NUMERICAL COMPUTATION FOR THE NON-CUTOFF RADIALLY SYMMETRIC HOMOGENEOUS BOLTZMANN EQUATION

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Abstract. For the non cutoff radially symmetric homogeneous Boltzmann equation with Maxwellian molecules, we give the numerical solutions using symbolic manipulations and spectral decomposition of Hermit functions. The initial data can belong to some measure space.

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1. Introduction

1.1. The Boltzmann equation. The Boltzmann equation, derived by Boltzmann in 1872 (and Maxwell 1866), models the behavior of a dilute gas (see [8]). As we know, Boltzmann has created a theory which described the movement of gases as balls which could bump and rebound against each other [11, 19]. This model can be considered by one of many cases which represent the so-called kinetic equation. Presently, the diversity of sciences and applications contains these models such as rarefied gas dynamics, semiconductor modeling, radiative transfer, and biological and social sciences. This type of equations is made by including a combination of a linear transport term and several interaction terms which provide the time evolution of the distribution of particles in the phase space. The equation that bears his name is the following

$$\partial_t f + v \cdot \nabla_x f = Q(f, f)$$

where $f = f(t, x, v) \geq 0$ is the probability density to find a particle at the time $t$, on the position $x$ and with velocity $v$ where the physical and the velocity space are located in three dimensions. The term $v \cdot \nabla_x f$ describes the free action of particles and $Q(f, f)$ is a bilinear operator which describes the binary collision process. It is called the Boltzmann collision operator and given by

$$Q(g, f)(v) = \int_{\mathbb{R}^3} \int_{S^2} B(v - v_s, \sigma)(g(v'_s)f(v') - g(v_s)f(v))dv_se\sigma$$

where for $\sigma \in S^2$, the symbols $v'_s$ and $v'$ are abbreviations for the expressions,

$$v' = \frac{v + v_s}{2} + \frac{|v - v_s|}{2} \sigma, \quad v'_s = \frac{v + v_s}{2} - \frac{|v - v_s|}{2} \sigma,$$

which are obtained in such a way that collision preserves momentum and kinetic energy, namely

$$v'_s + v' = v + v_s, \quad |v'_s|^2 + |v'|^2 = |v|^2 + |v_s|^2$$

where $|\cdot|$ is the Euclidean norm on $\mathbb{R}^3$. Note that $v$, $v'$ are the velocities before collision and $v_s$, $v'_s$ the velocities after collision.

The non-negative cross section $B(z, \sigma)$ depends only on $|z|$ and the scalar product $\frac{z}{|z|} \cdot \sigma = \cos \theta$ where $\theta$ is the deviation angle. Without loss of generality, we may assume that this cross section is supported on the set $\cos \theta \geq 0$. See for instance [30] for more details on the cross section and [38] for a general collision kernel. For physical models, it usually takes the form

$$B(v - v_s, \sigma) = \Phi(|v - v_s|)b(\cos \theta), \quad \cos \theta = \frac{v - v_s}{|v - v_s|} \cdot \sigma, \quad 0 \leq \theta \leq \frac{\pi}{2},$$

where $\Phi(|v - v_s|) = |v - v_s|^\gamma$ is a kinetic factor and $\gamma > -3$. 
In this work, we consider the spatially homogeneous case, that means the density distribution \( f = f(t, v) \) depends on the variables \( t \geq 0, v \in \mathbb{R}^3 \) and is uniform with respect to \( x \). So that the Boltzmann equation reads as

\[
\begin{aligned}
\partial_t f &= Q(f, f), \\
f(0, v) &= F(v)
\end{aligned}
\]  

(1.1)

where the initial data \( F \) is depends only on \( v \). For the collision kernel, we study only the Maxwellian molecules and non-cutoff cases (see [13, 12, 14, 15, 16, 24, 31]), that means the kinetic factor \( \Phi \equiv 1 \) and

\[
b(\cos \theta) \approx \frac{1}{|\theta|^{2+2s}}, \quad 0 < s < 1, \quad \theta \in \left(0, \frac{\pi}{2}\right).
\]  

(1.2)

1.2. Results on the Boltzmann equation. With the previous assumption (the non-cutoff case) on the cross-section, there is existence of a weak solution for the Boltzmann equation (1.1) for a positive initial value \( F \in L^1_{2+\delta}(\mathbb{R}^3) \). See [37] and many others.

Moreover, it is well-known that there is a regularization effect in Sobolev and Schwartz or analytic spaces for any time \( t > 0 \) (we refer the reader to [13, 12] and recently [1]) and that the solutions converge to the Gaussian when the time tends to infinity ([24]).

An important point that our distribution lives in a multidimensional space: this reason make us think that we have a numerical problem because in this case the computational cost is more or less forbidden [19]. The study of the numerical part for kinetic equations is not obvious due to many difficulties come from the computational cost. To clarify more, we mention two of these difficulties: It is clear the appearing of multiple scales, and then to get out of the resolution of the stiff dynamics, one should build suitable numerical schemes [25, 26, 2, 17, 27, 18]. The other one is that the collision operator is defined by multidimensional integrals and to compute one should solve it point by point as physical space [35, 21]. To treat kinetic equations numerically, there is several ways which are used over the centuries until now: probabilistic numerical methods such as Direct Simulation Monte Carlo (DSMC) schemes [11, 3], and, deterministic numerical methods such as finite volume, semi-Lagrangian and spectral schemes [19].

There are two important deterministic methods which are used in the past decades: the discrete velocity method (DVM) [23, 36, 6, 9, 33, 10] and the Fourier spectral method (FSM) [7, 34, 35, 22, 10]. Due to its discrete nature, the DVM preserves positivity of the distribution function, the H-theorem and the exact conservation of mass, energy and momentum. Note that the Fourier spectral method is based on two main things: the truncation of the collision operator and the restriction of the distribution function to an appropriate cube, for more details see [35, 32].
Our goal is to present an alternative method to solve formally and numerically the homogeneous Boltzmann equation in the non-cutoff case. In this work, we consider the radial symmetric case and we use a spectral method: we first compute the spectral coefficients of the solution with a formal computation software (Maple® 13; the codes can be provided). We then approximate these exact solutions and check the numerical results.

The used method helps us to motivate our work in several ways: It let us in the physical view understand more the behavior of the solutions and as we compute the first exact projections of the solutions on the spectral basis, that is in the numerical view, some other algorithms can be tested in the non-cutoff case (recall that the explicit 2D “BKW” solutions, obtained independently in [4, 28] are used to test the accuracy of the numerical methods in the case of a regular collision kernel $B \equiv 1$, see for example [10]). Finally, we do hope that our work will give some clues to formulate new mathematical conjectures.

The paper is organized as follows. In section 2, we state the main theoretical results. The numerical details and algorithms are provided in section 3. Sections 4 and 5 present the numerical results of the Boltzmann equation with different initial data for the Cauchy problem: we discuss in section 4 the results for a small $L^2$ initial data (bi-Gaussian); in section 5, we consider the case of a measure initial data. After that, we give a conclusion for this work. The paper ends with an appendix where we set some technical results.

2. Theoretical results

In this section, we present some theoretical parts: we begin by linearizing the Boltzmann equation and giving the spectral decomposition of this equation.

2.1. Linearization of the Boltzmann equation. We remark that $Q(\mu, \mu) = 0$ where the Gaussian function is defined by

$$\mu(v) = \frac{1}{(2\pi)^{3/2}} e^{-\frac{|v|^2}{2}}$$

and it is a stationary solution of the Boltzmann equation. We consider now a perturbation $g$ of the Gaussian. Then the solution $f$ of (1.1) can be written as

$$f(t, v) = \mu(v) + \sqrt{\mu(v)} g(t, v),$$

$$F(v) = \mu(v) + \sqrt{\mu(v)} G(v).$$

It is easy to show that $g$ is a solution of the Cauchy problem

$$\left\{ \begin{array}{l}
\partial_t g + \mathcal{L}(g) = \Gamma(g, g), \\
g|_{t=0} = g(0, v) = G(v)
\end{array} \right. \quad (2.1)$$
where
\[ L(g) = \frac{-1}{\sqrt{\mu}} \left[ Q(\sqrt{\mu}g, \mu) + Q(\mu, \sqrt{\mu}g) \right] \]
is a linear operator and
\[ \Gamma(g, h) = \frac{1}{\sqrt{\mu}} Q(\sqrt{\mu}g, \sqrt{\mu}h) \]
is a nonlinear operator. We decompose the solution of (2.1) into a linear and nonlinear part:
\[
g(t, v) = e^{-tL} G(v) + e^{-tL} h(t, v)
\]
where \( e^{\alpha L} \) is the exponential of the linear operator defined by its spectral decomposition (see below) and the new function \( h(t, v) \) satisfies the following equation
\[
\begin{aligned}
\partial_t h &= e^{tL} \Gamma(e^{-tL} (G + h), e^{-tL} (G + h)), \\
h(0, v) &= 0.
\end{aligned}
\]
(2.2)
The linearized operator \( L \) is a positive unbounded symmetric operator on \( L^2(\mathbb{R}^3) \) (see [11, 29, 30, 31]) with the kernel
\[ \mathcal{N} = \text{span} \left\{ \sqrt{\mu}, \sqrt{\mu}v_1, \sqrt{\mu}v_2, \sqrt{\mu}v_3, \sqrt{\mu}v^2 \right\} . \]
From a rescaling argument (see Appendix 7.1), we can always assume that the initial condition \( G \) satisfies
\[ G \in \mathcal{N}^\perp. \]
In [30], For the radially symmetric case, the authors show that the linear Boltzmann operator behaves like the fractional harmonic oscillator \( \mathcal{H}^s \) \((0 < s < 1)\) with
\[ \mathcal{H} = -\Delta + \frac{|v|^2}{4}. \]
We study in the next section the spectral properties of the operators \( L \) and \( \Gamma \).

2.2. The spectral problem. We introduce now an orthonormal basis of \( L^2_s(\mathbb{R}^3) \) the radial symmetric functions of \( L^2(\mathbb{R}^3) \) involving the generalized Laguerre polynomials \( L_n^{[\ell+\frac{3}{2}]} \): for that, we set for any \( n \geq 0 \)
\[
\varphi_n(v) = \left( \frac{n!}{\sqrt{2\Gamma(n + 3/2)}} \right)^{1/2} e^{-|v|^2/4} L_n^{[\ell+\frac{3}{2}]} \left( \frac{|v|^2}{2} \right) \frac{1}{\sqrt{4\pi}} \]
(2.3)
where \( \Gamma(\cdot) \) is the standard gamma function, for all \( x > 0 \),
\[ \Gamma(x) = \int_0^{+\infty} t^{x-1} e^{-tx} dx \]
and the Laguerre polynomial $L_n^{(\alpha)}$ of order $\alpha$, degree $n$ is

$$L_n^{(\alpha)}(x) = \sum_{r=0}^{n} (-1)^{n-r} \frac{\Gamma(\alpha + n + 1)}{r!(n-r)!\Gamma(\alpha + n - r + 1)} x^{n-r}.$$ 

We have the spectral decomposition for the linear Boltzmann operator

$$\mathcal{L} \varphi_n = \lambda_n \varphi_n \quad n \geq 0,$$

with $\varphi_0 = \sqrt{\mu}$, $\lambda_0 = 0$ and for $n \geq 1$

$$\lambda_n = 2 \int_{0}^{\frac{\pi}{4}} \beta(\theta) \left( 1 - (\sin \theta)^{2n} - (\cos \theta)^{2n} \right) d\theta \quad (2.4)$$

where $\beta(\theta)$ is defined from the collision kernel (see (1.2))

$$\beta(\theta) = \sin \theta b(\cos \theta) \approx \frac{1}{|\theta|^{1+2s}}, \quad 0 < s < 1, \quad \theta \in \left(0, \frac{\pi}{2}\right]. \quad (2.5)$$

The two families $(\varphi_n(v))_{n \geq 0}$ and $(\lambda_n)_{n \geq 0}$ represent the eigenvectors and the eigenvalues of $\mathcal{L}$. Remark that this diagonalization of the linearized Boltzmann operator with Maxwellian molecules is also verified in the cutoff case (see [5, 11, 20, 29, 30]).

We consider the spectral expansion

$$g(t, v) = \sum_{n=0}^{\infty} g_n(t) \varphi_n(v), \quad G(v) = \sum_{n=0}^{\infty} G_n \varphi_n(v) \quad (2.6)$$

where $g_n(t) = \langle g(t, \cdot), \varphi_n(\cdot) \rangle_{L^2}$ and $G_n = \langle G, \varphi_n \rangle_{L^2}$. By definition, we have

$$e^{-t\mathcal{L}} G(v) = \sum_{n=0}^{\infty} e^{-\lambda_n t} G_n \varphi_n(v).$$

It is the solution of the equation

$$\begin{cases} 
\partial_t g^{fin} + \mathcal{L} g^{fin} = 0, \\
g^{fin}(0, v) = G(v).
\end{cases}$$

Then the operator $\Gamma$ satisfies

$$\Gamma(\varphi_p, \varphi_q) = \mu_{pq} \varphi_{p+q}$$

where the non-linear eigenvalues are given by

$$\mu_{pq} = \left( \frac{(2p+2q+1)}{(2p+1)(2q+1)} \mathcal{C}_{2p+2q}^{2p} \right)^{\frac{1}{2}} \int_{|\theta| \leq \frac{\pi}{4}} \beta(\theta) (\sin \theta)^{2p} (\cos \theta)^{2q} d\theta \quad (2.7)$$

for $p \geq 1, q \geq 0$ and

$$\mu_{0q} = -\int_{|\theta| \leq \frac{\pi}{4}} \beta(\theta) (1 - (\cos \theta)^{2q}) d\theta.$$
for \( q \geq 1 \). Following [31], we therefore derive from (2.1) the following
infinite system of ordinary differential equations:

\[
\begin{cases}
g'_0(t) = 0, & \ g'_1(t) = 0, \\
\text{for all } n \geq 2, \\
g'_n(t) + \lambda_n g_n(t) = \sum_{p+q=n} \mu_{pq} g_p(t) g_q(t)
\end{cases}
\]  

(2.8)

with the initial conditions (see (2.6))

\[ g_n(0) = G_n \quad \text{for } n \geq 0. \]

The goal is to study the behavior of each function \( t \to g_n(t) \).

In the rest, we will focus on the computation and properties of this
intermediate solution.

**Proposition 2.1.** We assume that \( G \in \mathcal{N}^\perp \). Then the intermediate
solution \( h(t,v) \) defined by (2.2) satisfies

\[ h(t,v) = \sum_{n=0}^{\infty} h_n(t) \varphi_n(v) \]  

(2.9)

where \( h_0 \equiv h_1 \equiv h_2 \equiv h_3 \equiv 0 \) and for all \( n \geq 4 \)

\[ h_n(t) = \sum_{p+q=n} \int_0^t \mu_{pq} e^{-(\lambda_p+\lambda_q-\lambda_n)s} \left( G_p + h_p(s) \right) \left( G_q + h_q(s) \right) ds. \]  

(2.10)

**Remark 2.2.** As we have seen before, we divide the function \( g \) in two
parts as follows:

\[ g(t,v) = \sum_{n=0}^{\infty} e^{-\lambda_n t} G_n \varphi_n(v) + \sum_{n=0}^{\infty} e^{-\lambda_n t} h_n(t) \varphi_n(v), \]  

(2.11)

therefore the formal solution \( f(t,v) \) can be written as

\[ f(t,v) = \mu(v) + \sqrt{\mu(v)} \sum_{n=0}^{\infty} \left( e^{-\lambda_n t} G_n + e^{-\lambda_n t} h_n(t) \right) \varphi_n(v). \]  

(2.12)

**Proof of proposition 2.1.** : As \( G \in \mathcal{N}^\perp \), we get \( G_0 = G_1 = 0 \) and we
can verify from (2.8) that

\[ g_0(t) = g_1(t) = 0, \ g_2(t) = G_2 e^{-\lambda_2 t}, \ g_3(t) = G_3 e^{-\lambda_2 t} \]

and therefore \( h_0 \equiv h_1 \equiv h_2 \equiv h_3 \equiv 0 \). By (2.8), we may write

\[ g_n(t) = e^{-\lambda_n t} G_n + e^{-\lambda_n t} h_n(t) \]  

(2.13)

and

\[ g'_n(t) + \lambda_n g_n(t) = \sum_{p+q=n, 2 \leq p, q \leq n-2} \mu_{pq} g_p(t) g_q(t). \]  

(2.14)
We plug again the value of $g_n$ from (2.13) into the equation (2.14) and we get

$$h_n'(t) = e^{\lambda_n t} \sum_{2 \leq p, q \leq n - 2} \mu_{pq} g_p(t) g_q(t).$$

Note that $h_n(0) = g_n(0) = 0$. Finally, plugging the expression of $g_p$ and $g_q$ from (2.13) into the previous equation and integrating we prove (2.10). Concerning the exact expression of the eigenvalue $\lambda_n$ and $\mu_{pq}$, see [31]. This concludes the proof. □

We introduce now the following notations. For a $k$-uplet $\alpha \in \mathbb{N}^k$,

$$\Lambda_\alpha = \lambda_{\alpha_1} + \lambda_{\alpha_2} + \cdots + \lambda_{\alpha_k},$$

$$G^\alpha = G_{\alpha_1} \times G_{\alpha_2} \cdots \times G_{\alpha_k}.$$

**Proposition 2.3.** For each integer $n \geq 4$, we define $I_n$ a set of admissible indices

$$I_n = \left\{ \alpha \in \mathbb{N}^k \mid k \in \mathbb{N}^*, \, \alpha_i \geq 2, \, |\alpha| = n \right\}.$$

Then for each multi-index $\alpha, \beta, \in I_n$ there exists some real coefficients $c_{\beta}^\alpha$ which depends only on $\lambda_2, \ldots, \lambda_n$ and $\mu_{pq}$ for $2 \leq p, q \leq n - 2$, $p + q \leq n$ such that

$$h_n(t) = \sum_{\alpha, \beta \in I_n} c_{\beta}^\alpha G^\alpha \left( 1 - e^{-(\Lambda_\beta - \lambda_n)t} \right).$$

(2.15)

**Proof.** We compute directly from (2.10)

$$h_4(t) = c_{(2,2)}^{(2,2)} G_2^2 \left( 1 - e^{-(\Lambda_{(2,2)} - \lambda_4)t} \right)$$

where

$$c_{(2,2)}^{(2,2)} = \frac{\mu_{22}}{(\Lambda_{(2,2)} - \lambda_4)}$$

and

$$h_5(t) = c_{(2,3)}^{(2,3)} G_2 G_3 \left( 1 - e^{-(\Lambda_{(2,3)} - \lambda_5)t} \right) + c_{(3,2)}^{(3,2)} G_3 G_2 \left( 1 - e^{-(\Lambda_{(3,2)} - \lambda_5)t} \right)$$

where

$$c_{(2,3)}^{(2,3)} = \frac{\mu_{23}}{(\Lambda_{(2,3)} - \lambda_5)} \quad \text{and} \quad c_{(3,2)}^{(3,2)} = \frac{\mu_{32}}{(\Lambda_{(3,2)} - \lambda_5)}.$$

We prove the result by induction. Then we can suppose that (2.15) is true for each $h_{n'}$ $(4 \leq n' \leq n - 1)$. We will use the integral expression (2.10) of $h_n$. We consider two integers $p, q$ such that $2 \leq p, q \leq n - 2$ and $p + q = n$. Then from (2.15)

$$h_p(t) = \sum_{\alpha, \beta \in I_p} c_{\beta}^\alpha G^\alpha \left( 1 - e^{-(\Lambda_\beta - \lambda_p)t} \right),$$

$$h_q(t) = \sum_{\alpha', \beta' \in I_q} c_{\beta'}^{\alpha'} G^{\alpha'} \left( 1 - e^{-(\Lambda_{\beta'} - \lambda_q)t} \right).$$
From the integral formula (2.10) we get

\[ h_n(t) = \int_0^t \sum_{p+q=n}^{2 \leq p, q \leq n-2} (A + B + C + D) \, ds \]

with

\[
A = \mu_{pq} G_p G_q e^{-(\lambda_p + \lambda_q - \lambda_n) s},
\]

\[
B = \sum_{\alpha', \beta' \in I_q} \mu_{pq} \, c_{\beta'}^\alpha G_p G_\alpha' \left( e^{-(\lambda_p + \lambda_q - \lambda_n) s} - e^{-(\lambda_p + \Lambda_{\beta'} - \lambda_n) s} \right),
\]

\[
C = \sum_{\alpha, \beta \in I_p} \mu_{pq} \, c_{\beta}^\alpha G_\alpha G_q \left( e^{-(\lambda_p + \lambda_q - \lambda_n) s} - e^{-(\Lambda_\beta + \lambda_q - \lambda_n) s} \right),
\]

\[
D = \sum_{\alpha, \beta \in I_p} \sum_{\alpha', \beta' \in I_q} \mu_{pq} \, c_{\beta}^\alpha c_{\beta'}^{\alpha'} G_\alpha G_\alpha' \times \left( e^{-(\lambda_p + \lambda_q - \lambda_n) s} - e^{-(\Lambda_\beta + \lambda_q - \lambda_n) s} - e^{-(\Lambda_\beta + \Lambda_{\beta'} - \lambda_n) s} + e^{-(\Lambda_\beta + \Lambda_{\beta'} - \lambda_n) s} \right).
\]

Expanding each previous terms and integrating over \([0, t]\), we get the result (2.15) since each number \(\lambda_p + \lambda_q - \lambda_n, \Lambda_\beta + \lambda_q - \lambda_n, \Lambda_\beta + \Lambda_{\beta'} - \lambda_n\) are positive from the next lemma and \(|\alpha| = |\beta| = p, |\alpha'| = |\beta'| = q\) and \(p + q = n\).

**Lemma 2.4.** The linear eigenvalues \(\lambda_n\) for the non-cutoff radially symmetric spatially homogeneous Boltzmann equation

\[ \lambda_n = \int_{|\theta| \leq \frac{\pi}{4}} \beta(\theta) \left( 1 - (\sin \theta)^{2n} - (\cos \theta)^{2n} \right) d\theta, \quad n \geq 2, \]

verify the following property

\[ \lambda_{\alpha_1 + \cdots + \alpha_k} < \lambda_{\alpha_1} + \cdots + \lambda_{\alpha_k} (= \Lambda_\alpha) \]

for multi-index \(\alpha \in (\mathbb{N} \setminus \{0, 1\})^k\).

**Proof.** By [31], we may write

\[ \lambda_{\alpha_1 + \alpha_2} < \lambda_{\alpha_1} + \lambda_{\alpha_2}, \]

then by iteration, we have

\[ \lambda_{(\alpha_1 + \cdots + \alpha_k) + \alpha_{k+1}} < \lambda_{\alpha_1 + \cdots + \alpha_k + \alpha_{k+1}} < (\lambda_{\alpha_1} + \cdots + \lambda_{\alpha_k}) + \lambda_{\alpha_{k+1}}. \]

**3. Numerical computations**

From now on, for sake of simplicity, we consider the specific case \(s = \frac{1}{2}\) and

\[ \beta(\theta) = (\sin \theta)^{-2}. \]

For the general case \(s \in [0, 1]\) and other kernel \(\beta\) which satisfies (2.5), we can compute some numerical approximations of the eigenvalues. We think that the results do not change.
3.1. Computation of the eigenvalues. By the following assumption
\[ \beta(\theta) \approx 0 \quad \text{for} \quad |\theta| \geq 2, \]
we obtain (see [30])
\[ \lambda_n \approx \sqrt{n} \]  \tag{3.1}
where the linear eigenvalues \( \lambda_n \) of \( L \) was defined in (2.4). We recall the
value of \( \lambda_n \) for \( n \geq 2 \):
\[ \lambda_n = 2 \int_0^{\pi/2} \beta(\theta) \left( 1 - (\sin \theta)^{2n} - (\cos \theta)^{2n} \right) d\theta. \]

We compute the exact and approximate values of \( \lambda_n \) by the following
algorithm:
\[
\begin{align*}
\lambda_0 &\leftarrow 0 \\
\text{for } n \text{ from 1 to } N \text{ do} \\
\text{expr} &\leftarrow \text{algebraic simplification of } 1 - \sin^{2n} \theta - \cos^{2n} \theta \\
\lambda_n^{\text{exact}} &\leftarrow \text{symbolic computation of } 2 \int_0^{\pi/2} \text{expr} d\theta \\
\lambda_n^{\text{approx}} &\leftarrow \text{numerical computation of } \lambda_n^{\text{exact}}
\end{align*}
\]
The “algebraic simplification” of “expr” removes the singularity when
\( \theta \to 0 \) coming from the collision kernel \( \beta(\theta) = \sin^{-2} \theta \) (see (2.5)). It consists in a factorization of trigonometric polynomials. The symbolic computation of \( \lambda_n^{\text{exact}} \) is reduced to compute the exact integral
of a trigonometric polynomial. Then \( \lambda_n^{\text{exact}} \) is approached numerically with a number of significant digits (equal to 10 in 1). The approximation is easily controlled by the estimate of the relative error \( |\lambda_n^{\text{exact}} - \lambda_n^{\text{approx}}| / \lambda_n^{\text{exact}} \). Using the software Maple\textsuperscript{R} 13, we finally get the numerical table 1.

| \( \lambda_n \) | Exact value | Approximate value | Relative error |
|----------------|-------------|------------------|----------------|
| \( \lambda_1 \) | 0           | 0                | –              |
| \( \lambda_2 \) | 1 + \frac{1}{2} \pi | 2.570796327      | 8.0 \times 10^{-11} |
| \( \lambda_3 \) | \frac{3}{2} + \frac{3}{4} \pi | 3.85619490      | 5.0 \times 10^{-11} |
| \( \lambda_4 \) | \frac{23}{12} + \frac{1}{16} \pi | 4.861909780    | 1.2 \times 10^{-10} |
| \( \lambda_5 \) | \frac{55}{32} + \frac{35}{32} \pi | 5.727783632    | 8.2 \times 10^{-11} |
| \( \lambda_{10} \) | \frac{61717}{16128} + \frac{109305}{65536} \pi | 9.070756042     | 9.0 \times 10^{-11} |
| \( \lambda_{15} \) | \frac{41349267}{8200192} + \frac{35102025}{16777216} \pi | 11.61545300 | 3.2 \times 10^{-10} |
| \( \lambda_{20} \) | \frac{60225247403}{996683904} + \frac{83945001525}{34359738368} \pi | 13.75454524 | 2.5 \times 10^{-11} |

Table 1. Symbolic and numerical computation of \( \lambda_n \).
The approximation of eigenvalues can be controlled to be sufficiently precise for upcoming computations. For a general kernel $\beta(\theta)$, there is in general no more explicit values. But some classical numerical methods can be easily applied. Nevertheless, there is no more any algebraic simplification, and it is necessary to treat carefully the singularity.

3.2. Computation of the nonlinear eigenvalues. We recall the coefficients $\mu_{pq}$ from (2.7): for some $p, q \geq 1$

$$
\mu_{pq} = \sqrt{\frac{(2p + 2q + 1)}{(2p + 1)(2q + 1)}} \frac{C^{2p}}{2^{p+2q}} \int_{|\theta| \leq \frac{\pi}{4}} \beta(\theta)(\sin \theta)^{2p}(\cos \theta)^{2q} d\theta.
$$

We compute the exact and the approximate value of $\mu_{pq}$ (again with a relative error $\approx 10^{-10}$) for $1 \leq p + q \leq N$ by the following algorithm:

for $p$ from 1 to $N$ do
   for $q$ from 0 to $N-p$ do
      $expr \leftarrow$ symbolic computation of $2 \int_{0}^{\frac{\pi}{4}} \sin^{2p-2} \theta \cos^{2q} \theta d\theta$
      $\mu_{pq}^{\text{exact}} \leftarrow \sqrt{\frac{(2p+2q+1)}{(2p+1)(2q+1)}} \frac{C^{2p}}{2^{p+2q}} \times expr$
      $\mu_{pq}^{\text{approx}} \leftarrow$ numerical computation of $\mu_{pq}^{\text{exact}}$

We present in the table the results for $p+q = n = 2, \ldots, 5, 20$. The singularity coming from the collision kernel $\beta(\theta) = \sin^{-2} \theta$ is removed by a simple simplification (remark the exponent $(2p-2)$ of the sinus term of $\mu_{pq}$). Again for a general collision kernel, the values of these nonlinear eigenvalues can be approximated by classical numerical methods.

| $n = 2$ | $n = 3$ | $n = 4$ | $n = 5$ | $n = 20$ |
|---------|---------|---------|---------|---------|
| $\mu_{1,1} \approx 2.35$ | $\mu_{1,2} \approx 2.88$ | $\mu_{1,3} \approx 3.29$ | $\mu_{1,4} \approx 3.62$ | $\mu_{1,19} \approx 6.68$ |
| $\mu_{2,1} \approx 0.519$ | $\mu_{2,2} \approx 0.702$ | $\mu_{2,3} \approx 0.84$ | $\mu_{2,18} \approx 1.55$ |
| $\mu_{3,1} \approx 0.196$ | $\mu_{3,2} \approx 0.30$ | $\mu_{3,17} \approx 0.75$ |
| $\mu_{4,1} \approx 0.084$ | $\mu_{4,16} \approx 0.46$ |
| $\mu_{19,1} \approx 10^{-5}$ |

Table 2. Numerical computation of $\mu_{pq}$.
3.3. Numerical solutions of the linear problem. We introduce from (2.11) the approximation of the linear solution

\[ g^{\text{fin}}_N(t, v) = \sum_{n=0}^{N} e^{-\lambda_n t} G_n \varphi_n(v) \]  

(3.2)

where the reals \( G_n \) are the given initial spectral coefficients. In order to compute the value of the linear solution, we use the formula (2.3) of the eigenfunction \( \varphi_n \) which involves the generalized Laguerre polynomials \( L_n^{(\ell+\frac{1}{2})} \). We get the following algorithm :

for \( n \) from 0 to \( N \) do

\[ \varphi_n(v) \leftarrow \left( \frac{n!}{\sqrt{2\Gamma(n+3/2)}} \right)^{1/2} e^{-\frac{|v|^2}{4}} L_n^{\frac{1}{2}} \left( \frac{|v|^2}{2} \right) \frac{1}{\sqrt{4\pi}} \]

Finally, we obtain the linear solution by the sum :

\[ g^{\text{fin}}_N(t, v) \leftarrow 0 \]

for \( n \) from 2 to \( N \) do

\[ g^{\text{fin}}_N(t, v) \leftarrow g^{\text{fin}}_N(t, v) + e^{-\lambda_n t} G_n \varphi_n(v) \]

We estimate the \( L^2 \) theoretical error \( (g^{\text{fin}} - g^{\text{fin}}_N) \) for the different initial data \( G \) used for computation in the next sections.

**Proposition 3.1.** We consider the solution of the following linear problem

\[
\begin{cases}
\partial_t g^{\text{fin}} + \mathcal{L} g^{\text{fin}} = 0, \\
g^{\text{fin}}(0, v) = G(v).
\end{cases}

(3.3)

We have the following estimates :

1) For initial data \( G \in L^2 \),

\[ \|g^{\text{fin}}(t, \cdot) - g^{\text{fin}}_N(t, \cdot)\|_{L^2} \leq e^{-c\sqrt{N} t} \|G\|_{L^2}. \]

2) For the measure initial data \( G \) defined by (5.1) (see also proposition 7.3), there exist some constants \( C > 0 \) and \( c > 0 \) such that for \( t > 0 \)

\[ \|g^{\text{fin}}(t, \cdot) - g^{\text{fin}}_N(t, \cdot)\|_{L^2} \leq \frac{1}{t^b} e^{-\gamma \sqrt{N} t}. \]

**Proof.** The solution of (3.3) is

\[ g^{\text{fin}}(t, v) = \sum_{n=0}^{\infty} e^{-\lambda_n t} G_n \varphi_n(v). \]

The exact error in \( L^2 \) is

\[ \|g^{\text{fin}}(t, \cdot) - g^{\text{fin}}_N(t, \cdot)\|_{L^2}^2 = \sum_{n=N+1}^{\infty} e^{-2\lambda_n t} |G_n|^2. \]

1) If \( G \in L^2(\mathbb{R}^3) \), then as we have from (3.1)

\[ \|g^{\text{fin}}(t, \cdot) - g^{\text{fin}}_N(t, \cdot)\|_{L^2}^2 = \sum_{n=N+1}^{\infty} e^{-2\lambda_n t} |G_n|^2 \leq e^{-2c\sqrt{N} t} \|G\|_{L^2}^2. \]
We can deduce that the exact error tends to zero when $N$ tends to infinity.

2) We suppose now that $F$ is the measure initial data $\mu + \delta$. We can approximate the spectral coefficients $G_n$ of $G$ by $n^4$ and by 3.1 we can then find some positive constants $c$ and $C$ such that

$$\|g^{\text{fin}}(t, \cdot) - g_N^{\text{fin}}(t, \cdot)\|_{L^2} \leq C \sum_{n=N+1}^{\infty} e^{-c\sqrt{n}t} n^2.$$ 

We consider the function $\rho_t$ defined on $\mathbb{R}_+$ by $\rho_t(x) = e^{-c\sqrt{x}t} x^2$. So that $\rho_t$ is positive, continuous and decreasing for $x \geq 16/(ct)^2$, therefore by using the Cauchy integral criterion, we can write the following inequality:

$$\|g^{\text{fin}}(t, \cdot) - g_N^{\text{fin}}(t, \cdot)\|_{L^2} \leq \frac{C}{t^b} e^{-\gamma\sqrt{n}t} \xrightarrow{N \to \infty} 0$$

where $b$ and $\gamma$ are some positive constants.

3.4. **Numerical solutions of the non-linear part.** Concerning the nonlinear part $g^{\text{nt}} = e^{-it\cdot h}$ of the solution, we consider the partial series

$$g_N^{\text{nt}}(t, v) = \sum_{n=0}^{N} e^{-\lambda_n t} h_n(t) \varphi_n(v). \quad (3.4)$$

We then use the decomposition of $h$ in the spectral basis (2.9) and the integral formula (2.10) to compute $h_n(t)$. Therefore we solve explicitly the system (2.8) by the following algorithm:

$$S \leftarrow 0$$

for $n$ from 4 to $N$ do

$$S \leftarrow S + \mu_{pq} (G_p + h_p(t)) (G_q + h_q(t)) e^{-(\lambda_p + \lambda_q - \lambda_n)t}$$

$h_n(t) \leftarrow$ symbolic computation of $\int_0^t S$

The exact computation of the integral $\int_0^t S$ is straightforward since, from proposition 2.3, the symbolic expression $S$ is an linear combination of exponential terms $e^{at}$. We get the exact following solutions of the system of integral formula (2.10):

$h_0 = h_1 = h_2 = h_3 = 0$,

$$h_4 = \frac{\mu_{22}}{\lambda_2 + \lambda_2 - \lambda_4} G_2^2 \left(1 - e^{-(\lambda_2 + \lambda_2) t}\right),$$

$$h_5 = \frac{\mu_{23} + \mu_{32}}{\lambda_2 + \lambda_3 - \lambda_5} G_2 G_3 \left(1 - e^{-(\lambda_2 + \lambda_3) t}\right), \ldots$$
From the symbolic expression of $h_n$ we compute the numerical approximation:

\[
\begin{align*}
    h_0 &= h_1 = h_2 = h_3 = 0,
    
    h_4 &= 2.51 G_2^2 \left( 1 - e^{-0.279 t} \right),
    
    h_5 &= 1.62 G_2 G_3 \left( 1 - e^{-0.698 t} \right),
    
    h_6 &= 0.322 G_3^2 \left( 1 - e^{-1.20 t} \right) + 1.17 \left( 1 - e^{-0.928 t} \right) G_2 G_4
         + \left( -2.95 e^{-0.928 t} + 0.677 + 2.26 e^{-1.20 t} \right) G_2^3,
    
    h_7 &= 0.501 G_2 G_5 \left( 1 - e^{-1.09 t} \right) + 0.220 \left( 1 - e^{-1.51 t} \right) G_3 G_4
         + \left( 0.201 + 0.478 e^{-1.79 t} - 0.274 e^{-1.51 t} - 0.407 e^{-1.09 t} \right) G_2^2 G_3,
    
    \ldots
\end{align*}
\]

We finally get from (3.4) the approximation $g_n^{nl}$ of the nonlinear part of the solution $g^{nl}(t, v)$ by the following algorithm:

\[
\begin{align*}
    g_n^{nl}(t, v) &\leftarrow 0 \\
    \text{for } n \text{ from 2 to } N \text{ do} \\
    g_n^{nl}(t, v) &\leftarrow g_n^{nl}(t, v) + e^{-\lambda_n t} h_n(t) \varphi_n(v)
\end{align*}
\]

The symbolic and numerical computation of the nonlinear part of the solution plays the main difficulty of our method. We analyze the computation time and rounding error in the next section.

3.5. **Discussions on the symbolic computation.** From the computation of the linear (3.2) and nonlinear (3.4) part, we calculate the approximated solution of the Boltzmann equation (1.1)

\[
f_N = \mu + \sqrt{\mu} (g_{N}^{lin} + g_{N}^{nl}).
\]  

The method using the software Maple®13 and its internal function "\texttt{int}(f(x), x = a..b)" for symbolic computation of integrals seems limited to a number $N$ around 20, since for $N = 20$, the number of terms of $h_{20}$ is around 5000 and the computation time is around 50 seconds. Moreover, they are both exponentially increasing (see Figure 1). We now estimate the truncation and rounding error due to the software computations. For a regular $L^2$ initial data we have computed the solution $f_N$ for $N = 20$ and different number of digits (we can control the number of digits that Maple®13 uses when making calculations with software floating-point numbers). We set $P_1$ and $P_2$ two numbers of digits and we compare the two numerical solutions $f_N^{P_1}$ and $f_N^{P_2}$ computed respectively using $P_1$ and $P_2$. We define the rounding relative error as

\[
\text{error} = \frac{\|f_N^{P_1} - f_N^{P_2}\|_{\infty}}{\|f_N^{P_2}\|_{\infty}}
\]

and we get the following results for different choices of ($P_1, P_2$) : We check from the table 3 that the relative error is roughly 10 times the
Figure 1. Number of terms for $h_n$ and computation time in seconds for $h_1, \ldots, h_n$.

| $(P_1, P_2)$ | error |
|--------------|-------|
| (10,20)      | 3.8 $10^{-4}$ |
| (20,30)      | 2.3 $10^{-15}$ |
| (30,40)      | 2.0 $10^{-20}$ |
| (40,50)      | 8.2 $10^{-24}$ |
| (50,100)     | 1.4 $10^{-40}$ |

Table 3. Relative rounding off error.

The precision of the computation of $f_{P_i}^N$. The figure 2 represents the computation time of the solution $f_N$ with a regular $L^2$ initial data and for different numbers of digits. The computations of the solution $f_N$ was run on a computer having 8 Xeon processors 2.33 GHz with 8 GB of memory. The method using Maple$^\text{⃝}$ 13 on this computer seems limited to a number around $N = 20$. Surprisingly, the computation time is roughly the same (around 90 seconds) for a number of digits between 20 and 1000. The main part of this time is therefore used for algebraic manipulation.

We present in the two upcoming sections the results of the computation for different initial values.
4. Radial bi-Gaussian initial value

We set the initial data:

$$\tilde{F}(w) = \frac{1}{(2\pi)^{\frac{3}{2}}} \left( \exp \left( -\frac{1}{2} |w| + 1 \right)^2 + \exp \left( -\frac{1}{2} |w| - 1 \right)^2 \right).$$

We next rescale the initial data following lemma 7.1. We show in figure 3 the spectral approximation $F_N(v)$ of the initial data $F(v)$ such that $F_N(v) = \mu(v) + \sqrt{\mu(v)} G_N(v)$ where $G_N(v) = \sum_{n=0}^{N} G_n \varphi_n(v)$. We then compute the solutions $h_n(t)$ from the proposition 2.1 for $n = 4, 5, \ldots, N$ with $N = 20$. For each integer $n$, the function $t \rightarrow h_n(t)$ is monotone and tends to a finite limit when $t$ tends to infinity (See Figure 4 (a)). We recall that $h_n(t)$ is a finite sum of decreasing exponential terms (see section 3.4). Since the initial data $G$ is a regular function, the spectral coefficients $G_n$ are exponentially decreasing. The numerical computation of $h_n(t)$ shows also that $\|h_n\|_{\infty}$ is exponentially decreasing with respect to $n$ (See Figure 4 (b)). In this special case, the linear part $e^{-t\mathcal{L}} G$ and the nonlinear part $e^{-t\mathcal{L}} h$ have roughly the same behavior. We present in figure 5 the graph of the linear part and nonlinear part and the ratio in $L^2$-norm

$$R_N(t) = \frac{\|g_N^{in}(t, \cdot)\|_{L^2}}{\|g_N^{out}(t, \cdot)\|_{L^2}} = \frac{\left( \sum_{n=4}^{N} |e^{-\lambda_n t} h_n(t)|^2 \right)^{\frac{1}{2}}}{\left( \sum_{n=2}^{N} |e^{-\lambda_n t} G_n|^2 \right)^{\frac{1}{2}}}. \quad (4.1)$$

We observe that the nonlinear part is very small compared to the other. We remark that in this case the series (3.2) and (3.4) behave as

\begin{figure}
\centering
\includegraphics[width=0.7\textwidth]{approximation_initial_data.png}
\caption{Approximation of the initial data}
\end{figure}
Figure 4. Behavior of the nonlinear part $h_n$.

Figure 5. Comparison of the nonlinear part with respect to the linear part (see (3.2), (3.4), (4.1), (4.2)).

\[ g_{\text{fin}}(t,v) \approx e^{-\lambda_2 t} G_2 \varphi_2(v), \]
\[ g_{\text{fin}}^n(t,v) \approx 2.51 e^{-\lambda_4 t} G_2^2 (1 - e^{-(2\lambda_2-\lambda_4)t}) \varphi_4(v), \]

because the terms of $h_n(t)$ are composed of products of terms which are numerically converging to zero. The quotient (for $G_2 \neq 0$) of the two previous approximations behaves closely like the ratio (see (4.1))

\[ R_N(t) \approx \tilde{R}(t) \overset{\text{def}}{=} e^{-(\lambda_4-\lambda_2)t} |G_2| (1 - e^{-(2\lambda_2-\lambda_4)t}). \] (4.2)

We finally compute (see figure 6) the solution $f = \mu + \sqrt{\mu} g$ using the spectral Hermite eigenfunctions $\varphi_n(v)$ and the expansion (2.12) of $g$ in this basis. Since the function $g(t,\cdot) \in N^\perp$ for all time $t \geq 0$, the approximate solution $f_N(t,\cdot)$ is naturally orthogonal to $\varphi_0$ and $\varphi_1$. Therefore is a conservation of the mass and the energy. Finally, we check that the approximate solution $f_N$ converges to the Gaussian function when the time tends to infinity.
5. Numerical results for initial measure data

We consider the initial measure data

\[ \tilde{F} = \text{gaussian} + \text{Dirac} = \mu + \delta. \]

Following the lemma and rescaling the solution \( F(v) = 2^{-\frac{5}{2}} \tilde{F}(2^{-\frac{1}{2}} v) \), we get the normalized initial data

\[
\begin{align*}
    F(v) &= 2^{-\frac{5}{2}} \mu(2^{-\frac{1}{2}} v) + 2^{-1} \delta(v), \\
    G(v) &= 2^{-\frac{13}{4}} \pi^{-\frac{3}{4}} - \sqrt{\mu(v)} + 2^{-\frac{1}{4}} \pi^{\frac{3}{4}} \delta(v). 
\end{align*}
\]

(5.1)

We verify that \( \langle G, \varphi_0 \rangle = \langle G, \varphi_1 \rangle = 0 \) and therefore \( G \in \mathcal{N}^\perp \). We then compute the spectral coefficients for \( n \geq 0 \) (see proposition 7.1):

\[ G_n = \langle G, \varphi_n \rangle = \frac{1 + (-1)^n}{2} \left( \frac{(2n + 1)!}{2^{2n}(n!)^2} \right)^{\frac{1}{2}}. \]

Note that the coefficients \( G_{2n+1} \) are equal to zero and we have the following approximation of \( G \):

\[ G(v) \approx \sum_{n=1}^{\infty} n^{\frac{1}{2}} \varphi_{2n}(v). \]

We set \( F_{\text{reg}}(v) = 2^{-\frac{5}{2}} \mu(2^{-\frac{1}{2}} v) \) the regular part of the distribution \( F \). We check in the left figure 7 that the approximate initial data behaves as a Dirac function.
Remark that to capture the approximation of the regular part $F_{\text{reg}}$, we have to rescale the cote $y$-coordinate. We observe the oscillations of $F_N$ which are expected since the functions $F_N$ approach the Dirac function when $N$ tends to infinity (see the right figure 7). We now focus on the evolution problem. As the initial data is a distribution, we can check that the linear part of the solution is singular:

$$\|g^{\text{lin}}(t, \cdot)\|_{L^2}^2 = \sum_{n=2}^{\infty} G_n^2 e^{-2 \lambda_n t} \approx \frac{1}{t^\alpha}, \quad \text{when } t \to 0 \quad (5.2)$$

for some $\alpha > 0$ (since $G_{2n} \approx n^{\frac{3}{4}}$ and $\lambda_{n} \approx n^{\frac{1}{2}}$). We next compute the nonlinear part $h_n(t)$ of the solution (see the left figure 8).

$$t \mapsto h_n(t) \text{ for } n = 4, 5, \ldots, 20. \quad \sup_{t \geq 0} \frac{|h_n(t)|}{|G_n|} \text{ for } n = 4, 6, \ldots, 20.$$

Figure 7. Approximation of the initial data.

Figure 8. Behavior of the nonlinear part $h_n$.

We observe some numerical evidences that these functions are increasing less than a power of $n$:

$$\sup_{t \geq 0} |h_n(t)| \leq C \ n^\alpha$$
with $a$ close to 1. Since $G_n \approx n^{3/4}$, the behavior of a term of the series $(g_{N}^{\text{lin}}(t) + g_{N}^{\text{nl}}(t))$ is dominated by the nonlinear part.

We next calculate the linear part and nonlinear part of solution (see Figure 9). The numerical computations in the left figure show that $\|g_{20}^{\text{lin}}\|_{L^2}$ is dominated by the nonlinear part.

We then compute the numerical approximation $f_N$ of the solution $f$ for $N = 20$ and we check that the solution behaves as a Dirac function as $t \to 0$ and tends to the Gaussian as $t \to \infty$ (see Figure 10).

We observe some other numerical evidences that the series converges in $L^2$ for $t > 0$ and the solution converges to a Gaussian as $t \to \infty$.

6. CONCLUSION

We have considered the perturbation $g$ of the solution $f$ of the Boltzmann equation defined by

$$f = \mu + \sqrt{\mu} g \quad \text{where} \quad g(t, v) = \sum_n g_n(t) \varphi_n(v)$$

and we have studied the behavior of the spectral coefficients

$$g_n(t) = e^{-\lambda_n t}(G_n + h_n(t)), \quad g_n(0) = G_n.$$
We have then computed formally the spectral coefficients $h_n(t)$ for $n = 0, 1, \ldots, N$ with $N = 20$. We have checked also the results for small $L^2$ initial data and distribution type initial data $\mu + \delta$.

- For small $L^2$ initial data, our method was tested with several $L^2$ initial conditions: $F$ is a sum of two Gaussian, $G_n = \frac{\delta_n}{\pi}$, $G_n = \frac{1}{\pi}$. The results show that there are some numerical evidences that the spectral series $\sum_n e^{-\lambda_n t}(G_n + h_n(t))\varphi_n$ is convergent in $L^2$ for any time $t \geq 0$ and the solution converges to a Gaussian. Moreover, for large times, the linear part $G_n$ is preponderant with respect to the non-linear part $h_n(t)$.

- For the distribution type initial data $\mu + \delta$, the simulations show some numerical evidences that the spectral series converges in $L^2$ and there is a regularization of the solution for $t > 0$.

We have computed the formal solutions of the spectral coefficients $h_n$ of the solution of the Boltzmann equation. If there exists a regular solution for $t > 0$, then the solutions $h_n$ are the exact projections of the solution on the spectral basis. These calculations were made in the case of a non-cutoff kernel. The numerical results are coherent for small $L^2$ initial data or for the distribution case $\mu + \delta$. There is conservation of the mass, momentum and energy of the approximated solution (since $g_N(t, \cdot)$ is orthogonal to the kernel $\mathcal{N}$ for all time). Moreover the approximated solution $f_N(t, \cdot)$ (defined in (3.5)) converges to a Gaussian when $t$ tends to infinity.
Acknowledgments
The authors wish to thank Chao-Jiang Xu for interesting discussions.

7. Appendix

7.1. Rescaling of the solution. We consider a radial solution \( \tilde{f}(s, w) \) of the Boltzmann equation:

\[
\begin{align*}
\partial_s \tilde{f} &= Q(\tilde{f}, \tilde{f}), \\
\tilde{f}|_{t=0} &= \tilde{F}.
\end{align*}
\]

Lemma 7.1. We consider the functions \( f(t, v) \) and \( F(v) \) defined by the change of variable

\[
\begin{align*}
f(t, v) &= \alpha \tilde{f}(\frac{v}{\alpha} t, \beta v), \\
F(v) &= \alpha \tilde{F}(\beta v)
\end{align*}
\]

where

\[
\alpha = \left( \frac{1}{3} \int_{\mathbb{R}^3} w^2 \tilde{F}(w) \, dw \right)^{\frac{2}{3}} \quad \text{and} \quad \beta = \left( \frac{1}{3} \int_{\mathbb{R}^3} w^2 \tilde{F}(w) \, dw \right)^{\frac{1}{3}}.
\]

Therefore \( f(t, v) \) is a solution of the Boltzmann equation (1.1) with initial data \( F \). Moreover, if we set \( F = \mu + \sqrt{\mu} G \), we then have \( G \in \mathcal{N}^\perp \).

Remark 7.2. If \( F \) is such that

\[
\begin{align*}
\int_{\mathbb{R}^3} F(v) \, dv &= \int_{\mathbb{R}^3} \mu(v) \, dv = 1, \\
\int_{\mathbb{R}^3} v^2 F(v) \, dv &= \int_{\mathbb{R}^3} v^2 \mu(v) \, dv = 3,
\end{align*}
\]

then the function \( G \) defined by \( G = \frac{1}{\sqrt{\mu}} (F - \mu) \) belongs to \( \mathcal{N}^\perp \).

Proof. It is easy to check that \( f(t, v) \) is a solution of the Boltzmann equation. Since \( G \) is a radial function, it is enough to check that

\[
\left( G, \sqrt{\mu} \right)_{L^2} = \left( G, |v|^2 \sqrt{\mu} \right)_{L^2} = 0.
\]

Recalling that \( \left( \varphi_p, \varphi_q \right)_{L^2} = \delta_{pq} \),

\[
\varphi_0 = \sqrt{\mu} \quad \text{and} \quad \varphi_1 = 6^{-\frac{1}{2}} (3 - |v|^2) \sqrt{\mu},
\]

it is equivalent to prove

\[
(F/\sqrt{\mu}, \varphi_0)_{L^2} = 1 \quad \text{and} \quad (F/\sqrt{\mu}, \varphi_1)_{L^2} = 0,
\]

which gives the equations

\[
\begin{align*}
\int_{\mathbb{R}^3} F(v) \, dv &= \int_{\mathbb{R}^3} \mu(v) \, dv = 1, \\
\int_{\mathbb{R}^3} |v|^2 F(v) \, dv &= \int_{\mathbb{R}^3} |v|^2 \mu(v) \, dv = 3.
\end{align*}
\]
Using the change of variable \( w = \beta v \), we can check that if we set the values of \( \alpha \) and \( \beta \) given in the lemma, the previous equations are fulfilled.

\[ \square \]

7.2. **Measure initial data.** We define the following distribution initial data:

\[ \tilde{F} = \mu + \delta. \]

Following the rescaling of lemma 7.1, we compute

\[
\langle \tilde{F}, 1 \rangle = \int_{\mathbb{R}^3} \mu(v) \, 1 \, dv + \langle \delta, 1 \rangle = 2, \\
\langle \tilde{F}, v^2 \rangle = \frac{1}{3} \int_{\mathbb{R}^3} \mu(v) \, v^2 \, dv + \langle \delta, v^2 \rangle = 1
\]

and then \( \alpha = 2^{-\frac{5}{2}} \) and \( \beta = 2^{-\frac{1}{2}} \). Using the change of variable \( w = \beta v \), we get the new rescaled distribution initial data

\[
F = \alpha \tilde{F} \circ (\beta \text{Id}) = 2^{-\frac{5}{2}} \left( \mu(2^{-\frac{1}{2}} \cdot) + (2^{\frac{1}{2}})^3 \delta \right).
\]

**Proposition 7.3.** We consider the initial data

\[
F = 2^{-\frac{5}{2}} \left( \mu(2^{-\frac{1}{2}} \cdot) + (2^{\frac{1}{2}})^3 \delta \right)
\]

and we set \( G \) such that \( F = \mu + \sqrt{\mu} G \). Then we have

\[
G = -\sqrt{\mu} + 2^{-\frac{33}{4}} \pi^{-\frac{3}{4}} + 2^{-\frac{1}{4}} \pi^{\frac{3}{4}} \delta.
\]

We consider the coordinates \( G_n = \langle G, \varphi_n \rangle \) of the distribution \( G \) in the spectral basis \( (\varphi_n)_n \). We can check that

\[
G_0 = G_1 = 0
\]

and for all integer \( n \geq 2 \),

\[
G_n = \langle G, \varphi_n \rangle = \frac{1 + (-1)^n}{2} \left( \frac{(2n + 1)!}{2^{2n(n!)^2}} \right)^{\frac{1}{2}}. \tag{7.1}
\]

**Proof.** The expression of \( G \) follows from the definition of the Gaussian \( \mu \). We then compute

\[
\langle G, \varphi_n \rangle = -\langle \varphi_0, \varphi_n \rangle_{L^2} + 2^{-\frac{33}{4}} \pi^{-\frac{3}{4}} \langle 1, \varphi_n \rangle_{L^2} + 2^{-\frac{1}{4}} \pi^{\frac{3}{4}} \varphi_n(0)
\]

and the conclusion results directly from lemma 7.5. \[ \square \]

We consider now a special Gaussian approximation \( F_\varepsilon \in L^2 \) of the distribution initial data \( F = \mu + \delta \) and we obtain some spectral stability result in this case.

**Proposition 7.4.** We consider the initial data for \( \varepsilon > 0 \)

\[
\tilde{F}_\varepsilon(w) = \mu(w) + \frac{1}{\varepsilon^3} \mu \left( \frac{w}{\varepsilon} \right).
\]
Following lemma 7.1, the rescaled initial data of $\tilde{F}_\varepsilon$ is $F_\varepsilon = \mu + \sqrt{\mu} G_\varepsilon$ where $G_\varepsilon \in \mathcal{N}^\perp$ and

$$G_\varepsilon(v) = -\sqrt{\mu(v)} + 2^{-\frac{5}{4}} (1 + \varepsilon^2)^{3/2} \left( \sqrt{\mu(v)} + \frac{1}{\varepsilon^3} \sqrt{\mu(v/\varepsilon)} \right).$$

Then we have the following limit in the sense of distribution as $\varepsilon \to 0$:

$$F_\varepsilon \to F = 2^{-\frac{5}{4}} (\mu(2^{-\frac{1}{4}}) + (2^{\frac{3}{4}})^3 \delta),$$

$$G_\varepsilon \to G = -\sqrt{\mu} + 2^{-\frac{11}{16}} \pi^{-\frac{3}{4}} + 2^{-\frac{1}{4}} \pi^{\frac{3}{4}} \delta.$$
where
\[ I_1 = \left( -\sqrt{\mu}, \varphi_n \right)_{L^2} = \left( -\varphi_0, \varphi_n \right)_{L^2} = -\delta_{0,n}, \]
\[ I_2 = \left( \sqrt{\mu(\varepsilon)}, \varphi_n \right)_{L^2} = \left( 2^{\frac{3}{2}}\pi^{\frac{3}{2}} \right) \varphi_n(0) \frac{(1 - \varepsilon^2)^n}{(1 + \varepsilon^2)^{n+3/2}}, \]
\[ I_3 = ( -1 )^n I_2. \]
Finally we get:
\[ G_{\varepsilon,n} = -\delta_{0,n} + \frac{1 + (-1)^n (1 - \varepsilon^2)^n}{2} \left( \frac{(2n + 1)!}{2^n n! (n!)^2} \right)^{\frac{1}{2}}. \]

7.3. Some results on the spherical harmonics. We recall that
\[ \varphi_n(v) = \left( \frac{n!}{\sqrt{2\Gamma(n + 3/2)}} \right)^{1/2} e^{-\frac{|v|^2}{4}} L_n^{(\frac{3}{2})} \left( \frac{|v|^2}{2} \right) \frac{1}{\sqrt{4\pi}} \]
where the Laguerre polynomial \( L_n^{(\alpha)} \) of order \( \alpha \), degree \( n \) is
\[ L_n^{(\alpha)}(x) = \sum_{r=0}^{n} (-1)^{n-r} \frac{\Gamma(\alpha + n + 1)}{r!(n-r)!\Gamma(\alpha+n-r+1)} x^{n-r}. \]

**Lemma 7.5.** For \( a > 0 \) and \( n \geq 0 \) we have
\[ \varphi_n(0) = \frac{1}{(2\pi)^{\frac{3}{2}}} \left( \frac{(2n + 1)!}{2^n n! (n!)^2} \right)^{\frac{1}{2}}, \]
\[ \int_{\mathbb{R}^3} \varphi_n(v) dv = (-1)^{n} 2^{3/2} \pi^{\frac{3}{2}} \varphi_n(0), \]
\[ \left( \sqrt{\mu(a)}, \varphi_n \right)_{L^2} = \left( 2^{\frac{3}{2}}\pi^{\frac{3}{2}} \right) \varphi_n(0) \frac{(1 - a^2)^n}{(1 + a^2)^{n+3/2}}. \]

**Proof.** These equalities come from classical properties of the Hermite functions (we have checked them using Maple\(^{\text{®}}\)13 for integers \( n \leq 20 \)). \( \square \)

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