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concordance:Advection-Diffusion.tex:Advection-Diffusion.Rnw:1 71 1 1 3 1
5 6 1 1 2 1 3 1 1 1 3 12 1 1 20 23 0 1 3 10 1 1 3 2 0 4 1 1 2 5 0 1 3 4 1 1 2 9 0 1
7 4 1 4 0 1 3 1 1

Modeling Advection and Diffusions

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The movement of compounds in the environment is driven by two processes, advection and diffusion. Of course, these processes occur in three dimensions, but for this class we'll begin with one dimensional processes before getting to more complicated examples. Typeset using the Sweave function in R and L^AT_EX using a Tufte:1983, Tufte:1997 and style.

1 Session Outcomes

1. Describe Advection Mathematically
2. Analyze 1-dimensional movement using advection equations
3. Describe Diffusions mathematically
4. Analyze 1-dimensional movement using fick's law.
5. Two dimensional analysis of advection

2 Material Transport

3 R as a Calculator

2.4. Steady-state solution of 2-D PDEs

Function steady.2D efficiently

nds the steady-state of 2-dimensional problems. Karline Soetaert 13 In the following model $= D_x \cdot \nabla^2 C + p(x,y)$ a substance C is consumed at a quadratic rate ($r C^2$), while dispersing in X- and Y-direction. At certain positions (x,y) the substance is produced (rate p). The model is solved on a square (100*100) grid. There are zero- u_x boundary conditions at the 4 boundaries. The term $D_x \cdot \nabla^2 C$ where $\text{Flux} = -D_x \cdot \nabla C$ i.e. it is the negative of the u_x gradient, where the u_x is due to diffusion. In the numerical approximation for the u_x , the concentration gradient is approximated as the subtraction of two matrices, with the columns or rows shifted (e.g. $\text{Conc}[2:n,] - \text{Conc}[1:(n-1),]$). The u_x gradient is then also approximated by subtracting entire matrices (e.g. $\text{Flux}[2:(n+1),] - \text{Flux}[1:(n),]$). This is very fast. The zero- u_x at the boundaries is imposed by binding a column or row with 0-s.

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i diffusion2D j- function(t,conc,par) + Conc j- matrix(nr=n,nc=n,data=conc)
vector to 2-D matrix + dConc j- -r*Conc*Conc consumption + BND j- rep(1,n)
boundary concentration + + constant production in certain cells + dConc[ii]j-
dConc[ii] + p + + diffusion in X-direction; boundaries=imposed concentration +
+ Flux j- -Dx * rbind(rep(0,n),(Conc[2:n,]-Conc[1:(n-1),]),rep(0,n))/dx + dConc
j- dConc - (Flux[2:(n+1),]-Flux[1:n,])/dx + + diffusion in Y-direction + Flux j-
-Dy * cbind(rep(0,n),(Conc[,2:n]-Conc[,1:(n-1)]),rep(0,n))/dy + dConc j- dConc
- (Flux[,2:(n+1)]-Flux[,1:n])/dy + + return(list(as.vector(dConc))) + i

```

After specifying the values of the parameters, 10 cells on the 2-D grid where there will be substance produced are randomly selected (ii).

14 Package rootSolve : roots, gradients and steady-states in R 0.0 0.2 0.4 0.6 0.8 1.0 0.0 0.2 0.4 0.6 0.8 1.0 2-D diffusion+production x y Figure 5: Steady-state solution of the nonlinear 2-Dimensional model i parameters i dy j- dx j- 1 grid size i Dy j- Dx j- 1.5 diffusion coeff, X- and Y-direction i r j- 0.01 2-nd-order consumption rate (/time) i p j- 20 0-th order production rate (CONC/t) i n j- 100 i 10 random cells where substance is produced at rate p i ii j- trunc(cbind(runif(10)*n+1,runif(10)*n+1)) i The steady-state is found using function steady.2D. It takes as arguments a.o. the dimensionality of the problem (dimens) and lrw=1000000, the length of the work array needed by the solver. If this value is set too small, the solver will return with the size needed. It takes about 0.5 second to solve this 10000 state variable model.

```

i Conc0 j- matrix(nr=n,nc=n,10.) i print(system.time( i not working yet...
i i ST3 j- steady.2D(Conc0,func=diffusion2D,parms=NULL,pos=TRUE,dimens=c(n,n),
lrw=1000000,atol=1e-10,rtol=1e-10,ctol=1e-1))) i user system elapsed 1.044
0.032 1.076 The S3 image method is used to generate the steady-state plot.
i image(ST3,main="2-D diffusion+production", xlab="x", ylab="y")

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