Implicit Methods

- There are second set of multi-step methods, which are known as "implicit" methods.
 - "implicit" => not directly revealed
- Here it means that the value of the function at the later time is not provided in an "explicit" formula, but in an equation
- Since future data is used an iterative method must be used to iterate an initial guess to convergence
- Could use Runge-Kutta or Adams Bashforth to start the initial value problem.

Backward Euler

- We approximated the derivative at the initial point.
- In backward let us approximate it at the final point
- Find y_{n+1} so that

$$y_{n+1} = y_n + h f(t_{n+1}; y_{n+1})$$
:

• Taylor series derivation

$$y(t) = y(t+h) - hy'(t+h) + \frac{1}{2} h^2 y''(\xi)$$
$$y_{n+1} = y_n + hf_{n+1}$$

- How can we use it? Must solve a non-linear equation
- Generally not used in this way, but as a "correction step" in a "predictor-corrector" scheme.

Modified Euler Method

The Modified Euler method uses the slope at both old and the new location and is a predictorcorrector technique.

$$y_{n+1} = y_n + \Delta h \left(\frac{y'_n + y'_{n+1}}{2} \right) + O(\Delta h^2)$$

The method uses the average slope between the two locations.

Modified Euler Method

The algorithm will be:

$$y'_{n} = f(x_{n}, y_{n})$$

$$y'_{n+1} = y_{n} + \Delta h y'_{n}$$
Initial guess of the value

$$y'^{*}_{n+1} = f(x_{n+1}, y^{*}_{n+1})$$
Updated value

$$y_{n+1} = y_{n} + \Delta h \left(\frac{y'_{n} + {y'}^{*}_{n+1}}{2}\right)$$

MEM: Improves order of the method

If we were to look at the Taylor series expansion

$$y'_{n+1} = y_n + h y'_n + \frac{1}{2}h^2 y''_n + O(h^3)$$

Use a forward difference to represent the 2nd derivative

$$y'_{n+1} = y_n + h y'_n + \frac{1}{2} h^2 \left(\frac{y'_{n+1} - y'_n}{h} + O(h) \right) + O(h^3)$$
$$= y_n + h y'_n + \frac{1}{2} h y'_{n+1} - \frac{1}{2} h y'_n + O(h^3)$$
$$= y_n + h \left(\frac{y'_{n+1} + y'_n}{2} \right) + O(h^3)$$

Predictor Corrector methods

- **P** (predict): Guess y_{n+1} (e.g., using Euler's method).
- **E** (evaluate): Evaluate $f_{n+1} = f(t_{n+1}; y_{n+1})$.
- C (correct): Plug the current guess in, to get a new guess:

$$y_{n+1} = y_n + h_n f_{n+1}$$
:

• **E**: Evaluate

$$f_{n+1} = f(t_{n+1}; y_{n+1}).$$

- Repeat the CE steps if necessary.
- We call this a PECE (or PE(CE)^{*k*}) scheme.

One Step Method

The one-step techniques

- These methods allow us to vary the step size.
- Use only one initial value.
- After each step is completed the past step is "forgotten: We do not use this information.

Explicit and One-Step Methods

Up until this point we have dealt with:

- Euler Method
- Runge-Kutta Methods

These methods are called explicit methods, because they use only the information from previous steps.

Moreover these are one-step methods

Multi-Step Methods

The principle behind a multi-step method is to use past values, y and/or dy/dx to construct a polynomial that approximate the derivative function.

- Represent f(x,y) as a polynomial in x using known values over the past few steps.
- E.g., using Lagrangian form and equal steps, we have for 3 steps
- $(-2h, f_{-2}) (-h, f_{-1}), (0, f_0)$
- So the polynomial is
- $f(x) = f_{-2}(x+h)x/(2h^2) f_{-1}(x+2h)x/h^2 + f_0(x+h)(x+2h)/(2h^2)$ = $(x^2(f_{-2}+2f_{-1}+f_0)+hx(f_{-2}+4f_{-1}+3f_0)+2h^2f_0)/2h^2$
- Integrate from (x_i, x_{i+1})

 $y_{n+1} = y_n + h\left(\frac{23}{12}f(t_n,y_n) - \frac{16}{12}f(t_{n-1},y_{n-1}) + \frac{5}{12}f(t_{n-2},y_{n-2})\right)$

Multi-Step Methods

These methods are known as explicit schemes because the use of current and past values are used to obtain the future step.

The method is initiated by using either a set of known results or from the results of a Runge-Kutta to start the initial value problem.

Adam Bashforth Method (4 Point) Example



4 Point Adam Bashforth

From the 4th order Runge Kutta f(0,1) = 1.0000 f(0.1,1.104829) = 1.094829 f(0.2,1.218597) = 1.178597f(0.3,1.340141) = 1.250141

The 4 Point Adam Bashforth is:

$$\Delta y = \frac{0.1}{24} \left[55f_{0.3} - 59f_{0.2} + 37f_{0.1} - 9f_0 \right]$$

4 Point Adam Bashforth

The results are:

$$\Delta y = \frac{0.1}{24} \begin{bmatrix} 55(1.250141) - 59(1.178597) \\ +37(1.094829) - 9(1) \end{bmatrix}$$

= 0.128038

Upgrade the values

y(0.4) = 1.340141 + 0.128038 = 1.468179f(0.4, 1.468179) = 1.308179

4 Point Adam Bashforth Method -Example

х	Adam Bashforth	f(x,y)	sum	4th order Runge-Kutta	Exact
0	1	1		1	1
0.1	1.104828958	1.094829		1.104828958	1.104829
0.2	1.218596991	1.178597		1.218596991	1.218597
0.3	1.34014081	1.250141	30.72919	1.34014081	1.340141
0.4	1.468179116	1.308179	31.94617	1.468174786	1.468175
0.5	1.601288165	1.351288	32.78612	1.601278076	1.601279
0.6	1.737896991	1.377897	33.20969	1.737880409	1.737881
0.7	1.876270711	1.386271	33.17302	1.876246365	1.876247
0.8	2.014491614	1.374492	32.62766	2.014458009	2.014459
0.9	2.150440205	1.34044	31.52015	2.150395695	2.150397
1	2.281774162	1.281774	29.79136	2.281716852	2.281718

The values for the Adam Bashforth

4 Point Adam Bashforth Method -Example

The explicit Adam Bashforth method gave solution gives good results without having to go through large number of calculations.



Modified Euler's Method Example

Consider

$$\frac{dy}{dx} = x + y$$

The initial condition is: y(0) = 1The step size is: $\Delta h = 0.02$ The analytical solution is:

$$y(x) = 2e^x - x - 1$$

Modified Euler's Method Example

The results are:

				Estimated	Solution	Average	Exact	Error
Xn	y _n	y'n	hy'n	y_{n+1}	y' _{n+1}	$h(y'_{n+}y'_{n+1})/2$	Solution	
0	1.00000	1.00000	0.02000	1.02000	1.04000	0.02040	1.00000	0.00000
0.02	1.02040	1.04040	0.02081	1.04121	1.08121	0.02122	1.02040	0.00000
0.04	1.04162	1.08162	0.02163	1.06325	1.12325	0.02205	1.04162	-0.00001
0.06	1.06366	1.12366	0.02247	1.08614	1.16614	0.02290	1.06367	-0.00001
0.08	1.08656	1.16656	0.02333	1.10989	1.20989	0.02376	1.08657	-0.00001
0.1	1.11033	1.21033	0.02421	1.13453	1.25453	0.02465	1.11034	-0.00001
0.12	1.13498	1.25498	0.02510	1.16008	1.30008	0.02555	1.13499	-0.00002
0.14	1.16053	1.30053	0.02601	1.18654	1.34654	0.02647	1.16055	-0.00002
0.16	1.18700	1.34700	0.02694	1.21394	1.39394	0.02741	1.18702	-0.00002
0.18	1.21441	1.39441	0.02789	1.24229	1.44229	0.02837	1.21443	-0.00003
0.2	1.24277	1.44277	0.02886	1.27163	1.49163	0.02934	1.24281	-0.00003

Implicit Multi-Step Methods

The main method is Adams Moulton Method

Three Point Adams-Moulton Method

$$\Delta y = \frac{h}{12} \left[5f_{i+1} + 8f_i - f_{i-1} \right]$$

Four Point Adams-Moulton Method

$$\Delta y = \frac{h}{24} \left[9f_{i+1} + 19f_i - 5f_{i-1} + f_{i-2} \right]$$

Implicit Multi-Step Methods

•The method uses what is known as a Predictor-Corrector technique.

•explicit scheme to estimate the initial guess

- •uses the value to guess the future y^* and $dy/dx = f^*(x,y^*)$
- Using these results, apply Adam Moulton method

Implicit Multi-Step Methods

Adams third order Predictor-Corrector scheme.

Use the Adam Bashforth three point explicit scheme for the initial guess.

$$y_{i+1}^* = y_i + \frac{\Delta h}{12} [23f_i - 16f_{i-1} + 5f_{i-2}]$$

Use the Adam Moulton three point implicit scheme to take a second step.

$$y_{i+1} = y_i + \frac{\Delta h}{12} \left[5f_{i+1}^* + 8f_i - f_{i-1} \right]$$



4 Point Adam Bashforth

From the 4th order Runge Kutta

$$f(0,1) = 1.0000$$

$$f(0.1,1.104829) = 1.094829$$

$$f(0.2,1.218597) = 1.178597$$

The 3 Point Adam Bashforth is:

$$\Delta y = \frac{0.1}{12} \left[23f_{0.2} - 16f_{0.1} + 5f_{0.0} \right]$$

3 Point Adam Moulton Predictor-Corrector Method

The results of explicit scheme is:

$$\Delta y = \frac{0.1}{12} [23(1.178597) - 16(1.094829) + 5(1)]$$

= 0.121587

The functional values are:

y*(0.3) = 1.218597 + 0.121587 = 1.340184f*(0.3,1.340184) = 1.250184

3 Point Adam Moulton Predictor-Corrector Method

The results of implicit scheme is:

$$\Delta y = \frac{0.1}{12} [5(1.250184) + 8(1.178597) - 1(1.094829)]$$

= 0.121541

The functional values are:

y(0.3) = 1.218597 + 0.121541 = 1.340138f(0.3, 1.340184) = 1.250138

3 Point Adam Moulton Predictor-Corrector Method

The values for the Adam Moulton

	Adam Moulton Three Point Predictor-Corrector Scheme							
х	у	f	sum	у*	f*	sum		
0	1	1						
0.1	1.104829	1.094829						
0.2	1.218597	1.178597	0.121587	1.340184	1.250184	0.121541		
0.3	1.340138	1.250138	0.128081	1.468219	1.308219	0.12803		
0.4	1.468168	1.308168	0.133155	1.601323	1.351323	0.133098		
0.5	1.601266	1.351266	0.136659	1.737925	1.377925	0.136597		
0.6	1.737863	1.377863	0.138429	1.876291	1.386291	0.138359		
0.7	1.876222	1.386222	0.13828	2.014502	1.374502	0.138204		
0.8	2.014425	1.374425	0.136013	2.150438	1.340438	0.135928		
0.9	2.150353	1.340353	0.131404	2.281757	1.281757	0.13131		
1	2.281663	1.281663	0.124206	2.405869	1.195869	0.124102		

3 Point Adam Moulton Predictor-Corrector Method



Nonlinearity

- In general the quantity on the right hand side, *f*, in the standard form can be a nonlinear function of *t* and *y*.
- Nonlinearity implies multiple solutions and "chaos"
- Also has a bearing on how well a numerical solver can integrate the ODE

Linearized Diff Eq.

- Standard form $\frac{dy}{dt} = f(t, y)$ $y(t_0) = y_0$
- Local behavior of the solution to a differential equation near any point (t_c, y_c) can be analyzed by expanding f(t,y)in a two-dimensional Taylor series.

$$f(t, y) = f(t_c, y_c) + \alpha(t - t_c) + J(y - y_c) + \dots$$

• where

$$\alpha = \partial f / \partial t (t_c, y_c)$$
$$J = \partial f / \partial y (t_c, y_c)$$

- (We already used such expansions for deriving the RK method)

- These equations are linear and can consider the three terms on the rhs separately
- Behavior of differential equation governed by the structure of the Jacobian matrix *J*

Linearized differential equations

For a system of differential equations with n components,

$$\frac{d}{dt} \begin{bmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_n(t) \end{bmatrix} = \begin{bmatrix} f_1(t, y_1, \dots, y_n) \\ f_2(t, y_1, \dots, y_n) \\ \vdots \\ f_n(t, y_1, \dots, y_n) \end{bmatrix}$$

the Jacobian is an n-by-n matrix of partial derivatives

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial y_1} & \frac{\partial f_1}{\partial y_2} & \cdots & \frac{\partial f_1}{\partial y_n} \\ \frac{\partial f_2}{\partial y_1} & \frac{\partial f_2}{\partial y_2} & \cdots & \frac{\partial f_2}{\partial y_n} \\ \vdots & \vdots & & \vdots \\ \frac{\partial f_n}{\partial y_1} & \frac{\partial f_n}{\partial y_2} & \cdots & \frac{\partial f_n}{\partial y_n} \end{bmatrix}$$

Jacobian properties

• Interesting equation

$$\frac{dy}{dt} = Jy$$

• In one dimension it can be integrated to obtain the local solution

 $y=C \exp(J t)$

- Express solution similarly for a system.
- Use the eigendecomposition of the Jacobian matrix

$$J=V\Lambda V^{-1}$$

- V matrix with columns as eigenvectors
- $-\Lambda$ eigenvalues arranged as a diagonal matrix
- Why? It removes the coupling of terms in the right hand side by diagonalizing the matrix
- y' = Jy. So $V^{-1}y' = V^{-1}JVV^{-1}y$
- Let V x = y so $x = V^{-1}y$
- transforms the local system of equations to

•
$$dx_k/dt = \lambda_k x_k$$
 $x_k(t) = e^{\lambda_k(t-t_c)} x(t_c)$

- A single component x_k(t) has the following behaviors according to λ_k = μ_k + i v_k
- If μ_k is positive it grows
- It decays if μ_k is negative,
- and oscillates if v_k is nonzero.
- Example: harmonic oscillator $d^2 y / dt^2 = -y$
- s a linear system. The Jacobian is simply the matrix
- J = [0 1] [-10]
- has purely imaginary eigenvalues

Eigenvalues of examples considered Another example from the book The vector y(t) has formed.

$$\begin{aligned} \ddot{u}(t) &= -u(t)/r(t)^3 \\ \ddot{v}(t) &= -v(t)/r(t)^3 \end{aligned}$$

where

г

$$r(t) = \sqrt{u(t)^2 + v(t)^2}$$

$$J = \frac{1}{r^5} \begin{bmatrix} 0 & 0 & r^2 & 0\\ 0 & 0 & 0 & r^5\\ 2y_1^2 - y_2^2 & 3y_1y_2 & 0 & 0\\ 3y_1y_2 & 2y_2^2 - y_1^2 & 0 & 0 \end{bmatrix}$$

- one eigenvalue is real and positive, so that component is growing.
- One eigenvalue is real and negative, corresponding to a decaying component.
- Two eigenvalues are purely imaginary, corresponding to oscillatory components.

The vector
$$y(t)$$
 has four components

$$y(t) = \begin{bmatrix} u(t) \\ v(t) \\ \dot{u}(t) \\ \dot{v}(t) \end{bmatrix}$$

The differential equation is

$$\dot{y}(t) = \begin{bmatrix} \dot{u}(t) \\ \dot{v}(t) \\ -u(t)/r(t)^3 \\ -v(t)/r(t)^3 \end{bmatrix}$$

$$\lambda = \frac{1}{r^{3/2}} \begin{bmatrix} \sqrt{2} \\ i \\ -\sqrt{2} \\ -i \end{bmatrix}$$

Jacobian and ode behavior

- $J = \partial f / \partial y$
- Then a **single** ODE is
 - stable at a point (t_c, y_c) if $J(t_c, y_c) < 0$.
 - **unstable** at a point (t_c, y_c) if $J(t_c, y_c) > 0$.
 - stiff at a point (t_c, y_c) if $J(t_c, y_c) \ll 0$.

• A system of ODEs is

- stable at a point (t_c, y_c) if the real part of all the eigenvalues of the matrix $J(t_c, y_c)$ are negative (converse if some are positive)
- stiff at a point (t_c, y_c) if the real parts of more than one eigenvalue of $J(t_c, y_c)$ are negative and wildly different.

Stiffness

- Stiffness
 - A problem is stiff if the solution being sought is varying slowly, but there are nearby solutions that vary rapidly, so the numerical method must take small steps to obtain satisfactory results.
- Example problem: A match is lit and the fire grows as a ball of flame. until it reaches a critical size. It then remains at that size because the amount of oxygen being consumed by the combustion in the interior of the ball balances the amount available through the surface.
- Let y(t) represent the ball radius. y^2 is proportional to the surface area while y^3 to the volume
 - $y' = y^2 y^3$ $- y(0) = \eta$ $- 0 \le t \le 2/\eta$

Solution using regular and stiff-solver

- choose η=0.01 and 0.0001
- Solve with RK45
- Observe
- Solve with ode23s
- Observe