

# Visualization of Advection-Diffusion in Unsteady Fluid Flow

G. K. Karch, F. Sadlo, D. Weiskopf, C.-D. Munz, T. Ertl

## **Appendix**

This is supplementary material to the paper “Visualization of Advection-Diffusion in Unsteady Fluid Flow”. We provide a pseudo code for the GPU implementation of the method presented in the paper, and point out important implementation aspects.

---

**Input:**  $\bar{\phi}(t)$ , polynomial order  $o_p$

**Output:**  $\bar{\phi}(t + \delta t)$

- 1: **for** each dimension **do**
- 2:     **for all** thread blocks in parallel **do**
- 3:         **for all** threads in parallel **do** {thread corresponds to cell  $i$ }
- 4:              $\tilde{\phi}_i(\tau = t) = \bar{\phi}_i(t)$
- 5:              $v_{left} := -v(x_{i-\frac{1}{2}})$  if  $v(x_{i-\frac{1}{2}}) < 0$ , otherwise  $v_{left} := 0$
- 6:              $v_{right} := v(x_{i+\frac{1}{2}})$  if  $v(x_{i+\frac{1}{2}}) > 0$ , otherwise  $v_{right} := 0$
- 7:             **for**  $k = 0$  **to**  $o_p$  **do** {for each reconstruction polynomial}
- 8:                 **for**  $j = 1$  **to**  $o_p$  **do** {for each coefficient except  $w_0^k$ }
- 9:                     compute  $w_j^k$  from Eq. 14
- 10:                 **end for**
- 11:                  $\sigma_k = \sum_{m=0}^{o_p-1} (w_{m+1}^k \cdot \sum_{n=0}^{o_p-1} (\Sigma_{m,n} \cdot w_{n+1}^k))$
- 12:                  $\tilde{\omega}_k = \lambda_k / (\epsilon + \sigma_k)^4$  {Eq. 15 to 16}
- 13:                  $\omega_k := \text{normalize}(\tilde{\omega}_k)$
- 14:             **end for**
- 15:             compute  $\phi_i^{WENO}$  and its coefficients  $w_j^{i,WENO}$  from Eq. 15
- 16:             **if** active diffusion **then**
- 17:                 store  $w_j^{k=o_p/2}$  {coefficients from central polynomial}
- 18:             **end if**
- 19:              $\bar{f}_{i-}^l = v_{left} \cdot \phi_i^{WENO}(x_{i-\frac{1}{2}})$ ,  $\bar{f}_{i-}^r = v_{right} \cdot \phi_i^{WENO}(x_{i+\frac{1}{2}})$
- 20:             **if** active diffusion **then**
- 21:                  $\bar{d}_{i-}^l = D_\phi \sum_{j=1}^{o_p} (w_j^{k=o_p/2} \cdot (x_{i-\frac{1}{2}})^{j-1}) / 2$
- 22:                  $\bar{d}_{i-}^r = -D_\phi \sum_{j=1}^{o_p} (w_j^{k=o_p/2} \cdot (x_{i+\frac{1}{2}})^{j-1}) / 2$
- 23:             **else**
- 24:                  $\bar{d}_{i-}^l = \bar{d}_{i-}^r = 0$
- 25:             **end if**
- 26:             **for**  $j = 0$  **to**  $n - 1$  **do** {loop over prediction steps}
- 27:                  $\tilde{\phi}_i(\tau + \delta t/n) = \tilde{\phi}_i(\tau) + w_1^{i,WENO} \cdot (v_{i-\frac{1}{2}} + v_{i+\frac{1}{2}}) / 2 \cdot \delta t/n$  {Eq. 22}
- 28:                 perform steps 7 to 25
- 29:                 accumulate  $\bar{f}_{i-}^l$ ,  $\bar{f}_{i-}^r$ ,  $\bar{d}_{i-}^l$ , and  $\bar{d}_{i-}^r$
- 30:                  $\tau := \tau + \delta t/n$
- 31:             **end for**
- 32:              $\bar{f}_{i+} = \bar{f}_{(i+1)-}^l + \bar{f}_{(i-1)-}^r$ ,  $\bar{d}_{i+} = \bar{d}_{(i+1)-}^l + \bar{d}_{(i-1)-}^r$
- 33:              $\delta \bar{\phi}_i = (\bar{f}_{i+} - \bar{f}_{i-}^l - \bar{f}_{i-}^r + \bar{d}_{i+} - \bar{d}_{i-}^l - \bar{d}_{i-}^r) \delta t$
- 34:             **return**  $\bar{\phi}_i(t + \delta t) = \bar{\phi}_i(t) + \delta \bar{\phi}_i$
- 35:         **end for** all threads
- 36:     **end for** all thread blocks
- 37: **end for** each dimension

---

## Notes

- All indices are 0-based (e.g., first vector element is indexed with 0).
- The following variables are stored in shared memory (i.e., fast GPU memory at block scope):  $\bar{\phi}$ ,  $\tilde{\phi}$ ,  $\omega_k$ ,  $w_j^{i,WENO}$ ,  $\bar{f}_{i-}^l$  together with  $\bar{d}_{i-}^l$ , and  $\bar{f}_{i-}^r$  with  $\bar{d}_{i-}^r$ . Whereas we have to store fluxes at the right and left face separately, the fluxes at a given face (e.g.,  $\bar{f}_{i-}^r$  and  $\bar{d}_{i-}^r$ ) are combined.
- Line 5 and 6:  
Velocity  $v(x_{i-\frac{1}{2}})$  is trilinearly interpolated in space and linearly interpolated in time between  $v^{t_n}(x_{i-\frac{1}{2}})$  and  $v^{t_{n+1}}(x_{i-\frac{1}{2}})$ , where  $v^{t_n}$  and  $v^{t_{n+1}}$  ( $t_n \leq t$ ,  $t + \delta t \leq t_{n+1}$ ) are two consecutive steps of the simulated velocity field, stored in 3D texture.  $v^t(x_{i+\frac{1}{2}})$  is computed accordingly. For each cell we consider only outflow velocity directly, i.e., velocity which is used for calculation of  $\bar{f}_{i-}^l$  and  $\bar{f}_{i-}^r$ . The upper-script  $l$  and  $r$  are used to differentiate between fluxes on the right and left face—see comment on lines 19 and 21.
- Line 9:  
The matrices  $\mathbf{L}_k$  (one for each reconstruction polynomial  $\phi_k$ ),  $dim(\mathbf{L}_k) = (o_p + 1, o_p + 1)$  are stored in the GPU constant memory. In order to compute coefficients  $w_j^k$  at a given cell, values  $\phi$  from neighboring cells must be available for each thread in a block. These values are therefore stored in shared memory. The coefficient  $w_0^k$  is not needed as it does not influence the oscillation indicator  $\sigma$ .
- Line 11:  
The oscillation matrix  $\Sigma$ ,  $dim(\Sigma) = (o_p, o_p)$  is precomputed and stored in GPU constant memory.
- Line 19 and 21:  
Only fluxes that go outside a given cell  $i$  are computed explicitly.  $\bar{f}_{i+} + \bar{d}_{i+}$  are evaluated from  $\bar{f}_{(i+1)-}^l + \bar{d}_{(i+1)-}^l$  and  $\bar{f}_{(i-1)-}^r + \bar{d}_{(i-1)-}^r$ , that is, from outflux at left face from cell  $i + 1$  and outflux at right face from cell  $i - 1$ . In order to access these values from a given cell  $i$  they are stored in GPU shared memory.