Solving Differential Equations in R (book) - ODE examples

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Abstract

This vignette contains the R-examples of chapter 4 from the book:

Soetaert, K., Cash, J.R. and Mazzia, F. (2012). Solving Differential Equations in R. that will be published by Springer.

Chapter 4. Solving Ordinary Differential Equations in R.

Here the code is given without documentation. Of course, much more information about each problem can be found in the book.

Keywords: ordinary differential equations, initial value problems, examples, R.

1. A differential Equation Comprising One Variable

```
r <- 1 ; K <- 10
yini <- c(y = 2)
derivs <- function(t, y, parms)
   list(r * y * (1-y/K))
times <- seq(from = 0, to = 20, by = 0.2)
out   <- ode(y = yini, times = times, func = derivs,
          parms = NULL)
head(out, n = 3)
```

```
     time         y
[1,]  0.0 2.000000
[2,]  0.2 2.339222
[3,]  0.4 2.716436
```

```
#
yini <- c(y = 12)
out2 <- ode(y = yini, times = times, func = derivs,
          parms = NULL)

plot(out, out2, main = "logistic growth", lwd = 2)
```
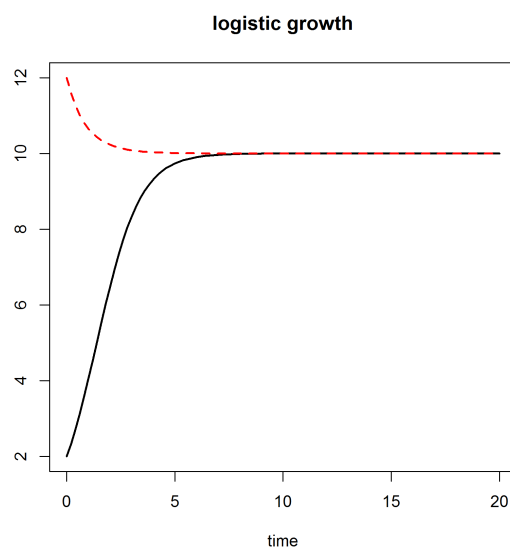
Figure 1: A simple initial value problem, solved twice with different initial conditions. See book for more information.
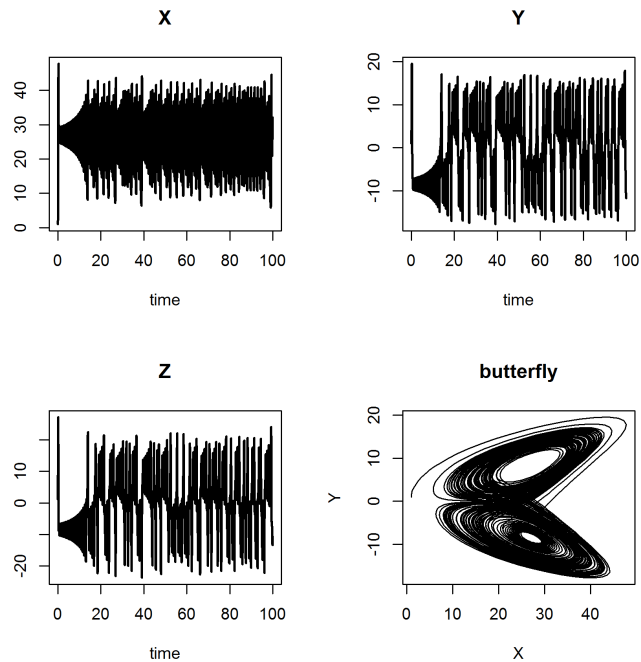
Figure 2: Solution of the lorenz equation. See book for more information.

2. Multiple Variables: the Lorenz Model

```
a      <- -8/3 ; b <- -10; c <- 28
yini <- c(X = 1, Y = 1, Z = 1)
Lorenz <- function (t, y, parms) {
 with(as.list(y), {
    dX <- a * X + Y * Z
    dY <- b * (Y - Z)
    dZ <- -X * Y + c * Y - Z
    list(c(dX, dY, dZ))
 })
}
times <- seq(from = 0, to = 100, by = 0.01)
out    <- ode(y = yini, times = times, func = Lorenz, parms = NULL)

plot(out, lwd = 2)
plot(out[,"X"], out[,"Y"], type = "l", xlab = "X",
  ylab = "Y", main = "butterfly")
```
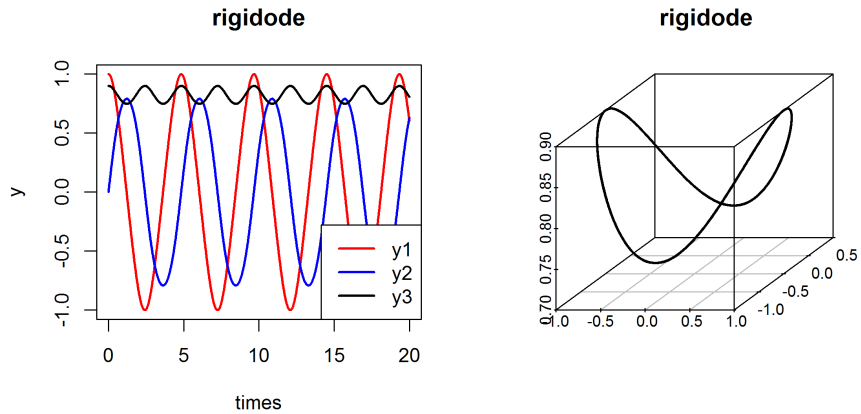
Figure 3: The rigid body equations. See book for more information.

3. Rigid Body Equations

```
yini  <- c(1, 0, 0.9)
rigidode <- function(t, y, parms) {
  dy1 <- -2  * y[2] * y[3]
  dy2 <- 1.25* y[1] * y[3]
  dy3 <- -0.5* y[1] * y[2]
  list(c(dy1, dy2, dy3))
}
times <- seq(from = 0, to = 20, by = 0.01)
out   <- ode (times = times, y = yini, func = rigidode,
              parms = NULL, method = rkMethod("rk45ck"))
head (out, n = 3)


      time           1           2           3
[1,]  0.00 1.0000000 0.00000000 0.9000000
[2,]  0.01 0.9998988 0.01124950 0.8999719
[3,]  0.02 0.9995951 0.02249603 0.8998875
```

4. Arenstorff Orbits

```
Arenstorff <- function(t, y, p) {
  D1 <- ((y[1] + mu1)^2 + y[2]^2)^(3/2)
  D2 <- ((y[1] - mu2)^2 + y[2]^2)^(3/2)
  dy1 <- y[3]
  dy2 <- y[4]
  dy3 <- y[1] + 2*y[4] - mu2*(y[1]+mu1)/D1 - mu1*(y[1]-mu2)/D2
  dy4 <- y[2] - 2*y[3] - mu2*y[2]/D1 - mu1*y[2]/D2
  return(list( c(dy1, dy2, dy3, dy4) ))
}
mu1   <- 0.012277471
mu2   <- 1 - mu1
yini  <- c(y1 = 0.994, y2 = 0,
          dy1 = 0, dy2 = -2.00158510637908252240537862224)
times <- seq(from = 0, to = 18, by = 0.01)
out <- ode(func = Arenstorff, y = yini, times = times,
           parms = 0, method = "ode45")
yini2 <- c(y1 = 0.994, y2 = 0,
           dy1 = 0, dy2 = -2.03173262955573368357302057924)
out2 <- ode(func = Arenstorff, y = yini2, times = times,
            parms = 0, method = "ode45")
yini3 <- c(y1 = 1.2, y2 = 0,
           dy1 = 0, dy2 = -1.049357510)
out3 <- ode(func = Arenstorff, y = yini3, times = times,
            parms = 0, method = "ode45")

plot(out, out2, out3, which = c("y1", "y2"),
     mfrow = c(2, 2), col = "black", lwd = 2)
plot(out[  ,c("y1", "y2")], type = "l", lwd = 2,
     xlab = "y1", ylab = "y2", main = "solutions 1,2")
lines(out2[ ,c("y1", "y2")], lwd = 2, lty = 2)
plot(out3[ ,c("y1", "y2")], type = "l", lwd = 2, lty = 3,
     xlab = "y1", ylab = "y2", main = "solution 3")
```
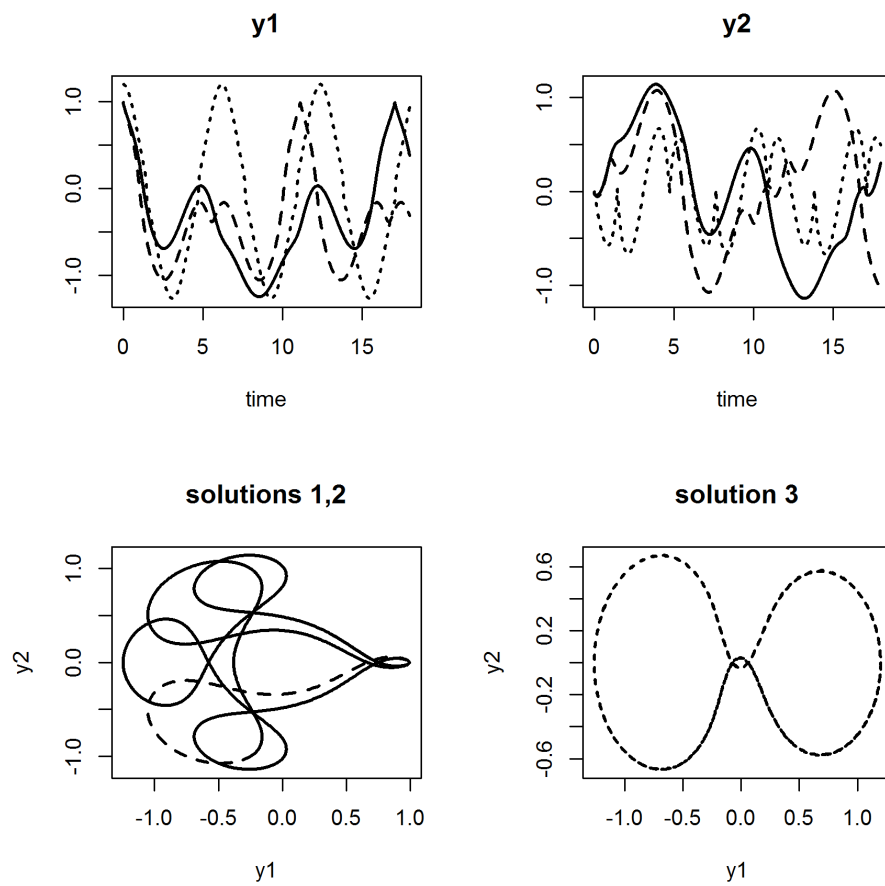
Figure 4: The Arenstorff problem. See book for more information.

5. Seven Moving Stars

```
pleiade <- function (t, Y, pars) {
   x <- Y[1:7]
   y <- Y[8:14]
   u <- Y[15:21]
   v <- Y[22:28]

   distx <- outer(x, x, FUN = function(x, y) x - y)
   disty <- outer(y, y, FUN = function(x, y) x - y)

   rij3 <- (distx^2 + disty^2)^(3/2)

   fx <- starMass * distx / rij3
   fy <- starMass * disty / rij3

   list(c(dx = u,
          dy = v,
          du = colSums(fx, na.rm = TRUE),
          dv = colSums(fy, na.rm = TRUE)))
}
starMass <- 1:7
yini<- c(x1= 3, x2= 3, x3=-1, x4=-3,     x5= 2, x6=-2,    x7= 2,
         y1= 3, y2=-3, y3= 2, y4= 0,     y5= 0, y6=-4,    y7= 4,
         u1= 0, u2= 0, u3= 0, u4= 0,     u5= 0, u6=1.75, u7=-1.5,
         v1= 0, v2= 0, v3= 0, v4=-1.25, v5= 1, v6= 0,    v7= 0)
print(system.time(
out <- ode(func = pleiade, parms = NULL, y = yini,
          method = "adams", times = seq(0, 3, 0.01))))


  user   system elapsed
   0.1      0.0     0.1


par(mfrow = c(3, 3))
for (i in 1:7) {
  plot(out[,i+1], out[,i+8], type = "l",
      main = paste("star ",i), xlab = "x", ylab = "y")
  points (yini[i], yini[i+7])
}
plot(out[, 2:8], out[, 9:15], type = "p", cex = 0.5,
     main = "ALL", xlab = "x", ylab = "y")
text(yini[1:7], yini[8:14], 1:7)
matplot(out[,"time"], out[, c("u1", "u7")], type = "l",
    lwd = 2, col = c("black", "grey"), lty = 1,
    xlab = "time", ylab = "velocity", main = "stars 1, 7")
abline(v = c(1.23, 1.68), lty = 2)
```
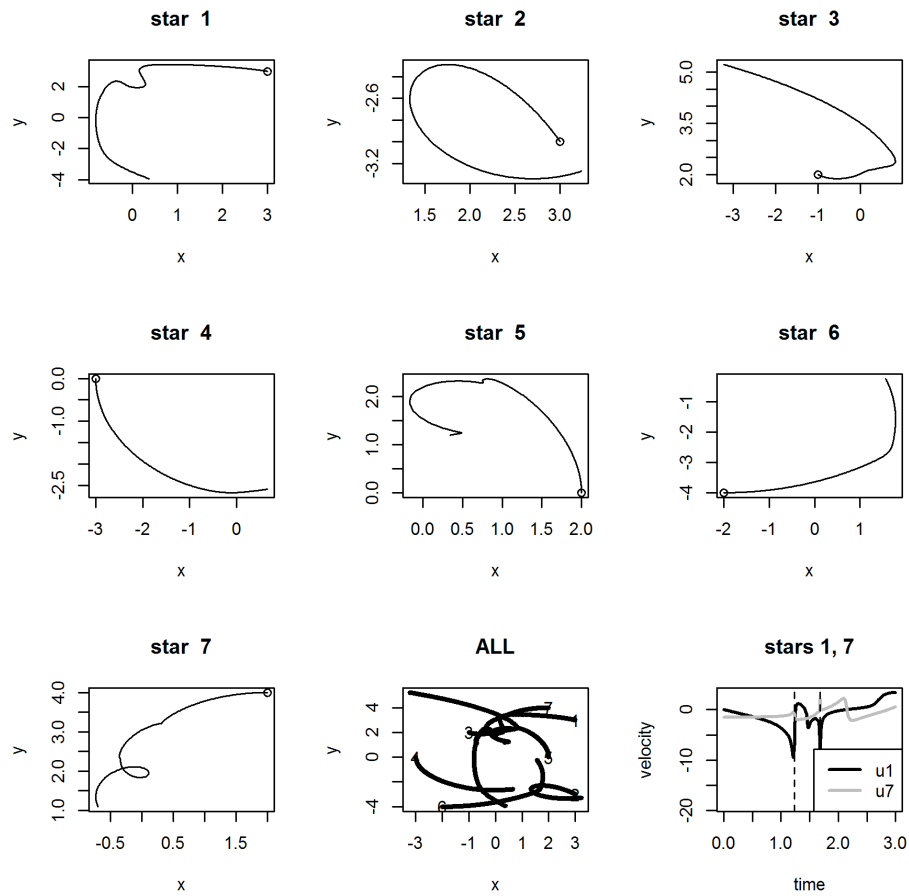
Figure 5: The pleiades problem. See book for more information.

```
legend("bottomright", col = c("black", "grey"), lwd = 2,
       legend = c("u1", "u7"))
```

5.1. A stiff Chemical Example

```
load(file = "Light.rda")
head(Light, n = 4)


        day     irrad
1 0.0000000   0.0000
2 0.3333333   0.0000
3 0.3541667 164.2443
4 0.3750000 204.7486


irradiance <- approxfun(Light)
irradiance(seq(from = 0, to = 1, by = 0.25))


[1]   0.0000   0.0000 698.8911 490.4644   0.0000


k3 <- 1e-11; k2 <- 1e10; k1a <- 1e-30
k1b <- 1; sigma <- 1e11
yini <- c(O = 0, NO = 1.3e8, NO2 = 5e11, O3 = 8e11)
chemistry <- function(t, y, parms) {
  with(as.list(y), {

    radiation <- irradiance(t)
    k1  <- k1a + k1b*radiation

    dO   <-  k1*NO2 - k2*O
    dNO  <-  k1*NO2         - k3*NO*O3 + sigma
    dNO2 <- -k1*NO2         + k3*NO*O3
    dO3  <-          k2*O - k3*NO*O3
    list(c(dO, dNO, dNO2, dO3), radiation = radiation)
  })
}
times <- seq(from = 8/24, to = 5, by = 0.01)
out <- ode(func = chemistry, parms = NULL, y = yini,
           times = times, method = "bdf")

plot(out, type = "l", lwd = 2 )
```
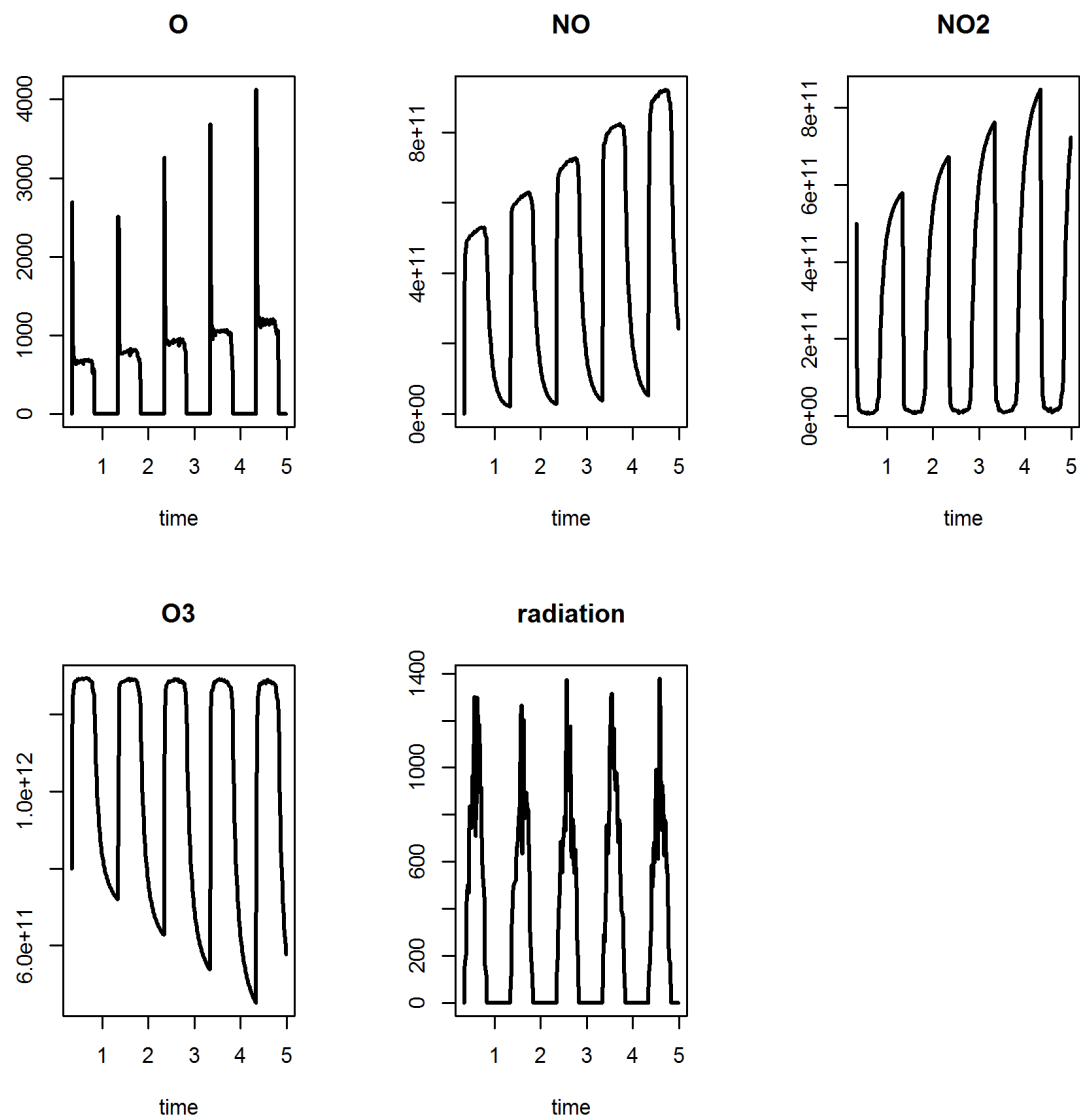
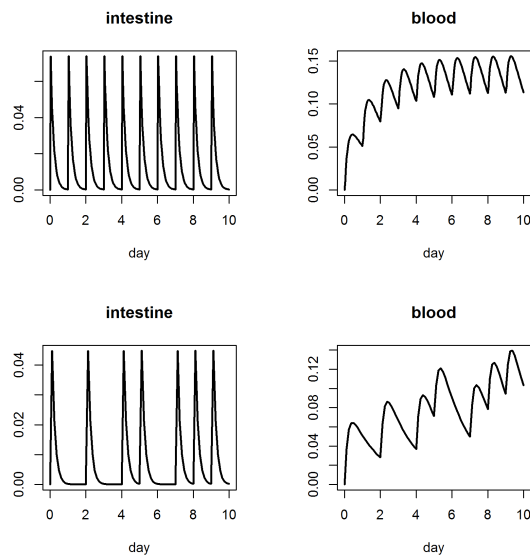Figure 6: The atmospheric chemistry model. See book for more information.

Figure 7: The 2-compartment pharmacokinetic model. See book for more information.

6. Pharmacokinetic Models

6.1. first example

```
a <- 6; b <- 0.6
yini <- c(intestine = 0, blood = 0)
pharmacokinetics <- function(t, y, p) {
  if ( (24*t) %% 24 <= 1)
    uptake <- 2
  else
    uptake <- 0
  dy1 <- - a* y[1] + uptake
  dy2 <-   a* y[1] - b *y[2]
  list(c(dy1, dy2))
}
times <- seq(from = 0, to = 10, by = 1/24)
out <- ode(func = pharmacokinetics, times = times,
           y = yini, parms = NULL)
times <- seq(0, 10,  by = 3/24)
out2 <- ode(func = pharmacokinetics, times = times,
            y = yini, parms = NULL, method = "impAdams")
```
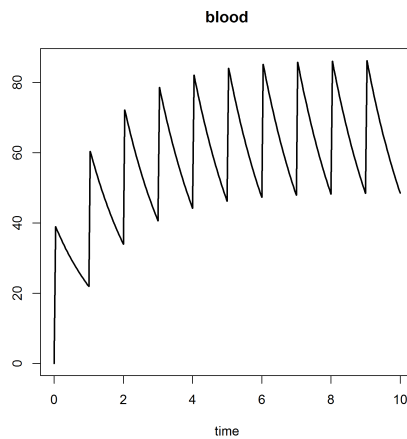
Figure 8: The 1-compartment pharmacokinetic model. See book for more information.

6.2. second example

```
b    <- 0.6
yini <- c(blood = 0)
pharmaco2 <- function(t, blood, p) {
  dblood <-  - b * blood
  list(dblood)
}
injectevents <- data.frame(var = "blood",
                           time =  0:20,
                           value = 40,
                           method = "add")
head(injectevents, n=3)


    var time value method
1 blood    0    40    add
2 blood    1    40    add
3 blood    2    40    add


 times <- seq(from = 0, to = 10, by = 1/24)
 out2 <- ode(func = pharmaco2, times = times, y = yini,
            parms = NULL, method = "impAdams",
            events = list(data = injectevents))


 plot(out2, lwd = 2)
```
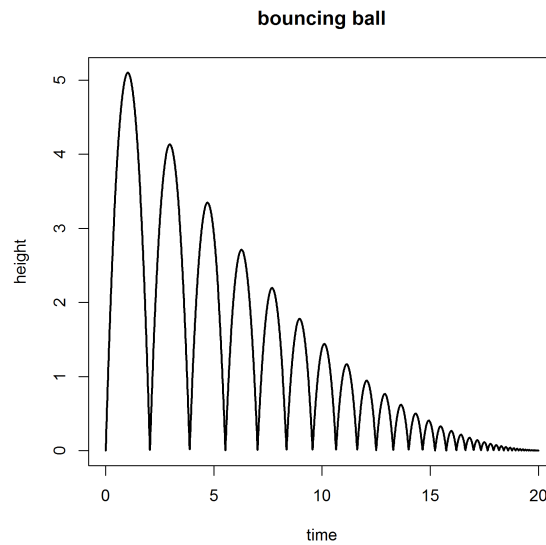
bouncing ball

Figure 9: The bouncing ball model. See book for more information.

7. A Bouncing Ball

```
yini  <- c(height = 0, velocity = 10)
ball <- function(t, y, parms) {
  dy1 <- y[2]
  dy2 <- -9.8

  list(c(dy1, dy2))
}
rootfunc <- function(t, y, parms) y[1]
eventfunc <- function(t, y, parms) {
 y[1] <- 0
 y[2] <- -0.9*y[2]
 return(y)
}
times <- seq(from = 0, to = 20, by = 0.01)
out <- ode(times = times, y = yini, func = ball,
           parms = NULL, rootfun = rootfunc,
           events = list(func = eventfunc, root = TRUE))

plot(out, which = "height", lwd = 2,
     main = "bouncing ball", ylab = "height")
```
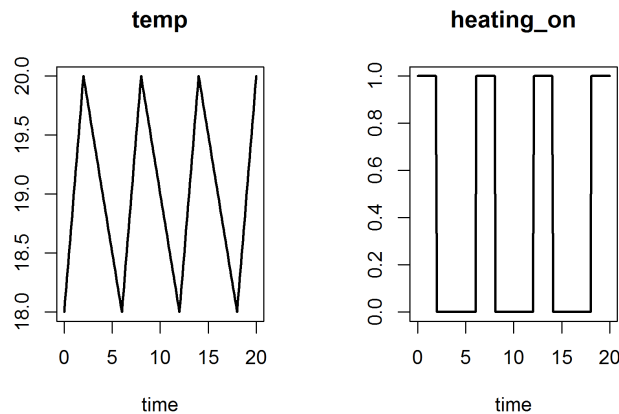
Figure 10: The temperature model. See book for more information.

8. Temperature in a Climate-controlled Room

```
yini  <- c(temp = 18, heating_on = 1)
temp <- function(t, y, parms) {
  dy1 <- ifelse(y[2] == 1, 1.0, -0.5)
  dy2 <- 0
  list(c(dy1, dy2))
}
rootfunc <- function(t, y, parms) c(y[1]-18, y[1]-20)
eventfunc <- function(t, y, parms) {
 y[1] <- y[1]
 y[2] <- ! y[2]
 return(y)
}
times <- seq(from = 0, to = 20, by = 0.1)
out <- lsode(times = times, y = yini, func = temp,
          parms = NULL, rootfun = rootfunc,
          events = list(func = eventfunc, root = TRUE))

attributes(out)$troot

[1]  2  6  6  6  8 12 14 18

plot(out, lwd = 2)
```

9. Method Selection

```
yini  <- c(y = 2, dy = 0)
Vdpol <- function(t, y, mu)
   list(c(y[2],
          mu * (1 - y[1]^2) * y[2] - y[1]))
times <- seq(from = 0, to = 30, by = 0.01)
nonstiff <- ode(func = Vdpol, parms = 1,    y = yini,
                times = times)
diagnostics(nonstiff)


--------------------
lsoda return code
--------------------

  return code (idid) =  2
  Integration was successful.


--------------------
INTEGER values
--------------------

  1 The return code : 2
  2 The number of steps taken for the problem so far: 3004
  3 The number of function evaluations for the problem so far: 6009
  5 The method order last used (successfully): 7
  6 The order of the method to be attempted on the next step: 7
  7 If return flag =-4,-5: the largest component in error vector 0
  8 The length of the real work array actually required: 52
  9 The length of the integer work array actually required: 22
 14 The number of Jacobian evaluations and LU decompositions so far: 0
 15 The method indicator for the last succesful step,
          1=adams (nonstiff), 2= bdf (stiff): 1
 16 The current method indicator to be attempted on the next step,
          1=adams (nonstiff), 2= bdf (stiff): 1


--------------------
RSTATE values
--------------------

  1 The step size in t last used (successfully): 0.01
  2 The step size to be attempted on the next step: 0.01
  3 The current value of the independent variable which the solver has reached: 30.00947
  4 Tolerance scale factor > 1.0 computed when requesting too much accuracy: 0
  5 The value of t at the time of the last method switch, if any: 0
```

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