

ASC Report No. 32/2015

On hypocoercive BGK models

F. Achleitner, A. Arnold, E.A. Carlen

Institute for Analysis and Scientific Computing —
Vienna University of Technology — TU Wien
www.asc.tuwien.ac.at ISBN 978-3-902627-05-6

Most recent ASC Reports

- 31/2015 *M. Feischl, T. Führer, M. Niederer, S. Strommer, A. Steinboeck, and D. Praetorius*
Efficient numerical computation
of direct exchange areas
in thermal radiation analysis
- 30/2015 *B. Düring, P. Fuchs, and A. Jüngel*
A higher-order gradient flow scheme
for a singular one-dimensional
diffusion equation
- 29/2015 *C. Erath and D. Praetorius*
Adaptive finite volume methods with convergence rates
- 28/2015 *W. Auzinger, O. Koch, M. Schöbinger, E. Weinmüller*
A new version of the code bvpsuite for singular BVPs in ODEs: Nonlinear solver
and its application to m-Laplacians.
- 27/2015 *C. Lehrenfeld, J. Schöberl*
High order exactly divergence-free Hybrid Discontinuous Galerkin Methods for
unsteady incompressible flows
- 26/2015 *A. Jüngel, and S. Schuchnigg*
Entropy-dissipating semi-discrete Runge-Kutta schemes for nonlinear diffusion
equations
- 25/2015 *W. Auzinger, H. Hofstätter, D. Ketcheson, O. Koch*
Practical splitting methods for the adaptive integration of nonlinear evolution
equations.
Part I: Construction of optimized schemes and pairs of schemes
- 24/2015 *J. Burkotová, M. Hubner, I. Rachunková, and E.B. Weinmüller*
Asymptotic properties of Kneser solutions to nonlinear second order ODEs with
regularly varying coefficients
- 23/2015 *C. Abert, G. Hrkac, D. Praetorius, M. Ruggeri, D. Suess*
Coupling of dynamical micromagnetism and a stationary spin drift-diffusion
equation: A step towards a fully implicit spintronics frame
- 22/2015 *F. Achleitner, A. Arnold and D. Stürzer*
Large-time behavior in non-symmetric Fokker-Planck equations

Institute for Analysis and Scientific Computing
Vienna University of Technology
Wiedner Hauptstraße 8–10
1040 Wien, Austria

E-Mail: admin@asc.tuwien.ac.at

WWW: <http://www.asc.tuwien.ac.at>

FAX: +43-1-58801-10196

ISBN 978-3-902627-05-6

© Alle Rechte vorbehalten. Nachdruck nur mit Genehmigung des Autors.



On hypocoercive BGK models

Franz Achleitner, Anton Arnold, Eric A. Carlen

Abstract We study hypocoercivity for a class of linear BGK models for discrete and continuous phase spaces. We develop methods for constructing entropy functionals that prove exponential rates of relaxation to equilibrium.

1 Introduction

This paper is concerned with the large time behavior of linear BGK models (named after the physicists Bhatnagar-Gross-Krook [8]) for a phase space density $f(x, v, t)$; $x, v \in \mathbb{R}^d$, satisfying the kinetic evolution equation

$$f_t + v \cdot \nabla_x f - \nabla_x V \cdot \nabla_v f = \mathbf{Q}f := M_{T(t)}(v) \int_{\mathbb{R}^d} f(x, v, t) \, dv - f(x, v, t), \quad t \geq 0, \quad (1.1)$$

with some given confinement potential $V(x)$ and where M_T denotes the normalized Maxwellian at some temperature T :

$$M_T(v) = (2\pi T)^{-d/2} e^{-|v|^2/2T}.$$

The function $T(t)$ is defined so that the energy is conserved:

Franz Achleitner

Vienna University of Technology, Institute of Analysis and Scientific Computing, Wiedner Hauptstr. 8-10, A-1040 Wien, Austria, e-mail: franz.achleitner@tuwien.ac.at

Anton Arnold

Vienna University of Technology, Institute of Analysis and Scientific Computing, Wiedner Hauptstr. 8-10, A-1040 Wien, Austria, e-mail: anton.arnold@tuwien.ac.at

Eric A. Carlen

Department of Mathematics, Rutgers University, 110 Frelinghuysen Rd., Piscataway NJ 08854, USA, e-mail: carlen@math.rutgers.edu

$$\int_{\mathbb{R}^d \times \mathbb{R}^d} \left[\frac{|v|^2}{2} + V(x) \right] f(x, v, t) \, dx dv = \int_{\mathbb{R}^d \times \mathbb{R}^d} \left[\frac{|v|^2}{2} + V(x) \right] f(x, v, 0) \, dx dv =: E_0 .$$

This is achieved in case

$$T(t) := \frac{2}{d} \left[E_0 - \int_{\mathbb{R}^d} V(x) \rho(x, t) \, dx \right] , \quad (1.2)$$

where $\rho(x, t) := \int_{\mathbb{R}^d} f(x, v, t) \, dv$, which completes the specification of the equation.

This model differs from the usual BGK model in that the Maxwellian M_T has a spatially constant temperature with zero momentum. This is already a simplification of the standard BGK model in which M_T would be replaced by the local Maxwellian corresponding to f ; i.e., the local Maxwellian with the same hydrodynamic moments as f . However, it is still non-linear since $T(t)$ depends linearly on f , but then M_T depends nonlinearly on T . This simplified equation arises in certain models of thermostated systems [7]. In fact, a further simplification is studied in this context: We take $d = 1$ replace the spatial domain \mathbb{R}^d by the unit circle \mathbf{T}^1 , and then dispense with the confining potential. In this case, the equation (1.2) for the temperature reduces to $T(t) = (2/d)E_0$, independent of t . This is the linear BGK model that we investigate:

$$f_t + v f_x = \mathbf{Q}f := M_T(v) \int_{\mathbb{R}} f(x, v, t) \, dv - f(x, v, t), \quad t \geq 0. \quad (1.3)$$

Under sufficient growth assumptions on V as $|x| \rightarrow \infty$, the unique normalized steady state of (1.1) is

$$f^\infty(x, v) = \exp \left(-\frac{1}{T_\infty} \left[V(x) + \frac{|v|^2}{2} \right] \right),$$

where the normalization constant shall be included in V and T_∞ such that the energy associated to f^∞ is E_0 . For the simplified linear equation (1.3), the unique steady state is even simpler; it is $f^\infty = M_T$, uniform on the circle, provided we normalize so that the total mass (also conserved) is one.

We shall study the rate at which normalized solutions of (1.3) approach the steady state $f^\infty = M_T$ as $t \rightarrow \infty$. This problem is interesting since the collision mechanism drives the local velocity distribution towards M_T , but a more complicated mechanism involving the interaction of the streaming term $v \partial_x$ and the collision operator \mathbf{Q} is responsible for the emergence of spatial uniformity.

To elucidate this key point, let us define the operator \mathbf{L} by

$$\mathbf{L}f(x, v) := -v \partial_x f(x, v) + \mathbf{Q}f(x, v) .$$

Then the evolution equation (1.3) can be written $f_t = \mathbf{L}f$. Let \mathcal{H} denote the weighted space $L^2(\mathbf{T}^1 \times \mathbb{R}; M_T^{-1})$. Then \mathbf{L} is self-adjoint on \mathcal{H} , $\mathbf{L}f^\infty = 0$, and a simple computation shows that if $f(t)$ is a solution of (1.3),

$$\frac{d}{dt} \|f(t) - f^\infty\|_{\mathcal{H}}^2 = 2\langle f(t), \mathbf{L}f(t) \rangle_{\mathcal{H}} = 2\langle f(t), \mathbf{Q}f(t) \rangle_{\mathcal{H}} = -2\|f - M_T \rho\|_{\mathcal{H}}^2,$$

where, as before, $\rho(x, t) := \int_{\mathbb{R}} f(x, v, t) dv$. Thus, while the norm $\|f(t) - f^\infty\|_{\mathcal{H}}$ is monotone decreasing, the derivative is zero whenever $f(t)$ has the form $f(t) = M_T \rho$ for any smooth density ρ . In particular, the inequality

$$\langle f - f^\infty, \mathbf{L}(f - f^\infty) \rangle_{\mathcal{H}} \leq -\lambda \|f - f^\infty\|_{\mathcal{H}}^2 \quad (1.4)$$

is valid in general for $\lambda = 0$, but for no positive value of λ . If (1.4) were valid for some $\lambda > 0$, we would have had $\|f(t) - f^\infty\|_{\mathcal{H}}^2 \leq e^{-t\lambda} \|f(0) - f^\infty\|_{\mathcal{H}}^2$ for all solutions of our equation, and we would say that the evolution equation is *coercive*. However, while this is not the case, it does turn out that one still has constants $1 < C < \infty$ and $\lambda > 0$ such that

$$\|f(t) - f^\infty\|_{\mathcal{H}}^2 \leq C e^{-t\lambda} \|f(0) - f^\infty\|_{\mathcal{H}}^2. \quad (1.5)$$

(The fact that there exist initial data $f(0) \neq f^\infty$ for which the derivative of the norm is zero shows that necessarily $C > 1$.) In Villani's terminology (see §3.2 of [20]), this means that our evolution equation is *hypocoercive*.

While many hypocoercive equations have been studied in recent years [20, 12, 3], BGK models were so far only analyzed in form of a two velocity toy model [12] (see also §4.1 below). The fact that normalized solutions of (1.3) converge exponentially fast at *some* rate to f^∞ is a consequence of a probabilistic analysis of such equations in [7]: In fact, equation (1.3) is the Kolmogorov forward equation for a certain Markov process, and as shown in [7] an argument based on a Doeblin condition yields exponential convergence. However, this approach relies on compactness arguments and does not yield explicit values for C or λ . We shall discuss another approach to the problem of establishing hypocoercivity for such models that does yield explicit – and quite reasonable – values for C and λ . To this end, our main tool will be variants of the *entropy–entropy production method*. Our main result will be a decay estimate for (1.3):

Theorem 1. [decay estimate for (1.3)] *Fix unit temperature $T = 1$. There exists an entropy functional $e(f)$ satisfying*

$$2\|f - M_1\|_{L^2(M_1^{-1})}^2 \geq e(f) \geq \frac{1}{4}\|f - M_1\|_{L^2(M_1^{-1})}^2$$

such that for all solutions $f(t)$ of (1.3) with $e(f(0)) < \infty$,

$$e(f(t)) \leq e(f(0))e^{-t \cdot 0.547592\dots}.$$

Before turning to our main investigation we study some still simpler models with a finite number of positions and velocities: In §2 we analyze coercive BGK models with just two velocities using relative entropies. Since this approach fails for discrete hypocoercive BGK models (considered in §3), their analysis will be based on

spectral methods and Lyapunov's direct method. §4 is concerned with the space-inhomogeneous BGK model (1.3) and its discrete velocity analogs. In that section we shall finally give the proof of Theorem 1.

2 Discrete coercive BGK models

In this section we consider space-homogeneous BGK models with two-velocities. Our main tool in the investigation is the relative entropy, which is defined as follows (see §2.2 of [4] for more details):

Definition 1. (a) Let J be either \mathbb{R}^+ or \mathbb{R} . A scalar function $\psi \in C(\bar{J}) \cap C^2(J)$ satisfying the conditions

$$\psi(1) = 0, \quad \psi \geq 0, \quad \psi'' \geq 0, \quad \text{on } J \quad (2.1)$$

(and hence also $\psi'(1) = 0$) is called *entropy generator*.

(b) Let $f_1 \in L^1(\mathbb{R}^{2d})$, $f_2 \in L^1_+(\mathbb{R}^{2d})$ with $\int \int f_1 \, dx \, dv = \int \int f_2 \, dx \, dv = 1$ and $\frac{f_1}{f_2}(x, v) \in \bar{J}$ a.e. (w.r.t. the measure $f_2(dx \, dv)$). Then

$$e_\psi(f_1|f_2) := \int \int_{\mathbb{R}^{2d}} \psi\left(\frac{f_1}{f_2}\right) f_2 \, dx \, dv \geq 0 \quad (2.2)$$

is called a *relative entropy* of f_1 with respect to f_2 with generating function ψ .

In applications, the most important examples are the logarithmic entropy $e_1(f_1|f_2)$, generated by

$$\psi_1(\sigma) = \sigma \ln \sigma - \sigma + 1,$$

and the power law entropies $e_p(f_1|f_2)$, generated by

$$\psi_p(\sigma) = \sigma^p - 1 - p(\sigma - 1), \quad p > 1. \quad (2.3)$$

Except for the quadratic entropy e_2 we shall always use $J = \mathbb{R}^+$. There is also a second family of power law entropies $\hat{e}_p(f_1|f_2)$ generated by

$$\hat{\psi}_p(\sigma) = |\sigma - 1|^p, \quad p > 1. \quad (2.4)$$

The above definition clearly shows that $e_\psi(f_1|f_2) = 0$ iff $f_1 = f_2$. In the next section we shall hence try to prove that solutions $f(t)$ to BGK models satisfy $e_\psi(f(t)|f^\infty) \rightarrow 0$ as $t \rightarrow \infty$. For the entropies e_p , $p \geq 1$ such a convergence in relative entropy then also implies L^1 -convergence, due to the *Csiszár-Kullback inequality*:

$$\|f_1 - f_2\|_{L^1(\mathbb{R}^{2d})}^2 \leq 2e_1(f_1|f_2) \leq \frac{2}{p(p-1)} e_p(f_1|f_2),$$

where we used $\psi_1(\sigma) \leq \psi_p(\sigma)/\psi_p''(1)$, $\sigma \geq 0$ in the second inequality. For the entropies defined in (2.4) one has a substitute for the Csiszár-Kullback inequality,

namely the identity

$$\hat{e}_p(f_1|f_2) = \|f_1 - f_2\|_{L^p(f_2^{1-p})}^p.$$

To illustrate the standard entropy method on a very simple example, we first revisit the ODE (1.10) from [4] for the vector $f(t) = (f_1(t), f_2(t))^\top \in \mathbb{R}^2$:

$$\begin{aligned} \frac{d}{dt}f &= \lambda \mathbf{A}f, \quad t \geq 0, \\ f(0) &= f^I \in \mathbb{R}^2, \end{aligned} \quad (2.5)$$

with the parameter $\lambda > 0$, and the matrix \mathbf{A} has BGK form:

$$\mathbf{A} := \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} = 2 \left[\begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes (1, 1) - \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right]. \quad (2.6)$$

This ODE can be seen as an x -homogeneous variant of (1.1) with just two discrete velocities. In fact, on the right hand side of (2.6), the column vector $(\frac{1}{2}, \frac{1}{2})^\top$ corresponds to the Maxwellian $M(v)$ in the BGK equation (1.1), and the row vector $(1, 1)$ corresponds to the velocity integral. The symmetric matrix \mathbf{A} has an eigenvalue 0 with corresponding eigenvector $f^\infty := (\frac{1}{2}, \frac{1}{2})^\top$ and an eigenvalue -2. Hence \mathbf{A} is coercive on $\{f^\infty\}^\perp$. Since each column of \mathbf{A} sums up to 0, the ‘‘total mass’’ of the system, i.e. $f_1(t) + f_2(t)$, stays constant in time. Hence, we shall assume w.l.o.g. that f^I is normalized, i.e. $f_1^I + f_2^I = 1$. Thus, as $t \rightarrow \infty$, $f(t) = f^\infty + (f^I - f^\infty) e^{-2\lambda t}$ converges to f^∞ exponentially with rate 2λ . For $f_{1,2}^I \geq 0$ we have $f_{1,2}(t) \geq 0$.

In analogy to Definition 1 we introduce for (2.5) (with $n = 2$) the relative entropy generated by ψ :

$$e_\psi(f(t)|f^\infty) := \sum_{j=1}^n \psi\left(\frac{f_j(t)}{f_j^\infty}\right) f_j^\infty. \quad (2.7)$$

Its time derivative under the flow of (2.5) reads

$$\begin{aligned} \frac{d}{dt}e_\psi(f(t)|f^\infty) &= -\lambda(f_1 - f_2) \left[\psi'\left(\frac{f_1(t)}{f_1^\infty}\right) - \psi'\left(\frac{f_2(t)}{f_2^\infty}\right) \right] \\ &=: -I_\psi(f(t)|f^\infty) = -2\lambda \psi''(\zeta)(f_1 - f_2)^2 \leq 0, \end{aligned} \quad (2.8)$$

where ζ is an intermediate value between $2f_1(t)$ and $2f_2(t)$. $I_\psi(f(t)|f^\infty)$ denotes the *Fisher information* (of $f(t)$ w.r.t. f^∞).

As pointed out in [4], it is not obvious to bound this Fisher information from below directly by a multiple of the relative entropy (except for quadratic entropies). The goal of such an estimate would be to establish the exponential decay of the relative entropy. Hence, it is the essence of the entropy method to consider the entropy dissipation rate: Differentiating (2.8) once more in time gives

$$\begin{aligned}
R_\psi(f(t)|f^\infty) &:= -\frac{d}{dt}I_\psi(f(t)|f^\infty) \\
&= 2\lambda I_\psi(f(t)|f^\infty) + \lambda^2 (f_1(t) - f_2(t))^2 \left[\psi''\left(\frac{f_1(t)}{f_1^\infty}\right) \frac{1}{f_1^\infty} + \psi''\left(\frac{f_2(t)}{f_2^\infty}\right) \frac{1}{f_2^\infty} \right].
\end{aligned} \tag{2.9}$$

Due to $\psi'' \geq 0$ the second term is nonnegative. Hence,

$$-\frac{d}{dt}I_\psi(f(t)|f^\infty) \geq 2\lambda I_\psi(f(t)|f^\infty).$$

And this yielded in [4] the exponential decay of $I_\psi(f(t)|f^\infty)$ and of $e_\psi(f(t)|f^\infty)$ at the *sub-optimal* rate 2λ . But this procedure can be improved easily to give the following sharp result:

Theorem 2. *Let the convex entropy generator ψ satisfy either: ψ'' is convex on J ; or ψ' is concave on $(0, 1)$ along with ψ' is convex on $(1, \infty)$. Then the solution to (2.5) satisfies*

$$I_\psi(f(t)|f^\infty) \leq e^{-4\lambda t} I_\psi(f^I|f^\infty), \quad t \geq 0, \tag{2.10}$$

$$e_\psi(f(t)|f^\infty) \leq e^{-4\lambda t} e_\psi(f^I|f^\infty), \quad t \geq 0. \tag{2.11}$$

Proof. Case 1: ψ'' convex on J

We have for $0 \leq s \leq 1$:

$$s\psi''(\sigma_2) + (1-s)\psi''(\sigma_1) \geq \psi''(s\sigma_2 + (1-s)\sigma_1).$$

Integrating this inequality over $s \in [0, 1]$ yields $\forall \sigma_1 \neq \sigma_2 \in J$:

$$\frac{\psi''(\sigma_1) + \psi''(\sigma_2)}{2} \geq \kappa \frac{\int_{\sigma_1}^{\sigma_2} \psi''(\sigma) d\sigma}{\sigma_2 - \sigma_1} = \kappa \frac{\psi'(\sigma_2) - \psi'(\sigma_1)}{\sigma_2 - \sigma_1}, \tag{2.12}$$

where $\kappa = 1$ is introduced only for later reference.

We now recall that $f_1^\infty = f_2^\infty$. Hence, (2.9) and (2.12) give

$$\frac{d}{dt}I_\psi(f(t)|f^\infty) \leq -4\lambda I_\psi(f(t)|f^\infty), \tag{2.13}$$

and (2.10) follows. As usual in the entropy method, one next integrates (2.13) in time (from t to ∞) to obtain

$$\frac{d}{dt}e_\psi(f(t)|f^\infty) \leq -4\lambda e_\psi(f(t)|f^\infty),$$

and this finishes the proof for the case ψ'' convex.

Case 2: ψ' concave on $(0, 1)$ along with ψ' convex on $(1, \infty)$

We may assume without loss of generality that $f_1 > f_2$. Then $f_1/f_1^\infty > 1 > f_2/f_2^\infty$, and by the tangent line inequality for the concave function $\psi'|_{(0,1)}$,

$$0 = \psi'(1) \leq \psi' \left(\frac{f_2}{f_2^\infty} \right) + \psi'' \left(\frac{f_2}{f_2^\infty} \right) \left(\frac{f_2^\infty - f_2}{f_2^\infty} \right).$$

Likewise, using the tangent line inequality for the convex function $\psi'|_{(1,\infty)}$,

$$\psi' \left(\frac{f_1}{f_1^\infty} \right) \leq \psi'(1) + \psi'' \left(\frac{f_1}{f_1^\infty} \right) \left(\frac{f_1 - f_1^\infty}{f_1^\infty} \right) = \psi'' \left(\frac{f_1}{f_1^\infty} \right) \left(\frac{f_1 - f_1^\infty}{f_1^\infty} \right).$$

Altogether we have

$$\psi'' \left(\frac{f_1}{f_1^\infty} \right) \left(\frac{f_1 - f_1^\infty}{f_1^\infty} \right) \geq \psi' \left(\frac{f_1}{f_1^\infty} \right) \quad \text{and} \quad \psi'' \left(\frac{f_2}{f_2^\infty} \right) \left(\frac{f_2^\infty - f_2}{f_2^\infty} \right) \geq -\psi' \left(\frac{f_2}{f_2^\infty} \right). \quad (2.14)$$

Now continuing to assume that $f_1 > f_2$, and using the fact that $f_1^\infty = f_2^\infty$ so that $f_1 - f_2 = 2(f_1 - f_1^\infty) = 2(f_2^\infty - f_2)$,

$$\begin{aligned} (f_1 - f_2) & \left[\psi'' \left(\frac{f_1}{f_1^\infty} \right) \frac{1}{f_1^\infty} + \psi'' \left(\frac{f_2}{f_2^\infty} \right) \frac{1}{f_2^\infty} \right] \\ & = 2(f_1 - f_1^\infty) \psi'' \left(\frac{f_1}{f_1^\infty} \right) \frac{1}{f_1^\infty} + 2(f_2^\infty - f_2) \psi'' \left(\frac{f_2}{f_2^\infty} \right) \frac{1}{f_2^\infty} \\ & \geq 2 \left[\psi' \left(\frac{f_1}{f_1^\infty} \right) - \psi' \left(\frac{f_2}{f_2^\infty} \right) \right]. \end{aligned}$$

Therefore,

$$\lambda^2 (f_1 - f_2)^2 \left[\psi'' \left(\frac{f_1}{f_1^\infty} \right) \frac{1}{f_1^\infty} + \psi'' \left(\frac{f_2}{f_2^\infty} \right) \frac{1}{f_2^\infty} \right] \geq 2\lambda I_\psi(f|f^\infty).$$

Again from (2.9) we obtain (2.13). \square

Remark:

1. Concerning the logarithmic and power law entropies from (2.3) one easily verifies: ψ_p satisfies the condition $\psi^{IV} \geq 0$ on J (or the inequality (2.12)) exactly for $p \in [1, 2] \cup [3, \infty)$.
2. For ψ_p with $p \in (2, 3)$, inequality (2.12) holds with $\kappa = \frac{p-1}{2}$ (but not for any larger constant κ). This follows from $g_p(z) := z^{p-2} + 1 - \frac{z^{p-1}-1}{z-1} > 0$ on \mathbb{R}^+ and $g_p(0) = 0$, which can be verified by elementary computations. Hence, for $p \in (2, 3)$, the entropy method yields exponential decay of $e_p(f(t)|f^\infty)$ with the reduced rate $2(\kappa + 1)\lambda = (p + 1)\lambda$:

$$e_p(f(t)|f^\infty) \leq e^{-(p+1)\lambda t} e_{\psi_p}(f^I|f^\infty), \quad t \geq 0.$$

But the decay estimates (2.11), (2.10) are in general false for $p \in (2, 3)$.

In an alternative approach, one can verify for $2 < p < 3$ the estimates

$$\psi_p(\sigma) \leq \psi_3(\sigma), \quad \forall \sigma \geq 0; \quad \psi_3(\sigma) \leq C_p \psi_p(\sigma), \quad \forall 0 \leq \sigma \leq 2,$$

where $[0, 2]$ is the maximum range of values for $\frac{f_1}{f_1^\infty}$ and $\frac{f_2}{f_2^\infty}$. Here the constant is $C_p = \frac{\psi_3(2)}{\psi_p(2)} = \frac{4}{2^{p-1}-p}$. With (2.11) this implies

$$e_p(f(t)|f^\infty) \leq e^{-4\lambda t} e_3(f^I|f^\infty) \leq C_p e^{-4\lambda t} e_p(f^I|f^\infty), \quad t \geq 0.$$

Hence, the entropies e_p , $p \in (2, 3)$ still decay with the optimal rate 4λ , but at the price of the multiplicative constant $C_p > 1$.

3. The relative entropies \hat{e}_p , $p \geq 2$ from (2.4) satisfy the second set of assumptions in Theorem 2. Note that ψ''' does not have to be continuous at $\sigma = 1$.

2.1 Multi-velocity BGK models

Now, we consider discrete space-homogeneous BGK models in \mathbb{R}^n : The evolution of a vector $f(t) = (f_1(t), f_2(t), \dots, f_n(t))^\top \in \mathbb{R}^n$ is governed by

$$\begin{cases} \frac{d}{dt} f = 2\lambda \mathbf{A} f, & t \geq 0, \\ f(0) = f^I \in \mathbb{R}^n, \end{cases} \quad (2.15)$$

for some $\lambda > 0$ and a matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ in BGK form

$$\mathbf{A} = \begin{pmatrix} \rho_1 \\ \vdots \\ \rho_n \end{pmatrix} \otimes (1, \dots, 1) - \mathbf{I} \quad (2.16)$$

with $\rho = (\rho_1, \dots, \rho_n)^\top \in (0, 1)^n$ such that $\sum_{j=1}^n \rho_j = 1$.

Such a matrix \mathbf{A} has a simple eigenvalue 0 with left eigenvector $l_1 = (1, \dots, 1)^\top$ and right eigenvector $r_1 = \rho$, and an eigenvalue -1 with geometric multiplicity $n - 1$. Since each column of \mathbf{A} sums up to 0, the ‘‘total mass’’ of system (2.15) stays constant in time, i.e. $\sum_{j=1}^n f_j(t) = \sum_{j=1}^n f_j^I$.

Matrix $\mathbf{A} = (a_{jk})_{j,k=1,\dots,n}$ has only non-negative off-diagonal coefficients a_{jk} ($j \neq k$); such matrices are called *essentially non-negative* or Metzler matrices [18]. An essentially non-negative matrix \mathbf{A} induces via (2.15) a semi-flow which preserves non-negativity of its initial datum f^I , i.e. $f_j^I \geq 0$ for all $j = 1, \dots, n$, implies $f_j(t) \geq 0$ for all $t \geq 0$.

Remark: An essentially non-negative matrix is called *Q-matrix* (or *W-matrix* in [19]) if it has an eigenvalue 0 with right eigenvector $(1, \dots, 1)^\top$. *Q*-matrices are the infinitesimal generators of continuous-time Markov processes with finite state space [16].

In the following, we consider normalized positive initial data f^I , i.e. $\sum_{j=1}^n f_j^I = 1$, such that the solution f of (2.15) is positive and normalized for all $t \geq 0$. Thus, as

$t \rightarrow \infty$, $f(t) = f^\infty + (f^I - f^\infty) e^{-2\lambda t}$ converges to the normalized steady state $f^\infty := \rho$ exponentially with rate 2λ .

The study of the long-time behavior of solutions f to (2.15) is a classical topic, an approach via entropy methods can be found in [19, 17]. Note that Perthame [17, §6.3] considers essentially positive matrices (i.e. off-diagonal elements are positive) to simplify the presentation. However, the results generalize to irreducible Q -matrices, since only the non-negativity of off-diagonal elements is used, see also [17, Remark 6.2]. While [17, Proposition 6.5] establishes only exponential decay in entropy, we aim at the optimal decay rate in the entropy approach.

We consider the time derivative of the relative entropy (2.7) under the flow of (2.15)

$$\frac{d}{dt} e_\psi(f(t)|f^\infty) = \sum_{j=1}^n \psi' \left(\frac{f_j(t)}{f_j^\infty} \right) 2\lambda (f_j^\infty - f_j(t)) =: -I_\psi(f(t)|f^\infty) \leq 0 \quad (2.17)$$

which is non-positive due to the properties (2.1) of an entropy generator (ψ' is an increasing function with $\psi'(1) = 0$). Next, we compute the second order derivative of $e_\psi(f(t)|f^\infty)$ w.r.t. time:

$$\begin{aligned} R_\psi(f(t)|f^\infty) &:= -\frac{d}{dt} I_\psi(f(t)|f^\infty) = \frac{d}{dt} \sum_{j=1}^n \psi' \left(\frac{f_j(t)}{f_j^\infty} \right) \frac{d}{dt} f_j \\ &= \sum_{j=1}^n \psi' \left(\frac{f_j(t)}{f_j^\infty} \right) \frac{d^2}{dt^2} f_j + \sum_{j=1}^n \psi'' \left(\frac{f_j(t)}{f_j^\infty} \right) \frac{1}{f_j^\infty} \left(\frac{d}{dt} f_j \right)^2 \\ &= 2\lambda I_\psi(f(t)|f^\infty) + \sum_{j=1}^n \psi'' \left(\frac{f_j(t)}{f_j^\infty} \right) \frac{1}{f_j^\infty} \left(\frac{d}{dt} f_j \right)^2 \geq 2\lambda I_\psi(f(t)|f^\infty), \end{aligned}$$

since $\mathbf{A}^2 = -\mathbf{A}$ and $\psi'' \geq 0$. This yields the non-optimal entropy dissipation rate 2λ . To obtain a better entropy dissipation rate, we want to estimate the neglected term via

$$\sum_{j=1}^n \psi'' \left(\frac{f_j(t)}{f_j^\infty} \right) \frac{1}{f_j^\infty} \left(\frac{d}{dt} f_j \right)^2 \geq \mu I_\psi(f(t)|f^\infty) \geq 0 \quad (2.18)$$

for some $\mu > 0$.

Theorem 3. Let $\rho = (\rho_1, \dots, \rho_n)^\top \in (0, 1)^n$ such that $\sum_{j=1}^n \rho_j = 1$ and let the convex entropy generator $\psi \in C^2(J)$ satisfy for some $\mu > 0$ and all $u = (u_1, \dots, u_n)^\top \in [0, 1]^n$ with $\sum_{j=1}^n u_j = 1$:

$$\sum_{j=1}^n \psi'' \left(\frac{u_j}{\rho_j} \right) \frac{1}{\rho_j} (\rho_j - u_j)^2 \geq \frac{\mu}{2\lambda} \sum_{j=1}^n \psi' \left(\frac{u_j}{\rho_j} \right) (u_j - \rho_j). \quad (2.19)$$

Then, for all non-negative normalized initial data f^I , the solution f to (2.15) satisfies

$$I_\psi(f(t)|f^\infty) \leq e^{-(2\lambda+\mu)t} I_\psi(f^t|f^\infty), \quad t \geq 0, \quad (2.20)$$

$$e_\psi(f(t)|f^\infty) \leq e^{-(2\lambda+\mu)t} e_\psi(f^t|f^\infty), \quad t \geq 0. \quad (2.21)$$

Proof. The solution f to (2.15) is positive and normalized for all $t > 0$. Under Assumption (2.19) on ψ , we obtain the estimates (2.18), and

$$\frac{d}{dt} I_\psi(f(t)|f^\infty) \leq -(2\lambda + \mu) I_\psi(f(t)|f^\infty), \quad (2.22)$$

hence (2.20) follows. Next, one integrates (2.22) in time (from t to ∞) to obtain

$$\frac{d}{dt} e_\psi(f(t)|f^\infty) \leq -(2\lambda + \mu) e_\psi(f(t)|f^\infty),$$

and this finishes the proof. \square

For the quadratic entropy generator ψ_2 inequality (2.19) holds with $\mu = 2\lambda$. Thus we recover the optimal decay rate 4λ in (2.20)–(2.21). For the logarithmic entropy generator ψ_1 an estimate for μ in (2.19) has been given in [11, 9] as

$$\frac{\mu}{2\lambda} \geq \sqrt{\rho_{\min}(1 - \rho_{\min})} \quad \text{with } \rho_{\min} = \min_{j=1, \dots, n} \rho_j.$$

Next, we consider entropy generators ψ in the sense of Definition 1, such that ψ' is concave on $(0, 1)$ along with ψ' convex on $(1, \infty)$. Thus, for $f_1 \geq f_1^\infty > 0$ and $f_2^\infty \geq f_2 > 0$, the inequalities (2.14) continue to hold. Distinguishing the cases $u_j < \rho_j$, $u_j > \rho_j$ and the trivial case $u_j = \rho_j$, we deduce for all $j = 1, \dots, n$,

$$\psi''\left(\frac{u_j}{\rho_j}\right) \frac{1}{\rho_j} (\rho_j - u_j)^2 \geq \psi'\left(\frac{u_j}{\rho_j}\right) (u_j - \rho_j),$$

hence (2.19) holds with $\mu = 2\lambda$. However, for the entropy generators $\hat{\psi}_p$ in (2.4) with $p \geq 2$ the optimal value is $\mu = (p - 1)2\lambda$.

In the following, we restrict ourselves to $n = 2$ and determine the best constant for some polynomial entropy generators:

Lemma 1. *Let $\rho_1, \rho_2 \in (0, 1)$ with $\rho_1 + \rho_2 = 1$. The entropy generator $\psi(\sigma)$ satisfies condition (2.19) with*

$$1 \geq \frac{\mu}{2\lambda} = \begin{cases} 1 & \text{for } \psi(\sigma) = \psi_2(\sigma), \\ 2 \min\{\rho_1, \rho_2\} & \text{for } \psi(\sigma) = \psi_3(\sigma), \\ 2 - 2\sqrt{1 - 3\rho_2(1 - \rho_2)} > 0 & \text{for } \psi(\sigma) = \psi_4(\sigma). \end{cases}$$

Proof. For $n = 2$, the assumptions on ρ and u in (2.19) imply

$$-(\rho_2 - u_2) = \rho_1 - u_1 = \rho_1(u_1 + u_2) - u_1 = \rho_1 u_2 - \rho_2 u_1 = \rho_1 \rho_2 \left(\frac{u_2}{\rho_2} - \frac{u_1}{\rho_1} \right).$$

Thus condition (2.19) is equivalent to

$$\sum_{j=1}^2 \psi''\left(\frac{u_j}{\rho_j}\right) \frac{1}{\rho_j} \rho_1^2 \rho_2^2 \left(\frac{u_2}{\rho_2} - \frac{u_1}{\rho_1}\right)^2 \geq \frac{\mu}{2\lambda} \sum_{j=1}^2 \psi'\left(\frac{u_j}{\rho_j}\right) (-1)^j \rho_1 \rho_2 \left(\frac{u_2}{\rho_2} - \frac{u_1}{\rho_1}\right) \geq 0.$$

Setting $v_1 := u_1/\rho_1$ and $v_2 := u_2/\rho_2$, we deduce for $\psi_p(\sigma)$, $p > 1$,

$$(p-1)[v_1^{p-2}\rho_2 + v_2^{p-2}\rho_1](v_1 - v_2)^2 \geq \frac{\mu}{2\lambda}[v_1^{p-1} - v_2^{p-1}](v_1 - v_2) \quad \forall v_1, v_2 \geq 0.$$

Moreover, for $v_2 > 0$, dividing by v_2^p and defining $z := v_1/v_2$, we obtain

$$(p-1)[z^{p-2}\rho_2 + \rho_1](z-1)^2 \geq \frac{\mu}{2\lambda}[z^{p-1} - 1](z-1) \quad \forall z \geq 0.$$

We show the statement for the quartic entropy generator $\psi_4(\sigma)$, the (simpler) proof for quadratic and cubic entropy generators is omitted. For ψ_4 , condition (2.19) is equivalent to

$$g(z) := z^2(3\rho_2 - \tilde{\mu}) - \tilde{\mu}z + 3\rho_1 - \tilde{\mu} \geq 0 \quad \forall z \geq 0$$

with $\tilde{\mu} = \mu/(2\lambda)$. Evaluating $g(z)$ at $z = 0$ and taking the limit $z \rightarrow \infty$, we deduce the necessary conditions $3\rho_1 \geq \tilde{\mu}$ and $3\rho_2 > \tilde{\mu}$, respectively. The minimum of $g(z)$ on $z \in (0, \infty)$ is zero, iff $\tilde{\mu}$ solves $\tilde{\mu}^2 - 4(3\rho_1 - \tilde{\mu})(3\rho_2 - \tilde{\mu}) = 0$. This quadratic polynomial has a simple positive zero given by $\tilde{\mu}_0 = 2 - 2\sqrt{1 - 3\rho_2(1 - \rho_2)} > 0$, since $\rho_1 + \rho_2 = 1$.

The expression $\tilde{\mu}_0 = 2 - 2\sqrt{1 - 3\rho_2(1 - \rho_2)} > 0$ attains its maximum 1 for $\rho_2 \in (0, 1)$ at $\rho_2 = 1/2$. \square

Remark: The quadratic entropy $\psi_2(\sigma)$ satisfies Assumption (2.19) with $\mu = 2\lambda$ for all $f_1^\infty, f_2^\infty \in (0, 1)$. The cubic entropy $\psi_3(\sigma)$ and the quartic entropy $\psi_4(\sigma)$ satisfy (2.19) with $\mu = 2\lambda$ only for $f_1^\infty = f_2^\infty = \frac{1}{2}$.

2.2 Lyapunov's direct method

We consider an ODE for a vector $f(t) = (f_1(t), f_2(t), \dots, f_n(t))^\top \in \mathbb{R}^n$:

$$\begin{cases} \frac{d}{dt}f = \mathbf{A}f, & t \geq 0, \\ f(0) = f^I \in \mathbb{R}^n, \end{cases} \quad (2.23)$$

for some real matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$. The origin 0 is a steady state of (2.23). The stability of the trivial solution $f^0(t) \equiv 0$ is determined by the eigenvalues of matrix \mathbf{A} :

Theorem 4. *Let $\mathbf{A} \in \mathbb{R}^{n \times n}$ and let λ_j ($j = 1, \dots, n$) denote the eigenvalues of \mathbf{A} (counted with their multiplicity).*

- (S1) The equilibrium f^0 of (2.23) is stable if and only (i) $\Re\lambda_j \leq 0$ for all $j = 1, \dots, n$; and (ii) all eigenvalues with $\Re\lambda_j = 0$ are non-defective¹.
- (S2) The equilibrium f^0 of (2.23) is asymptotically stable if and only if $\Re\lambda_j < 0$ for all $j = 1, \dots, n$.
- (S3) The equilibrium f^0 of (2.23) is unstable in all other cases.

To study the stability for f^0 via Lyapunov's direct method, a first guess for a Lyapunov function $V(f)$ is the (squared) Euclidean norm $V(f) = \|f\|_2^2$. The derivative of $V(f)$ along solutions $f(t)$ of (2.23) satisfies

$$\frac{d}{dt}V(f(t)) = \langle f(t), (\mathbf{A}^\top + \mathbf{A})f(t) \rangle.$$

Thus the derivative depends only on the symmetric part $\frac{1}{2}(\mathbf{A}^\top + \mathbf{A})$ of a matrix \mathbf{A} . Hence the choice $V(f) = \|f\|_2^2$ is only suitable for symmetric matrices \mathbf{A} .

To study the stability of $f^0(t) \equiv 0$ w.r.t. (2.23) for a general \mathbf{A} , it is standard to consider the generalized (squared) norm

$$V(f) := \langle f, \mathbf{P}f \rangle \quad \text{for some symmetric, positive definite matrix } \mathbf{P} \in \mathbb{R}^{n \times n}.$$

The derivative of $V(f)$ along solutions $f(t)$ of (2.23) satisfies

$$\frac{d}{dt}V(f(t)) = \langle \mathbf{A}f(t), \mathbf{P}f(t) \rangle + \langle f(t), \mathbf{P}\mathbf{A}f(t) \rangle = \langle f(t), \mathbf{R}f(t) \rangle, \quad (2.24)$$

with matrix $\mathbf{R} := \mathbf{A}^\top \mathbf{P} + \mathbf{P}\mathbf{A}$. Conclusions on the stability of f^0 are possible, depending on the (negative) definiteness of \mathbf{R} , see e.g. [15, Proposition 7.6.1].

To determine the decay rate of an asymptotically stable steady state, we shall use the following algebraic result.

Lemma 2. *For any fixed matrix $\mathbf{C} \in \mathbb{C}^{n \times n}$, let $\mu := \min\{\Re\{\lambda\} \mid \lambda \text{ is an eigenvalue of } \mathbf{C}\}$. Let $\{\lambda_j \mid 1 \leq j \leq j_0\}$ be all the eigenvalues of \mathbf{C} with $\Re\{\lambda_j\} = \mu$, only counting their geometric multiplicity.*

If all λ_j ($j = 1, \dots, j_0$) are non-defective, then there exists a Hermitian, positive definite matrix $\mathbf{P} \in \mathbb{C}^{n \times n}$ with

$$\mathbf{C}^* \mathbf{P} + \mathbf{P} \mathbf{C} \geq 2\mu \mathbf{P}, \quad (2.25)$$

where \mathbf{C}^* denotes the Hermitian transpose of \mathbf{C} . Moreover, (non-unique) matrices \mathbf{P} satisfying (2.25) are given by

$$\mathbf{P} := \sum_{j=1}^n b_j w_j \otimes \bar{w}_j^\top, \quad (2.26)$$

where w_j ($j = 1, \dots, n$) denote the eigenvectors of \mathbf{C}^* , and $b_j \in \mathbb{R}^+$ ($j = 1, \dots, n$) are arbitrary weights.

¹ An eigenvalue is defective if its geometric multiplicity is strictly less than its algebraic multiplicity.

Remark: Lemma 2 is the complex analog of [3, Lemma 4.3] or [1, Lemma 2.6]. In particular, if $\mathbf{C} \in \mathbb{R}^{n \times n}$ is a real matrix, then the statement (2.25) of Lemma 2 holds true for real, symmetric, positive definite matrices $\mathbf{P} \in \mathbb{R}^{n \times n}$. Moreover, the case of defective eigenvalues is also treated in [3, 1].

If $\mathbf{A} \in \mathbb{R}^{n \times n}$ has only eigenvalues with negative real parts, then the origin is the unique and asymptotically stable steady state $f^0 = 0$ of (2.23). Due to Lemma 2, there exists a symmetric, positive definite matrix $\mathbf{P} \in \mathbb{R}^{n \times n}$ such that $\mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A} \leq -2\mu \mathbf{P}$ where $\mu = \min |\Re \lambda_j|$. Thus, the derivative of $V(f) := \langle f, \mathbf{P} f \rangle$ along solutions of (2.23) satisfies

$$\frac{d}{dt} V(f(t)) \leq -2\mu V(f(t)) \quad \text{with} \quad \mu = \min |\Re \lambda_j|, \quad (2.27)$$

which implies $V(f(t)) \leq e^{-2\mu t} V(f^I)$ and $\|f(t)\|^2 \leq c e^{-2\mu t} \|f^I\|^2$ for some $c > 0$ by equivalence of norms on \mathbb{R}^n .

In contrast, we consider next matrices $\mathbf{A} \in \mathbb{R}^{n \times n}$ having only eigenvalues with non-positive real part. More precisely, let \mathbf{A} satisfy

- (A1) \mathbf{A} has a simple eigenvalue $\lambda_1 = 0$ with left eigenvector $w_1^\top \in \mathbb{R}^n$ and right eigenvector $v_1 \in \mathbb{R}^n$;
- (A2) the other eigenvalues λ_j ($j = 2, \dots, n$) of \mathbf{A} have negative real part;

Then, the space of steady states of (2.23) consists of $\text{span}\{v_1\}$, and solutions to (2.23) will typically not decay to 0. More precisely, if f is a solution of ODE (2.23) with initial datum f^I satisfying $\langle w_1, f^I \rangle = c$ for some $c \in \mathbb{R}$, then $\langle w_1, f(t) \rangle = c$ for all $t \geq 0$. Therefore we aim to prove the convergence of solutions $f(t)$ of (2.23) for an initial datum f^I (normalized in the sense of $\langle w_1, f^I \rangle = 1$) to the unique steady state $f^\infty \in \text{span}\{v_1\}$ (again normalized as $\langle w_1, f^\infty \rangle = 1$).

Lemma 3. *Let $\mathbf{A} \in \mathbb{R}^{n \times n}$ satisfy (A1)–(A2) with non-defective eigenvalues λ_j for $j = 1, \dots, n$. If f is a solution of (2.23) for some normalized initial datum f^I (i.e. $\langle w_1, f^I \rangle = 1$), then*

$$\|f(t) - f^\infty\| \leq c \|f^I - f^\infty\| e^{-\lambda_* t}, \quad t \geq 0, \quad (2.28)$$

where $\lambda_* := \min_{\lambda_j \neq 0} |\Re \lambda_j|$ and some constant $c > 0$.

Proof. To present a unified approach for symmetric and non-symmetric matrices \mathbf{A} satisfying (A1)–(A2), we consider again the “distorted” vector norm $\|f\|_{\mathbf{P}} := \sqrt{\langle f, \mathbf{P} f \rangle}$, and the relative entropy-type functional

$$E_{\psi_2}(f(t)|f^\infty) := \|f(t) - f^\infty\|_{\mathbf{P}}^2$$

with some real, symmetric and positive definite matrix \mathbf{P} to be determined. Its derivative satisfies

$$\frac{d}{dt} E_{\psi_2}(f(t)|f^\infty) = \left\langle (f - f^\infty), (\mathbf{A}^\top \mathbf{P} + \mathbf{P} \mathbf{A})(f - f^\infty) \right\rangle.$$

Every matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ induces an orthogonal decomposition of \mathbb{R}^n via

$$\mathbb{R}^n = \ker(\mathbf{A}) \oplus \operatorname{ran}(\mathbf{A}^\top) = \ker(\mathbf{A}^\top) \oplus \operatorname{ran}(\mathbf{A}).$$

Thus, there exists an orthogonal projection from \mathbb{R}^n onto $\operatorname{ran}(\mathbf{A})$, which is represented by a matrix $\mathbf{P}_1 \in \mathbb{R}^{n \times n}$ with $\mathbf{P}_1^2 = \mathbf{P}_1$. Due to assumption (A1), matrix \mathbf{A}^\top has a one-dimensional kernel which is spanned by w_1 , hence, $\mathbf{P}_1 w_1 = 0$. Since w_1^\top is a left eigenvector of \mathbf{A} for the eigenvalue 0, a solution f of (2.23) for a normalized initial datum f^I (i.e. $\langle w_1, f^I \rangle = 1$) is again normalized, i.e. $\langle w_1, f(t) \rangle = 1$ for all $t \geq 0$. Thus, $\langle w_1, f(t) - f^\infty \rangle \equiv 0$ iff $\langle w_1, f^I - f^\infty \rangle = 0$, which implies $f(t) - f^\infty \in \operatorname{ran}(\mathbf{A})$ for all $t \geq 0$. Moreover,

$$\frac{d}{dt} E_{\psi_2}(f(t)|f^\infty) = \left\langle \mathbf{P}_1(f - f^\infty), \mathbf{P}_1^\top (\mathbf{A}^\top \mathbf{P} + \mathbf{P}\mathbf{A}) \mathbf{P}_1 \mathbf{P}_1(f - f^\infty) \right\rangle.$$

In order to prove

$$\mathbf{P}_1^\top (\mathbf{A}^\top \mathbf{P} + \mathbf{P}\mathbf{A}) \mathbf{P}_1 \leq -2\lambda_* \mathbf{P}_1^\top \mathbf{P} \mathbf{P}_1 \quad (2.29)$$

we consider the modified matrix $\tilde{\mathbf{A}} := \mathbf{A} - \lambda_* v_1 \otimes w_1^\top \in \mathbb{R}^{n \times n}$. Due to (A1)–(A2) and the assumptions in Lemma 3, $\tilde{\mathbf{A}}$ has only non-defective eigenvalues with negative real part. Due to Lemma 2, there exists a real, symmetric, positive-definite matrix \mathbf{P} such that $\tilde{\mathbf{A}}^\top \mathbf{P} + \mathbf{P}\tilde{\mathbf{A}} \leq -2\lambda_* \mathbf{P}$. This implies (2.29) since $\mathbf{P}_1^\top ((v_1 \otimes w_1^\top)^\top \mathbf{P} + \mathbf{P}(v_1 \otimes w_1^\top)) \mathbf{P}_1 = 0$. Therefore we conclude

$$\frac{d}{dt} E_{\psi_2}(f(t)|f^\infty) = -2\lambda_* E_{\psi_2}(f(t)|f^\infty), \quad (2.30)$$

and $E_{\psi_2}(f(t)|f^\infty) \leq E_{\psi_2}(f^I|f^\infty) e^{-2\lambda_* t}$ follows. Moreover, $0 \leq \lambda_{P,\min} \mathbf{I} \leq \mathbf{P} \leq \lambda_{P,\max} \mathbf{I}$, where $\lambda_{P,\min} > 0$ is the smallest eigenvalue and $\lambda_{P,\max} > 0$ is the biggest eigenvalue of \mathbf{P} . Therefore, $\lambda_{P,\min} \|f\|_2^2 \leq \|f\|_{\mathbf{P}}^2 \leq \lambda_{P,\max} \|f\|_2^2$ and (2.28) follows. \square

Remark: For a symmetric matrix \mathbf{A} , the choice $\mathbf{P} = \mathbf{I}$ is admissible and one recovers the optimal decay rate and constant $c = 1$ in estimate (2.28).

3 A discrete hypocoercive BGK model

In this section we consider an example for a discrete version (both in x and v) of (1.1). More precisely, we consider the evolution of a vector $f(t) = (f_j(t); j = 1, \dots, 4)^\top \in \mathbb{R}^4$, where its four components may correspond to the following points in the $x-v$ -phase space: $(1, 1)$, $(1, -1)$, $(-1, -1)$, $(-1, 1)$, in this order. Its evolution is given by

$$\begin{aligned} \frac{d}{dt} f &= (\mathbf{A} + \mathbf{B})f, \quad t \geq 0, \\ f(0) &= f^I \in \mathbb{R}^4. \end{aligned} \quad (3.1)$$

Similarly to (2.6), the matrix \mathbf{A} has BGK form:

$$\mathbf{A} := \frac{1}{2} \begin{pmatrix} -1 & 1 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 1 & -1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} - \mathbf{I}, \quad (3.2)$$

where the first summand on the r.h.s. is the projection onto the kernel of \mathbf{A} ,

$$\ker \mathbf{A} = \text{span}[(1100)^\top, (0011)^\top].$$

In (3.1), the matrix \mathbf{B} is skew-symmetric and reads

$$\mathbf{B} := \begin{pmatrix} 0 & -1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \\ -1 & 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}. \quad (3.3)$$

\mathbf{B} corresponds to a discretization of the transport operator in (1.1) by symmetric finite differences. We remark that (3.1) does not preserve positivity but, as we shall show, the hypocoercivity of (1.1). Motivated by the theory of hyperbolic systems, one may also replace the transport operator by an upwind discretization with a then non-symmetric matrix $\tilde{\mathbf{B}}$. Then, the resulting system would preserve positivity. But it would be coercive rather than hypocoercive. Here we opt to discuss the situation with \mathbf{B} given in (3.3).

The spectrum of $\mathbf{A} + \mathbf{B}$ is given by $0, -\frac{1}{2} \pm \frac{\sqrt{15}}{2}i, -1$. The unique, (in the 1-norm) normalized steady state of (3.1) is given by $f^\infty = w_1 = \frac{1}{4}(1111)^\top$, which spans the kernel of $\mathbf{A} + \mathbf{B}$. Eigenvectors of the non-trivial eigenvalues are given by $w_{2,3} := (\sqrt{5}, \pm\sqrt{3}i, -\sqrt{5}, \mp\sqrt{3}i)^\top$ and $w_4 := (1, -1, 1, -1)^\top$, and all three of them have mass 0. This shows that $\frac{1}{2}$ is the sharp decay rate of any (normalized) $f(t)$ towards f^∞ . But this ‘‘spectral gap’’ of size $\frac{1}{2}$ disappears in the symmetric part of the matrix: $\sigma(\mathbf{A} + \frac{\mathbf{B} + \mathbf{B}^\top}{2}) = \{0, 0, -1, -1\}$. Hence, the matrix $\mathbf{A} + \mathbf{B}$ is only *hypocoercive* on $\{f^\infty\}^\perp$ (as defined by C. Villani, see §3.2 of [20]). But using an appropriate similarity transformation of $\mathbf{A} + \mathbf{B}$ one can again recover the sharp decay rate of the hypocoercive BGK-model (3.1) via energy or entropy methods.

Due to Lemma 3, if f^I is normalized, then the solution to (3.1) satisfies (for any norm on \mathbb{R}^4)

$$\|f(t) - f^\infty\| \leq c e^{-t/2} \|f^I - f^\infty\|, \quad t \geq 0,$$

with some generic constant $c > 0$.

4 Space-inhomogeneous BGK models

In this section we study the large-time behavior of the BGK equation (1.3) on $L^2(\mathbf{T} \times \mathbb{R}; M_T^{-1})$ with periodic boundary conditions in x . We start with the x -Fourier series of f :

$$f(x, v, t) = \sum_{k \in \mathbb{Z}} f_k(v, t) e^{ikx},$$

and obtain the following evolution equation for the spatial modes f_k , $k \in \mathbb{Z}$:

$$\partial_t f_k + ikv f_k = \mathbf{Q} f_k = M_T(v) \int_{\mathbb{R}} f_k(v, t) dv - f_k(v, t), \quad k \in \mathbb{Z}; t \geq 0. \quad (4.1)$$

Since the BGK operator \mathbf{Q} projects onto the centered Maxwellian at temperature T , it is natural to consider (4.1) in the basis spanned by the Hermite functions (in v). This is natural for the following reason:

The Hermite polynomials (for temperature T) are the system of orthonormal polynomials that one obtains by applying the Gram-Schmidt orthonormalization procedure to the sequence of monomials $\{v^\ell\}$ in $L^2(M_T)$; let $P_\ell(v)$ denote the ℓ th Hermite polynomial. The *Hermite functions* themselves are the functions of the form $H_\ell(v) = P_\ell(v)M_T(v)$, and evidently these are orthonormal in $L^2(M_T^{-1})$. This is the space in which we work.

The key fact concerning the Hermite functions is that multiplication by v acts on them in a very simple way, and this is relevant since the action of our streaming operator on the k th mode is multiplication by ikv . In fact, the reason for the simple nature of its action is very general and thus applies to generalizations of the Hermite functions. Since we use this below, we explain the simple action from a general point of view, using only the fact that M_T is even.

Note that multiplication by v is evidently self adjoint on $L^2(M_T^{-1})$. Also, for each ℓ , $vH_\ell(v)$ is in the span of $\{H_0, \dots, H_{\ell+1}\}$. Hence, for $m > \ell + 1$

$$0 = \langle H_m, vH_\ell \rangle_{L^2(M_T^{-1})} = \langle H_\ell, vH_m \rangle_{L^2(M_T^{-1})}$$

from which we conclude that the ℓ, m matrix elements of multiplication by v are zero for $|\ell - m| > 2$. Finally, by the symmetry of M_T , the diagonal matrix elements are all zero. Hence, *in the Hermite basis, multiplication by v is represented by a tridiagonal symmetric matrix that is zero on the main diagonal*. The operator \mathbf{Q} is evidently diagonal in the Hermite basis. Hence the operator $\mathbf{L}_k := -ikv + \mathbf{Q}$ has a simple tridiagonal structure. We shall see that the matrix representing ikv is

$$ik\sqrt{T} \begin{pmatrix} 0 & \sqrt{1} & 0 & \cdots \\ \sqrt{1} & 0 & \sqrt{2} & 0 \\ 0 & \sqrt{2} & 0 & \sqrt{3} \\ \vdots & 0 & \sqrt{3} & \ddots \end{pmatrix}$$

while $\mathbf{Q} = \text{diag}(0, -1, -1, \dots)$.

The infinite tridiagonal matrix representing $\mathbf{L}_k = -ikv + \mathbf{Q}$ in the Hermite basis is still not easy to analyze directly. We cannot compute its eigenfunctions in closed form, and hence cannot apply formula (2.26) to implement Lyapunov's method.

However, we can do this for a related family of discrete velocity models, since then we are dealing with finite matrices. The discrete models, using the binomial approximation to the Gaussian distribution, are sufficiently close in structure to the continuous velocity BGK model that they suggest an ansatz for the \mathbf{P} operator that specifies the entropy function norm. In fact, a complete solution of a 2-velocity model provides the essential hint for proving hypocoercivity of the continuous velocity BGK model.

We shall present the details of this analysis in §4.3 below. Here, the above remark only serves as a motivation for our analysis of discrete velocity models, which are velocity discretizations of the BGK equation (1.3). We shall start with the two velocity case, and then discuss its generalization to n velocities.

4.1 A two velocity BGK model

In this section we revisit the following hyperbolic system, which can be considered as a kinetic equation with the two velocities $v = \pm\sigma$, and some parameter $\sigma > 0$:

$$\partial_t f_{\pm} \pm \sigma \partial_x f_{\pm} = \pm \frac{1}{2} (f_- - f_+), \quad t \geq 0 \quad (4.2)$$

for the distributions $f_{\pm}(x, t)$ of right- and left-moving particles, 2π -periodic in x . The matrix of the interaction term on the r.h.s. has the form

$$\frac{1}{2} \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix},$$

and hence (4.2) is also of BGK-form. Due to the conservation of the total mass $\int_0^{2\pi} (f_+(x, t) + f_-(x, t)) dx$ of (4.2), its unique steady state is $f_+^{\infty} = f_-^{\infty} = \text{const} = \frac{1}{4\pi} \int_0^{2\pi} (f_+^I(x) + f_-^I(x)) dx$.

This toy model (with the choice $\sigma = 1$) was analyzed in §1.4 of [12] to illustrate the hypocoercivity method presented there. As for (4.1), we Fourier transform (4.2) in x and expand it in the discrete velocity basis $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}$. This yields for each mode $k \in \mathbb{Z}$ the following decoupled ODE-system:

$$\frac{d}{dt} u_k = -\mathbf{C}_k u_k, \quad \mathbf{C}_k = \begin{pmatrix} 0 & ik\sigma \\ ik\sigma & 1 \end{pmatrix}, \quad (4.3)$$

with $u_k(t) \in \mathbb{C}^2$, $k \in \mathbb{Z}$. The matrices $-\mathbf{C}_k$ have the eigenvalues $-\frac{1}{2} \pm \sqrt{\frac{1}{4} - k^2 \sigma^2}$ in the case $|k| \leq \frac{1}{2\sigma}$ and $-\frac{1}{2} \pm i\sqrt{k^2 \sigma^2 - \frac{1}{4}}$ in the case $|k| > \frac{1}{2\sigma}$. Hence, as $t \rightarrow \infty$, $u_0(t)$ converges to an eigenvector of the 0-eigenvalue, i.e. $u_0^{\infty} = (f_+^{\infty} + f_-^{\infty}, 0)^{\top}$, with

the exponential rate $\lambda_0 := 1$. All modes $u_k(t)$ with $k \neq 0$ converge to $u_k^\infty = 0$ with an exponential rate determined by the spectral gap of the matrix \mathbf{C}_k . For simplicity we shall assume here that $\frac{1}{2\sigma} \notin \mathbb{N}$. This avoids defective eigenvalues of the matrices \mathbf{C}_k , but they could be included as discussed in Lemma 4.3 of [3]. The spectral gap of the low modes (i.e. for $0 < |k| < \frac{1}{2\sigma}$) is $\lambda_k := \frac{1}{2} - \sqrt{\frac{1}{4} - k^2\sigma^2}$, and it is $\lambda_k := \frac{1}{2}$ for the high modes. Hence, the exponential decay rate of the sequence of modes $\{u_k(t)\}_{k \in \mathbb{Z}}$ is given by the decay of the modes $k = \pm 1$: $\lambda := \min_{k \in \mathbb{Z}} \{\lambda_k\} = \Re(\frac{1}{2} - \sqrt{\frac{1}{4} - \sigma^2})$. By Plancherel's theorem this is then also the convergence rate of $f(t) = (f_+(t), f_-(t))^\top$ towards the steady state $f^\infty = (f_+^\infty, f_-^\infty)^\top$.

The goal of entropy methods is to prove this exponential decay towards equilibrium, possibly with the sharp rate, by constructing an appropriate Lyapunov functional. In the hypocoercive method developed in [12] the authors obtained, for the case $\sigma = 1$ and the quadratic entropy, a decay rate bounded above by $\frac{1}{5}$. But the sharp rate for this case is $\lambda = \frac{1}{2}$. We shall now construct a refined Lyapunov functional that captures the sharp decay rate.

Following Lemma 2(i) we introduce the positive definite transformation matrices $\mathbf{P}_0 := \mathbf{I}$,

$$\mathbf{P}_k := \begin{pmatrix} 4k^2\sigma^2 & -2ik\sigma \\ 2ik\sigma & 2 - 4k^2\sigma^2 \end{pmatrix}, \quad \text{for } 0 < |k| < \frac{1}{2\sigma},$$

and

$$\mathbf{P}_k := \begin{pmatrix} 1 & \frac{-i}{2k\sigma} \\ \frac{i}{2k\sigma} & 1 \end{pmatrix}, \quad \text{for } |k| > \frac{1}{2\sigma}. \quad (4.4)$$

In the latter case, \mathbf{P}_k is unique only up to a multiplicative constant, which is chosen here such that $\text{Tr} \mathbf{P}_k = n = 2$. We define the ‘‘distorted’’ vector norms for each mode u_k :

$$\|u_k\|_{\mathbf{P}_k} := \sqrt{\langle u_k, \mathbf{P}_k u_k \rangle}.$$

Due to the ODE (4.3) and the matrix inequality (2.25) it satisfies

$$\frac{d}{dt} \|u_k\|_{\mathbf{P}_k}^2 = -\bar{u}_k^\top (\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k) u_k \leq -2\lambda_k \|u_k\|_{\mathbf{P}_k}^2, \quad k \in \mathbb{Z} \setminus \{0\}, \quad (4.5)$$

and hence

$$\|u_k(t) - u_k^\infty\|_{\mathbf{P}_k} \leq e^{-\lambda_k t} \|u_k(0) - u_k^\infty\|_{\mathbf{P}_k}, \quad t \geq 0, \quad k \in \mathbb{Z}. \quad (4.6)$$

With this motivation we define the following norm as a Lyapunov functional for the sequence of modes:

$$E(\{u_k\}_{k \in \mathbb{Z}}) := \sqrt{\sum_{k \in \mathbb{Z}} \|u_k\|_{\mathbf{P}_k}^2}. \quad (4.7)$$

From (4.6) we obtain

$$E(\{u_k(t) - u_k^\infty\}) \leq e^{-\lambda t} E(\{u_k(0) - u_k^\infty\}), \quad t \geq 0,$$

with $\lambda = \min_{k \in \mathbb{Z}} \{\lambda_k\}$. Due to Plancherel's theorem, this is also a norm for the corresponding distributions $f = (f_+, f_-)^\top$:

$$E(\{u_k\}) = \|Bf\|_{L^2(0, 2\pi; \mathbb{R}^2)},$$

where B is a (nonlocal) bounded operator on $L^2(0, 2\pi; \mathbb{R}^2)$ with bounded inverse. More precisely, $B = I + K$, where K is a compact operator with $\|K\| < 1$, since $\mathbf{P}_k \xrightarrow{|k| \rightarrow \infty} \mathbf{I}$ (cf. (4.4)). This implies the sought-for exponential decay of $f(t)$ with sharp rate:

Theorem 5. *Let $\frac{1}{2\sigma} \notin \mathbb{N}$. Then the solution to (4.2) satisfies*

$$\|f(t) - f^\infty\|_{L^2(0, 2\pi; \mathbb{R}^2)} \leq c e^{-\lambda t} \|f^I - f^\infty\|_{L^2(0, 2\pi; \mathbb{R}^2)}, \quad t \geq 0,$$

with $\lambda = \Re\left(\frac{1}{2} - \sqrt{\frac{1}{4} - \sigma^2}\right)$ and some generic constant $c > 0$.

4.2 A multi-velocity BGK model

We now turn to a discrete velocity model analog of the linear BGK equation (1.3), and we shall establish its hypocoercivity. Fixing unit temperature T , recall that as a consequence of the Central Limit Theorem, the measure $M_1(v)dv$ is the (weak) limit of a sequence of discrete probability measures $\{\mu_n\}$ where

$$\mu_n := \sum_{j=0}^n 2^{-n} \binom{n}{j} \delta_{(2j-n)/\sqrt{n}},$$

where δ_y denotes the unit mass at $y \in \mathbb{R}$. Each of the probability measures μ_n , $n \in \mathbb{N}$, has zero mean and unit variance.

The Hermite polynomials have a natural discrete analog, namely the *Krawtchouk polynomials*. A good reference containing proofs of all of the facts we use below is the survey [10]. (We are only concerned with a special family of the more general Krawtchouk polynomials discussed in [10], namely the $s = 2$ case in the terminology used there.) The standard Krawtchouk polynomials of order m are a set of $n + 1$ polynomials $K_{n,m}$; $m = 0, \dots, n$ that are orthogonal with respect to the probability measure

$$\omega_n = \sum_{j=0}^n 2^{-n} \binom{n}{j} \delta_j,$$

and are given by the following generating function:

$$(1+t)^{n-v} (1-t)^v = \sum_{m=0}^n t^m K_{n,m}(v). \quad (4.8)$$

The leading coefficient of $K_{n,m}$ has the sign $(-1)^m$. One has the orthogonality relations

$$\int_{\mathbb{R}} K_{n,m} K_{n,\ell} d\omega_n = \begin{cases} \binom{n}{m} & m = \ell, \\ 0 & m \neq \ell. \end{cases} \quad (4.9)$$

Then the *discrete Hermite polynomials* $H_{n,m}$ are defined by

$$H_{n,m}(v) := (-1)^m \binom{n}{m}^{-1/2} K_{n,m} \left(\frac{n}{2} + \frac{\sqrt{n}}{2} v \right) \quad \text{for } m = 0, 1, \dots, n; v \in \mathbb{R}. \quad (4.10)$$

Then $\{H_{n,0}, \dots, H_{n,n}\}$ is the set of $n+1$ polynomials that are orthogonal with respect to μ_n , and hence are an orthonormal basis for $L^2(\mathbb{R}; \mu_n)$, and for each m and v , $\lim_{n \rightarrow \infty} H_{n,m}(v) = \frac{1}{\sqrt{m!}} H_m(v)$. The analog of the crucial Hermite–recurrence relation (4.13) for the Krawtchouk polynomials is

$$(m+1)K_{n,m+1} = (n-2v)K_{n,m} - (n-m+1)K_{n,m-1}.$$

Rewriting this in terms of the discrete Hermite polynomials, one obtains

$$vH_{n,m}(v) = \sqrt{m+1} \binom{n-m}{n}^{1/2} H_{n,m+1}(v) + \sqrt{m} \binom{n-m+1}{n}^{1/2} H_{n,m-1}(v). \quad (4.11)$$

Notice that this reduces to (4.13) in the limit $n \rightarrow \infty$ (up to the multiplication by the standard Gaussian).

We are now ready to produce a discrete velocity analog of (1.3) in continuous x -space. The phase space is $\mathbf{T} \times [v_0, \dots, v_n]$ where the discrete velocity $v_j = (2j-n)/\sqrt{n}$. Our phase spaces density at time t is a vector $\mathbf{f}(x, t)$ with $n+1$ non-negative entries $f_0(x, t), \dots, f_n(x, t)$, such that

$$\sum_{j=0}^n \left(\int_{\mathbf{T}} f_j(x, t) dx \right) = 1.$$

We associate to $\mathbf{f}(x, t)$ the probability measure on the phase space given by

$$\sum_{j=0}^n f_j(x, t) \delta_{(2j-n)/\sqrt{n}}.$$

The discrete unit Maxwellian (of order n) is the vector $\mathbf{m} = 2^{-n} \left(\binom{n}{0}, \binom{n}{1}, \dots, \binom{n}{n} \right)^\top$. Then the order n discrete analog of (1.3) is the equation

$$\partial_t \mathbf{f}(x, t) + \mathbf{V} \partial_x \mathbf{f}(x, t) = \mathbf{m} \left(\sum_{j=0}^n f_j(x, t) \right) - \mathbf{f}(x, t), \quad t \geq 0; x \in \mathbf{T}, \quad (4.12)$$

with the $(n+1) \times (n+1)$ matrix $\mathbf{V} = \text{diag}(v_0, \dots, v_n)$. Proceeding as for (4.1) yields the evolution equation for the spatial modes $\mathbf{f}_k(t)$, $k \in \mathbb{Z}$. Expanding \mathbf{f}_k in the discrete

Hermite basis $\{H_{n,m}(v_j); m, j = 0, \dots, n\}$, we obtain for each k the equation

$$\partial_t \hat{\mathbf{f}}_k + ik\mathbf{L}_1 \hat{\mathbf{f}}_k = \mathbf{L}_2 \hat{\mathbf{f}}_k, \quad t \geq 0; k \in \mathbb{Z},$$

where the vector $\hat{\mathbf{f}}_k(t) \in \mathbb{C}^{n+1}$ represents the basis coefficients of $\mathbf{f}_k(t)$. As before \mathbf{L}_2 is the $(n+1) \times (n+1)$ matrix $\mathbf{L}_2 = \text{diag}(0, -1, -1, \dots)$, and \mathbf{L}_1 is the symmetric tridiagonal matrix whose diagonal entries are all zero, and whose superdiagonal sequence is given by

$$[\mathbf{L}_1]_{m,m+1} = \sqrt{m+1} \left(\frac{n-m}{n} \right)^{1/2}; \quad m = 0, 1, \dots, n-1.$$

For example, with $n = 4$,

$$\mathbf{L}_1 = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & \sqrt{3/2} & 0 & 0 \\ 0 & \sqrt{3/2} & 0 & \sqrt{3/2} & 0 \\ 0 & 0 & \sqrt{3/2} & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix}.$$

Next we discuss the time decay of the solution to (4.12) towards $\mathbf{f}^\infty = \mathbf{m}$. We shall focus on the example with order $n = 4$, but the other cases behave similarly. Computing for the modes $k = \pm 1$ the eigenvalues of $\mp i\mathbf{L}_1 + \mathbf{L}_2$ we find two complex pairs and one real eigenvalue $\lambda_0 = -0.526948302245121\dots$ which has the least negative real part, and hence determines the exponential decay rate of $\mathbf{f}_{\pm 1}(t)$. This situation for higher $|k|$ is similar, but even better, with faster decay. To see this we write the eigenvalue equation for the matrices $-ik\mathbf{L}_1 + \mathbf{L}_2, k \in \mathbb{Z}$ as

$$h_0(\lambda) := \lambda(\lambda+1)^4 = -k^2(\lambda+1)^2(5\lambda+1) - k^4(4\lambda + \frac{5}{2}) =: -k^2 h_2(\lambda) - k^4 h_4(\lambda).$$

The function h_0 is negative on $(-1, 0)$, $-h_2$ on $(-\frac{1}{5}, 0]$ and $-h_4$ on $(-\frac{5}{8}, 0]$ (cf. Figure 1). For $k \neq 0$, the function $k^2 h_2(\lambda) + k^4 h_4(\lambda)$ has exactly one real zero, $\tilde{\lambda}(k)$, and it is nonnegative on $[\tilde{\lambda}(k), 0]$. For each fixed $k \in \mathbb{Z}$, the function $k^2 h_2 + k^4 h_4$ is strictly increasing w.r.t. λ . Hence, the above eigenvalue equation has exactly one real zero $\lambda_0(k)$, and it lies in $(-\frac{5}{8}, 0]$. For each fixed $\lambda \in [\tilde{\lambda}(k), 0]$, the function $k^2 h_2 + k^4 h_4$ is strictly increasing w.r.t. increasing $|k|$. Hence, $\lambda_0(k)$ decreases monotonically (w.r.t. $|k|$) towards $-\frac{5}{8}$.

This proves that the 5 velocity model is hypocoercive, at least in the norm E defined in (4.7) (with the transformation matrices \mathbf{P}_k now corresponding to $-\mathbf{C}_k := \mp ik\mathbf{L}_1 + \mathbf{L}_2$). The sharp decay rate is given by $\lambda_0 = -0.526948302245121\dots$.

To establish a uniform-in- k spectral gap was already cumbersome for the case $n = 4$, and it becomes even more involved for larger n . In the following section we present a much simpler strategy, at the price of giving up sharpness of the decay rate.

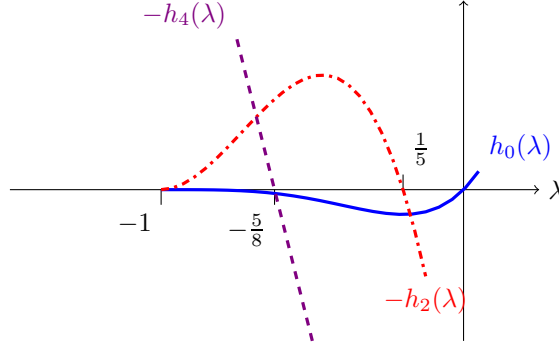


Fig. 1 Functions appearing in the eigenvalue equation of $-ik\mathbf{L}_1 + \mathbf{L}_2$; solid blue curve: $h_0(\lambda)$; red dash-dotted curve: $-h_2(\lambda)$; purple dashed line: $-h_4(\lambda)$. (colors only online)

4.3 A continuous velocity BGK model

Using the probabilists' Hermite polynomials,

$$H_m(v) := (-1)^m e^{\frac{v^2}{2}} \frac{d^m}{dv^m} e^{-\frac{v^2}{2}}, \quad m \in \mathbb{N}_0,$$

we define the normalized Hermite functions

$$g_m(v) := (2\pi m!)^{-1/2} H_m(v) e^{-\frac{v^2}{2}}, \quad \text{and} \quad \tilde{g}_m(v) := \frac{1}{\sqrt{T}} g_m\left(\frac{v}{\sqrt{T}}\right).$$

They satisfy

$$\int_{\mathbb{R}} \tilde{g}_m(v) \tilde{g}_n(v) M_T^{-1}(v) dv = \delta_{mn}$$

and the recurrence relation

$$v \tilde{g}_m(v) = \sqrt{T} [\sqrt{m+1} \tilde{g}_{m+1}(v) + \sqrt{m} \tilde{g}_{m-1}(v)]. \quad (4.13)$$

In the basis $\{\tilde{g}_m\}_{m \in \mathbb{N}_0}$ Equation (4.1) becomes

$$\partial_t \hat{\mathbf{f}}_k + ik\sqrt{T} \mathbf{L}_1 \hat{\mathbf{f}}_k = \mathbf{L}_2 \hat{\mathbf{f}}_k, \quad t \geq 0; k \in \mathbb{Z}. \quad (4.14)$$

Here, the “infinite vector” $\hat{\mathbf{f}}_k(t) \in l^2(\mathbb{N}_0)$ is the representation of the function $f_k(v, t) \in L^2(\mathbb{R}; M_T^{-1})$ in the Hermite function basis, and the operators $\mathbf{L}_1, \mathbf{L}_2$ are represented by “infinite matrices” as

$$\mathbf{L}_1 = \begin{pmatrix} 0 & \sqrt{1} & 0 & \cdots \\ \sqrt{1} & 0 & \sqrt{2} & 0 \\ 0 & \sqrt{2} & 0 & \sqrt{3} \\ \vdots & 0 & \sqrt{3} & \ddots \end{pmatrix}, \quad \mathbf{L}_2 = \text{diag}(0, -1, -1, \cdots). \quad (4.15)$$

Next we shall prove the exponential decay of (4.14), using a modified strategy compared to §4.2. For the 5 velocity model there, it was possible (with some effort) to determine the sharp spectral gap of the matrices $-ik\mathbf{L}_1 + \mathbf{L}_2$, uniform in all modes k . But since this seems not (easily) possible for the infinite dimensional case in (4.14), we shall construct now approximate transformation matrices \mathbf{P}_k that yield at least a (reasonable) lower bound on the spectral gap, and hence on the decay rate. For simplicity we set now $T = 1$, as the temperature could be “absorbed” into the parameter k by scaling.

Let \mathbf{A} be an $(n+1) \times (n+1)$ tridiagonal matrix that is zero on the main diagonal. That is $A_{i,j} = 0$ unless $j = i+1$ or $i = j-1$. We further suppose that \mathbf{A} is real and symmetric, so that \mathbf{A} is characterized by the numbers a_1, \dots, a_n where $a_j = A_{j-1,j}$. Let $\mathbf{B} = \text{diag}(0, -1, \dots, -1)$. Finally, for $k \in \mathbb{Z}$, consider the matrix $-\mathbf{C}_k := -ik\mathbf{A} + \mathbf{B}$.

In the simplest case $n = 1$ with $a = 1$, we obtain

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} 0 & 0 \\ 0 & -1 \end{pmatrix} \quad \text{and} \quad -\mathbf{C}_k = \begin{pmatrix} 0 & -ik \\ -ik & -1 \end{pmatrix}.$$

For this matrix \mathbf{C}_k , the transformation matrix \mathbf{P}_k was already computed in (4.4) (with $\sigma = 1$). For $k \neq 0$ a simple computation yields

$$\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k = \mathbf{P}_k$$

so that with this choice of \mathbf{P}_k , Lyapunov’s method yields exponential decay of the ODE-sequence $\frac{d}{dt} u_k = -\mathbf{C}_k u_k$, $k \in \mathbb{Z}$ at the optimal rate $e^{-t/2}$ (cf. §4.1).

We now turn to $n > 1$. For $k \neq 0$ define \mathbf{P}_k to be the $(n+1) \times (n+1)$ matrix whose upper left 2×2 block is $\begin{pmatrix} 1 & -i\alpha/k \\ i\alpha/k & 1 \end{pmatrix}$, where $0 < \alpha < k$ is a parameter to be chosen below, and with the remaining diagonal entries being 1, and all other entries being 0. Then the eigenvalues of \mathbf{P}_k are $(k + \alpha)/k$, 1 and $(k - \alpha)/k$, so that \mathbf{P}_k is positive definite, and close to the identity for large k .

We take $-\mathbf{C}_k := -ik\mathbf{A} + \mathbf{B}$ as above. Then

$$\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k = -ik(\mathbf{A}\mathbf{P}_k - \mathbf{P}_k\mathbf{A}) - (\mathbf{B}\mathbf{P}_k + \mathbf{P}_k\mathbf{B}),$$

and its upper left 3×3 block reads

$$\begin{pmatrix} 2a_1\alpha & -i\alpha/k & a_2\alpha \\ i\alpha/k & 2 - 2a_1\alpha & 0 \\ a_2\alpha & 0 & 2 \end{pmatrix}.$$

The lower right $(n-2) \times (n-2)$ block is -2 times the identity, the off diagonal blocks are zero. In all of our finite dimensional approximations to (4.15) we have $a_1 = 1$. The value of a_2 is different for the different discrete velocity models, but to simplify matters, we only present calculations for $a_2 = \sqrt{2}$, which is the value for the limiting continuous velocity model.

The determinants of the upper left 2×2 and 3×3 blocks read, respectively,

$$\delta_2(\alpha, k) = \alpha \left(4 - \left(4 + \frac{1}{k^2} \right) \alpha \right) \quad \text{and} \quad \delta_3(\alpha, k) = 4\alpha \left((\alpha - 2)(\alpha - 1) - \frac{\alpha}{2k^2} \right).$$

For each k , $\delta_3(\alpha, k)/\alpha$ has two positive roots, and is negative between them. Hence our matrix is positive definite when α lies between zero and the smaller positive root of $\delta_3(\alpha, k)/\alpha$. This root is least when $k = 1$, when it has the value $\frac{7-\sqrt{17}}{4} \approx 0.719$. Hence, by Sylvester's criterion, $\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k$ is positive definite for all $k \neq 0$ if and only if $\alpha \in (0, \frac{7-\sqrt{17}}{4}) \approx (0, 0.719)$. Note that also $\delta_2(\alpha, k) > 0$ for these α, k .

When α is in this range, our 3×3 matrix has three positive eigenvalues $\lambda_1, \lambda_2, \lambda_3$ which we may take to be arranged in increasing order. Then

$$\sqrt{\lambda_1} = \frac{\sqrt{\delta_3(\alpha, k)}}{\sqrt{\lambda_2 \lambda_3}} \geq \frac{2\sqrt{\delta_3(\alpha, k)}}{\lambda_2 + \lambda_3} > \frac{\sqrt{\delta_3(\alpha, k)}}{2}$$

since the trace of our matrix is 4. Hence, the least eigenvalue λ_1 of our 3×3 matrix satisfies

$$\lambda_1 = \lambda_1(\alpha, k) \geq \frac{1}{4} \delta_3(\alpha, k).$$

Hence we choose $\alpha = \alpha_k$ to maximize $\delta_3(\alpha, k)$ between its first two roots. Its maximal value, $\delta_3(\alpha_k, k)$, depends on k , but it is easily seen to be least for $k = 1$ with $\alpha_1 = \frac{1}{3}$. Simple computations and estimates then yield $\lambda_1(\alpha_k, k) \geq \frac{1}{4} \delta_3(\alpha_1, 1) = 17/54$ for all k .

Since we always take $\alpha < 1$, the largest eigenvalue of the matrix \mathbf{P}_k (defined with $\alpha = \alpha_k$) is no more than 2, uniformly in k . Hence

$$\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k \geq \frac{17}{54} \mathbf{I} \geq \frac{17}{108} \mathbf{P}_k \tag{4.16}$$

uniformly in k . Thus in each Fourier mode, we at least have exponential decay (of a quadratic type entropy) at the rate $17/108$ (by proceeding as in (4.5)).

Since this is also uniform in n , we obtain a bound for the continuous velocity model. Let the infinite matrix \mathbf{P}_k be the positive matrix using the optimal value of α in the k th mode, and regarded as a bounded operator on $L^2(M_T^{-1})$ through its action on Hermite modes. Define the entropy function by

$$e(f) := \sum_{k \in \mathbb{Z}} \langle (f_k(v) - M_1(v)), \mathbf{P}_k (f_k(v) - M_1(v)) \rangle_{L^2(M_1^{-1})}.$$

We obtain that, for solutions $f(t)$ of our BGK equation (1.3) or, equivalently, (4.1),

$$\frac{d}{dt}e(f(t)) \leq -\frac{17}{108}e(f(t)),$$

giving exponential relaxation.

The least eigenvalue of \mathbf{P}_k , $1 - \alpha/k$, is at least $1 - \frac{7-\sqrt{17}}{4} > \frac{1}{4}$ uniformly in k , and hence we have the inequality

$$e(f) \geq \frac{1}{4} \|f - M_1\|_{L^2(M_1^{-1})}^2.$$

The above method to establish exponential decay is simple to apply but does not give the sharp decay rate (it is off by a factor of about 9). Hence we shall now sketch how to improve on it. The essence of the above method is to use an ansatz for the transformation matrix \mathbf{P}_k , namely to use for its upper left 2×2 block the matrix from the 2 velocity case. Using instead larger blocks, will most likely improve the decay rate.

As a second alternative we shall now present an improvement of the crucial matrix inequality (4.16): In the inequality

$$\mathbf{C}_k^* \mathbf{P}_k + \mathbf{P}_k \mathbf{C}_k - 2\mu \mathbf{P}_k \geq \mathbf{0} \quad (4.17)$$

we shall choose $\mu \in [0, 1]$ as large as possible (related to the matrix inequality (2.25)). The upper left 3×3 block of the l.h.s. of this matrix reads

$$\mathbf{D} := \begin{pmatrix} 2\alpha - 2\mu & -i\alpha(1 - 2\mu)/k & \sqrt{2}\alpha \\ i\alpha(1 - 2\mu)/k & 2 - 2\alpha - 2\mu & 0 \\ \sqrt{2}\alpha & 0 & 2 - 2\mu \end{pmatrix}.$$

We shall first derive strict inequalities on μ to obtain the positive definiteness of this matrix, using Sylvester's criterion. From $\mathbf{D}_{0,0}$ we deduce the first condition $0 \leq \mu < \alpha$. The determinant of the upper left 2×2 block reads

$$\delta_2(\mu; \alpha, k) = 4(\alpha - \mu)(1 - \alpha - \mu) - \frac{\alpha^2}{k^2}(1 - 2\mu)^2.$$

Since the last term increases with $|k|$, it suffices to consider δ_2 for $k = 1$. Next we want to establish the positivity of

$$\frac{\delta_2(\mu; \alpha, 1)}{4(1 - \alpha^2)} = \mu^2 - \mu + \alpha \frac{1 - 5\alpha/4}{1 - \alpha^2}.$$

The zero order term of this quadratic polynomial is positive on the relevant α -interval $(0, \frac{7-\sqrt{17}}{4}) \subset (0, \frac{4}{5})$, taking its maximum value $\frac{1}{4}$ at $\alpha = \frac{1}{2}$. For that limiting case, the r.h.s. reads $(\mu - \frac{1}{2})^2$, and for $0 < \alpha < \frac{1}{2}$, $\delta_2(\mu; \alpha, 1)$ always has a zero in the interval $(0, \frac{1}{2})$. This discussion yields the second condition $0 \leq \mu < \frac{1}{2}$, related to $\alpha < \frac{1}{2}$.

Next we consider the positivity of the determinant of the upper left 3×3 block, which reads

$$\delta_3(\mu; \alpha, k) = 8(1-\mu)(\alpha-\mu)(1-\alpha-\mu) - 4\alpha^2(1-\alpha-\mu) - 2\frac{\alpha^2}{k^2}(1-\mu)(1-2\mu)^2.$$

For the same reason as before, we only have to consider the case $k = 1$. For the resulting cubic polynomial in μ we want to find its largest zero in the interval $[0, \frac{1}{2}]$ w.r.t. the parameter $\alpha \in [0, \frac{1}{2}]$. By numerical inspection we find that $\alpha_0 \approx 0.4684$ yields $\delta_3(\mu; \alpha_0, 1) \geq 0$ for $\mu \in [0, 0.273796\dots]$. This yields the third condition on μ and shows that the matrix inequality (4.17) holds with $\mu_0 := 0.273796\dots$, uniformly in $k \neq 0$. This somewhat more involved discussion shows that the decay rate can be improved to $2\mu_0 \approx 0.547592$. This finished the proof of Theorem 1.

Remark: To appreciate the above decay rate μ_0 (since $e(f)$ is a quadratic functional), we compare it to a numerical computation of the spectral gap of the “infinite matrices” $-ik\mathbf{L}_1 + \mathbf{L}_2$, $k \in \mathbb{Z}$ from (4.15). To this end we cut out the upper left $n \times n$ submatrix for large values of n . For increasing n the spectral gap approaches 0.6973. Hence our decay rate is off by only a factor of about 2.5. If one desired a closer bound, one could work with a \mathbf{P} matrix with a larger block, say 3×3 , in the upper left.

Acknowledgements The second author (AA) was supported by the FWF-doctoral school “Dispersion and dispersion in non-linear partial differential equations”. The third author (EC) was partially supported by U.S. N.S.F. grant DMS 1501007.

References

1. Achleitner, F., Arnold, A., Stürzer, D.: Large-time behavior in non-symmetric Fokker-Planck equations. to appear in *Rivista di Matematica della Università di Parma* (2015)
2. Arnold, A., Carlen, E., and Ju, Q.: Large-time behavior of non-symmetric Fokker-Planck type equations. *Commun. Stoch. Anal.* **2**, no. 1, 153–175 (2008)
3. Arnold, A., and Erb, J. Sharp entropy decay for hypocoercive and non-symmetric Fokker-Planck equations with linear drift. arXiv preprint arXiv:1409.5425 (2014)
4. Arnold, A., Markowich, P., Toscani, G., and Unterreiter, A. On convex Sobolev inequalities and the rate of convergence to equilibrium for Fokker-Planck type equations. *Comm. Partial Differential Equations* **26**, no. 1-2, 43–100 (2001)
5. Bakry, D., and Émery, M. Hypercontractivité de semi-groupes de diffusion. *C. R. Acad. Sci. Paris Sér. I Math.* **299**, no. 15, 775–778 (1984)
6. Bakry, D., and Émery, M. Diffusions hypercontractives. In *Séminaire de probabilités, XIX, 1983/84*, vol. 1123 of *Lecture Notes in Math.* Springer, Berlin 177–206 (1985)
7. Bergmann, P., and Lebowitz, J.L. A new approach to nonequilibrium processes, *Phys. Rev.* **99**, no. 2, 578–587 (1955)
8. Bhatnagar, P.L., Gross, E.P., and Krook, M. A Model for Collision Processes in Gases. I. Small Amplitude Processes in Charged and Neutral One-Component Systems. *Physical Review* **94**, no. 3, 511–525 (1954)
9. Bobkov, S.G., and Tetali, P. Modified logarithmic Sobolev inequalities in discrete settings. *J. Theoret. Probab.* **19**, no. 2, 289–336 (2006)

10. Coleman, R. On Krawtchouk polynomials. arXiv preprint arXiv:1101.1798 (2011)
11. Diaconis, P., and Saloff-Coste, L. Logarithmic Sobolev inequalities for finite Markov chains. *Ann. Appl. Probab.* **6**, no. 3, 695–750, (1996)
12. Dolbeault, J., Mouhot, C., and Schmeiser, C. Hypocoercivity for linear kinetic equations conserving mass. *Trans. Amer. Math. Soc.* **367**, no. 6, 3807–3828 (2015)
13. Horn, R.A., and Johnson, C.R. *Matrix analysis*. Cambridge University Press, Cambridge, 2nd ed. (2013)
14. Horn, R.A., and Johnson, C.R. *Topics in matrix analysis*. Cambridge University Press, Cambridge, (1991)
15. Michel, A.N., Hou, L., and Liu, D. *Stability of dynamical systems*. Birkhäuser/Springer, 2nd ed. (2015)
16. Norris, J.R. *Markov chains*. Cambridge University Press, Cambridge, (1997)
17. Perthame, B. *Transport Equations in Biology*. Birkhäuser, Basel (2007)
18. Seneta, E., *Nonnegative matrices and Markov chains*, Springer Series in Statistics, 2nd ed., Springer-Verlag, New York, (1981)
19. van Kampen, N.G. *Stochastic processes in physics and chemistry*. Elsevier, 3rd ed. (2007)
20. Villani, C. Hypocoercivity. *Mem. Amer. Math. Soc.* **202**, no. 950 (2009)