

NETWORK-NORM ERROR ESTIMATION USING INTERPOLATION OF SPACES
 AND APPLICATION TO DIFFERENTIAL EQUATIONS

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In order to illustrate the main results of this paper (theorems 1 and 2), the following Cauchy boundary problem for evolutionary equations is considered:

$$(1) \begin{cases} \frac{\partial}{\partial t} u(x, t) - H\left(\frac{\partial}{\partial x}, x, t\right) u(x, t) = 0 \\ x \in (-\infty, \infty) = \mathbb{R}, \quad 0 < t \leq T \\ u(x, 0) = f(x), \quad x \in \mathbb{R}, \quad f \in L_p(\mathbb{R}) \quad (1 \leq p \leq \infty). \end{cases}$$

Here $L_p(\mathbb{R})$ is defined as usually. The differential operator H is supposed to allow an unique representation of the solution by a solution operator denoted by G : $u(x, t) = [G(t)f](x)$.

Problem (1) is being solved numerically by means of problem (2):

$$(2) \begin{cases} u_h(x, t+d) = \sum_{\alpha \in I} c_\alpha u_h(x+\alpha h, t), \quad I\text{-finite}, \quad h \leq 1 \text{ without l.o.g.}, \\ u_h(x, 0) = f(x) \\ x \text{ and } t \text{ belong to the uniform networks } \Sigma_h \text{ and } \Omega_h, \text{ resp.:} \\ x \in \Sigma_h = \{x_\mu : x_\mu = \mu h, \mu = 0, \pm 1, \pm 2, \dots\}, \\ t \in \Omega_h = \{t_\nu : t_\nu = \nu d; \nu = 0, 1, \dots, N, \quad Nd = T\}, \end{cases}$$

c_α and I may depend on x and t . The solution operator of (2) is $G_h(t) : u_h(x, t) = [G_h(t)f](x)$. The error operator is then $E_h = G_h - G$. All these operators may depend on x , as well as on t .

Many works deal with obtaining estimates of the error $u_h - u$. These are of two types: i) discrete (network) - norm estimates requiring: a) stability, b) approximation and c) f has derivatives of some order (v. [6]); ii) integral L_p - norm error estimates using a) and b) only (these require information about the initial value f a.e. on R instead of \sum_h only - v. [1], [2]).

The purpose of the present work is to derive discrete-norm error estimates which, like the integral-norm ones, require conditions a) and b) only, i.e. without the assumption of additional smoothness of f . In order to obtain these estimates, a method is developed (v. also [5]), which is quite general and can be applied to a wide variety of error estimation problems not necessarily related to differential equations. Here applications are restricted to the model evolutionary problem (1) (2) with concrete simple differential operators H only. The technique of estimation includes:

A. Average moduli of smoothness: $\tau_k(f; \delta)_{L_p} = \|\omega_k(f, x; \delta)\|_{L_p}$ where $\omega_k(f, x; \delta) = \sup \left\{ \left| \Delta_h^k f(t) \right| : t, t+kh \in \left[x-k\delta/2, x+k\delta/2 \right] \right\}$, where $\Delta_h^k f(t) = \sum_{m=0}^k (-1)^{k+m} \binom{k}{m} f(t+mh)$, $f \in L_p(R)$, bounded, $\delta > 0$, $0 < h \leq \delta$.

(For history, properties and applications of these moduli v. e.g. [4], [7]). Here and henceforth every function $f \in L_p$ is distinguished from its class of equivalence and considered defined by a concrete value in every point of R . The following space is to be dealt with:

$$A_p(R) = \left\{ f : \|f\|_{A_{p,h}} = \|f\|_{L_p} + \tau_1(f; h)_{L_p} < \infty, h > 0 \right\}.$$

For every $h_1, h_2 > 0$ the norms $\|\cdot\|_{A_{p,h_1}}$, $\|\cdot\|_{A_{p,h_2}}$ are equivalent on A_p (with constants dependent on h_1, h_2).

B. Besov spaces B_{pq}^s , interpolation and embedding results.

C. A-spaces (v. [4]), defined by ($1 \leq p, q \leq \infty$, $0 < s \leq r$)

$$A_{pq}^s(R) = \left\{ f : \|f\|_{A_{pq}^s} = \|f\|_{L_p} + \left(\int_0^\infty (t^{-s} \tau_r(f; t)_{L_p})^q \frac{dt}{t} \right)^{1/q} < \infty \right\}$$

Note that for $\frac{1}{p} < s \leq r$ $A_{pq}^s = B_{pq}^s$ and for $0 < s < \frac{1}{p}$ $A_{pq}^s \subsetneq B_{pq}^s$.

D. Spaces M_p of Fourier multipliers on L_p (v. e.g. [1]).

The discrete-norm spaces to be used are the following:

$$l_h^p(\sum_h) = \left\{ f : R \rightarrow R, \|f\|_{l_h^p(\sum_h)} < \infty \right\}, \text{ where}$$

$$\|f\|_{1_h^p(\Sigma_h)} = \left(\sum_{\mu=-\infty}^{\infty} h |f(x_\mu)|^p \right)^{1/p}, \quad x_\mu \in \Sigma_h, \quad 1 \leq p \leq \infty.$$

$$\tilde{1}_h^p(\Sigma_h) = \left\{ f : \mathbb{R} \rightarrow \mathbb{R}, \|f\|_{1_h^p(\Sigma_h)} \right\}, \quad \text{where}$$

$$\|f\|_{\tilde{1}_h^p(\Sigma_h)} = \left(\sum_{\mu=-\infty}^{\infty} h \sup \left\{ |f(x)|^p, x \in \left[x_\mu - \frac{h}{2}, x_\mu + \frac{h}{2} \right] \right\} \right)^{1/p}.$$

It can be proved that for every $\xi \in \mathbb{R}$ $\tilde{1}_h^p(\Sigma_{h+\xi}) = \tilde{1}_h^p(\Sigma_h) = A_{p,h}(\mathbb{R})$ (equivalence of norms with constants independent of h). With respect to the applications it should be mentioned here that for some important classes of operators, stability in L_p implies stability: $A_p \rightarrow 1_h^p(\Sigma_h)$. (Such as, for instance, all difference operators G_h appearing in (2) or convolution operators $Ef = \check{\xi} * f$, with $\check{\xi}$ - Fourier multiplier on L_p , $\check{\xi} = F^{-1}(\xi)$ - inverse Fourier transform of ξ , and $\xi \in S$ (Schwartz' space) - v. Appl. 1).

The next theorem is an useful generalization of a theorem of V. Popov (v. [3], [4], [7]).

Theorem 1. (v. also [5]). Let E be a Lipschitz operator: $A_p \rightarrow 1_h^p(\Sigma_h)$ with:

$$(A) \text{ For every } f, g \in A_p \quad \|Ef - Eg\|_{1_h^p(\Sigma_h)} \leq c_1 \|f - g\|_{A_{p,h}^{\sigma/r}}$$

(B) For every $f \in W_p^r$ (Sobolev space, i.e. $f, f^{(r)} \in L_p$, $f^{(r-1)} \in AC$ - the space of all absolutely continuous functions) it is fulfilled that

$$\|Ef\|_{1_h^p(\Sigma_h)} = c_2 h^\sigma (\|f\|_{L_p} + \|f^{(r)}\|_{L_p}), \quad r > 0,$$

where $0 < \sigma \leq r$; c_1, c_2 do not depend on h ; $h \leq 1$ (without l.o.g.).

Then, for every $f \in A_p$, $0 < s \leq r$,

$$\begin{aligned} \|Ef\|_{1_h^p(\Sigma_h)} &= c(r, c_1, c_2) (h^\sigma \|f\|_{L_p} + \mathcal{C}_r(f; h^{\frac{\sigma}{r}})_{L_p}) \leq \\ &\leq c(r, c_1, c_2) h^{\frac{\sigma s}{r}} \|f\|_{A_{p,\infty}^s}. \end{aligned}$$

When applied to some error operator E (usually linear bounded), this theorem yields a network-norm interpolation estimate. It may happen, however, that the 1_h^p -norm estimate of condition (B) is not available, while there is an analogous one in L_p -norm. In this case the next theorem may prove useful:

Theorem 2. Let E be a Lipschitz operator: $A_p \rightarrow 1_h^p(\Sigma_h)$, with:

$$(A) \text{ For every } f, g \in A_p, \quad \|Ef - Eg\|_{1_h^p(\Sigma_h)} = c_1 \|f - g\|_{A_{p,h}^{M(r)/r}} ;$$

(B) For every $f \in W_p^r$, $r \geq 1$,

$$\|Ef\|_{L_p} = c_2 h^{\sigma_r} (\|f\|_{L_p} + \|f^{(r)}\|_{L_p}).$$

(C) For every $f \in W_p^{r-1}$,

$$\|Ef\|_{L_p} \leq c_3 h^{\sigma_{r-1}} (\|f\|_{L_p} + \|f^{(r-1)}\|_{L_p}), \text{ where:}$$

$0 < \sigma_r, \sigma_{r-1} \leq \sigma_r$; c_1, c_2, c_3 do not depend on h ; $h \leq 1$ (without l.o.g.).

(D) For every $f \in W_p^1$, $Ef \in AC$ and E commutes with $\frac{\partial}{\partial x}$.

Then, for every $f \in A_p$, $0 < s \leq r$,

$$\|Ef\|_{1_h^p(\Sigma_h)} = c(r, c_1, c_2, c_3) h^{\frac{s\mu(r)}{r}} \|f\|_{A_{p\infty}^s},$$

where $\mu(r) = \min \{ \sigma_r, \sigma_{r-1} + 1 \}$.

Both these theorems hold also for a bounded interval instead of R and can be modified for a non-uniform network as well. They are the main technique (especially Theorem 2) in the following applications:

Application 1. Parabolic equation of arbitrary order (see [1]). For denotations, see (1), (2). In this case $H(\frac{\partial}{\partial x}, x, t) = \frac{\partial^{2m}}{\partial x^{2m}}$, $m > 0$ integer. (We may consider $H(\xi) = \xi^{2m}$).

The operators G and G_h are defined on the network by the ^{con}volutions with Fourier multipliers:

$$G(\nu d)f = F^{-1} (\exp(-\nu dH(\cdot)))_* f$$

$$(3) \quad G_h(\nu d)f = F^{-1} (\exp(-\nu dH_h(\cdot)))_* f, \text{ where}$$

$$H_h(\xi) = -\frac{\ln e(h\xi)}{d}, \text{ where}$$

$$e(h\xi) = \sum_{\alpha \in I} c_\alpha e^{i\alpha h\xi}; \quad c_\alpha, I - \text{constants (see (2)) and}$$

$$F(G_h(\nu d)f)(\xi) = [e(h\xi)]^\nu \widehat{f}(\xi); \quad F(f) = \widehat{f} - \text{Fourier transform of } f.$$

The next theorem is an integral-norm estimate for this problem:

Theorem A. (v. [1]). Let the following be fulfilled:

(A) $G_h(t)$ - stable in L_p (this contains the case of $\sum_{\alpha \in I} |c_\alpha| \leq \leq 1 + cd, c > 0$),

(B) $H_h(\xi)$ approximates $H(\xi)$ of the order $r > 0$ (for the definition of this approximation, see [1]). ($d = kh^{2m}$).

Then, $\|E_h(t)f\|_{L_p(\mathbb{R})} \leq ch^s \|f\|_{B_{p\infty}^s(\mathbb{R})}$,

$t = \nu d = \nu kh^{2m}, \nu = 0, 1, 2, \dots, N; 0 < s \leq r$.

Thence, Theorem 2 and some lemmas imply:

Theorem 3. Let the assumptions of theorem A hold, with $f \in A_p$:

Then, $\|E_h(t)f\|_{1_h^p(\sum_h)} \leq ch^s \|f\|_{A_{p\infty}^s(\mathbb{R})}$.

Corollary 1.: If for this problem $f^{(s)} \in L_p(\mathbb{R}), 1 \leq s \leq r$, then,

$$\|E_h(t)f\|_{1_h^p(\sum_h)} = O(h^s \omega_{r-s}(f^{(s)}; h)_{L_p(\mathbb{R})} + h^r).$$

Here the integral modulus of smoothness $\omega_k(f; h)_{L_p}$ is defined as usually.

Corollary 2. If $Vf^{(s-1)} < \infty$ (the variation of f), $1 \leq s \leq r$, then

$$\|Eh(t)f\|_{1_h^p(\sum_h)} = O(h^{s-1+1/p}), \quad 1 \leq p \leq \infty.$$

Application 2. Hyperbolic equation of the first order (v. [2]).

Using denotations (1), (2), (3), we have $H(-\frac{\partial}{\partial x}, x, t) = \frac{\partial}{\partial x}$, i.e. $H(\xi) = \xi$. For this problem the following L_p -norm estimate holds:

Theorem B. (v. [2]). Let conditions (A) and (B) of Theorem A be fulfilled for $G(t)$ with $H(\xi) = \xi$, $d = kh$, and a corresponding $G_h(t)$.

Then, $\|E_h(t)f\|_{L_p(\mathbb{R})} \leq ch^{\frac{sr}{r+1}} \|f\|_{B_{p\infty}^s(\mathbb{R})}, 0 < s < r+1$.

Our corresponding discrete-norm estimate is given by:

Theorem 4. Let the assumptions of theorem B. hold, with $f \in A_p$:

Then, $\|E_h(t)f\|_{1_h^p(\sum_h)} = ch^{\frac{sr}{r+1}} \|f\|_{A_{p\infty}^s(\mathbb{R})}$.

Corollary 3. If $f^{(s)} \in L_p(\mathbb{R})$, $1 \leq s < r+1$, then

$$\| E_h(t)f \|_{1_h^p(\Sigma_h)} = O(h^{\frac{sr}{r+1}} \omega_{r+1-s}(f^{(s)}; h^{\frac{r}{r+1}})_{L_p(\mathbb{R})} + h^r) .$$

Corollary 4. If $\forall f^{(s)} < \infty$, $0 \leq s \leq r$, then

$$\| E_h(t)f \|_{1_h^p(\Sigma_h)} = O(h^{\frac{(s+1)r}{r+1} - 1 + \frac{1}{p}}), \quad 1 \leq p \leq \infty .$$

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