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# **E: Variance Estimation**

Fish 559; Day 2: 15h30-17h00

# Variance Estimation

- There are a variety of ways to compute standard errors and confidence intervals for model parameters and model outputs. This lecture considers two of these:
  - asymptotic methods; and
  - likelihood profile.

# Asymptotic Methods

# Asymptotic Standard Errors and Correlations-I

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- TMB automatically calculates asymptotic standard errors and parameter correlations:

$$\Sigma = \left[ -\frac{\partial^2 \ell n L}{\partial \theta_i \partial \theta_j} \right]^{-1}$$

- The parameter estimates and their asymptotic standard errors can be extracted using the `sdreport(model)` call.

# Asymptotic Standard Errors and Correlations-II

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- The default asymptotic variance-covariance calculations relate to the **parameters** of the model. However, we often want uncertainty estimates for **derived quantities**.
- The way to do this is via the **delta** method:

$$\text{Cov}[g(\underline{\theta}), h(\underline{\theta})] \approx \sum_i \sum_j \frac{\partial g(\underline{\theta})}{\partial \theta_i} \frac{\partial h(\underline{\theta})}{\partial \theta_j} \text{Cov}[\theta_i, \theta_j]$$

# Asymptotic Standard Errors and Correlations-III

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- To obtain asymptotic variances for derived parameters simply add

`ADREPORT(variable)`

to the CPP file.

- Note: the results from asymptotic theory are approximations and may be bad approximations.

# Asymptotic Standard Errors and Correlations-IV

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- Hints to speed things up:
  - Only calculate the derived variables when they are needed:

```
if (Do_sd==1)
  { code; }
```

- Look at the correlation matrix (which you can get by inverting the Hessian matrix; "hessian=T" in the MakeADFun call). If you see large ( $>0.80$ ) correlations among the parameters, it may be time to consider reparameterizing the model!

## Exercise-I

- Use the age-structured model from the last lecture and the dataset LECTE.DAT (a dataset with error) and compute:
  - the variance-covariance matrix for the parameters;
  - the standard error for the ratio:

$$B_{20} / B_1$$



## Exercise-II

- The correlation matrix:

33	Sel50	1.0000							
34	Sel95	<b>0.9663</b>	1.0000						
35	logFish	0.6845	0.6354	1.0000					
36	logFish	0.7331	0.6794	0.7865	1.0000				
37	logFish	0.7521	0.6950	0.7468	<b>0.8133</b>	1.0000			
38	logFish	0.7597	0.6992	0.7034	0.7712	<b>0.8081</b>	1.0000		
39	logFish	0.7597	0.6949	0.6708	0.7372	0.7792	<b>0.8175</b>	1.0000	
40	logFish	0.7643	0.6962	0.6444	0.7057	0.7470	0.7887	<b>0.8250</b>	1.0000

"Large" values  
in bold

- The standard error of  $B_{20} / B_1$ :  
0.038787



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Likelihood profile

# Likelihood profile-I

- A likelihood profile is:

$$\max_{\underline{\theta}} (\ell \mathbf{n} L(\mathbf{D} \mid \underline{\theta} \mid) \mid g(\underline{\theta}) = \alpha)$$

1. Fit the model to find the ML parameter estimates and find the corresponding negative log-likelihood.
2. Select a set of fixed values for the quantity of interest.
3. Minimize the negative log-likelihood fixing the quantity to each value in turn.
4. Plot the difference between the negative log-likelihood from step 1 and those from step 3

# Likelihood profile-II (Confidence intervals)

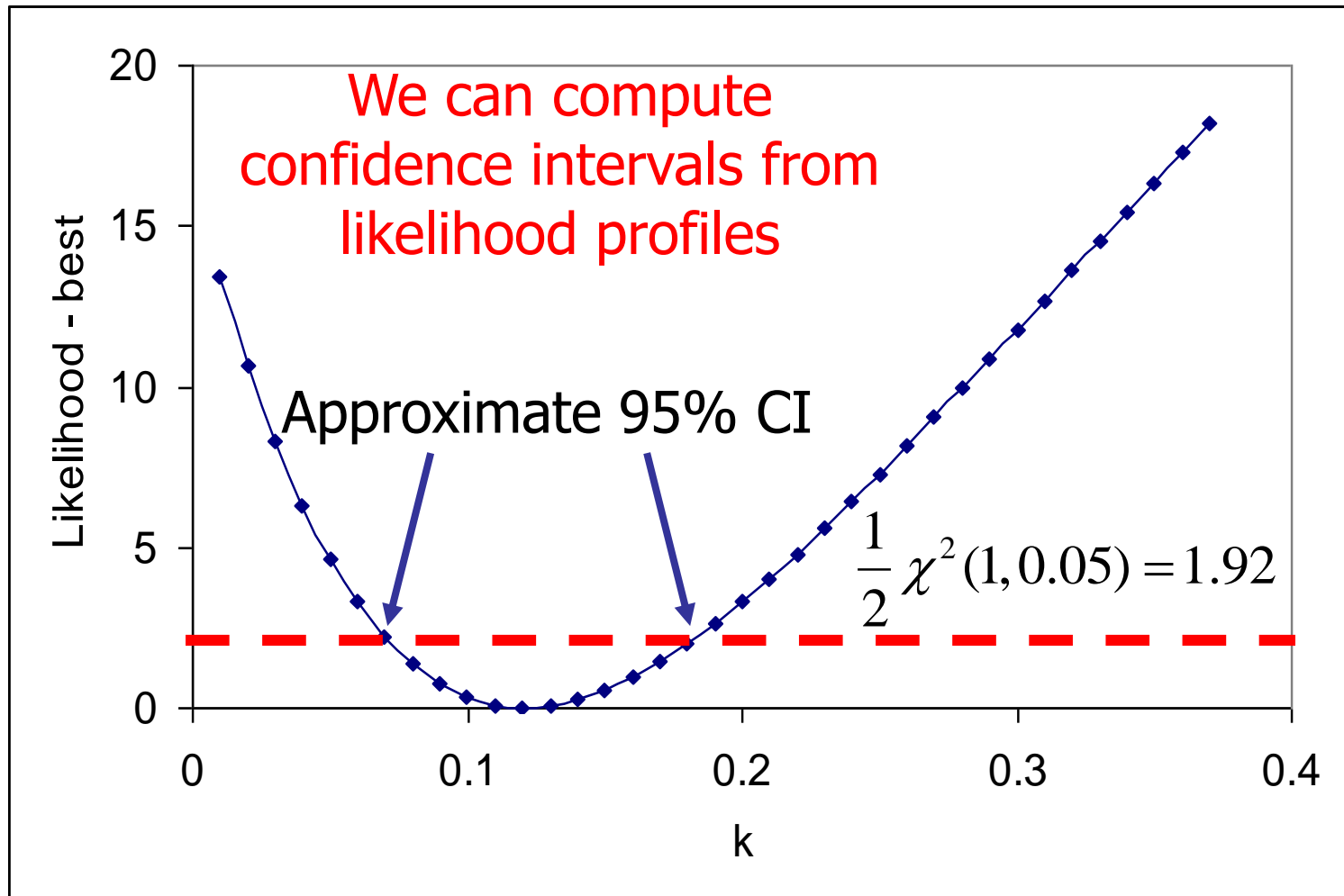
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- An  $100-x\%$  confidence interval for  $p$  parameters is determined by finding the values for the parameter(s) for which:

$$\ell n L - \ell n L_{MLE} = 0.5 \chi^2(p, x/100)$$

- $\ell n L_{MLE}$  is the negative log-likelihood corresponding to the maximum likelihood estimates.

# Likelihood profile-III



# Likelihood Profile-IV

## (How to create one)

- To obtain a likelihood profile for any parameter (lets say x). Call the function:

```
prof <- tmbprofile(model,"x")
```

- You can use the profile to determine confidence intervals for a parameter for which a profile has been determined using:

```
print confint(prof,level=0.8)
```



# Likelihood Profile-V

## (Using profiles)

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- Profiles can be used to evaluate the reliability of asymptotic variance estimates (check for non-symmetric profiles).
- Profiles can be used to assess the information content of the data relative to some parameter.
- Profiles can be used to assess data conflicts by plotting a the components of the likelihood.

# Likelihood Profile-VI

(Constructing likelihood profiles for derived parameters)

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- Add code that involves the computing the square of the difference between the quantity and a (read-in) target value.
- Run the program for many different target values.
- Construct a likelihood profile manually.