Runge Kutta Methods

Bernd Schröder

1. **Taylor's Formula**. If the function y is n + 1 times differentiable, then for any h there is a c between x and x + h so that

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- 4. If we can capture more than the first two terms of the Taylor expansion, we could get a global error proportional to $(\Delta x)^n$. This would be good, because Δx is usually small, so a higher power of Δx would be even smaller.

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$$= y(x) + \left(\frac{1}{2}y'(x) + \frac{1}{2}y'(x+h)\right)h + \left(\frac{y'''(c)}{3!} - \frac{y'''(\tilde{c})}{4}\right)h^{3}$$

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Runge-Kutta Methods

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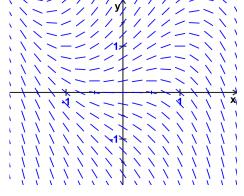
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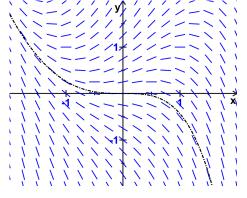
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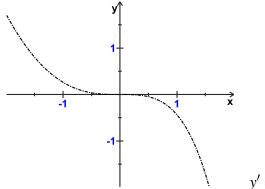
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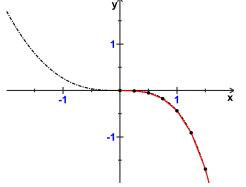
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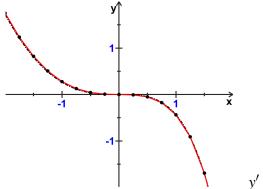
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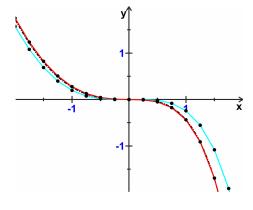
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- 3. So how do we translate this into a formula for differential equations that has high accuracy?
- 4. Match as many terms of Taylor's formula as possible. The remainder term gives the order of the error.
- 5. The global error in Simpson's rule is $\sim (\Delta x)^4$, so we must match the first four terms of the Taylor expansion.

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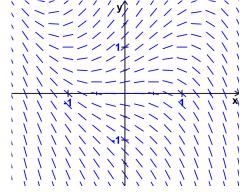
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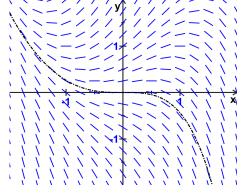
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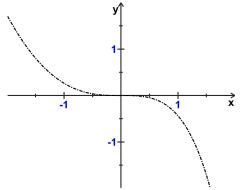
Runge-Kutta Methods



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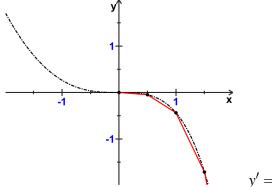


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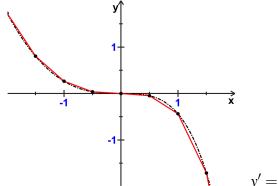
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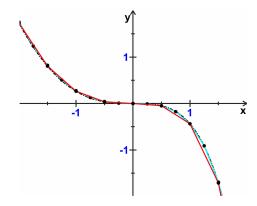
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$$y_{n+1} = y_n + \frac{1}{6} (k_1 + 2k_2 + 2k_3 + k_4), \quad \text{where}$$

$$k_1 = F(x_n, y_n) \Delta x,$$

$$k_2 = F\left(x_n + \frac{1}{2} \Delta x, y_n + \frac{1}{2} k_1\right) \Delta x,$$

$$k_3 = F\left(x_n + \frac{1}{2} \Delta x, y_n + \frac{1}{2} k_2\right) \Delta x,$$

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 y_n will be an approximation for the value of the solution y at x_n .

A j^{th} order **Runge-Kutta procedure** computes an approximate solution to the differential equation y' = F(x,y) by computing a sequence of values $y_{n+1} := y_n + a_1k_1 + a_2k_2 + \cdots + a_mk_m$ so that y_{n+1} agrees with the j^{th} order Taylor polynomial of y at the previous evaluation point.

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Setting up a formula for increments to match a Taylor polynomial actually leads to a system of equations for the parameters in the setup of the approximation formulas. So there is more than one second order Runge-Kutta method. Same goes for higher orders. Some of this freedom can be used to improve performance for certain types of equations.

Error Analysis