TOTAL VARIATION AND ERROR ESTIMATES
FOR SPECTRAL VISCOSITY APPROXIMATIONS

EITAN TADMOR

Dedicated with appreciation to Heinz-Otto Kreiss on his 60th birthday

ABSTRACT. We study the behavior of spectral viscosity approximations to non-linear scalar conservation laws. We show how the spectral viscosity method compromises between the total-variation bounded viscosity approximations—which are restricted to first-order accuracy—and the spectrally accurate, yet unstable, Fourier method. In particular, we prove that the spectral viscosity method is $L^1$-stable and hence total-variation bounded. Moreover, the spectral viscosity solutions are shown to be Lip$^*$-stable, in agreement with Oleinik's E-entropy condition. This essentially nonoscillatory behavior of the spectral viscosity method implies convergence to the exact entropy solution, and we provide convergence rate estimates of both global and local types.

1. THE SPECTRAL VISCOSITY APPROXIMATION

We are concerned here with spectral approximations of the scalar conservation law

\begin{equation}
\frac{\partial}{\partial t} u(x, t) + \frac{\partial}{\partial x} f(u(x, t)) = 0, \quad u(x, 0) = u_0(x) \in BV.
\end{equation}

To single out a unique physically relevant weak solution, (1.1a) is complemented with an entropy condition such that for all convex $U$'s (e.g., [7, 12])

\begin{equation}
\frac{\partial}{\partial t} U(u) + \frac{\partial}{\partial x} F(u) \leq 0, \quad F(u) \equiv \int U'(\xi) f'(\xi) d\xi.
\end{equation}

We want to solve the $2\pi$-periodic initial value problem (1.1a)–(1.1b) by spectral methods. To this end, we use an $N$-trigonometric polynomial, $u_N(x, t) = \sum_{k=-N}^{N} \hat{u}_k(t) e^{ikx}$, to approximate the spectral (or pseudospectral) projection of the exact entropy solution, $P_N u$. Starting with $u_N(x, 0) = P_N u_0(x)$, the standard Fourier method reads, e.g., [5, 2, 1],

\begin{equation}
\frac{\partial}{\partial t} u_N + \frac{\partial}{\partial x} P_N f(u_N) = 0.
\end{equation}

Together with one's favorite ODE solver, (1.2) gives a fully discrete spectral method for the approximate solution of (1.1a).
Although the spectral method (1.2) is a spectrally accurate approximation of the conservation law (1.1a), in the sense that its local error does not exceed
\[ \| (I - P_N) f (u_N (\cdot, t)) \|_{H^{-s}} \leq \text{Const} \cdot N^{-s} \| u_N \|_{L^2}, \quad \forall s \geq 0, \]
the spectral solution, \( u_N (x, t) \), need not approximate the corresponding entropy solution, \( u(x, t) \). Indeed, counterexamples provided in [13, 14] show that the spectral approximation (1.2) lacks entropy dissipation, which is inconsistent with the entropy condition (1.1b). Consequently, the spectral approximation (1.2) supports spurious Gibbs oscillations which prevent strong convergence to the exact solution of (1.1).

To suppress these oscillations, without sacrificing the overall spectral accuracy, we consider instead the Spectral Viscosity (SV) approximation
\[ \frac{\partial}{\partial t} u_N (x, t) + \frac{\partial}{\partial x} P_N f (u_N (x, t)) = \varepsilon_N \frac{\partial}{\partial x} Q_N \ast \frac{\partial}{\partial x} u_N (x, t). \]

The left-hand side of (1.4) is the standard approximation of (1.1a). On the right-hand side, it is augmented by spectral viscosity which consists of the following three ingredients: a vanishing viscosity amplitude of size \( \varepsilon_N \downarrow 0 \), a viscosity-free spectrum of size \( m_N \gg 1 \), and a viscosity kernel, \( Q_N (x, t) = \sum_{|k|=m_N} \hat{\Omega}_k (t) e^{i k x} \), activated only on high wave numbers \( |k| \geq m_N \), which can be conveniently implemented in the Fourier space as
\[ \varepsilon_N \frac{\partial}{\partial x} Q_N \ast \frac{\partial}{\partial x} u_N (x, t) \equiv -\varepsilon_N \sum_{|k|=m_N} k^2 \hat{\Omega}_k (t) \hat{u}_k (t) e^{i k x}. \]

We deal with real viscosity kernels \( Q_N (x, t) \) with increasing Fourier coefficients, \( \hat{\Omega}_k \equiv \hat{\Omega}_{|k|} \), which satisfy
\[ 1 - \left( \frac{m_N}{|k|} \right)^{2q} \leq \hat{\Omega}_k (t) \leq 1, \quad |k| \geq m_N, \quad \text{for some fixed } q \geq 1, \]
and we let the spectral viscosity parameters, \( (\varepsilon_N, m_N) \), lie in the range
\[ \varepsilon_N \sim \frac{1}{N^{\theta} \log N}, \quad m_N \sim N^{\frac{q}{2}}, \quad \theta < 1. \]

We remark that this choice of spectral viscosity parameters is small enough to retain the formal spectral accuracy of the overall approximation, since
\[ \left\| \varepsilon_N \frac{\partial}{\partial x} Q_N \ast \frac{\partial}{\partial x} u_N (\cdot, t) \right\|_{H^{-s}} \leq \text{Const} \cdot N^{-\frac{s q}{2}} \| u_N (\cdot, t) \|_{L^2}, \quad \forall s \geq 2. \]

At the same time, it is sufficiently large to enforce the correct amount of entropy dissipation that is missing otherwise, when either \( \varepsilon_N = 0 \) or \( m_N = N \). Indeed, it was shown in [13]–[15], [8] that the SV approximation (1.4)–(1.6)q has a bounded entropy production in the sense that
\[ \varepsilon_N \left\| \frac{\partial}{\partial x} u_N (x, t) \right\|_{L^2_{\text{loc}} (x, t)}^2 \leq \text{Const}, \]
and this together with an \( L^\infty \)-bound imply—by compensated compactness arguments—that the SV approximation \( u_N \) converges to the unique entropy solution of (1.1).
Observe that in the limit case \( q = \infty \), the SV method \((1.4), (1.5)_{\infty} - (1.6)_{\infty}\), coincides with the usual viscosity approximation, \( \frac{\partial}{\partial t} u_\epsilon(x, t) + \frac{\partial}{\partial x} P_N f(u_\epsilon(x, t)) = \varepsilon_N \frac{\partial^2}{\partial x^2} u_\epsilon(x, t) \). But of course, the spectral accuracy \((1.7)\) is lost in this limit case.

In this paper we show that the SV method \((1.4) - (1.6)_{q}\), while maintaining the spectral accuracy \((1.7)\), also shares the essentially nonoscillatory behavior of standard viscosity approximations. In particular, in §3 we show that the SV solution is total-variation bounded. Moreover, in the genuinely nonlinear case, \( f'' > 0 \), the SV solution is Lip\(^+\)-stable, in agreement with Oleinik’s E-condition. We conclude that the SV approximation converges to the exact entropy solution of \((1.1)\), and we provide various error estimates.

2. A total-variation bound

The presence of spectral viscosity on the right of \((1.4)\) is responsible for a rapid decay of the Fourier coefficients located toward the end of the computed spectrum. This spectral decay result was proved in [8] for the special case of Burgers’ equation, \( f(u) = \frac{1}{2} u^2 \), following the argument of [4]. The general case was analyzed by S. Schochet, [11], where it was shown that the following spectral decay estimate holds [11, Theorem 1]:

\[
\| (I - P_k) f(u_N(\cdot, t)) \| \leq K_s \left( \frac{k}{N} \right)^{-s} \cdot \left[ N^{-(s(1-\theta))} + k^{-r} e^{-e_N N^s} \right], \quad \forall k \geq N, \quad 0 < t \leq T.
\]

Here, \( r \) and \( s \) are related to the smoothness of the data—the initial data \( u_N(\cdot, 0) \) and the flux \( f(\cdot) \); \( r \geq 0 \) is related to the smoothness of the initial data, \( u_0 \)—the initial smoothness being measured by the requirement that

\[
\max_{k < N} k^r \| (I - P_k) u_N(\cdot, 0) \| \leq \text{Const} ;
\]

and \( s \) is any sufficiently large integer, \( s \geq s_0(r) \), which is related to the degree of smoothness of \( f(\cdot) \) measured by the constants \( K_s \)—constants which may depend on \( \| f \|_{C^r} \) (as well as \( \| u_N \|_{L^\infty} \) and \( \theta \)), but otherwise are independent of \( N \).

The last estimate shows that the discretization error as well as its spatial derivatives, \( \partial_k^p (I - P_N) f(u_N(\cdot, t)) \), become spectrally small independently of whether the underlying entropy solution is smooth or not. Indeed, using the dyadic decomposition \( \partial_k^p (I - P_N) f(u_N) = \sum_{j=0}^{\infty} \partial_k^p P_{2^{j+1}N} (I - P_{2^jN}) f(u_N) \) and applying the above estimate with \( k = 2^j N, \quad j = 0, 1, \ldots \), we obtain (consult [11, (4.9)])

\[
(2.1)_p \quad \left\| \frac{\partial^p}{\partial x^p} (I - P_N) f(u_N(\cdot, t)) \right\| \leq K_s \cdot \left[ N^{-(s(1-\theta)+p)} + N^{-r+p} e^{-e_N N^{2-s}} \right], \quad \forall s \geq s_0.
\]

Remark. As noted in [11, §3], the above smoothness requirements are by no means optimal. For the sake of technical convenience, we therefore assume throughout the rest of the paper that the flux, \( f(\cdot) \), is sufficiently smooth (e.g., \( K_s < \infty \) for \( s \) large enough so that the first term on the right in \((2.1)_p\) is negligible for, say, \( s(1-\theta) > 2 \)). Observe that then the spatial derivatives
of the truncation error are spectrally small, provided the initial data \( u_N(\cdot, 0) \) are sufficiently smooth so that the second term on the right of (2.1)_p applies with \( r > p \) (otherwise, an initial layer of size \( < \frac{1}{N} \) may be formed, which is smoothed out exponentially fast once the spectral viscosity becomes effective).

We conclude that the SV approximation is governed by the viscosity-like equation
\[
(2.2) \quad \frac{\partial}{\partial t} u_N(x, t) + \frac{\partial}{\partial x} f(u_N(x, t)) = \varepsilon_N \frac{\partial}{\partial x} \mathcal{Q}_N \ast \frac{\partial}{\partial x} u_N(x, t) + \ldots ,
\]
where the missing term ... on the right refers to the spectrally small discretization error (2.1)_₁.

Equation (2.2) is similar to the usual viscosity approximation
\[
(2.3) \quad \frac{\partial}{\partial t} u_e(x, t) + \frac{\partial}{\partial x} f(u_e(x, t)) = \varepsilon \frac{\partial^2}{\partial x^2} u_e(x, t).
\]
(In fact, the SV method (1.4)−(1.6)_q coincides with the viscosity equation (2.3) in the limit case \( q = \infty \).) To quantify this similarity, we rewrite (2.2) in the equivalent form
\[
(2.4a) \quad \frac{\partial}{\partial t} u_N(x, t) + \frac{\partial}{\partial x} f(u_N(x, t)) = \varepsilon_N \frac{\partial^2}{\partial x^2} u_N(x, t) - \varepsilon_N \frac{\partial}{\partial x} R_N(x, t) \ast \frac{\partial}{\partial x} u_N(x, t)
\]
\[
= \frac{\partial^2}{\partial x^2} u_N(x, t) - \varepsilon_N \frac{\partial}{\partial x} \mathcal{R}_N(x, t) \ast \frac{\partial}{\partial x} u_N(x, t)
\]
\[
= \frac{\partial}{\partial x} (I - P_N) f(u_N),
\]
where
\[
(2.4b) \quad \mathcal{R}_N(x, t) = \sum_{k=\pm N} \mathcal{R}_k(t) e^{ikx}, \quad \hat{\mathcal{R}}_k(t) = \left\{ \begin{array}{ll}
1, & |k| < m_N, \\
1 - \hat{Q}_k(t), & |k| \geq m_N.
\end{array} \right.
\]

Apart from the spectrally small truncation error on the right, the SV approximation (2.4a) differs from (2.3) by the additional term involving the 'residual kernel', \( R_N(x, t) \), on its right-hand side. We claim that this kernel is 'sufficiently small'.

**Lemma 2.1.** Consider the SV kernel \( \mathcal{Q}_N(x, t) \) subject to the SV parameterization (1.5)_q−(1.6)_q. Then \( \mathcal{R}_N(x, t) \equiv \mathcal{D}_N(x) - \mathcal{Q}_N(x, t) \) satisfies
\[
(2.5) \quad \left\| \frac{\partial^{2s}}{\partial x^{2s}} \mathcal{R}_N(\cdot, t) \right\|_{L^1} \leq \text{Const} \cdot m_N^{2s} \log N, \quad 0 \leq s \leq q.
\]

**Remark.** The inequality (2.5)_₁ followed by (1.6)_q imply the bound
\[
\varepsilon_N \| \frac{\partial^2}{\partial x^2} \mathcal{R}_N(\cdot, t) \|_{L^1} \leq \text{Const},
\]
which plays an essential role in our foregoing discussion. In practice it was found that the latter bound is minimized if we let the monotonically increasing (respectively decreasing) SV coefficients, \( \mathcal{Q}_k \) (respectively \( \hat{\mathcal{R}}_k \)), to depend smoothly on the relative wave number \( \frac{k}{N} \).

**Proof of Lemma 2.1.** We first recall

[8, Lemma A.1]: For any symmetric \( N \)-trigonometric polynomial, \( u_N(x) = \sum_{k=0}^N \hat{u}_k \cos kx \), with monotonically decreasing coefficients, \( 0 \leq \hat{u}_k \leq 1 \), there holds
\[ \|w_N(x)\|_{L^1} \leq \text{Const} \cdot \log N. \]

(Consideration of the Dirichlet kernel, \(D_N(x) \equiv 2 \sum_{k=0}^{N} e^{ikx}\), shows that the last estimate is sharp.)

Consider now the symmetric \(N\)-trigonometric polynomial

\[ \frac{1}{m_N^{2s}} \frac{\partial^{2s}}{\partial x^{2s}} R_N(x, t) = 2 \sum_{k=0}^{N} \frac{k^{2s}}{m_N^{2s}} \hat{R}_k \cos kx, \quad s \leq q. \]

According to the SV parameterization in (1.5)\(_q\), it has monotonically decreasing Fourier coefficients which satisfy (for \(s \leq q\)) \(\frac{k^{2s}}{m_N^{2s}} \hat{R}_k \equiv \frac{k^{2s}}{m_N^{2s}} [1 - \hat{Q}_k] \leq 1\). By (2.6), the \(L^1\)-norm of such polynomial does not exceed \(\text{Const} \cdot \log N\), and (2.5)\(_s\) follows, for

\[ \begin{align*}
\left\| \frac{\partial^{2s}}{\partial x^{2s}} R_N(\cdot, t) \right\|_{L^1} &= m_N^{2s} \left\| \sum_{k=-N}^{N} \frac{k^{2s}}{m_N^{2s}} \hat{R}_k(t) e^{ikx} \right\|_{L^1} \\
&\leq \text{Const} \cdot m_N^{2s} \log N, \quad s \leq q.
\end{align*} \]

Equipped with Lemma 2.1, one can show now that the SV approximation (2.4)—like the viscosity approximation (2.3)—is \(L^1\)-stable and total-variation bounded. The necessary estimate in this direction is included in the following

**Lemma 2.2** (\(L^1\)-stability). Let \(u_N\) and \(v_N\) be two different solutions of the SV approximation (1.4). Then there exist constants \(C_N \sim N^{-(1-\frac{1}{r})\theta}\) such that the following estimate holds:

\[ \|u_N(\cdot, t) - v_N(\cdot, t)\|_{L^1} \]

\[ \leq e^{C_N t} \left[ \|u_N(\cdot, 0) - v_N(\cdot, 0)\|_{L^1} + \left\| \frac{\partial}{\partial x} (I - P_N)[f(u_N) - f(v_N)] \right\|_{L^1([0,t])} \right]. \]

(2.7)

**Remark.** Taking into account the truncation error spectral decay (2.1)\(_1\), then inequality (2.7) provides us with the announced \(L^1\)-stability of the form

\[ \|u_N(\cdot, t) - v_N(\cdot, t)\|_{L^1} \]

\[ \leq e^{C_N t} \left[ \|u_N(\cdot, 0) - v_N(\cdot, 0)\|_{L^1} + K_s \cdot (t N^{-s(1-\theta)+1}) + N^{-r+1+\theta} \right]. \]

**Proof of Lemma 2.2.** The difference \(u_N - v_N\) satisfies

\[ \begin{align*}
\frac{\partial}{\partial t} (u_N - v_N) + \frac{\partial}{\partial x} (f(u_N) - f(v_N)) \\
= \varepsilon_N \frac{\partial^2}{\partial x^2} (u_N - v_N) - \varepsilon_N \frac{\partial^2}{\partial x^2} R_N \ast (u_N - v_N) \\
+ \frac{\partial}{\partial x} (I - P_N)[f(u_N) - f(v_N)].
\end{align*} \]
One may proceed now with an $L^1$-estimate in a standard fashion: we integrate against $\text{sgn}(u_N - v_N)$, and in view of $(2.5)_1$ and $(2.1)_1$, we obtain

\[
\frac{d}{dt}||u_N(\cdot, t) - v_N(\cdot, t)||_{L^1}
\leq \varepsilon_N \left\| \frac{\partial^2}{\partial x^2} R_N(\cdot, t) \right\|_{L^1} ||u_N(\cdot, t) - v_N(\cdot, t)||_{L^1}
+ \left\| \frac{\partial}{\partial x} (I - P_N)[f(u_N) - f(v_N)] \right\|_{L^1}
\leq C_N \cdot ||u_N(\cdot, t) - v_N(\cdot, t)||_{L^1} + K_S \cdot (N^{-s(1-\theta)+1} + N^{-r+1} e^{-N^2-\theta t}),
\]

where according to $(1.6)_q$ one has $C_N \sim \varepsilon_N m_N^2 \log N \sim N^{-(1-\frac{1}{q})\theta}$. The assertion $(2.7)$ now follows. \(\square\)

Application of Lemma 2.2 with $v_N(\cdot, t) = u_N(\cdot + \Delta x, t)$ shows that the total variation of the SV solution,

\[
||u_N(\cdot, t)||_{BV} \equiv \sup_{\Delta x} \frac{1}{\Delta x} ||u_N(x + \Delta x, t) - u_N(x, t)||_{L^1},
\]

does not exceed

\[
e^{C_N t} \left[ ||u_N(\cdot, 0)||_{BV} + \left\| \frac{\partial^2}{\partial x^2} (I - P_N)f(u_N) \right\|_{L^1(x, [0, t])} \right].
\]

In fact, if $q > 1$, then $(1.6)_q$ implies that $C_N \sim N^{-(1-\frac{1}{q})\theta} \downarrow 0$, and together with the spectral decay estimate $(2.1)_2$ we conclude

**Corollary 2.3** (Total-variation boundedness). The SV approximation $(1.4)$–$(1.6)_q$ with $q > 1$ is essentially nonoscillatory, in the sense that the increase of its initial total variation is $o(1)$,

\[
||u_N(\cdot, t)||_{BV} \leq (1 + \Theta(N^{-(1-\frac{1}{q})\theta}))
\times \left[ ||u_N(\cdot, 0)||_{BV} + K_s \cdot (tN^{-s(1-\theta)+2} + N^{-r+\theta}) \right].
\]

**Remarks.** 1. Corollary 2.3 tells us that the SV solution $u_N(\cdot, t)$ is total-variation bounded (independently of $N$), provided its initial data, $u_N(\cdot, 0)$, are. (Observe that the BV-smoothness of $u_N(\cdot, 0)$ allows us to use $(2.1)_p$ with $r \geq \frac{1}{2}$, and hence the contribution of the truncation error, which is bounded from above by the second term on the right of $(2.8)$, is negligible, at least for $\theta < \frac{1}{2}$. For $\frac{1}{2} \leq \theta < 1$, however, one might need a slightly stronger assumption regarding the initial smoothness, e.g., $u_N(\cdot, 0) \in W^{\frac{1}{2}, 1}$; we shall not explore this issue here since, as noted above, the smoothness requirements for $(2.1)_p$ to hold are not optimal to begin with.)

To guarantee the BV-boundedness of $u_N(\cdot, 0)$ without sacrificing spectral accuracy, one can preprocess the exact initial data $u_0$ prescribed in $(1.1a)$. For example, de la Vallée Poussin's filter,

\[
(2.9) \quad u_N(x, 0) = V P_N u_0 \equiv \sum_{k=-N}^{N} \sigma_k \hat{u}_k(0) e^{ikx}, \quad \sigma_k = \begin{cases} 
1, & |k| \leq \frac{N}{2}, \\
2 - \frac{k}{N}, & |k| > \frac{N}{2}
\end{cases}
\]
yields a spectrally accurate approximation of $u_0$,

$$\|(I - V P_N)u_0\|_{H^{-s}} \leq \text{Const} \cdot N^{-\frac{s}{2}}\|u_0\|_{L^2}, \quad \forall s \geq 0;$$

moreover, since $\|V P_N u_0\|_{BV} \leq 3\|u_0\|_{BV}$, it follows from Corollary 2.3 that the SV approximation (1.4)–(1.6)$_q$, subject to the preprocessed initial data (2.9), is total-variation bounded (independently of $N$) for arbitrary BV-initial data $u_0$.

Finally, we note that such initial preprocessing might be necessary, since the unfiltered (pseudo-)spectral projection, $\|P_N u_0\|_{BV}$, may grow as much as $\mathcal{O}(\log N)$ for arbitrary BV-initial data. Of course, it can be avoided if the initial data are smooth enough, say in $H^1$, for then $\|P_N u_0\|_{BV} \leq \|u_0\|_{H^1} < \infty$.

2. The BV-estimate (2.8) shows how the SV method maintains the delicate tradeoff between spectral accuracy and TV-stability: According to (2.8), the total variation of the SV solution (1.4)–(1.6)$_q$ with $q = 1$ may grow by a factor of $\mathcal{O}(1)$ times its initial variation; as $q$ increases, this growth factor approaches one—in agreement with the Total-Variation Diminishing (TVD) property of the exact solution, but at the same time, the spectral accuracy estimate (1.7) ‘deteriorates’. Thus, the SV method (1.4)–(1.6)$_q$ can be viewed as a compromise between the first-order TVD viscosity approximation (2.3) (which corresponds to $q = \infty$), and the spectrally accurate, yet unstable, Fourier approximation (1.2) (which corresponds to $q = 0$).

Similarly, the $L^1$-stability (2.7) approaches the $L^1$-contraction of the exact entropy solution as we increase the amount of spectral viscosity by letting $q \uparrow \infty$.

3. The total-variation boundedness of the SV solution implies that (a subsequence of) $u_N(x, t)$ converges strongly to a limit $u(x, t)$, which is a weak solution of (1.1a). To conclude that this limit is the unique entropy solution of (1.1), it remains to verify that $u$ satisfies the entropy condition (1.1b). To this end, we multiply (2.4a) by $U'(u_N)$, obtaining

$$\frac{\partial}{\partial t} U(u_N) + \frac{\partial}{\partial x} F(u_N) = \frac{\partial}{\partial x} \left( \epsilon_N U'(u_N) \frac{\partial u_N}{\partial x} \right) - \epsilon_N U''(u_N) \left( \frac{\partial u_N}{\partial x} \right)^2$$

$$+ \frac{\partial}{\partial x} \left( \epsilon_N u_N R_N * \frac{\partial u_N}{\partial x} \right) - \epsilon_N \frac{\partial u_N}{\partial x} R_N * \frac{\partial u_N}{\partial x} = \frac{\partial}{\partial x} I + \text{II} + \frac{\partial}{\partial x} \text{III} + \text{IV}.$$ 

Since the entropy function $U$ is convex, II $\leq 0$. This, together with the straightforward estimates (based on (2.5)$_0$, (2.8) and (3.3)$_0$ below)

$$\|\text{II}\|_{L^1} + \|\text{III}\|_{L^1} \leq \epsilon_N \|R_N\|_{L^1} \|u_N\|_{L^\infty} \|u_N\|_{BV} \leq \text{Const} \cdot \epsilon_N \log N \|u_0\|_{BV} \rightarrow 0,$$

$$\|\text{IV}\|_{L^1} \leq \epsilon_N \|R_N\|_{L^\infty} \|u_N\|_{BV} \leq \text{Const} \cdot \epsilon_N m_N \log N \|u_0\|_{BV}^2 \rightarrow 0,$$

imply that $u = \lim u_N$ satisfies the entropy inequality (1.1b), and convergence of (the whole sequence of) $u_N$ to the entropy solution follows.

4. The total-variation bound indicated above implies the usual decay

$$(2.10) \quad |\hat{u}_k(t)| \leq \frac{\|u_N(\cdot, t)\|_{BV}}{|k| + 1},$$
which in turn implies, by the Parseval identity, an $L^2(x)$-bounded entropy production estimate (uniformly in time) of the type indicated earlier in (1.8),
\[
\frac{1}{N} \left\| \frac{\partial}{\partial x} u_N(\cdot, t) \right\|_{L^2_{loc}(x)}^2 \leq \text{Const}, \quad 0 \leq t \leq T.
\]

3. Convergence rate estimates

In this section we restrict our attention to the genuinely nonlinear conservation law (1.1) where $f'' \geq \alpha > 0$.

We say that a family of approximate solutions $\{u_N(x, t)\}$ is Lip+$^+$-stable, if there exists a constant (independent of $N$), such that the following estimate is fulfilled$^1$:
\[
\|u_N(\cdot, t)\|_{\text{Lip}^+} \leq \text{Const}_T, \quad 0 \leq t \leq T.
\]
Recall that the viscosity approximation $u_\varepsilon$ as well as the entropy solution of the nonlinear conservation law (1.1) with $f'' \geq \alpha > 0$ satisfy Oleinik’s E-condition, e.g., [7, 15],
\[
\|u_\varepsilon(\cdot, t)\|_{\text{Lip}^+} \leq \frac{1}{\|u_0\|_{\text{Lip}^+} + \alpha t}, \quad t \geq 0.
\]
In particular, they are Lip+$^+$-stable as long as their initial data $u_0$ are Lip+$^+$-bounded. We want to show that the SV approximation (1.4) is also Lip+$^+$-stable.

We remark that the BV-bound (2.8) does not exclude the possibility of small high-frequency oscillations. (By conservation, Lip+$^+$- implies BV-stability, but not vice versa). Such ‘unphysical’ oscillations may violate the Lip+$^+$-stability of the SV solution. In order to prevent such Lip+$^+$-unstable oscillations, we therefore need to slightly increase the amount of spectral viscosity. We achieve this (without sacrificing formal spectral accuracy) by requiring the spectral viscosity parameters to lie in the range $(1.5)_q- (1.6)_q^+$ with $q \geq \frac{3}{2}$.

As before, the Lip+$^+$-stability of the SV method hinges on the (small) size of the ‘residual’ kernel, $R_N(x, t)$, which distinguishes the SV approximation (2.4a) from the Lip+$^+$-stable viscosity approximation (2.3). To this end, we first state

**Lemma 3.1.** Consider the SV kernel $Q_N(x, t)$ subject to the SV parameterization $(1.5)_q- (1.6)_q^+$. Then $R_N(x, t) \equiv D_N(x) - Q_N(x, t)$ satisfies
\[
(3.3)_s \quad \left\| \frac{\partial \phi}{\partial x^{2s}} R_N(\cdot, t) \right\|_{L^\infty} \leq \text{Const} \cdot m_N^{2s+1} \log N, \quad 0 \leq s \leq q - \frac{1}{2}.
\]

**Proof.** By $(1.5)_q^+$, $\hat{R}_k \equiv 1 - \hat{Q}_k \leq (\frac{m_N}{|k|})^{2q}$. The lemma follows from the straightforward estimate
\[
\left\| \frac{\partial \phi}{\partial x^{2s}} R_N(\cdot, t) \right\|_{L^\infty} \leq \sum_{|k| \leq N} |k|^{2s} |\hat{R}_k| \leq \sum_{|k| < m_N} |k|^{2s} + \sum_{|k| \geq m_N} \frac{m_N^{2s+1}}{|k|^{2q}} \left( \frac{m_N}{|k|} \right)^{2q}
\]
\[
\leq \text{Const} \cdot m_N^{2s+1} + \text{Const} \cdot m_N^{2q} \cdot \begin{cases} m_N^{2s+2q+1} & \text{if } s < q - \frac{1}{2} \\ \log N & \text{if } s = q - \frac{1}{2} \end{cases}
\]
\[
\leq \text{Const} \cdot m_N^{2s+1} \log N. \quad \square
\]

$^1$We let $\|\phi\|_{\text{Lip}^+}$, $\|\phi\|_{\text{Lip}^{+\prime}}$ and $\|\phi\|_{\text{Lip}^{+\prime}}$ denote respectively, $ess \sup_{x \neq y} |(\phi(x) - \phi(y))/(x - y)|$, $ess \sup_{x \neq y} [(\phi(x) - \phi(y))/(x - y)]_+$ and $\sup_{x \neq y} (\phi(x) - \phi(y))/\|\psi\|_{\text{Lip}}$. 

Equipped with Lemma 3.1, we now turn to the Lip$^+$-stability proof of the SV method, stating

**Lemma 3.2 (Lip$^+$-stability).** The SV approximation with $q > \frac{1}{2}$ satisfies the Lip$^+$-stability estimate

$$
(3.4) \quad \|u_N(\cdot, t)\|_{\text{Lip}^+} \leq \frac{1 + \sqrt{\frac{6\kappa}{\alpha}}}{\|u_N(\cdot, 0)\|_{\text{Lip}^+}^{-1} + \alpha_N t}, \quad \alpha_N \equiv \frac{\tanh(\sqrt{\alpha c_N t})}{\sqrt{\alpha c_N t}} \sim \alpha,
$$

where the vanishingly small constants, $c_N$, are given by

$$
c_N \sim N^{-\theta(1-\frac{1}{2q})} \cdot \|u_N(\cdot, 0)\|_{\text{BV}} + K_s \cdot (N^{-s(1-\theta)+\frac{1}{2}} + N^{-r+\frac{1}{2}+\theta}).
$$

**Remark.** The constants $c_N$ involved in the Lip$^+$-bound (3.4) have two major contributions: the second term on the right represents an upper bound of the truncation error, which is spectrally small—provided the initial data $u_N(\cdot, 0)$ are sufficiently smooth, say $u_N(\cdot, 0) \in W^{2,1}$ so that (3.4) holds with $r \geq \frac{1}{2}$ (otherwise, an initial layer may be formed, after which the spectral viscosity becomes effective and drives the truncation error spectrally small). Apart from this spectrally small contribution, we have $c_N \sim N^{-\theta(1-\frac{1}{2q})}$; here we observe that, as before, when the amount of spectral viscosity increases with $q$, the Lip$^+$-bound in (3.4) becomes tighter, in agreement with (3.2), and in particular, the two Lip$^+$-bounds coincide in the fully viscous limit $q = \infty$.

**Proof of Lemma 3.2.** Differentiation of (2.4a) yields, for $w_N(x, t) \equiv \frac{\partial}{\partial x} u_N(x, t)$,

$$
\frac{\partial}{\partial t} w_N(x, t) + f'(u_N(x, t)) \frac{\partial}{\partial x} w_N(x, t) + f''(u_N(x, t))w_N^2(x, t)
$$

$$
= \varepsilon_N \frac{\partial^2}{\partial x^2} w_N(x, t) - \varepsilon_N \frac{\partial^2}{\partial x^2} R_N \ast w_N(x, t) + \frac{\partial^2}{\partial x^2} (I - P_N)f(u_N),
$$

which implies that $\|u_N(\cdot, t)\|_{\text{Lip}^+} = \max_x[w_N(x, t)_{\text{Lip}^+}]$ satisfies the differential inequality

$$
\frac{d}{dt} \|u_N(\cdot, t)\|_{\text{Lip}^+} + \alpha \|u_N(\cdot, t)\|_{\text{Lip}^+}^2
$$

$$
\leq \varepsilon_N \left\| \frac{\partial^2}{\partial x^2} R_N(\cdot, t) \right\|_{L^\infty} \cdot \|u_N(\cdot, t)\|_{\text{BV}} + \left\| \frac{\partial^2}{\partial x^2} (I - P_N)f(u_N) \right\|_{L^\infty}
$$

$$
\equiv I_N + II_N.
$$

We recall that according to (3.3)$\_1$,

$$
(3.6a) \quad I_N = \varepsilon_N \left\| \frac{\partial^2}{\partial x^2} R_N(\cdot, t) \right\|_{L^\infty} \|u_N\|_{L^\infty(\text{BV}, [0, t])}
$$

$$
\leq \text{Const} \cdot N^{-\theta(1-\frac{1}{2q})} \cdot \|u_N\|_{L^\infty(\text{BV}, [0, t])};
$$

moreover, according to (2.1)$\_2$, together with Sobolev’s inequality, we have

$$
(3.6b) \quad II_N = \left\| \frac{\partial^2}{\partial x^2} (I - P_N)f(u_N) \right\|_{L^\infty} \leq K_s \cdot (N^{-s(1-\theta)+\frac{1}{2}} + N^{-r+\frac{1}{2}+\theta} e^{-N^{-2-\theta} t}).
$$

Equipped with (3.6a)-(3.6b) we return to (3.5), obtaining

$$
\frac{d}{dt} |u_N(\cdot, t)|_{\text{Lip}^+} + \alpha |u_N(\cdot, t)|_{\text{Lip}^+}^2 \leq c_N, \quad c_N \equiv I_N + II_N \downarrow 0,
$$
which in turn implies the desired Lip$^+$-stability of $u_N$,
\[
\|u_N(\cdot, t)\|_{\text{Lip}^+} \leq \frac{c_N \tanh(\sqrt{\alpha N}t) + \sqrt{\alpha N} \|u_N(\cdot, 0)\|_{\text{Lip}^+}}{\alpha_N + \alpha \tanh(\sqrt{\alpha N}t) \|u_N(\cdot, 0)\|_{\text{Lip}^+}}.
\]
\[
\leq \frac{1 + \sqrt{\frac{\beta \alpha}{\alpha}}}{\|u_N(\cdot, 0)\|_{\text{Lip}^+}^{-1} + \alpha_N t}, \quad \alpha_N \equiv \alpha \frac{\tanh(\sqrt{\alpha N}t)}{\sqrt{\alpha N}} \sim \alpha.
\]

We now recall the main result of [9] (see also [16]), which deals with the convergence rate of Lip$^+$-stable approximations.

**Theorem 3.3** [9]. Let \{\(u_N(x, t)\)\} be a family of Lip$^+$-stable approximate solutions of the conservation law (1.1), with Lip$^+$-bounded initial data. Assume that \{\(u_N(x, t)\)\} are Lip$'$-consistent of order \(\varepsilon\),
\[
\|u_N(x, 0) - u_0(x)\|_{\text{Lip}'} + \left\| \frac{\partial}{\partial t} u_N(\cdot, t) + \frac{\partial}{\partial x} f(u_N(\cdot, t)) \right\|_{\text{Lip}'}
\]
\[
\leq \text{Const}_T \cdot \varepsilon, \quad 0 \leq t \leq T.
\]

Then the following error estimates hold:
\[
\begin{align*}
(3.8a) \quad & \|u_N(\cdot, t) - u(\cdot, t)\|_{W^{-s,p}} \leq \text{Const} \cdot \varepsilon^{\frac{s+p}{2p}}, \quad 1 \leq p < \infty, \quad -\frac{1}{p} \leq s \leq 1; \\
(3.8b) \quad & |u_N(x, t) - u(x, t)| \leq C_1 \cdot \varepsilon^{\frac{1}{3}}, \quad 0 \leq t \leq T; \\
(3.8c) \quad & |u_N(x, t) \ast \psi_r - u(x, t)| \leq C_r \cdot \varepsilon^{\frac{r}{2r}}, \quad 1 \leq r < \infty.
\end{align*}
\]

Here, \(\psi_r\) is any \(r\)-th order mollifier, and the constants \(C_r \sim 1 + \|u(\cdot, t)\|_{\text{Lip}'} \|W^{s,p}_\infty\|\) measure the local smoothness of the entropy solution in the \(\Theta(\varepsilon^{\frac{1}{3}})\)-neighborhood of \(x\).

We have shown that the SV approximation is Lip$^+$-stable and hence convergent to the exact entropy solution of (1.1). To estimate the convergence rate with the help of Theorem 3.3, it remains to verify the order of its Lip$'$-consistency. To this end, we note

1. The initial data, \(u_N(x, 0) = P_N u_0\), are Lip$'$-consistent of order \(\sim \frac{1}{N}\) with the BV-initial data \(u_0\), for by (2.10),
\[
(3.9a) \quad \|P_N u_0 - u_0\|_{\text{Lip}'} \leq \sum_{|k| > N} \left| \frac{\hat{u}_k(0)}{k} \right| \leq \frac{1}{N} \|u_0\|_{BV}.
\]

2. The SV approximation (1.4) is Lip$'$-consistent with the conservation law (1.1a) of order \(\varepsilon_N \log N \sim N^{-\theta}\), for
\[
(3.9b) \quad \left\| \frac{\partial}{\partial t} u_N(\cdot, t) + \frac{\partial}{\partial x} f(u_N(\cdot, t)) \right\|_{\text{Lip}'} \leq \varepsilon_N \|Q_N(\cdot, t) + \frac{\partial}{\partial x} u_N(\cdot, t)\|_{\text{Lip}'}
\]
\[
\leq \varepsilon_N \|Q_N(\cdot, t)\|_{L^1} \|u_N(\cdot, t)\|_{BV} \leq \text{Const} \cdot N^{-\theta} \|u_N(\cdot, 0)\|_{BV}.
\]

Here, the first inequality follows from (2.2) by ignoring the spectrally small discretization error (2.1); the second is an obvious use of Young's inequality; and the third inequality uses (2.5) and (2.6), which show that the \(L^1\)-norm of the viscosity kernel, \(Q_N(\cdot, t) \equiv D_N(\cdot) - R_N(\cdot, t)\), does not exceed \(\text{Const} \cdot \log N\).
In summary, we find that the $\text{Lip}^+$-stable SV approximation (1.4) is $\text{Lip}^+$-consistent of order $\varepsilon \sim N^{-\theta}$, and using Theorem 3.3, we conclude

**Theorem 3.4** (Convergence rate estimates). Consider the $2\pi$-periodic nonlinear conservation law (1.1) with $\text{Lip}^+$ initial data. Then the SV approximation (1.4), (1.5)$_q$ (1.6)$_q$ with $q \geq \frac{1}{2}$, converges to the entropy solution of (1.1), and the following error estimates hold for $1 \leq p$, $r < \infty$, $-1/p \leq s \leq 1$:

\begin{align}
(3.10) \quad \|u_N(\cdot, t) - u(\cdot, t)\|_{W^{-s,p}} & \leq \text{Const} \cdot N^{-\frac{\theta}{2} + \frac{1}{2} p}, \quad 0 < t_0 \leq t \leq T; \\
(3.11) \quad |u_N(x, t) - u(x, t)| & \leq C_1 \cdot N^{-\frac{r}{2}}, \quad 0 < t_0 \leq t \leq T; \\
(3.12) \quad |u_N(x, t) \ast \psi_r - u_N(x, t)| & \leq C_r \cdot N^{-\frac{r}{2} + \theta}, \quad 0 < t_0 \leq t \leq T.
\end{align}

**Remarks.** 1. Theorem 3.4 requires the initial data of the SV method, $u_N(x, 0)$, to be $\text{Lip}^+$-bounded independently of $N$. Consequently, one might need to pre-process the prescribed initial data $u_0$ unless they are smooth enough to begin with. The de la Vallée Poussin preprocessing in (2.9) will guarantee this requirement for arbitrary $\text{Lip}^+$-bounded initial data $u_0$.

2. The error estimates (3.10), (3.11) are not uniform in time as $t_0 \downarrow 0$, unless the initial data are sufficiently smooth to guarantee the uniformity (in time) of the $\text{Lip}^+$ bound (3.4); consult the remark following Lemma 3.2. For arbitrary $\text{Lip}^+$ initial data, $u_0$, an initial layer may be formed, after which the spectral viscosity becomes effective and guarantees the spectral decay of the discretization error indicated in (2.1)$_p$.

3. According to (3.11) and (3.12), the pointwise convergence rate of the SV solution in smooth regions of the entropy solution is of order $\sim N^{-\frac{1}{2}}$, and by postprocessing the SV solution, this convergence rate can be made arbitrarily close to $N^{-1}$. In fact, numerical experiments reported in [14] show that by postprocessing the SV solution using the spectrally accurate mollifier of [3],

\[ \psi_r(x) = \psi_0(x)D_n(x), \quad n \sim \varepsilon^{-\frac{1}{2}} \]

we recover the pointwise values in smooth regions of the entropy solution within spectral accuracy.

4. According to (3.10) with $(s, p) = (0, 1)$, the SV approximation has an $L^1$-convergence rate of order $\sim N^{-\frac{1}{2}}$, in agreement with [11, §5]. This corresponds to the usual $L^1$-convergence rate of order $\frac{1}{2}$ for monotone difference approximations, [6, 10].

**Bibliography**


15. ——, *Semi-discrete approximations to nonlinear systems conservation laws; consistency and $L^\infty$-stability imply convergence*, ICASE Report No. 88-41.


School of Mathematical Sciences, Tel-Aviv University, Tel-Aviv 69978, Israel

E-mail address: tadmor@taurus.bitnet