

# Linear Algebra, Optimization and Solving Ordinary Differential Equations Using Scilab

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Linear Algebra

Optimization

Solving Ordinary Differential Equations

# System of Linear Equations

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- ▶ Number of Equations may or may not be equal to number of unknowns.

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 $x=A \setminus b$
- ▶ Or use command  
`[x,ker]=linsolve(A,b)`
- ▶ To find Kernel(nullspace) of a system separately use  
`ker=kernel(A)`



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- ▶ `[Q,R]=qr(A)` //QR-Decomposition

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$$x_1 + 2x_2 + 3x_3 = 2$$

- ▶ Use `linsolve`

- ▶ Try this for previously obtained solution

```
A*x A*(x+ker) //In this case kernel is a line
```



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Solve  $x^2 + 3x + 2 = 0$

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def f('y=f(x)', 'y=x^2+3*x+2')
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where  $x_0$  is initial guess.
- ▶ One can also define function  $f : R^n \rightarrow R^n$  and solve it for zero locations.

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- ▶ `[f,xopt]=optim(list(NDcost,myf),x0)`

## For Example

- ▶ Minimize:

$$f(x, y) = (x + y)^2 + x + y + 2$$

- ▶ Gradient of the Function  $f$

$$\nabla f = (2(x + y) + 1 \quad 2(x + y) + 1)$$

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- ▶ If  $f : R^n \rightarrow R^m$ , then  $g$  is Jacobian a  $m \times n$  Matrix.



# Hessian

- ▶  $[g, H] = \text{derivative}(f, x)$  is the calling sequence
- ▶ for a function  $f : R^n \rightarrow R$   
g is the gradient of  $f$   
and H is Hessian matrix of  $f$

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 $t$  is the time instants at which solution is needed.  
'myode' is external function which defines the differential equation.
- ▶ Higher Order Equations must be made into first order equations of form  $\dot{x} = Ax + Bu$ .

## Examples

- ▶ Solve the differential equation
$$\frac{d^2\theta}{dt^2} + \frac{g}{L}\sin(\theta) = 0$$
- ▶ Take  $g = 9.8 \text{ m/s}^2$   $L = 1 \text{ m}$
- ▶ Check the plot of solution against time using `plot2d(t,x(1,:))` and `plot2d(t,x(2,:))`
- ▶ Also obtain the phase plane plot using `plot2d(x(1,:),x(2,:))`

# Thank You!

▶ [www.scilab.org](http://www.scilab.org)

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- ▶ [www.scilab.org](http://www.scilab.org)
- ▶ "Modeling And Simulation in Scilab/Scicos", by S.L.Campbell, J. Chancelier, R. Nikoukah.