

AEM 2012: Dynamics Numerical Integration

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- Kinematics - the geometry of motion.
- Dynamics - the science of motion; how forces interact with bodies causing deflections and/or motion.

What kinematics and dynamics is really about is *deriving the differential equations that describe the motion of a body.*

Equation of Motion of a Damped Pendulum



For example, consider a pendulum with a rod of negligible mass with a particle of mass m at its end. In class we derived

$$\underline{v}^{yw/a} = \underline{\mathcal{F}}_b^T \begin{bmatrix} -\ell\dot{\theta} \\ 0 \\ 0 \end{bmatrix}, \quad \underline{a}^{yw/a/a} = \underline{\mathcal{F}}_b^T \begin{bmatrix} -\ell\ddot{\theta} \\ -\ell\dot{\theta}^2 \\ 0 \end{bmatrix}.$$

From our free-body diagram we determined

$$\underline{f}^{yg} = \underline{\mathcal{F}}_a^T \begin{bmatrix} 0 \\ -mg \\ 0 \end{bmatrix} = \underline{\mathcal{F}}_b^T \begin{bmatrix} mg \sin(\theta) \\ mg \cos(\theta) \\ 0 \end{bmatrix}, \quad \underline{f}^{yp} = \underline{\mathcal{F}}_b^T \begin{bmatrix} 0 \\ f_{b2}^{yp} \\ 0 \end{bmatrix}.$$

We will now also consider a drag force acting on the particle modeled as a linear viscous damping force:

$$\underline{f}^{yd} = -c \underline{v}^{yw/a} = \underline{\mathcal{F}}_b^T \begin{bmatrix} c\ell\dot{\theta} \\ 0 \\ 0 \end{bmatrix}.$$

Using Newton's Second Law, we get

$$\begin{aligned} \vec{f}^{yg} + \vec{f}^{yp} + \vec{f}^{yd} &= m \vec{a}^{yw/a/a} \\ \underline{\mathcal{F}}_b^T \begin{bmatrix} mg \sin(\theta) \\ mg \cos(\theta) \\ 0 \end{bmatrix} + \underline{\mathcal{F}}_b^T \begin{bmatrix} 0 \\ f_{b2}^{yp} \\ 0 \end{bmatrix} + \underline{\mathcal{F}}_b^T \begin{bmatrix} c\ell\dot{\theta} \\ 0 \\ 0 \end{bmatrix} &= \underline{\mathcal{F}}_b^T \begin{bmatrix} -m\ell\ddot{\theta} \\ -m\ell\dot{\theta}^2 \\ 0 \end{bmatrix} \end{aligned}$$

Removing the vectrices and rearranging the equations gives

$$m\ell\ddot{\theta} + c\ell\dot{\theta} + mg \sin(\theta) = 0, \quad (1)$$

and

$$f_{b2}^{yp} = -m \left(\ell\dot{\theta}^2 + g \cos(\theta) \right). \quad (2)$$

Equation (1) is our differential equation of motion, while equation (2) is our expression for the reaction force.



Notice that

$$m\ell\ddot{\theta} + c\ell\dot{\theta} + mg\sin(\theta) = 0 \quad (3)$$

is a *nonlinear* ordinary differential equation (ODE)!

How do we find a solution to this nonlinear ODE?

Often nonlinear ODEs cannot be solved analytically; as such, we resort to **numerical integration methods** to numerically find solutions given a set of initial conditions (ICs).

The first thing we need to do is put our ODE into *first-order* form:

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), t)$$

where $\mathbf{x} = [x_1 \ x_2 \ \dots \ x_{n-1} \ x_n]^\top$ are the n states of the system.

For example, to write the equation of motion of the pendulum in first-order form let $x_1 = \theta$ and $x_2 = \dot{x}_1 = \dot{\theta}$ so that

$$m\ell\dot{x}_2 + c\ell x_2 + mg \sin(x_1) = 0,$$

or alternatively,

$$\dot{x}_2 = \frac{1}{m\ell} (-c\ell x_2 - mg \sin(x_1)).$$

We can now write out the entire ODE in first-order form as

$$\underbrace{\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix}}_{\dot{\mathbf{x}}(t)} = \underbrace{\begin{bmatrix} x_2(t) \\ \frac{1}{m\ell} (-c\ell x_2(t) - mg \sin(x_1(t))) \end{bmatrix}}_{\mathbf{f}(\mathbf{x}(t), t)} \quad (4)$$

Notice that all the “dots” are on the left-hand side, while each of the states, x_1 and x_2 , are on the right-hand side.

- Note that we have not “changed” the system; we’ve simply taken a second-order scalar ODE (i.e., Eq. (3)) and written it as a first-order matrix ODE (i.e., Eq. (4)).
- The form of $\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), t)$ is not unique. For instance

$$\underbrace{\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix}}_{\dot{\mathbf{x}}(t)} = \underbrace{\begin{bmatrix} \frac{1}{m\ell} (-c\ell x_1(t) - mg \sin(x_2(t))) \\ x_1(t) \end{bmatrix}}_{\mathbf{f}(\mathbf{x}(t), t)}$$

is also correct; here $x_2 = \theta$ and $\dot{x}_2 = x_1 = \dot{\theta}$.

- Denote

$$\mathbf{x}_k = \mathbf{x}(t_k)$$

and $h = t_{k+1} - t_k$ as the step-size.

- Approximate $\dot{\mathbf{x}}$ using a simple forward difference:

$$\dot{\mathbf{x}}(t) \doteq \frac{\mathbf{x}(t_{k+1}) - \mathbf{x}(t_k)}{t_{k+1} - t_k} = \frac{\mathbf{x}_{k+1} - \mathbf{x}_k}{h}$$

- Use the above approximation in $\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), t)$ yielding

$$\begin{aligned}\frac{\mathbf{x}_{k+1} - \mathbf{x}_k}{h} &= \mathbf{f}(\mathbf{x}_k, t_k), \\ \mathbf{x}_{k+1} &= \mathbf{x}_k + h\mathbf{f}(\mathbf{x}_k, t_k).\end{aligned}$$

- Given $\mathbf{x}_1 = \mathbf{x}(0)$ at $t_1 = 0$, “march forward in time”.
- Issues: poor accuracy (global error on the order of h); numerical stability.



- RK4 integration is based on a higher-order Taylor series approximation of $\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), t)$.

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$$\mathbf{x}_{k+1} = \mathbf{x}_k + \frac{1}{6} (\mathbf{k}_1 + 2\mathbf{k}_2 + 2\mathbf{k}_3 + \mathbf{k}_4)$$

where

$$\mathbf{k}_1 = h\mathbf{f}(\mathbf{x}_k, t_k)$$

$$\mathbf{k}_2 = h\mathbf{f}\left(\mathbf{x}_k + \frac{1}{2}\mathbf{k}_1, t_k + \frac{1}{2}h\right)$$

$$\mathbf{k}_3 = h\mathbf{f}\left(\mathbf{x}_k + \frac{1}{2}\mathbf{k}_2, t_k + \frac{1}{2}h\right)$$

$$\mathbf{k}_4 = h\mathbf{f}(\mathbf{x}_k + \mathbf{k}_3, t_k + h)$$

- Given $\mathbf{x}_1 = \mathbf{x}(0)$ at $t_1 = 0$, “march forward in time”.
- Global error on the order of h^4 ; local error on the order of h^5 .

- `ode45` in `matlab` uses the fourth order Runge-Kutta-Fehlberg method, which also gives an error estimate at each step.
- It is variable step size integrator; the solver is able to choose a step size which meets the error tolerance specified.
- Absolute tolerance (`AbsTol`) and Relative tolerance (`RelTol`).
 - At each time-step, the solver estimates the local error, $e_{k,i}$, in the i^{th} state, $x_{k,i}$.
 - At that time step the simulation converges if

$$|e_{k,i}| \leq \max(\text{RelTol} \cdot |x_{k,i}|, \text{AbsTol}).$$

- When state values are large, `RelTol` dictates convergence. As the states approach zero, `AbsTol` dictates convergence.
- In general, values for `AbsTol` and `RelTol` are problem dependent.

* <http://www.mathworks.com/help/simbio/ref/absolutetolerance.html>.



These slides are based on slides by Prof. James Richard Forbes (McGill).