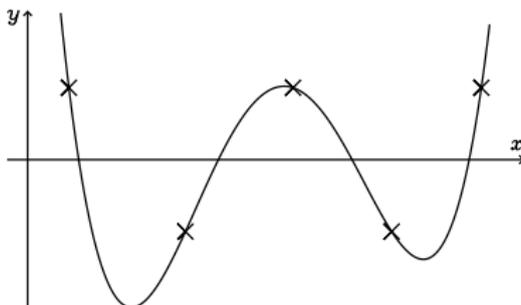


# Finite-differences - mesh drift and superconvergence

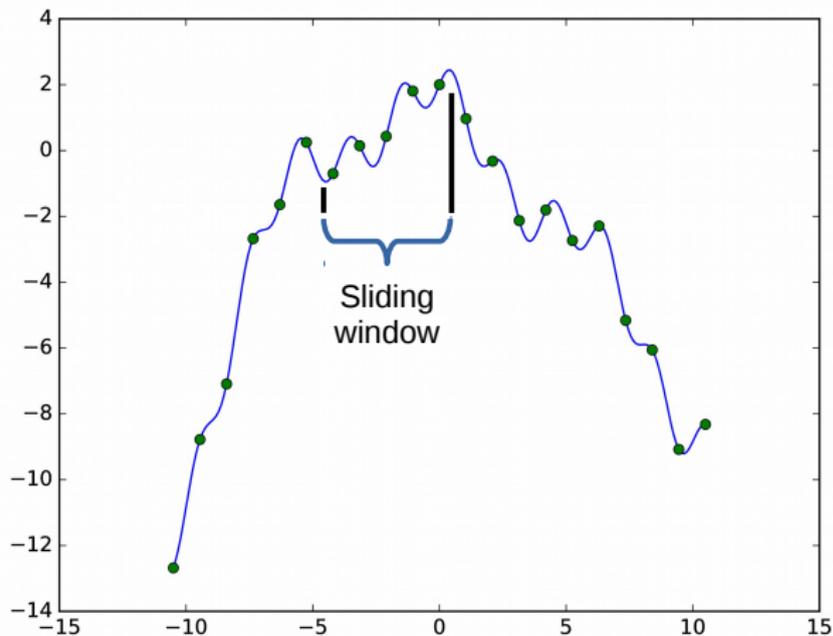
Daniel R. Reese

December 5<sup>th</sup>, 2017



- 1 Reminders on finite-differences
- 2 Mesh drift
- 3 Superconvergence
- 4 Conclusion

# A few reminders on finite-differences



# A few reminders on finite-differences

## What we know

- a set of  $n$  distinct points:  $x_1 < x_2 < \dots < x_n$
- the values of  $f$  at those points:  $f(x_1), f(x_2), \dots, f(x_n)$

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- a set of  $n$  distinct points:  $x_1 < x_2 < \dots < x_n$
- the values of  $f$  at those points:  $f(x_1), f(x_2), \dots, f(x_n)$

## What we would like to know

$f^{(d)}(z)$  for some point  $z$

# A few reminders on finite-differences

We carry out a Taylor expansion around  $z$ :

$$f(x_1) = f(z) + f'(z)(x_1 - z) + \dots + f^{(n-1)} \frac{(x_1 - z)^{n-1}}{(n-1)!} + \mathcal{O}(x_1^n)$$

$$f(x_2) = f(z) + f'(z)(x_2 - z) + \dots + f^{(n-1)} \frac{(x_2 - z)^{n-1}}{(n-1)!} + \mathcal{O}(x_2^n)$$

...

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- search for coefficients  $a_i^d(x)$  such that
 
$$\sum_{i=1}^n a_i^d(z) f(x_i) = f^{(d)}(z) + \sum_{i=1}^n a_i^d(z) \mathcal{O}(x_i^n)$$
- in other words,  $\sum_{i=1}^n a_i^d(z) \frac{(x_i - z)^k}{k!} = \delta_k^d$

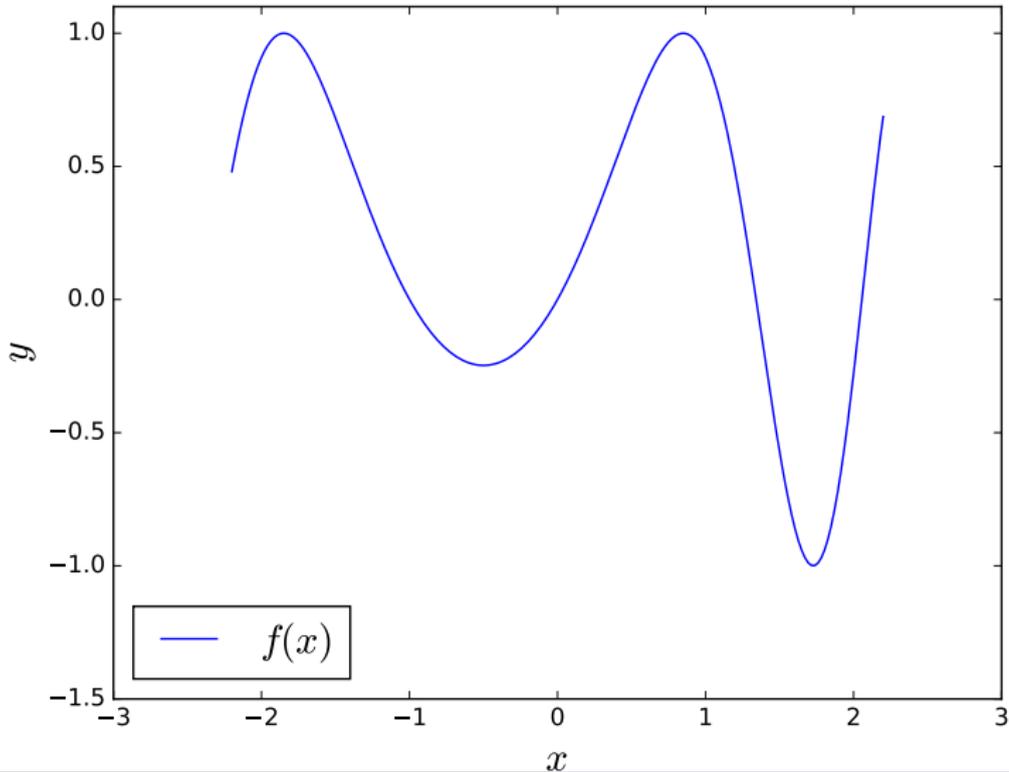
# A few reminders on finite-differences

## Solution

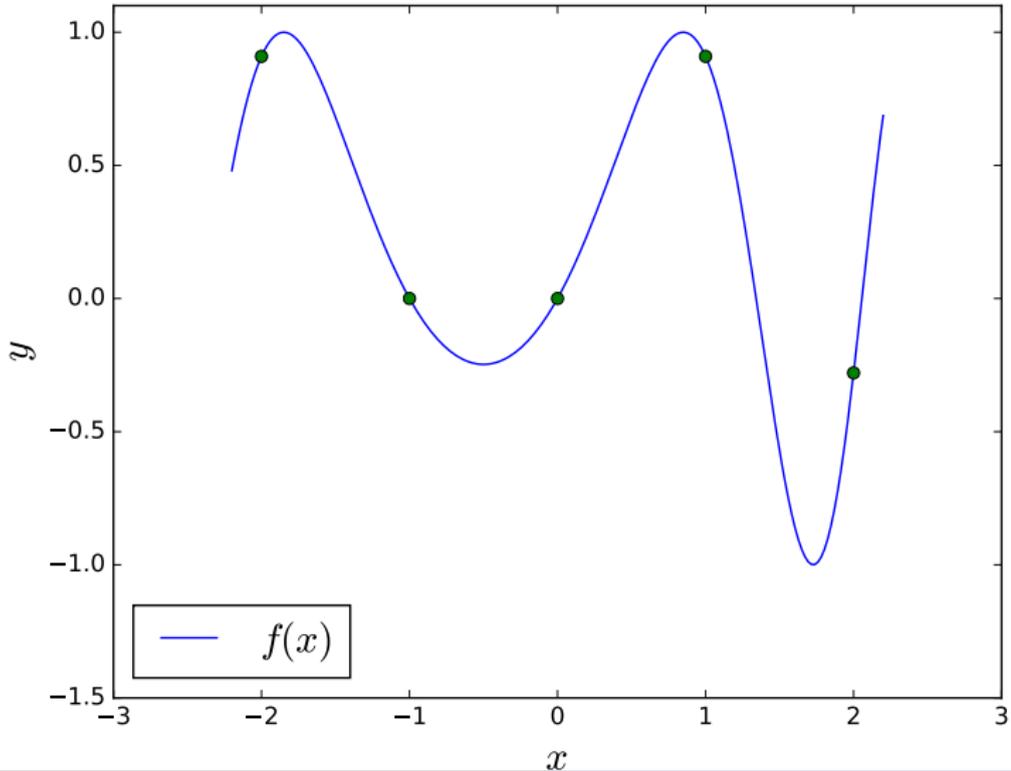
$$a_i^d(z) = L_i^{(d)}(z)$$

where  $L_i(z)$  is the Lagrange interpolation polynomial such that  $L_i(x_j) = \delta_i^j$  (e.g. Fornberg (1988), also see [proof](#) in the appendix).

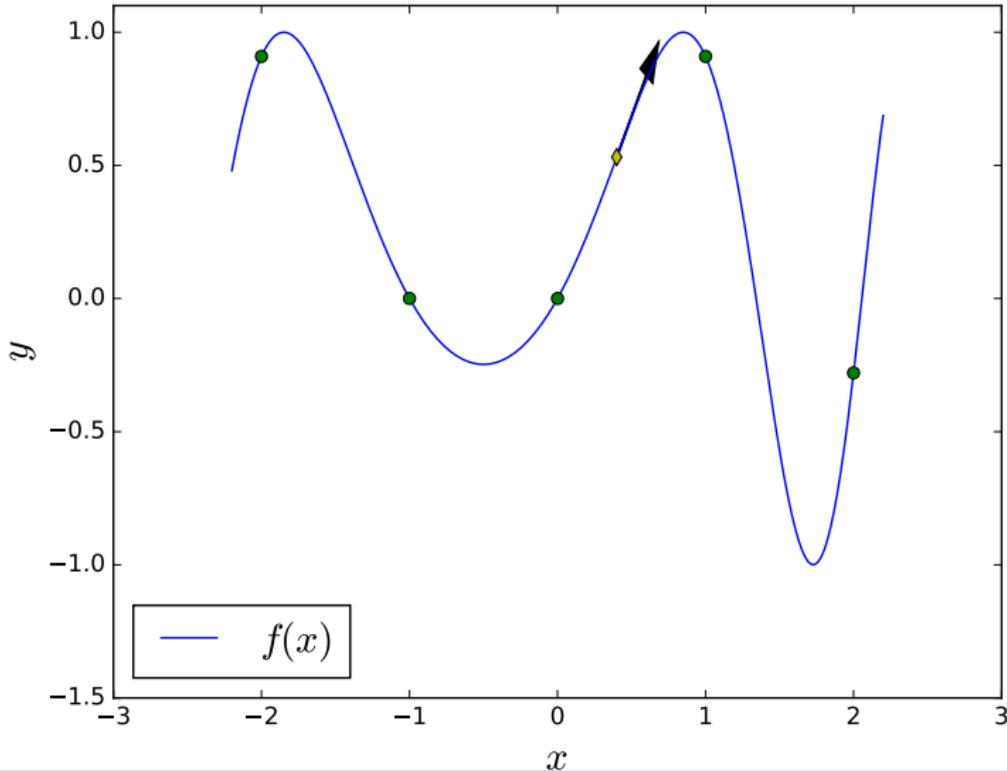
# Another way of looking at this



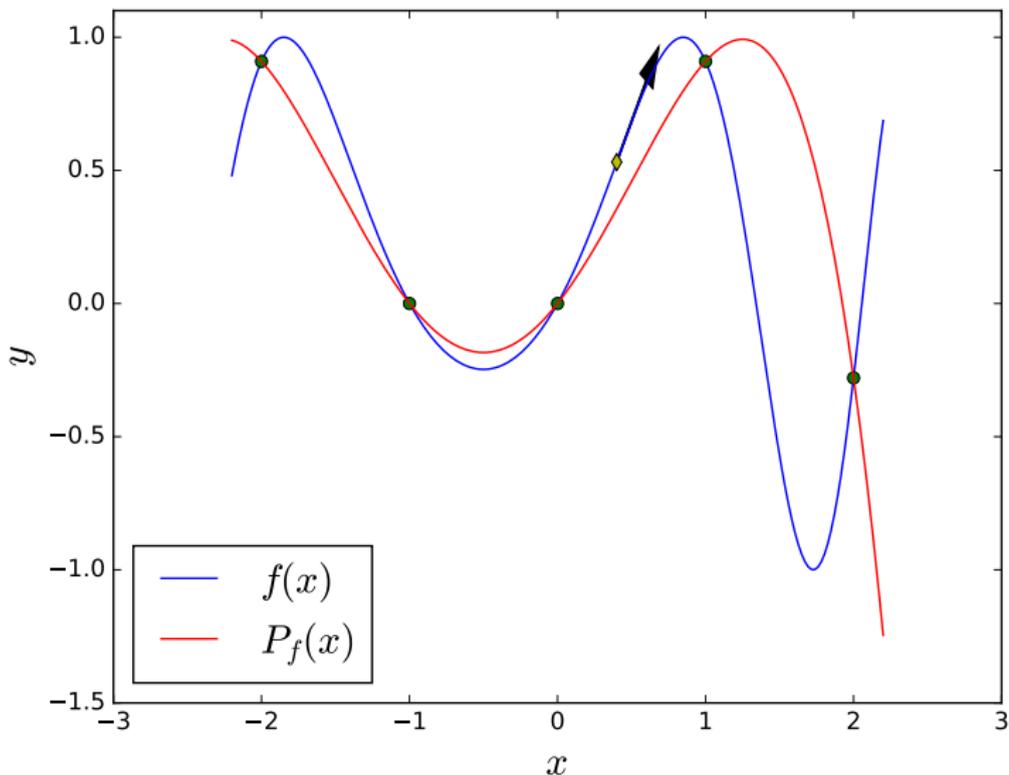
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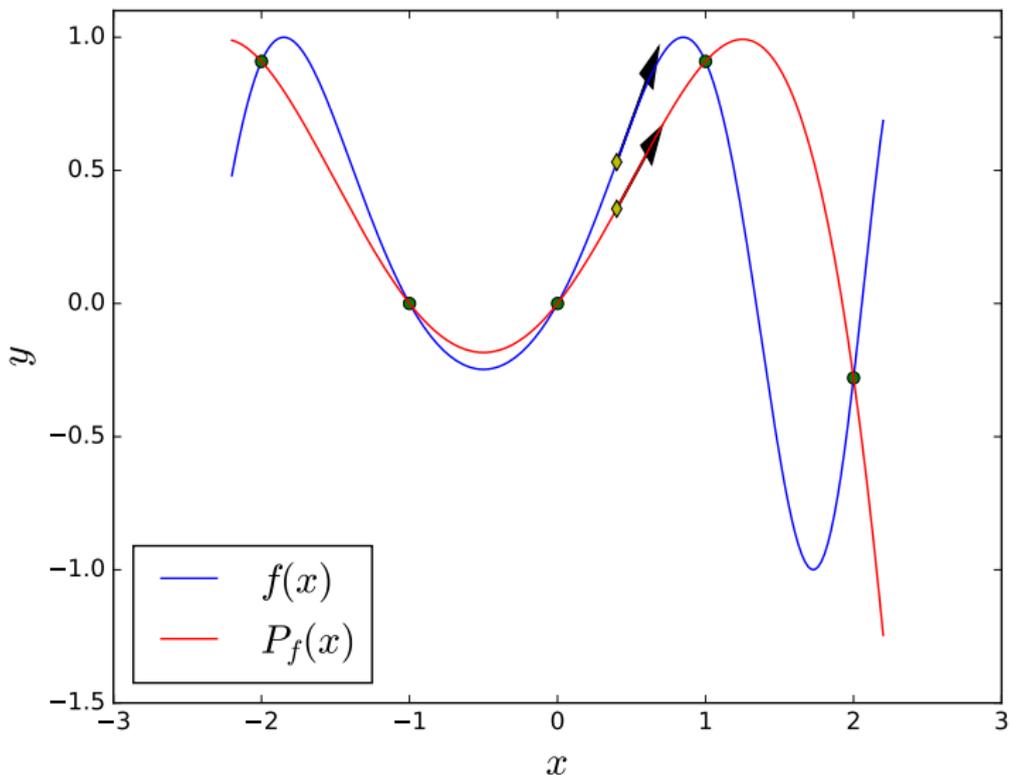
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# A few simple examples

For a uniform grid:

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i} + \mathcal{O}(x)$$

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1}))}{x_{i+1} - x_{i-1}} + \mathcal{O}(x^2)$$

$$f'(x_i) = \frac{-f(x_{i+2}) + 8f(x_{i+1}) - 8f(x_{i-1}) + f(x_{i-2}))}{12(x_{i+1} - x_i)} + \mathcal{O}(x^4)$$

$$f''(x_i) = \frac{f(x_{i+1}) - 2f(x_i) + f(x_{i-1}))}{(x_{i+1} - x_i)^2} + \mathcal{O}(x^2)$$

# Mesh drift

## Definition

- a numerical instability where even and odd grid points decouple
- it typically leads to a “jigsaw” type behaviour in the solution

# Mesh drift – an example

An eigenvalue problem

$$\begin{aligned}v' &= -\omega u \\u' &= \omega v \\u(0) &= u(1) = 0\end{aligned}$$

# Mesh drift – an example

## An eigenvalue problem

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## Analytical solution

$$\begin{aligned}u(x) &= A \sin(\omega x) \\v(x) &= A \cos(\omega x) \\ \omega &= k\pi, \text{ where } k \text{ is an integer}\end{aligned}$$

# Mesh drift – an example

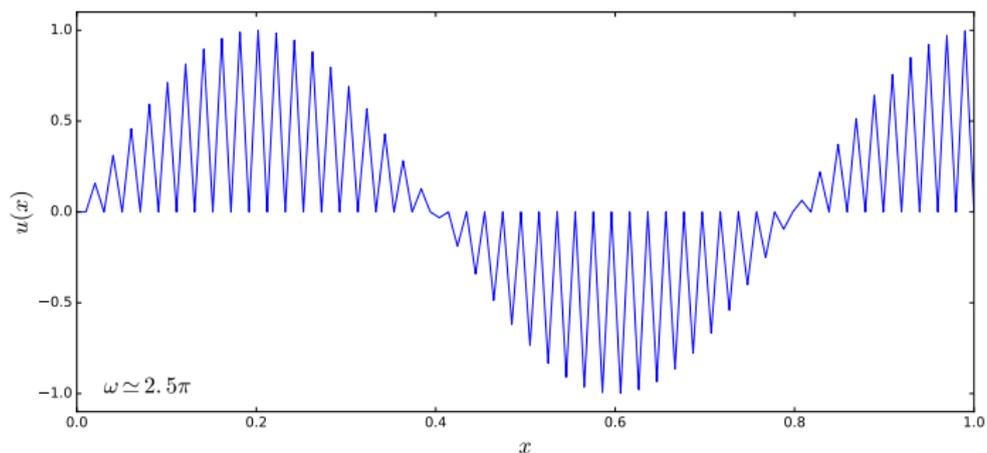
## Numerical setup

$$\begin{aligned}
 -\omega u_i &= \frac{v_{i+1} - v_{i-1}}{2\Delta x}, & \omega v_i &= \frac{u_{i+1} - u_{i-1}}{2\Delta x}, & 2 \leq i \leq N-1, \\
 -\omega u_N &= \frac{v_N - v_{N-1}}{\Delta x}, & \omega v_1 &= \frac{u_2 - u_1}{\Delta x}, \\
 u_1 &= 0, & u_N &= 0,
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# Mesh drift – an example

## Numerical setup 2

- use 4<sup>th</sup> order scheme:

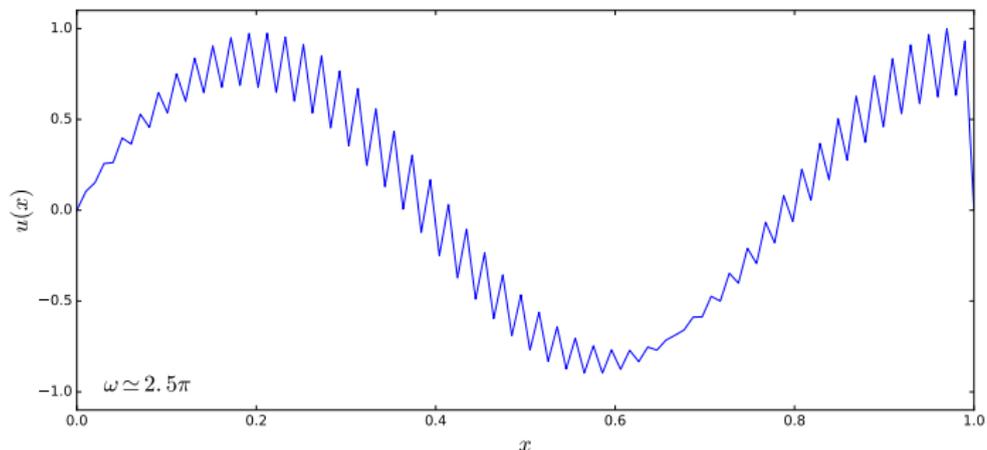
$$f'(x_i) \simeq \frac{-f(x_{i+2}) + 8f(x_{i+1}) - 8f(x_{i-1}) + f(x_{i-2}))}{12(x_{i+1} - x_i)}$$

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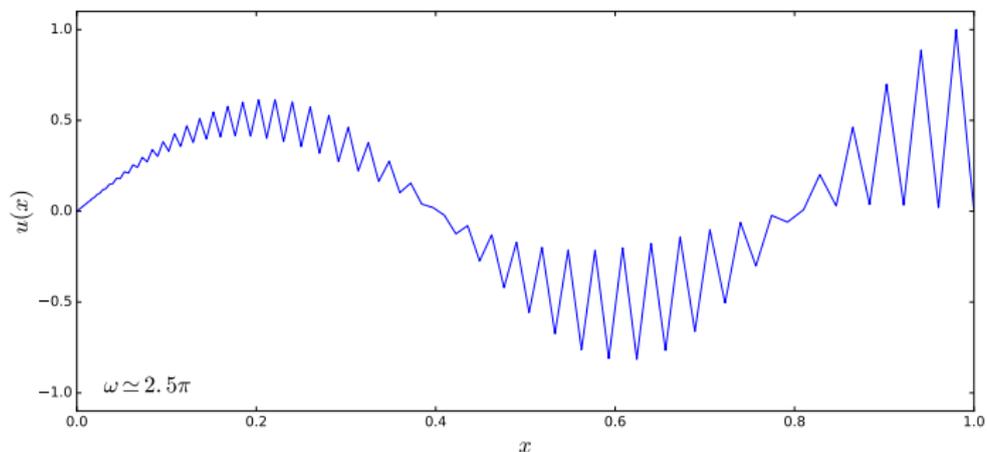
## Numerical setup 3

- use 2<sup>nd</sup> order scheme + non uniform grid

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# Causes of mesh drift

## What are not the causes

- finite-difference expressions with only even or odd grid points
- an even or an odd number of grid points

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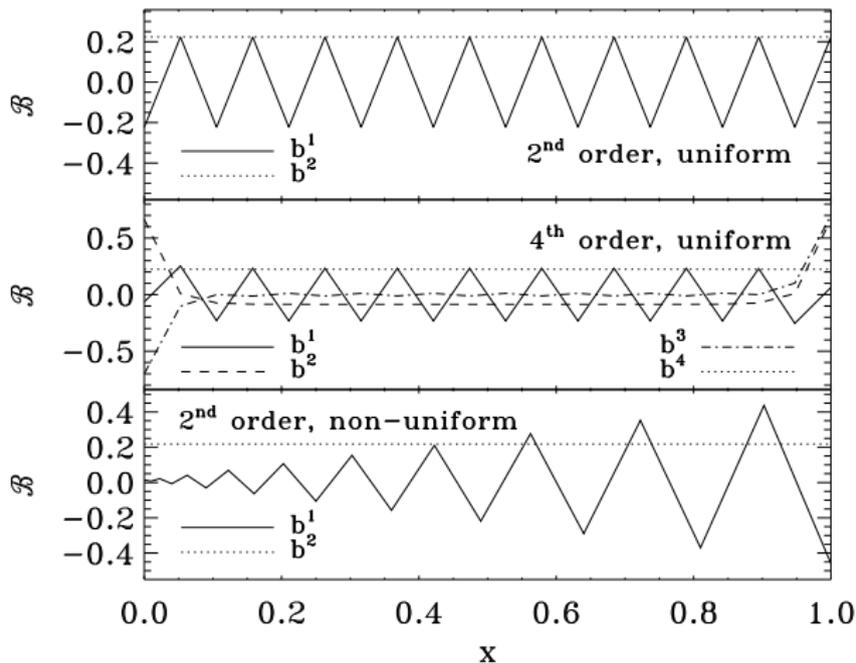
## What are the causes

- finite-difference expressions which allow oscillatory solutions to

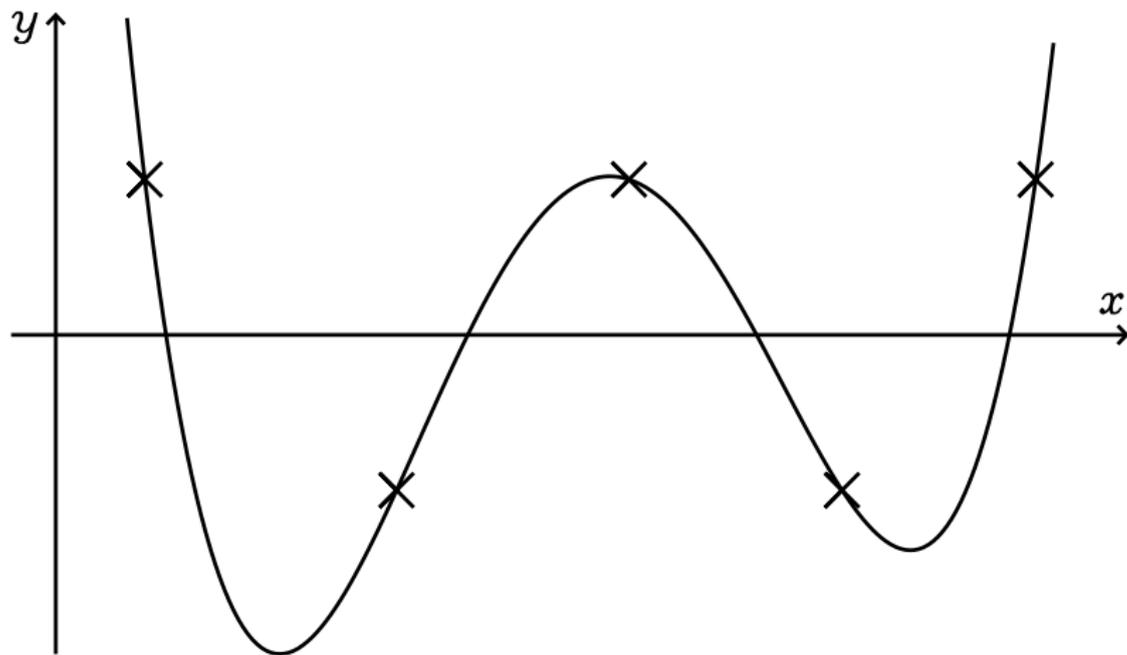
$$f' = 0$$

# Causes of mesh drift

Solutions of  $f' = 0$



# Causes of mesh drift

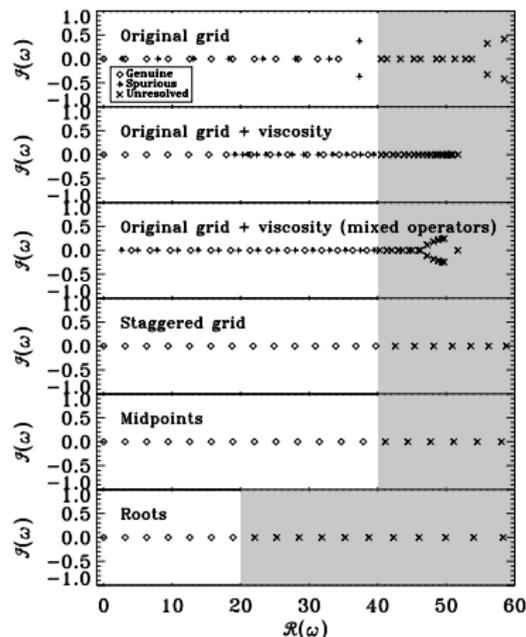


# Remedies for mesh drift

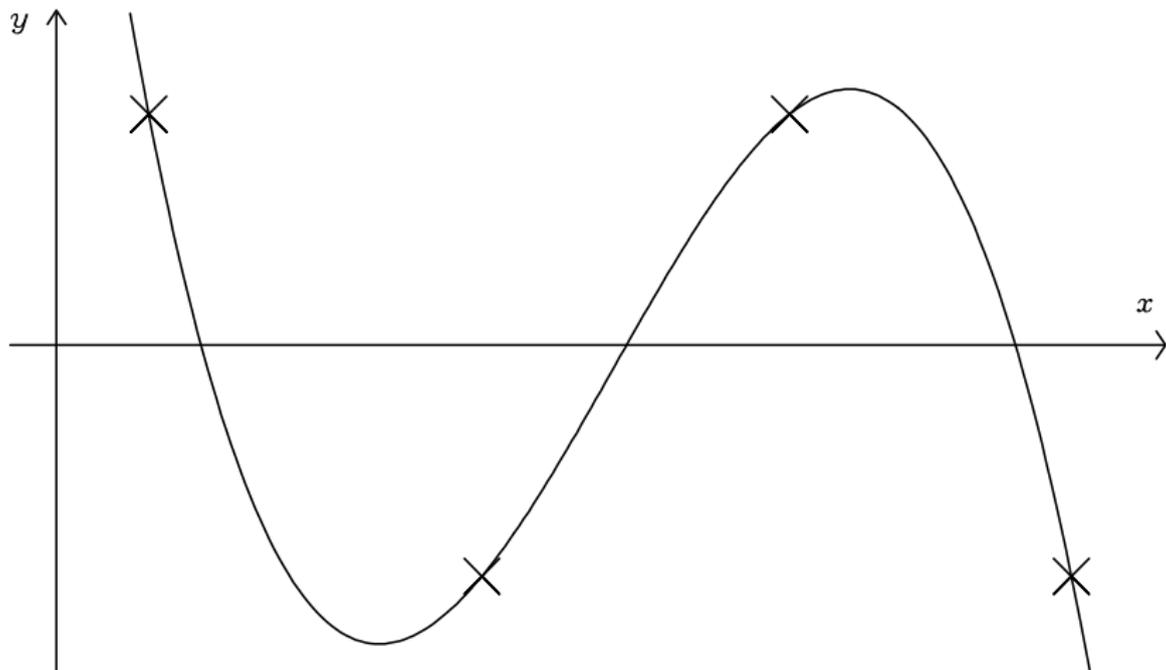
- 1 **Numerical viscosity:** introduce terms with higher order derivatives to damp oscillations
- 2 **Staggered grid:** different variables are defined on different grids
- 3 **Alternate grid:** keep variables on the same grid, but enforce differential equations on a different grid

# Remedies for mesh drift

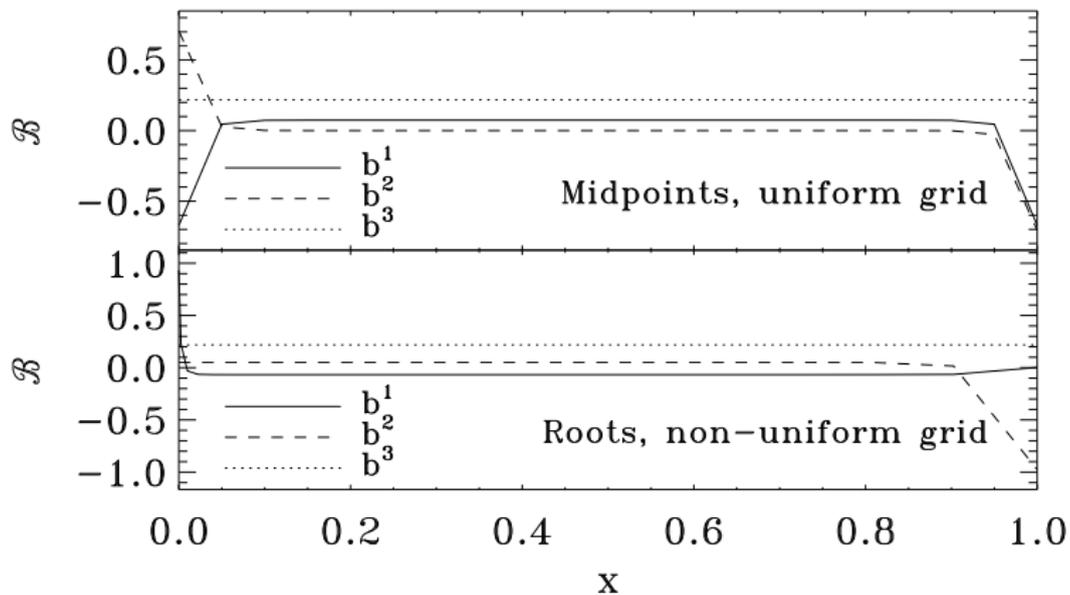
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# Remedies of mesh drift



# Remedies for mesh drift



# Remedies for mesh drift

The alternate grid strategy also lends itself to “superconvergence” ...

# Superconvergence

## Definition

- usually, finite-differences for  $n$  points are accurate to order  $n - d$ , where  $d$  is the order of the derivative
- in some cases, it is accurate to a higher order – this is called “superconvergence” (e.g. Sadiq & Viswanath 2013)

# Superconvergence

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- usually, finite-differences for  $n$  points are accurate to order  $n - d$ , where  $d$  is the order of the derivative
- in some cases, it is accurate to a higher order – this is called “superconvergence” (e.g. Sadiq & Viswanath 2013)

**Question:** is it possible to cause superconvergence by carefully choosing the point,  $z$ , where you calculate the finite differences?

# Superconvergence

**Answer:** yes!

We carry out a Taylor expansion around  $z$ :

$$f(x_1) = f(z) + \dots + f^{(n-1)} \frac{(x_1 - z)^{n-1}}{(n-1)!} + f^{(n)} \frac{(x_1 - z)^n}{n!} + \mathcal{O}(x_1^{n+1})$$

$$f(x_2) = f(z) + \dots + f^{(n-1)} \frac{(x_2 - z)^{n-1}}{(n-1)!} + f^{(n)} \frac{(x_2 - z)^n}{n!} + \mathcal{O}(x_2^{n+1})$$

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- the FD coefficients verify  $\sum_{i=1}^n a_i^d(z) f(x_i) = f^{(d)}(z) + R$ , where:

$$R = \underbrace{\sum_{i=1}^n a_i^d(z) f^{(n)}(z) \frac{(x_i - z)^n}{n!}}_{R_n(z)} + \sum_{i=1}^n a_i^d(z) \mathcal{O}((x_i - z)^{n+1})$$

- superconvergence if  $R_n(z) = 0$

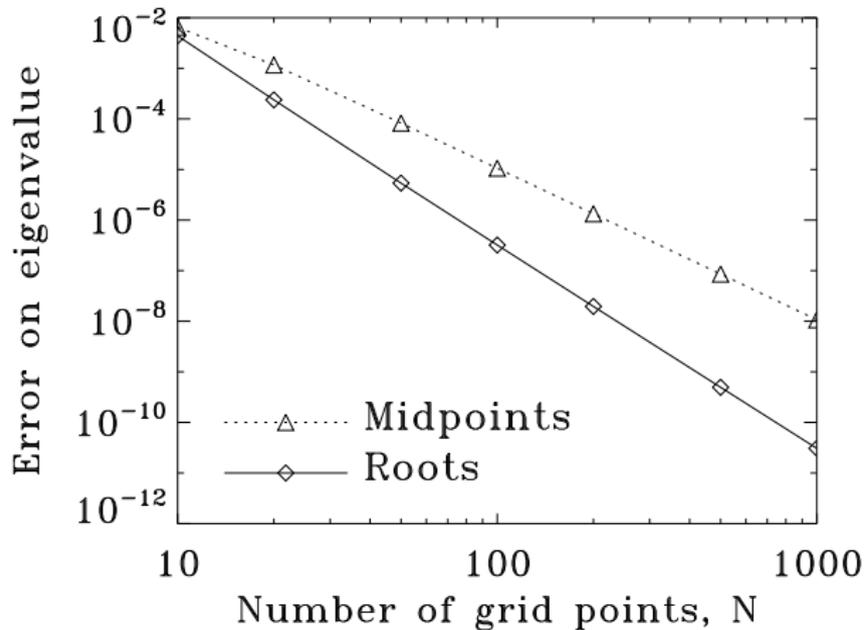
# Superconvergence

- it can be shown that (see [appendix](#)):

$$R_n(z) = -\frac{f^{(n)}(z)}{n!} Q^{(d)}(z) \text{ where } Q(z) = \prod_{i=1}^n (z - x_i)$$

- hence, by choosing one of the roots of  $Q^{(d)}(z)$ , superconvergence is achieved
- furthermore, mesh drift continues to be suppressed

# Superconvergence



# Conclusion

- suppression of mesh-drift through alternate grid strategy (i.e. the variables are defined on one grid, but the equations are enforced on another grid)
- superconvergence can be achieved systematically and combines nicely with the suppression of mesh-drift
- for more details, see Reese (2013, A&A 555, A148) and references therein

# Derivation of finite-difference coefficients

We carry out a Taylor expansion around  $z$ :

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$$\sum_{i=1}^n a_i^d(z) f(x_i) = f^{(d)}(z) + \sum_{i=1}^n a_i^d(z) \mathcal{O}(x_i^n)$$
- in other words,  $\sum_{i=1}^n a_i^d(z) \frac{(x_i - z)^k}{k!} = \delta_k^d$

# Derivation of finite-difference coefficients

In matrix form, this can be expressed as follows:

$$\begin{bmatrix} 1 & 1 & \dots & 1 \\ x_1 - z & x_2 - z & \dots & x_n - z \\ \vdots & \vdots & \ddots & \vdots \\ \frac{(x_1 - z)^{n-1}}{(n-1)!} & \frac{(x_2 - z)^{n-1}}{(n-1)!} & \dots & \frac{(x_n - z)^{n-1}}{(n-1)!} \end{bmatrix} \begin{bmatrix} a_1^0 & a_1^1 & \dots & a_1^{n-1} \\ a_2^0 & a_2^1 & \dots & a_2^{n-1} \\ \vdots & \vdots & \ddots & \vdots \\ a_n^0 & a_n^1 & \dots & a_n^{n-1} \end{bmatrix} = \mathcal{I}$$

where  $\mathcal{I}$  is the identity matrix.

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where  $\mathcal{I}$  is the identity matrix.

However, a matrix and its inverse commute:

$$\begin{bmatrix} a_1^0 & a_1^1 & \dots & a_1^{n-1} \\ a_2^0 & a_2^1 & \dots & a_2^{n-1} \\ \vdots & \vdots & \ddots & \vdots \\ a_n^0 & a_n^1 & \dots & a_n^{n-1} \end{bmatrix} \begin{bmatrix} 1 & 1 & \dots & 1 \\ x_1 - z & x_2 - z & \dots & x_n - z \\ \vdots & \vdots & \ddots & \vdots \\ \frac{(x_1 - z)^{n-1}}{(n-1)!} & \frac{(x_2 - z)^{n-1}}{(n-1)!} & \dots & \frac{(x_n - z)^{n-1}}{(n-1)!} \end{bmatrix} = \mathcal{I}$$

# Derivation of finite-difference coefficients

In algebraic form, this becomes:  $\sum_{k=0}^{n-1} a_i^k \frac{(x_j - z)^k}{k!} = \delta_i^j$

We introduce the polynomial:  $P(x) = \sum_{k=0}^{n-1} a_i^k \frac{(x-z)^k}{k!}$

From the above properties,  $\begin{cases} P(x_i) = 1 \\ P(x_j) = 0 \quad \forall j \neq i \end{cases}$

From this we recognise the Lagrange interpolation polynomial:

$$P(x) = L_i(x) = \frac{\prod_{j \neq i} (x - x_j)}{\prod_{j \neq i} (x_i - x_j)}$$

The  $a_i^d$  coefficients simply come from the Taylor expansion of  $P$  around  $z$ :

$$a_i^d = L_i^{(d)}(z)$$



# Proof for $R_n(z)$

We carry out a Taylor expansion around  $z$ :

$$f(x_1) = f(z) + \dots + f^{(n-1)} \frac{(x_1 - z)^{n-1}}{(n-1)!} + f^{(n)} \frac{(x_1 - z)^n}{n!} + \mathcal{O}(x_1^{n+1})$$

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- the FD coefficients verify  $\sum_{i=1}^n a_i^d(z) f(x_i) = f^{(d)}(z) + R$ , where:

$$R(z) = \underbrace{\sum_{i=1}^n a_i^d(z) f^{(n)}(z) \frac{(x_i - z)^n}{n!}}_{R_n(z)} + \sum_{i=1}^n a_i^d(z) \mathcal{O}\left((x_i - z)^{n+1}\right)$$

# Proof for $R_n(z)$

- we calculate  $R(z)$  for the function  $f_0(x) = x^n$
- one has:  $R = -f_0^{(d)}(z) + \sum_{i=1}^n a_i^d(z)x_i^n$
- furthermore,  $a_i^d(z) = L_i^{(d)}(z)$
- let us define the polynomial  $P(x) = -x^n + \sum_{i=1}^n L_i(x)x_i^n$
- it turns out that  $P^{(d)}(z) = R$
- furthermore:  $P(x_i) = -x_i^n + x_i^n = 0, \quad 1 \leq i \leq n$
- but,  $P$  is of degree  $n$  with a leading coefficient of  $-1$
- hence:  $P(x) = -\prod_{i=1}^n (x - x_i)$
- however:  $R = \sum_{i=1}^n a_i^d(z)f_0^{(n)}(z)\frac{(x_i - z)^n}{n!} = \sum_{i=1}^n a_i^d(z)(x_i - z)^n,$
- hence:

$$\sum_{i=1}^n a_i^d(z)(x_i - z)^n = P^{(d)}(z)$$

