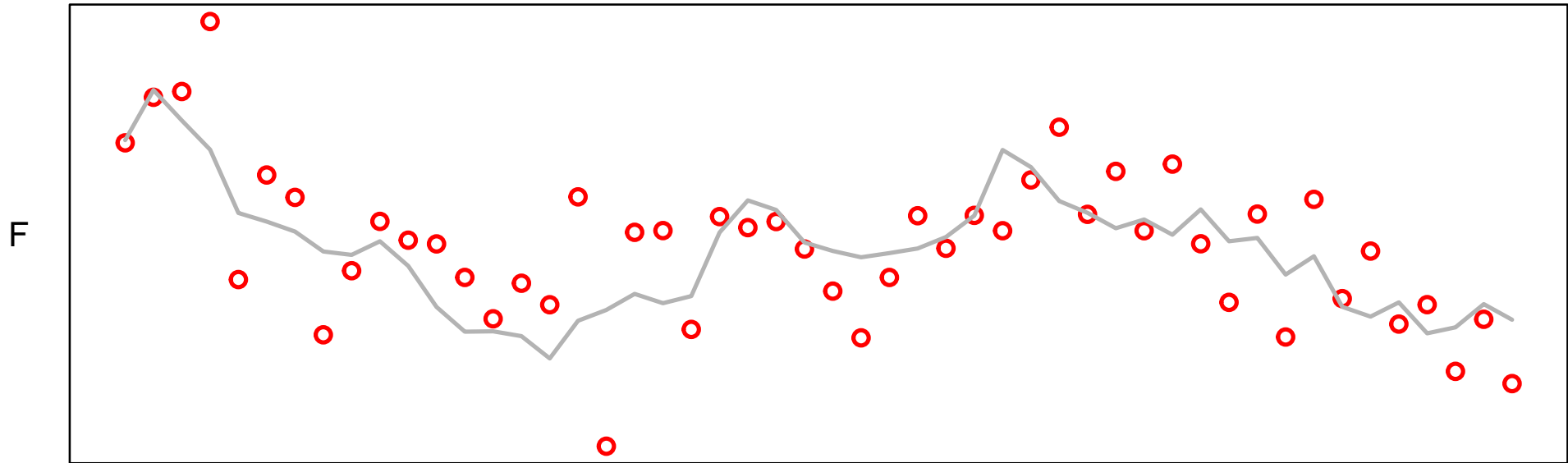


# A basic statistical assessment model

Anders Nielsen, Ethan Lawler, & Sean Anderson

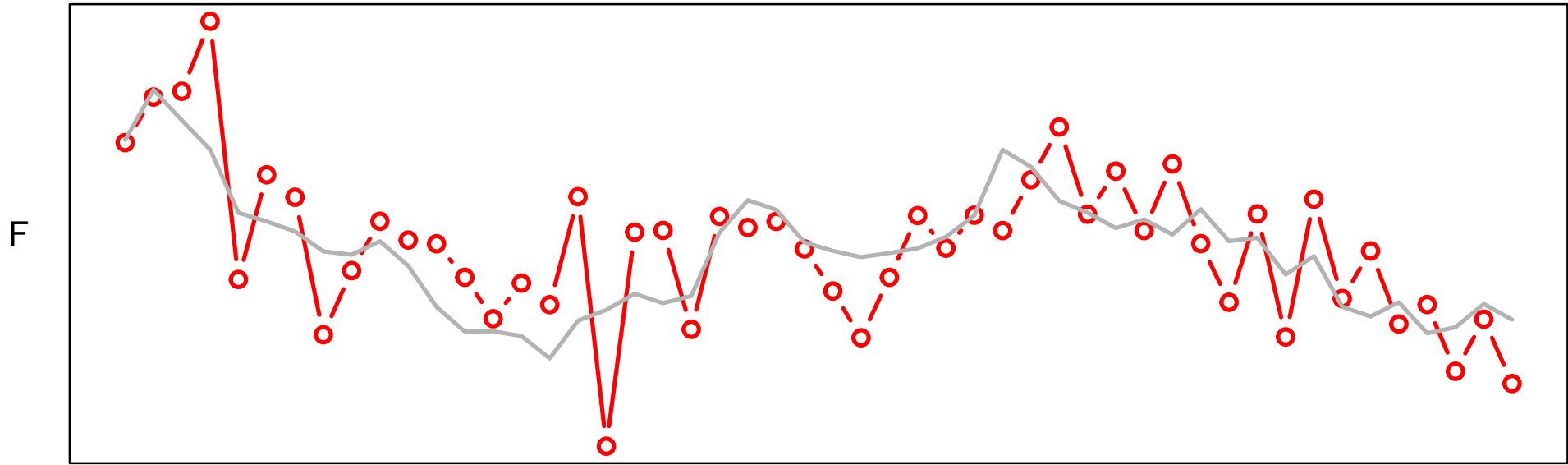
an@aqua.dtu.dk

# The problem in a nutshell



- Consider this example:
  - The true underlying fishing mortality (here grey) we can never observe
  - We only observe  $Y$  (here red circles) which is  $F + \text{'noise'}$
  - The key question is: How do we estimate/predict the gray process?

# Deterministic model estimates



- If we assume no observation error the estimate of  $F$  is  $Y$
- Too fluctuating
- No quantification of uncertainties
- Does this remind you of anything?

# Fish Stock Assessment

**Problem:** How many fish (relative or absolute) are left in the ocean?

**Data:**

$C_{a,y}$ : Yearly catches (divided into age-classes)

$I_{a,y}$ : Scientific surveys

	Year e.g. 1963–2022								
Age e.g. 1–7									
				$C_{a,y}$					

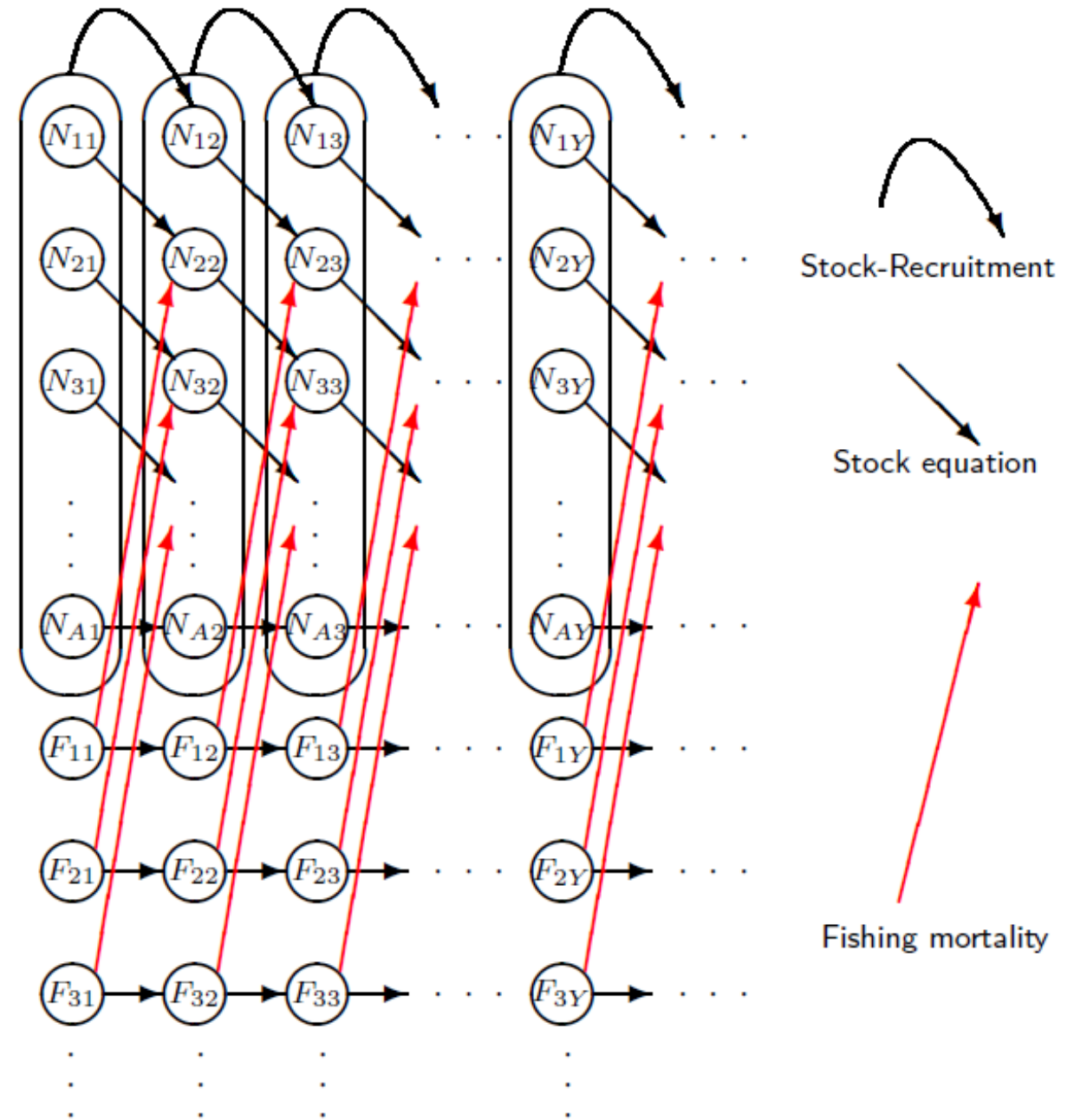
	Year e.g. 1983–2022								
Age e.g. 1–5									
				$I_{a,y}$					

Often we have catches ( $C_{f,a,y}$ ) from more than one fleet and indices ( $I_{s,a,y}$ ) from more than one survey, but here we keep it simple.

**Important output:** “Reference points” SSB,  $\overline{F}$ , ...

# Assessment Models

- Based on a standard set of equations describing the structure of the system
- Account for selectivity
- Account for different assumptions about 'natural' mortality
- can account for different effort numbers
- Can include different assumptions about recruitment
- All of these things have to be accounted for



# Basic equations

**Stock equation:** The number of fish in a cohort is expected to follow:

$$\frac{dN_t}{dt} = -(F_t + M_t)N_t$$

If  $F$  and  $M$  are assumed constant within each year we get:

$$N_{a+1,y+1} = N_{a,y}e^{-(F_{a,y}+M_{a,y})}$$

**Catch equation:** The number of fish in a cohort after one year can be separated into:

$$N_{a,y} = \underbrace{N_{a+1,y+1}}_{\text{survived}} + \underbrace{C_{a,y} + D_{a,y}}_{\text{died}}$$

The part of the cohort that was caught during the year is:

$$C_{a,y} = \frac{F_{a,y}}{F_{a,y} + M_{a,y}} \left( 1 - e^{-(F_{a,y}+M_{a,y})} \right) N_{a,y}$$

**Stock–recruitment equation:** Obviously connected — Different opinions about how

# Calculation example

- Imagine that a cohort consists of  $N_0 = 1000$  fish at time 0 and we know  $M = 0.2$  and  $F = 0.1$ , then we can calculate

```
> N0 <- 1000; F <- 0.1; M <- 0.2
```

```
> N1 <- N0*exp(-F-M)
```

```
> C0 <- F/(F+M)*(1-exp(-F-M))*N0
```

- This gives  $N_1 \approx 741$  survivors and a catch of 0 year olds of  $C_0 \approx 86$
- We can continue the calculation and find the number surviving to age 2 and the catch of 1 year olds:

```
> N2 <- N1*exp(-F-M)
```

```
> C1 <- F/(F+M)*(1-exp(-F-M))*N1
```

- These calculations can be gathered in a loop — try it. Start with:

```
> N0<-1000; F<-.1; M<-.2
```

```
> N<-c(N0, rep(NA,19))
```

- In addition what is the cohort size at time 1.75?

# Deterministic models

- A deterministic model is a model where **observation noise is ignored**
- Typically catches are assumed known without error
- Most commonly applied fish stock assessment models are (semi-)deterministic
- These algorithms work (very simply put) by:
  - 0: Guess the number of survivors  $N_{A+1,y}$  and  $N_{a,Y+1}$
  - 1: Back calculate ( $\nwarrow$ ) all  $N_{a,y}$  by subtracting catch and natural mortality
  - 2: Use surveys to adjust all  $N_{a,y}$  and update survivors accordingly
  - 3: Repeat 1-3 until survivors converge
- Doing 0-1 just ones is known as Virtual Population Analysis

		Year e.g. 88–08							
Age e.g. 1–7		$\nwarrow$					$\nwarrow$		
			$\nwarrow$					$\nwarrow$	
				$\nwarrow$	$C_{a,y}$				$\nwarrow$
					$\nwarrow$				
						$\nwarrow$			

$\vdots$   
 $N_{a,Y+1}$   
 $\vdots$

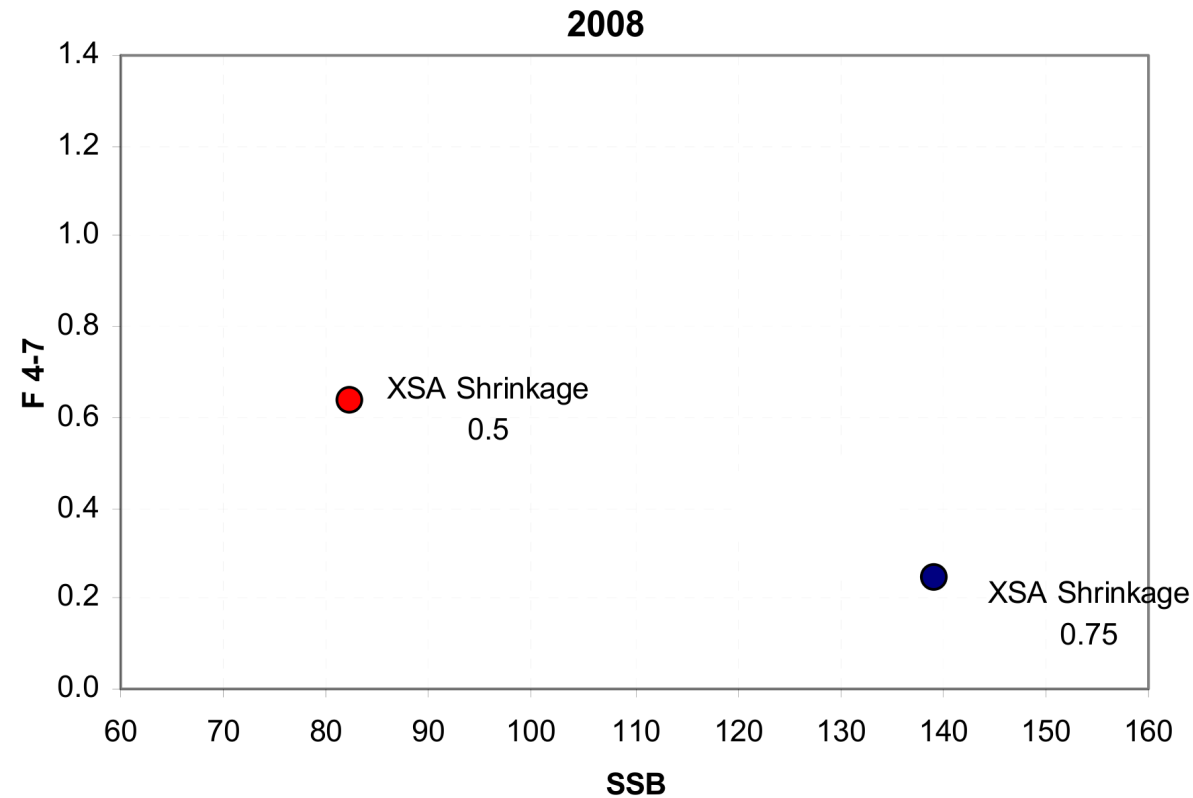
$\cdots N_{A+1,y} \cdots$



# Features of deterministic models

- + Super fast to compute
- + Fairly simple to explain the path from data to stock numbers (especially VPA)
- Difficult to explain why it works (converges), and what a solution mean
- These algorithms contain many ad-hoc settings (shrinkage, tapered time weights, ...) that makes them less objective
- No quantification of uncertainties within model
- ? What exactly is the model
  - The assumptions are difficult to identify and verify
  - With no clearly defined model more ad-hoc methods are needed to make predictions
  - No framework for comparing models (different settings)

# Example: F-shrinkage for Eastern Baltic Cod



- These differences are not small and theoretical
- There are no objective way to choose between these two deterministic approaches
- Things would be simpler if we had a statistical model

# Now we know

- **Assessment** Basic equations:

- Stock equation: The number of fish in a cohort is expected to follow:

$$N_{a+1,y+1} = N_{a,y} e^{-(F_{a,y} + M_{a,y})}$$

- and further if we need the stock size some fraction  $\tau$  into a year

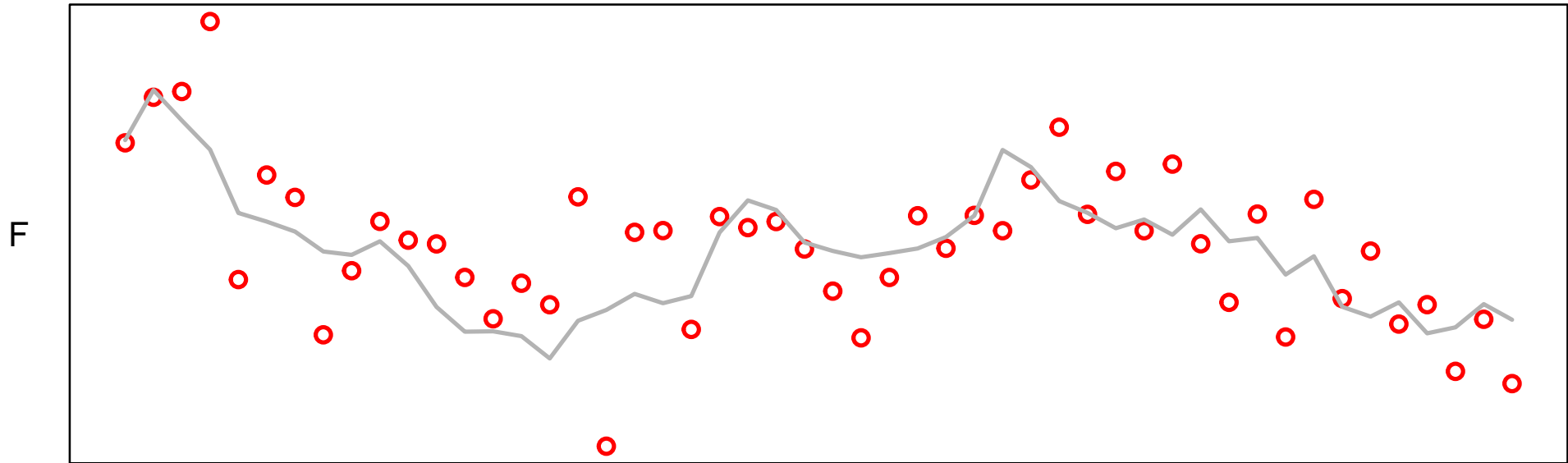
$$N_{a+\tau,y+\tau} = N_{a,y} e^{-(F_{a,y} + M_{a,y})\tau}$$

- Catch equation: The part of the cohort that was caught during the year is:

$$C_{a,y} = \frac{F_{a,y}}{F_{a,y} + M_{a,y}} \left( 1 - e^{-(F_{a,y} + M_{a,y})} \right) N_{a,y}$$

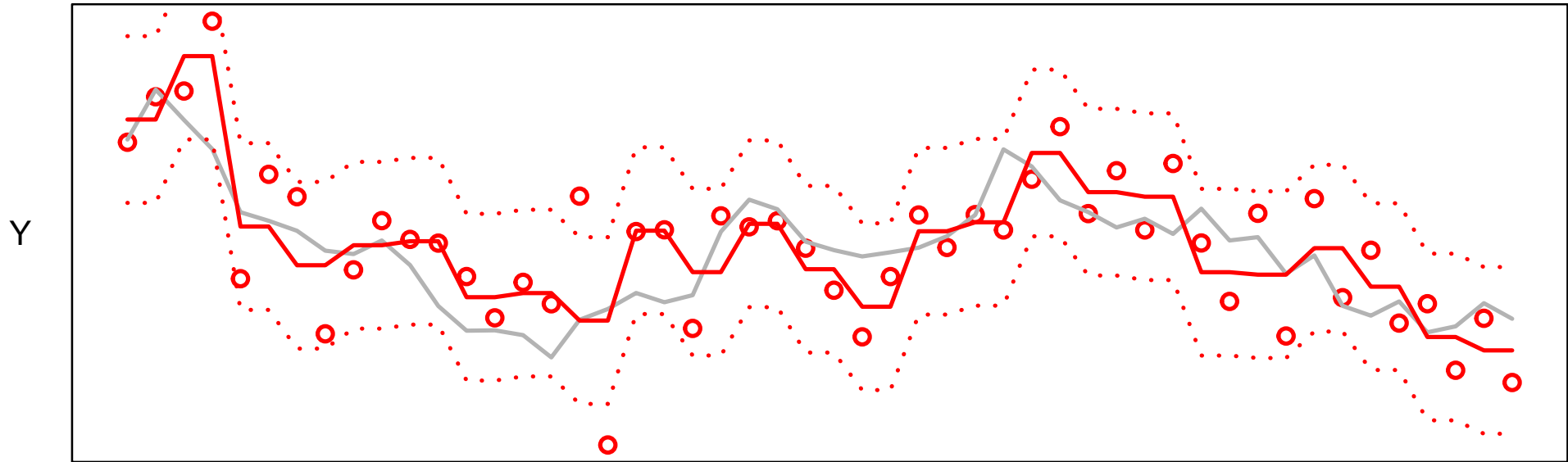
- **Deterministic assessment models** What is the main assumption in a deterministic model?

# Allowing observation noise



- How do we:
  - Estimate model parameters?
  - Estimate uncertainties?
  - Compare models?

# Model with observation noise



- It allows for observation noise
- We first had to group the observations (here pairs)
- Choice is arbitrary
- The reconstructed track appear OK
- The model contains a lot (26) of model parameters
- Uncertainties are estimated but the confidence interval seems too wide

# Now we know

- **Assessment** Basic equations:

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- Catch equation: The part of the cohort that was caught during the year is:

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- **Deterministic assessment models** Assume catches are known without error
- **Statistical/stochastic assessment models** Allow observation noise:
  - Explicit about assumptions w.r.t. observation noise
  - Complete solution for: estimation, uncertainty, model comparison

# A full parametric statistical model

- The log catches are assumed to follow:

$$\log(C_{a,y}) \sim \mathcal{N} \left( \log \left( \frac{F_{a,y}}{Z_{a,y}} (1 - e^{-Z_{a,y}}) N_{a,y} \right), \sigma_c^2 \right), \text{ where}$$

$$F_{a,y} = F_y F_a, \text{ with } F_{a=5} = F_{a=6} = F_{a=7} = 1, \text{ and } Z_{a,y} = F_{a,y} + M_{a,y}$$

- The log catches from the survey are assumed to follow:

$$\log(I_{a,y}) \sim \mathcal{N} \left( \log \left( Q_a e^{-Z_{a,y} T} N_{a,y} \right), \sigma_s^2 \right), \text{ where}$$

$T$  is the fraction into the year where the survey is taken, and  $Q_a$  is catchability parameter.

- The stock sizes are assumed to follow:

$$N_{a,y} = N_{a-1,y-1} e^{-Z_{a-1,y-1}}$$

Notice that it does not define  $N$  in the first year and for the youngest age.

- So the model parameters are the undefined  $N$ 's,  $F_y$ ,  $F_a$ ,  $Q_a$ ,  $\sigma_c$ , and  $\sigma_s$

# A full parametric statistical model — implementation

- Data is in `fsa.RData` (let's study the content)
- An implementation in `★RTMB★` is in `basicfsa.R`
- The code is short, so let's identify the parts about  $N$ ,  $F$ ,  $C$ , and  $I$



## Exercise 3.1: Weighting of information sources

- Weighting to different data sources is a concept often used and often debated in fish stock assessment
- If the likelihood is correct, then the models should be self-weighting
- But imagine that we wanted to give double weight to our survey
- How could we achieve that by changing the code?
- Try two approaches:
  - Double the survey observations
  - Fix the variance to be half of the estimated variance for the survey observations
- Hint: Notice that the parameter is the log of the standard deviation. The variance is the standard deviation squared
- Take home: The weight is the inverse of the variance

## Exercise 3.2: Using robust distribution for observations

- The normal distribution is fairly sensitive to outliers

- Standard model uses pdf:  $\phi\left(\frac{x-\mu}{\sigma}\right) \frac{1}{\sigma}$

- Robust model will use pdf:

$$\left( (1-p)\phi\left(\frac{x-\mu}{\sigma}\right) + p\psi\left(\frac{x-\mu}{\sigma}\right) \right) \frac{1}{\sigma}$$

, where the  $p$  is an input (not estimated)

- Where  $\phi$  is pdf of  $N(0,1)$  and  $\psi$  is pdf of heavy-tailed distribution (e.g.  $t_3$ ).

- **Exercise:** In the simple assessment mode replace 5 (or 10) observations with huge outliers. Then run the assessment as is and compare to replacing with the model where the robust density is used for the observations.

