

# POMP Model Analysis

Sijia Wang

2025-06-11

```
library(conflicted)
conflict_prefer("map", "pomp")
conflict_prefer("filter", "dplyr")

library(pomp)
library(ggplot2)
library(tidyverse)
library(dplyr)
library(doFuture)
library(doRNG)
library(foreach)
knitr::opts_chunk$set(echo = TRUE)
```

## POMP Model Fitting

We use the **orange panel data between days 41 and 107** to build and analyze a Partially Observed Markov Process (POMP) model for the three-species food web system.

```
# Load the data
X123 <- read.csv("NoEvoData=SE=Feb14 2012.csv")

# Prepare orange panel data (log-transformed and mean-centered)
orange_data <- data.frame(
  day = X123$day,
  algae = log(X123$Algae.orange1),
  flagellates = log(X123$Flag.orange1),
  rotifers = log(pmax(X123$Rot.orange1, 1e-3))
)

# Subset data between day 41 and 107
orange_data <- orange_data[orange_data$day >= 41 & orange_data$day <= 107, ]

# Simple mean-centering
orange_data <- orange_data %>%
  mutate(
    algae = algae,
    flagellates = flagellates,
    rotifers = rotifers
  )
```

```

rmeasure <- Csnippet("
  algae = rnorm(log(A), sigma_A);
  flagellates = rnorm(log(F), sigma_F);
  rotifers = rnorm(log(R), sigma_R);
")

dmeasure <- Csnippet("
  // Convert states to mean-centered log scale
  lik = dnorm(algae, log(A), sigma_A, 1) +
        dnorm(flagellates, log(F), sigma_F, 1) +
        dnorm(rotifers, log(R), sigma_R, 1);

  if (!give_log) lik = exp(lik);
")

# Updated parameter names to include the means
paramnames <- c(
  "delta", "kA", "kR", "r", "g", "h",
  "alphaA", "alphaF", "eta", "IF",
  "S_0", "A_0", "R_0", "F_0",
  "sigma_A", "sigma_F", "sigma_R",
  "sigma_proS", "sigma_proA",
  "sigma_proF", "sigma_proR"
)

# Process model (same as before)
rproc <- euler(
  step.fun = Csnippet("
    double dS = delta * (1 - S) - S * r * A / (kA + S);
    double dA = A * (
      r * S / (kA + S)
      - g * R / (kR + A + alphaF * F)
      - h * F / (1 + alphaA * A)
      - delta
    );
    double dR = R * (
      g * A / (kR + A + alphaF * F)
      + eta * F / (kR + A + alphaF * F)
      - delta
    );
    double dF = F * (
      h * A / (1 + alphaA * A)
      - eta * R / (kR + A + alphaF * F)
      - delta
    ) + IF;

    double eS = rnorm(0, sigma_proS * sqrt(dt));
    double eA = rnorm(0, sigma_proA * sqrt(dt));
    double eR = rnorm(0, sigma_proR * sqrt(dt));
    double eF = rnorm(0, sigma_proF * sqrt(dt));

    S += dS * dt + eS;
    A += dA * dt + eA;
  ")
)

```

```

    R += dR * dt + eR;
    F += dF * dt + eF;

    if (S < 80) S = 80;
    if (A < 0) A = 1e-3;
    if (R < 0) R = 1e-3;
    if (F < 0) F = 1e-3;

    if (S > 1e6) S = 1e6;
    if (A > 1e8) A = 1e8;
    if (R > 1e6) R = 1e6;
    if (F > 1e6) F = 1e6;
  ),
  delta.t = 0.25
)

rinit <- Csnippet("
  S = S_0;
  A = A_0;
  R = R_0;
  F = F_0;
")

# Parameter transformation
pt <- parameter_trans(
  log = c("delta", "kA", "kR", "r", "g", "h",
    "alphaA", "alphaF", "eta", "IF",
    "S_0", "A_0", "R_0", "F_0",
    "sigma_A", "sigma_F", "sigma_R",
    "sigma_proS", "sigma_proA",
    "sigma_proF", "sigma_proR")
)

# Create POMP model
pomp_model <- pomp(
  data = orange_data,
  times = "day",
  t0 = 40,
  rprocess = rproc,
  rmeasure = rmeasure,
  dmeasure = dmeasure,
  rinit = rinit,
  statenames = c("S", "A", "R", "F"),
  paramnames = paramnames,
  obsnames = c("algae", "flagellates", "rotifers"),
  partrans = pt
)

# Get initial conditions
init_row <- X123[X123$day == 40, ]

# Parameter vector including the means
params <- c(

```

```

delta = 0.25,
kA = 0.2, # smaller kA, algae can grow faster even at low nutrient levels
r = 0.25, # algae growth rate
kR = 1.6, # lower kR, predators (F and R) can respond strongly even when prey is scarce
g = 0.2, # higher g, rotifers grow faster from consuming algae.
h = 0.5, # higher h, algae are grazed faster by flagellates, and flagellates grow faster
alphaA = 2.33, # Higher A, flagellates saturate faster
alphaF = 0.4, # rotifer's feeding rate on flagellates
eta = 3.6, # rotifer feeding on flagellates
IF = 0.3, # rotifer's feeding saturation constant
S_0 = 80,
A_0 = init_row$Algae.orange1,
R_0 = init_row$Rot.orange1,
F_0 = init_row$Flag.orange1,
sigma_A = 0.1,
sigma_F = 0.1,
sigma_R = 0.1,
sigma_proS = 0.1,
sigma_proA = 2,
sigma_proF = 0.5,
sigma_proR = 0.1
)

```

```

# Simulate and plot
sim <- simulate(pomp_model, params = params, nsim = 1)
sim_df <- as.data.frame(sim)
summary(sim_df)

```

```

##      day      algae      flagellates      rotifers
## Min.   : 41.0   Min.   :8.851   Min.   :2.150   Min.   : -1.66936
## 1st Qu.: 57.5   1st Qu.:9.033   1st Qu.:2.607   1st Qu.: -0.54541
## Median : 74.0   Median :9.088   Median :2.940   Median : -0.33298
## Mean   : 74.0   Mean   :9.096   Mean   :2.918   Mean   : -0.29119
## 3rd Qu.: 90.5   3rd Qu.:9.158   3rd Qu.:3.184   3rd Qu.: 0.07498
## Max.   :107.0   Max.   :9.323   Max.   :3.666   Max.   : 0.65749
##      S      A      R      F
## Min.   :80   Min.   :8757   Min.   :0.1858   Min.   : 9.333
## 1st Qu.:80   1st Qu.:8887   1st Qu.:0.6067   1st Qu.:13.710
## Median :80   Median :9055   Median :0.7117   Median :17.440
## Mean   :80   Mean   :9065   Mean   :0.8236   Mean   :19.776
## 3rd Qu.:80   3rd Qu.:9227   3rd Qu.:1.0581   3rd Qu.:24.718
## Max.   :80   Max.   :9433   Max.   :1.7782   Max.   :37.362

```

```

# Prepare observed data for plotting
obs_long <- orange_data %>%
  pivot_longer(cols = c(algae, flagellates, rotifers),
               names_to = "species",
               values_to = "log_abundance") %>%
  mutate(type = "Observed")

# Prepare simulation data - convert to log scale to match observations
sim_long <- sim_df %>%
  mutate(

```

```

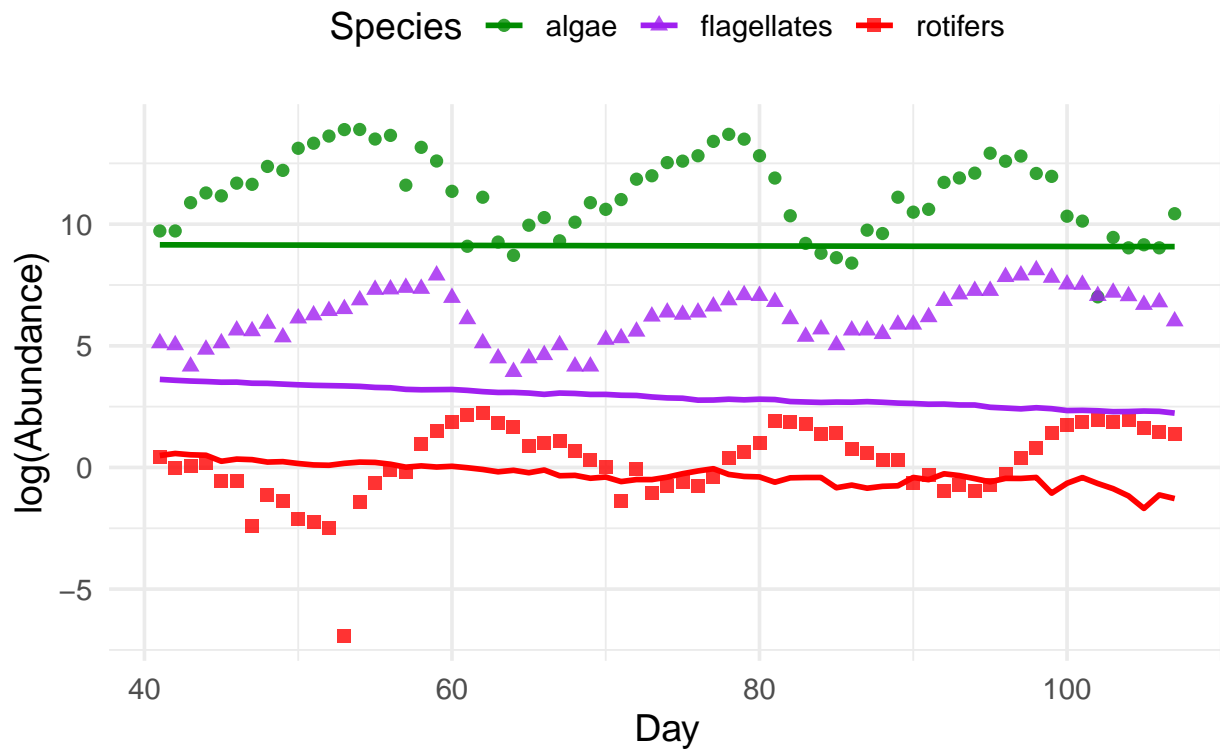
    algae = log(pmax(A, 1e-6)),
    flagellates = log(pmax(F, 1e-6)),
    rotifers = log(pmax(R, 1e-6))
  ) %>%
  select(day, algae, flagellates, rotifers) %>%
  pivot_longer(cols = c(algae, flagellates, rotifers),
               names_to = "species",
               values_to = "log_abundance") %>%
  mutate(type = "Simulated")

# Combine and plot
combined_data <- bind_rows(obs_long, sim_long)

ggplot(combined_data, aes(x = day, y = log_abundance, color = species)) +
  geom_point(data = filter(combined_data, type == "Observed"),
            aes(shape = species), size = 2, alpha = 0.8) +
  geom_line(data = filter(combined_data, type == "Simulated"),
           linewidth = 1) +
  scale_color_manual(values = c(
    algae = "green4",
    flagellates = "purple",
    rotifers = "red"
  )) +
  scale_shape_manual(values = c(
    algae = 16, flagellates = 17, rotifers = 15
  )) +
  labs(
    title = "POMP Simulation vs. Observed Data",
    x = "Day",
    y = "log(Abundance)",
    color = "Species",
    shape = "Species"
  ) +
  theme_minimal(base_size = 14) +
  theme(
    legend.position = "top",
    plot.title = element_text(hjust = 0.5, face = "bold")
  )

```

## POMP Simulation vs. Observed Data



```
library(future)
library(doParallel)
```

```
## Loading required package: iterators
```

```
## Loading required package: parallel
```

```
library(foreach)
plan(multisession)
```

```
# Step 1: initial guess
init_params <- params
```

```
# Step 2: define random walk standard deviations for each parameter
rw_sd_values <- rw_sd(
  kA      = 0.02,
  kR      = 0.02,
  r        = 0.02,
  g        = 0.02,
  h        = 0.02,
  alphaA   = 0.2,
  alphaF   = 0.2,
  eta      = 0.02,
  IF       = 0.02,
  sigma_A  = 0.02,
```

```

sigma_F = 0.02,
sigma_R = 0.02,
sigma_proS = 0.2,
sigma_proA = 0.5,
sigma_proF = 0.2,
sigma_proR = 0.02,
A_0 = ivp(0.02),
F_0 = ivp(0.02),
R_0 = ivp(0.02),
S_0 = ivp(0.02)
)

cores <- as.numeric(Sys.getenv('SLURM_NTASKS_PER_NODE', unset=NA))
if(is.na(cores)) cores <- detectCores()
registerDoParallel(cores)
ggplot2::theme_set(ggplot2::theme_bw())

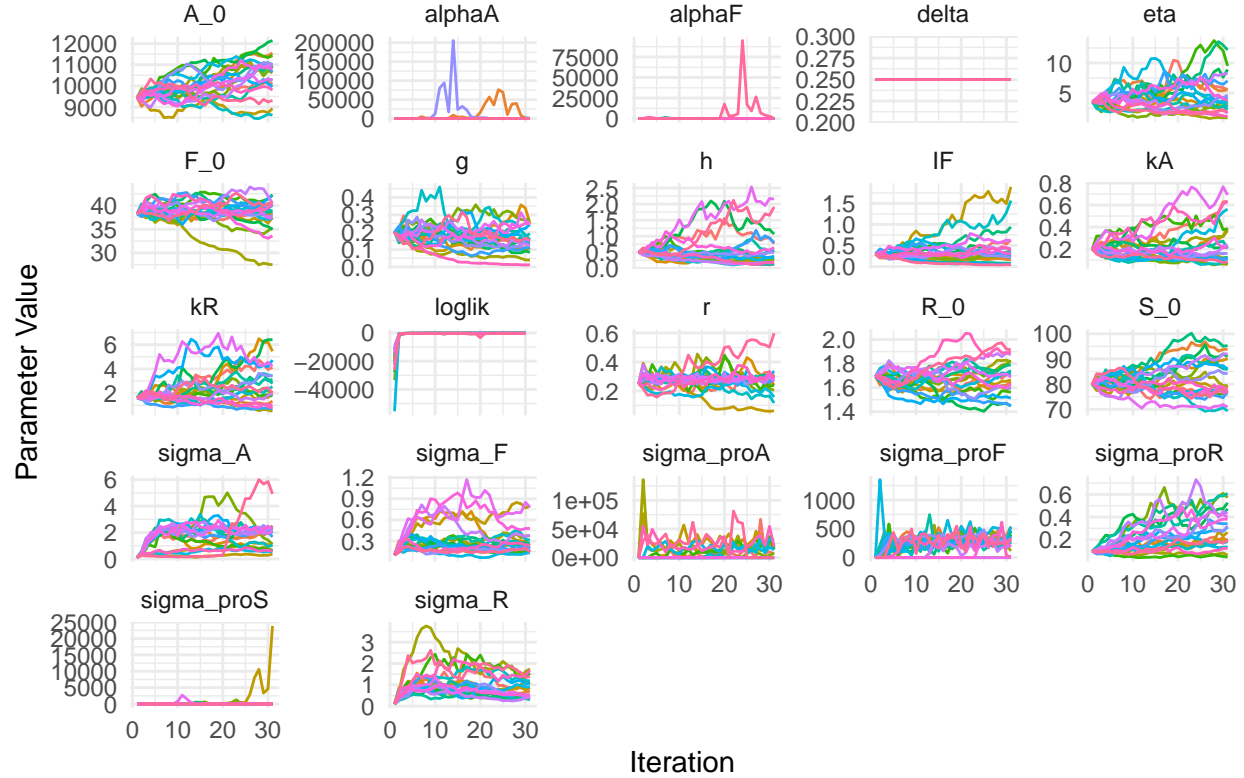
stopifnot(packageVersion("pomp")>="5.0")

# Perform multiple mif2 fits in parallel to obtain multiple parameter estimates
mifs_local <- foreach(
  i = 1:20,
  .combine = c,
  .options.future = list(seed = 89898975),
  .packages = "pomp"
) %dopar% {
  mif2(
    pomp_model,
    params = init_params,
    Np = 1000,          # Number of particles
    Nmif = 30,          # Number of mif iterations
    cooling.fraction.50 = 0.5, # Cooling fraction
    rw.sd = rw_sd_values # Random walk standard deviations
  )
}

#parameter traces
mifs_local |>
  traces() |>
  melt() |>
  ggplot(aes(x = iteration, y = value, group = .L1, color = factor(.L1))) +
  geom_line() +
  guides(color = "none") +
  facet_wrap(~ name, scales = "free_y") +
  theme_minimal() +
  labs(
    title = "Parameter Traces from Multiple mif2 Runs",
    x = "Iteration",
    y = "Parameter Value"
  )

```

## Parameter Traces from Multiple mif2 Runs



```
local_search <- foreach(
  mf = mifs_local,
  .combine = rbind
) %dopar% {
  evals <- replicate(10, logLik(pfilter(mf, Np = 5000)))
  ll <- logmeanexp(evals, se = TRUE)
  mf %>% coef() %>% bind_rows() %>%
    bind_cols(loglik = ll[1], loglik.se = ll[2])
}

bind_rows(local_search) %>%
  filter(is.finite(loglik)) %>%
  filter(loglik.se < .5) %>%
  arrange(-loglik) -> best_searches

head(best_searches)
```

```
## # A tibble: 6 x 23
##   delta    kA    r    kR    g    h alphaA  alphaF  eta    IF    S_0    A_0
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 0.250 0.210 0.287 6.41  0.0837 1.07   5.26 0.00515 8.92 0.340 90.9 12144.
## 2 0.25  0.127 0.296 3.13  0.155  0.513  3.56 0.171  12.2 0.945 95.3 10820.
## 3 0.250 0.208 0.282 2.79  0.191  0.566  2.87 0.0849  2.21 0.0783 85.9 10260.
## 4 0.250 0.240 0.282 1.96  0.124  0.197  1.25 0.274  8.26 0.427 92.0 11048.
## 5 0.250 0.111 0.278 0.709 0.110  0.798  4.75 2.50   7.03 0.349 77.2 10651.
```



```
## 6 0.250 0.634 0.300 4.12 0.113 0.765 3.80 0.000519 5.60 0.629 79.5 10083.
## # i 11 more variables: R_0 <dbl>, F_0 <dbl>, sigma_A <dbl>, sigma_F <dbl>,
## # sigma_R <dbl>, sigma_proS <dbl>, sigma_proA <dbl>, sigma_proF <dbl>,
## # sigma_proR <dbl>, loglik <dbl>, loglik.se <dbl>
```

```
summary(best_searches$loglik)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -983.6  -345.8  -293.0  -398.1  -275.6  -273.4
```

```
params_best <- best_searches[1, ] %>% unlist() %>% as.numeric()
names(params_best) <- names(best_searches)
simulated_data <- simulate(
  pomp_model,
  params = params_best,
  nsim = 5,
  format = "data.frame",
  include.data = TRUE
)

sim_df <- as.data.frame(simulated_data, include.data = FALSE)
summary(sim_df)
```

```
##      day      .id      algae      flagellates      rotifers
## Min.   : 41    data:67    Min.   : 4.128    Min.   : -7.358    Min.   : -6.908
## 1st Qu.: 57     1 :67     1st Qu.: 8.854    1st Qu.: 5.853    1st Qu.: 1.411
## Median : 74     2 :67     Median :10.534    Median : 6.500    Median : 3.646
## Mean   : 74     3 :67     Mean   :10.348    Mean   : 5.650    Mean   : 3.503
## 3rd Qu.: 91     4 :67     3rd Qu.:11.859    3rd Qu.: 7.083    3rd Qu.: 5.920
## Max.   :107     5 :67     Max.   :16.717    Max.   : 8.470    Max.   : 8.045
##
##      S      A      R      F
## Min.   :80    Min.   :12556    Min.   : 0.2309    Min.   : 0.001
## 1st Qu.:80    1st Qu.:20043    1st Qu.: 12.7748    1st Qu.: 369.392
## Median :80    Median :31643    Median : 88.6258    Median : 694.015
## Mean   :80    Mean   :38029    Mean   : 321.8617    Mean   : 884.632
## 3rd Qu.:80    3rd Qu.:52272    3rd Qu.: 469.4543    3rd Qu.:1196.150
## Max.   :80    Max.   :94365    Max.   :2652.8085    Max.   :3642.942
## NA's   :67    NA's   :67      NA's   :67      NA's   :67
```

```
# Prepare observed data for plotting
obs_long <- orange_data %>%
  pivot_longer(cols = c(algae, flagellates, rotifers),
    names_to = "species",
    values_to = "log_abundance") %>%
  mutate(type = "Observed")

# Prepare simulation data - convert to log scale to match observations
sim_long <- sim_df %>%
  filter(.id != "data") %>%
  mutate(
    algae = log(pmax(A, 1e-6)),
```

```

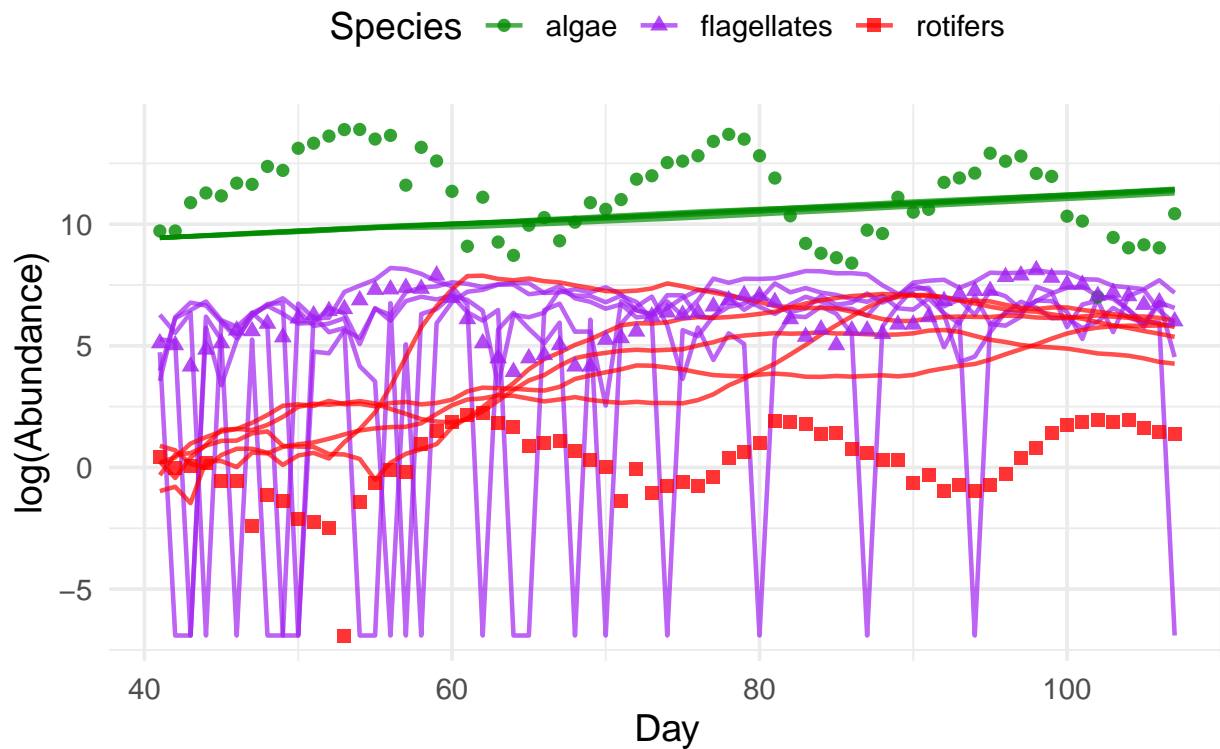
    flagellates = log(pmax(F, 1e-6)),
    rotifers = log(pmax(R, 1e-6))
  ) %>%
  select(.id, day, algae, flagellates, rotifers) %>%
  pivot_longer(cols = c(algae, flagellates, rotifers),
               names_to = "species",
               values_to = "log_abundance") %>%
  mutate(type = "Simulated")

# Combine and plot
combined_data <- bind_rows(obs_long, sim_long)

ggplot() +
  geom_line(data = filter(sim_long, is.finite(log_abundance)),
            aes(x = day, y = log_abundance, color = species, group = interaction(.id, species)),
            linewidth = 0.8, alpha = 0.7) +
  geom_point(data = obs_long,
             aes(x = day, y = log_abundance, color = species, shape = species),
             size = 2, alpha = 0.8) +
  scale_color_manual(values = c(
    algae = "green4",
    flagellates = "purple",
    rotifers = "red"
  )) +
  scale_shape_manual(values = c(
    algae = 16, flagellates = 17, rotifers = 15
  )) +
  labs(
    title = "POMP Simulation (5 Paths) vs. Observed Data",
    x = "Day",
    y = "log(Abundance)",
    color = "Species",
    shape = "Species"
  ) +
  theme_minimal(base_size = 14) +
  theme(
    legend.position = "top",
    plot.title = element_text(hjust = 0.5, face = "bold")
  )

```

## POMP Simulation (5 Paths) vs. Observed Data



```
library(doFuture)
library(iterators)
plan(multisession)
library(purrr)

# Step 1: define narrowed random parameter ranges
best_param <- coef(mifs_local[[1]])
param_range <- runif_design(
  lower = best_param * 0.5,
  upper = best_param * 1.5,
  nseq = 100
)

# Step 2: pick a starting mif2 object from local search
mf_start <- mifs_local[[1]]

# Step 3: Run mif2 for each random parameter set
bake(file = "mif2_global.rds", seed = 888888, {
  foreach(p = iter(param_range, "row"), .combine = c,
    .options.future = list(seed = TRUE)) %dofuture% {
    mif2(
      mf_start,
      Nmif = 200,
      params = unlist(p),
      Np = 1000,
      cooling.fraction.50 = 0.5,

```

```

    rw.sd = rw_sd_values
  )
}
}) -> mif2_results

# Step 4: Evaluate log-likelihoods
bake(file = "pfilter_ll.rds", seed = 999999, {
  foreach(m = mif2_results, .combine = rbind,
    .options.future = list(seed = TRUE)) %dofuture% {
    replicate(20, logLik(pfilter(m, Np = 5000))) |>
      logmeanexp(se = TRUE)
  }
}) -> loglik_results

# Step 5: Combine and sort
mif2_params <- map_dfr(mif2_results, coef)

mif2_global_summary <- mif2_params %>%
  bind_cols(
    loglik = loglik_results[, 1],
    loglik.se = loglik_results[, 2]
  ) %>%
  arrange(desc(loglik))

# Step 6: Filter good results
best_searches <- mif2_global_summary %>%
  filter(is.finite(loglik), loglik.se < 0.5) %>%
  arrange(-loglik)

head(best_searches)

```

```

## # A tibble: 6 x 23
##   delta    kA      r    kR      g      h alphaA  alphaF    eta    IF    S_0    A_0
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 0.262 1.10 0.246 1.16 0.218 2.90 12.6 2.97e+1 1.06 0.550 80.0 5144.
## 2 0.259 0.867 0.172 0.226 0.143 0.269 1.29 8.58e-1 0.380 1.88 97.1 6512.
## 3 0.263 0.890 0.288 0.155 0.121 0.502 2.60 8.99e-3 7.72 1.26 87.2 14250.
## 4 0.134 0.166 0.113 2.28 0.0331 0.761 44.8 9.01e-4 7.82 1.04 127. 11409.
## 5 0.205 0.264 0.159 0.322 0.175 0.118 0.653 1.22e+0 0.296 0.812 78.5 6617.
## 6 0.356 0.558 0.344 0.0817 0.188 2.85 9.74 1.25e-2 10.5 0.925 63.5 13664.
## # i 11 more variables: R_0 <dbl>, F_0 <dbl>, sigma_A <dbl>, sigma_F <dbl>,
## #   sigma_R <dbl>, sigma_proS <dbl>, sigma_proA <dbl>, sigma_proF <dbl>,
## #   sigma_proR <dbl>, loglik <dbl>, loglik.se <dbl>

```

```
summary(best_searches$loglik)
```

```

##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -359.7 -290.1 -282.1 -282.6 -270.2 -254.6

```

```

# Simulate best
params <- best_searches[1, paramnames] |> unlist()

```

```

simulated_data <- simulate(
  pomp_model,
  params = params,
  nsim = 1,
  format = "data.frame",
  include.data = TRUE
)

summary(simulated_data)

```

```

##      day      .id      algae      flagellates
## Min.   : 41.00  data:67  Min.   :-9.143  Min.   :-7.591
## 1st Qu.: 57.25   1 :67  1st Qu.:10.726  1st Qu.: 4.637
## Median : 74.00           Median :11.930  Median : 5.532
## Mean   : 74.00           Mean   :11.356  Mean   : 4.700
## 3rd Qu.: 90.75           3rd Qu.:12.688  3rd Qu.: 6.309
## Max.   :107.00          Max.   :14.687  Max.   : 8.114
##
##      rotifers      S      A      R
## Min.   :-7.03468  Min.   :    80  Min.   :    0  Min.   :0.0010
## 1st Qu.: -0.58056  1st Qu.:    80  1st Qu.:143326  1st Qu.:0.7922
## Median : 0.42530  Median :1000000  Median :234618  Median :1.6570
## Mean   :-0.05763  Mean   : 552275  Mean   :239837  Mean   :2.1220
## 3rd Qu.: 1.28994  3rd Qu.:1000000  3rd Qu.:348831  3rd Qu.:3.2400
## Max.    : 2.23569  Max.   :1000000  Max.   :499083  Max.   :6.6582
##              NA's   :67      NA's   :67      NA's   :67
##
##      F
## Min.   : 0.001
## 1st Qu.: 51.821
## Median :164.964
## Mean   :179.914
## 3rd Qu.:256.592
## Max.   :667.235
## NA's   :67

```

```

sim_df <- as.data.frame(simulated_data, include.data = FALSE)

# Prepare observed data for plotting
obs_long <- orange_data %>%
  pivot_longer(cols = c(algae, flagellates, rotifers),
    names_to = "species",
    values_to = "log_abundance") %>%
  mutate(type = "Observed")

# Prepare simulation data - convert to log scale to match observations
sim_long <- sim_df %>%
  filter(.id != "data") %>%
  mutate(
    algae = log(pmax(A, 1e-6)),
    flagellates = log(pmax(F, 1e-6)),
    rotifers = log(pmax(R, 1e-6))
  ) %>%
  select(.id, day, algae, flagellates, rotifers) %>%

```

```

pivot_longer(cols = c(algae, flagellates, rotifers),
              names_to = "species",
              values_to = "log_abundance") %>%
mutate(type = "Simulated")

# Combine and plot
combined_data <- bind_rows(obs_long, sim_long)

ggplot() +
  geom_line(data = filter(sim_long, is.finite(log_abundance)),
            aes(x = day, y = log_abundance, color = species, group = interaction(.id, species)),
            linewidth = 0.8, alpha = 0.7) +
  geom_point(data = obs_long,
             aes(x = day, y = log_abundance, color = species, shape = species),
             size = 2, alpha = 0.8) +
  scale_color_manual(values = c(
    algae = "green4",
    flagellates = "purple",
    rotifers = "red"
  )) +
  scale_shape_manual(values = c(
    algae = 16, flagellates = 17, rotifers = 15
  )) +
  labs(
    title = "POMP Simulation (5 Paths) vs. Observed Data",
    x = "Day",
    y = "log(Abundance)",
    color = "Species",
    shape = "Species"
  ) +
  theme_minimal(base_size = 14) +
  theme(
    legend.position = "top",
    plot.title = element_text(hjust = 0.5, face = "bold")
  )

```

## POMP Simulation (5 Paths) vs. Observed Data

