

(3.31) Scilab.

*Scilab*'s built-in function *pinv* returns the pseudoinverse of a matrix. For example (see Example (3.27) in Lecture 27):

```
--> A = [1,1;1,1;1,1]

A = [3x2 double]

    1.    1.
    1.    1.
    1.    1.

--> pinv(A)

ans = [2x3 double]

    0.1666667    0.1666667    0.1666667
    0.1666667    0.1666667    0.1666667
```

The built-in function *backslash* (`\`) is used to solve a system of linear equations. Specifically, if  $A \in \mathbb{R}^{r \times c}$  is a matrix and  $b \in \mathbb{R}^r$  is a column, after the assignment:

$$x = A \backslash b$$

we have:<sup>1</sup>

- if  $r = c$  and  $c_1(A) \leq \frac{1}{10u}$   
then:  
 $x$  is an approximation of the solution of the system  $Ax = b$  computed with a procedure equivalent to the application of the EGPP, SA, SI procedures;
- if  $(r = c \text{ and } c_1(A) > \frac{1}{10u})$  or  $r > c$   
then:  
 $x$  is an approximation of an element of  $S_{\text{MQ}}(A,b)$  - usually *not* the minimum norm one - computed with a procedure that uses a QR factorization of  $A$ .

For example (see Example (3.25) of Lecture 27):

```
--> A = [1,1;1,1]

A = [2x2 double]

    1.    1.
    1.    1.
```

---

<sup>1</sup> Let  $N$  be a norm in  $\mathbb{R}^n$ . In *Scilab*, when  $A \in \mathbb{R}^{n \times n}$  is a *non-invertible* matrix, we set:  $c_N(A) = +\infty$ .

Lecture 28 - 2

```
--> b = [1;0]
```

```
b = [2x1 double]
```

```
1.  
0.
```

```
--> x = A\b
```

```
x = [2x1 double]
```

```
0.5000000  
0.
```

```
--> y = pinv(A) * b
```

```
y = [2x1 double]
```

```
0.2500000  
0.2500000
```

The built-in function `qr` returns an approximation to a QR factorization of a matrix, even if it is not square. Specifically, if  $A \in \mathbb{R}^{r \times c}$  where  $r > c$ , after the assignment:

$$[Q,R] = qr(A)$$

the matrix  $Q \in \mathbb{R}^{r \times r}$  is an approximation of the orthogonal matrix calculated with the Householder method (Remark (2.21) of Lecture 17) applied to  $A$  and  $R \in \mathbb{R}^{r \times c}$  is a matrix all whose elements under the main diagonal are equal to zero (i.e it is an upper triangular matrix). For example:

```
--> A = [1,0;1,1;1,1]
```

```
A = [3x2 double]
```

```
1.    0.  
1.    1.  
1.    1.
```

```
--> [Q,R] = qr(A)
```

```
Q = [3x3 double]
```

```
-0.5773503    0.8164966   -8.756D-17  
-0.5773503   -0.4082483   -0.7071068  
-0.5773503   -0.4082483    0.7071068
```

```
R = [3x2 double]
```

```
-1.7320508   -1.1547005  
0.           -0.8164966  
0.           0.
```

To obtain an approximation of a QR factorization of A as defined in Definition (3.28) of Lecture 27 one can use the qr function as follows:

```
--> [U,T] = qr(A,'e')
```

```
U = [3x2 double]
```

```
-0.5773503    0.8164966
-0.5773503   -0.4082483
-0.5773503   -0.4082483
```

```
T = [2x2 double]
```

```
-1.7320508   -1.1547005
 0.           -0.8164966
```

The factors U,T are obtained from the factors Q,R by eliminating, respectively, the third column of Q and the third row of R. In fact, if we perform the product QR column by column, we observe that, if  $q_1, q_2, q_3$  are the columns of Q and  $r_{ij}$  are the elements of R, we have:

$$QR = (r_{11} q_1 + 0 q_2 + 0 q_3, r_{12} q_1 + r_{22} q_2 + 0 q_3) = UT$$

#### (4) NUMERICAL METHODS FOR ORDINARY DIFFERENTIAL EQUATIONS

##### (4.01) Example (damped harmonic oscillator).

The motions of a damped harmonic oscillator are described by the *differential equation*:

$$(E) \quad x''(t) + a x'(t) + b x(t) = 0$$

where the unknown is the real-valued *function*  $x(t)$ . This is a *second-order* differential equation (linear, with constant coefficients, homogeneous). A *solution* to the equation is an *at least two-times differentiable* real-valued function  $y(t)$  that satisfies the equality  $y''(t) + a y'(t) + b y(t) = 0$  for *all*  $t$  in  $\mathbb{R}$ . The differential equation determines *all* possible motions of the oscillator (equation (E) has *infinitely many* solutions). Each of the motions is identified by the *initial conditions*:

$$(CI) \quad x(t_0) = x_0, \quad x'(t_0) = v_0$$

The *Cauchy Problem* is the problem to *find the solutions to the differential equation that satisfy the initial conditions*.

The second-order differential equation (E) is *equivalent* to a *system of two first-order equations*. Equivalence in this case means that: if  $y(t)$  is a solution to equation (E), then, set:

$$u_1(t) = y(t), \quad u_2(t) = y'(t)$$

Then:

$$u_1'(t) = u_2(t) \quad , \quad u_2'(t) = -a u_2(t) - b u_1(t)$$

Hence the column  $(u_1(t), u_2(t))^t$  is a solution of the system:

$$(S) \quad x_1'(t) = x_2(t) \quad , \quad x_2'(t) = -a x_2(t) - b x_1(t)$$

On the contrary: let  $(y_1(t), y_2(t))^t$  be a solution of system (S) and set  $y(t) = y_1(t)$ . We have:  $y'(t) = y_1'(t) = y_2(t)$  and  $y''(t) = y_1''(t) = y_2'(t) = -a y_2(t) - b y_1(t)$  that is:

$$y''(t) + a y'(t) + b y(t) = 0$$

hence  $y(t)$  is a solution of equation (E). Moreover,  $y(t)$  is a solution of the Cauchy Problem:

$$x''(t) + a x'(t) + b x(t) = 0 \quad ; \quad x(t_0) = x_0 \quad , \quad x'(t_0) = v_0$$

if and only if  $(y(t), y'(t))^t$  is a solution of the Cauchy Problem:

$$x_1'(t) = x_2(t) \quad , \quad x_2'(t) = -a x_2(t) - b x_1(t) \quad ; \quad x_1(t_0) = x_0 \quad , \quad x_2(t_0) = v_0$$

(4.02) Remark.

The procedures we will describe are designed to approximate the solution of the Cauchy Problem:

$$(\S) \quad x'(t) = F(t, x(t)) \quad , \quad x(t_0) = x_0$$

for  $t$  in a *bounded* interval  $[t_0, t_f]$ . The *unknown* of the problem is the function  $x(t)$  with values in  $\mathbb{R}^n$ ; the *data* are: *the function*  $F$  defined in  $\mathbb{R} \times \mathbb{R}^n$  with values in  $\mathbb{R}^n$ , *the instants*  $t_0$  and  $t_f > t_0$  and *the column*  $x_0$  in  $\mathbb{R}^n$ .

The previous statement presupposes that the solution to problem (§) *exists and is unique*. A further hypothesis will also be necessary to *describe the numerical procedures*.

(4.03) Hypothesis (existence and uniqueness).

For every  $\underline{t}$  in  $\mathbb{R}$  and  $\underline{x}$  in  $\mathbb{R}^n$  there is only one solution to the differential equation:

$$x'(t) = F(t, x(t))$$

which verifies the initial condition:

$$x(\underline{t}) = \underline{x}$$

We will denote such a solution by:  $y(t; \underline{x}, \underline{t})$ .

(4.04) Definition (numerical method).

A *numerical method for approximating the solution of the Cauchy Problem* (§) on  $[t_0, t_f]$  is a procedure that constructs, based on the value of a user-controlled parameter  $E$ , real numbers  $t(0) = t_0, \dots, t(N)$  in  $[t_0, t_f]$ , columns  $x(0) = x_0, \dots, x(N)$  in  $\mathbb{R}^n$  and, for  $k = 0, \dots, N$ , suggests to use  $x(k)$  as an approximation of  $y(t(k); x_0, t_0)$ .

The numbers  $t(0), \dots, t(N)$  are called *integration instants* and, for  $k = 0, \dots, N-1$ , the number  $h(k) = t(k+1) - t(k)$  is called the *integration step at instant  $t(k)$* .

A *Scilab* implementation of a numerical method has the following structure:

```
function [T,X] = NumericalMethod(x0,t0,t_f,F,E)

k = 0; t(0) = t0; x(0) = x0;
while t(k) < t_f,
    CHOOSE h(k) based on the value of E;
    COMPUTE x(k+1);
    t(k+1) = t(k) + h(k);
    k = k+1;
end;

endfunction
```

The output variables are, respectively, the row  $T$  and the matrix  $X$  such that:

$$T = (t(0), \dots, t(N)) \quad , \quad X = (x(0), \dots, x(N))$$

A numerical method is specified by the procedures of *choosing*  $h(k)$  and *computing*  $x(k+1)$ .