# Estimation of constant stock-recruitment parameters for mixed fisheries Management Strategy Evaluation

Ghassen Halouani & Cóilín Minto

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# FishKOSM Project

### FishKOSM Project

# Mixed fisheries, MSY ranges and Management Strategy Evaluation

Investigate the performance of MSY reference values and ranges to provide practical and operational advice on the management of mixed demersal fisheries.

### FishKOSM Project

FLBEIA toolbox which facilitates the development of bio-economic impact assessments of fisheries management strategies.

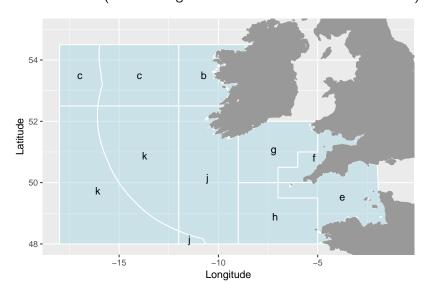
The simulation is divided in two worlds:

- the operating model (OM, the real world)
- the management procedure model (MPM, the perceived world)

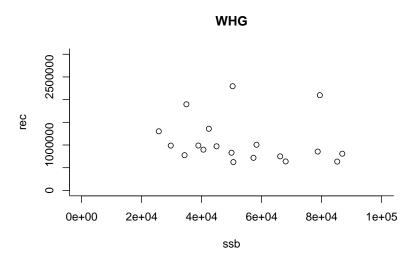
# Estimation of S-R parameters

#### Study area

#### Divisions 7.e-k (Eastern English Channel and Southern Celtic Seas)



#### Data



 $<sup>^{0}</sup>$ Data from the ICES Stock Assessment Graphs database <a href="http://sg.ices.dk">http://sg.ices.dk</a> for the period (1999 - 2017)

### Standard estimation of the S-R parameters

#### The Beverton and Holt stock recruitment relationship

$$R = \frac{\alpha B}{\beta + B}$$

#### Where

• R: the recruitment

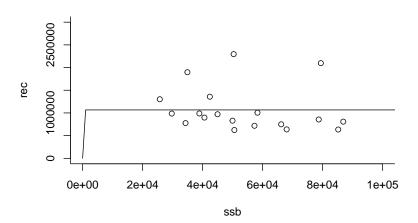
• B: the spawing stock biomass

ullet  $\alpha$  : the maximum rectruitment

•  $\beta$  : the spawing stock biomass needed to produce  $\alpha/2$ 

#### Standard estimation of the S-R parameters

Direct estimation of  $\alpha$  and  $\beta$  using a likelihood function



#### The Beverton and Holt stock recruitment relationship

$$R = \frac{B}{\alpha + \beta B}$$

#### Where

R: the recruitement

• B: the spawing stock biomass

ullet  $\alpha$  : the inverse of the initial slope of the curve

•  $\beta$ : the inverse of asymptotic recruitment

 $\alpha$  and  $\beta$  parameters

$$\alpha = \frac{S_0}{R_0} \frac{1 - h}{4h}$$
$$\beta = \frac{5h - 1}{4hR_0}$$

#### Where

- $R_0$ : the recruitment when F=0
- h: the steepness parameter defined as the proportion of unfished recruitment  $R_0$  produced by 20% of unfished population (spawning biomass  $S_0$ )
- $S_0$ : the spawing stock biomass when at F=0

<sup>&</sup>lt;sup>0</sup>from Mangel et al. 2009

The Beverton-Holt spawner-recruit function expressed with steepness parameter

$$R = \frac{0.8R_0hS}{0.2S_0(1-h) + (h-0.2)S}$$

#### Where

- R: the recruitement
- R<sub>0</sub>: the unfished recruitment
- h: the steepness parameter defined as the proportion of unfished recruitment  $R_0$  produced by 20% of unfished population (spawning biomass  $S_0$ )
- S: the spawing stock biomass

<sup>&</sup>lt;sup>0</sup>from Mace and Doonan 1988

Etimation of N by age when F = 0

$$N_{a+1} = N_a e^{-M_a}$$

N : Abundance of Whiting

M<sub>a</sub>: Natural mortality

• *a* : age

Etimation of S by age when F = 0

$$S_0 = \sum_{a=0}^{T} N_a W_a Mat_a$$

#### Where

•  $S_0$ : the unfished spawing biomass

•  $N_a$ : the number at age

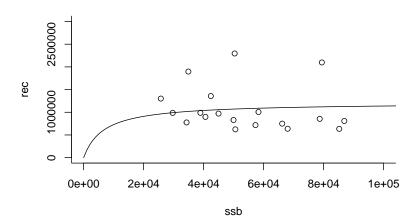
•  $W_a$ : the weight at age

• Mata: the maturity at age

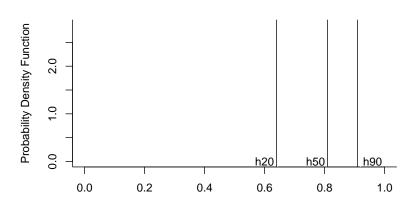
#### Using a fixed value for the steepness parameter

- Estimation of  $\alpha$  and  $\beta$  using a likelihood function
- h = 0.81

# Estimation of the S-R parameters using **the steepness h**Using a fixed value for the steepness parameter

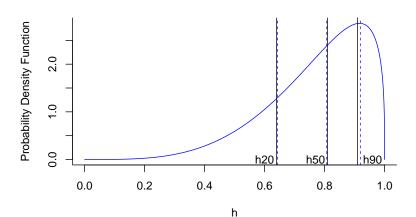


Estimated from (Myers et al. 1999)

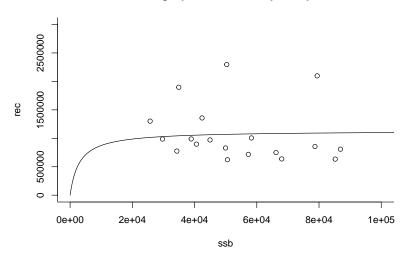


h

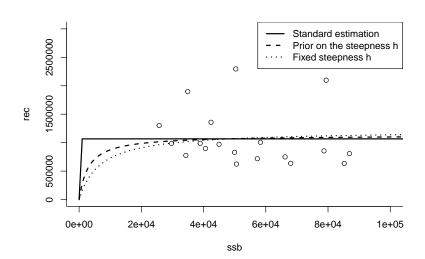
Estimated from (Myers et al. 1999)



#### WHG SRR using a prior on the steepness parameter h



#### Estimation of the S-R parameters



### R Code

### Standard estimation of the S-R parameters

#### Direct estimation of $\alpha$ and $\beta$ using a likelihood function

```
LL <- function(par){
    a <- exp(par[1])
    b <- exp(par[2])
    c <- exp(par[3])

R <- a*ssb / (b*ssb)

11 = dlnorm(rec, meanlog = log(R)-(c^2)/2, sdlog = c, log = T)
    -sum(11)
}

LL_opt_WHG <- optim(par = log(c(15e3, 1e4, 2)), fn = LL)</pre>
```

#### Using a fixed value for the steepness parameter

```
## likelihood function
LL <- function(par) {
  R0 \leftarrow exp(par[1])
  c <- exp(par[2])
  M <- M # vector of natural mortality by age
  N \leftarrow rep(NA, 7)
  N[1] <- RO
  for (i in 1:6) {
    N[i+1] \leftarrow N[i] * exp(-M[i])
  data$N <-N
  S \leftarrow apply(data, 1, function(x) x[2]*x[3]*x[4])
  SO \leftarrow sum(S)
  R \leftarrow 0.8*R0*h*ssb / (0.2*S0*(1-h) + ssb*(h-0.2))
  11 = dlnorm(rec, meanlog = log(R) - (c^2)/2, sdlog = c, log = T)
  -sum(11)
LL_opt <- nlminb(start = log(c(10000, 0.2)), objective = LL)
```

#### Adding a prior on the steepness parameter Estimation of the distribution of the prior

```
# adding a prior for the steepness parameter
zmed <- 0.81
z20 <- 0.64
z80 <- 0.91

## sum of square of the difference at z20 and z80
ssq <- function(alpha) {
  beta <- (alpha - 1/3) / zmed - alpha + 2/3
  z20.pred <- qbeta(p = 0.2, shape1 = alpha, shape2 = beta)
  z80.pred <- qbeta(p = 0.8, shape1 = alpha, shape2 = beta)
  return((z20.pred-z20)^2 + (z80.pred-z80)^2)
}
fit <- optim(par = 2, fn = ssq, method = "Brent", lower = 1, upper = 50)
alpha.hat <- fit$par
beta.hat <- (alpha.hat - 1/3) / zmed - alpha.hat + 2/3</pre>
```

```
# Likelihood function : Estimation of RO using a prior en h
LL <- function(par) {
  # parameters to estimate
  R0 \leftarrow exp(par[1])
  sdev <- exp(par[2])
  h <- exp(par[3])
  # abundance by age class
  N \leftarrow rep(NA, 8)
  N[1] <- RO
  for (i in 1:7) {
    N[i+1] \leftarrow N[i]*exp(-M[i])
  # 50
  data$N <- N
  S \leftarrow apply(data, 1, function(x) x[2]*x[3]*x[4])
  SO \leftarrow sum(S)
  R \leftarrow 0.8*R0*h*ssb / (0.2*S0*(1-h) + ssb*(h-0.2))
  11 = dlnorm(rec, meanlog = log(R)-(sdev^2)/2, sdlog = sdev, log = T)
  11p <- 11 + dbeta(h, shape1 = alpha.hat, shape2 = beta.hat) # adding the prior
  -sum(11p)
LL_opt \leftarrow nlminb(start = log(c(10000, 0.2, 0.8)), objective = LL)
RO.hat <- exp(LL_opt$par[1])
```