

Ex 1.

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$$\frac{\partial u}{\partial t} = -\sin nx \frac{\partial u}{\partial x}$$

We seek a trigonometric polynomial,

$$u_N(x,t) = \sum_{|n| \leq \frac{N}{2}} a_n(t) e^{inx}$$

and require that the residual

$$R_N(x,t) = \frac{\partial u_N(x,t)}{\partial t} + \sin nx \frac{\partial u_N(x,t)}{\partial x}$$

is orthogonal to \hat{B}_N .

$$R_N(x,t) = \sum_{|n| \leq \frac{N}{2}} \left(\frac{da_n(t)}{dt} + \frac{e^{ix} - e^{-ix}}{2i} (in) a_n(t) \right) e^{inx}$$

$$R_N(x,t) = \sum_{|n| \leq \frac{N}{2}} \frac{da_n(t)}{dt} e^{inx} + \frac{1}{2} \sum_{|n| \leq \frac{N}{2}} n e^{i(n+1)x} a_n(t) - \frac{1}{2} \sum_{|n| \leq \frac{N}{2}} n e^{i(n-1)x} a_n(t)$$

$$= \sum_{|n| \leq \frac{N}{2}} \frac{da_n(t)}{dt} e^{inx} + \frac{1}{2} \sum_{|n| \leq \frac{N}{2}} (n+1) e^{inx} a_{n+1}(t) - \frac{1}{2} \sum_{|n| \leq \frac{N}{2}} (n-1) e^{inx} a_{n-1}(t) \\ + \frac{N}{4} (e^{i\frac{N+2}{2}x} a_{\frac{N}{2}}(t) + e^{-i\frac{N+2}{2}x} a_{-\frac{N}{2}}(t))$$

where $a_{-(\frac{N}{2}+1)}(t) = a_{\frac{N}{2}+1}(t) = 0$.

Finally we get, for $|n| \leq \frac{N}{2}$

$$\frac{da_n(t)}{dt} + \frac{n+1}{2} a_{n+1}(t) - \frac{n-1}{2} a_{n-1}(t) = 0$$

with $a_{-(\frac{N}{2}+1)}(t) = a_{\frac{N}{2}+1}(t) = 0$.



We notice that

$$P_N \mathcal{L} \neq \mathcal{L} P_N,$$

and so the truncation error

$$P_N \mathcal{L} (I - P_N) u \neq 0.$$

In this case $a_n(t)$ are not equal to Fourier coefficients \hat{u}_n

$$\text{So } P_N u \neq U_N(x, t).$$

Ex 2. We could assume test functions are defined as

$$\psi_n(x) = \frac{2}{\pi} \sin(nx).$$

$$\psi_n(0) = \psi_n(\pi) = 0 \text{ for all } n.$$

We seek an approximation

$$U_N(x, t) = \sum_{n=0}^N \hat{u}_n(t) \sin(nx)$$

and require that the residual

$$R_N(x, t) = \frac{\partial U_N(x, t)}{\partial t} + \sin(x) \frac{\partial U_N(x, t)}{\partial x}$$

is orthogonal to the test space.

$$R_N(x, t) = \sum_{n=0}^N \frac{\partial \hat{u}_n(t)}{\partial t} \sin(nx) + \sin(x) \sum_{n=0}^N n \hat{u}_n(t) \cos(nx)$$

$$\cancel{R_N(x, t)} = \sum_{n=0}^N \frac{\partial \hat{u}_n(t)}{\partial t} \sin(nx) + \sum_{n=0}^N n \hat{u}_n(t) \frac{\sin((n+1)x) - \sin((n-1)x)}{2}.$$

By defining $\hat{u}_{-1}(t) = \hat{u}_{N+1}(t) = 0$, we could get



$$\begin{aligned}
 R_N &= \sum_{n=0}^N \frac{\partial \hat{u}_n(t)}{\partial t} \sin(nx) + \sum_{n=0}^N (n-1) \hat{u}_{n-1}(t) \frac{\sin(nx)}{2} + \frac{\hat{u}_N(t)}{2} N \sin((N+1)x) \\
 &\quad - \sum_{n=0}^N (n+1) \hat{u}_{n+1}(t) \frac{\sin(nx)}{2} - \frac{\hat{u}_0(t)}{2} \cdot 0 \sin(-x) \\
 &= \sum_{n=0}^N \left(\frac{\partial \hat{u}_n(t)}{\partial t} + \frac{n-1}{2} \hat{u}_{n-1}(t) - \frac{n+1}{2} \hat{u}_{n+1}(t) \right) \sin(nx) + \text{truncation error}
 \end{aligned}$$

Finally we obtain, for $0 \leq n \leq N$

$$\frac{\partial \hat{u}_n(t)}{\partial t} + \frac{n-1}{2} \hat{u}_{n-1}(t) - \frac{n+1}{2} \hat{u}_{n+1}(t) = 0$$

with $\hat{u}_{-1}(t) = \hat{u}_{N+1}(t) = 0$.

Ex 3.

① Assume $\psi_n = \frac{2}{\pi} \cos(nx)$

② Assume $u_N(x, t) = \sum_{n=0}^{N+2} \hat{u}_n(t) \cos(nx)$

③ $(R_N, \psi_n) = 0$ for $0 \leq n \leq N+2$

$$R_N = \sum_{n=0}^{N+2} \left(\frac{\partial \hat{u}_n(t)}{\partial t} - \frac{n+1}{2} \hat{u}_{n+1}(t) + \frac{n-1}{2} \hat{u}_{n-1}(t) \right) \cos(nx) + \text{truncation error}$$

with $\hat{u}_{-1}(t) = \hat{u}_{N+3}(t) = 0$

$$\Rightarrow \frac{\partial \hat{u}_n(t)}{\partial t} - \frac{n+1}{2} \hat{u}_{n+1}(t) + \frac{n-1}{2} \hat{u}_{n-1}(t) = 0$$

④
$$\begin{cases} \sum_{n=0}^{N+2} \hat{u}_n(t) = 0 \\ \sum_{n=0}^{N+2} (-1)^n \hat{u}_n(t) = 0 \end{cases}$$



Ex 4.

Burgers equation could be rewritten as

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} - \varepsilon \frac{\partial^2 u}{\partial x^2} = 0$$

We seek solution of the form:

$$u_N(x,t) = \sum_{|n| \leq \frac{N}{2}} a_n(t) e^{inx} = \sum_{j=0}^{N-1} u_N(x_j, t) g_j(x)$$

$$R_N(x,t) = \frac{\partial u_N(x,t)}{\partial t} + u_N(x,t) \frac{\partial u_N(x,t)}{\partial x} - \varepsilon \frac{\partial^2 u_N(x,t)}{\partial x^2}$$

Vanishes at a specified set of grids y_j , in this case.

We assume $y_j = x_j$

this results in:

$$\frac{d u_N(x_j, t)}{dt} + \sum_{k=0}^{N-1} (u_N(x_j, t) D_{jk}^{(1)} - \varepsilon D_{jk}^{(2)}) u_N(x_k, t)$$

$D^{(1)}$ and $D^{(2)}$ are the differentiation matrices.

