

# R Package: Metafolio

**Presented by:** Alicia Canales, Stephanie Luu,  
Olivia Somhegyi



# What does the package do?

Metafolio simulates meta populations and applies **financial portfolio optimization concepts** to those metapopulations. **This is a package that optimizes spatial conservation strategies.**

Goal of portfolio optimization is to **increase the return** on investment from portfolios and **minimize variance** on the returning portfolios.

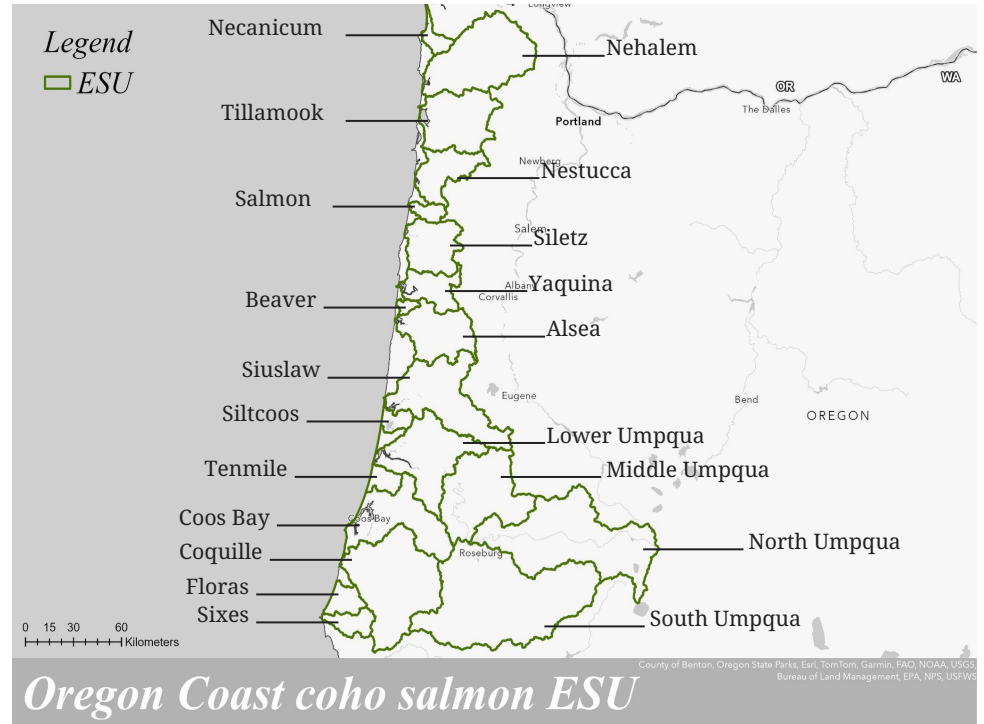
Metafolio is best suited for salmon populations, but can be applied to other species potentially.



# Background: What is a metapopulation? ?

Metapopulation is a **group of spatially separated populations** that **interact or exchange genetic material** at some level.

The Oregon Coast coho salmon has **21 individual populations**.



# What type of problems can metafolio address? 💡

Metafolio can help us understand:

1. How to make **optimal investments** to the metapopulations for spatial recovery planning
2. How to **choose optimal portfolios** that promote population stability (**reduce variance and increase population abundance by quantifying carrying capacity and productivity**)
3. How to identify optimal vs non optimal portfolios
4. What are the impacts of restorative actions on fish populations?



WOW!!



# A (very) brief overview of Portfolio Theory



**The overall goal of portfolio theory is to maximize returns for a given level of variance.**

Three variables need to be defined: the portfolio, the assets, and the weights. In the context of metafolio, they are defined as:

**Portfolio:** Metapopulation

**Assets:** Individual populations that make up the metapopulation

**Weights:** Capacity of the population

# Model Overview



## Ricker Model: Spawner-Returner Relationship

$$R_{i(t+1)} = S_{i(t)} e^{a_{i(t)} \left(1 - \frac{S_{i(t)}}{b_i}\right) + w_{i(t)}}$$

### Where:

$R$  = Fish Return

$S$  = Stock Abundance

$a$  = Productivity

$b$  = Carrying Capacity

$w$  = weight

$i$  = population

# Model Overview



## ## (2) Main functions within the package:

*meta\_sim()* :Runs a single iteration simulation. Can be baseline scenario prior to conservation intervention

*run\_cons\_plans()* : Used to run meta\_sim() for several iterations across multiple conservation strategies

## ## (2) Optimization functions:

*create\_asset\_weights()* :Creates a matrix of asset weights for optimization

*monte\_carlo\_portfolios()* :Runs the optimization using the weights created from create\_asset\_weights()

## ## (3) Plotting functions:

*plot\_sim\_ts()*

*plot\_cons\_plans()*

*plot\_efficient\_frontier()*

# Type of Data Needed?



1. Matrix of population abundances over time
2. Environmental parameters specific to a given spatial strategy of conservation
  - a. Thermal Tolerance



# Example Code continued...



## Starting with a base scenario:

```
## Example simulation - create environmental parameters
```

```
```{r}
arma_env_params <- list(mean_value = 16, # mean
                        ar = 0.1, # auto regressive parameter
                        sigma_env = 2, # std deviation of env signal
                        ma = 0) # moving average parameter
```
```

```
## Run base simulation
```

```
```{r}
## define base simulation:
base1 <- meta_sim(n_pop = 10, # simulate 10 population
                 env_params = arma_env_params,
                 env_type = "arma",
                 assess_freq = 5) # reassess the fishery every 5 years

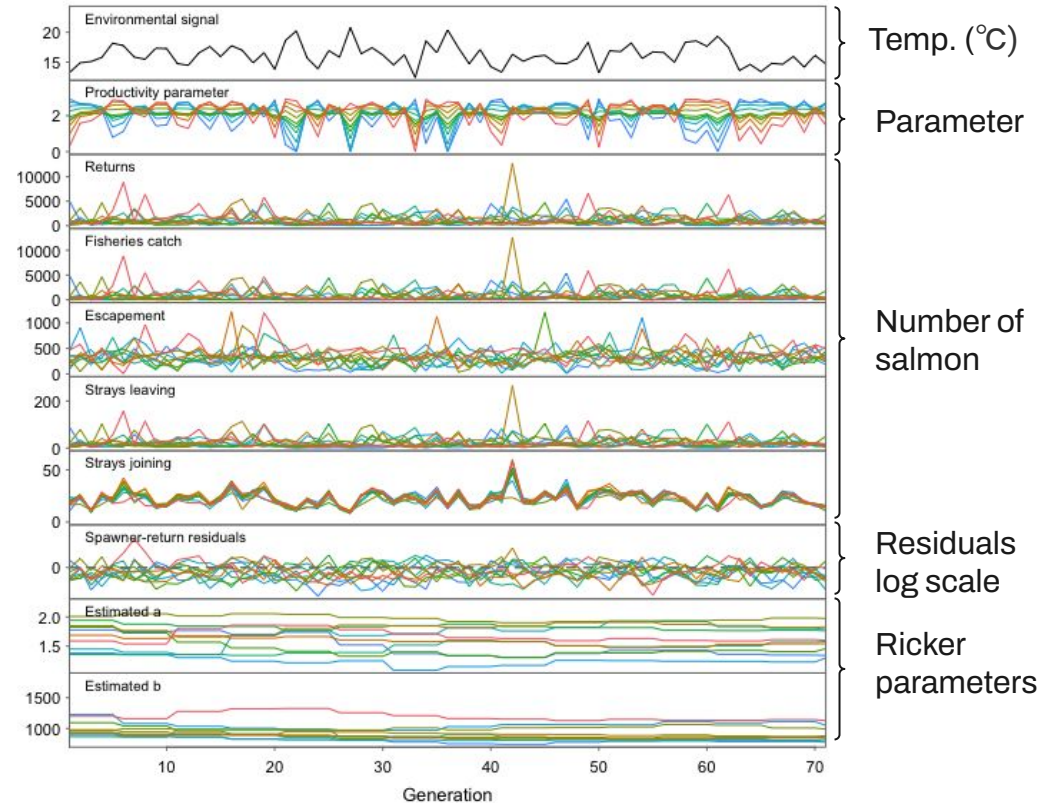
# plot base simulation
plot_sim_ts(base1, years_to_show = 70, burn = 30)
```
```

# Example Code Continued...



## Plot base scenario:

Baseline scenario at stationary environmental stochasticity and parameter values



# Example Code Continued...

Using the baseline scenario, we can explore different conservation strategies by creating weights

```
## Exploring prioritization strategies
```

```
```{r}
```

```
## create conservation plans
```

```
# balanced: conserve response diversity from across all possible responses
```

```
# one half: conserve one half of the response diversity
```

```
w_plans <- list()
```

```
w_plans[["balanced"]] <- c(5, 1000, 5, 1000, 5, 5, 1000, 5, 1000, 5)
```

```
w_plans[["one_half"]] <- c(rep(1000, 4), rep(5, 6))
```

```
```
```



|          |             |                             |
|----------|-------------|-----------------------------|
| w_plans  | list [2]    | List of length 2            |
| balanced | double [10] | 5 1000 5 1000 5 5 ...       |
| one_half | double [10] | 1000 1000 1000 1000 5 5 ... |

**Weight of 5: no investment in population; weight of 1000: investment in population**

```
```{r}
```

```
# create list of weights
```

```
w <- list()
```

```
for(i in 1:2) { # loop over plans
```

```
w[[i]] <- list()
```


```
for(j in 1:80) { # loop over iterations
```

```
w[[i]][[j]] <- matrix(w_plans[[i]], nrow = 1)
```

```
}
```

```
}
```

```
```
```



|       |           |                   |
|-------|-----------|-------------------|
| w     | list [2]  | List of length 2  |
| [[1]] | list [80] | List of length 80 |
| [[2]] | list [80] | List of length 80 |

## Example Code Continued...

Using our weights, we can run simulations to explore different prioritization strategies for conservation

```
# run simulation with conservation plans
```{r}
set.seed(1)

# run the conservation plan with our weights
arma_sp <- run_cons_plans(w,
                          env_type = "arma", # same as base scenario
                          env_params = arma_env_params) # same as base scenario

plot_cons_plans(arma_sp$plans_mv,
                 plans_name = c("Balanced", "One half"),
                 cols = c("#EE7C11", "#377EB8"), xlab = "Variance of growth rate",
                 ylab = "Mean growth rate")
```
```

# Results



## prioritizing populations for investment

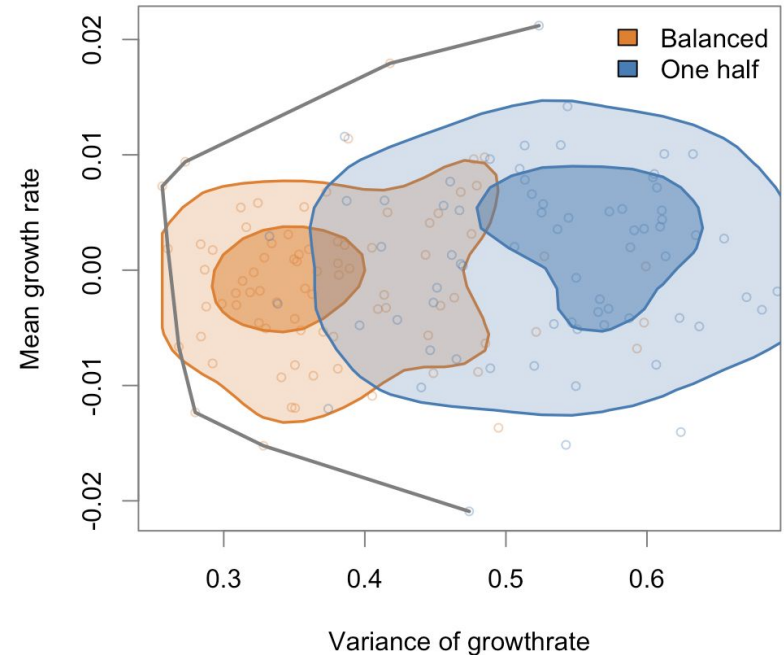
Figure shows **two spatial conservation strategies** in a risk–return space with stationary environmental stochasticity.

1. *All populations (10) are conserved (orange)*
2. *Half of the populations are conserved (blue)*

**Grey line** is the **efficient frontier** that represents the **minimum expected mean growth rate** for a given expected **variance in populations**.

An efficient portfolio will maximize returns, while minimizing variance.

**Dots** show **meta-populations** across all simulated meta populations (80 simulations).



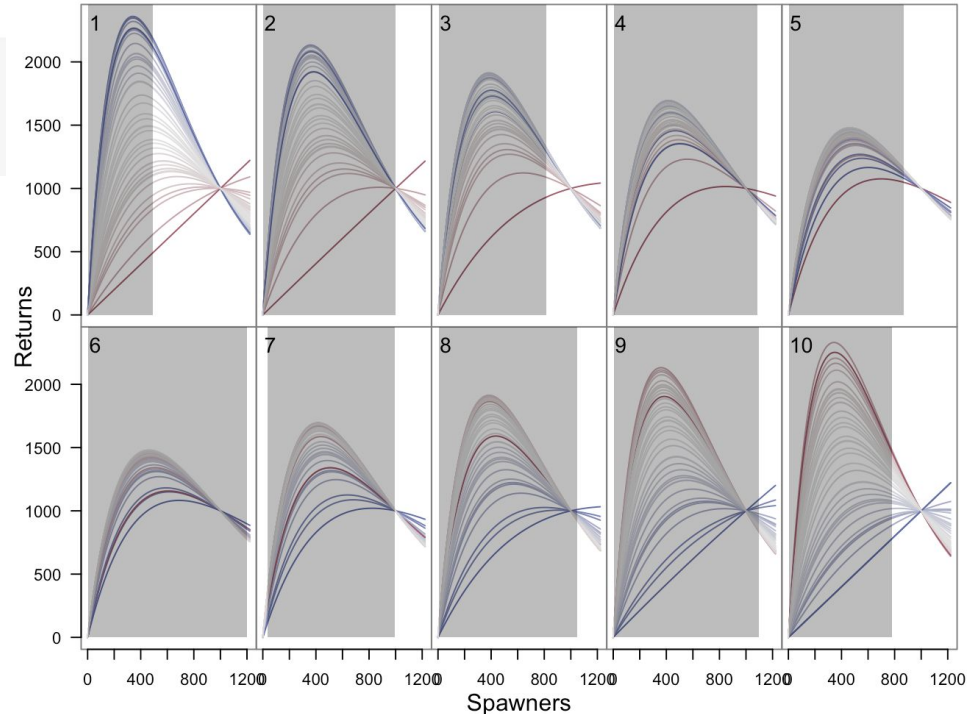
# Results



```
## Plotting additional functions
```

```
```{r}  
# visualize variability in Ricker alpha parameter  
plot_rickers(base1, pal = rep("black", 10))  
```
```

Each panel represents a different stream population. Population 1 is more productive in cool conditions and population 10 is more productive in warm conditions.



# Results



We can select the optimized portfolios by determining the efficient frontier

```
## Optimizing metapopulation portfolios
```

```
```{r}
```

```
set.seed(1)
```

```
weights_matrix <- create_asset_weights(n_pop = 6, n_sims = 3000,  
  weight_lower_limit = 0.001)
```

```
mc_ports <- monte_carlo_portfolios(weights_matrix = weights_matrix,  
  n_sims = 3000, mean_b = 1000)
```

```
```
```

```
```{r}|
```

```
ef_dat <- plot_efficient_portfolios(port_vals = mc_ports$port_vals)
```

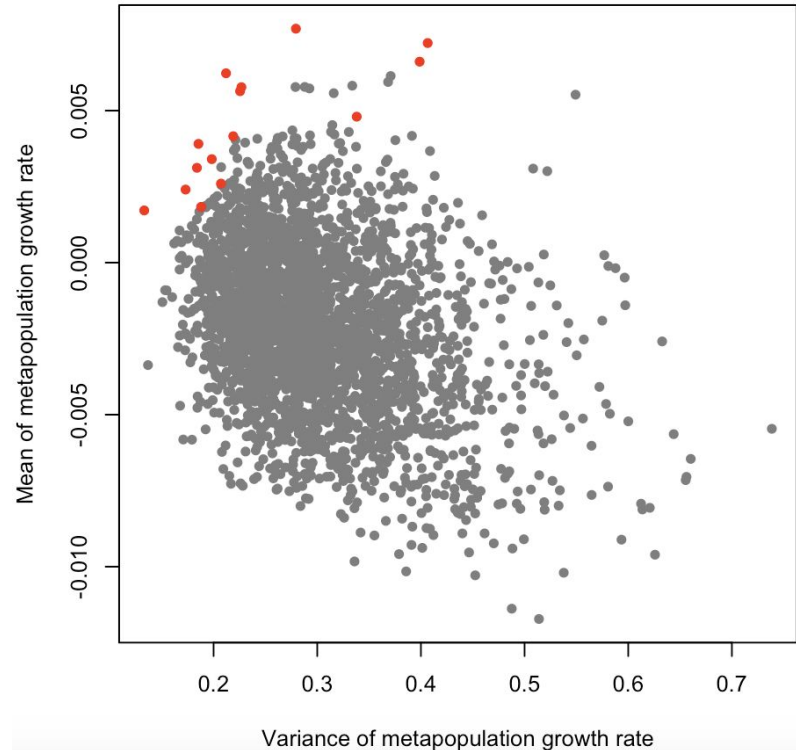
```
```
```

# Results



## visualizing thermal tolerance conditions and efficient frontier showing meta populations

**Red dots** are efficient portfolios that **maximize return of salmon for a given level of variance.**





# Advantages and Limitations of Metafolio

## Advantages:

- Metapopulation analysis
- Allows for the direct comparison of returns and variance from different conservation scenarios
- The results can aid in which type of conservation scenario is optimal
  - For example, should we conserve all thermal tolerances or just one type of thermal tolerance?
- In this example, the results can be used to help build metapopulation resilience in the face of climate change

## Limitations:

- Largely limited to salmon simulations
- Analysis requires a metapopulation
- Requires different forms of environmental data for a given conservation scenario
- Uses a single life-stage population model (Ricker model)

# Unique Insight



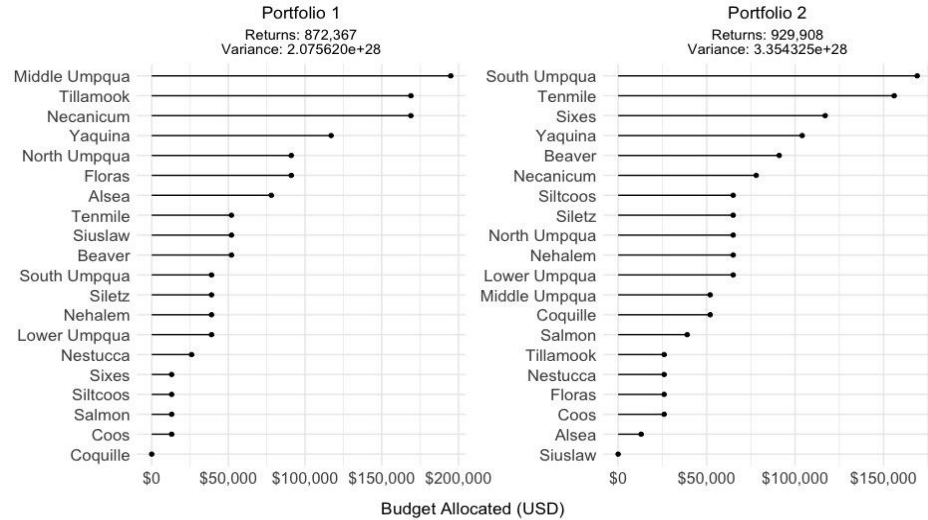
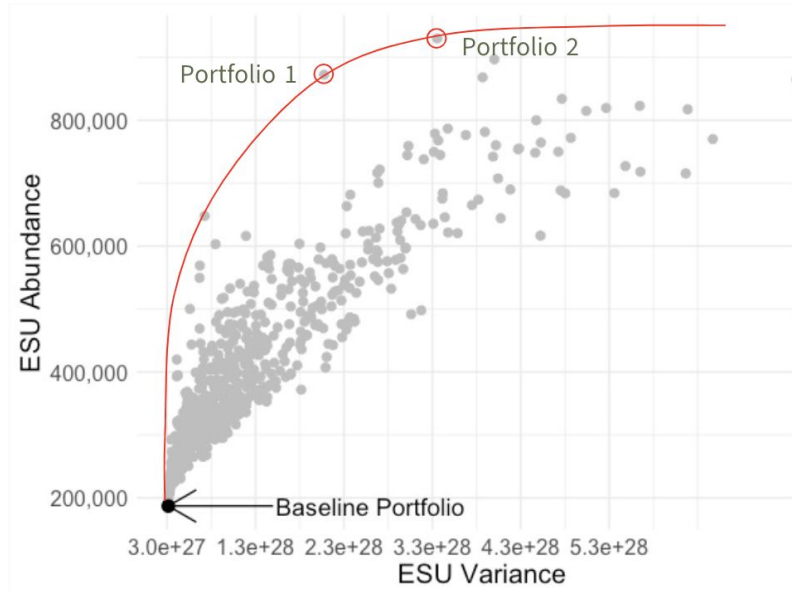
Portfolio theory can be applied in many different ways: salmon conservation, fishery management...etc.

A new application of MPT that our group developed as part of our Group Project with the Wild Salmon Center is: ***Applying an endogenous application of portfolio theory for spatial recovery planning for Oregon Coast coho salmon.***

Our research question is: ***How can we differentially allocate a fixed budget across a metapopulation to maximize salmon returns for a given level of variance?***

- ***Portfolio:*** Oregon Coast coho salmon
- ***Assets:*** 21 Individual populations
- ***Weights:*** Conservation Dollars

# Unique Insight





# Resources

## Vignette

An introduction to the `metafolio` R package

Sean C. Anderson<sup>1\*</sup>, Jonathan W. Moore<sup>1,2</sup>, Michelle M. McClure<sup>3</sup>,  
Nicholas K. Dulvy<sup>1</sup> Andrew B. Cooper<sup>2</sup>

## Metafolio Manual

### Package ‘metafolio’

October 20, 2023

**Type** Package

**Title** Metapopulation Simulations for Conserving Salmon Through  
Portfolio Optimization

**Version** 0.1.2

**Description** A tool to simulate salmon metapopulations and apply financial  
portfolio optimization concepts. The package accompanies the paper  
Anderson et al. (2015) <doi:10.1101/2022.03.24.485545>.

**License** GPL-2

**URL** <https://github.com/seananderson/metafolio>

**BugReports** <https://github.com/seananderson/metafolio/issues>

**VignetteBuilder** knitr

**Depends** R (>= 3.5.0)

**LinkingTo** Rcpp, RcppArmadillo

**Imports** Rcpp (>= 0.11.2), plyr, colorspace, MASS

**Suggests** knitr, TeachingDemos, RColorBrewer, reshape2

**RoxygenNote** 7.2.3

**Encoding** UTF-8

## Literature

Portfolio conservation of metapopulations under  
climate change

Sean C. Anderson<sup>1\*</sup> Jonathan W. Moore<sup>1,2</sup> Michelle M. McClure<sup>3</sup>  
Nicholas K. Dulvy<sup>1</sup> Andrew B. Cooper<sup>2</sup>

## MPT Related Resources

- *Identifying the Potential of anadromous salmonid habitat restoration with life cycle models:*
- *Optimizing provisions of ecosystem services suring modern portfolio theory*
- *Optimal Portfolio Design to reduce climate-related conservation uncertainty in the Prairie Pothole Region.*



# Portfolio Assignment



1. We evaluated 10 individual populations as our baseline meta population. **Change the number of populations in the baseline scenario within the `meta_sim()` function and plot the results.**
  - *How does this affect the base simulation plot (`plot_sim_ts`)?*
2. **Change the weight allocation for the ‘balanced’ conservation strategy so that all 10 populations receive a maximum investment (i.e. 1000) and plot the results.**
  - *How does the variance and mean of the growth rate shift?*
  - *Does it shift the efficiency frontier?*
  - *Of the two conservation plans, which portfolio is optimal? Why?*
3. **What are some other examples of environmental conditions for salmonids that would be beneficial to model using this package?**

**Thanks! Questions?**



