# AN ENTROPIC FOURIER METHOD FOR THE BOLTZMANN EQUATION* 

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#### Abstract

We propose an entropic Fourier method for the numerical discretization of the Boltzmann collision operator. The method, which is obtained by modifying a Fourier-Galerkin method to match the form of the discrete velocity method, can be viewed both as a discrete velocity method and as a Fourier method. As a discrete velocity method, it preserves the positivity of the solution and satisfies a discrete version of the H-theorem. As a Fourier method, it allows one to readily apply the FFT-based fast algorithms. A second-order convergence rate is validated by numerical experiments.


Key words. Fourier method, discrete velocity method, Boltzmann equation, modified Jackson filter, H-theorem, positivity

AMS subject classifications. $65 \mathrm{M} 70,65 \mathrm{R} 20,76 \mathrm{P} 05$
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1. Introduction. Gas kinetic theory describes the statistical behavior of a large number of gas molecules in the joint spatial and velocity space. It has been widely used to model gases outside the hydrodynamic regime, for example in the field of rarefied gas dynamics. Let $f(t, x, v)$ be the mass density distribution of the particles, depending on the time $t \in \mathbb{R}^{+}$, position $x \in \mathbb{R}^{d}(d \geq 2)$, and microscopic velocity $v \in \mathbb{R}^{d}$. Based on the molecular chaos assumption, the Boltzmann equation

$$
\begin{equation*}
\frac{\partial f}{\partial t}+v \cdot \nabla_{x} f=\mathcal{Q}[f, f], \quad f(0, x, v)=f^{0}(x, v) \tag{1.1}
\end{equation*}
$$

for the evolution of $f(t, x, v)$ was derived in [6] and has served as the fundamental equation in gas kinetic theory. When modeling the binary interaction between the particles, the Boltzmann collision operator $\mathcal{Q}[f, f]$ takes the form

$$
\begin{equation*}
\mathcal{Q}[f, f](v)=\int_{\mathbb{R}^{d}} \int_{\mathbb{S}^{d-1}} \mathcal{B}\left(v-v_{*}, \omega\right)\left[f\left(v^{\prime}\right) f\left(v_{*}^{\prime}\right)-f(v) f\left(v_{*}\right)\right] \mathrm{d} v_{*} \mathrm{~d} \omega \tag{1.2}
\end{equation*}
$$

for monatomic gases, where

$$
v^{\prime}=\frac{v+v_{*}}{2}+\frac{\left|v-v_{*}\right|}{2} \omega, \quad v_{*}^{\prime}=\frac{v+v_{*}}{2}-\frac{\left|v-v_{*}\right|}{2} \omega
$$

are the postcollisional velocities of two particles with precollisional velocities $v$ and $v_{*}$, and $\omega$ is the angular parameter of the collision. Here the variables $t$ and $x$ are omitted for simplicity, and we shall continue to do so when focusing only on the collision term. The collision kernel $\mathcal{B}$ is nonnegative and usually takes the form

$$
\begin{equation*}
\mathcal{B}\left(v-v_{*}, \omega\right)=b\left(\left|v-v_{*}\right|, \cos \theta\right), \quad \cos \theta=\left|\left(v-v_{*}\right) \cdot \omega\right| /\left|v-v_{*}\right| . \tag{1.3}
\end{equation*}
$$

[^0]The Boltzmann equation (1.1) guarantees that $f(t, x, v)$ remains nonnegative if the initial value $f(t=0, x, v)$ is also nonnegative [28]. The symmetry of the collision term (1.2) and the fact $\mathrm{d} v \mathrm{~d} v_{*}=\mathrm{d} v^{\prime} \mathrm{d} v_{*}^{\prime}$ imply that for any function $\psi(\cdot)$

$$
\begin{align*}
& \int_{\mathbb{R}^{d}} \psi(v) \mathcal{Q}[f, f](v) \mathrm{d} v  \tag{1.4}\\
& \quad=\frac{1}{4} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{S}^{d}-1}\left(\psi(v)+\psi\left(v_{*}\right)-\psi\left(v^{\prime}\right)-\psi\left(v_{*}^{\prime}\right)\right) \mathcal{B}\left[f\left(v^{\prime}\right) f\left(v_{*}^{\prime}\right)-f(v) f\left(v_{*}\right)\right] \mathrm{d} v_{*} \mathrm{~d} v \mathrm{~d} \omega .
\end{align*}
$$

Setting $\psi(v)=1, v,|v|^{2}$ gives rise to the conservation of the mass, momentum, and energy

$$
\begin{equation*}
\int_{\mathbb{R}^{d}} \mathcal{Q}[f, f](v) \mathrm{d} v=0, \quad \int_{\mathbb{R}^{d}} \mathcal{Q}[f, f](v) v \mathrm{~d} v=0, \quad \int_{\mathbb{R}^{d}} \mathcal{Q}[f, f](v)|v|^{2} \mathrm{~d} v=0, \tag{1.5}
\end{equation*}
$$

respectively. The famous H -theorem that states the monotonicity of the entropy

$$
\begin{align*}
& \int_{\mathbb{R}^{d}} \mathcal{Q}[f, f](v) \ln (f(v)) \mathrm{d} v \\
& \quad=\frac{1}{4} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{S}^{d-1}} \ln \left(\frac{f(v) f\left(v_{*}\right)}{f\left(v^{\prime}\right) f\left(v_{*}^{\prime}\right)}\right) \mathcal{B}\left[f\left(v^{\prime}\right) f\left(v_{*}^{\prime}\right)-f(v) f\left(v_{*}\right)\right] \mathrm{d} v_{*} \mathrm{~d} v \mathrm{~d} \omega \leq 0 \tag{1.6}
\end{align*}
$$

can also be obtained by setting $\psi(v)=\ln (f(v))$.
By introducing new variables $y=v^{\prime}-v$ and $z=v_{*}^{\prime}-v$ and carrying out algebra calculations, the Boltzmann collision operator can be rewritten as (see [7, 22, 19] for details)

$$
\begin{equation*}
\mathcal{Q}[f, f](v)=\int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \tilde{\mathcal{B}}(y, z) \delta(y \cdot z)[f(v+y) f(v+z)-f(v) f(v+y+z)] \mathrm{d} y \mathrm{~d} z . \tag{1.7}
\end{equation*}
$$

This is the well-known Carleman representation [7] of the Boltzmann collision operator, where $\tilde{\mathcal{B}}(y, z)$ is related to $\mathcal{B}\left(v-v_{*}, \omega\right)$ in (1.3) by

$$
\begin{equation*}
\tilde{\mathcal{B}}(y, z)=2^{d-1} \mathcal{B}\left(y+z, \frac{y-z}{|y-z|}\right)|y+z|^{2-d} . \tag{1.8}
\end{equation*}
$$

Though the Boltzmann equation serves as the fundamental equation in gas dynamics, its high-dimensional nature and the complexity of the collision operator pose difficulties for its numerical solution. A classical method is the direct Monte Carlo simulation [2], which uses simulated particles to mimic gas molecules and handles the collisions in a stochastic way. Though it treats the high dimensionality effectively, the convergence order is low and the numerical solution appears rather oscillatory.

With the rapid growth of computing power, it has become more practical to solve the Boltzmann equation with deterministic methods. For all deterministic approaches, the complexity of the collision integral poses the most serious difficulty for numerical computation. Therefore, this paper focuses on the spatially homogeneous case

$$
\begin{equation*}
\frac{\partial f}{\partial t}=\mathcal{Q}[f, f] \tag{1.9}
\end{equation*}
$$

for simplicity.
In the past decades, several deterministic schemes have been developed for the Boltzmann collision term. Two methods that have attracted the most attention are the discrete velocity method (DVM) $[13,27,4,22]$ and the Fourier-Galerkin method (FGM) [5, 23, 24], which will be reviewed in what follows.
1.1. Discrete velocity method. The DVM assumes that the particle velocity takes only values from a finite set. Consider a domain $\mathcal{D}_{T}=[-T, T]^{d}$ for the velocity variable $v$ that is discretized uniformly with step size $h=2 T / N$ for a positive integer $N$ (which is assumed to be odd for simplicity). By adopting the $d$-dimensional multiindex notation $k=\left(k_{1}, \ldots, k_{d}\right)$, one can denote the set of discrete velocity samples by

$$
\begin{equation*}
X=\left\{h \cdot k \mid k=\left(k_{1}, \ldots, k_{d}\right),-n \leq k_{1}, \ldots, k_{d} \leq n\right\} \tag{1.10}
\end{equation*}
$$

where $N=2 n+1$. In the rest of this paper, we use the lowercase letters $p, q, r$, and $s$ to denote the discrete velocity samples in $X$.

Using $F_{r}(t)$ for $r \in X$ as the numerical approximations of the distribution function $f(t, v)$ at the points in $X$, the governing equations of DVM for $F_{r}(t)$ are

$$
\begin{equation*}
\frac{\mathrm{d} F_{r}(t)}{\mathrm{d} t}=Q_{r}(t):=\sum_{p, q, s \in X} A_{p q}^{r s}\left(F_{p}(t) F_{q}(t)-F_{r}(t) F_{s}(t)\right), \quad r \in X \tag{1.11}
\end{equation*}
$$

Here $Q_{r}(t)$ serves as an approximation to $\mathcal{Q}[f, f](t, r)$. The coefficients $A_{p q}^{r s}$ are nonnegative constants and satisfy the conservation relations

$$
\begin{equation*}
A_{p q}^{r s} \neq 0 \Rightarrow p+q=r+s \quad \text { and } \quad|p|^{2}+|q|^{2}=|r|^{2}+|s|^{2} \tag{1.12}
\end{equation*}
$$

and the symmetry property

$$
\begin{equation*}
A_{p q}^{r s}=A_{q p}^{r s}=A_{r s}^{p q} \tag{1.13}
\end{equation*}
$$

The property (1.12) shows that the collisions in DVM also satisfy the momentum and energy conservation, and the property (1.13) implies that the collisions are also reversible as in the Boltzmann equation. These two facts guarantee that DVM maintains a number of fundamental physical properties of the continuous Boltzmann equation, such as (a) the positivity of the distribution function, (b) the exact conservation of mass, momentum, and energy, and (c) a discrete H -theorem.

More precisely, the values $F_{r}(t)$ for $r \in X$ are always nonnegative if the initial values $F_{r}(t=0)$ are nonnegative $[28,26]$. The symmetry relation (1.13) implies that

$$
\begin{equation*}
\sum_{r \in X} Q_{r} \psi_{r}=\frac{1}{4} \sum_{p, q, r, s \in X} A_{p q}^{r s}\left(\psi_{r}+\psi_{s}-\psi_{p}-\psi_{q}\right)\left(F_{p} F_{q}-F_{r} F_{s}\right) \tag{1.14}
\end{equation*}
$$

Combining (1.14) and the relations (1.12) gives rise to the conservation of mass, momentum, and energy in the discrete sense:

$$
\begin{equation*}
\sum_{r \in X} Q_{r}=0, \quad \sum_{r \in X} Q_{r} r=0, \quad \sum_{r \in X} Q_{r}|r|^{2}=0 \tag{1.15}
\end{equation*}
$$

After letting $\psi_{r}=\ln \left(F_{r}\right)$ in (1.14), one obtains a discrete version of the H-theorem for DVM,

$$
\begin{equation*}
\sum_{r \in X} Q_{r} \ln \left(F_{r}\right)=\frac{1}{4} \sum_{p, q, r, s \in X} A_{p q}^{r s} \ln \left(\frac{F_{r} F_{s}}{F_{p} F_{q}}\right)\left(F_{p} F_{q}-F_{r} F_{s}\right) \leq 0 \tag{1.16}
\end{equation*}
$$

using the nonnegativity of the coefficients $A_{p q}^{r s}$ and the monotonicity of the ln function. Notice that, in this argument, the symmetry relations (1.13), the nonnegativity of
$A_{p q}^{r s}$, and the nonnegativity of the initial values are all essential for derivation of the H-theorem.

As a direct discretization for the high-dimensional integral of the collision term, DVM has a rather high computational cost $O\left(N^{2 D+\delta}\right)$ for some $0<\delta \leq 1$ [8]. It is also difficult to achieve a decent convergence rate due to the insufficient collision pairs in the Cartesian grid used for velocity discretization. More precisely, in the 2D case, the rate of convergence of DVM introduced in [13] is only $O\left((1 / \log h)^{p}\right)$ with $p<1 / 2-1 / \pi[9]$. For the 3D case, the best rate of convergence of DVM is also slower than the first order [21, 22].

The method in $[18,29]$ tries to improve the accuracy of DVM by interpolation. While the mass, momentum, and energy are conserved in this scheme, positivity and the H-theorem fail to hold. The fast algorithm in [20] reduces the computational cost of DVM to $O\left(\bar{N}^{d} N^{d} \log (N)\right)$ with some parameter $\bar{N} \ll N$ for the hard sphere molecules, but it abandons the conservation of momentum and energy.
1.2. Fourier-Galerkin method. The Fourier-based methods assume that the distribution function $f(t, v)$ is supported (in the $v$ variable) in a ball $B_{R / 2}$ centered at the origin with radius $R / 2$. Under this assumption, it makes sense to focus on the functions $f(t, v)$ with $\operatorname{Supp}(f) \subset B_{R / 2}$. For those functions, $\operatorname{Supp}(\mathcal{Q}[f, f]) \subset B_{\sqrt{2} R / 2}$ and the collision term $\mathcal{Q}[f, f]$ reduces to a truncated version $\mathcal{Q}^{R}[f, f]$ defined as

$$
\begin{equation*}
\mathcal{Q}^{R}[f, f](v):=\int_{B_{R}} \int_{B_{R}} \tilde{\mathcal{B}}(y, z) \delta(y \cdot z)[f(v+y) f(v+z)-f(v) f(v+y+z)] \mathrm{d} y \mathrm{~d} z \tag{1.17}
\end{equation*}
$$

where the superscript in $\mathcal{Q}^{R}[f, f]$ denotes the truncation radius. In order to obtain a spectral approximation to the collision term, one restricts the domain of the distribution function $f(v)$ to the cube $\mathcal{D}_{T}=[-T, T]^{d}$ with $T \geq \frac{3 \sqrt{2}+1}{4} R$ in order to reduce aliasing. One then extends it periodically to the whole space. (See $[24,19]$ for details of the derivation.) After periodization, $f(v)$ can be written as a Fourier series

$$
\begin{equation*}
f(v)=\sum_{k \in \mathbb{Z}^{d}} \hat{f}_{k} E_{k}(v), \quad \hat{f}_{k}=\frac{1}{(2 T)^{d}} \int_{\mathcal{D}_{T}} f(v) E_{-k}(v) \mathrm{d} v \tag{1.18}
\end{equation*}
$$

where $E_{k}(v)=\exp \left(\frac{\mathbf{i} \pi}{T} k \cdot v\right)$. Substituting (1.18) into (1.17) gives rise to the following representation of the truncated collision operator:

$$
\begin{equation*}
\mathcal{Q}^{R}[f, f](v)=\sum_{l, m \in \mathbb{Z}^{d}}(\hat{B}(l, m)-\hat{B}(m, m)) \hat{f}_{l} \hat{f}_{m} E_{l+m}(v), \quad v \in \mathcal{D}_{T} \tag{1.19}
\end{equation*}
$$

where

$$
\begin{equation*}
\hat{B}(l, m):=\int_{B_{R}} \int_{B_{R}} \tilde{\mathcal{B}}(y, z) \delta(y \cdot z) E_{l}(y) E_{m}(z) \mathrm{d} y \mathrm{~d} z, \quad l, m \in \mathbb{Z}^{d} \tag{1.20}
\end{equation*}
$$

It is easy to check that the coefficients $\hat{B}(l, m)$ are real and satisfy the symmetry relations

$$
\begin{equation*}
\hat{B}(l, m)=\hat{B}(m, l)=\hat{B}(l,-m) \tag{1.21}
\end{equation*}
$$

In terms of the Fourier expansion,

$$
\begin{equation*}
\hat{\mathcal{Q}}^{R}[f, f]_{k}=\sum_{l, m \in \mathbb{Z}^{d}} \mathbf{1}(l+m-k)(\hat{B}(l, m)-\hat{B}(m, m)) \hat{f}_{l} \hat{f}_{m}, \quad k \in \mathbb{Z}^{d} \tag{1.22}
\end{equation*}
$$

Here $\mathbf{1}(\cdot)$ is the indicator function, equal to 1 at the origin and 0 otherwise.
The FGM in the literature (see, e.g., [24]) starts with a finite square grid of Fourier modes

$$
\begin{equation*}
K:=\left\{k \mid k=\left(k_{1}, \ldots, k_{d}\right),-n \leq k_{1}, \ldots, k_{d} \leq n\right\} \tag{1.23}
\end{equation*}
$$

and the subspace

$$
\begin{equation*}
\mathbb{P}_{N}=\operatorname{span}\left\{E_{k}(v) \mid k \in K\right\} \subset L_{\mathrm{per}}^{2}\left(\mathcal{D}_{T}\right) \tag{1.24}
\end{equation*}
$$

We shall denote the grid points in $K$ with lowercase letters $j, k, l, m$. FGM approximates the collision term (1.22) by projecting it to the subspace $\mathbb{P}_{N}$,

$$
\begin{equation*}
\hat{Q}_{k}^{G}:=\sum_{l, m \in K} \mathbf{1}(l+m-k)(\hat{B}(l, m)-\hat{B}(m, m)) \hat{F}_{l} \hat{F}_{m}, \quad k \in K, \tag{1.25}
\end{equation*}
$$

where $\hat{F}_{k}$ for $k \in K$ serve as the approximation of the Fourier modes $\hat{f}_{k}$ of the exact solution.

Putting together the above discussion, one arrives at the equations of the discrete Fourier coefficients $\hat{F}_{k}(t)$ for $k \in K$ for FGM [24],

$$
\left\{\begin{array}{l}
\frac{\mathrm{d} \hat{F}_{k}}{\mathrm{~d} t}=\hat{Q}_{k}^{\mathrm{G}}=\sum_{l, m \in K} \mathbf{1}(l+m-k)(\hat{B}(l, m)-\hat{B}(m, m)) \hat{F}_{l} \hat{F}_{m}  \tag{1.26}\\
\hat{F}_{k}(t=0)=\hat{F}_{k}^{0}
\end{array}\right.
$$

where $\hat{F}_{k}^{0}$ are the Fourier coefficients of the initial condition $f^{0}(v)$ restricted on $\mathcal{D}_{T}$.
Remark 1.1. The above description of the FGM is based on the Carleman representation of the Boltzmann collision operator (1.7). Starting from the classical form (1.2), one can also derive a relation similar to (1.25) (see [24] for details), while the definition of $\hat{B}(l, m)$ is slightly different:

$$
\begin{equation*}
\hat{B}(l, m)=\int_{B_{R}} \int_{\mathbb{S}^{d-1}} \mathcal{B}(g, \omega) E_{l}\left(\frac{1}{2}(g+|g| \omega)\right) E_{m}\left(\frac{1}{2}(g-|g| \omega)\right) \mathrm{d} g \mathrm{~d} \omega \tag{1.27}
\end{equation*}
$$

where $T \geq \frac{3+\sqrt{2}}{4} R$ in the definition of $E_{k}(\cdot)$. It is straightforward to check that these coefficients also satisfy the symmetry relation (1.21).

FGM achieves spectral accuracy, although the computational cost is still as high as $O\left(N^{2 d}\right)$ [24]. Two fast algorithms [19, 11] reduced the cost to $O\left(M N^{d} \log (N)\right)$ for the hard sphere molecules (the Maxwell molecules for the 2D case) [19] and to $O\left(M N^{d+1} \log (N)\right)$ for general collision kernels [11], where $M$ is the number of points used for discretizing the unit sphere $\mathbb{S}^{d-1}$.

Compared to DVM, the solution of FGM loses most of the aforementioned physical properties, including positivity, the conservation of momentum and energy, and the $\mathrm{H}-$ theorem. In [25], Pareschi and Russo proposed a positivity-preserving regularization of FGM by using the Fejér filter at the expense of spectral accuracy. Despite this, the solution fails to satisfy the H-theorem. The loss of the conservation can be fixed by a spectral Lagrangian strategy [12].
1.3. Motivation. DVM preserves a number of physical properties (such as positivity of the solution, the H -theorem, and exact conservation of mass, momentum, and energy) but suffers from high computational costs and low accuracies. FGM enjoys spectral accuracies and lower computational costs but sacrifices almost all physical properties except the mass conservation. In this paper, we aim for a tradeoff between the physical properties and the spectral accuracy.

As a fundamental property of the solution to the Boltzmann equation, the positivity of the distribution function helps establish the H -theorem, which is one of the properties crucial to guaranteeing the well-posedness of the discrete system. Therefore, it makes sense to maintain the positivity and the H-theorem, as long as it does not significantly sacrifice other properties such as numerical accuracy and efficiency. This paper is an initial study in this direction.

To achieve this goal, we first carefully study the reason behind the loss of the H-theorem in FGM by comparing it with DVM. With a few novel modifications to FGM, we propose an entropic Fourier method (EFM) that preserves the positivity, the mass conservation, and the H-theorem. In addition, the computational cost of this new method is the same as that of FGM.

The rest of the paper is organized as follows. In section 2, we first outline the key steps in developing EFM and state the main results of the paper. The details of the derivation and some deeper understandings of the model are provided in section 3. Section 4 presents the implementation of EFM and the numerical results. The paper ends with a discussion in section 5 .
2. Main result. This section outlines the overall procedure of our derivation and lists some key results. Detailed derivation and investigation will be given in section 3.

Aiming at developing an entropic Fourier method for the homogeneous Boltzmann equation, one works mainly with the evolution of the Fourier coefficients $\hat{F}_{k}(t)$. Recall the discrete Fourier transform

$$
\begin{equation*}
F_{p}=\sum_{k \in K} \hat{F}_{k} E_{k}(p), \quad \hat{F}_{k}=\frac{1}{N^{d}} \sum_{p \in X} F_{p} E_{-k}(p), \tag{2.1}
\end{equation*}
$$

where $X$ defined in (1.10) and $K$ defined in (1.23) are the sets of uniform samples in the velocity space and the Fourier domain, respectively.

Using the discrete Fourier transform, one can instead treat the point values $F_{p}$ as the degrees of freedom and write the numerical scheme in the DVM form (1.11). According to the derivation in section 1.1, the following condition is required in order to guarantee the H -theorem for DVM.

Condition 1. The DVM defined in (1.11) satisfies the following:

1. the coefficients $A_{p q}^{r s}$ satisfy the symmetry relation $A_{p q}^{r s}=A_{q p}^{r s}=A_{r s}^{p q}$;
2. the coefficients $A_{p q}^{r s}$ are nonnegative, i.e., $A_{p q}^{r s} \geq 0$;
3. the initial values are nonnegative, i.e., $F_{p}(t=0) \geq 0$ for any $p \in X$.

The general idea of our approach is to revise the existing FGM so that Condition 1 is fulfilled. Below we list the steps that lead to a numerical scheme that satisfies the H -theorem.

1. Apply the Fourier collocation method to (1.19). This leads to an approximation to (1.22) in the form

$$
\begin{equation*}
\hat{Q}_{k}^{C}=\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left[\hat{B}_{N}(l, m)-\hat{B}_{N}(m, m)\right] \hat{F}_{l} \hat{F}_{m}, \quad k \in K \tag{2.2}
\end{equation*}
$$

where $\mathbf{1}_{N}(l):=\mathbf{1}(l \bmod N)$ and $\hat{B}_{N}(l, m):=\hat{B}(l \bmod N, m \bmod N)$. Here $\bmod$ is the symmetric modulo function, i.e., each component of $l \bmod N$ ranging from $-n$ to $n$ (recall $N=2 n+1$ ). Using the relation between the Fourier coefficients and the values on collocation points (2.1), we can rewrite (2.2) as

$$
\begin{equation*}
Q_{r}^{\mathrm{C}}=\sum_{p, q, s \in X} A_{p q}^{r s}\left[F_{p} F_{q}-F_{r} F_{s}\right], \quad r \in X, \tag{2.3}
\end{equation*}
$$

where $A_{p q}^{r s}$ (given in (3.8)) is determined by the Fourier modes of the collision kernel $\hat{B}_{N}(\cdot, \cdot)$ and satisfies the symmetry relation (Condition 1.1).
2. A careful study shows that $A_{p q}^{r s}$ fails to be nonnegative. This can be fixed by applying a positivity-preserving filter to $\hat{B}_{N}(l, m)$, i.e.,

$$
\begin{equation*}
\hat{B}_{N}^{\sigma}(l, m):=\hat{B}_{N}(l, m) \sigma_{N}(l) \sigma_{N}(m), \quad l, m \in K \tag{2.4}
\end{equation*}
$$

where $\sigma_{N}(l)$ is the tensor-product of $d$, the one-dimensional modified Jackson filter $[17,30]$. The modified collision term takes the following form in the Fourier domain:

$$
\hat{Q}_{k}^{\sigma}=\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left[\hat{B}_{N}^{\sigma}(l, m)-\hat{B}_{N}^{\sigma}(m, m)\right] \hat{F}_{l} \hat{F}_{m}, \quad k \in K
$$

Using $\left(A^{\sigma}\right)_{p q}^{r s}$ to denote the coefficients determined by the new kernel modes $\hat{B}_{N}^{\sigma}(l, m)$ and writing

$$
Q_{r}^{\sigma}=\sum_{p, q, s \in X}\left(A^{\sigma}\right)_{p q}^{r s}\left[F_{p} F_{q}-F_{r} F_{s}\right], \quad r \in X
$$

one can verify that both the symmetry relation (Condition 1.1) and the nonnegativity (Condition 1.2) are satisfied.
3. To guarantee the positivity of the initial values (Condition 1.3), we adopt interpolation rather than orthogonal projection while discretizing the initial distribution function.
Main result. Summarizing the outline given above, we arrive at a new entropic Fourier method (EFM) that takes the following simple form:

$$
\left\{\begin{array}{l}
\frac{\mathrm{d} \hat{F}_{k}}{\mathrm{~d} t}=\hat{Q}_{k}^{\sigma}=\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left(\hat{B}_{N}^{\sigma}(l, m)-\hat{B}_{N}^{\sigma}(m, m)\right) \hat{F}_{l} \hat{F}_{m}  \tag{2.5}\\
\hat{F}_{k}(t=0)=\frac{1}{N^{d}} \sum_{r \in X} f(t=0, r) E_{-k}(r)
\end{array}\right.
$$

This method preserves several key physical properties, as guaranteed by the following theorem.

Theorem 2.1. If $f(t=0, v) \geq 0$ for $v \in \mathbb{R}^{d}$, then the solution $F_{r}(t)=\sum_{k \in K} \hat{F}_{k}(t)$ $\cdot E_{k}(r)$ for $r \in X$ of (2.5) satisfies for all $t>0$

$$
\begin{align*}
\text { conservation of mass: } & : \frac{\mathrm{d}}{\mathrm{~d} t} \sum_{r \in X} F_{r}(t)=0  \tag{2.6}\\
\text { nonnegativity: } & F_{r}(t) \geq 0, \quad r \in X  \tag{2.7}\\
\text { discrete } H \text {-theorem: } & : \frac{\mathrm{d}}{\mathrm{~d} t} \sum_{r \in X} F_{r}(t) \ln F_{r}(t) \leq 0 . \tag{2.8}
\end{align*}
$$

The proof is presented in section 3.5. Due to the positivity-preserving filter (3.21), the numerical accuracy of EFM in approximating the collision operator is second-order (see section 3.4 for details).

Another important result for EFM is the existence of fast algorithms. For FGM, the fast algorithms proposed in $[19,11]$ are based on the approximation of the kernel $\hat{B}_{N}(\cdot, \cdot)$ :

$$
\begin{equation*}
\hat{B}_{N}(l, m) \approx \sum_{t=1}^{M} \alpha_{l+m}^{t} \beta_{l}^{t} \gamma_{m}^{t}, \quad l, m \in K \tag{2.9}
\end{equation*}
$$

with the number of terms $M \ll N^{d}$. Since the filtered kernel $\hat{B}_{N}^{\sigma}(l, m)$ turns out to have a similar approximation,

$$
\begin{equation*}
\hat{B}_{N}^{\sigma}(l, m) \approx \sum_{t=1}^{M} \alpha_{l+m}^{t}\left(\sigma_{N}(l) \beta_{l}^{t}\right)\left(\sigma_{N}(m) \gamma_{m}^{t}\right) \tag{2.10}
\end{equation*}
$$

these fast algorithms still apply. Moreover, when the above approximation is applied, the H -theorem still holds. Detailed discussion will be given in section 4.1.
3. Entropic Fourier method. As shown in section 1.1, a discrete H-theorem can be obtained from the classical DVM, where the associated entropy function can be considered as a numerical quadrature for the integral of $f \ln f$. This requires the positivity of the distribution function, which can be guaranteed by the positivity of the discrete collision kernel $A_{p q}^{r s}$. In general, to preserve the Boltzmann entropy in the numerical scheme, the positivity of the numerical solution needs to be enforced in a certain sense due to the presence of $\ln f$ in the entropy function. However, in FGM, there is no guarantee of any form of positivity in the numerical solution, and hence the H -theorem does not hold.

In this paper, rather than enforcing the nonnegativity of the whole distribution function, we take a collocation approach and focus on the nonnegativity only at the collocation points. Based on this idea, we start from a collocation method for the homogeneous Boltzmann equation and write it as a DVM of the function values defined at the collocation points. One then tries to alter the coefficients to match the requirements in Condition 1 so that the H -theorem can be subsequently derived.

The three steps listed in section 2 are detailed in the first three subsections below. After that, section 3.4 compares the entropic Fourier method (EFM) with other Fourier methods.
3.1. Fourier collocation method in a DVM form. The mechanisms of DVM and FGM are quite different: DVM is concerned with the values of the distribution on discrete points, whereas FGM (1.25), as a Galerkin method, works on the Fourier modes of the distribution function. It is not straightforward how to link these two methods. Alternatively, we will consider another type of Fourier methods-the collocation method (also known as the pseudospectral method).
3.1.1. Fourier collocation method. In the Fourier collocation method (FCM), the collision term on the set $X$ is evaluated directly using (1.19):

$$
\begin{equation*}
Q_{r}^{\mathrm{C}}=\sum_{l, m \in K}\left(\hat{B}_{N}(l, m)-\hat{B}_{N}(m, m)\right) \hat{F}_{l} \hat{F}_{m} E_{l+m}(r), \quad r \in X \tag{3.1}
\end{equation*}
$$

where $\hat{B}_{N}(l, m):=\hat{B}(l \bmod N, m \bmod N)$, and $\bmod$ is the symmetric modulo function, i.e., each component of $l \bmod N$ ranging from $-n$ to $n($ recall $N=2 n+1$ ).

Since the above equation only uses the value of $\hat{B}_{N}$ in $K$, one can use $\hat{B}$ and $\hat{B}_{N}$ interchangeably here. Here we note that $\hat{B}_{N}(l, m)$ satisfy the symmetry relation (1.21).

The corresponding Fourier modes can be obtained by an inverse discrete Fourier transform:

$$
\begin{align*}
\hat{Q}_{k}^{C} & =\frac{1}{N^{d}} \sum_{r \in X} Q_{r} E_{-k}(r)  \tag{3.2}\\
& =\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left(\hat{B}_{N}(l, m)-\hat{B}_{N}(m, m)\right) \hat{F}_{l} \hat{F}_{m}
\end{align*}
$$

where $\mathbf{1}_{N}(l):=\mathbf{1}(l \bmod N)$.
If the initial value is smooth enough, due to the smoothing effect of the Boltzmann collision operator [1], both FGM and FCM have spectral accuracy [8]. Moreover, in some cases, FCM (3.2) is numerically more efficient, especially for the fast summation algorithms in [19, 11]. For example, in [19], the following approximation of $\hat{B}_{N}(l, m)$ is considered:

$$
\begin{equation*}
\hat{B}_{N}(l, m) \approx \sum_{t=1}^{M} \beta_{l}^{t} \gamma_{m}^{t} \tag{3.3}
\end{equation*}
$$

where $M \in \mathbb{N}^{+}$is the total number of quadrature points on the sphere. Then the collision term in this Galerkin method can be approximated by

$$
\begin{equation*}
\sum_{t=1}^{M} \sum_{l, m \in K} \mathbf{1}(l+m-k)\left[\left(\beta_{l}^{t} \hat{F}_{l}\right)\left(\gamma_{m}^{t} \hat{F}_{m}\right)-\hat{F}_{l}\left(\beta_{m}^{t} \gamma_{m}^{t} \hat{F}_{m}\right)\right] \tag{3.4}
\end{equation*}
$$

To evaluate (3.4) efficiently, one needs to utilize FFT-based convolutions. To obtain these coefficients, one needs the zero-padding technique to avoid aliasing. If one uses the same method to evaluate $\hat{Q}_{k}$ in (3.2), then no zero-padding is needed. Therefore, the collocation method shortens the length of vectors used in the Fourier transform, which makes the algorithm faster.
3.1.2. DVM form. To link FCM with DVM, we split the collision term (3.1) into the gain part $(+)$ and the loss part $(-)$ :

$$
\begin{equation*}
Q_{r}^{\mathrm{C},+}=\sum_{l, m \in K} \hat{B}_{N}(l, m) \hat{F}_{l} \hat{F}_{m} E_{l+m}(r), Q_{r}^{\mathrm{C},-}=\sum_{l, m \in K} \hat{B}_{N}(m, m) \hat{F}_{l} \hat{F}_{m} E_{l+m}(r), r \in X \tag{3.5}
\end{equation*}
$$

Noticing $E_{l+m}(r)=\sum_{k \in K} \mathbf{1}_{N}(l+m-k) E_{k}(r)$ and plugging (2.1) into the gain part yields

$$
\begin{equation*}
Q_{r}^{\mathrm{C},+}=\frac{1}{N^{2 d}} \sum_{\substack{l, m, k \in K \\ p, q \in X}} \mathbf{1}_{N}(l+m-k) \hat{B}_{N}(l, m) E_{-l}(p) E_{-m}(q) E_{k}(r) F_{p} F_{q} \tag{3.6}
\end{equation*}
$$

Since $\frac{1}{N^{d}} \sum_{s \in X} E_{j}(s)=\mathbf{1}_{N}(j)$, one can sum over $j$ to get $\frac{1}{N^{d}} \sum_{j \in K, s \in X} E_{j}(s)=1$. With this equation, one introduces two new indices to (3.6) by multiplying its righthand side with $\frac{1}{N^{d}} \sum_{j \in K, s \in X} E_{j}(s)$ :
$Q_{r}^{\mathrm{C},+}=\frac{1}{N^{3 d}} \sum_{\substack{l, m, k, j \in K \\ p, q, s \in X}} \mathbf{1}_{N}(l+m-k-j) \hat{B}_{N}(l-j, m-j) E_{-l}(p) E_{-m}(q) E_{k}(r) E_{j}(s) F_{p} F_{q}$.

If one introduces

$$
\begin{equation*}
A_{p q}^{r s}=\frac{1}{N^{3 d}} \sum_{l, m, k, j \in K} \mathbf{1}_{N}(l+m-k-j) \hat{B}_{N}(l-j, m-j) E_{-l}(p) E_{-m}(q) E_{k}(r) E_{j}(s), \tag{3.8}
\end{equation*}
$$

then the gain term is

$$
\begin{equation*}
Q_{r}^{\mathrm{C},+}=\sum_{p, q, s \in X} A_{p q}^{r s} F_{p} F_{q} \tag{3.9}
\end{equation*}
$$

Apparently, such a term does take the form of the gain term of DVM (1.11).
For the loss term, the identity

$$
\begin{equation*}
\sum_{p, q \in X} A_{p q}^{r s}=\frac{1}{N^{d}} \sum_{k, j} \mathbf{1}_{N}(k+j) \hat{B}_{N}(j, j) E_{k}(r) E_{j}(s) \tag{3.10}
\end{equation*}
$$

leads to the following derivation:
(3.11)

$$
\begin{aligned}
\sum_{p, q, s \in X} A_{p q}^{r s} F_{r} F_{s} & =\frac{1}{N^{d}} \sum_{k, j \in K, s \in X} \mathbf{1}_{N}(k+j) \hat{B}_{N}(j, j) E_{k}(r) E_{j}(s) F_{r} F_{s} \\
& =\frac{1}{N^{d}} \sum_{k, j, l, m \in K, s \in X} \mathbf{1}_{N}(k+j) \hat{B}_{N}(j, j) E_{k}(r) E_{j}(s) E_{l}(r) E_{m}(s) \hat{F}_{l} \hat{F}_{m} \\
& =\sum_{k, j, l, m \in K} \mathbf{1}_{N}(k+j) \mathbf{1}_{N}(j+m) \hat{B}_{N}(j, j) \hat{F}_{l} \hat{F}_{m} E_{k+l}(r) \\
& =\sum_{l, m \in K} \hat{B}_{N}(m, m) \hat{F}_{l} \hat{F}_{m} E_{l+m}(r)=Q_{r}^{\mathrm{C},-}
\end{aligned}
$$

In summary, one can write FCM in the following DVM form:

$$
\begin{equation*}
Q_{r}^{\mathrm{C}}=\sum_{p, q, s \in X} A_{p q}^{r s}\left(F_{p} F_{q}-F_{r} F_{s}\right) \tag{3.12}
\end{equation*}
$$

with $A_{p q}^{r s}$ given in (3.8). Finally, the symmetry relation (Condition 1.1)

$$
\begin{equation*}
A_{p q}^{r s}=A_{q p}^{r s}=A_{r s}^{p q} \tag{3.13}
\end{equation*}
$$

holds, as this can be easily seen by the symmetry relation of $\hat{B}_{N}(l, m)$ and switching the indices in (3.8).
3.2. Positivity preservation. As remarked earlier, in order to obtain an H theorem for FCM, one needs to ensure that all the coefficients $A_{p q}^{r s}$ are nonnegative (Condition 1.2). Below, we first show that $A_{p q}^{r s}$ as defined in (3.12) fail to be nonnegative, and then we apply a filter to recover nonnegativity.
3.2.1. Failure of positivity preservation in FCM. We start by simplifying the coefficients $A_{p q}^{r s}$ based on (3.8):

$$
\begin{align*}
A_{p q}^{r s} & =\frac{1}{N^{3 d}} \sum_{l, m, k, j \in K} \mathbf{1}_{N}(l+m-k-j) \hat{B}_{N}(l-j, m-j) E_{-l}(p) E_{-m}(q) E_{k}(r) E_{j}(s)  \tag{3.14}\\
& =\frac{1}{N^{3 d}} \sum_{l, m, k \in K} \hat{B}_{N}(m-k, l-k) E_{-l}(p-s) E_{-m}(q-s) E_{k}(r-s)
\end{align*}
$$

where one uses $j=l+m-k \bmod N$. By performing a change of variables $i=$ $(m-k) \bmod N$ and $j=(l-k) \bmod N$, we arrive at

$$
\begin{aligned}
A_{p q}^{r s} & =\frac{1}{N^{3 d}} \sum_{i, j, k \in K} \hat{B}_{N}(i, j) E_{-k-i}(p-s) E_{-k-j}(q-s) E_{k}(r-s) \\
& =\mathbf{1}_{N}(r+s-p-q) \frac{1}{N^{2 d}} \sum_{i, j \in K} \hat{B}_{N}(i, j) E_{-i}(p-s) E_{-j}(q-s) .
\end{aligned}
$$

By introducing

$$
\begin{equation*}
G(y, z)=\sum_{i, j \in K} \hat{B}_{N}(i, j) E_{-i}(y) E_{-j}(z), \quad y, z \in \mathcal{D}_{T} \tag{3.16}
\end{equation*}
$$

which is by definition a periodic function with period $\mathcal{D}_{T}$, one can write compactly

$$
\begin{equation*}
A_{p q}^{r s}=\frac{1}{N^{2 d}} \mathbf{1}_{N}(r+s-p-q) G(p-s, q-s), \quad p, q, r, s \in X \tag{3.17}
\end{equation*}
$$

In order to check whether $A_{p q}^{r s}$ is nonnegative, one just needs to check whether $G(\cdot, \cdot)$ is nonnegative on the collocation points in $X$. To get a better understanding of the function $G(\cdot, \cdot)$ as defined in (3.16), one applies the definition of $\hat{B}_{N}(\cdot, \cdot)$ to obtain

$$
\begin{equation*}
G(y, z)=\int_{B_{R}} \int_{B_{R}} \tilde{\mathcal{B}}\left(y^{\prime}, z^{\prime}\right) \delta\left(y^{\prime} \cdot z^{\prime}\right) \chi_{N}\left(y-y^{\prime}\right) \chi_{N}\left(z-z^{\prime}\right) \mathrm{d} y^{\prime} \mathrm{d} z^{\prime} \tag{3.18}
\end{equation*}
$$

Here $\chi_{N}$ is the Dirichlet kernel over $\mathcal{D}_{T}$ defined by

$$
\begin{equation*}
\chi_{N}(v)=\sum_{k \in K} E_{k}(v), \quad v \in \mathcal{D}_{T} \tag{3.19}
\end{equation*}
$$

and its discrete Fourier transform $\hat{\chi}_{N}(k)$ is equal to 1 for $k \in K$ and 0 on $\mathbb{Z}^{d} \backslash K$. By introducing a periodic function in $\mathcal{D}_{T}$

$$
\begin{equation*}
H(y, z)=\tilde{\mathcal{B}}(y, z) \delta(y \cdot z) \mathbf{1}(|y| \leq R) \mathbf{1}(|z| \leq R), \quad y, z \in \mathcal{D}_{T} \tag{3.20}
\end{equation*}
$$

one can write

$$
G=H *\left(\chi_{N} \otimes \chi_{N}\right)
$$

where the convolution is defined periodically in $\mathcal{D}_{T} \times \mathcal{D}_{T}$. Equivalently, $G(y, z)$ is also the truncated Fourier expansion of $H(y, z)$ by keeping only the frequencies in $K$.

Although $H(y, z)$ is nonnegative in the weak sense, its truncated Fourier approximation $G(y, z)$ fails to be so. For example, the values of $G$ for the kernel $\tilde{\mathcal{B}}(y, z) \equiv \frac{1}{\pi}$, $R=6$ in 2D are plotted in Figure 1. This clearly shows that negative values appear as expected. Therefore, in general, the H-theorem does not hold for FCM.
3.2.2. Filtering. In the previous subsection, one can see that if $\chi_{N}(\cdot)$ were a nonnegative function, then $G(y, z)$ would be nonnegative for any $y$ and $z$. Thus, in order to get nonnegative coefficients, a possible way is to replace the function $\chi_{N}$ by a nonnegative one. Note that $\chi_{N}(v)$ is a Dirichlet kernel, which is an approximation of the Dirac delta. As $N \rightarrow+\infty$, the function $\chi_{N}(\cdot)$ tends to the Dirac delta weakly in an oscillatory way. As pointed out in [10], the oscillation breaks the nonnegativity of the solution.


FIG. 1. The values of $G(y, z)$ with $N=32$ at $z=(T / 2, T / 2)$. The axes are the two components of $y$.

As mentioned earlier in section 2, we adopt the one-dimensional modified Jackson filter $[17,30]$ given by

$$
\begin{equation*}
\sigma_{N}(\beta)=\frac{(n+1-|\beta|) \cos \left(\frac{\pi|\beta|}{n+1}\right)+\sin \left(\frac{\pi|\beta|}{n+1}\right) \cot \left(\frac{\pi}{n+1}\right)}{n+1} \tag{3.21}
\end{equation*}
$$

where $N=2 n+1$ and $-n \leq \beta \leq n$. By a slight abuse of notation, the $d$-dimensional modified Jackson filter for a multi-index $k=\left(k_{1}, \ldots, k_{d}\right) \in K$ is defined through tensor product

$$
\sigma_{N}(k)=\prod_{i=1}^{d} \sigma_{N}\left(k_{i}\right)
$$

The modified kernel $\chi_{N}^{\sigma}(v)$ can then be defined as

$$
\chi_{N}^{\sigma}(v)=\sum_{k \in K} \sigma_{N}(k) E_{k}(v), \quad v \in \mathcal{D}_{T}
$$

Once $\chi_{N}$ is replaced with $\chi_{N}^{\sigma}$ in (3.18), the function $G(y, z)$ is substituted with $G^{\sigma}(y, z):=\left(G *\left(\chi_{N}^{\sigma} \otimes \chi_{N}^{\sigma}\right)\right)(y, z)$. A direct calculation shows that

$$
\begin{align*}
G^{\sigma}(y, z) & =\frac{1}{N^{2 d}} \int_{B_{R}} \int_{B_{R}} \tilde{\mathcal{B}}\left(y^{\prime}, z^{\prime}\right) \delta\left(y^{\prime} \cdot z^{\prime}\right) \chi_{N}^{\sigma}\left(y-y^{\prime}\right) \chi_{N}^{\sigma}\left(z-z^{\prime}\right) \mathrm{d} y^{\prime} \mathrm{d} z^{\prime}  \tag{3.22}\\
& =\frac{1}{N^{2 d}} \sum_{l, m \in K} \int_{B_{R}} \int_{B_{R}} \tilde{\mathcal{B}}\left(y^{\prime}, z^{\prime}\right) \delta\left(y^{\prime} \cdot z^{\prime}\right) \sigma_{N}(l) \sigma_{N}(m) E_{l}\left(y-y^{\prime}\right) E_{m}\left(z-z^{\prime}\right) \mathrm{d} y^{\prime} \mathrm{d} z^{\prime} \\
& =\frac{1}{N^{2 d}} \sum_{l, m \in K}\left[\sigma_{N}(l) \sigma_{N}(m) \hat{B}_{N}(l, m)\right] E_{-l}(y) E_{-m}(z) \\
& =\frac{1}{N^{2 d}} \sum_{l, m \in K} \hat{B}_{N}^{\sigma}(l, m) E_{-l}(y) E_{-m}(z)
\end{align*}
$$

where $\hat{B}_{N}^{\sigma}(l, m):=\hat{B}_{N}(l, m) \sigma_{N}(l) \sigma_{N}(m)$ as defined in (2.4). With $G^{\sigma}(y, z) \geq 0$ guaranteed, one can mimic (3.17) and define

$$
\begin{equation*}
\left(A^{\sigma}\right)_{p q}^{r s}=\frac{1}{N^{2 d}} \mathbf{1}_{N}(r+s-p-q) G^{\sigma}(p-s, q-s), \quad p, q, r, s \in X \tag{3.23}
\end{equation*}
$$

which are apparently nonnegative. Since replacing $\hat{B}_{N}(l, m)$ with $\hat{B}_{N}^{\sigma}(l, m)$ does not affect the symmetry relation (1.21), the new coefficients $\left(A^{\sigma}\right)_{p q}^{r s}$ also satisfy $\left(A^{\sigma}\right)_{p q}^{r s}=$ $\left(A^{\sigma}\right)_{q p}^{r s}=\left(A^{\sigma}\right)_{r s}^{p q}$.

From the above discussion, we now define the entropic collision term to be

$$
\begin{equation*}
\hat{Q}_{k}^{\sigma}:=\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left(\hat{B}_{N}^{\sigma}(l, m)-\hat{B}_{N}^{\sigma}(m, m)\right) \hat{F}_{l} \hat{F}_{m}, \quad k \in K, \tag{3.24}
\end{equation*}
$$

in the Fourier domain. In the velocity domain, it is equal to

$$
\begin{equation*}
Q_{r}^{\sigma}:=\sum_{p, q, s \in X}\left(A^{\sigma}\right)_{p q}^{r s}\left(F_{p} F_{q}-F_{r} F_{s}\right), \quad r \in X . \tag{3.25}
\end{equation*}
$$

3.3. Initial condition. In order to obtain the H -theorem, one needs to make sure that the initial data are nonnegative. Since the collision term in (3.25) depends only on the values at the collocation points in $X$, it is sufficient to make the initial data nonnegative at these collocation points. Consequently, it is natural to use sampling rather than orthogonal projection while preparing the discrete initial data $F_{r}^{0}$ for $r \in X$. More precisely,

$$
F_{r}^{0}=\mathcal{I}_{N} f^{0}:= \begin{cases}f^{0}(r), & f^{0} \text { is continuous, }  \tag{3.26}\\ \left(\varphi^{\epsilon} * f^{0}\right)(r) & \text { otherwise },\end{cases}
$$

where $\varphi^{\epsilon} \geq 0$ is a mollifier, such that $\left\|f^{0}-\varphi^{\epsilon} * f^{0}\right\|_{L^{2}}<\epsilon$ for $\epsilon$ sufficiently small. Once $\left\{F_{r}^{0}\right\}$ are ready, the corresponding Fourier coefficients $\left\{\hat{F}_{k}^{0}\right\}$ are computed via a fast Fourier transform.

At this point, all ingredients of the entropic Fourier method (EFM) are ready. The Cauchy problem of EFM takes the following form as a DVM:

$$
\left\{\begin{array}{l}
\frac{\mathrm{d} F_{r}}{\mathrm{~d} t}=Q_{r}^{\sigma}=\sum_{p, q, s \in X}\left(A^{\sigma}\right)_{p q}^{r s}\left(F_{p} F_{q}-F_{r} F_{s}\right),  \tag{3.27}\\
F_{r}(t=0)=F_{r}^{0} .
\end{array}\right.
$$

Equivalently, in the Fourier domain, EFM takes the form

$$
\left\{\begin{array}{l}
\frac{\mathrm{d} \hat{F}_{k}}{\mathrm{~d} t}=\hat{Q}_{k}^{\sigma}=\sum_{l, m \in K} \mathbf{1}_{N}(l+m-k)\left(\hat{B}_{N}^{\sigma}(l, m)-\hat{B}_{N}^{\sigma}(m, m)\right) \hat{F}_{l} \hat{F}_{m},  \tag{3.28}\\
\hat{F}_{k}(t=0)=\hat{F}_{k}^{0} .
\end{array}\right.
$$

Remark 3.1. The same technique can be applied to the Fourier method derived from the classical form of the Boltzmann collision operator (1.2) to obtain an entropic Fourier method. In fact, from (1.27) and following the definition of $G(y, z)$ in (3.16), one can directly obtain
$G(y, z)=\left(H *\left(\chi_{N} \otimes \chi_{N}\right)\right)(y, z), \quad H\left(g, g^{\prime}\right)=\mathcal{B}(g, \omega) \delta\left(|g|-\left|g^{\prime}\right|\right) \mathbf{1}(|g| \leq R), \quad g, g^{\prime} \in \mathcal{D}_{T}$.
Again, by (3.17), the positivity of $A_{p q}^{r s}$ depends only on the positivity of $G(y, z)$ at the collocation points. Therefore, replacing $\chi_{N}$ with $\chi_{N}^{\sigma}$ does the job.
3.4. Comparison. The derivation of EFM switched frequently between the language of Fourier methods and DVM for different purposes. As we have shown in (3.27) and (3.28), EFM can be regarded either as a Fourier method or as a special DVM.

In what follows, we provide a comparison between the EFM (entropic Fourier method), the FGM (Fourier-Galerkin method), and the Fourier collocation method $(\mathrm{FCM})$. To set up a uniform notation, let $\mathcal{Q}[\cdot ; \cdot, \cdot]$ be the general collision operator

$$
\begin{equation*}
\mathcal{Q}[C ; f, f](v)=\int_{\mathcal{D}_{T}} \int_{\mathcal{D}_{T}} C(y, z)[f(v+y) f(v+z)-f(v) f(v+y+z)] \mathrm{d} y \mathrm{~d} z \tag{3.29}
\end{equation*}
$$

with a collision kernel $C(\cdot, \cdot)$. Thus the truncated collision term (1.17) can be written as $\mathcal{Q}(H ; f, f)$ using the definition of $H$ in (3.20).

Notice that a special feature of a function in $\mathbb{P}_{N}$ is that it is uniquely defined via its function values at points in $X$ defined in (1.10). Therefore, $\left\{F_{p} \mid p \in X\right\}$ can be regarded as both a discrete set of values and the samples from the smooth periodic $F(v) \in \mathbb{P}_{N} \subset L_{\text {per }}^{2}\left(\mathcal{D}_{T}\right)$. By introducing two operators

$$
\mathcal{P}_{N}: f \rightarrow \chi_{N} * f, \quad \mathcal{S}_{N}^{\sigma}: f \rightarrow \chi_{N}^{\sigma} * f
$$

for the space $L_{\text {per }}^{2}\left(\mathcal{D}_{T}\right)$, the three methods are different approximations of $\mathcal{Q}(H ; f, f)$ with different initial values:

$$
\begin{array}{ll}
\mathrm{FGM}: \mathcal{P}_{N} \mathcal{Q}\left[\left(\mathcal{P}_{N} \otimes \mathcal{P}_{N}\right) H ; F, F\right], & F(t=0, v)=\mathcal{P}_{N} f(t=0, v), \\
\mathrm{FCM}: \mathcal{I}_{N} \mathcal{Q}\left[\left(\mathcal{P}_{N} \otimes \mathcal{P}_{N}\right) H ; F, F\right], & F(t=0, v)=\mathcal{P}_{N} f(t=0, v), \\
\mathrm{EFM}: \overline{\mathcal{I}_{N} \mathcal{Q}\left[\left(\mathcal{S}_{N}^{\sigma} \otimes \mathcal{S}_{N}^{\sigma}\right) H ; F, F\right],} & F(t=0, v)=\underline{\mathcal{I}_{N}} f(t=0, v), \tag{3.32}
\end{array}
$$

where $\mathcal{I}_{N}$ is the interpolation operator defined in (3.26).
The list (3.30)-(3.32) clearly shows how we change from FGM to EFM in our derivation. The last line (3.32) also shows that EFM provides an approximation of the original binary collision operator in the language of spectral methods. Below we will briefly review the basic properties of all three methods.

The method (3.30) stands for FGM as described in section 1.2. In the derivation, the kernel $K$ is not explicitly projected. However, (1.25) shows that the discrete collision operator depends only on $\left(\mathcal{P}_{N} \times \mathcal{P}_{N}\right) H$. By replacing the projection operator applied to $\mathcal{Q}$ with interpolation as in (3.2), we arrive at the Fourier collocation method (3.31) introduced in section 3.1. Since a direct projection of $H$ does not preserve the positivity of the kernel, the negative part of the discrete kernel may cause a violation of the H-theorem. Nevertheless, both of these methods have spectral accuracy in the velocity space.

To ensure the positivity of the discrete kernel, the filter $\mathcal{S}_{N}^{\sigma} \otimes \mathcal{S}_{N}^{\sigma}$ is applied in (3.32), and thus positive coefficients (3.23) are obtained. The method (3.32) also ensures the positivity of the approximation of $F$ at collocation points, and thus the discrete H-theorem follows.

However, the filter $S_{N}^{\sigma}$ has a smearing effect, which reduces the order of convergence. For any smooth periodic function $f \in L_{\text {per }}^{2}\left(\mathcal{D}_{T}\right)$, the $L^{2}$-error $\left\|f-\mathcal{S}_{N}^{\sigma} f\right\|_{2}$ is $O\left(N^{-2}\right)[16$, Chapter 4], and therefore EFM is at most second-order. On the other hand, if one splits the collision term of EFM into the gain part and the loss part and
lets $F=\mathcal{P}_{N} f$, then

$$
\begin{align*}
& Q^{\sigma,+}[F, F](r)=\sum_{l, m \in K} \hat{B}_{N}(l, m) \sigma_{N}(l) \hat{F}_{l} \sigma_{N}(m) \hat{F}_{m} E_{l+m}(r)=Q^{\mathrm{C},+}\left[\mathcal{S}_{N}^{\sigma} F, \mathcal{S}_{N}^{\sigma} F\right](r),  \tag{3.33}\\
& 3.34)  \tag{3.34}\\
& Q^{\sigma,-}[F, F](r)=\sum_{l, m \in K} \hat{B}_{N}(m, m) \hat{F}_{l} \sigma_{N}^{2}(m) \hat{F}_{m} E_{l+m}(r)=Q^{\mathrm{C},-}\left[F, \mathcal{S}_{N}^{\sigma} \mathcal{S}_{N}^{\sigma} F\right](r)
\end{align*}
$$

for any $r \in X$. Following the boundedness of the truncated collision operator proven in [24], one concludes that
$\left\|Q^{\sigma}[F, F]-Q^{\mathrm{C}}[F, F]\right\|_{2} \leq\left\|Q^{\sigma,+}[F, F]-Q^{\mathrm{C},+}[F, F]\right\|_{2}+\left\|Q^{\sigma,-}[F, F]-Q^{\mathrm{C},-}[F, F]\right\|_{2} \leq O\left(N^{-2}\right)$.
Hence, EFM has second-order accuracy in approximating the truncated collision operator if the distribution is smooth enough (due to the smoothing effect of the Boltzmann collision operator [1], we only need that the initial value is smooth enough). This order of convergence will be also numerically verified in the next section.
3.5. Proof of Theorem 2.1. This subsection provides the proof of Theorem 2.1.

Proof of Theorem 2.1. The argument in section 3.2.2 indicates that if $f(t=$ $0, v) \geq 0$, then the coefficients $A_{p q}^{r s}$ and the initial values $F_{r}^{0}$ satisfy all three conditions in Condition 1.

The symmetry relation (1.21) of $\hat{B}(l, m)$ and the definition of $\hat{B}_{N}^{\sigma}(l, m)$ indicate $\mathbf{1}_{N}(l+m)\left(\hat{B}_{N}^{\sigma}(l, m)-\hat{B}_{N}^{\sigma}(m, m)\right)=0$, i.e., $Q_{0}^{\sigma}=0$. Noticing that the zero frequency $\hat{F}_{0}(t)=\frac{1}{N^{d}} \sum_{r \in X} F_{r}(t)$, one can directly obtain the conservation of mass (2.6).

Since $f^{0}(v) \geq 0, F_{r}^{0} \geq 0$ by construction. If there exist $t^{\prime}>0$ and $r \in X$ such that $F_{r}\left(t^{\prime}\right)=0$ and $F_{p}\left(t^{\prime}\right) \geq 0$ for any other $p \in X$, then

$$
\left.\frac{\mathrm{d} F_{r}(t)}{\mathrm{d} t}\right|_{t=t^{\prime}}=\sum_{p, q, s \in X}\left(A^{\sigma}\right)_{p q}^{r s} F_{p} F_{q} \geq 0
$$

which indicates $F_{r}(t) \geq 0$ for all $t>0$ and $r \in X$.
The symmetry relation and nonnegativity of $\left(A^{\sigma}\right)_{p q}^{r s}$ indicate the discrete H theorem (2.8)

$$
\sum_{r \in X} Q_{r}^{\sigma} \ln \left(F_{r}\right)=\frac{1}{4} \sum_{p, q, r, s \in X}\left(A^{\sigma}\right)_{p q}^{r s} \ln \left(\frac{F_{r} F_{s}}{F_{p} F_{q}}\right)\left[F_{p} F_{q}-F_{r} F_{s}\right] \leq 0
$$

This completes the proof.
4. Numerical tests. This section describes several numerical tests to demonstrate the properties of EFM and to compare with the FGM in [24] and the positivitypreserving spectral method (PPSM) in [25].
4.1. Implementation. It is pointed out in section 2 that the fast algorithms in $[19,11]$ can be applied to EFM without affecting the H-theorem. To show this, one needs to check that the fast algorithms do not violate the first two conditions in Condition 1.

These fast algorithms are based on an approximation of $\hat{B}_{N}(l, m)(2.10)$ of the following form:

$$
\begin{equation*}
\hat{B}_{N}(l, m) \approx \hat{B}_{N, \text { fast }}(l, m):=\sum_{t=1}^{M} \alpha_{l+m}^{t} \beta_{l}^{t} \gamma_{m}^{t} \tag{4.1}
\end{equation*}
$$

After the filtering is applied, one obtains a similar approximation for $\hat{B}_{N}^{\sigma}(l, m)$ :

$$
\begin{equation*}
\hat{B}_{N}^{\sigma}(l, m) \approx \hat{B}_{N, \text { fast }}^{\sigma}(l, m):=\sum_{t=1}^{M} \alpha_{l+m}^{t}\left(\sigma_{N}(l) \beta_{l}^{t}\right)\left(\sigma_{N}(m) \gamma_{m}^{t}\right) \tag{4.2}
\end{equation*}
$$

It can be verified that both kernels satisfy the symmetry relation

$$
\begin{align*}
& \hat{B}_{N, \mathrm{fast}}(l, m)=\hat{B}_{N, \mathrm{fast}}(m, l)=\hat{B}_{N, \mathrm{fast}}(-l, m),  \tag{4.3}\\
& \hat{B}_{N, \mathrm{fast}}^{\sigma}(l, m)=\hat{B}_{N, \mathrm{fast}}^{\sigma}(m, l)=\hat{B}_{N, \mathrm{fast}}^{\sigma}(-l, m), \tag{4.4}
\end{align*}
$$

which indicates that Condition 1.1 is valid for the fast algorithms.
To see that the fast algorithms do not affect the nonnegativity of $G^{\sigma}(y, z)$, we use the fast algorithm in [19] with $d=2$ and $\tilde{B}=1$ as an example. The first step of this algorithm writes $y$ and $z$ in (1.20) in the polar coordinates $y=\rho e$ and $z=\rho_{*} e_{*}$ :

$$
\begin{equation*}
\hat{B}_{N}(l, m)=\frac{1}{4} \int_{\mathbb{S}^{1}} \int_{\mathbb{S}^{1}} \delta\left(e \cdot e_{*}\right)\left[\int_{-R}^{R} E_{l}(\rho e) \mathrm{d} \rho\right]\left[\int_{-R}^{R} E_{m}\left(\rho_{*} e_{*}\right) \mathrm{d} \rho_{*}\right] \mathrm{d} e \mathrm{~d} e_{*} . \tag{4.5}
\end{equation*}
$$

Let $\psi_{R}(l, e)=\int_{-R}^{R} E_{l}(\rho e) \mathrm{d} \rho$; then

$$
\hat{B}_{N}(l, m)=\frac{1}{4} \int_{\mathbb{S}^{1}} \int_{\mathbb{S}^{1}} \delta\left(e \cdot e_{*}\right) \psi_{R}(l, e) \psi_{R}\left(m, e_{*}\right) \mathrm{d} e \mathrm{~d} e_{*}
$$

Integrating it with respect to $e_{*}$ yields

$$
\begin{equation*}
\hat{B}_{N}(l, m)=\int_{0}^{\pi} \psi_{R}\left(l, e_{\theta}\right) \psi_{R}\left(m, e_{\theta+\pi / 2}\right) \mathrm{d} \theta \tag{4.6}
\end{equation*}
$$

Substituting (4.5) into (3.22) gives rise to

$$
\begin{equation*}
G^{\sigma}(y, z)=\frac{1}{4} \int_{\mathbb{S}^{1}} \int_{\mathbb{S}^{1}} \delta\left(e \cdot e_{*}\right)\left[\int_{-R}^{R} \chi_{N}^{\sigma}(\rho e-y) \mathrm{d} \rho\right]\left[\int_{-R}^{R} \chi_{N}^{\sigma}\left(\rho e_{*}-z\right) \mathrm{d} \rho_{*}\right] \mathrm{d} e \mathrm{~d} e_{*} \tag{4.7}
\end{equation*}
$$

Let $\phi_{R}^{\sigma}(y, e)=\int_{-R}^{R} \chi_{N}^{\sigma}(\rho e-y) \mathrm{d} \rho$. Apparently, $\phi_{R}^{\sigma}(y, e) \geq 0$ due to $\chi_{N}^{\sigma}(y) \geq 0$ for any $y \in \mathbb{R}^{2}$. Then integrating (4.7) with respect to $e_{*}$ yields

$$
\begin{equation*}
G^{\sigma}(y, z)=\int_{0}^{\pi} \phi_{R}^{\sigma}\left(y, e_{\theta}\right) \phi_{R}^{\sigma}\left(z, e_{\theta+\pi / 2}\right) \mathrm{d} \theta \tag{4.8}
\end{equation*}
$$

The idea of the fast algorithm is to replace the integration in (4.6) with a quadrature formula. More precisely, (4.6) is approximated by

$$
\begin{equation*}
\hat{B}_{N, \text { fast }}(l, m)=\sum_{t=1}^{M} \frac{\pi}{M} \psi_{R}\left(l, e_{\theta_{t}}\right) \psi_{R}\left(m, e_{\theta_{t}+\pi / 2}\right) \tag{4.9}
\end{equation*}
$$

Similarly to (4.9), one obtains

$$
\begin{equation*}
G_{\text {fast }}^{\sigma}(y, z)=\sum_{t=1}^{M} \frac{\pi}{M} \phi_{R}^{\sigma}\left(y, e_{\theta_{t}}\right) \phi_{R}^{\sigma}\left(z, e_{\theta_{t}+\pi / 2}\right) \tag{4.10}
\end{equation*}
$$

Since $\phi_{R}^{\sigma}(y, e) \geq 0$ for any $y \in \mathbb{R}^{2}, e \in \mathbb{S}^{1}, G_{\text {fast }}^{\sigma}(y, z) \geq 0$ for any $y, z \in \mathbb{R}^{2}$. Hence, the fast algorithm does not destroy the nonnegativity of $G_{\text {fast }}^{\sigma}(y, z)$.

As we pointed out in section 3.1, an aliased convolution can be directly used to calculate (2.5). Since the accuracy of EFM is only second-order, the smoothing filter is the main source of the error. In the fast algorithms, the number $M$ in (4.1) perhaps can be smaller than that in [19, 11].

In the above discussion, we only study the case when $N$ is odd. The case of even $N$ values can be reduced to the odd $(N-1)$ case by setting the coefficient of a mode $k$ to be 0 if any component of $k=\left(k_{1}, \ldots, k_{d}\right)$ is equal to $-N / 2$.

For the time discretization, the third-order strong stability-preserving RungeKutta method proposed in [14] is employed in the discretization of time. In all tests, the time step is chosen as $\Delta t=0.01$.
4.2. Numerical results. The test problems used here are solutions of the spacehomogeneous Boltzmann equation for Maxwell molecules $\left(\mathcal{B}(g, \omega)=\frac{1}{2 \pi}\right.$ in 2D and $\mathcal{B}(g, \omega)=\frac{1}{4 \pi}$ in 3D).

Example 1 (2D BKW solution). The first example is the well-known 2D Bobylev-Krook-Wu (BKW) solution, obtained independently in [3] and [15]. The exact solution takes the form

$$
\begin{equation*}
f(t, v)=\frac{1}{2 \pi S} \exp \left(-\frac{|v|^{2}}{2 S}\right)\left(\frac{2 S-1}{S}+\frac{1-S}{2 S^{2}}|v|^{2}\right), \tag{4.11}
\end{equation*}
$$

where $S=1-\exp (-t / 8) / 2$. The BKW solution allows one to check the accuracy, positivity of the solution, and the entropy of the proposed method. Here we set the truncation radius $R=6$ in the tests.


FIG. 2. Positivity error of FGM and FCM with the initial value prepared by orthogonal projection ( $P$ ) and interpolation (I) in the $\log _{10}$ scale. Since the positivity error of EFM is strictly zero, its result is not plotted in the figure.

Both FGM (1.26) and FCM (3.2) result in good approximations of the exact solution at the collocation points. However, the solutions are not nonnegative. Even if we use interpolation rather than orthogonal projection to prepare the initial values, the solutions of these methods still fail to be nonnegative. The positivity of the
solution is measured by the positivity error defined via

$$
\begin{equation*}
\epsilon:=\frac{\sum_{q \in X}\left|F_{q}\right|-\sum_{q \in X} F_{q}}{\sum_{q \in X}\left|F_{q}\right|} . \tag{4.12}
\end{equation*}
$$

Figure 2 shows that both FGM and FCM fail to preserve the positivity of the solution at the collocation points regardless of whether the initial value is given by orthogonal projection or interpolation. On the other hand, the positivity error of EFM is strictly zero, thanks to the modification in (3.22).

Table 1 summarizes the $\ell_{1}, \ell_{2}$, and $\ell_{\infty}$ errors of (3.27) at time $t=0.01$. Here the $\ell_{\mathrm{p}}$ relative errors for $\mathrm{p}=1,2, \infty$ are defined by

$$
\begin{equation*}
\frac{\|F-f\|_{\mathrm{p}}}{\|f\|_{\mathrm{p}}}=\frac{\left(\sum_{q \in X}\left|F_{q}-f(q)\right|^{\mathrm{p}}\right)^{1 / \mathrm{p}}}{\left(\sum_{q \in X}|f(q)|^{\mathrm{p}}\right)^{1 / \mathrm{p}}} \tag{4.13}
\end{equation*}
$$

where $f(q)$ is the exact solution at $q \in X$. The numerical results also show that the convergence rate of EFM is of second order.

TABLE 1
The $\ell_{1}, \ell_{2}$, and $\ell_{\infty}$ errors and convergence rates for the $B K W$ solution at time $t=0.01$ with $R=6$.

| $N$ | $\ell_{1}$ error | Rate | $\ell_{2}$ error | Rate | $\ell_{\infty}$ error | Rate |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 16 | $4.68 \times 10^{-3}$ |  | $3.23 \times 10^{-3}$ |  | $3.12 \times 10^{-3}$ |  |
| 32 | $1.72 \times 10^{-3}$ | 1.44 | $1.36 \times 10^{-3}$ | 1.25 | $1.40 \times 10^{-3}$ | 1.15 |
| 64 | $5.54 \times 10^{-4}$ | 1.64 | $4.56 \times 10^{-4}$ | 1.58 | $5.57 \times 10^{-4}$ | 1.34 |
| 128 | $1.55 \times 10^{-4}$ | 1.84 | $1.29 \times 10^{-4}$ | 1.82 | $1.73 \times 10^{-4}$ | 1.68 |
| 256 | $4.05 \times 10^{-5}$ | 1.93 | $3.42 \times 10^{-5}$ | 1.92 | $4.73 \times 10^{-5}$ | 1.87 |
| 512 | $1.03 \times 10^{-5}$ | 1.97 | $8.76 \times 10^{-6}$ | 1.96 | $1.22 \times 10^{-5}$ | 1.94 |

As discussed in section 4.1, the fast algorithm in [19] can be applied to EFM to accelerate the computation. In (4.6), the integration on $[0, \pi)$ can be reduced to $[0, \pi / 2)$ and $M$ is equal to the number of samples within $[0, \pi / 2)$. Table 2 presents the $\ell_{1}$ error for multiple values of $M$; notice that $M=2$ is good enough in practice, while in [19] the authors suggest $M \geq 4$.

Table 2
The $\ell_{1}$ error of $E F M$ with fast algorithm in [19] for multiple choices of $N$ and $M$.

| $N$ | $M=2$ | $M=3$ | $M=32$ |
| :---: | :---: | :---: | :---: |
| 16 | $4.6852 \times 10^{-3}$ | $4.6826 \times 10^{-3}$ | $4.6830 \times 10^{-3}$ |
| 32 | $1.7241 \times 10^{-3}$ | $1.7244 \times 10^{-3}$ | $1.7245 \times 10^{-3}$ |
| 64 | $5.5368 \times 10^{-4}$ | $5.5388 \times 10^{-4}$ | $5.5394 \times 10^{-4}$ |
| 128 | $1.5485 \times 10^{-4}$ | $1.5488 \times 10^{-4}$ | $1.5489 \times 10^{-4}$ |
| 256 | $4.0513 \times 10^{-5}$ | $4.0516 \times 10^{-5}$ | $4.0517 \times 10^{-5}$ |

In order to demonstrate that the proposed method satisfies the H -theorem numerically, we define a time-dependent discrete entropy function

$$
\begin{equation*}
\eta(t)=\left(\frac{2 T}{N}\right)^{d} \sum_{q \in X} F_{q}(t) \ln F_{q}(t) \tag{4.14}
\end{equation*}
$$

The evolution of the entropy, plotted in Figure 3, shows that as the number of discrete points $N$ increases, the discrete entropy converges to the one of the exact solution.


Fig. 3. The evolution of the entropy of EFM for multiple $N$ values.


Fig. 4. Comparison between PPSM and EFM at time $t=1$ with $N=64$ for the BKW solution.

As a comparison with PPSM, Figure 4 presents the numerical solutions in the $v_{1}$ direction of PPSM and EFM at $t=1$ with $N=32$. The smoothing filter used for EFM results in much less dissipation, thus leading to much better agreement with the exact solution. Finally, Figure 5 shows that as $N$ increases, the solution of EFM converges rapidly to the exact solution.

Example 2 (bi-Gaussian initial value). Another frequent example is a problem with the bi-Gaussian initial value

$$
\begin{equation*}
f(t=0, v)=\frac{1}{4 \pi}\left(\exp \left(-\frac{\left|v-u_{1}\right|^{2}}{2}\right)+\exp \left(-\frac{\left|v-u_{2}\right|^{2}}{2}\right)\right) \tag{4.15}
\end{equation*}
$$

where $u_{1}=(-2,0)^{\top}$ and $u_{2}=(2,0)^{\top}$. This is solved for the Maxwell molecules $(2 \mathrm{D}$ in velocity) with radius $R=6$. Figure 6 shows the numerical results of PPSM and EFM. The reference solution is calculated by the Fourier spectral method with $N=400$ and $R=8$. It is clear that the EFM solution is much closer to the reference solution. Figure 7 demonstrates that as $N$ increases, EFM converges rapidly to the reference solution.


Fig. 5. Numerical solution of EFM for multiple $N$ values with the BKW solution at time $t=1$ on different scales.


Fig. 6. Comparison between PPSM and EFM at $t=1$ with $N=64$ for the bi-Gaussian initial value.

Example 3 (discontinuous initial value). The initial condition given by

$$
f(t=0, v)= \begin{cases}\frac{\rho_{1}}{2 \pi T 1} \exp \left(-\frac{|v|^{2}}{2 T_{1}}\right) & \text { if } v_{1}>0  \tag{4.16}\\ \frac{\rho_{2}}{2 \pi T 2} \exp \left(-\frac{|v|^{2}}{2 T_{2}}\right) & \text { if } v_{1}<0\end{cases}
$$

in this example is discontinuous. Here $\rho_{1}=\frac{6}{5}$ and the values of $\rho_{2}, T_{1}$, and $T_{2}$ are


FIG. 7. Numerical solution of EFM for multiple $N$ values with the bi-Gaussian initial value at time $t=1$ on different scales.


Fig. 8. Profile of $F(t, v)$ with the discontinuous initial value (4.16) at time $t=0.5$.
uniquely determined by the following conditions:

$$
\int_{\mathbb{R}^{2}} f(0, v) \mathrm{d} v=\int_{\mathbb{R}^{2}} f(0, v)|v|^{2} / 2 \mathrm{~d} v=1, \quad \int_{\mathbb{R}^{2}} f(0, v) v \mathrm{~d} v=0
$$

The profile of the reference solution is presented in Figure 8, which is computed


Fig. 9. Numerical solution of EFM and FGM for multiple $N$ values with the discontinuous initial value (4.16) at time $t=0.5$ on different scales.
by EFM with $N=2048$. Due to the discontinuity in the initial value, the spectral accuracy of FGM is lost. In addition, the Gibbs phenomenon leads to oscillations in the initial value of FGM. In Figure 9, the plots around the discontinuity demonstrate that EFM has much better agreement as compared to FGM. The oscillations in FGM solutions exhibit large errors, and the amplitude of the oscillation decreases slowly as $N$ increases. On the contrary, there is no oscillation for EFM around the discontinuity, and the solution is always nonnegative.

Example 4 (3D BKW solution). The solution of this example is the exact 3D BKW solution, given by

$$
\begin{equation*}
f(t, v)=\frac{1}{(2 \pi S)^{3 / 2}} \exp \left(-\frac{|v|^{2}}{2 S}\right)\left(\frac{5 S-3}{2 S}+\frac{1-S}{2 S^{2}}|v|^{2}\right), \tag{4.17}
\end{equation*}
$$

where $S=1-2 \exp (-t / 6) / 5$. Similarly to the 2D case, we first check the accuracy of EFM. At time $t=0.01$, the $\ell_{1}, \ell_{2}$, and $\ell_{\infty}$ errors and the convergence rates are listed in Table 3. Similarly to the 2D case, the convergence rate is of the second order and the errors are rather small.

Table 3
The $\ell_{1}, \ell_{2}$, and $\ell_{\infty}$ errors and convergence rates for the $B K W$ solution at time $t=0.01$ with $R=6$.

| $N$ | $\ell_{1}$ error | Rate | $\ell_{2}$ error | Rate | $\ell_{\infty}$ error | Rate |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 16 | $4.08 \times 10^{-3}$ |  | $3.08 \times 10^{-3}$ |  | $3.56 \times 10^{-3}$ |  |
| 32 | $1.42 \times 10^{-3}$ | 1.52 | $1.12 \times 10^{-3}$ | 1.47 | $1.26 \times 10^{-3}$ | 1.50 |
| 64 | $4.07 \times 10^{-4}$ | 1.80 | $3.29 \times 10^{-4}$ | 1.76 | $3.72 \times 10^{-4}$ | 1.76 |
| 128 | $1.08 \times 10^{-4}$ | 1.91 | $8.85 \times 10^{-5}$ | 1.90 | $1.00 \times 10^{-4}$ | 1.89 |




Fig. 10. Numerical solution of EFM and PPSM for multiple $N$ values with the BKW solution at time $t=1$ on different scales.

As a comparison with PPSM, Figure 10 presents the numerical solutions on the $v_{1}$ direction of PPSM and EFM at $t=1$ with $N=32$. The plots clearly show that the smoothing filter used in EFM results in much less dissipation, thus leading to better agreement with the exact solution.
5. Discussion. The EFM proposed in this paper is a tradeoff between accuracy and preservation of physical properties. The resulting scheme can be viewed as both a discrete velocity method and a Fourier method. In terms of the convergence rate, it is better than DVM but slower than FGM. In terms of physical properties, it guarantees positivity, mass conservation, and a discrete H -theorem, while the momentum and energy conservation are lost. Regarding the computational cost, fast algorithms in $[19,11]$ remain valid for EFM. As for future work, we plan to study how to mitigate momentum and energy loss, where higher order accuracy is needed for long time simulation. The numerical implementation of the spatially inhomogeneous setting is also in progress.

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## REFERENCES

[1] J.-M. Barbaroux, D. Hundertmark, T. Ried, and S. Vugalter, Gevrey smoothing for weak solutions of the fully nonlinear homogeneous Boltzmann and Kac equations without cutoff for Maxwellian molecules, Arch. Ration. Mech. Anal., 225 (2017), pp. 601-661.
[2] G. A. Bird, Molecular Gas Dynamics and the Direct Simulation of Gas Flows, The Clarendon Press, Oxford University Press, New York, 1994.
[3] A. V. Bobylev, Exact solutions of the Boltzmann equation, Dokl. Akad. Nauk SSSR, 225 (1975), pp. 1296-1299 (in Russian).
[4] A. V. Bobylev, A. Palczewski, and J. Schneider, On approximation of the Boltzmann equation by discrete velocity models, C. R. Acad. Sci. Paris Sér. I Math., 320 (1995), pp. 639-644.
[5] A. V. Bobylev and S. RJasanow, Difference scheme for the Boltzmann equation based on the fast Fourier transform, Eur. J. Mech. B Fluids, 16 (1997), pp. 293-306.
[6] L. Boltzmann, Weitere Studien über das Wärmegleichgewicht unter Gasmolekülen, Wiener Berichte, 66 (1872), pp. 275-370.
[7] T. Carleman, Problemes mathématiques dans la théorie cinétique de gaz, Publ. Sci. Inst. Mittag-Leffler 2, Almqvist \& Wiksell, Uppsala, Sweden, 1957.
[8] G. Dimarco and L. Pareschi, Numerical methods for kinetic equations, Acta Numer., 23 (2014), pp. 369-520.
[9] L. Fainsilber, P. Kurlberg, and B. Wennberg, Lattice points on circles and discrete velocity models for the Boltzmann equation, SIAM J. Math. Anal., 37 (2006), pp. 1903-1922, https: //doi.org/10.1137/040618916.
[10] F. Filbet and C. Mouhot, Analysis of spectral methods for the homogeneous Boltzmann equation, Trans. Amer. Math. Soc., 363 (2011), pp. 1947-1980.
[11] I. M. Gamba, J. R. Haack, C. D. Hauck, and J. Hu, A fast spectral method for the Boltzmann collision operator with general collision kernels, SIAM J. Sci. Comput., 39 (2017), pp. B658-B674, https://doi.org/10.1137/16M1096001.
[12] I. M. Gamba and S. H. Tharkabhushanam, Spectral-Lagrangian methods for collisional models of non-equilibrium statistical states, J. Comput. Phys., 228 (2009), pp. 2012-2036.
[13] D. Goldstein, B. Sturtevant, and J. E. Broadwell, Investigations of the motion of discrete-velocity gases, Progr. Astronautics Aeronautics, 117 (1989), pp. 100-117.
[14] S. Gottlieb, C.-W. Shu, and E. Tadmor, Strong stability-preserving high-order time discretization methods, SIAM Rev., 43 (2001), pp. 89-112, https://doi.org/10.1137/ S003614450036757X.
[15] M. Krook and T. T. Wu, Exact solutions of the Boltzmann equation, Phys. Fluids, 20 (1977), pp. 1589-1595.
[16] G. G. Lorentz, Approximation of Functions, Athena Series, Holt, Rinehart, and Winston, New York, 1966.
[17] G. Meinardus, Approximation of Functions: Theory and Numerical Methods, Springer, New York, 1967.
[18] A. B. Morris, P. L. Varghese, and D. B. Goldstein, Improvement of a discrete velocity Boltzmann equation solver that allows for arbitrary post-collision velocities, in Rarefied Gas Dynamics: Proceedings of the 26th International Symposium (Kyoto, Japan), American Institute of Physics, Melville, NY, 2008, pp. 458-463.
[19] C. Mouhot and L. Pareschi, Fast algorithms for computing the Boltzmann collision operator, Math. Comp., 75 (2006), pp. 1833-1852.
[20] C. Mouhot, L. Pareschi, and T. Rey, Convolutive decomposition and fast summation methods for discrete-velocity approximations of the Boltzmann equation, ESAIM Math. Model. Numer. Anal., 47 (2013), pp. 1515-1531.
[21] A. Palczewski, J. Schneider, and A. V. Bobylev, A consistency result for a discretevelocity model of the Boltzmann equation, SIAM J. Numer. Anal., 34 (1997), pp. 18651883, https://doi.org/10.1137/S0036142995289007.
[22] A. V. Panferov and A. G. Heintz, A new consistent discrete-velocity model for the Boltzmann equation, Math. Methods Appl. Sci., 25 (2002), pp. 571-593.
[23] L. Pareschi and B. Perthame, A Fourier spectral method for homogeneous Boltzmann equations, Transport Theory Statist. Phys., 25 (1996), pp. 369-382.
[24] L. Pareschi and G. Russo, Numerical solution of the Boltzmann equation I: Spectrally accurate approximation of the collision operator, SIAM J. Numer. Anal., 37 (2000), pp. 12171245, https://doi.org/10.1137/S0036142998343300.
[25] L. Pareschi and G. Russo, On the stability of spectral methods for the homogeneous Boltzmann equation, Transport Theory Statist. Phys., 29 (2000), pp. 431-447.
[26] T. PŁatkowski and R. Illner, Discrete velocity models of the Boltzmann equation: A survey on the mathematical aspects of the theory, SIAM Rev., 30 (1988), pp. 213-255, https: //doi.org/10.1137/1030045.
[27] F. Rogier and J. Schneider, A direct method for solving the Boltzmann equation, Transport Theory Statist. Phys., 23 (1994), pp. 313-338.
[28] S. Simons, Is the solution of the Boltzmann equation positive?, Phys. Lett. A, 69 (1978/1979), pp. 239-240.
[29] P. L. Varghese, Arbitrary post-collision velocities in a discrete velocity scheme for the Boltzmann equation, in Proceedings of the 25th International Symposium on Rarefied Gas Dynamics, M. S. Ivanov and A. K. Rebrov, eds., Novosibirsk, Russia, 2007, pp. 225-232.
[30] A. Weiße, G. Wellein, A. Alvermann, and H. Fehske, The kernel polynomial method, Rev. Modern Phys., 78 (2006), 275.


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