# A UNIFORM FIRST-ORDER METHOD FOR THE DISCRETE ORDINATE TRANSPORT EQUATION WITH INTERFACES IN X,Y-GEOMETRY* 

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#### Abstract

A uniformly first-order convergent numerical method for the discrete-ordinate transport equation in the rectangle geometry is proposed in this paper. Firstly we approximate the scattering coefficients and source terms by piecewise constants determined by their cell averages. Then for each cell, following the work of De Barros and Larsen [1, 19], the solution at the cell edge is approximated by its average along the edge. As a result, the solution of the system of equations for the cell edge averages in each cell can be obtained analytically. Finally, we piece together the numerical solution with the neighboring cells using the interface conditions. When there is no interface or boundary layer, this method is asymptotic-preserving, which implies that coarse meshes (meshes that do not resolve the mean free path) can be used to obtain good numerical approximations. Moreover, the uniform first-order convergence with respect to the mean free path is shown numerically and the rigorous proof is provided.


Mathematics subject classification: 41A30, 41A60, 65D25.
Key words: Transport equation, Interface, Diffusion limit, Asymptotic preserving, Uniform numerical convergence, X,Y-geometry.

## 1. Introduction

The transport equation plays an important role in many physical applications, such as neutron transport, radiative transfer, high frequency waves in heterogeneous and random media, semiconductor device simulation and so on. One difficulty about solving this equation numerically is when its mean free path (the average distance a particle travels between two successive interactions with the background media) is small, which requires numerical resolution of the small scale. Historically researchers use the diffusion limit to approximate the solution when the cost is too much to solve the transport equation directly. This small scale is embodied by the introduction of a dimensionless parameter $\epsilon$ into the transport equation and the diffusion limit can be obtained when $\epsilon \rightarrow 0$.

In this paper we consider the steady state isotropic neutron transport equation with interfaces in the X,Y-geometry. The interface condition to be used is that the density of all directions is continuous at the interface, which often arises in neutron transport equations. There is another kind of interface condition which always arises in radiative transfer equations as an approximation of high frequency waves in random and heterogeneous media, where the energy flux is continuous $[2,14,16]$. As the interface condition is local, for the density continuous case, we only need to consider the one-dimensional interface analysis which is given in [15].

[^0]The diffusion limit for two-dimensional neutron transport equation with interfaces is derived in this paper. For simplicity, we assume that the incoming particle density at the boundaries is isotropic, in which case no boundary layer exists when $\epsilon \rightarrow 0$. A first-order numerical method uniform with respect to $\epsilon$ is proposed and investigated here. The most used determined method for this problem has been the discrete ordinate method which is a semi-discretization of the velocity field. Our method is based on the discrete ordinate form and the full discretization is also considered. As in [15], we approximate the coefficients by piecewise constants determined by their cell averages. Then for each cell, following the work of De Barros and Larsen [1,19], the solution at the cell edge is approximated by its average along the edge. The obtained system of equations for the cell edge averages can be solved analytically in each cell and we can get the solution of the whole domain by using the interface conditions to piece together the numerical solution with the neighboring cells. This method is a direct extension of the one-dimensional scheme proposed in [15].

Asymptotic-preserving (AP) schemes have been proved quite successfully for problems with small scales. A scheme with AP properly means it is a good scheme for the original equation, which in the limit as the small parameter goes to zero, becomes an effective scheme for the limit equation [10]. It was proved in [9] for the linear transport equation with boundary conditions that an AP scheme converges uniformly with respect to $\epsilon$. This implies we can solve the transport equation without resolving the small scale. For more applications of AP schemes we refer to $[5-7]$ for plasmas and fluids and $[3,11,12]$ for hyperbolic systems with stiff relaxations. When there is no interface or boundary layer, the method we propose here is proved to be AP and its uniform first-order accuracy is demonstrated numerically. The rigorous convergence proof relies heavily on the eigenfunction expansion of the constant coefficient, one-dimensional discrete-ordinate equations, but the idea is quite similar to the proof of uniform second-order convergence for one-dimensional case in [15]. AP means that at the interior of the materials where the solution varies slowly (no matter whether $\epsilon$ is big or small), we do not need to resolve $\epsilon$ to get accurate approximations, though the equation itself requires under-resolving. When using coarse meshes, most methods can not provide satisfactory results for problems with boundary layers even if they are AP inside. The method proposed in [15] seems the first effort that is uniformly convergent valid up to the boundary. It will be demonstrated numerically that our two-dimensional method is not valid at the layers, but this can be improved by resolving $\epsilon$ locally at the boundaries or interfaces.

Similar ideas can be found in $[1,17,19]$ but with auxiliary equations and in an iterative way. They only discussed piecewise constant case and the required storage is much more than that of our approach. Problems with coefficients depending on space are investigated in this paper and we also present its AP property and uniform accuracy with respect to $\epsilon$.

The arrangement of this paper is as follows. In section 2, we introduce the two-dimensional neutron transport equation and its discrete ordinate form, and derive their diffusion limit with interfaces. In section 3, the scheme is given and its AP property is proved in section 4. Several numerical examples are presented in section 5 to test the AP property, and the uniform accuracy is discussed. Finally we make some conclusions in section 6.

## 2. Neutron Transport Equation in Two-Dimensions

The steady state, isotropic, neutron transport equation in the X,Y-geometry reads as: for $\mathbf{z} \in \Omega \subset \mathbb{R}^{2}, \mu \in S=\left\{\mathbf{u} \in \mathbb{R}^{2}:|\mathbf{u}|=1\right\}$,

$$
\begin{equation*}
\mathbf{u} \cdot \nabla \Psi(\mathbf{z}, \mathbf{u})+\frac{\sigma_{T}(\mathbf{z})}{\epsilon} \Psi(\mathbf{z}, \mu)=\frac{1}{2 \pi}\left(\frac{\sigma_{T}(\mathbf{z})}{\epsilon}-\epsilon \sigma_{a}(\mathbf{z})\right) \int_{S} \Psi(\mathbf{z}, \mathbf{u}) d \mathbf{u}+\epsilon q(\mathbf{z}) \tag{2.1}
\end{equation*}
$$

with the boundary conditions

$$
\begin{equation*}
\Psi(\mathbf{z}, \mathbf{u})=\Psi_{\Gamma}^{-}, \quad \text { for }(\mathbf{z}, \mathbf{u}) \in \Gamma_{u}^{-}=\{\mathbf{z} \in \Gamma=\partial \Omega, \mathbf{u} \cdot n(\mathbf{z})<0\} \tag{2.2}
\end{equation*}
$$

Here $\sigma_{T}, \sigma_{a}, q$ are the total cross section, absorption cross section and source respectively. $\Psi(\mathbf{z}, \mathbf{u})$ is the function we want, which represent the density of the particles moving in direction $\mathbf{u}$ at position $\mathbf{z}$. For isotropic boundary condition, $\Psi_{\Gamma}^{-}$only depends on $\Gamma$ but is independent of $\mathbf{u}$. The diffusion limit can be obtained by introducing

$$
\begin{equation*}
\Psi=\sum_{n=0}^{\infty} \epsilon^{n} \Psi^{(n)} \tag{2.3}
\end{equation*}
$$

into (2.1) and equating the coefficients of different powers of $\epsilon$. The $\mathcal{O}(1 / \epsilon)$ equation is

$$
\begin{equation*}
\Psi^{(0)}=\frac{1}{2 \pi} \int_{S} \Psi^{(0)} d \mathbf{u} \tag{2.4a}
\end{equation*}
$$

The $\mathcal{O}(1)$ and mathcal $O(\epsilon)$ equations are

$$
\begin{align*}
\mathbf{u} \cdot \nabla \Psi^{(0)}+\sigma_{T} \Psi^{(1)} & =\frac{\sigma_{T}}{2 \pi} \int_{S} \Psi^{(1)} d \mathbf{u}  \tag{2.4b}\\
\mathbf{u} \cdot \nabla \Psi^{(1)}+\sigma_{T} \Psi^{(2)} & =\sigma_{T} \frac{1}{2 \pi} \int_{S} \Psi^{(2)} d \mathbf{u}-\sigma_{a} \frac{1}{2 \pi} \int_{S} \Psi^{(0)} d \mathbf{u}+q . \tag{2.4c}
\end{align*}
$$

Dividing both sides of $(2.4 \mathrm{~b})$ by $\sigma_{T}$ gives

$$
\Psi^{(1)}=-\frac{1}{\sigma_{T}} \mathbf{u} \cdot \nabla \Psi^{(0)}+\frac{1}{2 \pi} \int_{S} \Psi^{(1)} d \mathbf{u} .
$$

Then take the gradient of both sides of the above equation, and left dot the resulting equation by $\mathbf{u}$. Integrating over $S$, and using $\int_{S} \mathbf{u} d \mathbf{u}=(0,0)^{T}$, we have

$$
\int_{S} \mathbf{u} \cdot \nabla \Psi^{(1)} d \mathbf{u}=-\int_{S} \mathbf{u} \cdot \nabla\left(\frac{1}{\sigma_{T}} \mathbf{u} \cdot \nabla \Psi^{(0)}\right) d \mathbf{u}
$$

Finally integrating both sides of (2.4c) and noting $\int_{S} d \mathbf{u}=2 \pi, \int_{S} \mathbf{u} \cdot \mathbf{u} d \mathbf{u}=2 \pi$ yield

$$
\begin{equation*}
-\nabla \cdot\left(\frac{1}{2 \sigma_{T}} \nabla \Phi\right)+\sigma_{a} \Phi=q \tag{2.5}
\end{equation*}
$$

where $\Phi=\frac{1}{2 \pi} \int_{S} \Psi^{(0)} d \mathbf{u}$. This is the diffusion limit of (2.1) which means that when $\epsilon \rightarrow 0$, the solution of (2.1) becomes isotropic and can be approximated by the solution of (2.5).

In the Cartesian coordinate system, assume $\mathbf{z}=(x, y) \quad \mathbf{u}=(\mu, \nu)$. Let

$$
V=\{-2 M, \cdots,-1,1, \cdots, 2 M\} .
$$

The discrete-ordinates form of (2.1) is

$$
\begin{equation*}
\mu_{m} \frac{\partial}{\partial x} \psi_{m}+\nu_{m} \frac{\partial}{\partial y} \psi_{m}+\frac{\sigma_{T}}{\epsilon} \psi_{m}=\left(\frac{\sigma_{T}}{\epsilon}-\epsilon \sigma_{a}\right) \sum_{n \in V} \psi_{n} w_{n}+\epsilon q, \quad m \in V \tag{2.6}
\end{equation*}
$$

where $\mu_{m}^{2}+\nu_{m}^{2}=1$. For simplicity, the computational domain we consider here is a rectangle:

$$
\Omega=\{(x, y) \mid x \in[0, a], y \in[0, b]\}
$$

Other more complex domains can be approximated by rectangles, thus are direct extensions. When $\Omega$ is a rectangle, the boundary conditions become

$$
\begin{array}{ll}
\psi_{m}(0, y)=\psi_{L}(y), \quad \mu_{m}>0 ; & \psi_{m}(a, y)=\psi_{R}(y), \quad \mu_{m}<0 \\
\psi_{m}(x, 0)=\psi_{B}(x), \quad \nu_{m}>0 ; \quad \psi_{m}(x, b)=\psi_{T}(x), \quad \nu_{m}<0 \tag{2.7}
\end{array}
$$

The corresponding diffusion limit of (2.6) can be obtained similarly as for (2.1) by introducing

$$
\psi_{m}=\sum_{n=0}^{\infty} \epsilon^{n} \psi_{m}^{(n)}
$$

The equations corresponding to (2.4) are

$$
\begin{align*}
& \psi_{m}^{(0)}=\sum_{n \in V} w_{n} \psi_{n}^{(0)}  \tag{2.8a}\\
& \mu_{m} \partial_{x} \psi_{m}^{(0)}+\nu_{m} \partial_{y} \psi_{m}^{(0)}+\sigma_{T} \psi_{m}^{(1)}=\sigma_{T} \sum_{n \in V} w_{n} \psi_{n}^{(1)}  \tag{2.8b}\\
& \mu_{m} \partial_{x} \psi_{m}^{(1)}+\nu_{m} \partial_{y} \psi_{m}^{(1)}+\sigma_{T} \psi_{m}^{(2)}=\sigma_{T} \sum_{n \in V} w_{n} \psi_{n}^{(2)}-\sigma_{a} \sum_{n \in V} w_{n} \psi_{n}^{(0)}+q \tag{2.8c}
\end{align*}
$$

In order to get the same form of diffusion equation as (2.5), the quadrature set $\left\{\mu_{n}, \nu_{n}, w_{n}\right\}$ should satisfy

$$
\begin{align*}
& \sum_{n \in V} w_{n}=1, \quad \sum_{n \in V} w_{n} \mu_{n}=0, \quad \sum_{n \in V} w_{n} \nu_{n}=0  \tag{2.9a}\\
& \sum_{n \in V} w_{n} \mu_{n} \nu_{n}=0, \quad \sum_{n \in V} w_{n} \mu_{n}^{2}=\frac{1}{2}, \quad \sum_{n \in V} w_{n} \nu_{n}^{2}=\frac{1}{2} \tag{2.9~b}
\end{align*}
$$

and diffusion limit now is

$$
\begin{equation*}
-\partial_{x}\left(\frac{1}{2 \sigma_{T}} \partial_{x} \phi\right)-\partial_{y}\left(\frac{1}{2 \sigma_{T}} \partial_{y} \phi\right)+\sigma_{a} \phi=q \tag{2.10}
\end{equation*}
$$

where $\phi=\sum_{n \in V} w_{n} \psi_{n}^{(0)}$. Because of the isotropic boundary condition, we have

$$
\begin{equation*}
\left.\phi\right|_{x=0}=\psi_{L},\left.\quad \phi\right|_{x=a}=\psi_{R},\left.\quad \phi\right|_{y=0}=\psi_{B},\left.\quad \phi\right|_{y=b}=\psi_{T} \tag{2.11}
\end{equation*}
$$

As the commonly used $\left\{\mu_{n}, \nu_{n}, w_{n}\right\}$ are symmetric, we can assume for $\theta_{m} \in\left(0, \frac{\pi}{2}\right), m=$ $1, \cdots, M$,

$$
\begin{array}{ll}
\theta_{m}=\theta_{m+M}-\frac{\pi}{2}=\theta_{-m}-\pi=\theta_{-m-M}-\frac{3}{2} \pi, & m=1, \cdots, M \\
\mu_{m}=\cos \theta_{m}, \quad \nu_{m}=\sin \theta_{m}, & m \in V \\
w_{m}=w_{-m}=w_{-m-M}=w_{m+M}>0, & m=1, \cdots, M \tag{2.12c}
\end{array}
$$

It is easy to check that when $\left\{\mu_{n}, \nu_{n}, w_{n}\right\}$ satisfy (2.12), (2.9) holds automatically. The distribution of the quadrature set can be seen more clearly from Figure 2.1.


Fig. 2.1. The distribution of the quadrature set.

The problem we consider here is when some heterogeneous media are put together and the particles do not change their directions when passing through the interfaces. Note that Eq. (2.6) controls the movement of the particles, that means at the interfaces the coefficients $\sigma_{T}$, $\sigma_{a}$ and $q$ have discontinuities, but $\psi_{m}$ are continuous. For any interface lines $\alpha$, assuming the two different media are denoted by + and - , we have

$$
\begin{equation*}
\left.\psi_{m}^{+}\right|_{\alpha}=\left.\psi_{m}^{-}\right|_{\alpha} \tag{2.13}
\end{equation*}
$$

In this paper we only consider the case when the interfaces consist of pieces of lines parallel to the $x$ or $y$ coordinates. Because the discontinuities only occur in one-direction for each piece of interface line, we only need to use the one dimensional interface analysis. As proved in [15], in one-dimension where the space coordinate is $z$, the interface conditions for the diffusion limit are

$$
\phi^{+}=\phi^{-}, \quad \frac{1}{\sigma_{T}^{+}} \partial_{z} \phi^{+}=\frac{1}{\sigma_{T}^{-}} \partial_{z} \phi^{-} .
$$

Thus the interface conditions for our two-dimensional case are

$$
\begin{align*}
& \left.\phi^{+}\right|_{\alpha}=\left.\phi^{-}\right|_{\alpha}  \tag{2.14a}\\
& \left.\frac{1}{\sigma_{T}^{+}} \partial_{z} \phi^{+}\right|_{\alpha}=\left.\frac{1}{\sigma_{T}^{-}} \partial_{z} \phi^{-}\right|_{\alpha}, \tag{2.14b}
\end{align*}
$$

where $z$ is $x(y)$ when $\alpha$ is parallel to $y(x)$.

## 3. Derivation of the Scheme

The scheme is focused on the space discretization of the discrete ordinate form (2.6). The basic idea of this method includes two approximations:
i) to approximate the coefficients by piecewise constants;
ii) to approximate the solution on each cell edge by its cell edge average as in (3.6).

The details are as follows: First, generate a set of grid points

$$
G=\left\{\left(x_{i}, y_{j}\right) \mid 0=x_{0}<x_{1}<\cdots<x_{N}=a, 0=y_{0}<y_{1}<\cdots<y_{N^{\prime}}=b\right\}
$$

Let

$$
\triangle=_{i=0, \cdots, N-1 ; j=0, \cdots, N^{\prime}-1}\left\{\left|x_{i+1}-x_{i}\right|,\left|y_{j+1}-y_{j}\right|\right\}
$$

and for $i=0, \cdots, N-1, j=0, \cdots, N^{\prime}-1$,

$$
\begin{aligned}
& \sigma_{a i j}=\frac{1}{\left(x_{i+1}-x_{i}\right)\left(y_{j+1}-y_{j}\right)} \int_{y_{j}}^{y_{j+1}} \int_{x_{i}}^{x_{i+1}} \sigma_{a} d x d y \\
& \sigma_{T i j}=\frac{1}{\left(x_{i+1}-x_{i}\right)\left(y_{j+1}-y_{j}\right)} \int_{y_{j}}^{y_{j+1}} \int_{x_{i}}^{x_{i+1}} \sigma_{T} d x d y \\
& q_{i j}=\frac{1}{\left(x_{i+1}-x_{i}\right)\left(y_{j+1}-y_{j}\right)} \int_{y_{j}}^{y_{j+1}} \int_{x_{i}}^{x_{i+1}} q d x d y
\end{aligned}
$$

The first approximation of (2.6) is

$$
\begin{equation*}
\mu_{m} \frac{\partial}{\partial x} \tilde{\psi}_{m}+\nu_{m} \frac{\partial}{\partial y} \tilde{\psi}_{m}+\frac{\tilde{\sigma}_{T}}{\epsilon} \tilde{\psi}_{m}=\left(\frac{\tilde{\sigma}_{T}}{\epsilon}-\epsilon \tilde{\sigma}_{a}\right) \sum_{n \in V} \tilde{\psi}_{n} w_{n}+\epsilon \tilde{q}, \quad m \in V \tag{3.1}
\end{equation*}
$$

where $\tilde{\sigma}_{a}, \tilde{\sigma}_{T}, \tilde{q}$ are piecewise constants, i.e.,,

$$
\begin{equation*}
\tilde{\sigma}_{a}(x, y)=\sigma_{a i j}, \quad \tilde{\sigma}_{T}(x, y)=\sigma_{T i j}, \quad \tilde{q}(x, y)=q_{i j}, \quad(x, y) \in\left(x_{i}, x_{i+1}\right] \times\left(y_{j}, y_{j+1}\right] \tag{3.2}
\end{equation*}
$$

Now (3.1) can be solved by similar ideas as in $[1,17,19]$, but the approach is different. On each rectangle $\left[x_{i}, x_{i+1}\right] \times\left[y_{j}, y_{j+1}\right]$, we seek the solution of

$$
\begin{equation*}
\mu_{m} \partial \tilde{\psi}_{m}+\nu_{m} \partial y \tilde{\psi}_{m}+\frac{\sigma_{T i j}}{\epsilon} \tilde{\psi}_{m}=\left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} \tilde{\psi}_{n} w_{n}+\epsilon q_{i j}, \quad m \in V \tag{3.3}
\end{equation*}
$$

Let

$$
\begin{equation*}
\hat{\psi}_{m j}(x)=\frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \tilde{\psi}_{m} d y, \quad \check{\psi}_{m i}(y)=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \tilde{\psi}_{m} d x \tag{3.4}
\end{equation*}
$$

Integrating both sides of (3.3) from $y_{j}$ to $y_{j+1}$ gives

$$
\begin{align*}
& \mu_{m} \partial_{x} \hat{\psi}_{m j}(x)+\frac{\sigma_{T i j}}{\epsilon} \hat{\psi}_{m j}(x) \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \hat{\psi}_{n j}(x)+\epsilon q_{i j}-\nu_{m} \frac{\tilde{\psi}_{m}\left(x, y_{j+1}\right)-\tilde{\psi}_{m}\left(x, y_{j}\right)}{y_{j+1}-y_{j}}, \tag{3.5a}
\end{align*}
$$

and from $x_{i}$ to $x_{i+1}$,

$$
\begin{align*}
& \nu_{m} \partial_{y} \check{\psi}_{m i}(y)+\frac{\sigma_{T i j}}{\epsilon} \check{\psi}_{m i}(y) \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \check{\psi}_{n i}(y)+\epsilon q_{i j}-\mu_{m} \frac{\tilde{\psi}_{m}\left(x_{i+1}, y\right)-\tilde{\psi}_{m}\left(x_{i}, y\right)}{x_{i+1}-x_{i}} . \tag{3.5b}
\end{align*}
$$

Introduce the second approximation

$$
\begin{equation*}
\tilde{\psi}_{m}\left(x, y_{j+1}\right) \approx \check{\psi}_{m i}\left(y_{j+1}\right), \quad \tilde{\psi}_{m}\left(x, y_{j}\right) \approx \check{\psi}_{m i}\left(y_{j}\right) \tag{3.6a}
\end{equation*}
$$

in (3.5a) and

$$
\begin{equation*}
\tilde{\psi}_{m}\left(x_{i}, y\right) \approx \hat{\psi}_{m j}\left(x_{i}\right), \quad \tilde{\psi}_{m}\left(x_{i+1}, y\right) \approx \hat{\psi}_{m j}\left(x_{i+1}\right) \tag{3.6b}
\end{equation*}
$$

in (3.5b), we have

$$
\begin{align*}
& \mu_{m} \partial_{x} \hat{\psi}_{m j}(x)+\frac{\sigma_{T i j}}{\epsilon} \hat{\psi}_{m j}(x) \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \hat{\psi}_{n j}(x)+\epsilon q_{i j}-\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}, \quad m \in V, \tag{3.7a}
\end{align*}
$$

and

$$
\begin{align*}
& \nu_{m} \partial_{y} \check{\psi}_{m i}(y)+\frac{\sigma_{T i j}}{\epsilon} \check{\psi}_{m i}(y) \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \check{\psi}_{n i}(y)+\epsilon q_{i j}-\mu_{m} \frac{\hat{\psi}_{m j}\left(x_{i+1}\right)-\hat{\psi}_{m j}\left(x_{i}\right)}{x_{i+1}-x_{i}}, \quad m \in V . \tag{3.7b}
\end{align*}
$$

The boundary conditions for $\hat{\psi}_{m j}$ and $\check{\psi}_{m i}$ corresponding to (2.7) become

$$
\begin{align*}
& \left.\hat{\psi}_{m j}\right|_{x=0}=\frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \psi_{L} d y=\psi_{L j}, \quad \mu_{m}<0 ;  \tag{3.8a}\\
& \left.\hat{\psi}_{m j}\right|_{x=a}=\frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \psi_{R} d y=\psi_{R j}, \quad \mu_{m}>0 ;  \tag{3.8b}\\
& \left.\check{\psi}_{B i}\right|_{y=0}=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x^{i+1}} \psi_{B} d x=\psi_{B j}, \quad \nu_{m}<0 ;  \tag{3.8c}\\
& \left.\check{\psi}_{T i}\right|_{y=b}=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \psi_{T} d x=\psi_{T j}, \quad \nu_{m}>0 . \tag{3.8d}
\end{align*}
$$

Clearly there are $4 M N+4 M N^{\prime}$ boundary conditions in all.
The system of Eq. (3.7) can be solved analytically using the same idea as in [15] and the following part is trying to find the general solution of (3.7). We should emphasize here that in (3.7a), $\mu_{m}$ could be the same for different $m$. From (2.12), the same $\mu_{m}$ occurs in

$$
S_{\mu}=\left\{\mu_{n}, \mid n \in V\right\}
$$

twice at most. Define a new set

$$
\hat{S}=\left\{\hat{\mu}_{n}, \hat{w}_{n} \mid n \in \hat{V}\right\},
$$

where $\hat{V}=\{-\hat{M}, \cdots,-1,1, \cdots, \hat{M}\}$ in the following way: if $\mu_{m}$ occurs once in $S_{\mu}$, keep $\mu_{m}, w_{m}$ unchanged in $\hat{S}$, otherwise for $\mu_{m}=\mu_{m_{1}}, m \neq m_{1}$, let $\mu_{m}$ occur once as $\hat{\mu}_{m^{\prime}}$ in $\hat{S}$ and the corresponding $\hat{w}_{m^{\prime}}=w_{m}+w_{m_{1}} .2 \hat{M}$ is the order of this new set $\hat{S}$, in which $\hat{\mu}_{n}$ are different from each other. Obviously, in $S_{\mu}$ there are $4 M-2 \hat{M} \mu_{m}$ occurring twice and $4 \hat{M}-4 M$ occurring once.

Introduce some new variables

$$
\begin{array}{lll}
\hat{\varphi}_{m^{\prime} j}=\hat{\psi}_{m j}, & \mu_{m}=\hat{\mu}_{m^{\prime}} & \mu_{m} \text { occurs once in } \hat{S} \\
\hat{\varphi}_{m^{\prime} j}=\frac{w_{m} \hat{\psi}_{m j}+w_{m_{1}} \hat{\psi}_{m j}}{w_{m}+w_{m_{1}}}, & \mu_{m}=\mu_{m_{1}}=\hat{\mu}_{m^{\prime}} & \mu_{m} \text { occurs twice in } \hat{S} \tag{3.9b}
\end{array}
$$

and from (3.7a) $\hat{\varphi}_{m^{\prime} j}, m^{\prime} \in \hat{V}$ satisfy

$$
\begin{equation*}
\hat{\mu}_{m^{\prime}} \partial_{x} \hat{\varphi}_{m^{\prime} j}(x)+\frac{\sigma_{T i j}}{\epsilon} \hat{\varphi}_{m^{\prime} j}(x)=\left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} \hat{w}_{n} \hat{\varphi}_{n j}(x)+\epsilon q_{i j}-\hat{r}_{m^{\prime} i j}, \tag{3.10}
\end{equation*}
$$

where

$$
\hat{r}_{m^{\prime} i j}=\left\{\begin{array}{l}
\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}, \quad \mu_{m}=\hat{\mu}_{m^{\prime}}, \quad \mu_{m} \text { occurs once in } S_{\mu}  \tag{3.11}\\
\frac{1}{w_{m}+w_{m_{1}}}\left(w_{m} \nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}+w_{m_{1}} \nu_{m_{1}} \frac{\check{\psi}_{m_{1} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}\right) \\
\mu_{m}=\mu_{m_{1}}=\hat{\mu}_{m^{\prime}}, \quad \mu_{m} \text { occurs twice in } S_{\mu}
\end{array}\right.
$$

$\hat{r}_{m^{\prime} i j}$ are undetermined constants on $\left[x_{i}, x_{i+1}\right] \times\left[y_{j}, y_{j+1}\right]$. For this new system of equations (3.10), $\hat{\mu}_{m^{\prime}}$ are different from each other. Assume

$$
1>\hat{\mu}_{\hat{M}}>\cdots>\hat{\mu}_{1}>0>\hat{\mu}_{-1}>\cdots>\hat{\mu}_{-\hat{M}}>-1
$$

and the symmetry: $\hat{\mu}_{n}=-\hat{\mu}_{-n}, \hat{w}_{n}=\hat{w}_{-n}$ can be obtained from (2.12).
This new system of equations (3.10) is exactly the same as the one-dimensional system of equations solved in [15] but the $\hat{r}_{m i j}$ terms. We can find the $4 \hat{M}$ relations between $\left\{\varphi_{m j}\left(x_{i}\right), m \in\right.$ $\hat{V}\}$ and $\left\{\varphi_{m j}\left(x_{i+1}\right), m \in \hat{V}\right\}$ through the general solution of (3.10) as in [15]. As proved in Appendix of [15], we have the following theorem:

Theorem 3.1. Consider the equation

$$
\begin{equation*}
\sum_{n \in \hat{V}} \frac{\hat{w}_{n}}{1-\hat{\mu}_{n} \xi}=\frac{1}{1-\epsilon^{2} \frac{\sigma_{a i j}}{\sigma_{T i j}}} \tag{3.12}
\end{equation*}
$$

i) When $\sigma_{a i j} \neq 0$, (3.12) has $2 \hat{M}$ simple roots that occur in positive/negative pairs, let them be $\xi_{n}(1 \leq|n| \leq \hat{M})$. Assume

$$
\begin{equation*}
\hat{l}_{m}^{(n)}=\frac{1}{1-\hat{\mu}_{m} \xi_{n}} \tag{3.13}
\end{equation*}
$$

we have

$$
\sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{l}_{m}^{(n)}=\left\{\begin{array}{cc}
0 & n \neq k  \tag{3.14a}\\
c^{(k)} & n=k
\end{array}\right.
$$

where $\hat{c}^{(k)}$ satisfy

$$
\sum_{k \in \hat{V}} \frac{1}{c^{(k)}} \hat{l}_{m}^{(k)} l_{n}^{(k)}=\left\{\begin{array}{cl}
0 & m \neq n  \tag{3.14b}\\
\frac{1}{\hat{w}_{n} \hat{\mu}_{n}} & m=n .
\end{array}\right.
$$

ii) When $\sigma_{a i j}=0$, (3.12) has $2 \hat{M}-2$ simple roots appear in positive/negative pairs while 0 is a double root. Assume $\xi_{n}(1 \leq n \leq \hat{M}-1)$ is the unique (positive, simple) root in $\left(1 / \hat{\mu}_{n+1}, 1 / \hat{\mu}_{n}\right), \xi_{-n}=-\xi_{n}$ and $\hat{l}_{m}^{(n)}=\frac{1}{1-\hat{\mu}_{m} \xi_{n}} .(3.14 a)$ still holds for $k, n \in\{-\hat{M}+$ $1, \cdots, \hat{M}-1\}$. Moreover, defining

$$
c^{(-\hat{M})}=c^{(\hat{M})}=\sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m}^{2}=\frac{1}{2}
$$

we have

$$
\sum_{1 \leq|k| \leq \hat{M}-1} \frac{1}{c^{(k)}} l_{n}^{(k)} \hat{l}_{m}^{(k)}+\frac{\hat{\mu}_{n}}{\hat{c}^{(-\hat{M})}}+\frac{\hat{\mu}_{m}}{c^{(\hat{M})}}=\left\{\begin{array}{cc}
0 & m \neq n  \tag{3.15a}\\
\frac{1}{\hat{w}_{n} \hat{\mu}_{n}} & m=n
\end{array}\right.
$$

and

$$
\begin{equation*}
\sum_{m \in \hat{V}} \hat{\mu}_{m} \hat{w}_{m} \hat{l}_{m}^{(k)}=0, \quad \sum_{m \in \hat{V}} \hat{\mu}_{m}^{2} \hat{w}_{m} \hat{l}_{m}^{(k)}=0 \tag{3.15b}
\end{equation*}
$$

The derivation of the general solution of (3.10) is put into two different cases:
i) When $\sigma_{a i j} \neq 0$, multiplying both sides of (3.10) by $w_{m} \hat{l}_{m}^{(k)}$ and summing over $\hat{V}$ gives

$$
\begin{align*}
& \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \partial_{x} \hat{\varphi}_{m j}+\frac{\sigma_{T i j}}{\epsilon} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j} \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)}\left(\sum_{n \in \hat{V}} \hat{w}_{n} \hat{\varphi}_{n j}+\epsilon q_{i j}\right)-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{r}_{m i j} . \tag{3.16}
\end{align*}
$$

It follows from (3.12)-(3.13), that

$$
\sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)}=\frac{1}{1-\epsilon^{2} \frac{\sigma_{a i j}}{\sigma_{T i j}}}
$$

and

$$
\begin{aligned}
& \frac{\sigma_{T i j}}{\epsilon} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}-\left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{m \in V} \hat{w}_{m} \hat{l}_{m}^{(k)} \sum_{n \in V} \hat{w}_{n} \hat{\varphi}_{n j} \\
= & \frac{\sigma_{T i j}}{\epsilon}\left(\sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{\varphi}_{m j}\right) \\
= & \xi_{k} \frac{\sigma_{T i j}}{\epsilon} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}
\end{aligned}
$$

Thus (3.16) becomes

$$
\begin{align*}
& \partial_{x} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}+\xi_{k} \frac{\sigma_{T i j}}{\epsilon} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j} \\
= & \left(\sigma_{T i j}-\epsilon^{2} \sigma_{a i j}\right) q_{i j}-\sum_{m \in \hat{V}} \hat{w} \hat{l}_{m}^{(k)} \hat{r}_{m i j} . \tag{3.17}
\end{align*}
$$

Multiplying both sides of (3.17) by $\exp \left(\frac{\sigma_{T i j}}{\epsilon} \xi_{k} x\right)$ and integrating from $x_{i}$ to $x_{i+1}$ give

$$
\begin{align*}
& \exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i+1}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i+1}\right)-\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i}\right) \\
= & \frac{\epsilon}{\xi_{k} \sigma_{T i j}}\left(\left(\sigma_{T i j}-\epsilon^{2} \sigma_{a i j}\right) q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{r}_{m i j}\right) \\
& \left(\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i+1}\right)-\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i}\right)\right) . \tag{3.18}
\end{align*}
$$

These are $2 \hat{M}$ independent relations between $\left\{\hat{\varphi}_{m j}\left(x_{i+1}\right), m \in \hat{V}\right\},\left\{\hat{\varphi}_{m j}\left(x_{i}\right), m \in \hat{V}\right\}$ and $\left\{\hat{r}_{m i j}, m \in \hat{V}\right\}$.
ii) When $\sigma_{a i j}=0$, (3.12) has $2 \hat{M}-1$ eigenvalues from (ii) of Theorem 3.1. For $1 \leq|k| \leq$ $\hat{M}-1$, we can use the same discussion as for $\sigma_{a i j} \neq 0$ to obtain $2 \hat{M}-2$ independent relations between $\left\{\hat{\varphi}_{m j}\left(x_{i+1}\right), m \in \hat{V}\right\},\left\{\hat{\varphi}_{m j}\left(x_{i}\right), m \in \hat{V}\right\}$ and $\left\{\hat{r}_{m i j}, m \in \hat{V}\right\}$. For the other two relations, we can multiply both sides of (3.10) by $\hat{w}_{m^{\prime}}$ and $\hat{w}_{m^{\prime}} \hat{\mu}_{m^{\prime}}$, sum over $\hat{V}$ respectively and obtain

$$
\begin{align*}
& \partial_{x} \sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}=\epsilon q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{r}_{m i j},  \tag{3.19}\\
& \partial_{x} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m}^{2} \hat{\varphi}_{m j}+\frac{\sigma_{T i j}}{\epsilon} \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}=-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{r}_{m i j} . \tag{3.20}
\end{align*}
$$

Integrating both sides of (3.19) from $x_{i}$ to $x \in\left(x_{i}, x_{i+1}\right]$ gives

$$
\begin{equation*}
\sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}(x)-\sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}\left(x_{i}\right)=\left(x-x_{i}\right)\left(\epsilon q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{r}_{m i j}\right) . \tag{3.21}
\end{equation*}
$$

In particular when $x=x_{i+1}$, we have

$$
\begin{equation*}
\sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}\left(x_{i+1}\right)-\sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m} \hat{\varphi}_{m j}\left(x_{i}\right)=\left(x_{i+1}-x_{i}\right)\left(\epsilon q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{r}_{m i j}\right) \tag{3.22}
\end{equation*}
$$

Integrating both sides of (3.20) from $x_{i}$ to $x_{i+1}$ and using (3.21), one gets

$$
\begin{align*}
& \sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m}^{2} \hat{\varphi}_{m j}\left(x_{i+1}\right)-\sum_{m \in V} \hat{w}_{m} \hat{\mu}_{m}^{2} \hat{\varphi}_{m j}\left(x_{i}\right)+\frac{\sigma_{T i j}}{\epsilon}\left(x_{i+1}-x_{i}\right) \sum_{m \in \hat{V}} \hat{w} \hat{\mu}_{m} \hat{\varphi}_{m j}\left(x_{i}\right) \\
= & \frac{1}{2}\left(x_{i+1}-x_{i}\right)^{2}\left(\epsilon q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{r}_{m i j}\right)-\left(x_{i+1}-x_{i}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{r}_{m i j} . \tag{3.23}
\end{align*}
$$

Up to now, for $\sigma_{a i j} \neq 0$ and $\sigma_{a i j}=0$, we've found $2 \hat{M}$ independent relations of $\left\{\hat{\varphi}_{m j}\left(x_{i+1}\right)\right.$, $m \in \hat{V}\},\left\{\hat{\varphi}_{m j}\left(x_{i}\right), m \in \hat{V}\right\}$ and $\left\{\hat{r}_{m i j}, m \in \hat{V}\right\}$. When coming back to $\hat{\psi}_{m j}$, from (3.7), for $\mu_{m}=\mu_{m_{1}}=\hat{\mu}_{m^{\prime}}$, one has

$$
\begin{align*}
& \hat{\mu}_{m^{\prime}} \partial_{x}\left(\hat{\psi}_{m j}(x)-\hat{\psi}_{m_{1} j}(x)\right)+\frac{\sigma_{T i j}}{\epsilon}\left(\hat{\psi}_{m j}-\hat{\psi}_{m_{1} j}\right) \\
= & \nu_{m_{1}} \frac{\check{\psi}_{m_{1} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}-\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}} . \tag{3.24}
\end{align*}
$$

Multiplying both sides of the above equation by $\exp \left(\frac{\sigma_{T i j}}{\epsilon \hat{\mu}_{m^{\prime}}} x\right)$ and integrating from $x_{i}$ to $x_{i+1}$ yield

$$
\begin{align*}
& \exp \left(\frac{\sigma_{T i j}}{\epsilon \hat{\mu}_{m^{\prime}}} x_{i+1}\right)\left(\hat{\psi}_{m j}\left(x_{i+1}\right)-\hat{\psi}_{m_{1} j}\left(x_{i+1}\right)\right)-\exp \left(\frac{\sigma_{T i j}}{\epsilon \hat{\mu}_{m^{\prime}}} x_{i}\right)\left(\hat{\psi}_{m j}\left(x_{i}\right)-\hat{\psi}_{m_{1} j}\left(x_{i}\right)\right) \\
= & \frac{\epsilon \hat{\mu}_{m^{\prime}}}{\sigma_{T i j}}\left(\exp \left(\frac{\sigma_{T i j}}{\hat{\mu}_{m^{\prime}} \epsilon} x_{i+1}\right)-\exp \left(\frac{\sigma_{T i j}}{\hat{\mu}_{m^{\prime} \epsilon}} x_{i}\right)\right) \\
& \times\left(\nu_{m_{1}} \frac{\check{\psi}_{m_{1} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}-\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}\right) \tag{3.25}
\end{align*}
$$

Because there are $4 M-2 \hat{M}$ of $\mu_{m}$ occur twice as discussed previously, we have $4 M-2 \hat{M}$ equations of the form (3.25). From (3.9b)-(3.11), the $2 \hat{M}$ relations between $\left\{\hat{\varphi}_{m j}\left(x_{i+1}\right), m \in \hat{V}\right\}$, $\left\{\hat{\varphi}_{m j}\left(x_{i}\right), m \in \hat{V}\right\}$ and $\left\{\hat{r}_{m i j}, m \in \hat{V}\right\}$ obtained by solving (3.10) are in fact relations among $\left\{\hat{\psi}_{m j}\left(x_{i+1}\right), m \in V\right\},\left\{\hat{\psi}_{m j}\left(x_{i}\right), m \in V\right\},\left\{\check{\psi}_{m i}\left(y_{i+1}\right), m \in V\right\}$ and $\left\{\check{\psi}_{m i}\left(y_{i}\right), m \in V\right\}$. Combing with (3.25), we have $4 M$ relations among $\left\{\hat{\psi}_{m j}\left(x_{i+1}\right), m \in V\right\},\left\{\hat{\psi}_{m j}\left(x_{i}\right), m \in V\right\}$, $\left\{\check{\psi}_{m i}\left(y_{i+1}\right), m \in V\right\}$ and $\left\{\check{\psi}_{m i}\left(y_{i}\right), m \in V\right\}$. Similarly, the other $4 M$ relations can be found through (3.7b) by the same process. By now, there are $8 M$ equations for these $16 M$ variables and these are in fact a special finite difference scheme on the rectangle $\left[x_{i}, x_{i+1}\right] \times\left[y_{j}, y_{j+1}\right]$.

In summary, this finite difference scheme is as follows:

When $\sigma_{a i j} \neq 0$ :

$$
\begin{align*}
& \exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i+1}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i+1}\right)-\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i}\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i}\right) \\
= & \frac{\epsilon}{\xi_{k} \sigma_{T i j}}\left(\left(\sigma_{T i j}-\epsilon^{2} \sigma_{a i j}\right) q_{i j}-\sum_{m \in \hat{V}} \hat{w}_{m} \hat{l}_{m}^{(k)} \hat{r}_{m i j}\right) \\
& \times\left(\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i+1}\right)-\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i}\right)\right), \quad \leq|k| \leq \hat{M} \tag{3.26a}
\end{align*}
$$

where

$$
\operatorname{hat} \varphi_{m j}=\left\{\begin{array}{lll}
\hat{\psi}_{m^{\prime} j}, & \mu_{m^{\prime}}=\hat{\mu}_{m}, & \mu_{m^{\prime}} \text { occurs once in } S_{\mu} \\
\frac{w_{m^{\prime}} \hat{\psi}_{m^{\prime} j}+w_{m_{1}} \hat{\psi}_{m_{1} j}}{w_{m^{\prime}}+w_{m_{1}}}, & \mu_{m^{\prime}}=\mu_{m_{1}}=\hat{\mu}_{m}, & \mu_{m^{\prime}} \text { occurs twice in } S_{\mu}
\end{array}\right.
$$

and

$$
\hat{r}_{m i j}=\left\{\begin{array}{c}
\nu_{m^{\prime}} \frac{\check{\psi}_{m^{\prime} i}\left(y_{j+1}\right)-\check{\psi}_{m^{\prime} i}\left(y_{j}\right)}{y_{j}+1-y_{j}}, \quad \mu_{m^{\prime}}=\hat{\mu}_{m}, \quad \mu_{m^{\prime}} \text { occurs once in } S_{\mu} \\
\frac{1}{w_{m^{\prime}}+w_{m_{1}}}\left(w_{m^{\prime}} \nu_{m^{\prime}} \frac{\check{\psi}_{m^{\prime} i}\left(y_{j+1}\right)-\check{\psi}_{m^{\prime} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}+w_{m_{1}} \nu_{m_{1}} \frac{\check{\psi}_{m_{1} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}\right), \\
\mu_{m^{\prime}}=\mu_{m_{1}}=\hat{\mu}_{m}, \quad \mu_{m} \text { occurs twice in } S_{\mu} .
\end{array}\right.
$$

When $\mu_{m}$ occurs twice in $S_{\mu}$, let $\mu_{m^{\prime}}=\mu_{m_{1}}=\hat{\mu}_{m}$,

$$
\begin{align*}
& \exp \left(\frac{\sigma_{T i j}}{\epsilon \hat{\mu}_{m}} x_{i+1}\right)\left(\hat{\psi}_{m^{\prime} j}\left(x_{i+1}\right)-\hat{\psi}_{m_{1} j}\left(x_{i+1}\right)\right)-\exp \left(\frac{\sigma_{T i j}}{\epsilon \hat{\mu}_{m}} x_{i}\right)\left(\hat{\psi}_{m^{\prime} j}\left(x_{i}\right)-\hat{\psi}_{m_{1} j}\left(x_{i}\right)\right) \\
= & \frac{\epsilon \hat{\mu}_{m}}{\sigma_{T i j}}\left(\exp \left(\frac{\sigma_{T i j}}{\hat{\mu}_{m} \epsilon} x_{i+1}\right)-\exp \left(\frac{\sigma_{T i j}}{\hat{\mu}_{m} \epsilon} x_{i}\right)\right) \\
& \times\left(\nu_{m_{1}} \frac{\check{\psi}_{m_{1} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}-\nu_{m^{\prime}} \frac{\check{\psi}_{m^{\prime} i}\left(y_{j+1}\right)-\check{\psi}_{m^{\prime} i}\left(y_{j}\right)}{y_{j+1}-y_{j}}\right) . \tag{3.26b}
\end{align*}
$$

The other $4 M$ relations according to (3.7b) are:

$$
\begin{align*}
& \exp \left(\eta_{k} \frac{\sigma_{T i j}}{\epsilon} y_{j+1}\right) \sum_{m \in \check{V}} \check{w}_{m} \check{\nu}_{m} \check{l}_{m}^{(k)} \check{\varphi}_{m i}\left(y_{j+1}\right)-\exp \left(\eta_{k} \frac{\sigma_{T i j}}{\epsilon} y_{j}\right) \sum_{m \in \check{V}} \check{w}_{m} \check{\nu}_{m} \check{l}_{m}^{(k)} \check{\varphi}_{m i}\left(y_{j}\right) \\
= & \frac{\epsilon}{\eta_{k} \sigma_{T i j}}\left(\left(\sigma_{T i j}-\epsilon^{2} \sigma_{a i j}\right) q_{i j}-\sum_{m \in \tilde{V}} \check{w}_{m} \check{l}_{m}^{(k)} \check{r}_{m i j}\right) \\
& \times\left(\exp \left(\eta_{k} \frac{\sigma_{T i j}}{\epsilon} y_{j+1}\right)-\exp \left(\eta_{k} \frac{\sigma_{T i j}}{\epsilon} y_{j}\right)\right), \quad 1 \leq|k| \leq \check{M} \tag{3.26c}
\end{align*}
$$

and for $\nu_{m^{\prime}}$ occurs twice in $S_{\nu}$, let $\nu_{m^{\prime}}=\nu_{m_{1}}=\check{\nu}_{m}$,

$$
\begin{align*}
& \exp \left(\frac{\sigma_{T i j}}{\epsilon \check{\nu}_{m}} y_{j+1}\right)\left(\check{\psi}_{m^{\prime} i}\left(y_{j+1}\right)-\check{\psi}_{m_{1} i}\left(y_{j+1}\right)\right)-\exp \left(\frac{\sigma_{T i j}}{\epsilon \check{\nu}_{m}} y_{j}\right)\left(\check{\psi}_{m^{\prime} i}\left(y_{j}\right)-\check{\psi}_{m_{1} i}\left(y_{j}\right)\right) \\
= & \frac{\epsilon \check{\nu}_{m}}{\sigma_{T i j}}\left(\exp \left(\frac{\sigma_{T i j}}{\check{\nu}_{m} \epsilon} y_{j+1}\right)-\exp \left(\frac{\sigma_{T i j}}{\check{\nu}_{m} \epsilon} y_{j}\right)\right) \\
& \times\left(\mu_{m_{1}} \frac{\hat{\psi}_{m_{1} j}\left(x_{i+1}\right)-\hat{\psi}_{m_{1} j}\left(x_{i}\right)}{x_{i+1}-x_{i}}-\mu_{m^{\prime}} \frac{\hat{\psi}_{m^{\prime} j}\left(x_{i+1}\right)-\hat{\psi}_{m^{\prime} j}\left(x_{i}\right)}{x_{i+1}-x_{i}}\right) . \tag{3.26d}
\end{align*}
$$

Here $S_{\nu}, \check{S}=\left\{\check{\nu}_{n}, \check{w}_{n} \mid n \in \check{V}\right\}, \check{M}, \check{V}, \check{\varphi}, \check{r}, \check{l}_{m}^{(k)}$ are defined parallel to $S_{\mu}, \hat{S}=\left\{\hat{\mu}_{n}, \hat{w}_{n} \mid n \in \hat{V}\right\}$, $\hat{M}, \hat{V}, \hat{\varphi}, \hat{r}, \hat{l}_{m}^{(k)}$ respectively. When $\sigma_{a i j}=0$, the scheme can be easily written down by
replacing (3.26a) for $1 \leq|k| \leq \hat{M}$ by (3.26a) for $1 \leq|k| \leq \hat{M}-1$ and (3.22)(3.23). So are similar replacements for (3.26c).

When $\epsilon$ is small, the exponential terms in (3.26) can be extremely large and overflow may happen. To handle this problem, we can divide both sides of (3.26) by some factors to keep the power of the exponential terms negative. For example, when $\xi_{k}>0$ in (3.26a), dividing both sides by $\exp \left(\xi_{k} \frac{\sigma_{T i j}}{\epsilon} x_{i+1}\right)$ gives

$$
\begin{align*}
& \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i+1}\right)-\exp \left(-\xi_{k} \frac{\sigma_{T i j}}{\epsilon}\left(x_{i+1}-x_{i}\right)\right) \sum_{m \in \hat{V}} \hat{w}_{m} \hat{\mu}_{m} \hat{l}_{m}^{(k)} \hat{\varphi}_{m j}\left(x_{i}\right) \\
= & \frac{\epsilon}{\xi_{k} \sigma_{T i j}}\left(\left(\sigma_{T i j}-\epsilon^{2} \sigma_{a i j}\right) q_{i j}-\sum_{m \in \hat{V}} \hat{w} \hat{l}_{m}^{(k)} \hat{r}_{m i j}\right) \\
& \times\left(1-\exp \left(-\xi_{k} \frac{\sigma_{T i j}}{\epsilon}\left(x_{i+1}-x_{i}\right)\right)\right), \quad \leq|k| \leq \hat{M} . \tag{3.27}
\end{align*}
$$

The exponential terms in (3.27) are bounded by 1. Similar treatments can also bound all the other exponential terms in (3.26).

Now consider the matrix form of this special finite difference scheme. Let

$$
\hat{\psi}_{j}=\left(\hat{\psi}_{-2 M j}, \cdots, \hat{\psi}_{-1 j}, \hat{\psi}_{1 j}, \cdots, \hat{\psi}_{2 M j}\right)^{T}, \quad \check{\psi}_{i}=\left(\hat{\psi}_{-2 M i}, \cdots, \hat{\psi}_{-1 i}, \hat{\psi}_{1 i}, \cdots, \hat{\psi}_{2 M i}\right)^{T} .
$$

Then (3.26) can be expressed into the matrix form:

$$
\begin{equation*}
\hat{A}_{i j}\binom{\hat{\psi}_{j}\left(x_{i}\right)}{\hat{\psi}_{j}\left(x_{i+1}\right)}+\check{A}_{i j}\binom{\check{\psi}_{i}\left(y_{j}\right)}{\check{\psi}_{i}\left(y_{j+1}\right)}=\mathbf{b}_{i j} \tag{3.28}
\end{equation*}
$$

where $\hat{A}_{i j}, \check{A}_{i j}$ are $8 M \times 8 M$ matrices and $\mathbf{b}_{i j}$ is a $8 M \times 1$ vector. We piece together the numerical solution with the neighboring cells using the interface conditions (2.13), i.e., two neighboring cells sharing a common $\hat{\psi}$ or $\check{\psi}$.

Specifically, we can arrange the unknowns as

$$
\psi=(\hat{\psi}, \check{\psi})^{T}
$$

where

$$
\begin{aligned}
& \hat{\psi}=\left(\hat{\psi}_{0}\left(x_{0}\right), \hat{\psi}_{0}\left(x_{1}\right), \cdots, \hat{\psi}_{0}\left(x_{N}\right), \hat{\psi}_{1}\left(x_{0}\right), \cdots, \hat{\psi}_{1}\left(x_{N}\right), \cdots, \hat{\psi}_{N^{\prime}-1}\left(x_{0}\right), \cdots, \hat{\psi}_{N^{\prime}-1}\left(x_{N}\right)\right)^{T} \\
& \check{\psi}=\left(\check{\psi}_{0}\left(y_{0}\right), \check{\psi}_{0}\left(y_{1}\right), \cdots, \check{\psi}_{0}\left(y_{N^{\prime}}\right), \check{\psi}_{1}\left(y_{0}\right), \cdots, \check{\psi}_{1}\left(y_{N^{\prime}}\right), \cdots, \check{\psi}_{N-1}\left(y_{0}\right), \cdots, \check{\psi}_{N-1}\left(y_{N^{\prime}}\right)\right)^{T}
\end{aligned}
$$

The coefficient matrix is

$$
A=(\hat{A}, \check{A})
$$

where

$$
\hat{A}=\left(\begin{array}{ccccc}
\hat{A}_{00} & 0_{8 M \times 4 M(N-1)} & & & \\
0_{8 M \times 4 M} & \hat{A}_{10} & 0_{8 M \times 4 M(N-2)} & & \\
0_{8 M \times 4 M(N-1)} & \hat{A}_{N-10} & & & \\
0_{8 M \times 4 M} & 0_{8 M \times 4 M N} & \hat{A}_{01} & & \\
& & \ddots & \ddots & \\
& & & \hat{A}_{0 N^{\prime}-1} & 0_{8 M \times 4 M(N-1)} \\
& \ddots & \ddots & & \\
& & & 0_{8 M \times 4 M(N-1)} & \hat{A}_{N-1 N^{\prime}-1}
\end{array}\right)
$$

which is a $8 M N N^{\prime} \times 4 M N^{\prime}(N+1)$ matrix, and

$$
\check{A}=\left(\begin{array}{ccccc}
\check{A}_{00} & 0_{8 M \times 4 M\left(N^{\prime}-1\right)} & \check{A}_{10} & & \\
& 0_{8 M \times 4 M\left(N^{\prime}-1\right)} & \ddots & & \\
& & & & \check{A}_{N-10} \\
0_{4 M} & \check{A}_{01} & 0_{8 M \times 4 M\left(N^{\prime}-2\right)} & & \\
& & \ddots & \ddots & \\
& & & & \\
0_{8 M \times 4 M\left(N^{\prime}-1\right)} & \check{A}_{0 N^{\prime}-1} & \ddots & & \\
& \ddots & & 0_{8 M \times 4 M\left(N^{\prime}-1\right)} \\
& & & \check{A}_{N-1 N^{\prime}-1}
\end{array}\right)
$$

which is a $8 M N N^{\prime} \times 4 M N\left(N^{\prime}+1\right)$ matrix. When

$$
\mathbf{b}=\left(\mathbf{b}_{00}, \mathbf{b}_{10}, \cdots, \mathbf{b}_{N-10}, \mathbf{b}_{01}, \cdots, \mathbf{b}_{0 N^{\prime}-1}, \cdots, \mathbf{b}_{N-1 N^{\prime}-1}\right)^{T}
$$

we have

$$
\begin{equation*}
A \psi=\mathbf{b} \tag{3.29}
\end{equation*}
$$

These are $8 M N N^{\prime}$ equations for $\psi$; together with the $4 M N+4 M N^{\prime}$ boundary conditions (3.8d), the $8 M N N^{\prime}+4 M N+4 M N^{\prime}$ unknowns in $\psi$ are determined.

Remark. By integrating both sides of (3.1) from $x_{i}$ to $x_{i+1}$ and $y_{j}$ to $y_{j+1}$ gives

$$
\begin{align*}
& \mu_{m} \frac{\hat{\psi}_{m j}\left(x_{i+1}\right)-\hat{\psi}_{m j}\left(x_{i}\right)}{x_{i+1}-x_{i}}+\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}+\frac{\sigma_{T i j}}{\epsilon} \tilde{\varphi}_{m i j} \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \tilde{\varphi}_{n i j}+\epsilon q_{i j}, \quad m \in V, \tag{3.30}
\end{align*}
$$

which gives the relation for $\hat{\psi}_{m j}, \check{\psi}_{m i}$ defined in (3.4), where

$$
\begin{equation*}
\tilde{\varphi}_{m i j}=\frac{1}{\left(x_{i+1}-x_{i}\right)\left(y_{j+1}-y_{j}\right)} \int_{y_{j}}^{y_{j+1}} \int_{x_{i}}^{x_{i+1}} \tilde{\psi}_{m} d x d y \tag{3.31}
\end{equation*}
$$

Consider the approximating system of Eq. (3.7). By integrating both sides of (3.7a) from $x_{i}$ to $x_{i+1}$ and dividing them by $x_{i+1}-x_{i}$, one gets

$$
\begin{align*}
& \mu_{m} \frac{\hat{\psi}_{m j}\left(x_{i+1}\right)-\hat{\psi}_{m j}\left(x_{i}\right)}{x_{i+1}-x_{i}}+\nu_{m} \frac{\check{\psi}_{m i}\left(y_{j+1}\right)-\check{\psi}_{m i}\left(y_{j}\right)}{y_{j+1}-y_{j}}+\frac{\sigma_{T i j}}{\epsilon} \varphi_{m i j} \\
= & \left(\frac{\sigma_{T i j}}{\epsilon}-\epsilon \sigma_{a i j}\right) \sum_{n \in V} w_{n} \varphi_{n i j}+\epsilon q_{i j} \tag{3.32}
\end{align*}
$$

with

$$
\varphi_{m i j}=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \hat{\psi}_{m j} d x
$$

Then we integrate both sides of $(3.7 \mathrm{~b})$ from $y_{j}$ to $y_{j+1}$, divide them by $y_{j+1}-y_{j}$, compare the obtained equation with (3.32) and get

$$
\varphi_{m i j}=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \hat{\psi}_{m j} d x=\frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \check{\psi}_{m i} d y
$$

This consists with the definition of $\hat{\psi}_{m j}, \tilde{\psi}_{m i}$ in (3.4) and $\varphi_{m i j}$ is an approximation of $\tilde{\varphi}_{m i j}$ in (3.31). Moreover, that (3.32) is the same as (3.30) implies the second approximation conserves the equation $\tilde{\varphi}_{m}$ satisfies.

## 4. Asymptotic Preserving Property

This method is to solve (3.7) exactly. When $\epsilon \ll 1, \Delta / \epsilon \gg 1$, in each rectangle $\left[x_{i}, x_{i+1}\right] \times$ [ $y_{j}, y_{j+1}$ ], substituting

$$
\hat{\psi}_{m j}=\sum_{n=0}^{\infty} \epsilon^{n} \hat{\psi}_{m j}^{(n)}, \quad \check{\psi}_{m i}=\sum_{n=0}^{\infty} \epsilon^{n} \check{\psi}_{m i}^{(n)}
$$

into (3.7) and equating the same order of $\epsilon$ give

$$
\begin{align*}
& \hat{\psi}_{m j}^{(0)}=\sum_{n \in V} w_{n} \hat{\psi}_{n j}^{(0)}, \quad \check{\psi}_{m i}^{(0)}=\sum_{n \in V} w_{n} \check{\psi}_{n i}^{(0)},  \tag{4.1}\\
& \hat{\psi}_{m j}^{(1)}=-\frac{\mu_{m}}{\sigma_{T i j}} \partial_{x} \hat{\psi}_{m j}^{(0)}+\sum_{n \in V} w_{n} \hat{\psi}_{n j}^{(1)}-\frac{\nu_{m}}{\sigma_{T i j}} \frac{\hat{\psi}_{m i}^{(0)}\left(y_{j+1}\right)-\hat{\psi}_{m i}^{(0)}\left(y_{j}\right)}{y_{j+1}-y_{j}},  \tag{4.2a}\\
& \check{\psi}_{m i}^{(1)}=-\frac{\nu_{m}}{\sigma_{T i j}} \partial_{y} \check{\psi}_{m i}^{(0)}+\sum_{n \in V} w_{n} \check{\psi}_{n i}^{(1)}-\frac{\mu_{m}}{\sigma_{T i j}} \frac{\check{\psi}_{m j}^{(0)}\left(x_{i+1}\right)-\check{\psi}_{m j}^{(0)}\left(x_{i}\right)}{x_{i+1}-x_{i}}, \tag{4.2~b}
\end{align*}
$$

and

$$
\begin{align*}
& \mu_{m} \partial_{x} \hat{\psi}_{m j}^{(1)}+\sigma_{T i j} \hat{\psi}_{m j}^{(2)} \\
= & \sigma_{T i j} \sum_{n \in V} w_{n} \hat{\psi}_{n j}^{(2)}-\sigma_{a i j} \sum_{n \in V} w_{n} \hat{\psi}_{n j}^{(0)}+q_{i j}-\nu_{m} \frac{\check{\psi}_{m i}^{(1)}\left(y_{j+1}\right)-\check{\psi}_{m i}^{(1)}\left(y_{j}\right)}{y_{j+1}-y_{j}},  \tag{4.3a}\\
& \nu_{m} \partial_{y} \check{\psi}_{m i}^{(1)}+\sigma_{T i j} \check{\psi}_{m i}^{(2)} \\
= & \sigma_{T i j} \sum_{n \in V} w_{n} \check{\psi}_{n j}^{(2)}-\sigma_{a i j} \sum_{n \in V} w_{n} \check{\psi}_{n j}^{(0)}+q_{i j}-\nu_{m} \frac{\hat{\psi}_{m j}^{(1)}\left(x_{i+1}\right)-\hat{\psi}_{m j}^{(1)}\left(x_{i}\right)}{x_{i+1}-x_{i}} . \tag{4.3b}
\end{align*}
$$

For both equations in (4.3), multiplying both sides by $w_{m}$, summing over $V$ and using (4.1), (4.2), (2.9), one gets

$$
\begin{align*}
& -\partial_{x}\left(\frac{1}{2 \sigma_{T i j}} \partial_{x} \hat{\psi}_{j}^{(0)}\right)+\sigma_{a i j} \hat{\psi}_{j}^{(0)}=\frac{1}{2 \sigma_{T i j}} \frac{\partial_{y} \check{\psi}_{i}^{(0)}\left(y_{j+1}\right)-\partial_{y} \check{\psi}_{i}^{(0)}\left(y_{j}\right)}{y_{j+1}-y_{j}}+q_{i j},  \tag{4.4a}\\
& -\partial_{y}\left(\frac{1}{2 \sigma_{T i j}} \partial_{y} \check{\psi}_{i}^{(0)}\right)+\sigma_{a i j} \check{\psi}_{i}^{(0)}=\frac{1}{2 \sigma_{T i j}} \frac{\partial_{x} \hat{\psi}_{j}^{(0)}\left(x_{i+1}\right)-\partial_{x} \hat{\psi}_{j}^{(0)}\left(x_{i}\right)}{x_{i+1}-x_{i}}+q_{i j} \tag{4.4b}
\end{align*}
$$

where

$$
\begin{equation*}
\hat{\psi}_{j}^{(0)}=\sum_{n \in V} w_{n} \hat{\psi}_{n j}^{(0)}, \quad \check{\psi}_{i}^{(0)}=\sum_{n \in V} w_{n} \check{\psi}_{n i}^{(0)} \tag{4.5}
\end{equation*}
$$

Note the interface condition (2.13), the definitions of $\hat{\psi}_{m j}, \check{\psi}_{m i}$ in (3.4) and $\hat{\psi}_{j}^{(0)}, \check{\psi}_{i}^{(0)}$ in (4.5). By using the interface analysis as discussed previously, we can get the connection conditions for $\hat{\psi}_{j}^{(0)}, \check{\psi}_{i}^{(0)}$, which are themselves and their first-order derivative divided by $\sigma_{T i j}$ are continuous at the grid lines. Now, for the boundary conditions of $\hat{\psi}_{j}^{(0)}$ or $\check{\psi}_{i}^{(0)}$, we have

$$
\begin{equation*}
\left.\hat{\psi}_{j}^{(0)}\right|_{x=0}=\psi_{L j},\left.\quad \hat{\psi}_{j}^{(0)}\right|_{x=a}=\psi_{R j},\left.\quad \check{\psi}_{i}^{(0)}\right|_{y=0}=\psi_{B i}, \quad \check{\psi}_{i}^{(0)}=\psi_{T i} \tag{4.6}
\end{equation*}
$$

Considering (2.10), firstly approximating the coefficients by piecewise constants like in (3.1)(3.2) and integrating both sides from $y_{j}$ to $y_{j+1}$ and from $x_{i}$ to $x_{i+1}$ respectively, we have in the rectangle $\left[x_{i}, x_{i+1}\right] \times\left[y_{j}, y_{j+1}\right]$,

$$
\begin{align*}
& -\partial_{x}\left(\frac{1}{2 \sigma_{T i j}} \partial_{x} \hat{\phi}_{j}\right)-\frac{1}{2 \sigma_{T i j}} \frac{\partial_{y} \phi\left(x, y_{j+1}\right)-\partial_{y} \phi\left(x, y_{j}\right)}{y_{j+1}-y_{j}}+\sigma_{a i j} \hat{\phi}_{j}=q_{i j},  \tag{4.7a}\\
& -\partial_{y}\left(\frac{1}{2 \sigma_{T i j}} \partial_{y} \check{\phi}_{i}\right)-\frac{1}{2 \sigma_{T i j}} \frac{\partial_{x} \phi\left(x_{i+1}, y\right)-\partial_{x} \phi\left(x_{i}, y\right)}{x_{i+1}-x_{i}}+\sigma_{a i j} \check{\phi}_{i}=q_{i j}, \tag{4.7b}
\end{align*}
$$

where

$$
\hat{\phi}_{j}=\frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \phi d y, \quad \check{\phi}_{i}=\frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \phi d x
$$

By approximating

$$
\begin{align*}
& \partial_{y} \phi\left(x, y_{j+1}\right) \approx \frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \partial_{y} \phi\left(x, y_{j+1}\right) d x=\partial_{y} \check{\phi}_{i}\left(y_{j+1}\right),  \tag{4.8a}\\
& \partial_{y} \phi\left(x, y_{j}\right) \approx \frac{1}{x_{i+1}-x_{i}} \int_{x_{i}}^{x_{i+1}} \partial_{y} \phi\left(x, y_{j}\right) d x=\partial_{y} \check{\phi}_{i}\left(y_{j}\right),  \tag{4.8b}\\
& \partial_{x} \phi\left(x_{i+1}, y\right) \approx \frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \partial_{x} \phi\left(x_{i+1}, y\right) d y=\partial_{x} \hat{\phi}_{j}\left(x_{i+1}\right),  \tag{4.8c}\\
& \partial_{x} \phi\left(x_{i}, y\right) \approx \frac{1}{y_{j+1}-y_{j}} \int_{y_{j}}^{y_{j+1}} \partial_{x} \phi\left(x_{i}, y\right) d y=\partial_{x} \hat{\phi}_{j}\left(x_{i}\right), \tag{4.8d}
\end{align*}
$$

solving (4.7) in each cell analytically and piecing together the neighboring cells using common $\hat{\phi}_{j}\left(x_{i}\right), \frac{1}{\sigma_{T i j}} \partial_{x} \hat{\phi}_{j}\left(x_{i}\right)$ or $\check{\phi}_{i}\left(y_{j}\right), \frac{1}{\sigma_{T i j}} \check{\phi}_{i}\left(y_{j}\right)$, we can obtain a discretization of (2.10) with interface conditions (2.14b). This is just what $\hat{\psi}_{j}^{(0)}, \breve{\psi}_{i}^{(0)}$ in (4.4) satisfy. Moreover, from (2.11), $\hat{\phi}$ and $\check{\phi}$ satisfy the same boundary conditions as for $\hat{\psi}_{j}^{(0)}, \check{\psi}_{i}^{(0)}$ in (4.6). Thus this scheme is AP.

## 5. Numerical Example

The performance of the scheme described above will be illustrated in this section. In fact we can find numerically that when there is no boundary layer (the incoming density is homogeneous), this method is first-order convergent uniformly with respect to the mean free path $\epsilon$.

The numerical results are presented for several problems in two-dimension and the computational domain for all the test problems is $[0,1] \times[0,1]$. There is no specified 'best' quadrature set in two-dimension and the problem we are considered here is only the space discretization. From (2.12), we use $M=2$ and $\theta_{m}=\pi / 6, \pi / 3$ when $\theta_{m} \in(0, \pi / 2)$, the eight $\mu_{m}, \nu_{m}$ now are

$$
\begin{array}{clll}
\mu_{1}=0.866, & \mu_{2}=0.5, & \mu_{3}=-0.5, & \mu_{4}=-0.866 \\
\mu_{-1}=-0.866, & \mu_{-2}=-0.5, & \mu_{-3}=0.5, & \mu_{-4}=-0.866 \\
\nu_{1}=0.5, & \nu_{2}=0.866, & \nu_{3}=0.866, & \nu_{4}=0.5 \\
\nu_{-1}=-0.5, & \nu_{-2}=-0.866, & \nu_{-3}=-0.866, & \nu_{-4}=-0.5
\end{array}
$$

For $m=1, \cdots, 8$, let $w_{m}=1 / 8$. We use $\Delta x=\Delta y=1 / 64$ to get the "exact" solutions for all the numerical examples. In the figures and tables below, the average density defined areby

$$
\rho(x, y)=\sum_{m \in V} w_{m} \psi_{m}(x, y)
$$



Fig. 5.1. Example 1. "Exact" solution calculated by our method with $\Delta x=\Delta y=1 / 64$ for $\epsilon=0.5$ is represented by the cell average along $x$-axis $\check{\rho}$.

Assume $\hat{\rho}_{i j}, \check{\rho}_{i j}$ are the average of $\rho$ along the vertical and horizonal edges of the cell $\left[x_{i}, x_{i+1}\right] \times$ $\left[y_{j}, y_{j+1}\right]$, respectively, to be obtained by our method. The errors in the tables are the relative error given by

$$
\max \left\{\frac{\max _{i, j}\left\{\left|\hat{\rho}_{i j}-\int_{y_{j}}^{y_{j+1}} \rho\left(x_{i}, y\right) d y\right|\right\}}{\|\rho\|_{L^{\infty}(\Omega)}}, \frac{\max _{i, j}\left\{\left|\check{\rho}_{i j}-\int_{x_{i}}^{x_{i+1}} \rho\left(x, y_{j}\right) d x\right|\right\}}{\|\rho\|_{L^{\infty}(\Omega)}}\right\},
$$

where $\rho(x, y)$ is the exact average density.
Example 1. Consider

$$
\begin{array}{ll}
\psi_{L}=1, & \psi_{R}=1, \quad \psi_{B}=1 / 2, \quad \psi_{T}=1 / 2 \\
\sigma_{T}=1, & \sigma_{a}=1, \quad q=1
\end{array}
$$

In this example the medium is homogeneous. The scattering coefficients and the source are constants in the computational domain and there are particles coming in from outside, which

Table 5.1: Example 1. The error between the "exact" solution and the numerical solutions computed by the proposed method with different $\Delta x, \Delta y$ and $\epsilon$.

| $\epsilon$ | $\Delta x$ | $\Delta y$ | $\\|$ Error $\\|_{\infty}$ | ratio |
| :---: | :---: | :---: | :---: | :---: |
| 0.5 | $2^{-2}$ | $2^{-2}$ | $8.017 * 10^{-3}$ | - |
| 0.5 | $2^{-3}$ | $2^{-3}$ | $4.035 * 10^{-3}$ | 2.0 |
| 0.5 | $2^{-4}$ | $2^{-4}$ | $1.712 * 10^{-3}$ | 2.4 |
| 0.1 | $2^{-2}$ | $2^{-2}$ | $1.966 * 10^{-2}$ | - |
| 0.1 | $2^{-3}$ | $2^{-3}$ | $6.119 * 10^{-3}$ | 3.2 |
| 0.1 | $2^{-4}$ | $2^{-4}$ | $3.456 * 10^{-3}$ | 1.8 |
| 0.02 | $2^{-2}$ | $2^{-2}$ | $3.523 * 10^{-2}$ | - |
| 0.02 | $2^{-3}$ | $2^{-3}$ | $8.739 * 10^{-3}$ | 4.0 |
| 0.02 | $2^{-4}$ | $2^{-4}$ | $2.681 * 10^{-3}$ | 3.3 |



Fig. 5.2. Example 2. "Exact" solution calculated by our method with $\Delta x=\Delta y=1 / 64$ for $\epsilon=0.1$ is represented by the cell average along the $y$-axis $\hat{\rho}$.
are isotropic in all directions. There is no boundary layer in this problem and the "exact" solution which is calculated by our method with $\Delta x=\Delta y=1 / 64$ for $\epsilon=0.5$ are represented in Fig 5.1. In Table 5.1, the relative error between the "exact" solution and the numerical solutions computed by the method with different $\Delta x, \Delta y$ for different $\epsilon$ are shown and the uniform first order convergence can be seen easily.

## Example 2. Consider

$$
\begin{align*}
& \psi_{L}=0, \quad \psi_{R}=0, \quad \psi_{B}=0, \quad \psi_{T}=0  \tag{5.1}\\
& \sigma_{T}(x, y)=1, \quad \sigma_{a}(x, y)=1, \quad q(x, y)=1, \quad(x, y) \in[0,1 / 2] \times[0,1] \\
& \sigma_{T}(x, y)=2, \quad \sigma_{a}(x, y)=2, \quad q(x, y)=0, \quad(x, y) \in(1 / 2,1] \times[0,1]
\end{align*}
$$

Table 5.2: Example 2. The error between the "exact" solution and the numerical solutions computed by the propsed method with different $\Delta x, \Delta y$ and $\epsilon$.

| $\epsilon$ | $\Delta x$ | $\Delta y$ | $\\|$ Error $\\|_{\infty}$ | ratio |
| :---: | :---: | :---: | :---: | :---: |
| 0.5 | $2^{-2}$ | $2^{-2}$ | $1.755 * 10^{-2}$ | - |
| 0.5 | $2^{-3}$ | $2^{-3}$ | $7.318 * 10^{-3}$ | 2.4 |
| 0.5 | $2^{-4}$ | $2^{-4}$ | $3.689 * 10^{-3}$ | 2.0 |
| 0.1 | $2^{-2}$ | $2^{-2}$ | $2.951 * 10^{-2}$ | - |
| 0.1 | $2^{-3}$ | $2^{-3}$ | $8.359 * 10^{-3}$ | 3.5 |
| 0.1 | $2^{-4}$ | $2^{-4}$ | $3.283 * 10^{-3}$ | 2.5 |
| 0.02 | $2^{-2}$ | $2^{-2}$ | $4.069 * 10^{-2}$ | - |
| 0.02 | $2^{-3}$ | $2^{-3}$ | $1.025 * 10^{-2}$ | 4.0 |
| 0.02 | $2^{-4}$ | $2^{-4}$ | $2.497 * 10^{-3}$ | 4.1 |

The coefficients are piecewise constants and there is an interface along the vertical line $x=1 / 2$. The "exact" solution for $\epsilon=0.1$ is presented in Fig 5.2 and the relative error between the "exact" solution and the numerical solutions computed by the method with different $\Delta x$, $\Delta y$ for different $\epsilon$ are shown in Table 5.2. We can see the uniform first-order convergence easily and when $\epsilon$ is small the convergence order is even higher. This fact can also be observed in Table 5.
Example 3. Consider

$$
\begin{align*}
& \psi_{L}=0, \quad \psi_{R}=0, \quad \psi_{B}=0, \quad \psi_{T}=0  \tag{5.2a}\\
& \sigma_{T}(x, y)=\frac{1}{2}+2 x+y^{2}, \quad \sigma_{a}(x, y)=1, \quad q(x, y)=x  \tag{5.2b}\\
& \quad \text { for }(x, y) \in[0,1 / 4) \times[0,1] \cup[1 / 4,3 / 4) \cup[0,1 / 2), \\
& \sigma_{T}(x, y)=2, \quad \sigma_{a}(x, y)=2 y, \quad q(x, y)=0,  \tag{5.2c}\\
& \quad \\
& \quad \text { for }(x, y) \in[1 / 4,3 / 4) \times[1 / 2,1] \cup[3 / 4,1] \cup[0,1] .
\end{align*}
$$

The coefficients in this problem depend on the space variables and are discontinuous at the interface. There is no incoming particles from outside, thus no boundary layer will exist. The shape of the interface is displayed on the top of Fig. 5.3, while the bottom one shows the "exact" solution. The relative errors for different $\epsilon$ are displayed in Table 5.3, from which we can find that the convergence rate is between 1 and 2 .

## Example 4.

$$
\begin{align*}
& \psi_{L m}=5 \mu_{m}, \quad \mu_{m}>0, \quad \psi_{R m}=-5 \mu_{m}, \quad \mu_{m}<0,  \tag{5.3a}\\
& \psi_{B m}=5 \nu_{m}, \quad \nu_{m}>0, \quad \psi_{T m}=-5 \nu_{m}, \quad \nu_{m}<0 .  \tag{5.3~b}\\
& \sigma_{T}(x, y)=1+x+2 y, \quad \sigma_{a}(x, y)=1 / 2+x, \quad q(x, y)=0 \quad \epsilon=0.02 \tag{5.3c}
\end{align*}
$$

Table 5.3: Example 3. The error between the "exact" solution and the numerical solutions computed by the proposed method with different $\Delta x, \Delta y$ and $\epsilon$.

| $\epsilon$ | $\Delta x$ | $\Delta y$ | $\\|$ Error $\\|_{\infty}$ | ratio |
| :---: | :---: | :---: | :---: | :---: |
| 0.5 | $2^{-2}$ | $2^{-2}$ | $4.604 * 10^{-2}$ | - |
| 0.5 | $2^{-3}$ | $2^{-3}$ | $1.581 * 10^{-2}$ | 2.9 |
| 0.5 | $2^{-4}$ | $2^{-4}$ | $5.107 * 10^{-3}$ | 3.1 |
| 0.1 | $2^{-2}$ | $2^{-2}$ | $2.173 * 10^{-2}$ | - |
| 0.1 | $2^{-3}$ | $2^{-3}$ | $8.269 * 10^{-3}$ | 2.6 |
| 0.1 | $2^{-4}$ | $2^{-4}$ | $3.454 * 10^{-3}$ | 2.4 |
| 0.02 | $2^{-2}$ | $2^{-2}$ | $3.019 * 10^{-2}$ | - |
| 0.02 | $2^{-3}$ | $2^{-3}$ | $1.015 * 10^{-2}$ | 3.0 |
| 0.02 | $2^{-4}$ | $2^{-4}$ | $2.587 * 10^{-3}$ | 3.9 |

Table 5.4: Example 4. The error between the "exact" solution and the numerical solutions computed by the proposed method with different $\Delta x, \Delta y$ and $\epsilon$.

| $\Delta x$ | $\Delta y$ | $\\|$ Error $\\|_{\infty}$ |
| :---: | :---: | :---: |
| $2^{-2}$ | $2^{-2}$ | $1.111 * 10^{-2}$ |
| $2^{-3}$ | $2^{-3}$ | $2.844 * 10^{-3}$ |
| $2^{-4}$ | $2^{-4}$ | $4.873 * 10^{-3}$ |



Fig. 5.3. Example 3. "Exact" solution calculated by our method with $\Delta x=\Delta y=1 / 64$ for $\epsilon=0.02$ is represented by the cell average along the $y$-axis $\hat{\rho}$. (a): the shape of the interface; (b): the exact solution

The incoming particle density are different in each direction so there are boundary layers in this problem. We can see the layers in Fig 5.4. Table 5.4 gives the relative numerical errors calculated with different $\Delta x$ and $\Delta y$. The error for $\Delta x=\Delta y=2^{-4}$ is bigger than that with $\Delta x=\Delta y=2^{-3}$. The reason is that $\psi(x, y)$ changes very fast at the boundary, and we can no longer use the average along the grid lines to approximate $\psi(x, y)$ at the grid line as we do in (3.6). In [15], the uniform second-order convergence up to the boundary is proved for one-dimensional piecewise constant coefficient approximation even when boundary layers exist. This two-dimensional method we present here does not have such good properties. Though we can no longer use coarse meshes to obtain good numerical approximations at the boundary layer, the error will decay exponentially away from the boundary layer.
One typical mesh useful for resolving the boundary layer is displayed in Figure 5.5, which uses fine mesh at the boundary and coarse mesh inside. As the relation (3.28) only depends


Fig. 5.4. Example 4. "Exact" solution calculated by our method with $\Delta x=\Delta y=1 / 64$ for $\epsilon=0.02$ is represented, (a): the cell average along $x$-axis $\check{\rho}$; (b): the cell average along $y$-axis $\hat{\rho}$.
on the cell itself, the coding is exactly the same. We can see in Figure 5.6 that satisfactory approximation is obtained. Similarly, for the transport diffusion coexisting case where interface layers exist, same strategy can be used.

We close this section by making two remarks. First, using the same idea as in [15] for the one-dimensional method we can prove rigorously the uniform first-order convergence with respect to $\epsilon$. In fact, the error given from approximating the coefficients by piecewise constants is uniformly second-order and the cell edge average approximation has only uniform first-order convergence. Second, we observed from the numerical examples that when $\epsilon$ is small enough, the convergent order is almost two, which implies the limit scheme (4.7)-(4.8) for the diffusion limit (2.10) may be of second-order. Using the idea in [9], we can also obtain the uniform first


Fig. 5.5. Example 4. A mesh for resolving the boundary layers: for $x, y \in[0,1 / 16] \cup[15 / 16,1]$, $\Delta x, \Delta y=2^{-7}$ and for $x, y \in(1 / 16,15 / 16), \Delta x, \Delta y=2^{-4}$.


Fig. 5.6. Example 4. Numerical solution that is calculated with the mesh displayed in Figure 5 for $\epsilon=0.02$. (a): the cell average along $x$-axis $\check{\rho} ;(\mathrm{b}):$ the cell average along $y$-axis $\hat{\rho}$.
order convergence by its asympototic preserving property.

## 6. Conclusion

A uniformly first-order convergent numerical method for the discete-ordinate transport equation in the rectangle geometry is presented in this paper. This method is an extension of the method proposed in [15] for one-dimensional discrete-ordinate transport equation. Firstly we approximate the scattering coefficients and source terms by piecewise constants determined by their cell averages. Then the solution at the cell edges is approximated by their average along the cell edges. We then solve analytically the system of equations for the cell edge averages in each cell and piece together the numerical solution with the neighboring cells using the interface conditions. When there is no interface or boundary layer, we prove that this method is asymptotic-preserving and its first-order accuracy with respect to the mean free path is shown numerically.

The convergence rate can be improved by interpolating the cell edge averages between neighboring cells. Moreover we can extend this method to other kinds of meshes like parallelogram. For problems with interface or boundary layers, the method we discuss here can not give good approximation using coarse mesh, which is different with the one-dimensional case. This issue will be further investigated in the future. We will also extend this method to some other related transport equations.

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