



Lecture Notes

Week 7

Ordinary differential equations (ODEs)

Given $\frac{dy}{dt} = f(t, y(t))$ with $y(t=t_1) = y_1$.

Initial Value Problem: Can we estimate $y(t_2)$ at time $t_2 = t_1 + h$, $t_3 = t_1 + 2h$, ... ?

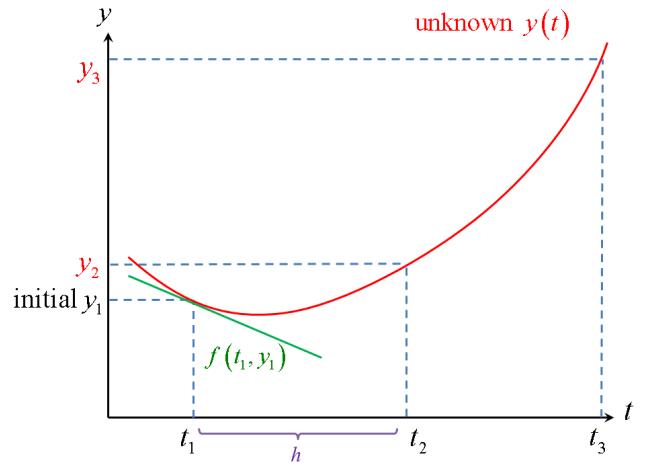
Solution: Fundamental Theorem of Calculus

$$y(t_2) = y(t_1) + \int_{t_1}^{t_2} f(t, y(t)) dt$$

f is known *But $y(t)$ is not known yet!*

Find approximate solutions for $y(t)$.

The form $\frac{dy}{dt} = f(t, y(t))$ is general



1. For a 1-D problem, e.g. $\frac{dy}{dt} = t^2 - y$, define $f=@(t, y) t.^2 - y;$
2. For a 2-D or higher dimensional problem, e.g. $\frac{dY_1}{dt} = Y_1 - Y_2 + 2$ and $\frac{dY_2}{dt} = -Y_1 + Y_2 + 4t$,

rewrite the system in vector form $\frac{d}{dt} \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \begin{bmatrix} Y_1 - Y_2 + 2 \\ -Y_1 + Y_2 + 4t \end{bmatrix}$, or $\frac{d\mathbf{Y}}{dt} = \mathbf{F}(t, \mathbf{Y}(t))$. Then define

$F=@(t, Y) [Y(1) - Y(2) + 2 ; -Y(1) + Y(2) + 4*t];$

3. For high-order ODEs, e.g. $\frac{d^2y}{dt^2} + 8\frac{dy}{dt} + 2y = 0$, define $Y_1 = y$ and $Y_2 = \frac{dY_1}{dt}$, then the high-order ODEs is transformed into a 2-D problem: $\frac{dY_1}{dt} = Y_2$ and $\frac{dY_2}{dt} = -2Y_1 - 8Y_2$. Then we can rewrite them in vector form $\frac{d}{dt} \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \begin{bmatrix} Y_2 \\ -2Y_1 - 8Y_2 \end{bmatrix}$ and define

$F=@(t, Y) [Y(2) ; -8*Y(1) - 2*Y(2)];$

Approximate Solutions

Euler's method

For small time step $t_2 = t_1 + h$, left sum applies to the integral: $\int_{t_1}^{t_2} f(t, \mathbf{y}(t)) dt \approx f(t_1, y_1)h$

$$y_2 \approx \underbrace{y_1 + f(t_1, y_1)h}_{\tilde{y}_2}$$

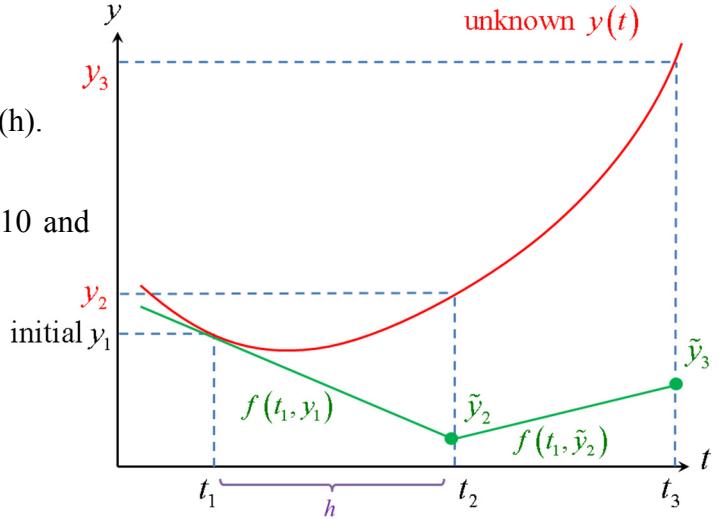
Global errors associated with Euler's method is $O(h)$.

e.g. 1-D problems: $\frac{dy}{dt} = t^2 - y$ with $y(t=0.5) = 10$ and

a time step size $h = 0.8$.

In Matlab,

```
f=@(t,y) t.^2 - y;
h=0.8;
y(1)=10;
t(1)=0.5;
y(2)=y(1)+h*f(t(1),y(1));
t(2)=t(1)+h;
```



Now we know \tilde{y}_2 , we can then further estimate other times $t_3 = t_1 + 2h$, $t_4 = t_1 + 3h$, ...

$$\begin{aligned} y_3 &\approx \underbrace{\tilde{y}_2 + f(t_2, \tilde{y}_2)h}_{\tilde{y}_3} \\ y_4 &\approx \underbrace{\tilde{y}_3 + f(t_2, \tilde{y}_3)h}_{\tilde{y}_4} \\ &\vdots \end{aligned}$$

In Matlab, this can easily be done by replacing the above one-step calculation with a **for** loop:

```
for j=1:n
    y(j+1)=y(j)+h*f(t(j),y(j));
    t(j+1)=t(j)+h;
end
```

For our example, if we are to estimate $y(t=3.7)$, then we calculate the first 5 (i.e. **n=4**) values:

```
t=[ 0.50 1.30 2.10 2.90 3.70]
y=[10.00 2.20 1.79 3.89 7.51]
```

e.g. Higher-dimensional problems: $\frac{dY_1}{dt} = Y_1 - Y_2 + 2$ and $\frac{dY_2}{dt} = -Y_1 + Y_2 + 4t$
with $Y_1(t=0.4) = -1$, $Y_2(t=0.4) = 0$ and $h = 0.2$.

```
F=@(t,Y) [Y(1)-Y(2)+2 ; -Y(1)+Y(2)+4*t] ;
h=0.2;
Y(:,1)=[-1;0];
t(1)=0.4;
for j=1:10
    Y(:,j+1)=Y(:,j)+h*F(t(j),Y(:,j));
    t(j+1)=t(j)+h;
end
Y1=Y(1,:);
Y2=Y(2,:);
```

After the above calculation, $\mathbf{Y1(end)} = Y_1(t=2.4)$ and $\mathbf{Y2(end)} = Y_2(t=2.4)$.

Implicit Euler's method

This time applies right sum to the integral:

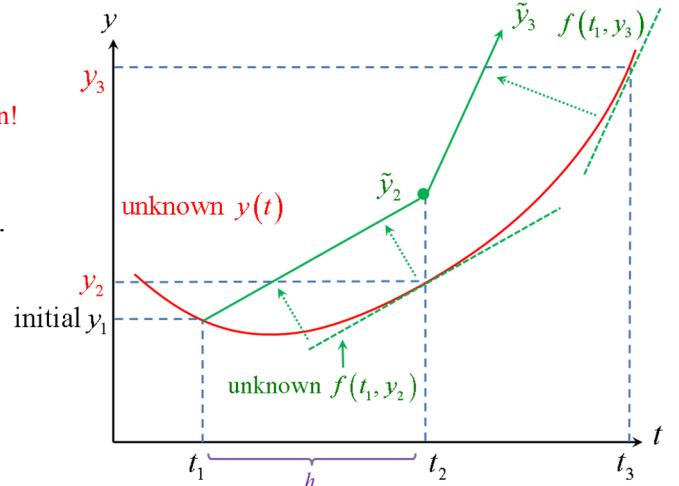
$$y_2 \approx y_1 + f(t_2, y_2)h$$

But y_2 is not known!

Answer: y_2 can be estimated as \tilde{y}_2 using **fzero**

Errors associated with Implicit Euler's method is $O(h)$.

```
f=@(t,y) t.^2 - y;
h=0.8;
y(1)=10;
t(1)=0.5;
t(2)=t(1)+h;
y(2)=fzero(@(z) z-y(1)-h*f(t(2),z),y(1));
```



** For multi-dimensional problems, use **fsolve**.

Prof. Brunton's way:

```
y(2)=(y(1)+t(2).^2)./(1+h);
```

For other times, $\tilde{y}_3 = \tilde{y}_2 + f(t_2, \tilde{y}_3)h$, $\tilde{y}_4 = \tilde{y}_3 + f(t_3, \tilde{y}_4)h$, ...

```
for j=1:n
    t(j+1)=t(j)+h;
    y(j+1)=fzero(@(z) z-y(j)-h*f(t(j+1),z),y(j));
end
```

In our example,

```
t =[ 0.50 1.30 2.10 2.90 3.70]
y =[10.00 6.31 5.46 6.77 9.85]
```

Implicit Trapezoidal Rule (optional; similar but not identical to **ode23tb**)

Try averaging the above two estimates.

This is the Trapezoidal Rule:

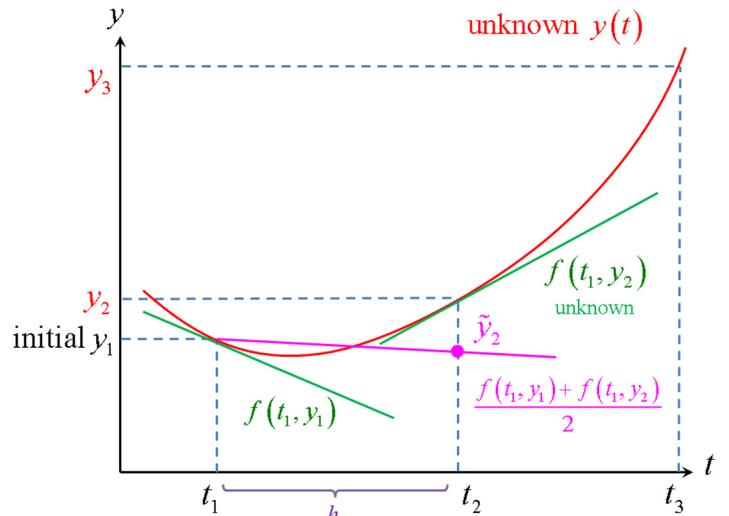
$$\begin{aligned} y_2 &= y_1 + \int_{t_1}^{t_2} f(t, y(t)) dt \\ &\approx y_1 + \frac{h}{2} [f(t_1, y_1) + f(t_2, y_2)] \end{aligned}$$

But still implicit. Global errors $\sim O(h^2)$.

Either:

1. Use **fzero** or **fsolve** to get y_2 .
2. Use Euler's method to estimate the y_2 on the right-hand side:

$$\begin{aligned} \tilde{y}_2 &= y_1 + hf(t_1, y_1) \\ y_2 &\approx y_1 + \frac{h}{2} [f(t_1, y_1) + f(t_2, \tilde{y}_2)] \end{aligned}$$



This is the Heun's method and is an example of the corrector-predictor methods.

4th-order Runge-Kutta (similar but not identical to **ode45**)

This time we try Simpson's rule:

$$\begin{aligned} y_2 &= y_1 + \int_{t_1}^{t_2} f(t, y(t)) dt \\ &\approx y_1 + \frac{(h/2)}{3} [f(t_1, y_1) + 4f(t_{1.5}, y_{1.5}) + f(t_2, y_2)] \end{aligned}$$

where $t_{1.5} = t_1 + \frac{h}{2}$. The mid-point $(t_{1.5}, y_{1.5})$ is constructed so that the width of the interval for the

Simpson rule is $h/2$. Apply the predictor-corrector method to estimate $y_{1.5}$ and y_2 on the right:

1. Define $k_1 = f(t_1, y_1)$. This is the slope at t_1 .
2. Use the forward Euler method to estimate $y_{1.5}$:

$$\tilde{y}_{1.5} = y_1 + \int_{t_1}^{t_1 + \frac{1}{2}h} f(t, y(t)) dt \approx y_1 + \frac{h}{2} f(t_1, y_1) = y_1 + \frac{h}{2} k_1$$

$\tilde{y}_{1.5}$ is the first estimate of $y_{1.5}$. Define $k_2 = f(t_{1.5}, \tilde{y}_{1.5})$.

3. Having $\tilde{y}_{1.5}$, now use backward Euler method to estimate $y_{1.5}$ again:

$$\hat{y}_{1.5} \approx y_1 + \frac{h}{2} f(t_{1.5}, \tilde{y}_{1.5}) = y_1 + \frac{h}{2} k_2$$

$\hat{y}_{1.5}$ is the 2nd estimate of $y_{1.5}$. Define $k_3 = f(t_{1.5}, \hat{y}_{1.5})$.

4. y_2 on the left hand side can now be estimated using k_3 :

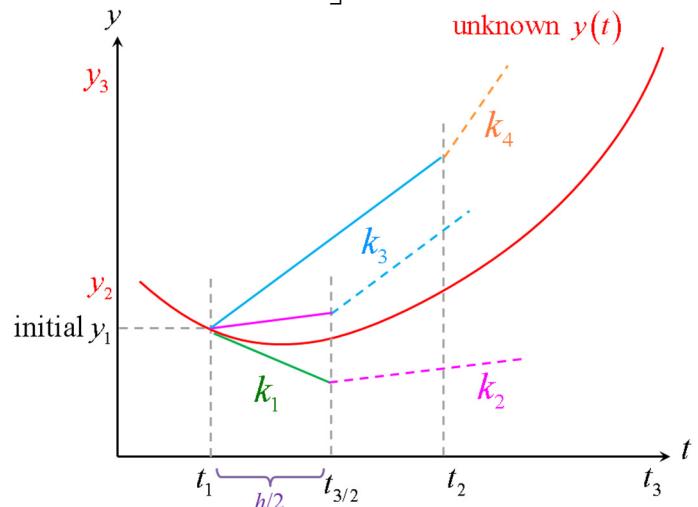
$$\hat{y}_2 \approx y_1 + h f(t_{1.5}, \hat{y}_{1.5}) = y_1 + h k_3.$$

\hat{y}_2 is the predictor for y_2 .

5. Now y_2 on the left hand side is given by substituting $\tilde{y}_{1.5}$, $\hat{y}_{1.5}$ and \hat{y}_2 to the right. To get a better estimate, we use both $\tilde{y}_{1.5}$ and $\hat{y}_{1.5}$ in $f(t_{1.5}, y_{1.5})$ but we average them out:

$$\begin{aligned} y_2 &\approx y_1 + \frac{(h/2)}{3} \left[f(t_1, y_1) + 4 \frac{f(t_{1.5}, \tilde{y}_{1.5}) + f(t_{1.5}, \hat{y}_{1.5})}{2} + f(t_2, \hat{y}_2) \right] \\ &= y_1 + \frac{h}{6} (k_1 + 2k_2 + 2k_3 + k_4) \end{aligned}$$

Global errors of estimate are the same as in Simpson's Rule $\sim O(h^4)$.



```

f=@(t,y) t.^2 - y;
h=0.8;
y(1)=10;
t(1)=0.5;

for j=1:n
    k1=f(t(j),y(j)); % Slope at t(j)
    k2=f(t(j)+h/2,y(j)+h/2*k1); % 1st Estimated slope at t(j)+h/2
    k3=f(t(j)+h/2,y(j)+h/2*k2); % 2nd Estimated slope at t(j)+h/2
    k4=f(t(j)+h,y(j)+h*k3); % Estimated slope at t(j)+h
    y(j+1)=y(j)+h/6*(k1+2*k2+2*k3+k4);
    t(j+1)=t(j)+h;
end

```

(Rule: Each k is used to estimate the next k .)

In our example,

```

t = [ 0.50 1.30 2.10 2.90 3.70]
y = [10.00 5.05 4.01 5.43 8.67]

```

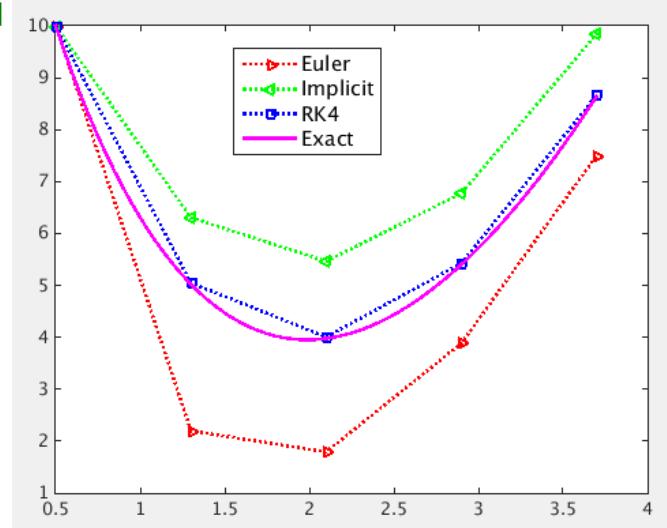
Exact solution in our 1-D example

$$y(t) = 8.75e^{0.5-t} + t^2 - 2t + 2$$

```

syms y(t)
yexact = dsolve(diff(y) == t.^2 - y, y(0.5) == 10)
yout = double(subs(yexact, [.5:.8:3.7]))
>> yout =[10.00 5.02 3.98 5.40 8.65]

```



Matlab's Functions

The above theories assume a fixed time step h . Matlab employs advanced algorithms that choose h automatically to minimize errors.

ode45 — Dormand-Prince Method

4/5th-order Runge-Kutta pair

Compare the 5-th and 4-th order integration to estimate the error term and then choose h to reduce the error term to desired accuracy.

```
[tout,yout]=ode45(f,[t1 tmax],y1)
```

The output **tout** has points spanning between **t1** and **tmax** and the points are determined automatically. **yout** are the values of **y** at **tout**.

```
[tout,yout]=ode45(f,[t1 t2 ... tmax],y1)
```

Returns **tout=[t1 t2 ... tmax]**. **yout** are the values of **y** at **t1, t2, ... tmax**.

In our example,

```
f=@(t,y) t.^2 - y ;
y1 = 10;
t1 = 0.5;
h = 0.8;
tmax = 3.7;
[tout,yout]=ode45(f,[t1:h:tmax],y1);

>> tout = [ 0.50 1.30 2.10 2.90 3.70]
>> yout = [ 10.00 5.02 3.98 5.40 8.65]
```

(Compare this with `[tout,yout]=ode45(f,[t1 tmax],y1);`)

Ode23tb — Implicit Trapezoidal-Backward Difference Method

Implicit 2/3th-order Trapezoidal pair

```
[t,y]=ode23tb(f,[t1 t2],y1)
```

The output **t** is determined automatically

```
[t,y]=ode23tb(f,[t1 t2 ... tmax],y1)
```

The output **tout=[t1 t2 ... tmax]**.