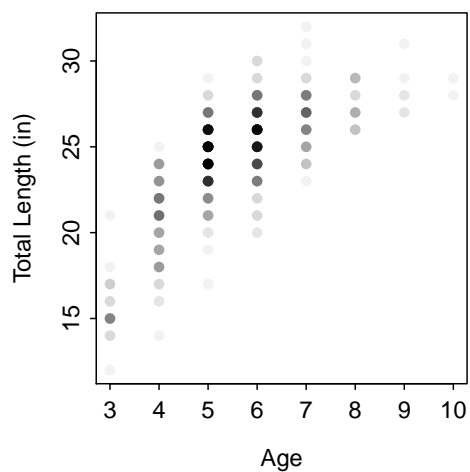


```
> library(FSA)          # Subset, fitPlot, vbModels, vbStart, vbFuns
> library(nlstools)     # overview
```

## 0.1 Brule River Rainbow Trout

```
> ## setwd("C:/aaaWork/Web/fishR/courses/Midwest2012/CourseMaterial/")
> d <- read.table("TroutBR.txt",header=TRUE)
> str(d)
'data.frame': 851 obs. of 3 variables:
 $ t1      : int  16 16 17 17 17 17 17 17 17 17 ...
 $ age     : int   4 4 2 3 3 3 3 3 3 4 ...
 $ species: Factor w/ 2 levels "Brown","Rainbow": 1 1 1 1 1 1 1 1 1 1 ...
> rbt <- Subset(d,species=="Rainbow")
> str(rbt)
'data.frame': 627 obs. of 3 variables:
 $ t1      : int  12 14 14 14 14 15 15 15 15 15 ...
 $ age     : int   3 3 3 3 4 3 3 3 3 3 ...
 $ species: Factor w/ 1 level "Rainbow": 1 1 1 1 1 1 1 1 1 1 ...
```

```
> clr <- rgb(0,0,0,0.05)
> plot(tl~age,data=rbt,col=clr,pch=16,xlab="Age",ylab="Total Length (in)")
```



## 0.2 Fit Traditional Model

```
> vbModels()
```

### FSA von Bertalanffy Parametrizations

Original: $E(L_t) = L_\infty - (L_\infty - L_0) e^{-Kt}$	Mooij: $E(L_t) = L_\infty - (L_\infty - L_0) e^{-\frac{\omega}{L_\infty} t}$
Typical: $E(L_t) = L_\infty \left(1 - e^{-K(t-t_0)}\right)$	Schnute: $E(L_t) = L_1 + (L_2 - L_1) \frac{1 - e^{-K(t-t_1)}}{1 - e^{-K(t_2-t_1)}}$
GalucciQuinn: $E(L_t) = \frac{\omega}{K} \left(1 - e^{-K(t-t_0)}\right)$	Francis: $E(L_t) = L_1 + (L_3 - L_1) \frac{1 - r^{2\frac{t-t_1}{t_3-t_1}}}{1 - r^2}$
	where $r = \frac{L_3 - L_2}{L_2 - L_1}$

```
> ( svb1 <- vbStarts(tl~age,data=rbt,type="typical") )
$Linf
[1] 28.67

$K
[1] 0.5242

$t0
[1] -1.429

> fit1 <- nls(tl~Linf*(1-exp(-K*(age-t0))),data=rbt,start=svb1)
> overview(fit1)
```

```
-----
Formula: tl ~ Linf * (1 - exp(-K * (age - t0)))

Parameters:
      Estimate Std. Error t value Pr(>|t|)
Linf  27.7118     0.2838   97.6    <2e-16
K      0.6324     0.0425   14.9    <2e-16
t0     1.7169     0.1016   16.9    <2e-16

Residual standard error: 1.78 on 624 degrees of freedom

Number of iterations to convergence: 5
Achieved convergence tolerance: 5.38e-08

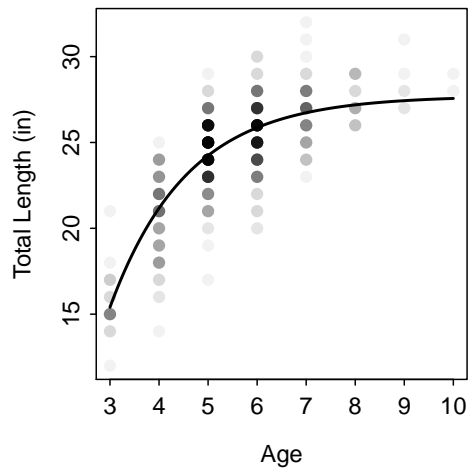
-----
Residual sum of squares: 1970

-----
Asymptotic confidence interval:
      2.5%    97.5%
Linf 27.154 28.2692
K     0.549 0.7159
t0    1.517 1.9164
```

```

-----
Correlation matrix:
      Linf      K      t0
Linf  1.0000 -0.9074 -0.7114
K     -0.9074  1.0000  0.9191
t0    -0.7114  0.9191  1.0000
> fitPlot(fit1,xlab="Age",ylab="Total Length (in)",main="",col.pt=clr,col.mdl="black")

```



### 0.3 Fit Galucci and Quinn Parameterization

```

> ( svb2 <- vbStarts(tl~age,data=rbt,type="GalucciQuinn") )
$omega
[1] 15.03

$K
[1] 0.5242

$t0
[1] -1.429
> ( vb2 <- vbFuns("GalucciQuinn",simple=TRUE) )
function(t,omega,K,t0) {
  (omega/K)*(1-exp(-K*(t-t0)))
}
<environment: 0x0529e800>
> fit2 <- nls(tl~vb2(age,omega,K,t0),data=rbt,start=svb2)
> overview(fit2)

-----
Formula: tl ~ vb2(age, omega, K, t0)

Parameters:
      Estimate Std. Error t value Pr(>|t|)
omega  17.5259     1.0172   17.2    <2e-16
K       0.6324     0.0425   14.9    <2e-16
t0      1.7169     0.1016   16.9    <2e-16

```

Residual standard error: 1.78 on 624 degrees of freedom

Number of iterations to convergence: 5

Achieved convergence tolerance: 9.19e-08

-----

Residual sum of squares: 1970

-----

Asymptotic confidence interval:

2.5% 97.5%

omega 15.528 19.5234

K 0.549 0.7159

t0 1.517 1.9164

-----

Correlation matrix:

omega K t0

omega 1.0000 0.9972 0.9382

K 0.9972 1.0000 0.9191

t0 0.9382 0.9191 1.0000

```
> fitPlot(fit2,xlab="Age",ylab="Total Length (in)",main="",col.pt=clr,col.mdl="black")
```

