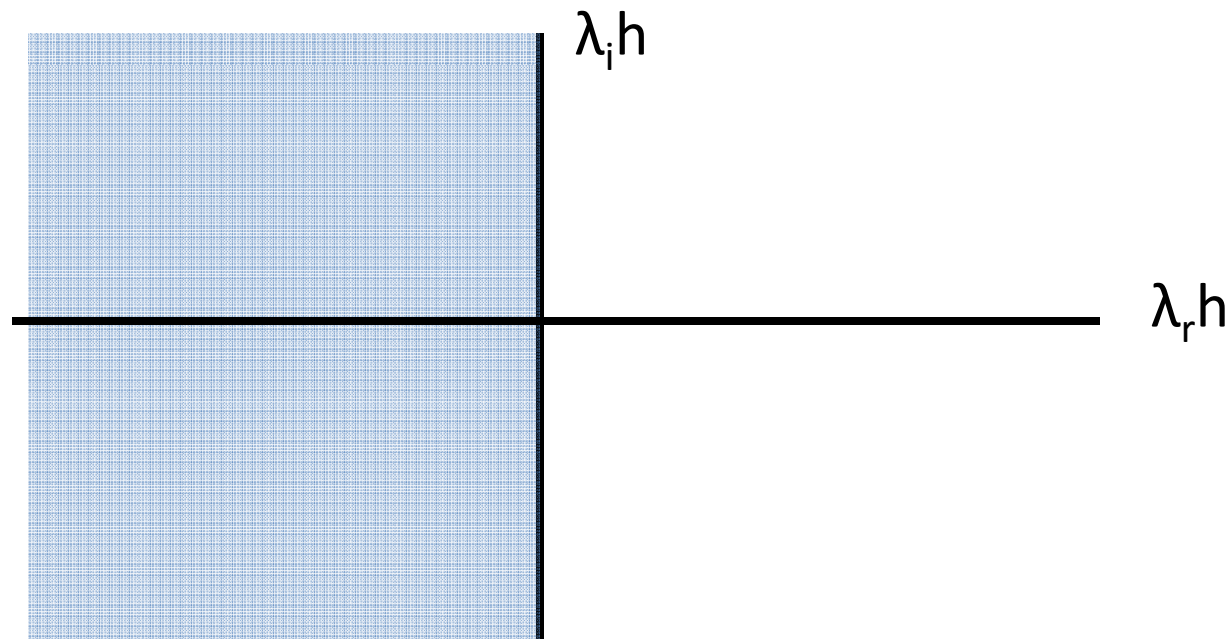


Linear Stability Analysis

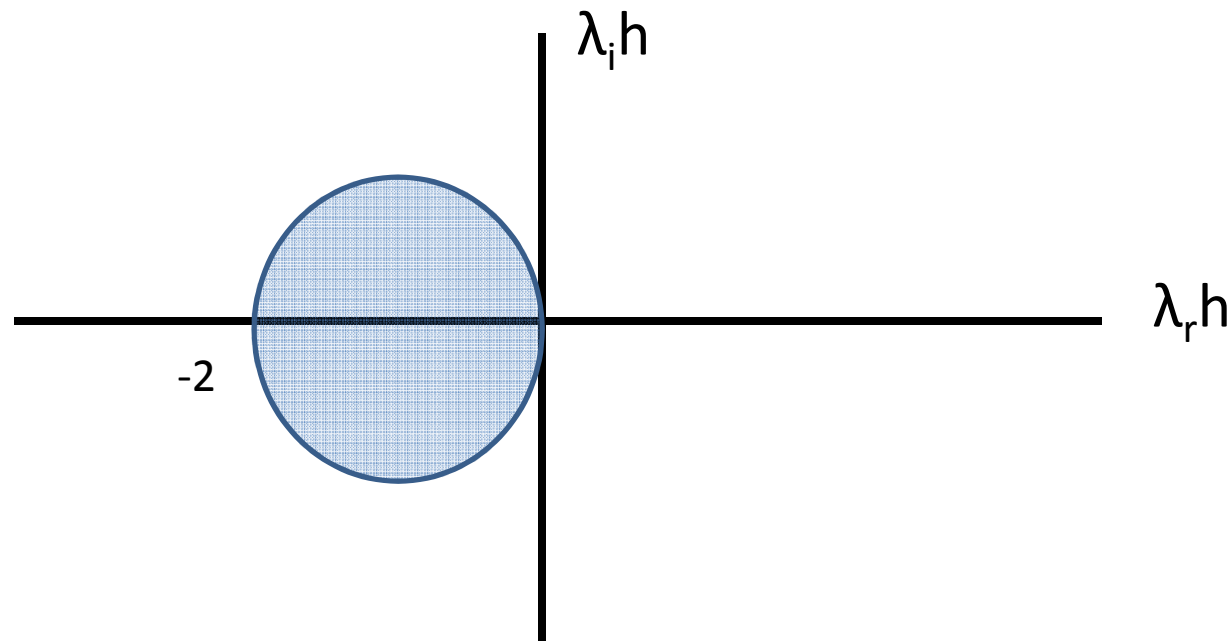
$$y = y_0 e^{n(\lambda_r h + i\lambda_i h)}$$

- The analytical solution is bounded for all negative λ_r
- Stability Region



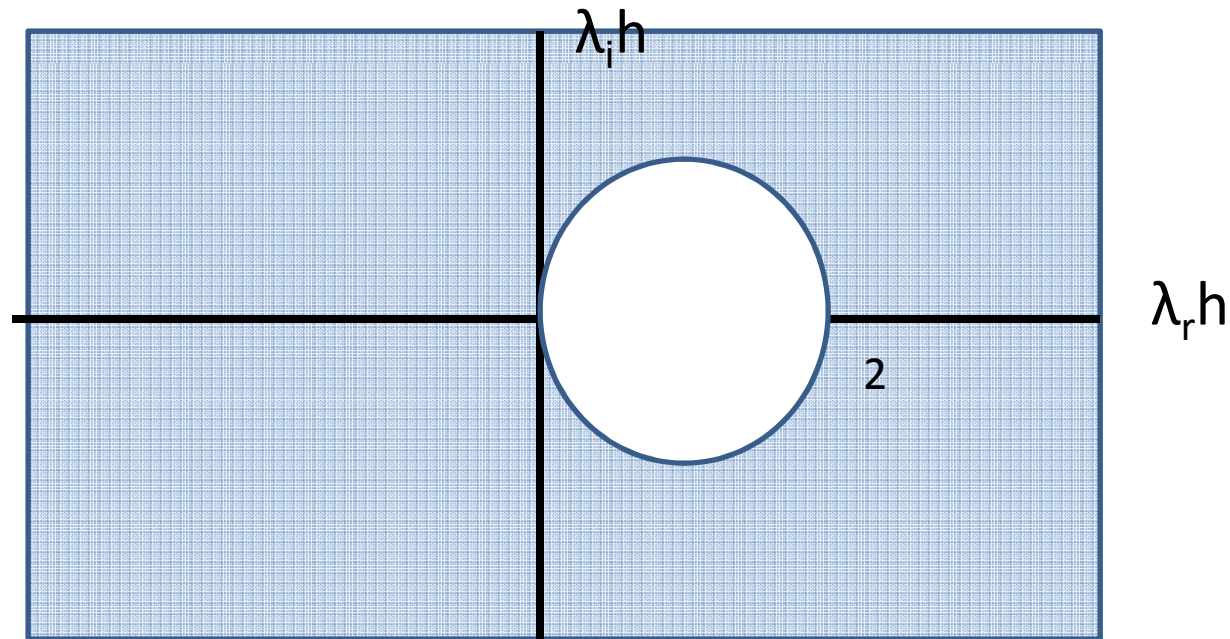
Linear Stability Analysis: Euler Forward

- The stability region is shown below: a circle of radius 1, centered at $(-1,0)$
- For real negative values of λ , the condition is $|\lambda h| \leq 2$



Linear Stability Analysis: Euler Backward

- The stability region is shown below: **outside** a circle of radius 1, centered at $(-1,0)$
- For real negative values of λ , the method is unconditionally stable



Linear Stability Analysis: Trapezoidal method

- For Trapezoidal method

$$y_{n+1} = y_n + h \frac{f(t_n, y_n) + f(t_{n+1}, y_{n+1})}{2} \Rightarrow y_{n+1} = y_n \frac{1 + \lambda_r h / 2 + i \lambda_i h / 2}{1 - \lambda_r h / 2 - i \lambda_i h / 2}$$

- The stability region is, therefore, given by

$$\left| \frac{1 + \lambda_r h / 2 + i \lambda_i h / 2}{1 - \lambda_r h / 2 - i \lambda_i h / 2} \right| \leq 1$$

- Which implies $\lambda_r h \leq 0$
- Same as that for the exact solution.
- Unconditionally stable, does not give bounded solution when the exact is not bounded!

Linear Stability Analysis: R-K method

- For 2nd order R-K method (Heun's)

$$y_{n+1} = y_n + h \frac{f(t_n, y_n) + f(t_{n+1}, y_n + hf(t_n, y_n))}{2}$$

$$\Rightarrow y_{n+1} = y_n \left(1 + \lambda h + \lambda^2 h^2 / 2 \right)$$

- The stability region is, therefore, the region inside the shape whose boundary is given by

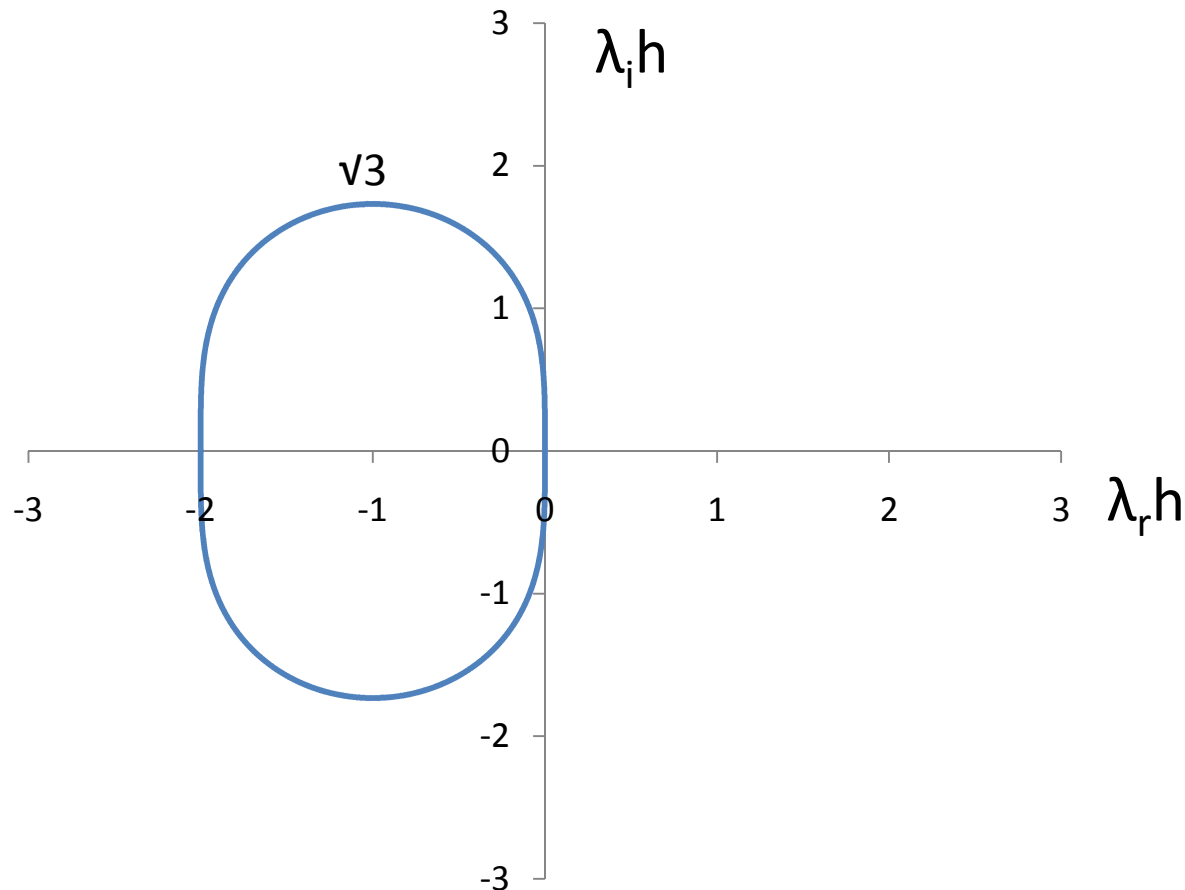
$$\left| 1 + \lambda h + \lambda^2 h^2 / 2 \right| = 1$$

- Or:

$$\lambda_i h = \pm \sqrt{-\lambda_r h(2 + \lambda_r h)} + 2 \sqrt{-\lambda_r h(2 + \lambda_r h)}$$

Stability Region for Heun's method

- The stability region is shown below: centered at $(-1,0)$, major axis = $2\sqrt{3}$, minor = 2
- For real negative values of λ , the condition is $|\lambda h| \leq 2$



Predictor-Corrector methods

- Implicit methods are stable but require solution of a nonlinear equation at each step
- Explicit methods require less computational effort per step but may need a very small time-step for stability
- Avoid the nonlinear equation solution, by predicting the “unknown” value using explicit method and then correcting it using implicit

Predictor-Corrector methods

- For example, Heun's method:

➤ Predictor: $y_{n+1}^p = y_n + hf(t_n, y_n)$

➤ Corrector: $y_{n+1}^c = y_n + \frac{h}{2}[f(t_n, y_n) + f(t_{n+1}, y_{n+1}^p)]$

- Why stop at one step only? Iterate using the corrected value in the implicit step.

$$y_{n+1}^{(0)} = y_n + hf(t_n, y_n)$$

$$y_{n+1}^{(i)} = y_n + \frac{h}{2}[f(t_n, y_n) + f(t_{n+1}, y_{n+1}^{(i-1)})]$$

- Repeat till convergence

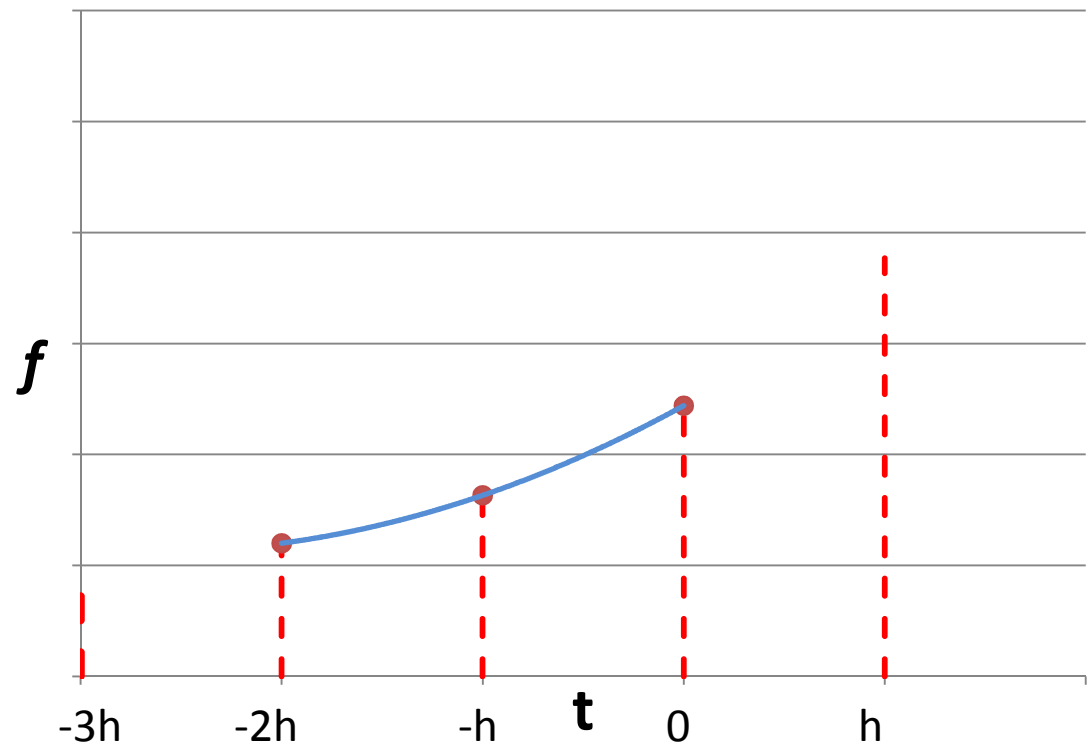
Predictor-Corrector : Milne's method

- Milne's method (multi-step):
- Non-self starting
- Uses Simpson's $1/3$ methodology
- **Predictor**: interpolate a quadratic using $n-2$, $n-1$, and n ; integrate over $n-3$ to $n+1$
- **Corrector**: interpolate a quadratic using $n-1$, n , and $n+1$; integrate over $n-1$ to $n+1$

Milne's method: Predictor

- Approximate f by a quadratic function:

$$\begin{aligned} f = & \frac{(t+h)t}{(-2h+h)(-2h)} f_{n-2} \\ & + \frac{(t+2h)t}{(-h+2h)(-h)} f_{n-1} \\ & + \frac{(t+2h)(t+h)}{(2h)(h)} f_n \end{aligned}$$



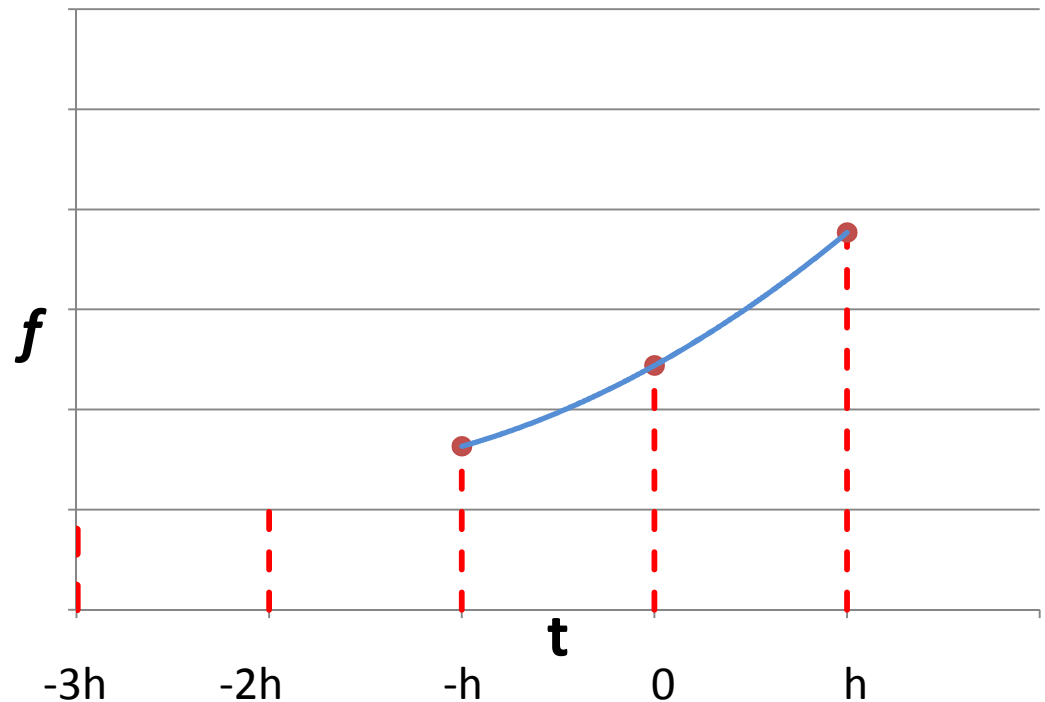
- Integrate from $-3h$ to h :

$$y_{n+1}^{(0)} = y_{n-3} + \int_{-3h}^h f dt = y_{n-3} + \frac{4h}{3} (2f_{n-2} - f_{n-1} + 2f_n)$$

Milne's method: Corrector

- Approximate f by a quadratic function:

$$\begin{aligned} f &= \frac{t(t-h)}{(-h)(-2h)} f_{n-1} \\ &+ \frac{(t+h)(t-h)}{(h)(-h)} f_n \\ &+ \frac{(t+h)t}{(2h)(h)} f_{n+1}^{(i-1)} \end{aligned}$$



- Integrate from $-h$ to h :

$$y_{n+1}^{(i)} = y_{n-1} + \frac{h}{3} [f_{n-1} + 4f_n + f(t_{n+1}, y_{n+1}^{(i-1)})]$$

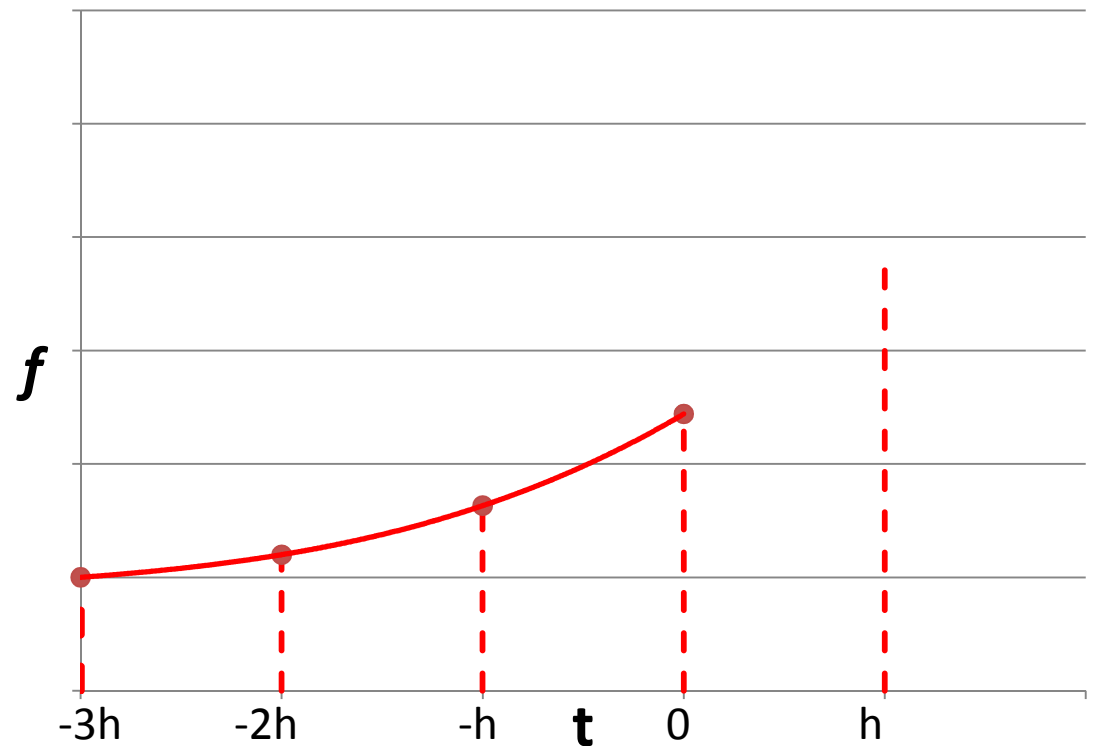
Predictor-Corrector : Adams method

- **Adams method:**
- Uses Adams-Bashforth (explicit) and Adams-Moulton (implicit)
- For Example, take the 4th order method
- **Predictor:** interpolate a cubic using $n-3$, $n-2$, $n-1$, and n ; integrate over n to $n+1$
- **Corrector:** interpolate a cubic using $n-2$, $n-1$, n , and $n+1$; integrate over n to $n+1$

Adams method: Predictor

- Approximate f by a cubic function:

$$\begin{aligned}
 f &= \frac{(t+2h)(t+h)t}{(-h)(-2h)(-3h)} f_{n-3} \\
 &+ \frac{(t+3h)(t+h)t}{(-h)(-h)(-2h)} f_{n-2} \\
 &+ \frac{(t+3h)(t+2h)t}{(2h)(h)(-h)} f_{n-1} \\
 &+ \frac{(t+3h)(t+2h)(t+h)}{(3h)(2h)(h)} f_n
 \end{aligned}$$



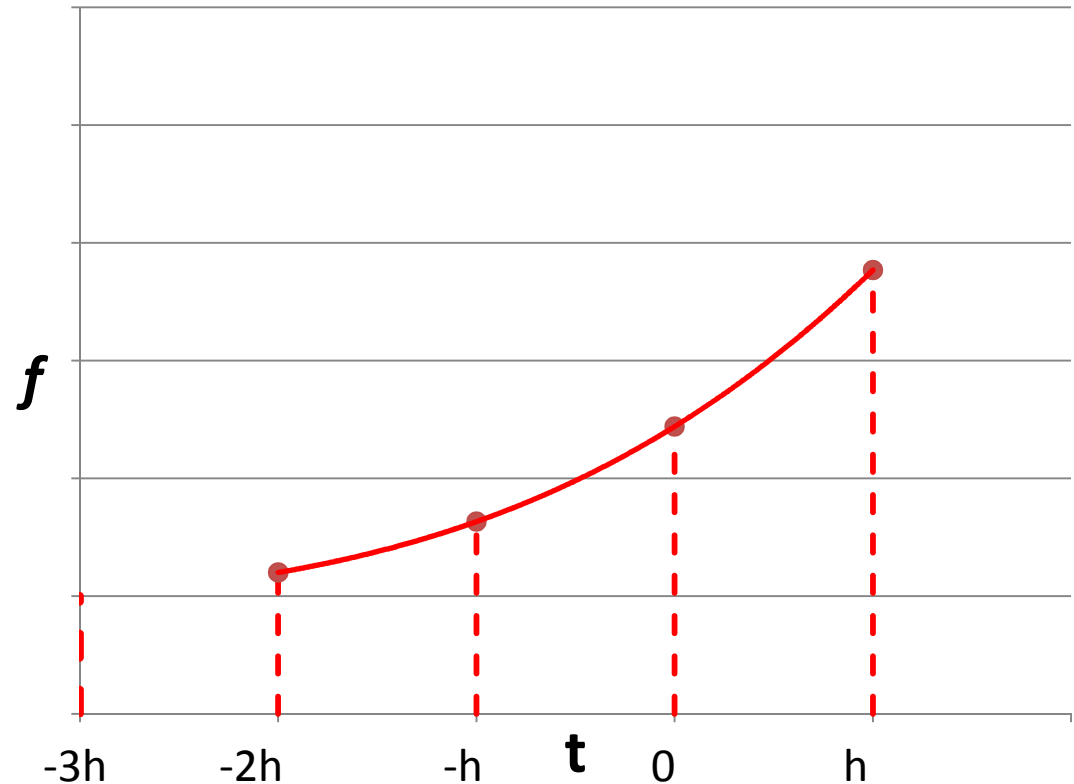
- Integrate from 0 to h:

$$y_{n+1}^{(0)} = y_n + \int_0^h f dt = y_n + h \left(-\frac{3}{8} f_{n-3} + \frac{37}{24} f_{n-2} - \frac{59}{24} f_{n-1} + \frac{55}{24} f_n \right)$$

Adams method: Corrector

- Approximate f by a cubic function:

$$\begin{aligned} f &= \frac{(t+h)t(t-h)}{(-h)(-2h)(-3h)} f_{n-2} \\ &+ \frac{(t+2h)t(t-h)}{(h)(-h)(-2h)} f_{n-1} \\ &+ \frac{(t+2h)(t+h)(t-h)}{(2h)(h)(-h)} f_n \\ &+ \frac{(t+2h)(t+h)t}{(3h)(2h)(h)} f_{n+1}^{(i-1)} \end{aligned}$$



- Integrate from 0 to h :

$$y_{n+1}^{(i)} = y_n + h \left(\frac{1}{24} f_{n-2} - \frac{5}{24} f_{n-1} + \frac{19}{24} f_n + \frac{3}{8} f_{n+1}^{(i-1)} \right)$$

System of ODEs

- If we have several dependent variables, y_i , i from 1 to m
- Derivatives could be functions of time and one or more y^s
- Initial conditions on all y^s should be given
- The system may be expressed as

$$\frac{dy_1}{dt} = f_1(t, y_1, y_2, \dots, y_m)$$

$$\frac{dy_2}{dt} = f_2(t, y_1, y_2, \dots, y_m) \quad y_1|_{(t=0)} = y_{1,0}; y_2|_{(t=0)} = y_{2,0}; \dots; y_m|_{(t=0)} = y_{m,0}$$

...

$$\frac{dy_m}{dt} = f_m(t, y_1, y_2, \dots, y_m)$$

Higher order ODEs

- If we have a higher order ODE, it could be converted into a system of ODEs

- For example,
$$c_2 \frac{d^2 y}{dt^2} + c_1 \frac{dy}{dt} + c_0 y = f(t)$$

- Could be expressed as (using $y_1=y$ and $y_2=dy/dt$):

$$\frac{dy_1}{dt} = f_1(t, y_1, y_2) = y_2$$

$$\frac{dy_2}{dt} = f_2(t, y_1, y_2) = \frac{f(t) - c_0 y_1 - c_1 y_2}{c_2}$$