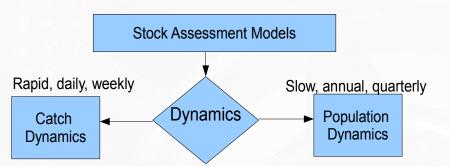
A4A



Catch Dynamics Stock Assessment Models for Data-Moderate Fisheries





$$\frac{dC}{dt} = f(E, N)$$

Fishing enters <u>directly</u> through nominal fishing effort at high frequency

Catch rate is the result of one observed predictor (effort) and one latent predictor (abundance)

Solutions lead to fishing operational stock assessment models

$$f(E, N) = qEN$$

$$\frac{dC}{dt} = f(F, N)$$

Fishing enters indirectly through its effect on fish survival at the scale of fish population dynamics (annual, quarter)

Catch rate is the result of two latent predictors (fish mortality and abundance)

Solutions lead to population dynamics stock assessment models

$$f(F,N)=FN$$

- •Population dynamics stock assessment models are vastly predominant.
- •Nominal effort is mostly used to fit catch rate models to generate annual indexes of fish abundance

A4A

Why, given that nominal effort is the single observed cause of catch?

- •Fish population dynamics is sexier than fishing operational dynamics
- •Fish population dynamics creates more jobs
- •In textbooks stock assessment is synonym with fish population dynamics (e.g. Quinn-Deriso)
- •Founders such as Baranov, Beverton, Holt, Ricker, were biologists?



Catch Dynamics Models

$$\frac{dC}{dt} = f(E, N)$$
$$f(E, N) = qEN$$

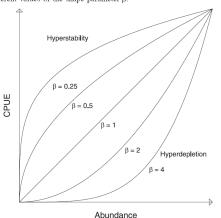
$$C(t) = q \int E(t) N(t) dt = q E(t) \int_{0}^{t} \left(N_{0} e^{-Mt} - \int_{0}^{t-\delta t} C(\tau) e^{-M(t-\delta t)} d\tau \right) dt$$

Switching to discrete time steps

$$C_{t} = q E_{t} N_{t} e^{-M/2} = q E_{t} \left(N_{0} e^{-Mt} - e^{-M/2} \sum_{i < t} C_{i} e^{-(t-i-1)M} \right) e^{-M/2}$$

This is Leslie-Davies-Chapman depletion model used to do stock assessment of some squid stocks and stocks that are data-poor

Fig. 1. Relationship between CPUE and abundance based on different values of the shape parameter β .



Harley, Myers,. Dunn CJFAS 58:1761, 2001

$$C = q \cdot f \cdot N;$$

$$q = \alpha f^{\lambda} N^{\beta};$$

$$C = \alpha f^{(1+\lambda)} N^{(1+\beta)}.$$

Bannerot, Austin TAFS 112:608, 1983

Catch Dynamics Models – Process Model

Enter hyperstability/hyperdepletion, let's call it "abundance response"

$$\frac{C_t}{E_t} = k N_t^{\beta} e^{-M/2} = k \left(N_0 e^{-Mt} - e^{-M/2} \sum_{i=1}^{t-1} C_i e^{-M(t-i-1)} \right)^{\beta} e^{-M/2}$$

Enter gear saturability/synergy, let's call it "effort response"

$$C_{t} = k E_{t}^{\alpha} N_{t}^{\beta} e^{-M/2} = k E_{t} \left(N_{0} e^{-Mt} - e^{-M/2} \sum_{i=1}^{t-1} C_{i} e^{-M(t-i-1)} \right)^{\beta} e^{-M/2}$$

Relax the depletion assumption, allow for replenishment

$$C_{t} = k E_{t}^{\alpha} N_{t}^{\beta} e^{-M/2} = k E_{t}^{\alpha} \left(N_{0} e^{-Mt} + \sum_{i=1}^{t} P_{i} e^{-M(t-i)} - e^{-M/2} \sum_{i=1}^{t-1} C_{i} e^{-M(t-i-1)} \right)^{\beta} e^{-M/2}$$

$$t > 0, C_{t} \ge 0, E_{t} \ge 0, k > 0, N_{0} > 0, \alpha > 0, \beta > 0, M > 0, -N_{0} e^{-M} \le P_{i} \le +\infty$$





Catch Dynamics Models – Observation Model

$$\begin{split} C_t &= k \, E_t^\alpha N_t^\beta e^{-M/2} = k \, E_t^\alpha \Biggl(N_0 e^{-Mt} + \sum_{i=1}^t P_i e^{-M(t-i)} - e^{-M/2} \sum_{i=1}^{t-1} C_i e^{-M(t-i-1)} \Biggr)^\beta e^{-M/2}, \\ t &> 0, C_t \geq 0, E_t \geq 0, k > 0, N_0 > 0, \alpha > 0, \beta > 0, M > 0, -N_0 e^{-M} \leq P_i \leq +\infty \\ \chi_t &= C_t + \varepsilon_t \quad Additive \\ \chi_t &= C_t e^{\varepsilon_t} \quad Multiplicative \\ \varepsilon_t \sim Normal\left(0, \sigma^2\right) \end{split}$$

The catch of one fleet at one time step is the result of two causes:

- Nominal effort (*E*)
- Stock abundance $(N_0, \{P_i\})$,

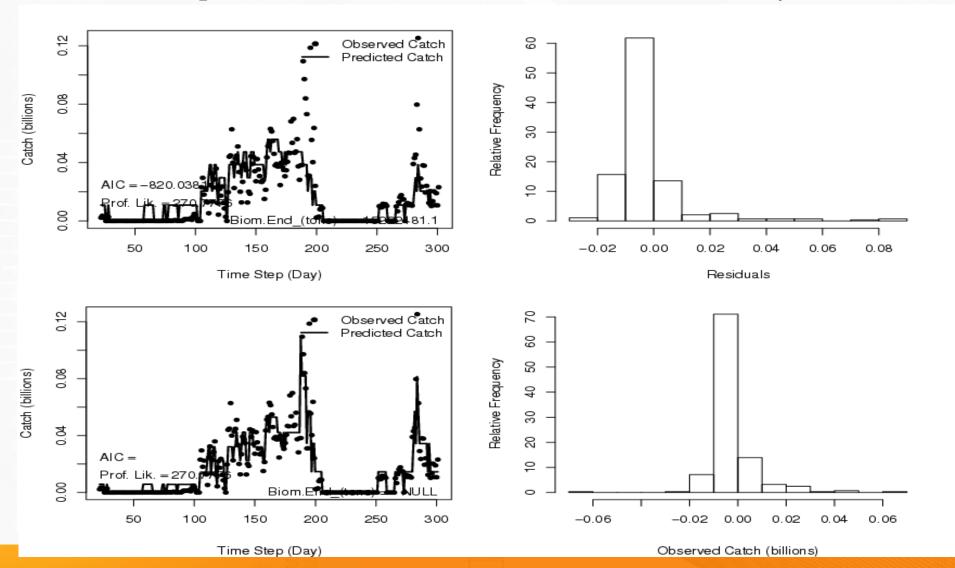
and five constants,

- A scaling parameter (k),
- An effort response parameter (α) ,
- An abundance response parameter (β),
- A natural mortality parameter (*M*),
- A stochastic response parameter (σ^2)



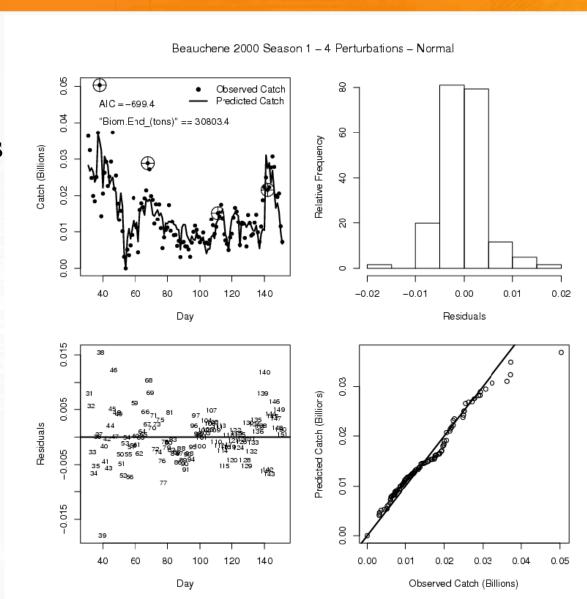
Pure Depletion AIC=-820

One Perturbation at Day 189=-999

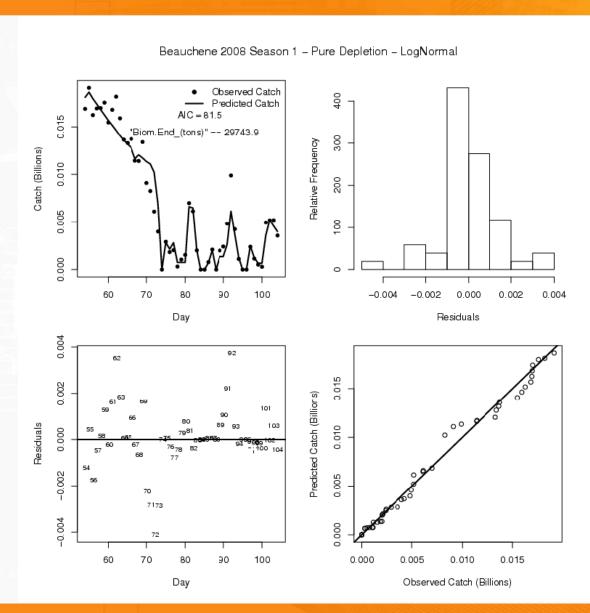




2000 Loligo summer season in the Falklands: a case of 4-perturbations

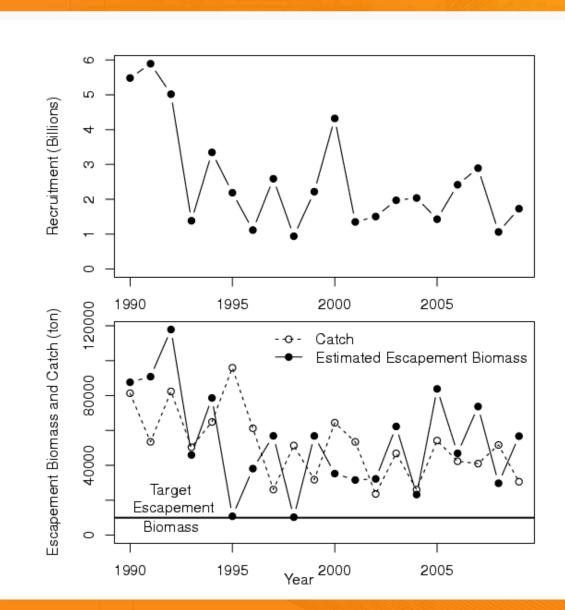


2008 Loligo summer season in the Falklands: a case of 0-perturbations, pure depletion





Real-time management to achieve a target





Catch Dynamics Models

	Populations Dynamics Models	Catch Dynamics Models
Data aggregation	Data aggregated annually/quarterly to scale it to fish populations dynamics	Data aggregated daily or weekly because the fishing operation has a fast dynamics
Biological comprehensiveness	Very comprehensive from the fish point of view, allows estimation of fish abundance and productivity	Limited, only allows estimation of fish abundance, but not productivity.
Fishery comprehensiveness	Quite comprehensive, allows estimation of gear selectivity, catchability.	Quite comprehensive, allows estimation of gear catchability, saturability.
Cost	Very expensive, requires armies of biologists measuring fish, counting rings in bones, carrying out surveys.	Very cheap, requires observers at sea or at ports armed just with a ruler and scales.
Reaction time	Slow, the status of the stock is estimated for years past, and management measures come two or more years after the stock is estimated at risk	Fast, the stock abundance status is realized for the current season and management measures can be implemented for the current season.
Implementation of management advice	Difficult, because of their slow dynamics (the dynamics of the fish population) management tend to be framed in terms of annual TAC, but catch limits are difficult to enforce	Easy, because these models at their rapid dynamics return estimates of stock status in real time so management can be based on effort limits.
Statistical underpinnings	Integrated likelihood/Bayesianity, observed catch at age, catch at length, indices are caused by several population and fisheries processes.	Nonlinear regression/recursive models, observed catch is caused by observed fishermen effort and latent fish abundance.

Catch	Dynamics	Models
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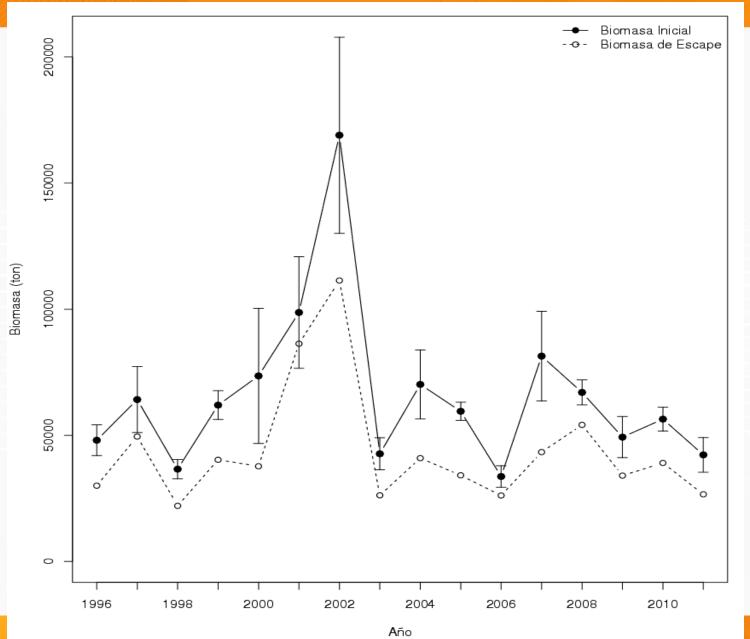
	Biomass	Season	Catch
'Data'	B0_1	1	C_1
	B0_2	2	C_2
	B0_S	S	C_S

Biomass Dynamics
$$B(s)=B(s-1)+rB(s-1)\left(1-\left(\frac{B(s-1)}{K}\right)^p\right)-C(s-1), s=1,...,S$$
 Model

Hierarchical Likelihood
$$\blacktriangleright l(r, K, p; \{B_s\}) = -\frac{1}{2} \sum_{s=1}^{S} \log(\hat{\sigma}_{B(s)}^2) - \frac{\sum_{s=1}^{S} (\log(\hat{B}(s)) - \log(\hat{B}(s)))^2}{2\hat{\sigma}_{\hat{B}(s)}^2}$$
Model
$$\hat{v}(\hat{MSY}) = \begin{bmatrix} \frac{\partial MSY}{\partial r} & \frac{\partial MSY}{\partial K} \end{bmatrix} \begin{bmatrix} \hat{v}(\hat{r}) & cov(\hat{r}, \hat{K}) \\ cov(\hat{K}, \hat{r}) & \hat{v}(\hat{K}) \end{bmatrix} \begin{bmatrix} \frac{\partial MSY}{\partial r} \\ \frac{\partial MSY}{\partial K} \end{bmatrix}$$

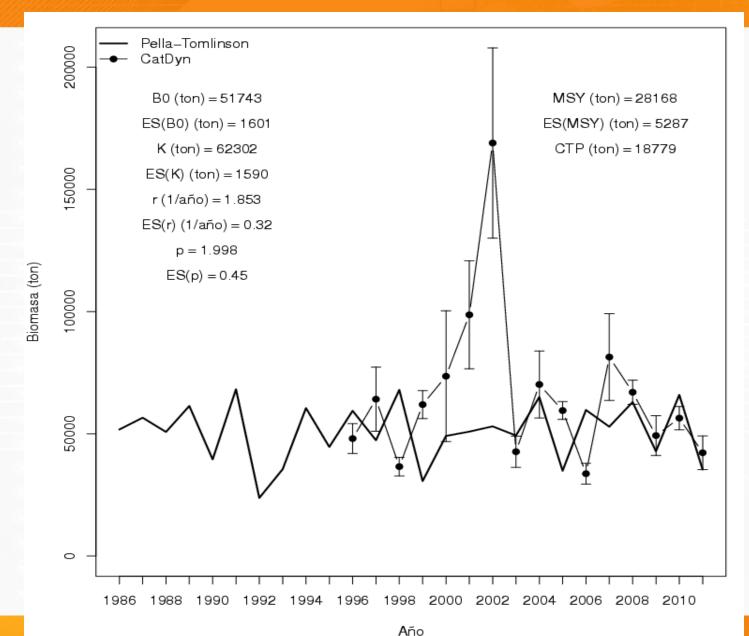
$$\hat{v}(\hat{MSY}) = \begin{bmatrix} \frac{\partial MSY}{\partial r} & \frac{\partial MSY}{\partial K} \end{bmatrix} \begin{bmatrix} \hat{v}(\hat{r}) & cov(\hat{r}, \hat{K}) \\ cov(\hat{K}, \hat{r}) & \hat{v}(\hat{K}) \end{bmatrix} \begin{bmatrix} \frac{\partial MSY}{\partial r} \\ \frac{\partial MSY}{\partial K} \end{bmatrix}$$

Catch Dynamics Models



Sea urchin Fishery In Southern Chile





Sea urchin Fishery In Southern Chile



Implementation of catch dynamics models

- •CatDyn is an R package available from CRAN that allow estimation of CDMs with 0, 1, ..., 4 perturbations, additive or multiplicative observation error, estimated or fixed natural Mortality.
- •The basic data needed is high-frequency (daily or weekly) time series of paired observations of catch in numbers and nominal effort in a measure that is approximately exact.
- •If the catch is recorded in biomass, then the basic data must also have a matching vector of mean bodymass per time step



Way ahead/work in progress:

- •Write down the multi-fleet version of the model
- •Write down the functions to fit the multi-annual production models to estimate productivity/MSY.
- •Set up simulations to test the models under perfect conditions and under conditions where assumptions are violated
- •Write the models in ADMB to speed up and secure convergence
- •Create a GUI to control the package from R, using PBSmodelling and PBSadmb
- •Apply the model to the following list of stocks:
 - •Redfish fished by a Spanish fleet in the North Sea
 - •Blue ling fished by an Icelandic fleet in Iceland EEZ
 - •Hake fished by two fleets in central Chile
 - •Squid fished by three fleets on the Eastern coast of the USA
 - •Shrimps and finfish in the Persian Gulf



Set up a spatially distributed stock with RandomFields

Harvest the stock by a group of agents over a period of time

Introduce random pulses of abundance increase during the harvest period

Observe the harvest process in a perfect manner or incompletely

Estimate abundance and fishing operational parameters with catdyn

Calculate bias, variance, mean square Error over many iterations

Nuclei may move or stay put; nuclei will decay Because of natural mortality Agents will differ in efficiency or will be Identical; error will be additive or multiplicative Pulses will be caused by recruitment or agents expanding area of opearation

Counting the right number of perturbations or missing a few

Effort will be an exact predictor

or a random variable, raised or

not