

# Building a simple model

## Assessment workshop

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# Outline

- 1 Modelling - data, parameters, predictions, obj fun, output
- 2 Sections - data, parameter, procedure, report
- 3 Data types - basic types, declarations
- 4 Exercise - estimate the mean

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# Statistical modelling

- **Read in data**

text file

- **Specify model**

relate parameters to data

- **Make predictions**

calculate fitted values to compare with observations

- **Specify objective function**

function to minimize, usually  $-\log L$

- **Minimize objective function**

some algorithm searches for the “best” parameter values

- **Write out results**

text file

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# ADMB sections

The corresponding sections in AD Model Builder are:

## **DATA\_SECTION**

read in data

## **PARAMETER\_SECTION**

declare parameters

## **PROCEDURE\_SECTION**

relate parameters to data, make predictions, specify the objective function  
[ADMB minimizes the objective function and writes out results]

## **(REPORT\_SECTION)**

write out verbose results (optional)

There are many other sections available in ADMB, and we will only use the first three sections in the first exercises

## hello.tpl

```
DATA_SECTION
  init_int n
  init_vector x(1,n)
  init_vector y(1,n)

PARAMETER_SECTION
  init_number a
  init_number b
  vector yfit(1,n)
  objective_function_value f

PROCEDURE_SECTION
  yfit=a+b*x;
  // Concentrated neglogL
  f=0.5*n*log(sum(square(y-yfit)));
```

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## Three basic data types

`int` integer

`double` floating point value

`dvariable` floating point value with derivative info

Estimated parameters are `dvariables`

Intermediate calculations and derived quantities are also `dvariables`

Each data type can form a vector, matrix, 3d array, ...

# Declaring objects

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Declaration	when inside	
	DATA_SECTION	PARAMETER_SECTION
<code>int</code>	integer	integer
<code>init_int</code>	integer from file	–
<code>number</code>	double	dvariable
<code>init_number</code>	double from file	dvariable to estimate
<code>init_bounded_number</code>	–	dvariable to estimate with bounds

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# Estimate the mean

Create a new model `mean.tpl` that estimates the mean of a vector

Use `hello.tpl` as a template

Implement two approaches, where the objective function is:

- 1 Residual sum of squares  
quick and dirty
- 2 Negative log likelihood  
evaluate uncertainty, estimate sigma

# Hints

## Residual sum of squares

$$\text{RSS} = \sum (y_i - \mu_i)^2$$

## Negative log likelihood

$$-\log L = [0.5n \log(2\pi)] + n \log \sigma + \frac{\text{RSS}}{2\sigma^2}$$