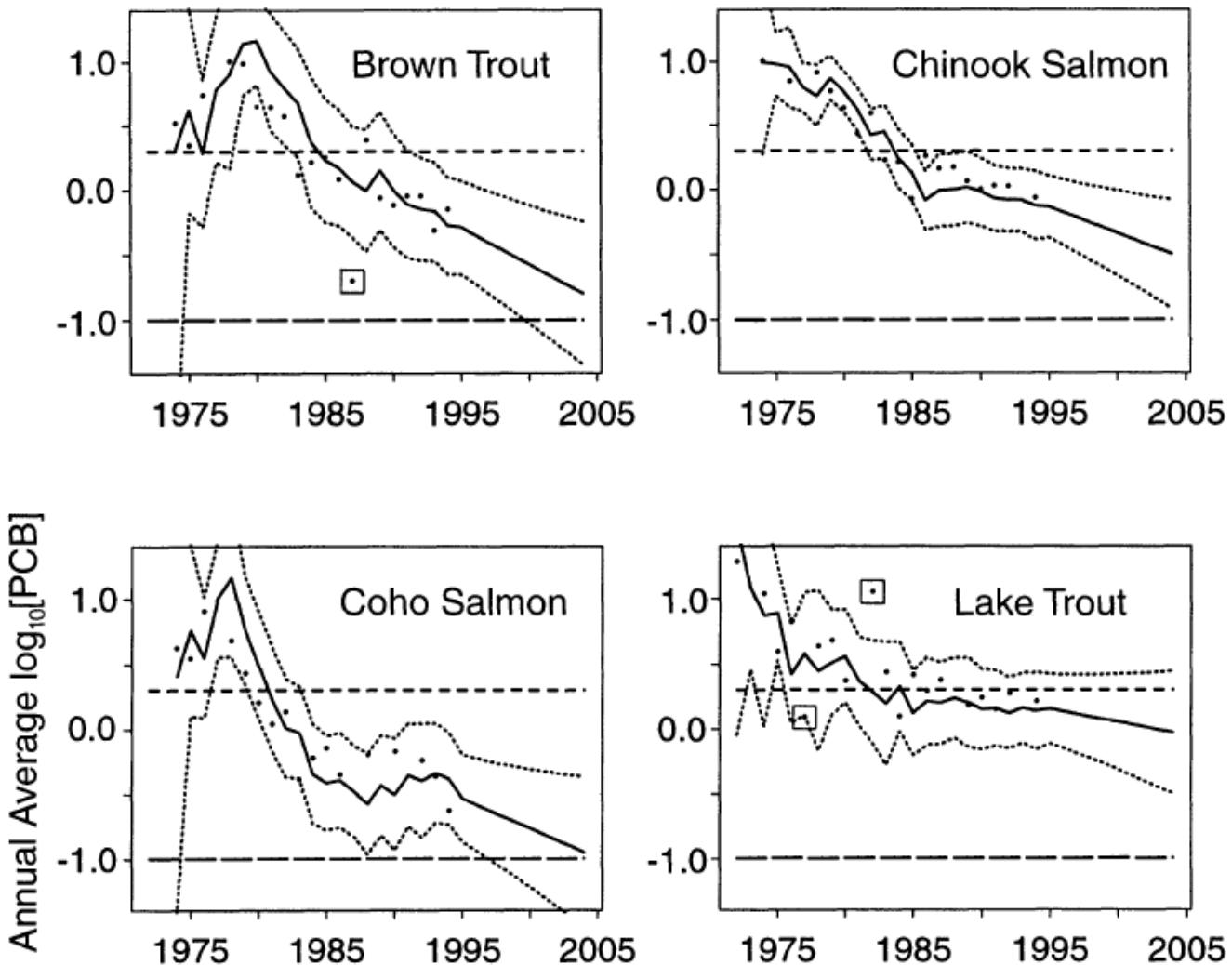
# Why use a DLM?

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DLMs are useful if/when:

- The underlying level (intercept) changes over time  
(eg, flow of the Nile R)
- The underlying growth (bias) changes over time  
(eg, PCB's in L Michigan salmonids)

# DLM with changing growth (bias)



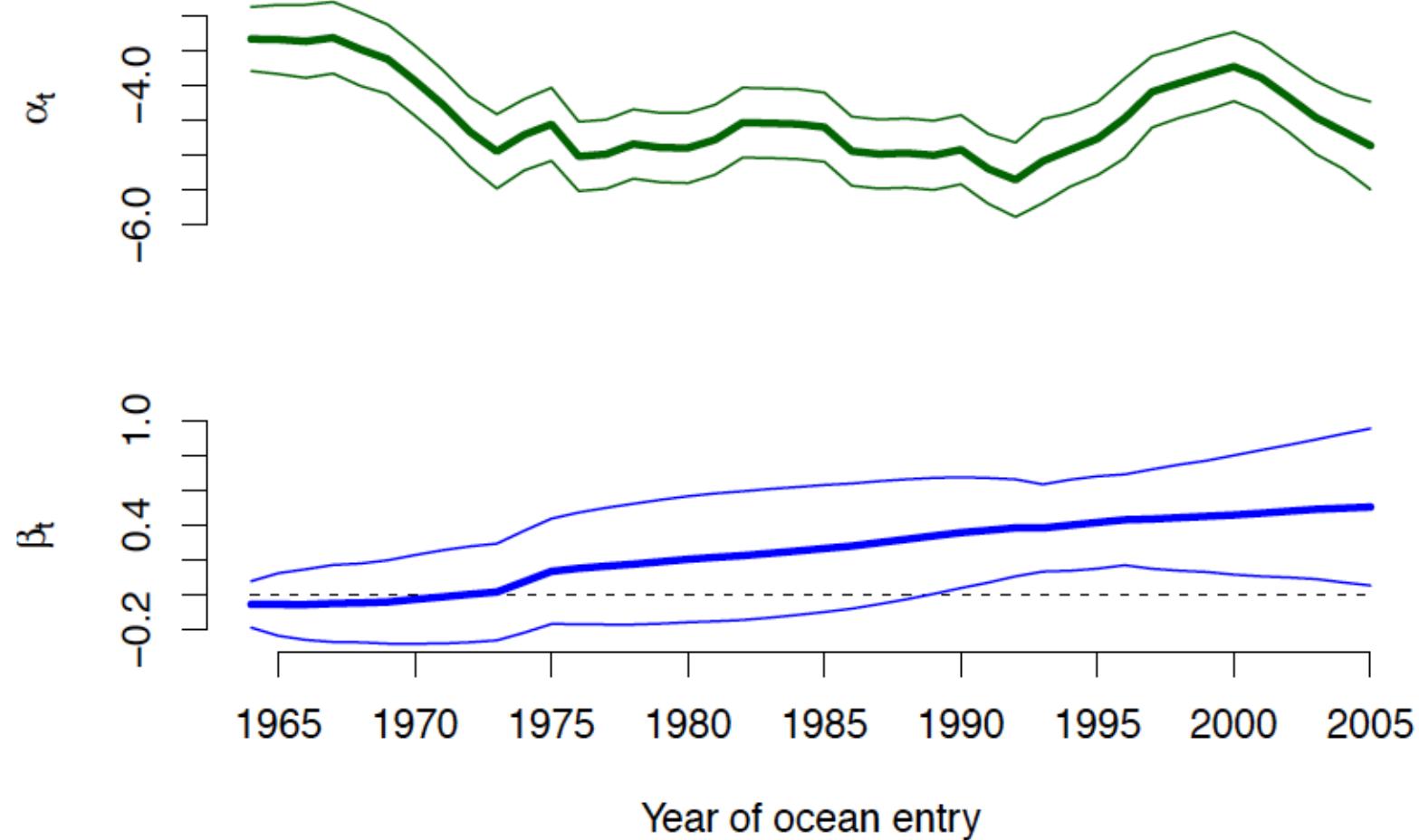
# Why use a DLM?

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DLMs are useful if/when:

- The underlying level (intercept) changes over time  
(eg, flow of the Nile R)
- The underlying growth (bias) changes over time  
(eg, PCB's in L Michigan salmonids)
- The relationship between the response and predictor (slope) changes over time (eg, effect of upwelling on salmon survival)

# DLM with changing effect (slope)



# From finance to fisheries: Using market models to evaluate returns versus risk for ESA-listed Pacific salmon

Mark Scheuerell

*NOAA Northwest Fisheries Science Center*

Jennifer Griffiths

*Stockholm University*

Eli Holmes

*NOAA Northwest Fisheries Science Center*

# Notions of risk in finance

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- Financial markets are rife with various forms of risk
- For simplicity, let's consider 2 broad categories:
  - 1) Systematic (market) risk is vulnerability to large-scale events or outcomes that affect entire markets (eg, natural disasters, govt policy, terrorism)
  - 2) Unsystematic (asset) risk is specific to particular securities or industries (eg, droughts affect commodities like corn, but not oil; bad batteries affect Boeing, but not Microsoft)

# Diversification

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- By holding a diverse collection of assets (a portfolio), one can reduce unsystematic, but not systematic, risk
- Investing is inherently risky, but (rationale) investors are risk averse
- That is, if presented with 2 portfolios offering equal returns, they should choose the less risky one
- Thus, investors expect to be compensated with higher returns for accepting more risk & vice-versa

# Estimating risks

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- Portfolios can only reduce unsystematic risks, so one should understand the systematic risk of an asset before it is added to a portfolio
- Sharpe (1963) outlined a model whereby returns of various assets are related through a combination of a common underlying influence & random factors
- Total risk = Systematic risk + Unsystematic risk

# The market model

- Many others (e.g., Treynor, Lintner, Beja) were also working on these ideas, which ultimately led to the “market model”

$$r_{a,t} = \alpha_a + \beta_a r_{m,t} + \nu_{a,t}$$

Manager's skill

Correlated volatility  
(sensitivity to systematic risk)

# Interpreting alpha

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Value of $\alpha$	Interpretation
$\alpha < 0$	Asset earns too little for its risk

\*

\*Expected value if market is “efficient” (*sensu* Fama 1970)

# Interpreting beta

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Value of $\beta$	Interpretation	Example
$\beta = 0$	Movement of asset is independent of market	Fixed-yield bond

# Capital Asset Pricing Model (CAPM)

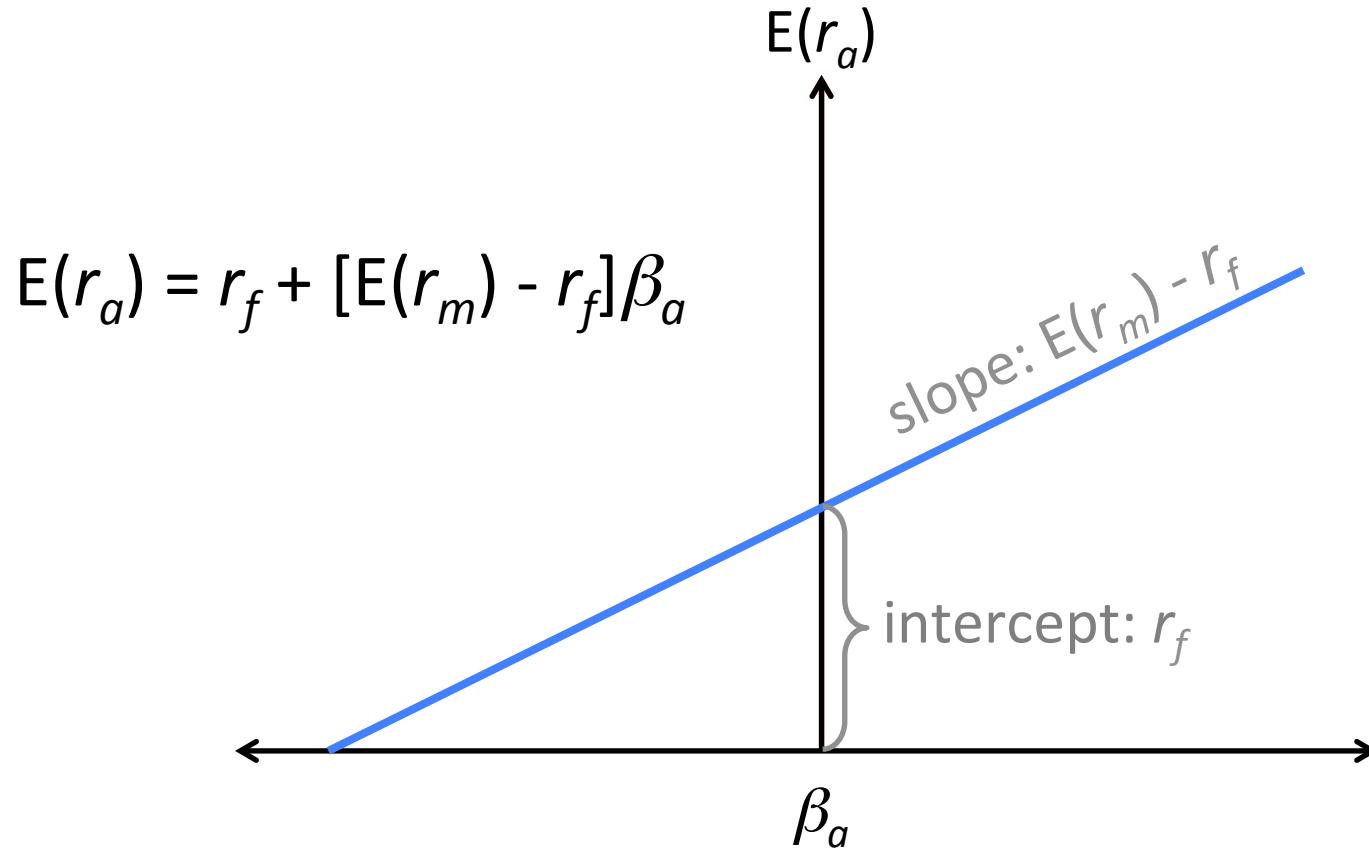
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- CAPM followed directly from the market model
- CAPM determines an expected rate of return necessary for an asset to be included in a portfolio based on:
  - 1) the asset's responsiveness to systematic risk ( $\beta$ );
  - 2) the expected return of the market; and
  - 3) the expected return of a risk-free asset (eg, US govt T-bills)
- CAPM is usually expressed via the security market line:

$$E(r_a) = r_f + [E(r_m) - r_f] \beta_a$$

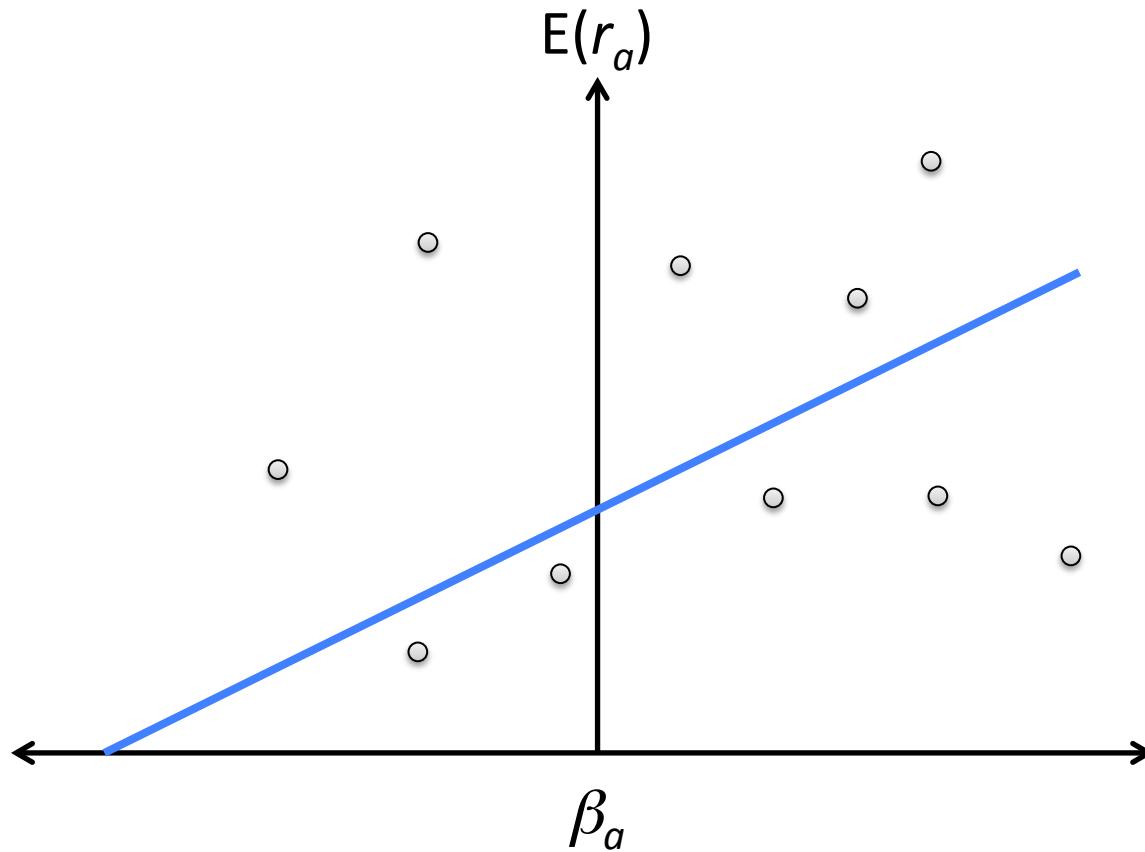
# Security market line (SML)

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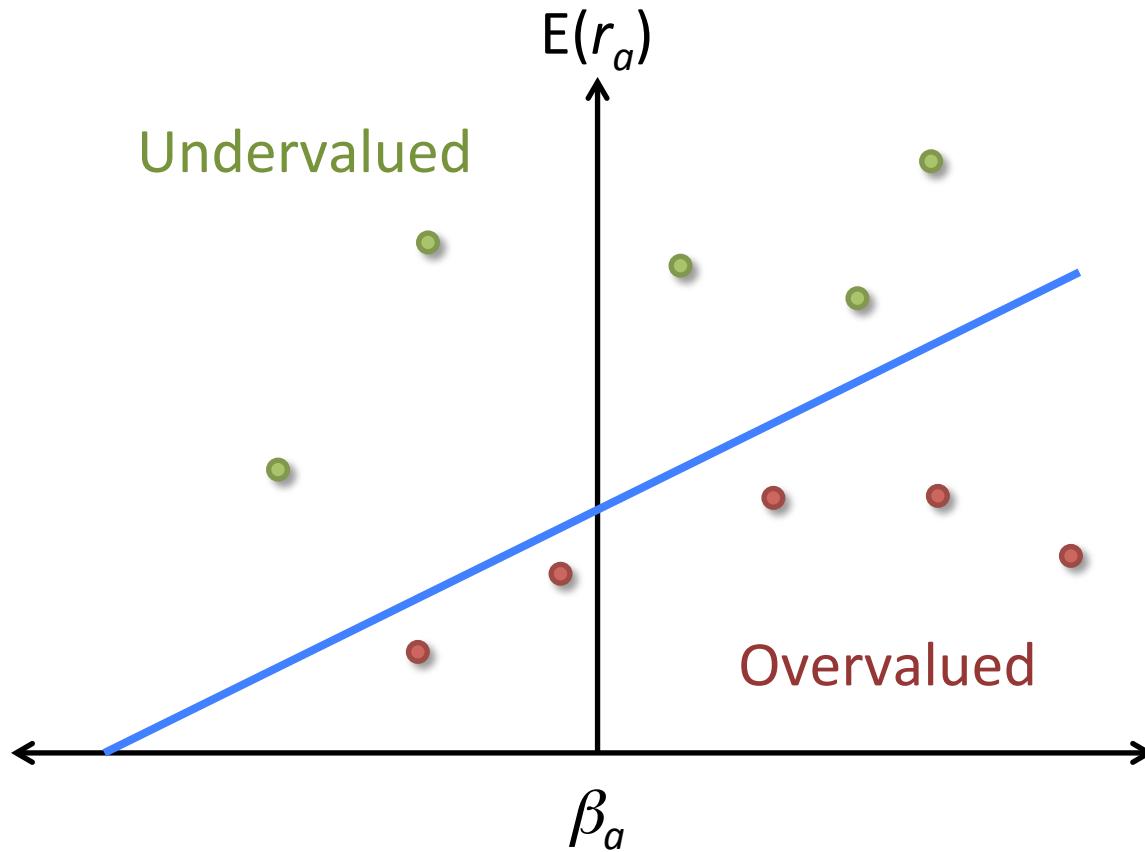
# Security market line

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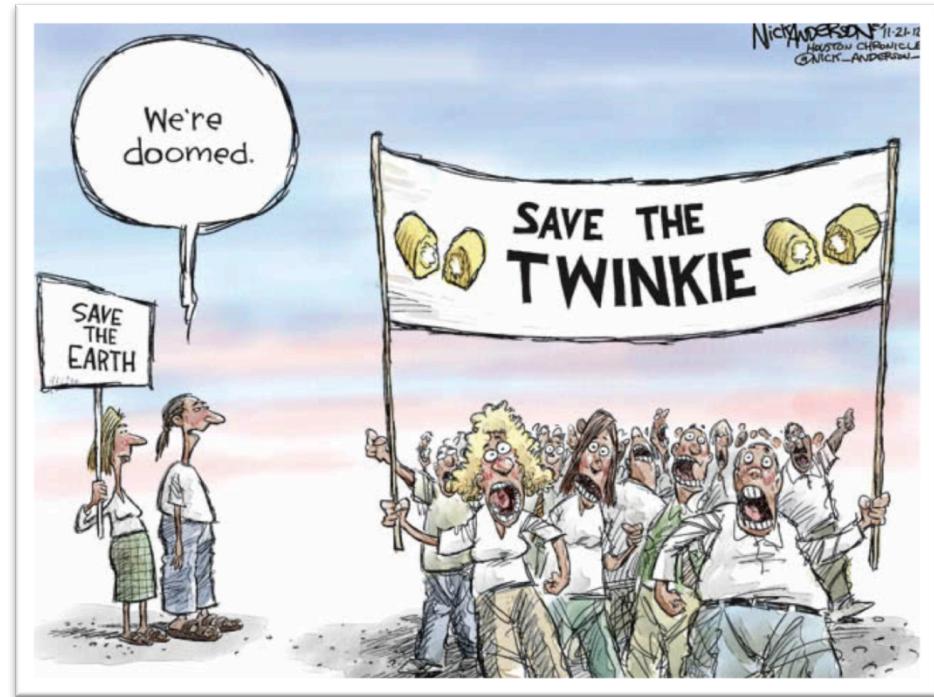
# Security market line

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# Ecological analogues

Many ecologists study “risk” & “returns” in a conservation context



# Portfolio ideas in nature

Vol 465 | 3 June 2010 | doi:10.1038/nature09060

nature

## LETTERS

### Population diversity and the portfolio effect in an exploited species

Daniel E. Schindler<sup>1</sup>, Ray Hilborn<sup>1</sup>, Brandon Chasco<sup>1</sup>, Christopher P. Boatright<sup>1</sup>, Thomas P. Quinn<sup>1</sup>, Lauren A. Rogers<sup>1</sup> & Michael S. Webster<sup>2</sup>

LETTER

### Synchronization and portfolio performance of threatened salmon

Jonathan W. Moore<sup>1</sup>, Michelle McClure<sup>1</sup>, Lauren A. Rogers<sup>2</sup>, & Daniel E. Schindler<sup>2</sup>

<sup>1</sup> Northwest Fisheries Science Center, National Marine Fisheries Service, Seattle, WA 98110, USA

<sup>2</sup> School of Aquatic and Fishery Sciences, University of Washington, Seattle, WA 98195, USA

Journal of Animal Ecology



*Journal of Animal Ecology* 2014, **83**, 157–167

doi: 10.1111/1365-2656.12117

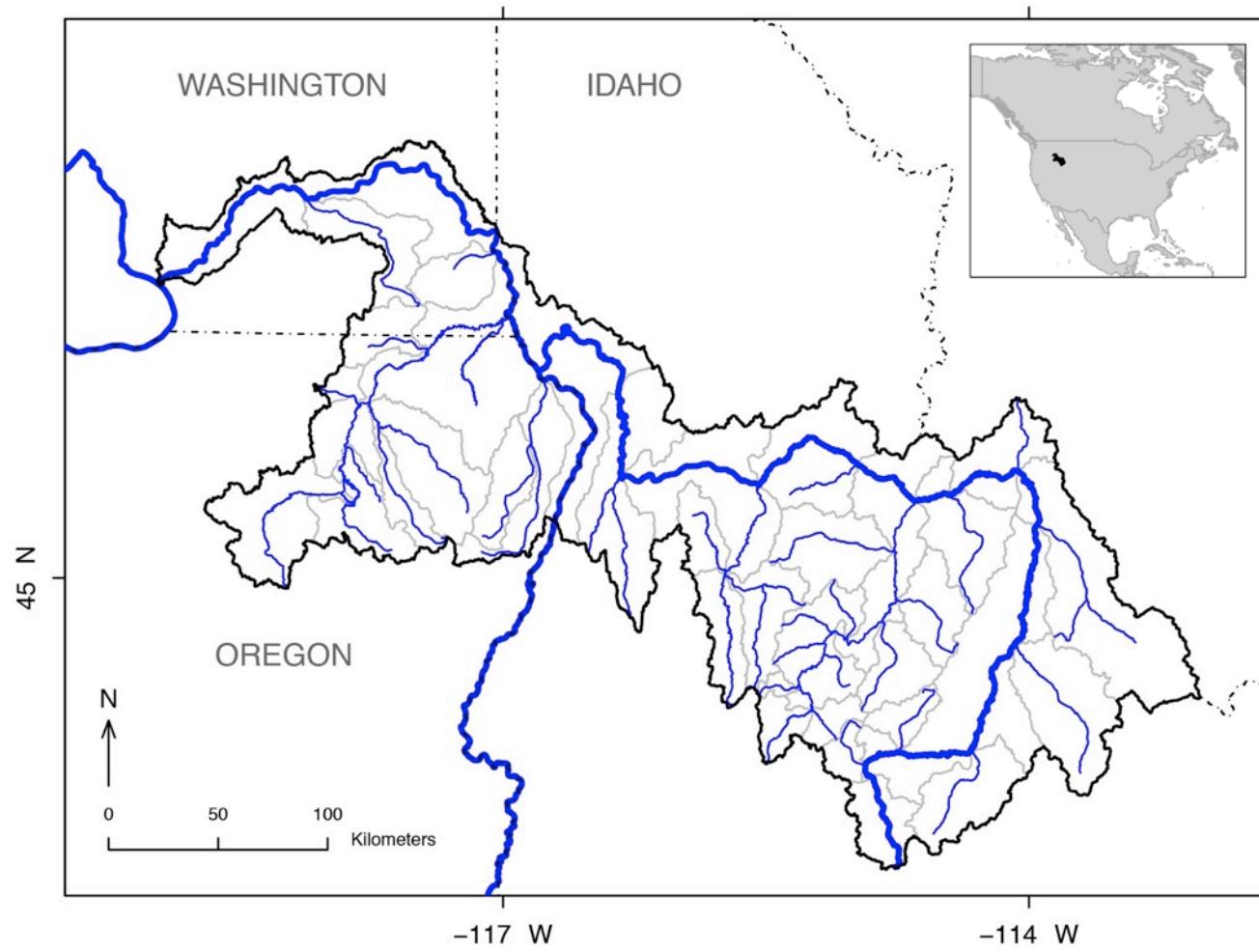
### Spatial variation buffers temporal fluctuations in early juvenile survival for an endangered Pacific salmon

James T. Thorson<sup>1\*</sup>, Mark D. Scheuerell<sup>1</sup>, Eric R. Buhle<sup>1</sup> and Timothy Copeland<sup>2</sup>

<sup>1</sup>National Oceanic and Atmospheric Administration, National Marine Fisheries Service, Northwest Fisheries Science Center, 2725 Montlake Blvd. East, Seattle, WA 98112, USA; and <sup>2</sup>Idaho Department of Fish and Game 1414 East Locust Lane, Nampa, ID 83686, USA

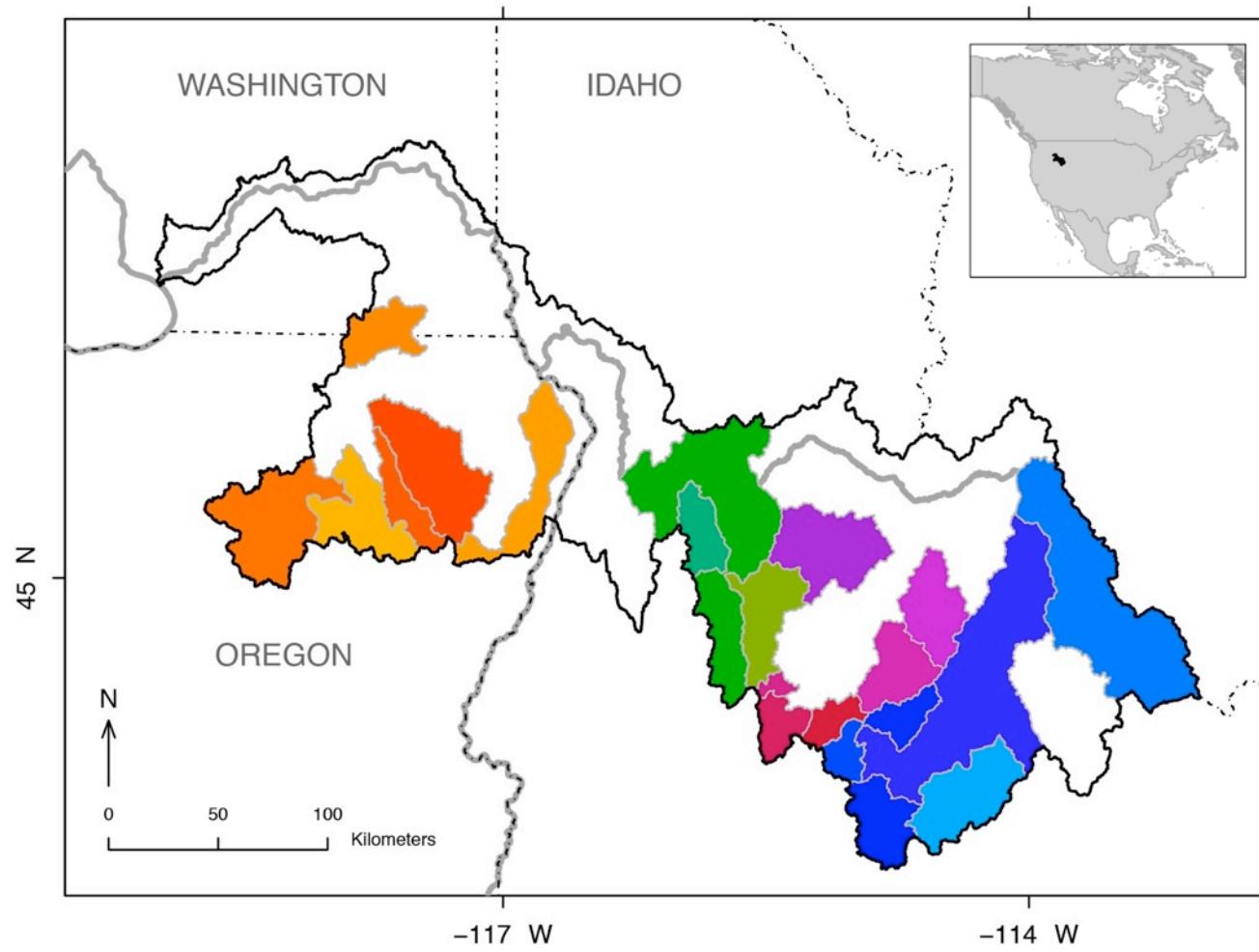


# Snake R spring/summer Chinook



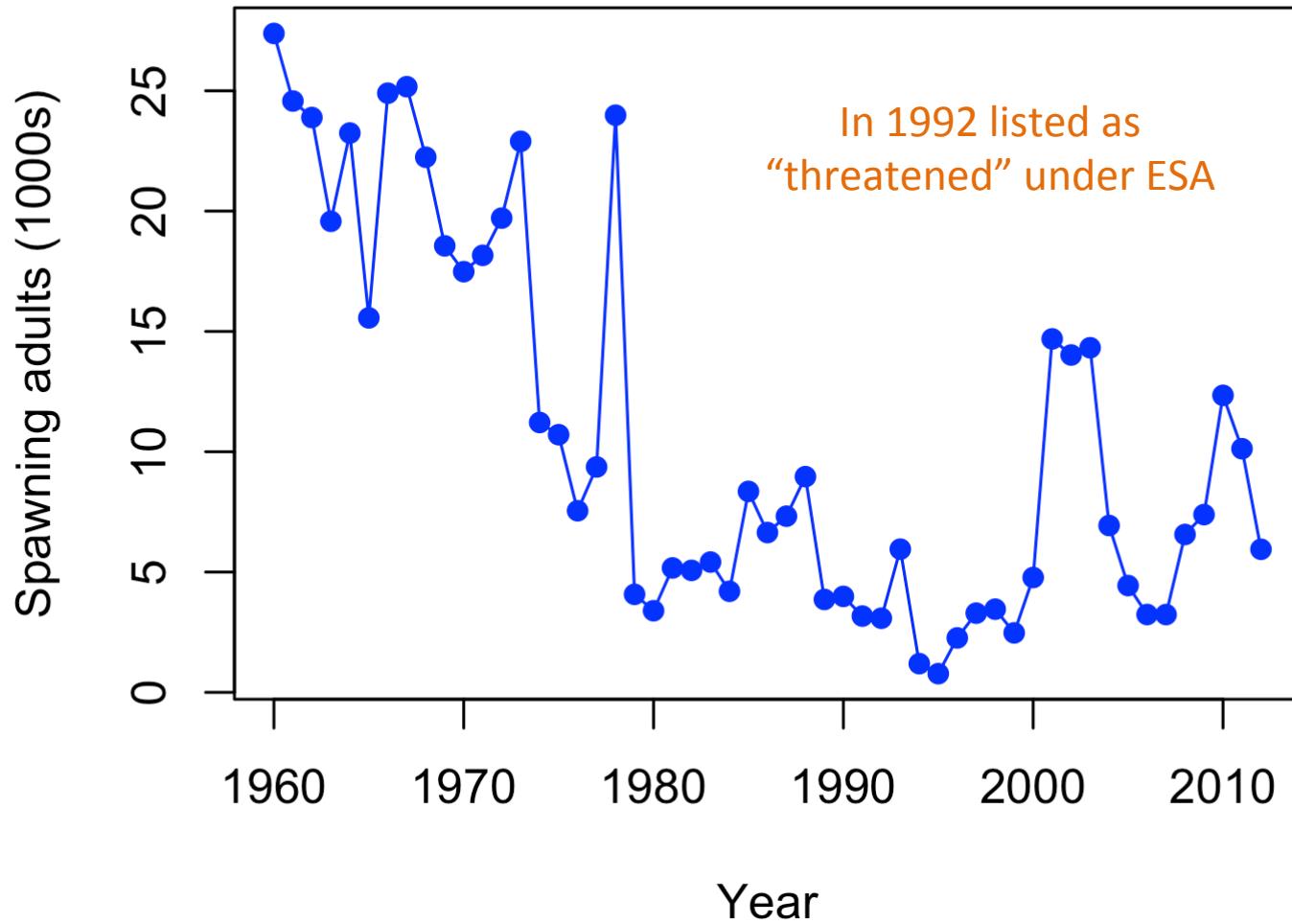
# Snake R spring/summer Chinook

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# Time series of abundance

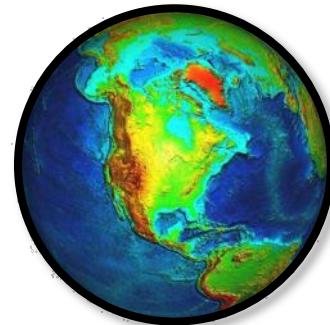
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## Reasons for decline

- ✓ Harvest
- ✓ Habitat degradation
- ✓ Hatchery operations
- ✓ Hydroelectric (& other) dams
- ✓ Climate
- ✓ Non-native species
- ✓ Marine-derived nutrients



# What are the recovery options?

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Recovery based on 4 “viable salmon” criteria:

1. Productivity
2. Abundance
3. Spatial structure
4. Diversity

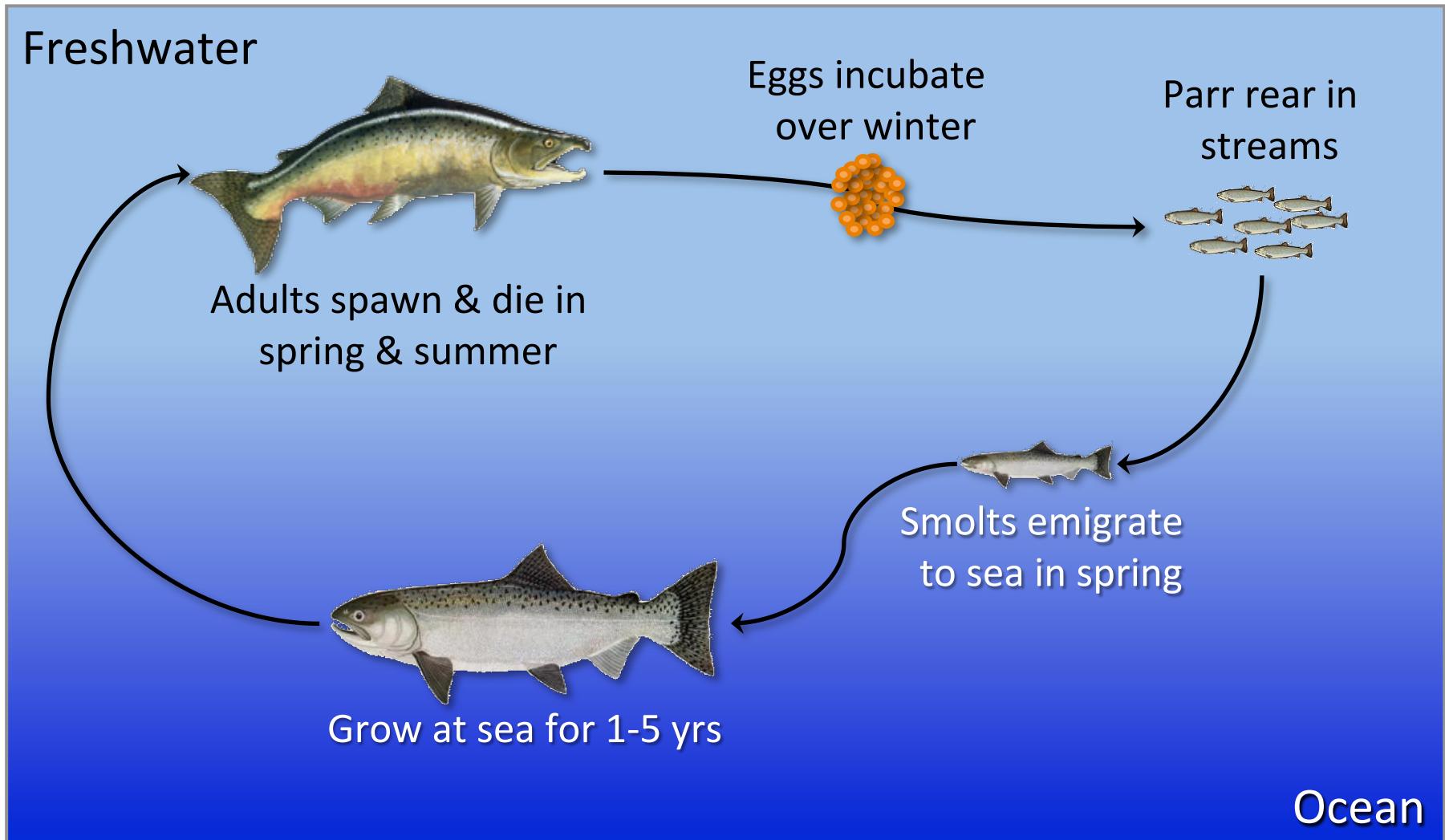
# Asset, market & risk-free indices

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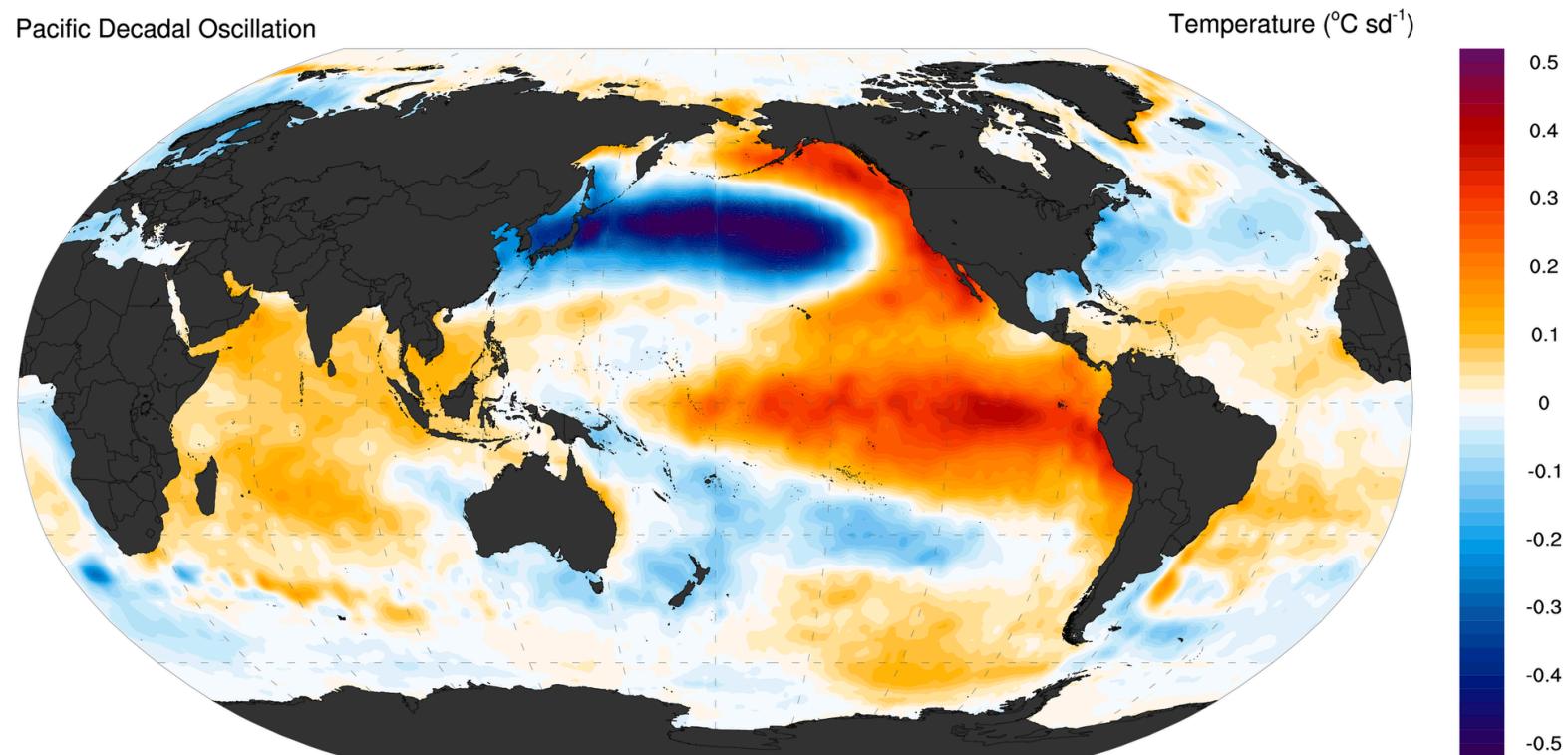
## Assets

- $\ln[\text{Recruits}/\text{Spawner}]$  (productivity)
- $\ln[\text{Spawners}/\text{ha}]$  (abundance)

# Salmon life cycle



# The Pacific Decadal Oscillation

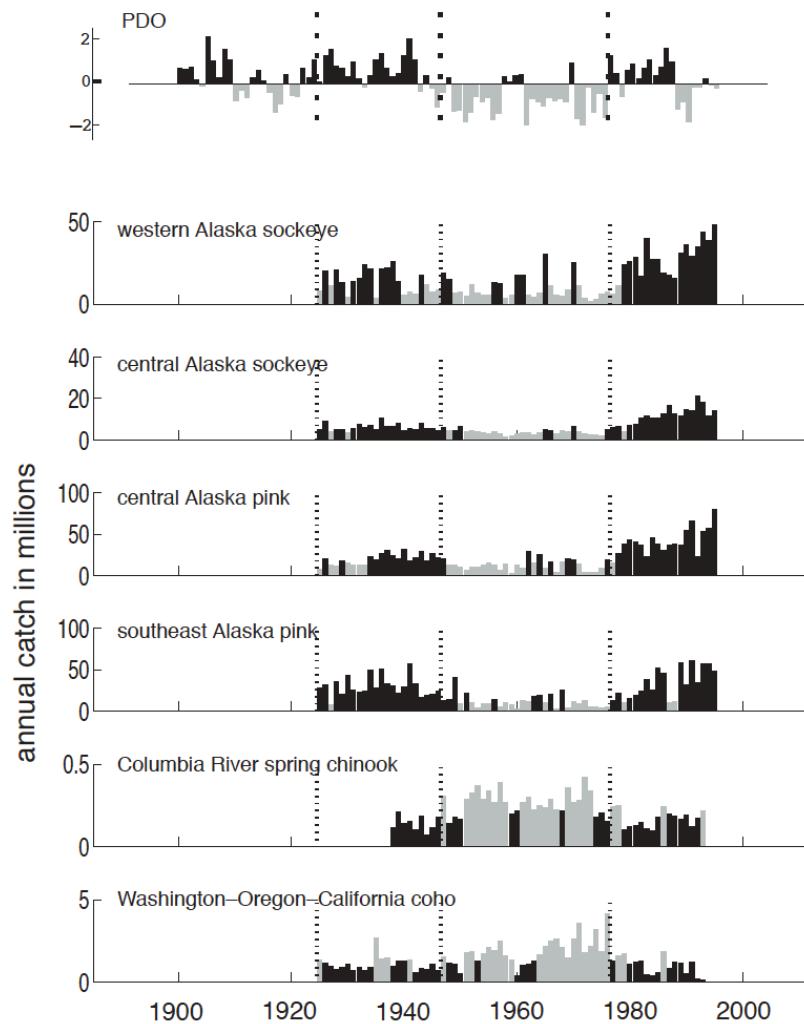


See Mantua et al. (1997) *Bull. Am. Meteor. Soc.*

# The Pacific Decadal Oscillation

2 take-home messages:

- 1) Salmon production related to ocean conditions
- 2) The PDO is a pretty good indicator of Ig-scale forcing



See Mantua et al. (1997) *Bull. Am. Meteor. Soc.*

# Asset, market & risk-free indices

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## Assets

- $\ln[\text{Recruits}/\text{Spawner}]$  (productivity)
- $\ln[\text{Spawners}/\text{ha}]$  (abundance)

## Market

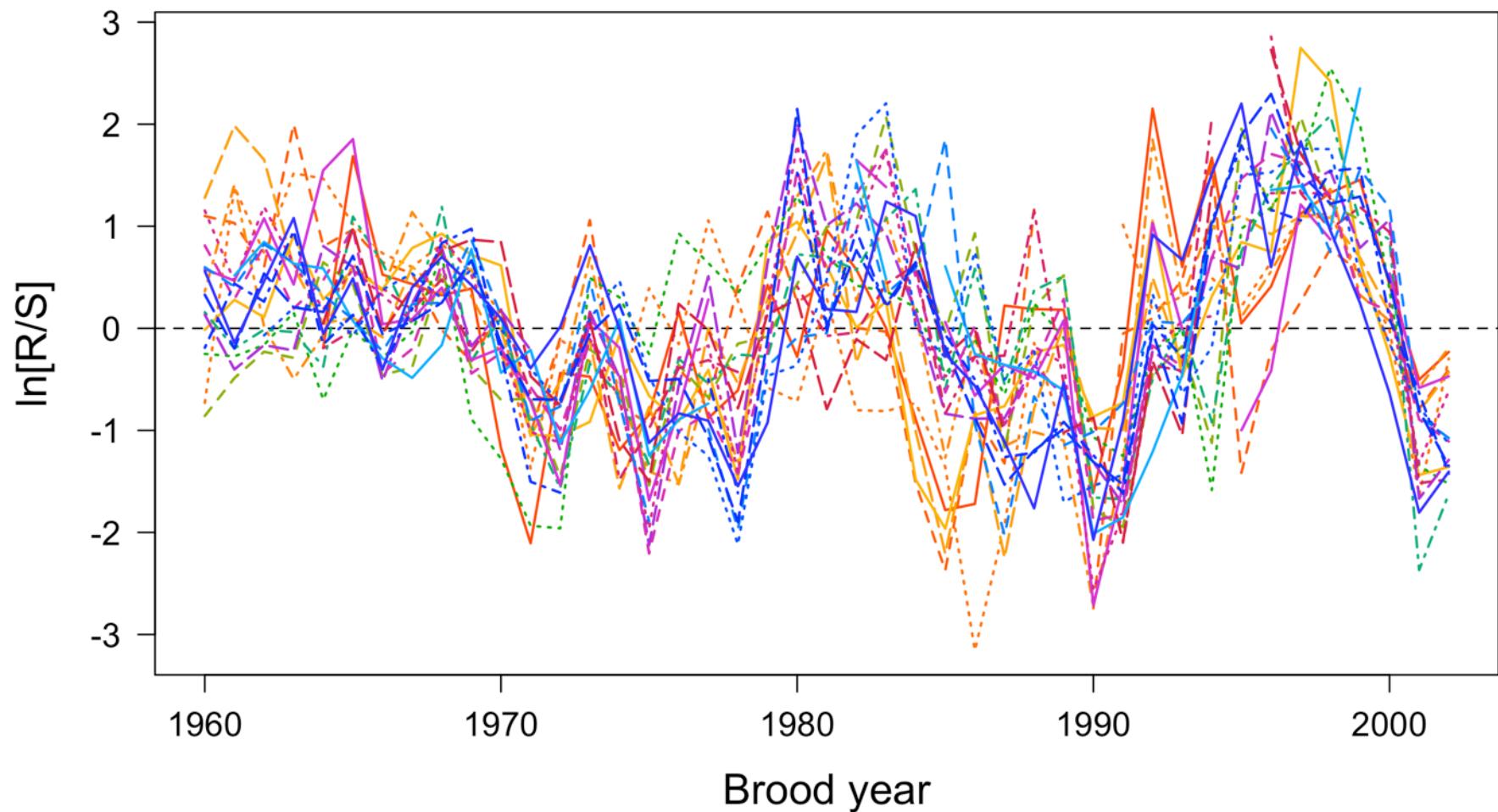
- Pacific Decadal Oscillation (PDO) in brood yr + 2

## Risk-free

- Replacement ( $\ln[R/S] = 0$ )
- $\ln[R/S]$  of John Day popn

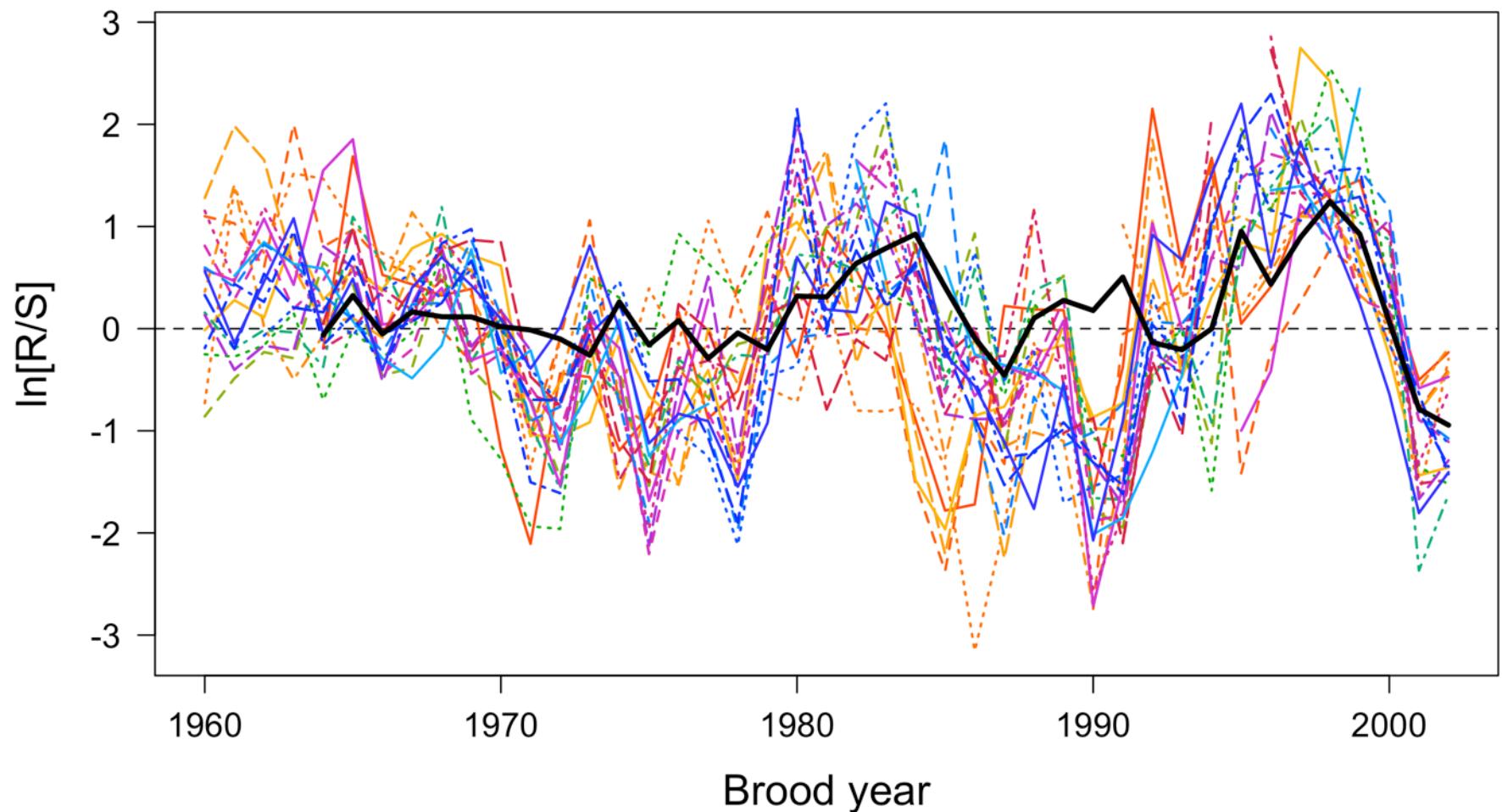
# Time series of “returns”

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# Time series of returns & risk-free

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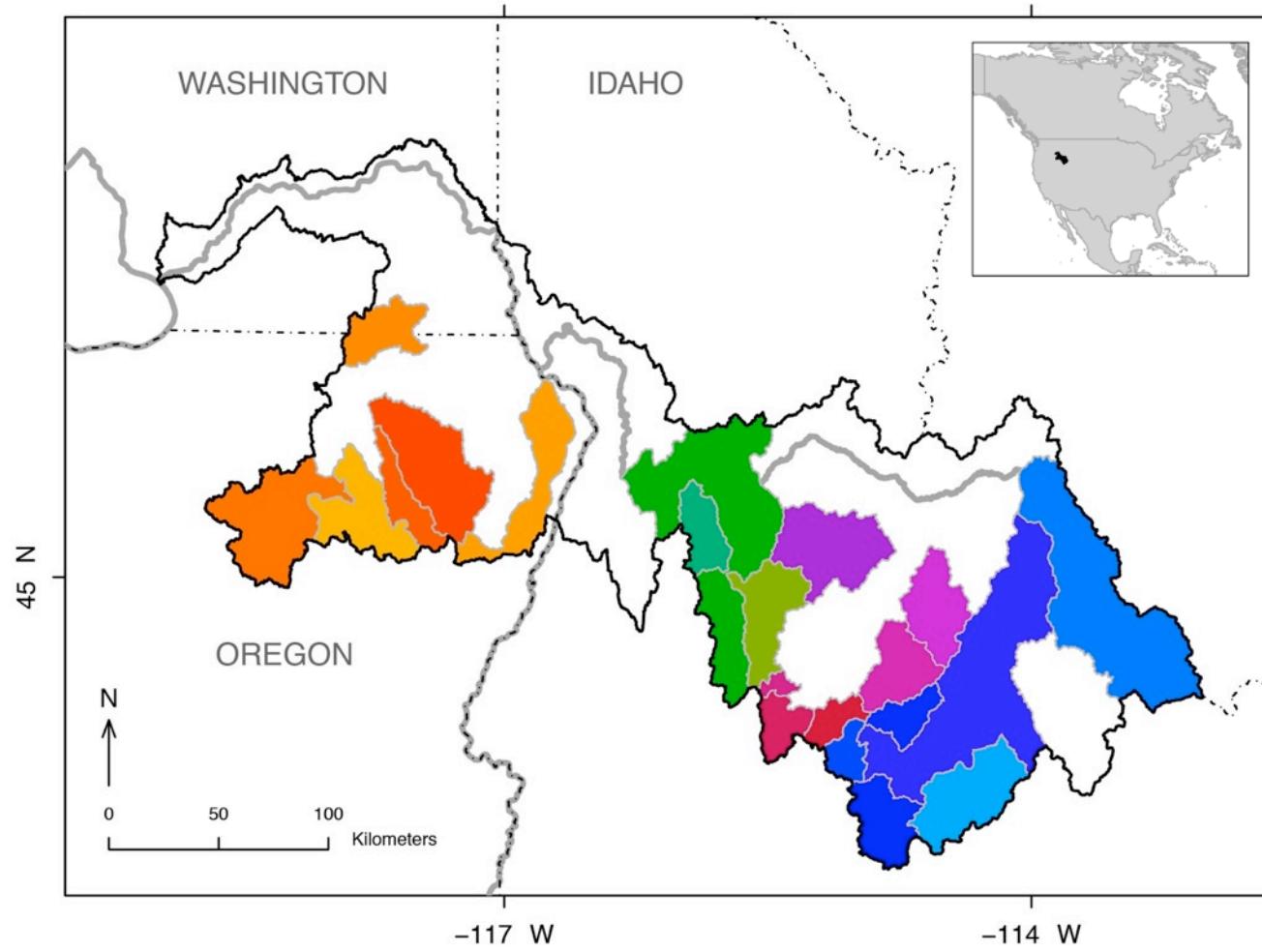
# Fitting the market model

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- In practice, the errors are often assumed to be Gaussian, and the model is solved via ordinary least squares
- This works well if the underlying relationship between asset & market is constant, but that rarely holds
- One option is to pass a moving window through the data, but window size affects accuracy & precision of  $\beta$
- Better choice is to use a dynamic linear model (DLM)

# Snake R spr/sum Chinook ESU

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# Multivariate DLM

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- Here we will examine multiple assets at once, so we need a multivariate (response) DLM
- First, the obs eqn

$$r_{a,t} = \mathbf{R}_{m,t} \boldsymbol{\theta}_{a,t} + v_{a,t} \quad v_t \sim N(0, \sigma)$$

becomes

$$\mathbf{R}_t = (\mathbf{R}_{m,t} \otimes \mathbf{I}_n) \boldsymbol{\theta}_t + \mathbf{v}_t \quad \mathbf{v}_t \sim MVN(\mathbf{0}, \Sigma)$$

# Multivariate DLM – obs eqn

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$$\underline{\mathbf{R}_t} = \underline{(\mathbf{R}_{m,t} \otimes \mathbf{I}_n)} \boldsymbol{\theta}_t + \mathbf{v}_t \quad \mathbf{v}_t \sim \text{MVN}(\mathbf{0}, \boldsymbol{\Sigma})$$

$$\begin{bmatrix} r_{1,t} \\ \vdots \\ r_{n,t} \end{bmatrix} = \begin{bmatrix} 1 & R_{m,t} \end{bmatrix} \otimes \begin{bmatrix} 1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

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# Multivariate DLM – obs eqn

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$$\mathbf{R}_t = \underbrace{(\mathbf{R}_{m,t} \otimes \mathbf{I}_n) \boldsymbol{\theta}_t}_{\text{matrix}} + \underbrace{\mathbf{v}_t}_{\text{vector}} \quad \mathbf{v}_t \sim \text{MVN}(\mathbf{0}, \boldsymbol{\Sigma})$$

$$\begin{bmatrix} r_{1,t} \\ \vdots \\ r_{n,t} \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 0 & 0 & r_{m,t} & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & 1 & 0 & 0 & r_{m,t} \end{bmatrix}}_{\text{matrix}} \begin{bmatrix} \alpha_{1,t} \\ \vdots \\ \alpha_{n,t} \\ \beta_{1,t} \\ \vdots \\ \beta_{n,t} \end{bmatrix} + \underbrace{\begin{bmatrix} v_{1,t} \\ \vdots \\ v_{n,t} \end{bmatrix}}_{\text{vector}}$$

# Multivariate DLM – evolution eqn

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- The evolution eqn

$$\boldsymbol{\theta}_{a,t} = \mathbf{G}_t \boldsymbol{\theta}_{a,t-1} + \mathbf{w}_t \quad \mathbf{w}_t \sim \text{MVN}(\mathbf{0}, \mathbf{Q})$$

becomes

$$\boldsymbol{\theta}_t = (\mathbf{G}_t \otimes \mathbf{I}_n) \boldsymbol{\theta}_{t-1} + \mathbf{w}_t \quad \mathbf{w}_t \sim \text{MVN}(\mathbf{0}, \mathbf{Q})$$

$$\mathbf{G}_t = \mathbf{I}_2 \Rightarrow \mathbf{G}_t \otimes \mathbf{I}_n = \mathbf{I}_{2n}$$

$$\boldsymbol{\theta}_t = \boldsymbol{\theta}_{t-1} + \mathbf{w}_t$$

# Multivariate DLM – evolution eqn

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$$\boldsymbol{\theta}_t = \boldsymbol{\theta}_{t-1} + \mathbf{w}_t \quad \mathbf{w}_t \sim \text{MVN}(\mathbf{0}, \mathbf{Q})$$

$$\begin{bmatrix} \alpha_{1,t} \\ \vdots \\ \alpha_{n,t} \\ \beta_{1,t} \\ \vdots \\ \beta_{n,t} \end{bmatrix} = \begin{bmatrix} \alpha_{1,t-1} \\ \vdots \\ \alpha_{n,t-1} \\ \beta_{1,t-1} \\ \vdots \\ \beta_{n,t-1} \end{bmatrix} + \begin{bmatrix} w_{1,t}^{(\alpha)} \\ \vdots \\ w_{n,t}^{(\alpha)} \\ w_{1,t}^{(\beta)} \\ \vdots \\ w_{n,t}^{(\beta)} \end{bmatrix}$$

# Multivariate DLM – evolution eqn

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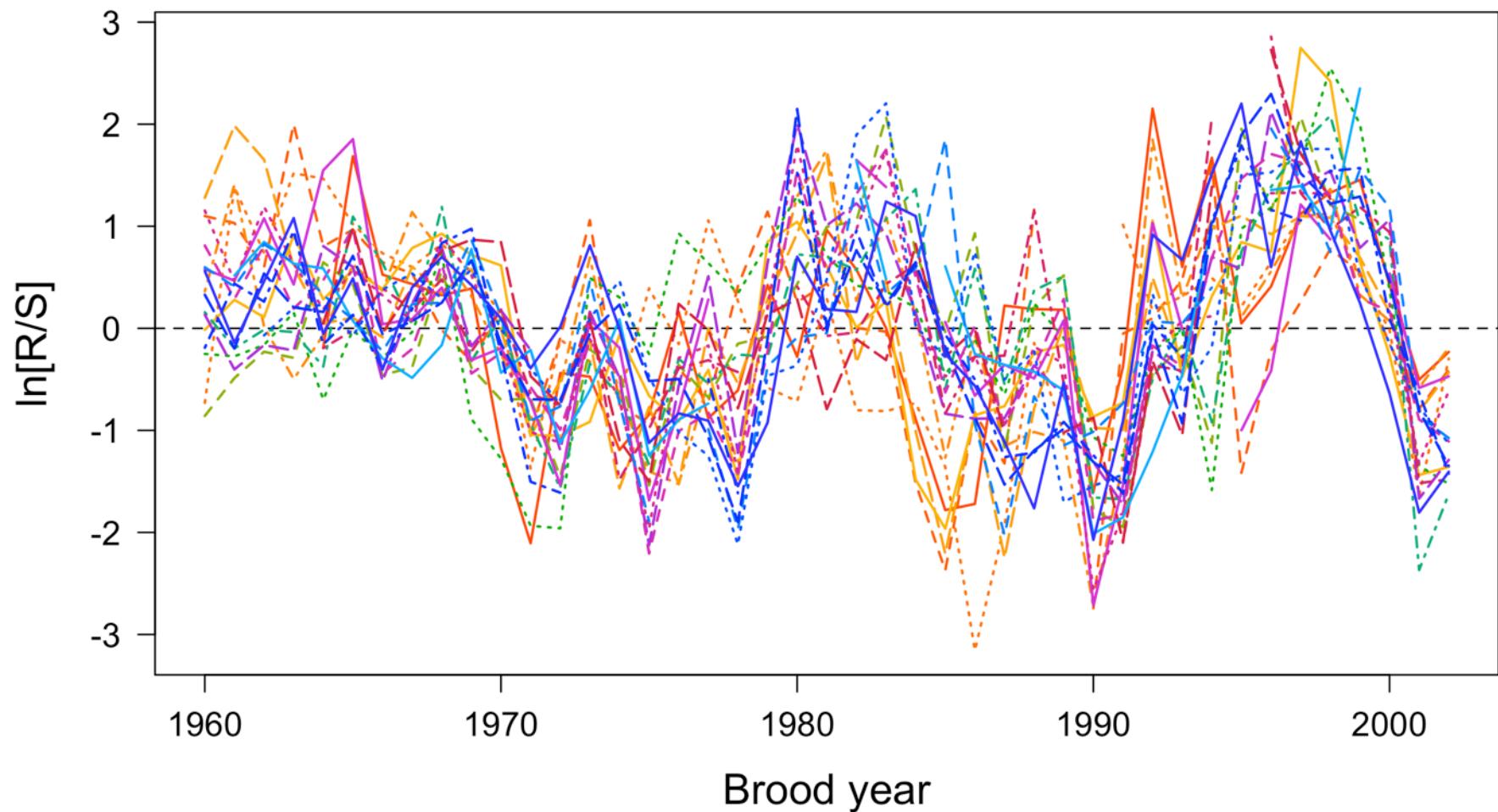
$$\boldsymbol{\theta}_t = \boldsymbol{\theta}_{t-1} + \mathbf{w}_t \quad \mathbf{w}_t \sim \text{MVN}(\mathbf{0}, \underline{\mathbf{Q}})$$

$$\mathbf{Q} = \begin{bmatrix} \mathbf{Q}^{(\alpha)} & \mathbf{0} \\ \mathbf{0} & \mathbf{Q}^{(\beta)} \end{bmatrix}$$

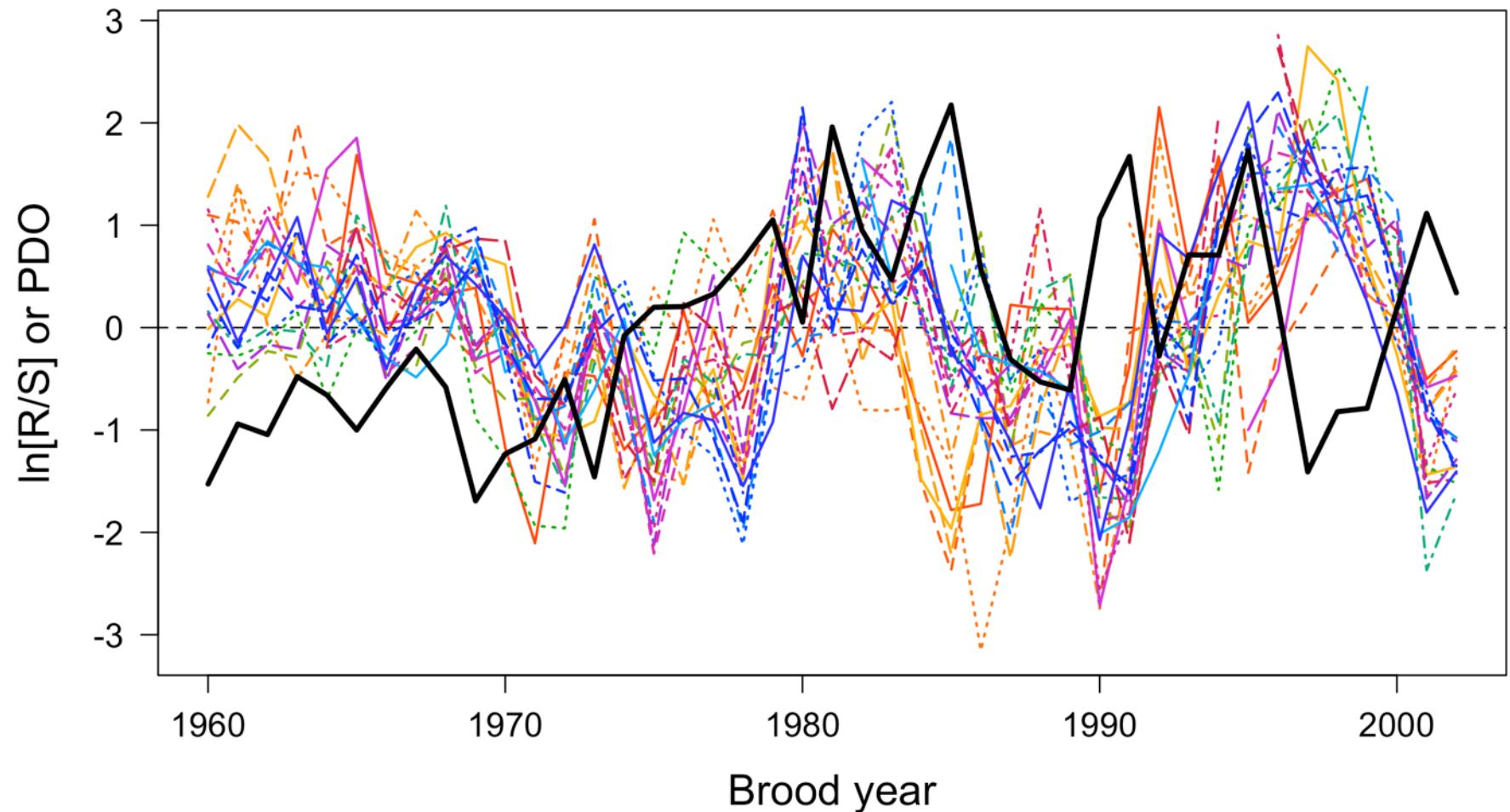
$$\mathbf{Q}^{(\cdot)} = \begin{bmatrix} q_1^{(\cdot)} & c_{21}^{(\cdot)} & \cdots & c_{n1}^{(\cdot)} \\ c_{12}^{(\cdot)} & q_2^{(\cdot)} & & c_{n2}^{(\cdot)} \\ \vdots & & \ddots & \vdots \\ c_{1n}^{(\cdot)} & c_{2n}^{(\cdot)} & \cdots & q_n^{(\cdot)} \end{bmatrix}$$

# Time series of “returns”

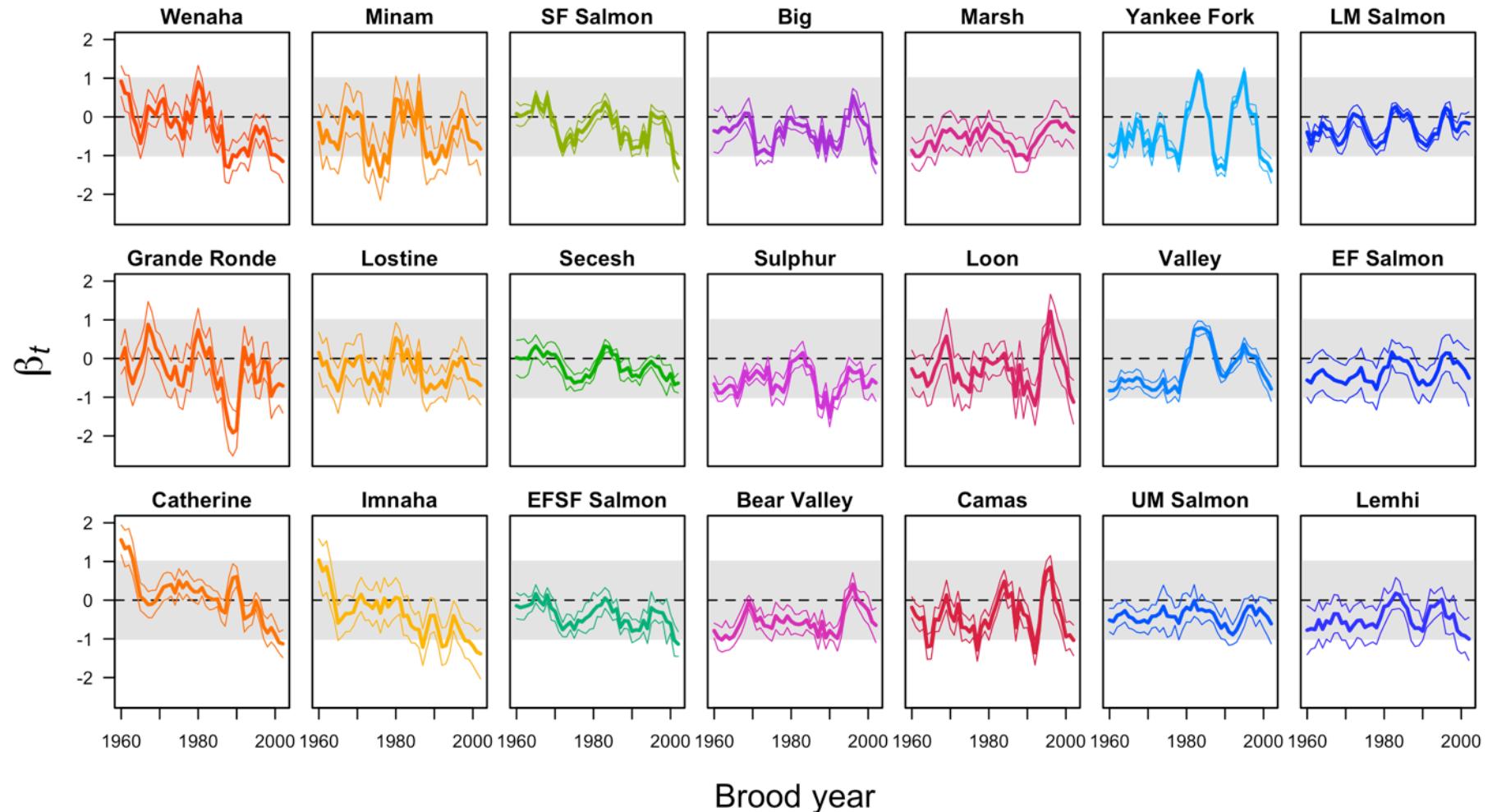
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# Time series of “returns” & PDO



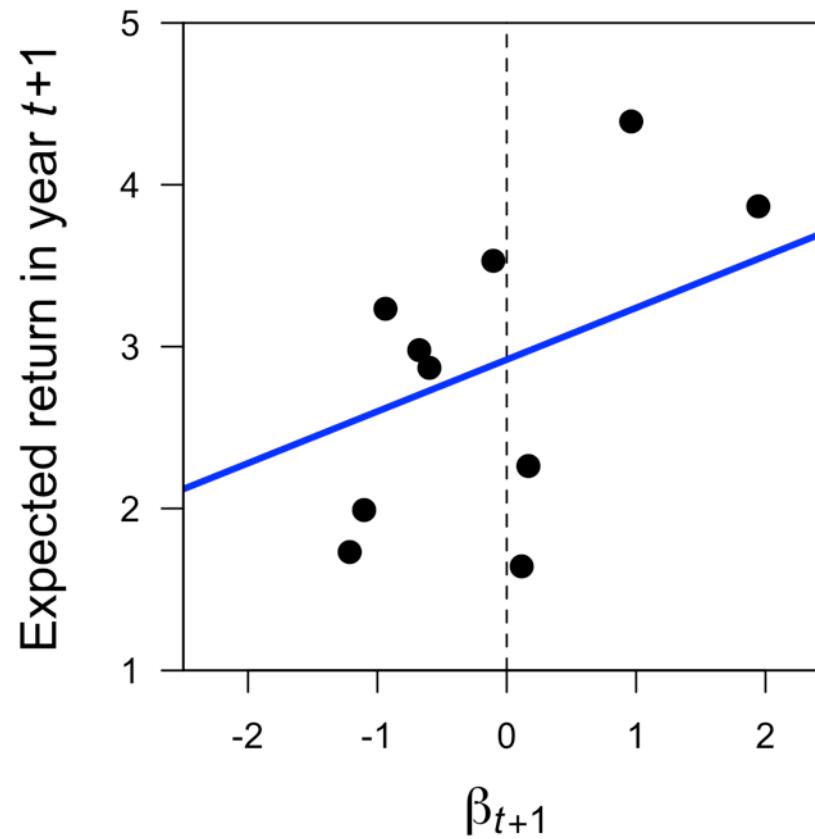
# Results – time series of betas



Market = PDO

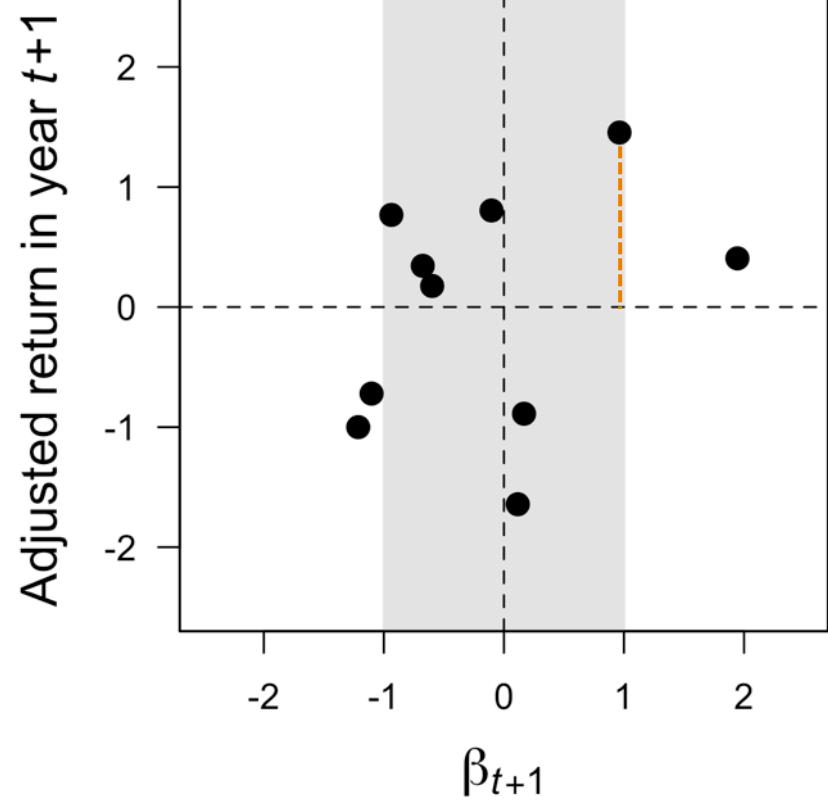
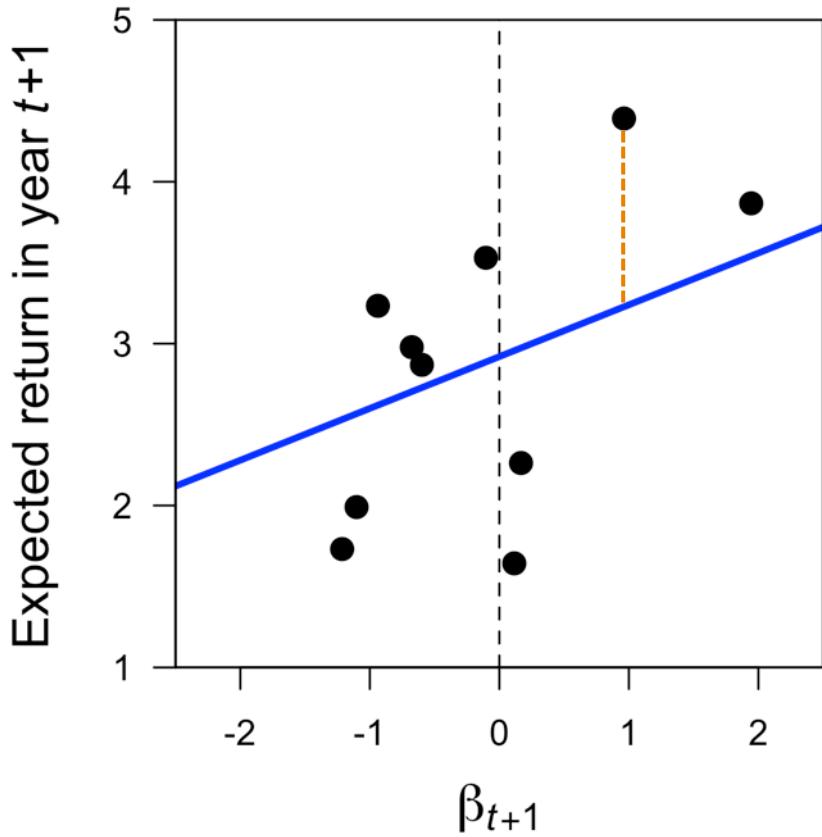
# Security market line

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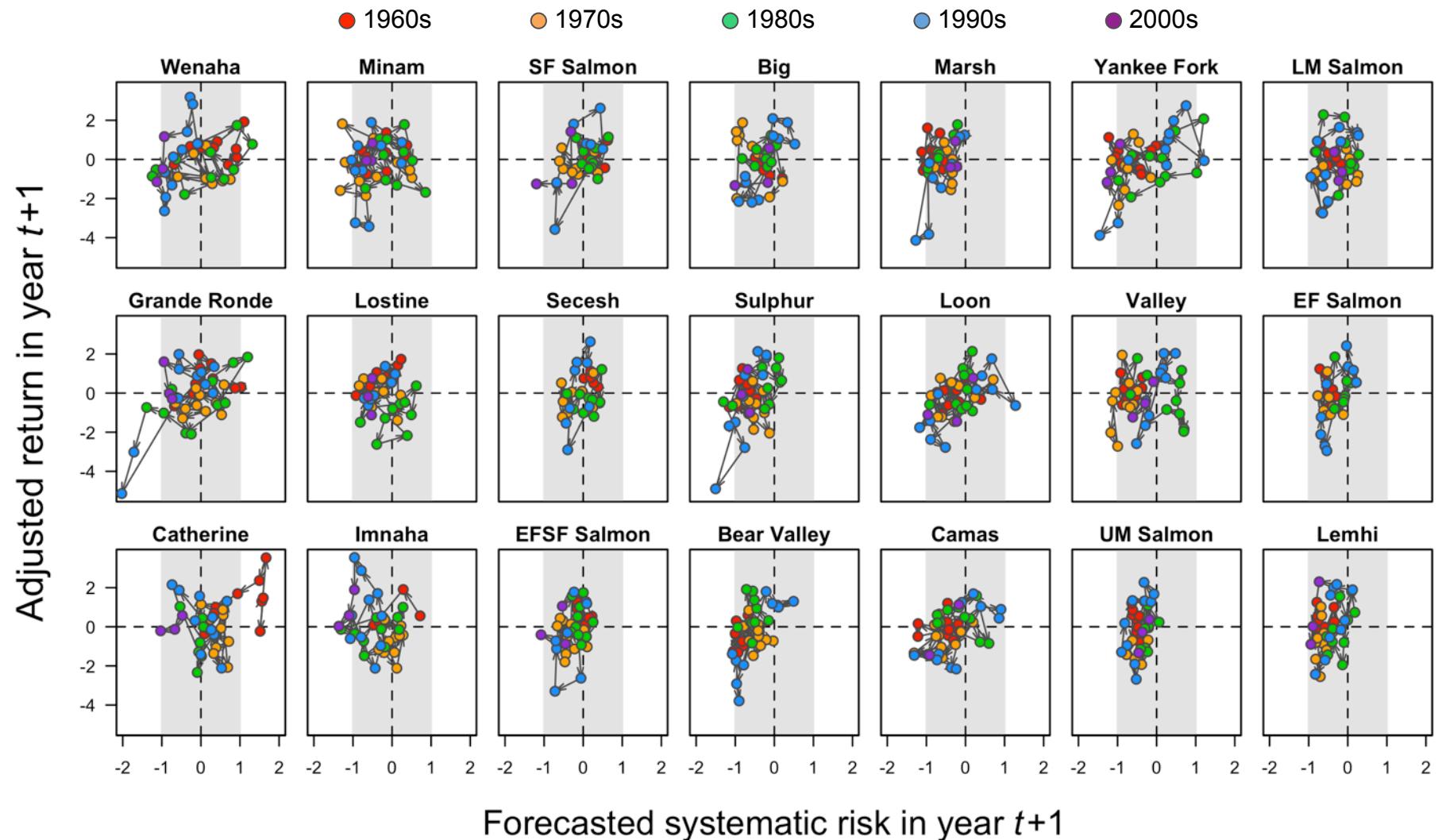


# Security market line

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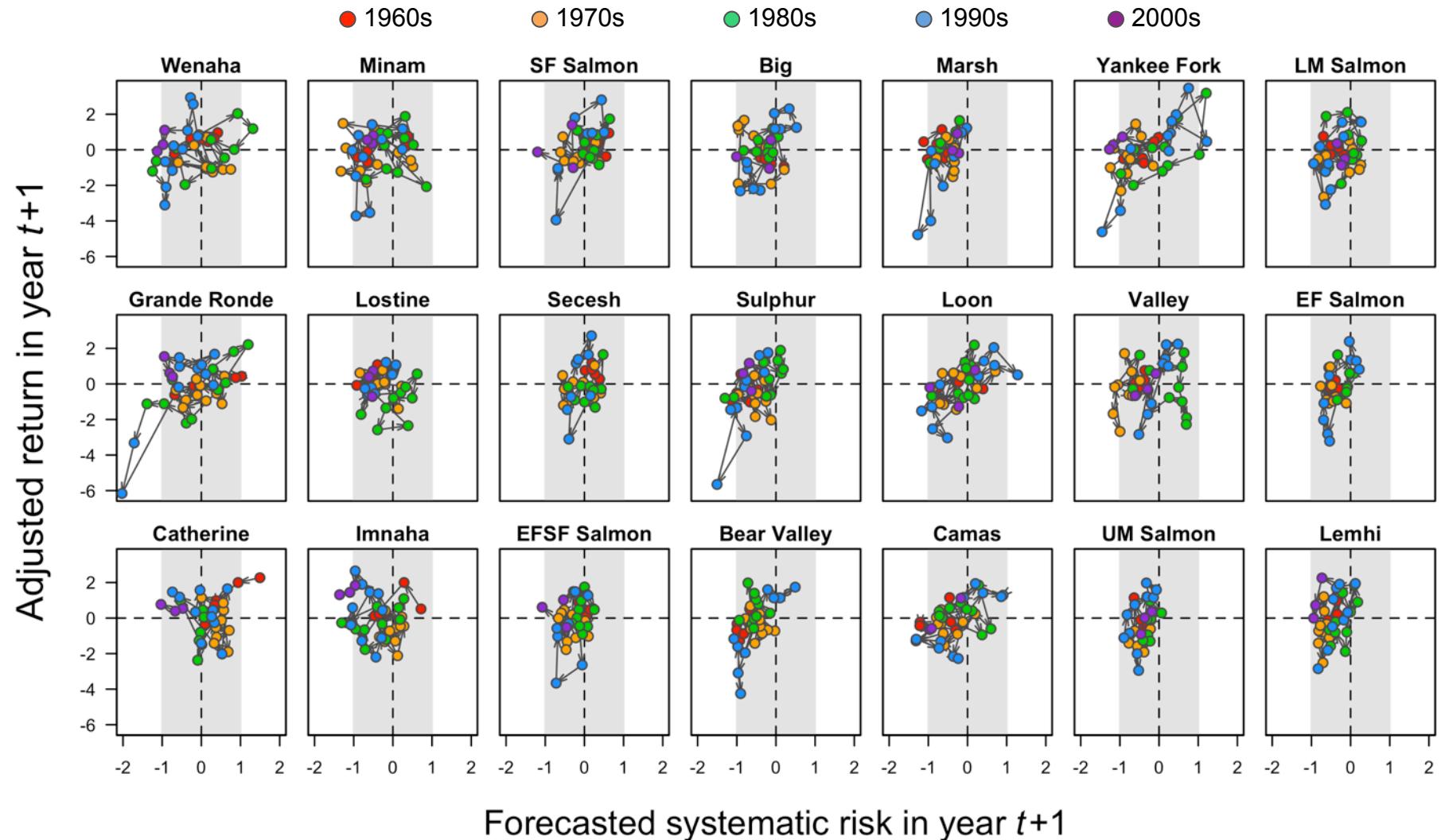


# Results – CAPM



*Market = PDO; Risk-free = 0*

# Results – CAPM



*Market = PDO; Risk-free = John Day*

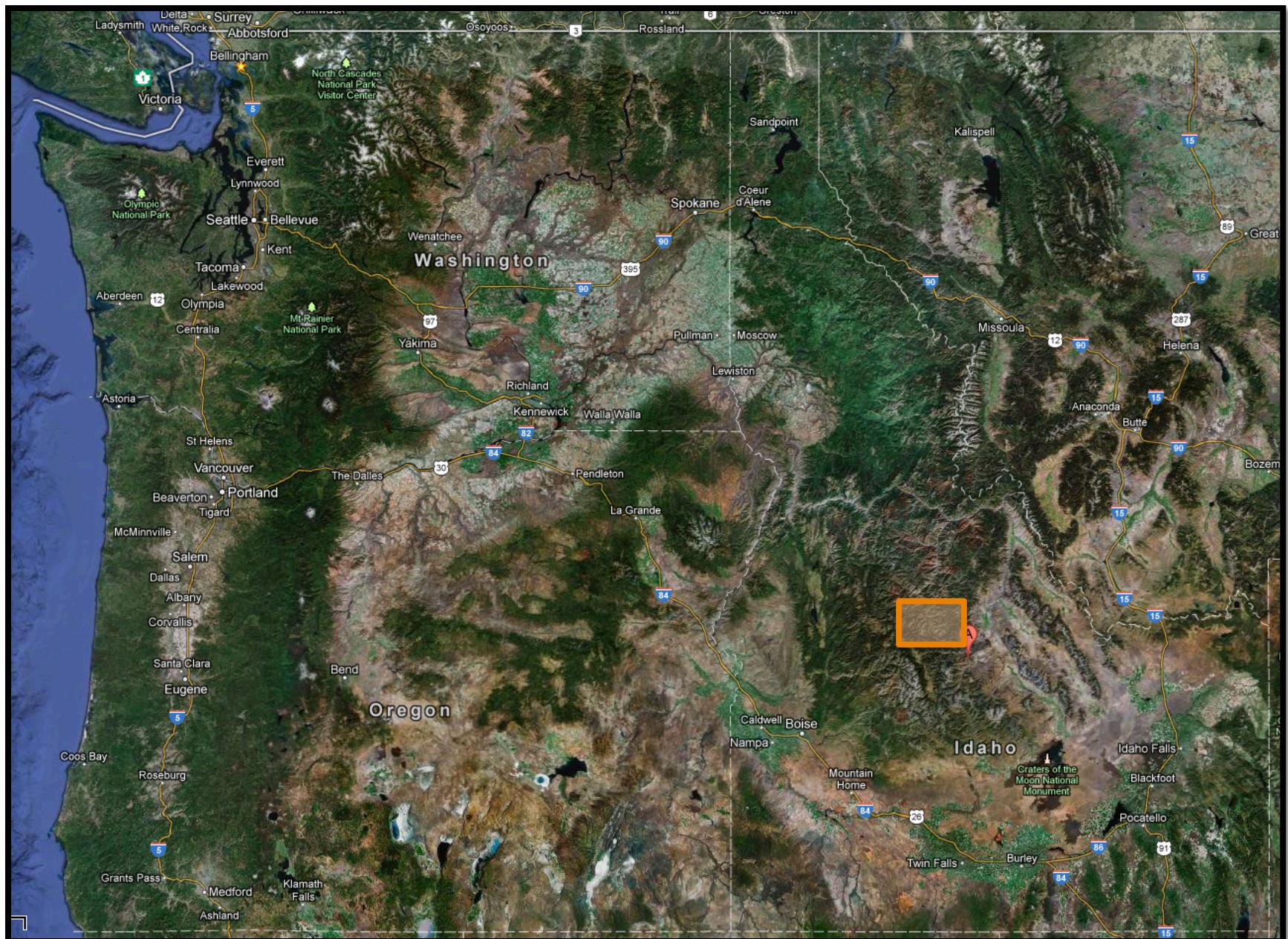


Image by Google



Image by Google

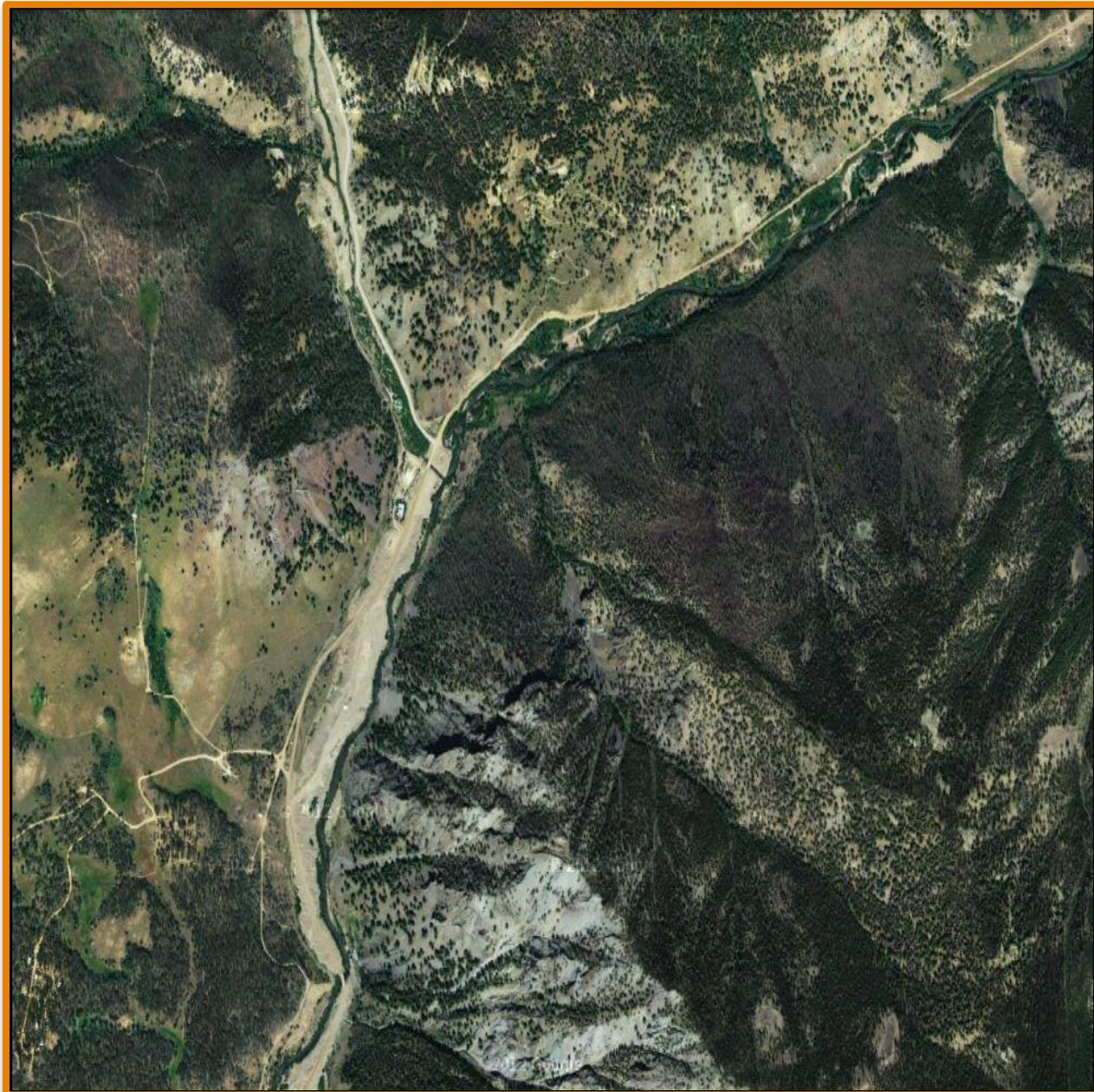


Image by Google



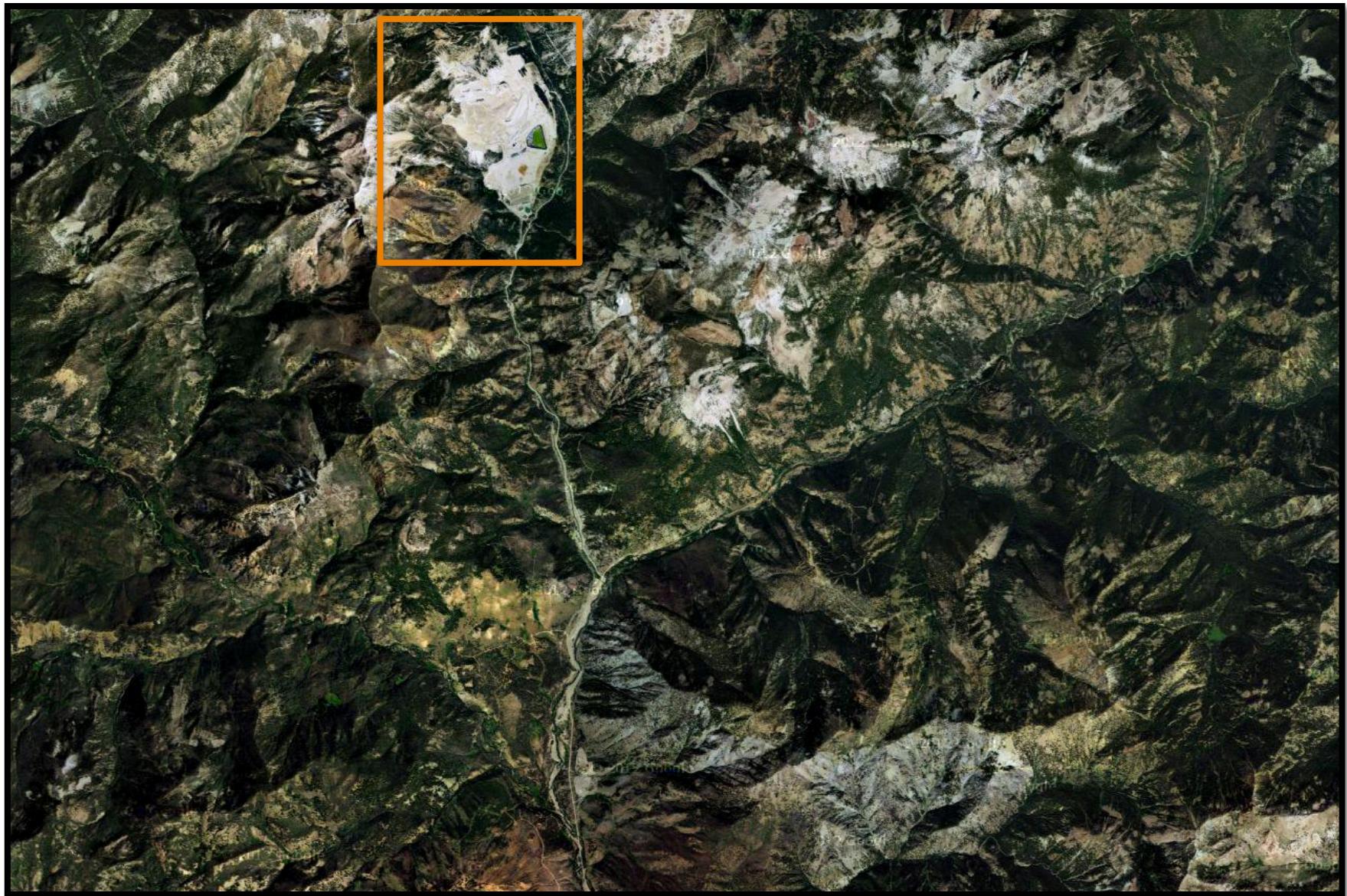




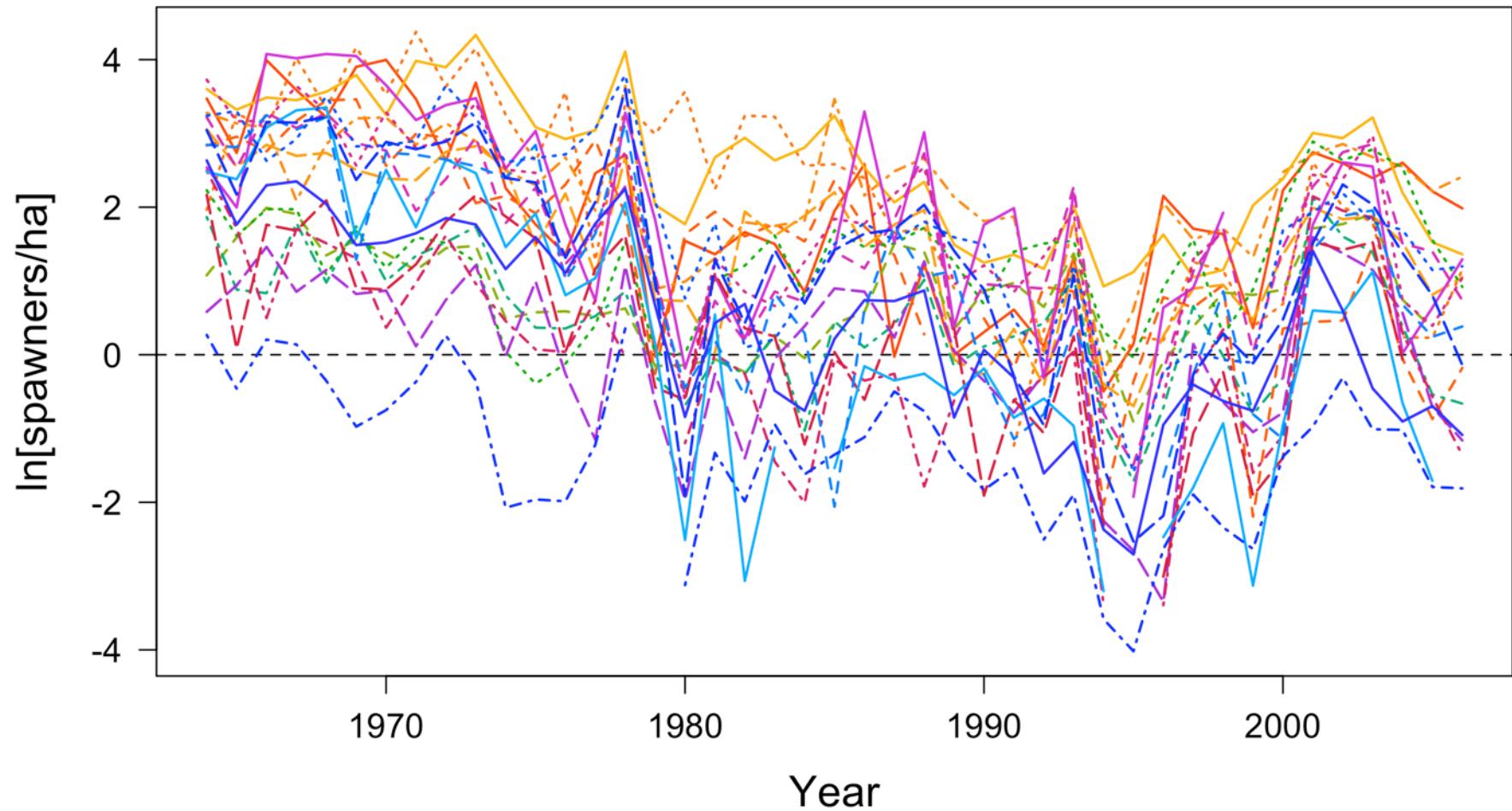




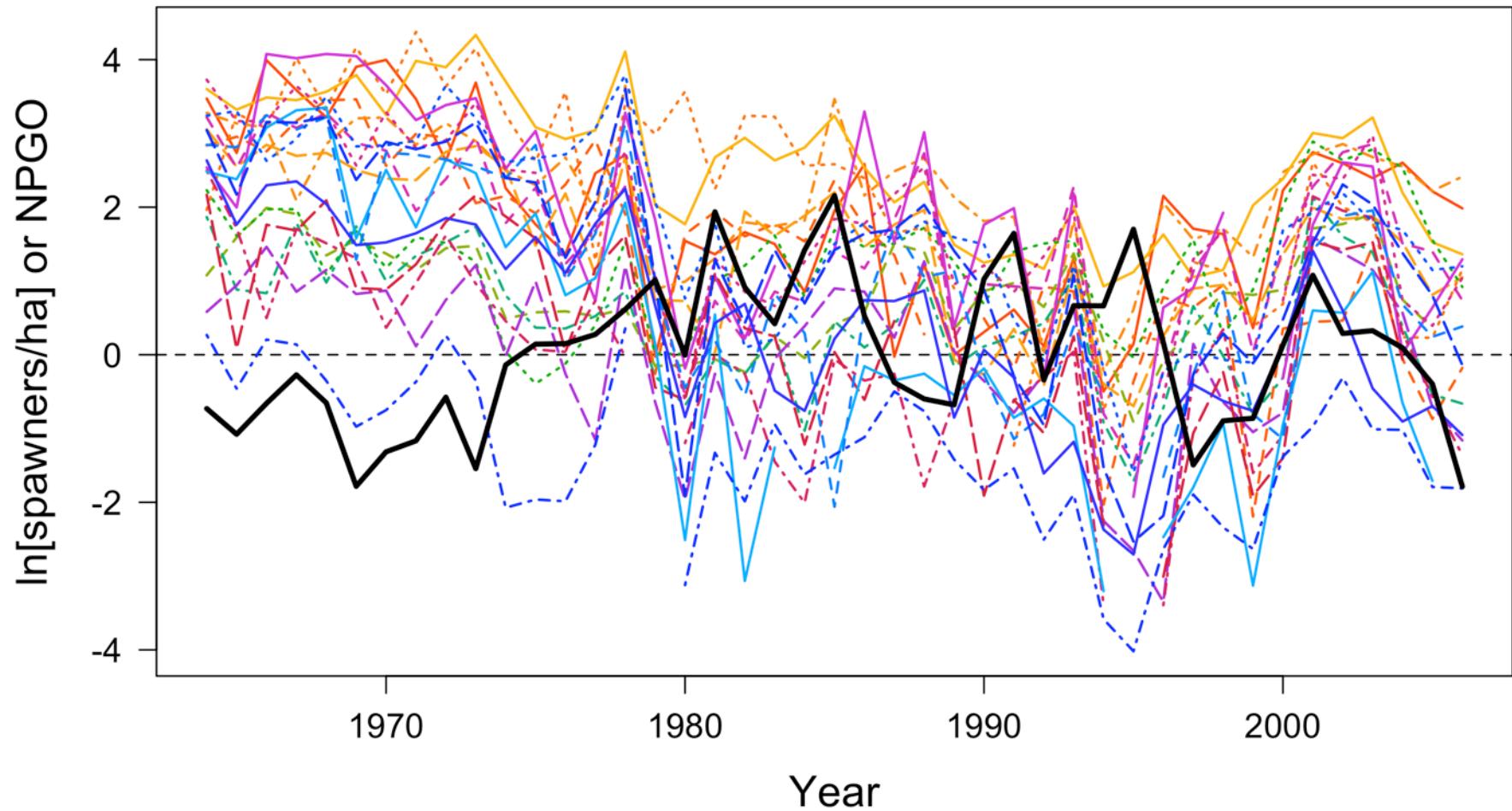
Photo by Jack Molan

# Time series of returns (Density)

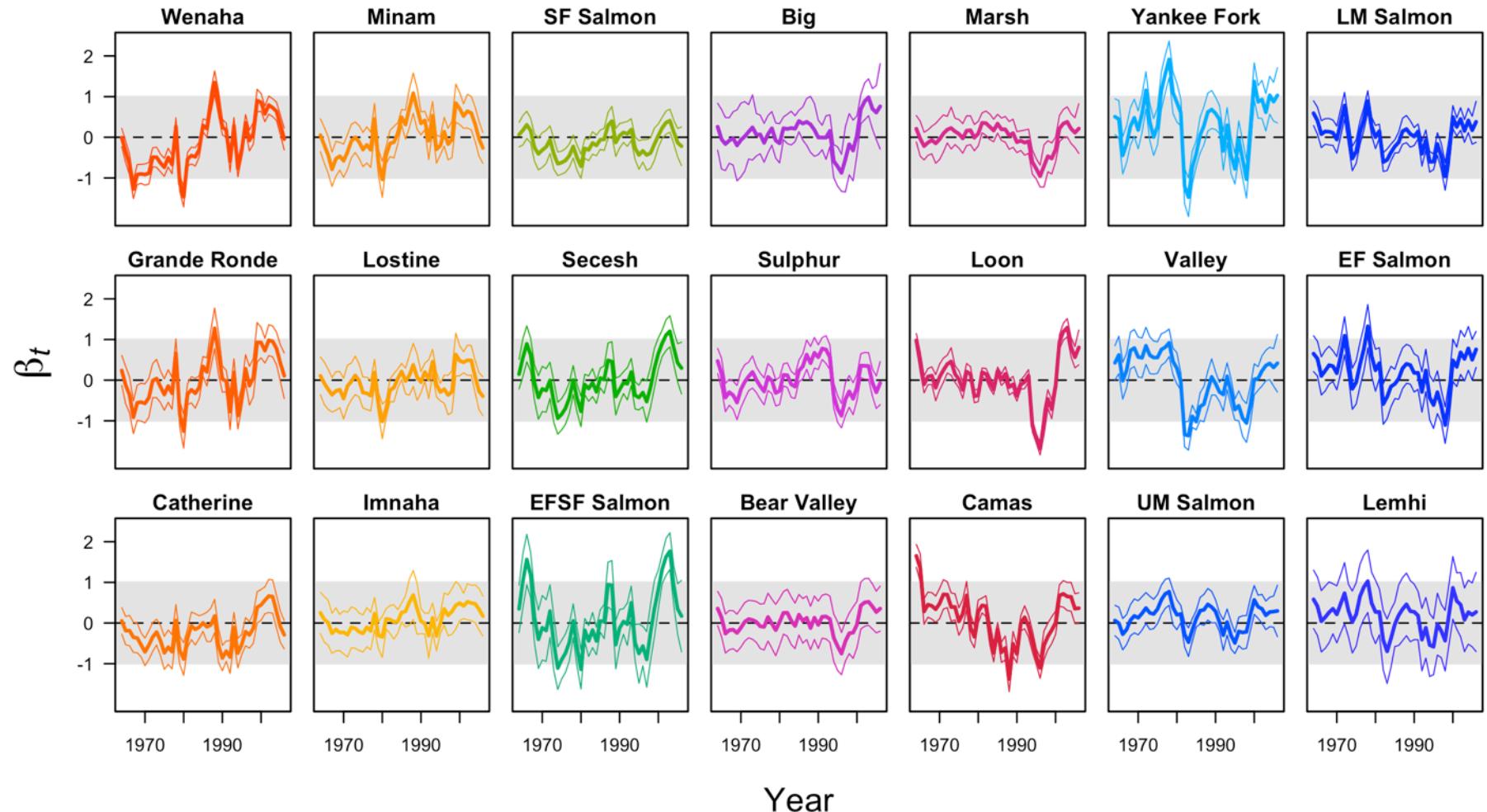
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# Time series of returns & PDO

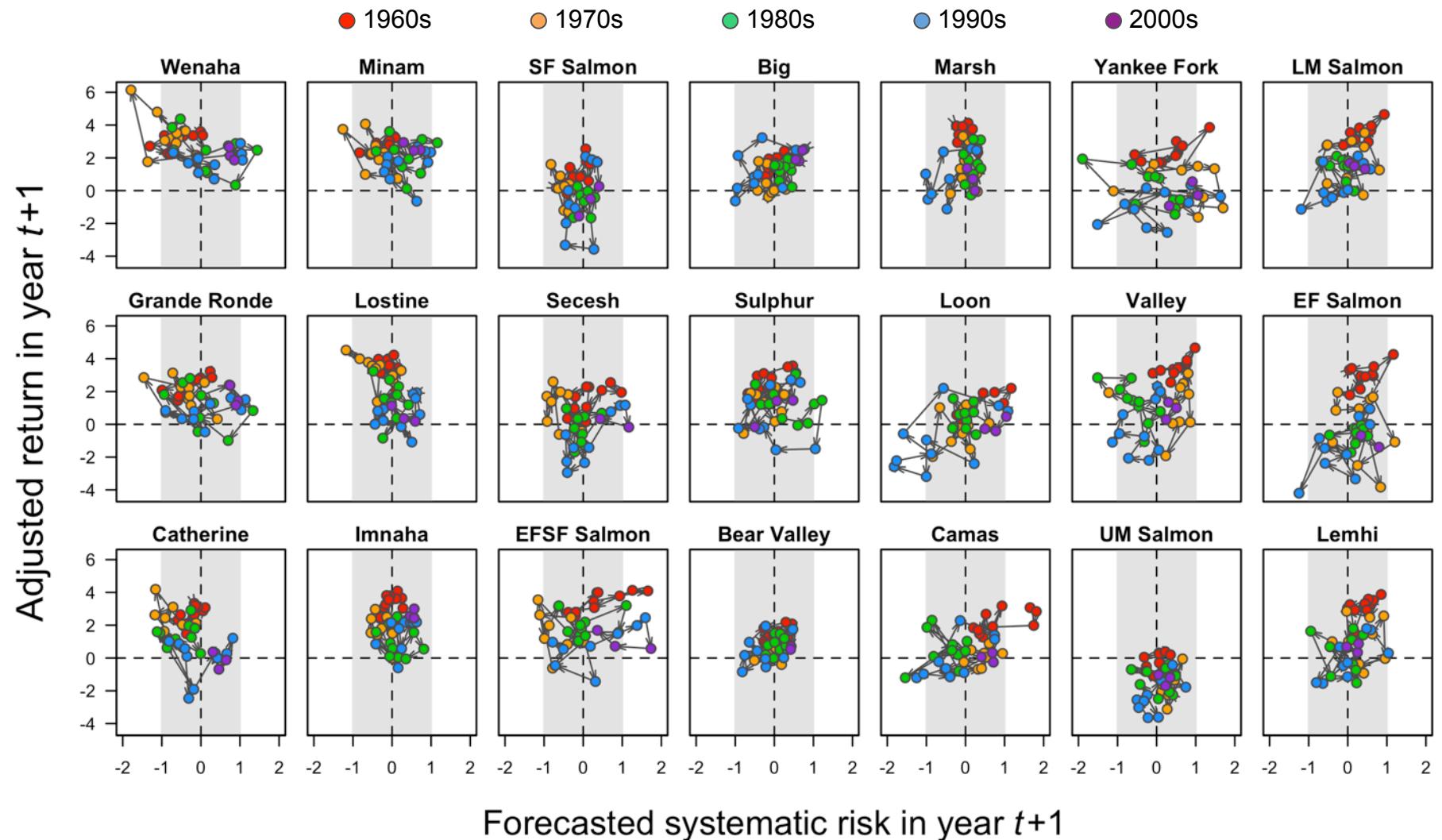


# Results – time series of betas



Market = PDO

# Results – CAPM



Market = PDO; Risk-free = 0