

Numerical Analysis.

HW 17 and 18 before you leave
HW 19 assigned.

Today: Multistep Methods. - use more than one past value of the numerical soln.

$$\underline{\text{Ex}} \quad y_{n+1} = y_n + \frac{h}{2} [3f(x_n, y_n) - f(x_{n-1}, y_{n-1})]$$

This is explicit, stable, and second order...
truncation error $\mathcal{O}(h^2)$

Lets see how these methods are derived using
Numerical Interpolation and Numerical Integration.

Start w/ our ODE $y' = f(x, y)$

$$\text{integrate over } [x_n, x_{n+1}] \quad y(n+1) = y(n) + \int_{x_n}^{x_{n+1}} f(x, y) dx$$

approximate $f(x, y)$ using
a polynomial
interpolant.
interpolate @ x_n, x_{n-1}

LINEAR INTERPOLATION or TRAPEZOID RULE.

$$P_1 = \frac{(x_n - x)f(x_{n+1}, y_{n+1})}{(x_n - x_{n+1})} + \frac{(x_{n+1} - x)f(x_n, y_n)}{(x_{n+1} - x_n)}$$

$$= \frac{1}{h} [(x_n - x)f(x_{n+1}, y_{n+1}) + (x - x_{n+1})f(x_n, y_n)]$$

$$\text{so } y(n+1) = y(n) + \int_{x_n}^{x_{n+1}} \frac{1}{h} [(x_n - x)f(x_{n+1}, y_{n+1}) + (x - x_{n+1})f(x_n, y_n)] dx$$

$$\text{Evaluate the integral.} \quad = y(n) + \frac{1}{h} \left[\int_{x_n}^{x_{n+1}} \left(-\frac{1}{2}(x_n - x)^2 f(x_{n+1}, y_{n+1}) + \frac{1}{2}(x - x_{n+1})^2 f(x_n, y_n) \right) dx \right]$$

$$= y(n) + \frac{1}{h} \left[-\frac{1}{2}h^2 f(x_{n+1}, y_{n+1}) + \frac{1}{2}(4h^2) f(x_n, y_n) - \frac{1}{2}h^2 f(x_n, y_n) \right]$$

$$= y(n) + \left[\frac{3h}{2} f(x_n, y_n) - \frac{h}{2} f(x_{n+1}, y_{n+1}) \right]$$

this gives our method:

$$y(n+1) = y(n) + \frac{h}{2} [3f(x_n, y_n) - f(x_{n+1}, y_{n+1})]$$

What is the error associated with this method?

$$\text{Error} = \int_0^h f(x, y) dx - \frac{h}{2} [3f(0, y(0)) - f(-h, y(-h))]$$

for simplicity write $f(0) \Rightarrow f(0, y(0))$
 $f(-h) \Rightarrow f(-h, y(-h))$

$$= \int_0^h f(x, y) dx - \frac{h}{2} [3f(0) - f(-h)]$$

Expand $f(x, y)$ in a Taylor Polynomial about $x=0$

$$f(x, y) = f(0) + xf'(0) + \frac{x^2}{2} f''(\xi) \quad \text{REMAINDER}$$

where $-h < \xi < h$

$$\text{Then } f(-h) = f(0) - hf'(0) + \frac{h^2}{2} f''(\xi)$$

Plug These in ...

$$\text{Error} = \int_0^h [f(0) + xf'(0) + \frac{x^2}{2} f''(\xi)] dx - \frac{h}{2} [3f(0) - [f(0) - hf'(0) + \frac{h^2}{2} f''(\xi)]]$$

Integrate ...

$$hf(0) + \frac{h^2}{2} f'(0) + \frac{h^3}{6} f''(\xi) - \frac{h}{2} [2f(0) + hf'(0) - \frac{h^2}{2} f''(\xi)]$$

$$\text{so Error} = \frac{h^3}{6} f''(\xi) + \frac{h^3}{4} f''(\xi) = \frac{5h^3}{12} f''(\xi)$$

we truncate the approximation.

Our error is $O(h^3)$

so our method is second order

NOTES ON ERROR AND APPROXIMATING ERROR ...

• If we know The REAL soln Then $\text{ERROR}_h = \text{REAL} - \text{APPROX}_h$
 $= y(x_n) - y_h(x_n)$

• Often in real life we do not know the real soln so we approximate the error using RICHARDSON'S.
 Always

HOW DO WE DERIVE RICHARDSON'S ...

1. Determine The order of the method: SECOND
2. With standard assumptions on differentiability of $f(x, y)$ and y , we can write

$$y(x_n) - y_h(x_n) \approx D(x_n)h^2 + O(h^3)$$

error decreases
as h^2 and there
is a function $D(x)$
that can predict
that...

for $D(x)$ a continuous unknown
function.

NOTE ORDER ONE: $D(x_n)h + O(h^2)$

we just can't know $D(x)$

3. Write down an approximate error for h and $2h$...

$$\begin{aligned} y(x_n) - y_h(x_n) &\approx D(x_n)h^2 \\ y(x_n) - y_{2h}(x_n) &\approx D(x_n)4h^2 \end{aligned}$$

4. Make the $D(x_n)$ disappear by combining these
• Here we do 4 times the top minus the bottom.

$$3y(x_n) - 4y_h(x_n) + y_{2h}(x_n) \approx 0$$

5. Rearrange to get $y(x_n) - y_h(x_n)$ on LHS.

$$3[y(x_n) - y_h(x_n)] - y_h(x_n) + y_{2h}(x_n) \approx 0$$

$$y(x_n) - y_h(x_n) \approx \frac{1}{3}[y_h(x_n) - y_{2h}(x_n)]$$

We can use this to estimate
the error in $y_h(x_n)$.

In practice we make sure the estimate $\frac{1}{3}[y_h(x_n) - y_{2h}(x_n)]$
is well below our error tolerance. Choose an
 h so that this error is very small.

PSEUDO CODE ... what do we need to do to implement this?

$$y(n+1) = y(n) + \frac{h}{2}[3f(x_n, y_n) - f(x_{n-1}, y_{n-1})]$$

note • The first step must use a one step method!
Usually EULER or RUNGE KUTTA.
depending on accuracy needed.

This approximation is an example of an ADAMS-BASHFORTH method.

- Considered to be less computationally expensive than RUNGE-KUTTA
 - we can save past evaluations of $f(x, y)$ so that we only evaluate it once per step.

General ADAMS-BASHFORTH Methods :

$$y(n+1) = y(n) + \int_{x_n}^{x_{n+1}} f(x, y) dx$$

before we used P_1 but we could always pick a higher order interpolating polynomial using previous grid points to interpolate.

$q+1$ points : $\{x_n, x_{n-1}, x_{n-2}, \dots, x_{n-q}\}$

$$\begin{aligned} \text{then } P_q(x) &= \sum_{j=0}^q L_{j,n}(x) f(x_{n-j}, y(x_{n-j})) \\ &= \sum_{j=0}^q L_{j,n}(x) y'(x_{n-j}) \end{aligned}$$

because $y' = f(x, y)$

here $L_{j,n}(x)$ are the Lagrange Basis Functions.

so

$$y(n+1) = y(n) + \int_{x_n}^{x_{n+1}} P_q(x) dx$$

$$= y(n) + h \sum_{j=0}^q w_{jq} f(x_{n-j}, y(x_{n-j})) + E_n$$

→ Truncation Error.

$$\text{where } hw_{jq} = \int_{x_n}^{x_{n+1}} L_{j,n}(x) dx$$

we must integrate each of the basis functions individually

$$\text{so } y_{n+1} = y_n + h \sum_{j=0}^q w_{jq} f(x_{n-j}, y(x_{n-j}))$$

is a $(q+1)$ -step, explicit method, with a truncation error of

$$O(h^{q+2})$$

$q = 0$:

$$y_{n+1} = y_n + hf(x_n, y_n) \quad [\text{Euler's method}]$$

$$E_n = \frac{1}{2}h^2 Y''(\xi_n)$$

$q = 1$: Use the notation $y'_k = f(x_k, y_k)$.

$$y_{n+1} = y_n + \frac{h}{2} [3y'_n - y'_{n-1}], \quad n \geq 1$$

$$E_n = \frac{5}{12}h^3 Y'''(\xi_n)$$

$q = 2$:

$$y_{n+1} = y_n + \frac{h}{12} [23y'_n - 16y'_{n-1} + 5y'_{n-2}]$$

$$E_n = \frac{3}{8}h^4 Y^{(4)}(\xi_n)$$

$q = 3$:

$$y_{n+1} = y_n + \frac{h}{24} [55y'_n - 59y'_{n-1} + 37y'_{n-2} - 9y'_{n-3}]$$

$$E_n = \frac{251}{720}h^5 Y^{(5)}(\xi_n)$$

EXAMPLES OF THESE (Hand out).

- NOTE ... The $q+1$ initial values must be computed with other methods...

Runge-Kutta or
a LOWER ORDER Adams-Bashforth.

- CONVERGENCE... Similar to Euler methods we can show that convergence depends on higher derivatives.

From the examples we see that

$$\max_{x_0 \leq x \leq b} |E_n| \leq d h^{q+2} \max_{x_0 \leq x \leq b} |y^{(q+2)}(x)|$$

- STABILITY ... not absolutely stable but we can choose an h that works.

- IN practice: we do a "convergence test"

Save all
this data →

1. Start with a "large" h value
2. solve the system
3. cut h in half
4. solve

continue doing this until h is quite small

Then Analyze:

- A. As h decreases the solutions should be approaching the REAL soln.
- B. The ratio of Richardson Errors
Should be decreasing with same
order as the METHOD.
- C. As you reduce h more and more
the numerical solutions should
vary less and less.

Recall the notation $y'_k = f(x_k, y_k)$, $k \geq 0$.

$q = 0$: It is the backward Euler method,

$$\begin{aligned}y_{n+1} &= y_n + hy'_{n+1} \\ \hat{E}_n &= -\frac{1}{2}h^2Y''(\xi_n)\end{aligned}$$

$q = 1$: It is the trapezoidal method,

$$\begin{aligned}y_{n+1} &= y_n + \frac{h}{2}[y'_{n+1} + y'_n] \\ \hat{E}_n &= -\frac{1}{12}h^3Y^{(3)}(\xi_n)\end{aligned}$$

$q = 2$:

$$\begin{aligned}y_{n+1} &= y_n + \frac{h}{12}[5y'_{n+1} + 8y'_n - y'_{n-1}] \\ \hat{E}_n &= -\frac{1}{24}h^4Y^{(4)}(\xi_n)\end{aligned}$$

$q = 3$:

$$\begin{aligned}y_{n+1} &= y_n + \frac{h}{24}[9y'_{n+1} + 19y'_n - 5y'_{n-1} + y'_{n-2}] \\ \hat{E}_n &= -\frac{19}{720}h^5Y^{(5)}(\xi_n)\end{aligned}$$

ADAMS - MOULTON METHODS...

only difference $y(n+1) = y(n) + \int_{x_n}^{x_{n+1}} f(x, y) dx$

approximate using polynomial interpolants at points starting with x_{n+1}

$$\{x_{n+1}, x_n, \dots, x_{n-q+1}\}$$

This means the method is always IMPLICIT.

EXAMPLES OF THESE...

general form

$$y_{n+1} = y_n + h \sum_{j=-1}^{q-1} v_{jq} f(x_{n-j}, y_{n-j})$$

where $h v_{jq} = \int_{x_n}^{x_{n+1}} \hat{L}_{jq}(x) dx$

↓
the lagrange basis
using the new points.

NOTES.

1 Higher order Adams - Moulton methods are not absolutely stable - but they are often more desirable for highly unstable or poorly conditioned problems.

2. To use these methods we:

A. Make an initial guess using an explicit method

B. use fixed point iteration to solve the implicit eqn.

often we use an Adams - Bashforth method with comparable accuracy.