Using AD Model Builder and R together: getting started with the R2admb package

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1 Installation

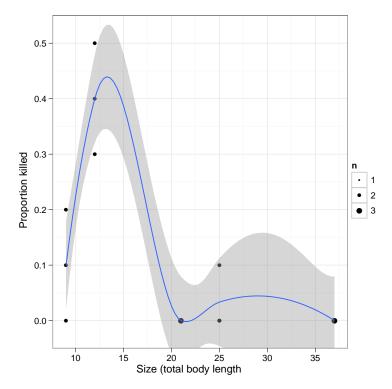
(Provide hints on where to find AD Model Builder ...)

- http://admb-project.org/
- http://admb-project.org/downloads
- http://code.google.com/p/admb-project/
- http://code.google.com/p/admb-project/downloads/list

Notes: (1) depending on your operating system etc., you may need to install a C++ compiler (in particular, the MacOS installation instructions will probably ask you to install gcc/g++ from the Xcode package); (2) you will need to have the scripts admb, adcomp, and adlink in the bin directory of your ADMB installation.

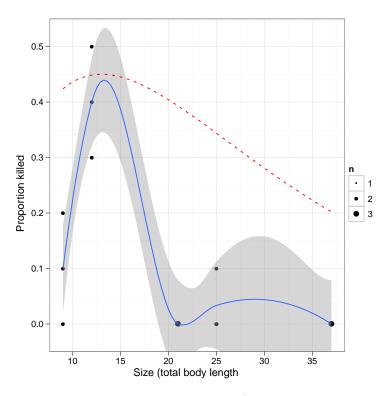
2 Basics

Initial view of the data (proportion killed):



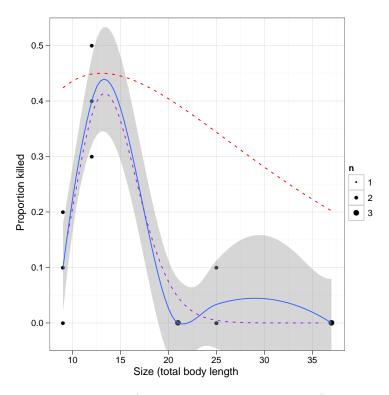
So if $p(\text{kill}) = c((S/d) \exp(1 - (S/d)))^g$ (peak occurs at S = d, peak height=c) then a reasonable set of first estimates would be c = 0.45, d = 13.

```
> startest <- stat_function(fun = function(x) { 0.45*((x/13)*exp(1-x/13)) }, 
+ ty=2, colour="red") > print(g1+startest)
```



(Not great, but

perhaps adequate as a starting point.)



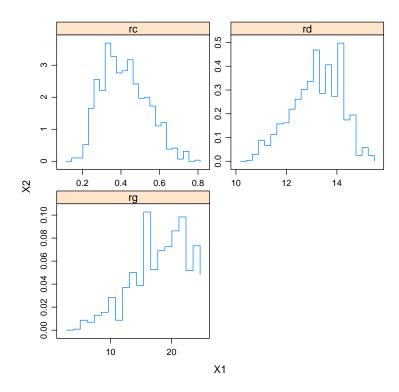
Here's the TPL (AD Model Builder definition) file:

```
DATA_SECTION
1
      init_int nobs
                                    // # of observations
2
      init_vector nexposed(1,nobs) // # exposed per trial
3
      init_vector TBL(1,nobs)
                                    // total body length
4
      init_vector Kill(1,nobs)
                                    // # killed per trial
5
    PARAMETER_SECTION
6
      init_bounded_number c(0,1) // baseline mort prob
7
      init_bounded_number d(0,50) // size scaling factor
8
      init_bounded_number g(-1,25) // size scaling power
9
      vector prob(1,nobs)
                              // per capita mort prob
10
      objective_function_value f
11
      sdreport_number rc;
                              // this & following for MCMC
12
      sdreport_number rd;
13
      sdreport_number rg;
14
    PROCEDURE_SECTION
15
                                   // set MCMC reporting
      rc = c; rd = d; rg = g;
16
      dvariable fpen=0.0;
                                   // penalty variable
17
```

```
// power-Ricker
18
      prob = c*pow(elem_prod(TBL/d,exp(1-TBL/d)),g);
19
      // penalties: constrain 0.001 <= prob <= 0.999
20
      prob = posfun(prob,0.001,fpen);
21
      f += 1000*fpen;
22
      prob = 1-posfun(1-prob,0.001,fpen);
23
      f += 1000*fpen;
24
      // binomial negative log-likelihood
25
      f -= sum( log_comb(nexposed,Kill)+
26
                elem_prod(Kill,log(prob))+
27
                elem_prod(nexposed-Kill,log(1-prob)));
28
   > setup_admb()
   > m1 <- do_admb("ReedfrogSizepred",
                    data=c(list(nobs=nrow(ReedfrogSizepred),
                      nexposed=rep(10,nrow(ReedfrogSizepred))),
                      ReedfrogSizepred),
                    param_list=list(c=0.45, d=13, g=1),
                    clean=TRUE)
   > unlink("reedfrogsizepred.tpl")
      Exercise the extractor methods:
   > m1
   Model file: reedfrogsizepred
   Negative log-likelihood: 12.8938
   Coefficients:
            С
                        d
    0.4138331 13.3508215 18.2479066
   > coef(m1)
             C.
                        d
    0.4138331 13.3508215 18.2479066
   > summary(m1)
   Model file: reedfrogsizepred
   Negative log-likelihood: 12.8938
   Coefficients:
```

```
Estimate Std. Error z value Pr(>|z|)
   0.4138
               0.1257
                        3.292 0.000996 ***
               0.8111 16.461 < 2e-16 ***
d 13.3508
g 18.2479
               6.0331
                        3.025 0.002489 **
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
> coef(summary(m1))
   Estimate Std. Error z value
                                      Pr(>|z|)
                0.12572 3.291705 9.958205e-04
c 0.4138331
d 13.3508215
                0.81107 16.460751 7.022349e-61
g 18.2479066
                6.03310 3.024632 2.489359e-03
> vcov(m1)
                     d
c 0.01580552 0.0578055 0.5043901
d 0.05780550 0.6578345 2.2464986
g 0.50439009 2.2464986 36.3982956
> logLik(m1)
[1] -12.8938
> deviance(m1)
[1] 25.7876
> AIC(m1)
[1] 31.7876
> m1MC <- do_admb("ReedfrogSizepred",</pre>
                data=c(list(nobs=nrow(ReedfrogSizepred),
                  nexposed=rep(10,nrow(ReedfrogSizepred))),
                  ReedfrogSizepred),
                  mcmc=TRUE,
                  param_list=list(c=0.45, d=13, g=1),
                  clean=TRUE)
> unlink("reedfrogsizepred.tpl")
```

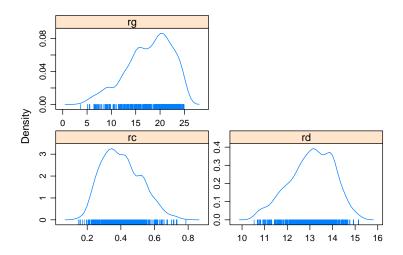
> print(plot(m1MC\$hist))



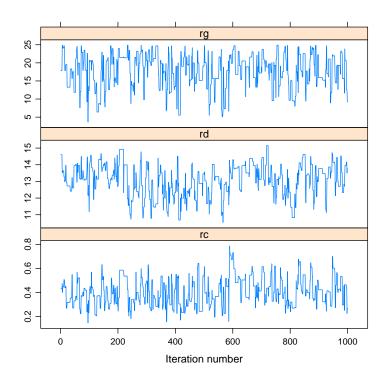
- > library(coda)
- > HPDinterval(as.mcmc(m1MC\$mcmc))

```
lower upper rc 0.2071513 0.6190827 rd 10.9495253 14.6161983 rg 8.7644415 24.8812550 attr(,"Probability") [1] 0.95
```

> print(densityplot(as.mcmc(m1MC\$mcmc)))



> print(xyplot(as.mcmc(m1MC\$mcmc)))



3 Random effects: gopher tortoise data