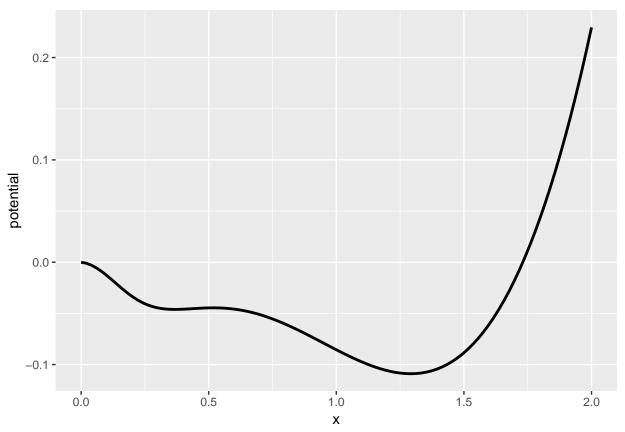
simulation & inference of transients

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```
library(tidyverse)
N <- 1e4
r <- 0.05; K <- 2
a <- 0.023; H <- 0.3; Q <- 3
x0 <- 0.2; sigma <- 0.02
growth <- function(x, r, K) x * r * (1 - x / K)
consumption <- function(x, a, H, Q) a * x^Q / (x^Q + H^Q)
theory <-
  tibble(x = seq(0,K, length.out = 100)) %>%
  mutate(g = growth(x, r, K),
         c = consumption(x, a, H, Q)) %>%
  mutate(potential = - cumsum(g - c)) %>%
  gather(curve, y, -x, -potential)
theory %>%
  ggplot(aes(x, y, col = curve)) +
  geom_line(lwd = 1)
  0.025 -
  0.020 -
  0.015 -
                                                                                     curve
  0.010 -
  0.005 -
  0.000 -
                          0.5
         0.0
                                           1.0
                                                            1.5
                                                                             2.0
```

```
theory %>%
ggplot(aes(x, potential)) +
geom_line(lwd = 1)
```



library(nimble)

```
## nimble version 0.8.0 is loaded.
## For more information on NIMBLE and a User Manual,
## please visit http://R-nimble.org.
##
## Attaching package: 'nimble'
## The following object is masked from 'package:stats':
##
##
       simulate
## define truncated normal distribution ----
dtruncnorm <- nimbleFunction(</pre>
 run = function(x = double(0), mean = double(0),
                 sd = double(0), log = integer(0, default = 0)) {
    returnType(double(0))
    log_prob <- dnorm(x = x, mean = mean, sd = sd, log = 1) -</pre>
      pnorm(q = 0, mean = -mean, sd = sd, log = 1)
    if(log) return(log_prob)
    else return(exp(log_prob))
rtruncnorm <- nimbleFunction(</pre>
 run = function(n = integer(0, default = 1), mean = double(0),
```

```
sd = double(0)) {
    returnType(double(0))
    if(n != 1) print("rtruncnorm only allows n = 1; using n = 1.")
    draw \leftarrow rnorm(n = 1, mean = mean, sd = sd)
    while(draw < 0)</pre>
      draw \leftarrow rnorm(n = 1, mean = mean, sd = sd)
    return(draw)
  })
# Define stochastic model in BUGS notation ----
may <- nimble::nimbleCode({</pre>
  log(r) ~ dnorm(mu_r, sd_r)
  log(K) ~ dnorm(mu_K, sd_K)
  log(a) ~ dnorm(mu_a, sd_a)
 log(H) ~ dnorm(mu_H, sd_H)
  log(Q) ~ dnorm(mu_Q, sd_Q)
  x0 ~ dtruncnorm(mu_x0, sd_x0)
  log(sigma) ~ dnorm(mu_sigma, sd_sigma)
  x[1] <- x0
  for(t in 1:(N - 1)){
    mu[t] \leftarrow x[t] + x[t] * r * (1 - x[t] / K) - a * x[t]^Q / (x[t]^Q + H^Q)
    x[t + 1] ~ dtruncnorm(mean = mu[t], sd = sigma)
})
# constants ----
constants <- list(</pre>
 mu_r = log(r), sd_r = 1,
 mu_K = log(K), sd_K = 1,
 mu_a = log(a), sd_a = 1,
 mu_H = log(H), sd_H = 1,
 mu_Q = log(Q), sd_Q = 1,
 mu_x0 = x0, sd_x0 = 1,
 mu_sigma = log(sigma), sd_sigma = 1
## inits ----
inits \leftarrow list("log_r" = log(r), "log_K" = log(K),
               \log_a = \log(a), \log_H = \log(H), \log_Q = \log(Q),
               "\log_{\text{sigma}}" = \log(\text{sigma}), "x0" = x0)
# define and compile model ----
model <- nimbleModel(code = may, constants = constants, inits = inits)</pre>
## defining model...
## Registering the following user-provided distributions: dtruncnorm .
## NIMBLE has registered dtruncnorm as a distribution based on its use in BUGS code. Note that if you m
## building model...
## setting data and initial values...
## running calculate on model (any error reports that follow may simply reflect missing values in model
## checking model sizes and dimensions... This model is not fully initialized. This is not an error. To
## model building finished.
cmodel <- compileNimble(model)</pre>
```

compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.

```
## compilation finished.
```

```
set.seed(1234)
simulate(cmodel, nodes = c('x', 'mu'))
df <- tibble(t = seq_along(cmodel$x), x = cmodel$x)

df %>% filter(t < 4000) %>% ggplot(aes(t, x)) + geom_line()
```

```
1.5-

1.0-

×

0.5-

0.00-

1000

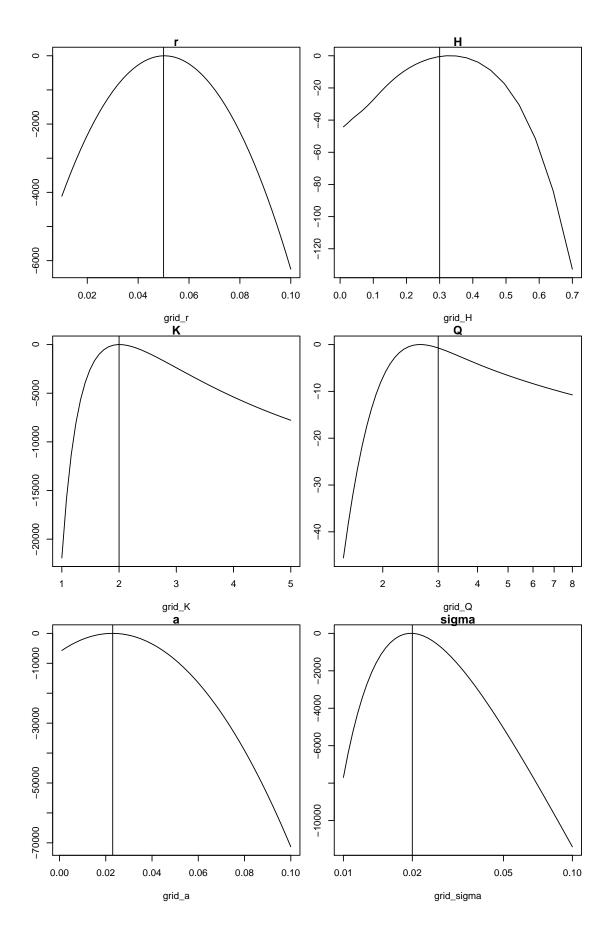
2000

3000

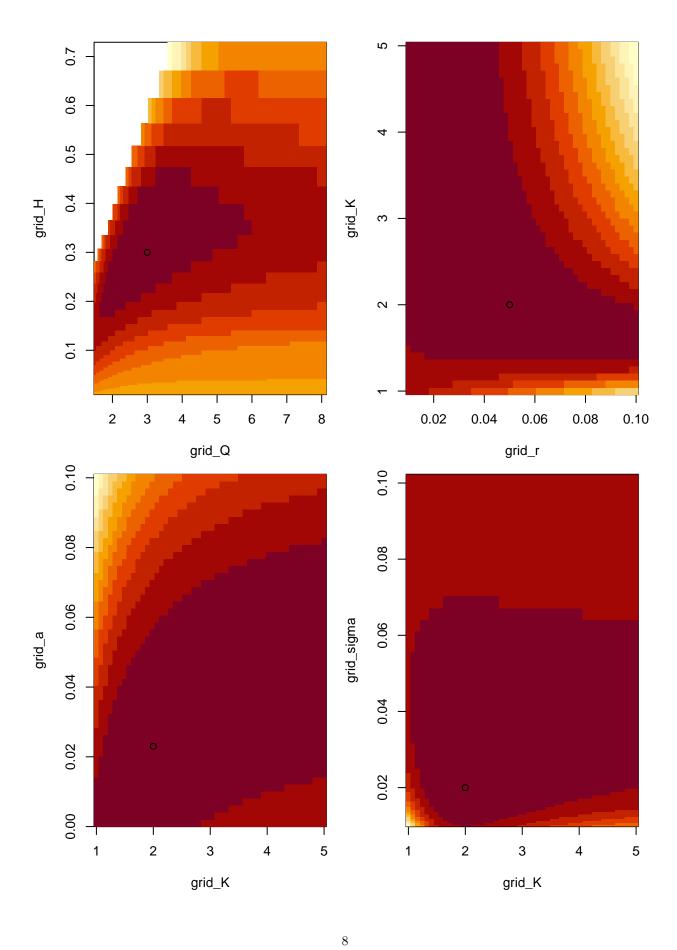
4000
```

```
dnorm_mod <- function(x, mean, sd, log = T){</pre>
  dnorm(x = x, mean = mean, sd = sd, log = log) -
    log(1 - pnorm(q = 0, mean = mean, sd = sd))
}
get_ll <- function(x, r, K, a, H, Q, sigma){</pre>
  TIME <- length(x)
  mu \leftarrow x + x * r * (1 - x / K) - a * x^Q / (x^Q + H^Q)
  sum(dnorm_mod(x = x[-1], mean = mu[-TIME], sd = sigma, log = T))
}
##
grid_r <- seq(1e-2, 1e-1, 1=5e1)
ll_r <- sapply(grid_r, function(r) get_ll(a = cmodel$a, x = cmodel$x, r = r, K = cmodel$K,</pre>
                                            H = cmodel$H, Q = cmodel$Q, sigma = cmodel$sigma))
##
grid_K <- seq(1, 5, l=5e1)
11_K <- sapply(grid_K, function(K) get_ll(a = cmodel$a, x = cmodel$x, r = cmodel$r, K = K,</pre>
                                            H = cmodel$H, Q = cmodel$Q, sigma = cmodel$sigma))
##
grid_a <- seq(1e-3, 1e-1, 1=5e1)
ll_a <- sapply(grid_a, function(a) get_ll(a = a, x = cmodel$x, r = cmodel$r, K = cmodel$K,
```

```
H = cmodel$H, Q = cmodel$Q, sigma = cmodel$sigma))
##
grid_H \leftarrow exp(seq(log(1e-2), log(0.7), l=5e1))
ll_H <- sapply(grid_H, function(H) get_ll(a = cmodel$a, x = cmodel$x, r = cmodel$r, K = cmodel$K,
                                           H = H, Q = cmodel$Q, sigma = cmodel$sigma))
grid_Q \leftarrow exp(seq(log(1.5), log(8), l=5e1))
ll_Q <- sapply(grid_Q, function(Q) get_ll(a = cmodel$a, x = cmodel$x, r = cmodel$r, K = cmodel$K,
                                           H = cmodel$H, Q = Q, sigma = cmodel$sigma))
grid_sigma \leftarrow exp(seq(log(1e-2), log(1e-1), l=5e1))
ll_sigma <- sapply(grid_sigma, function(sigma)</pre>
  get_11(a = cmodel$a, x = cmodel$x, r = cmodel$r, K = cmodel$K,
         H = cmodel$H, Q = cmodel$Q, sigma = sigma))
layout(matrix(1:6, 3, 2)); par(mar = c(4.1, 2, 1, 1))
plot(grid_r, (ll_r - max(ll_r)), type = "l", main = "r")
abline(v = cmodel$r)
plot(grid_K, (11_K - max(11_K)), type = "1", main = "K")
abline(v = cmodel$K)
plot(grid_a, (ll_a - max(ll_a)), type = "l", main = "a")
abline(v = cmodel$a)
plot(grid_H, (ll_H - max(ll_H)), type = "l", main = "H")
abline(v = cmodel$H)
plot(grid_Q, (11_Q - max(11_Q)), type = "1", main = "Q", log = "x")
abline(v = cmodel$Q)
plot(grid_sigma, (ll_sigma - max(ll_sigma)), log = "x", type = "l", main = "sigma")
abline(v = cmodel$sigma)
```

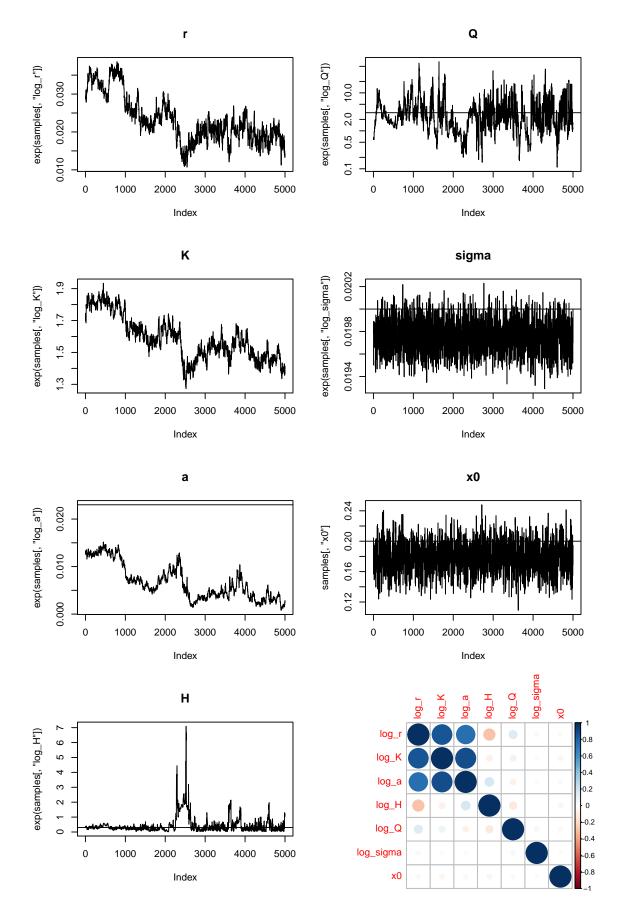


```
ll_QH <- apply(expand.grid(grid_Q, grid_H), 1, function(pair)</pre>
  if(pair[2] > (0.6 / 3) * pair[1] + 0.18 - (0.6 / 3)) NA else
  get_ll(a = cmodel$a, x = cmodel$x, r = cmodel$r, K = cmodel$K,
         H = pair[2], Q = pair[1], sigma = cmodel$sigma))
11_Ka <- apply(expand.grid(grid_K, grid_a), 1, function(pair)</pre>
  get_ll(a = pair[2], x = cmodel$x, r = cmodel$r, K = pair[1],
         H = cmodel$H, Q = cmodel$Q, sigma = cmodel$sigma))
11 rK <- apply(expand.grid(grid r, grid K), 1, function(pair)</pre>
  get_11(x = cmodel\$x, r = pair[1], K = pair[2], a = cmodel\$a,
         H = cmodel$H, Q = cmodel$Q, sigma = cmodel$sigma))
ll_Ksigma <- apply(expand.grid(grid_K, grid_sigma), 1, function(pair)</pre>
  get_ll(x = cmodel$x, r = cmodel$r, K = pair[1], a = cmodel$a,
         H = cmodel$H, Q = cmodel$Q, sigma = pair[2]))
layout(matrix(1:4, 2, 2)); par(mar = c(4, 4, 1, 1))
image(grid_Q, grid_H, matrix((ll_QH - max(ll_QH, na.rm = T)), length(grid_Q), length(grid_H)))
points(cmodel$Q, cmodel$H)
##
image(grid_K, grid_a, matrix((ll_Ka - max(ll_Ka)), length(grid_K), length(grid_a)))
points(cmodel$K, cmodel$a)
image(grid_r, grid_K, matrix((ll_rK - max(ll_rK)), length(grid_r), length(grid_K)))
points(cmodel$r, cmodel$K)
##
image(grid_K, grid_sigma, matrix((ll_Ksigma - max(ll_Ksigma)), length(grid_a), length(grid_sigma)))
points(cmodel$K, cmodel$sigma)
```



```
model$setData(x = cmodel$x)
cmodel <- compileNimble(model)</pre>
## compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.
## compilation finished.
## block sampler
system.time({
 mcmcConf <- configureMCMC(model)</pre>
 mcmcConf$getSamplers()
  # mcmcConf$removeSamplers(c())
 mcmcConf$addSampler(target = c("log_r", "log_K", "log_a",
                               "log_H", "log_Q", "log_sigma", "x0"),
                     type = 'RW_block')
  mcmcConf$getSamplers()
})
## Note: Assigning an RW_block sampler to nodes with very different scales can result in low MCMC effic
##
      user system elapsed
##
    8.602
           0.091
                    8.789
system.time({
 mcmc <- buildMCMC(mcmcConf)</pre>
  Cmcmc <- compileNimble(mcmc, project = model)</pre>
})
## compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.
## compilation finished.
##
      user system elapsed
           0.713 22.448
##
   20.385
##
n_iterations <- 1e4
system.time({
 Cmcmc$run(n_iterations, nburnin = n_iterations / 2)
})
## |-----|----|
## |-----|
     user system elapsed
## 176.546
           1.821 190.964
samples <- as.matrix(Cmcmc$mvSamples)</pre>
layout(matrix(1:8, 4, 2))
plot(exp(samples[, 'log_r']), type = "l", main = "r")
abline(h = r)
plot(exp(samples[, 'log_K']), type = "l", main = "K")
abline(h = K)
plot(exp(samples[, 'log_a']), type = "l", main = "a",
     ylim = range(exp(samples[, 'log_a']), a))
abline(h = a)
plot(exp(samples[, 'log_H']), type = "l", main = "H")
abline(h = H)
plot(exp(samples[, 'log_Q']), type = "l", main = "Q", log = "y")
```

```
abline(h = Q)
plot(exp(samples[, 'log_sigma']), type = "l", main = "sigma")
abline(h = sigma)
plot(samples[, 'x0'], type = "l", main = "x0")
abline(h = x0)
corrplot::corrplot(cor(samples[, c("log_r", "log_K", "log_a", "log_H", "log_Q", "log_sigma", "x0")]))
```



Notice a, r, and K are highly correlated. Taking out two of them stablizes somewhat, but dependence in the posterior distributions arises between other variables.

```
## block sampler
system.time({
 mcmcConf <- configureMCMC(model)</pre>
  mcmcConf$getSamplers()
  mcmcConf$removeSamplers(c("log_a", "log_K"))
  mcmcConf$addSampler(target = c("log_r", "log_H", "log_Q", "log_sigma", "x0"),
                      type = 'RW_block')
  mcmcConf$getSamplers()
})
## Note: Assigning an RW_block sampler to nodes with very different scales can result in low MCMC effic
##
      user system elapsed
     8.672
            0.081
                     8.844
system.time({
  mcmc <- buildMCMC(mcmcConf)</pre>
  Cmcmc <- compileNimble(mcmc, project = model, resetFunctions = T)</pre>
})
## compiling... this may take a minute. Use 'showCompilerOutput = TRUE' to see C++ compilation details.
## compilation finished.
##
      user system elapsed
##
   19.434
           0.585 20.355
##
n_iterations <- 1e4
system.time({
  Cmcmc$run(n_iterations, nburnin = n_iterations / 2)
})
## |-----|-----|
##
      user system elapsed
## 115.038
           0.824 117.418
samples <- as.matrix(Cmcmc$mvSamples)</pre>
layout(matrix(1:6, 3, 2))
plot(exp(samples[, 'log_r']), type = "l", main = "r")
abline(h = r)
plot(exp(samples[, 'log_H']), type = "l", main = "H")
abline(h = H)
plot(exp(samples[, 'log_Q']), type = "l", main = "Q", log = "y")
abline(h = Q)
plot(exp(samples[, 'log_sigma']), type = "l", main = "sigma")
abline(h = sigma)
plot(samples[, 'x0'], type = "l", main = "x0")
abline(h = x0)
corrplot::corrplot(cor(samples[, c("log_r", "log_H", "log_Q", "log_sigma", "x0")]))
```

