Exploiting timescale separation in micro and nano flows

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Abstract In this paper we describe how timescale separation in micro/nano flows can be exploited for computational acceleration. A modified version of the seamless heterogenous multiscale method (SHMM) is proposed: a multi-step SHMM. This maintains the main advantages of SHMM (e.g., re-initialisation of micro data is not required; temporal gearing (computational speed-up) is easily controlled; and it is applicable to full and intermediate degrees of timescale separation) while improving on accuracy and greatly reducing the number of macroscopic computations and micro/macro coupling instances required. The improved accuracy of the multi-step SHMM is demonstrated for two canonical one-dimensional transient flows (oscillatory Poiseuille and oscillatory Couette flow) and for rarefied-gas oscillatory Poiseuille flow.

Keywords: Heterogenous Multiscale Method (HMM); seamless HMM; rarefied oscillatory Poiseuille flow.

1. Introduction

Advances in micro and nano technologies are presenting new challenges for engineering science. In fluid dynamics, the number of flow systems that need an appreciation of the multiscale physics involved is increasing substantially. Modelling flows with a low degree of scale separation requires accounting for microscopic effects in the macroscopic behaviour: micro- and millisecond effects are important in micro and nano flows, but depend on the outcome of pico- or nanosecond molecular processes. The design of future technologies that exploit micro and nano scale flow components will therefore require the ability to resolve phenomena across scales of at least 8 orders of magnitude in space, and 10 orders of magnitude in time — this presents a formidable multiscale problem.

Any fluid flow could in principle be simulated by employing a suitable microscopic model over the entire flow domain. While this is practicable when studying the flows in a carbon nanotube, a small group of proteins in solution, or other very small-scale systems, in engineering problems the simulation domain is often much larger and more complex, making this approach computationally impractical. In many cases, the degree of scale separation may vary across the flowfield, as well as with time, and using a model of microscopic interactions in scale-separated regions is unnecessary because perfectly adequate macroscale models exist, in particular, the traditional Navier-Stokes-Fourier equations of fluid dynamics.

A more rational physical way of modelling the flow is to couple a macroscale model to a microscale model. The most popular of these hybrid approaches is domain decomposition, in which micro models are used in part of the computational domain and macro models on the remainder. But a generalised hybrid framework has been proposed that exploits scale separation, both in time and space, wherever it occurs in the domain in order to make computational efficiencies (the reader is referred to E et al. (2007), for an in-depth review). In this Heterogeneous Multiscale Method (HMM) the micro model is decoupled from the physical domain of the macro solver, and there is no direct communication between the different micro simulations. No a priori knowledge of the macro form of the fluid constitutive and boundary behaviour is needed: the micro models are used to provide this missing flow property data at different spatial locations. Each local micro simulation is in turn constrained by the local output of the macro solver.

The strategy for exploiting timescale separation in HMM is illustrated in Figure 1a. A macroscopic time step is set, $\Delta t$, which is small enough to adequately resolve macroscopic variations, but which is much larger than the microscopic time step, $\delta t$. 

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For each macroscopic time step there are performed multiple microscopic time steps, such that the time evolved microscopically is roughly equal to the microscopic relaxation time, $T_{\text{micro}}$; more than this, and computational effort is wasted.

A major improvement on this method has recently been proposed by E et al. (2009): the Seamless Heterogeneous Multiscale Method (SHMM). Here, the micro simulation does not need to be continually re-initialised between macro timesteps, which can be costly: the interaction between the micro and macro solvers is continuous (seamless), as illustrated in Figure 1b. SHMM exploits temporal scale separation in the given problem by running the macro and micro solvers on different clocks ($t$ and $\tau$, respectively). The computational time steps, although coupled as if they were of equivalent times, are asynchronous. The micro model is not reset at every coupling, instead it dynamically adapts to changes in the macro environment (manifested through the constraints applied from the macro solver). Implicit time-averaging of the micro data therefore replaces the explicit time-averaging of most other hybrid methods. This asynchronous coupling has the effect of “fast-forwarding” the macro constraints (from the perspective of the micro-solver); in timescale-separated conditions this produces a good approximate response to the macro constraints being applied in real time, provided the SHMM macro time step is sufficiently small. A further advantage of the SHMM approach is that it allows control over how aggressively timescale separation is exploited for computational efficiency (at the expense of error). The ratio between the scales of the micro and macro clocks (which we refer to in this paper as the temporal gearing) directly determines the computational saving relative to a simulation that is performed entirely using the micro model; in SHMM, the gearing can be set explicitly.

A drawback of SHMM is that in regions of high timescale separation, many more macroscopic time step calculations, and transfers of data, are performed than are necessary in HMM; these additional macroscopic computations may not always be negligible in cost. Furthermore, in such conditions, the HMM approach is more accurate, because at individual macro instants the HMM micro solution is fully relaxed to the actual macroscopic constraint and not a time-averaged one, as in SHMM.

In this paper we propose a modified time-stepping strategy for exploiting timescale separation: a multi-step SHMM. This approach adopts the same asynchronous coupling of two simulation clocks, and the continuous solution of the micro-solver (i.e. no re-initialisation is required), but allows the number of micro time steps per coupling instance, $N$, to be greater than 1; see Figure 1(c). Figure 2 shows the sequence of solution for each coupling instance. The multi-step number, $N$, provides an additional control, which can be increased in regions of high timescale separation (where $t \gg \tau$) to obtain maximum accuracy and to lower the macroscopic/coupling overhead, and reduced in regions of low timescale separation ($t \rightarrow \tau$) to approach a full microscale simulation. The multi-step approach can thus be viewed as a hybrid of HMM and SHMM, combining desirable features of both: simple control of the computational gearing (saving over a full micro simulation); no requirement for reinitialisation of the micro simulation; highest accuracy over the whole range of timescale separation; and fewest macroscopic computations and coupling instances performed in fully scale-separated conditions.

Figure 1: Alternative multi-scale time-stepping approaches: a) HMM; b) SHMM; and c) the Multi-Step SHMM

2. Time-stepping with timescale separation

In this paper we define a dimensionless timescale-separation number, $S$, as:

$$S = \frac{T_{\text{macro}}}{T_{\text{micro}}}, \quad (1)$$

For each macroscopic time step there are performed multiple microscopic time steps, such that the time evolved microscopically is roughly equal to the microscopic relaxation time, $T_{\text{micro}}$; more than this, and computational effort is wasted.
Figure 2: Sequence of solution for the multi-step SHMM, for a single coupling instance. (1) micro data compressed to macro model; (2) macro solver advancement; (3) constraints applied to micro model; (4) micro model advanced $N$ steps.

where $T_{\text{macro}}$ is a macroscopic time scale and $T_{\text{micro}}$ is a characteristic microscopic relaxation period. For $S \gg 1$, the micro and macro scales are purely separated; for $S = 1$ there is no separation (i.e. the micro and macro scales act over the same time frame); a “mixed” or transition regime between the two extremes is here defined as $1 < S < 10$. The exact division between the mixed and pure-separation regimes is arbitrary, in the same way as are the divisions of rarefaction regimes in terms of Knudsen number (e.g. slip, transition, free-molecular).

Here we define the temporal gearing of a hybrid approach, $g$, as the macroscopic time evolved per unit of microscopic time evolved. If the macro solver costs are negligible, this is equivalent to the computational saving compared to a full micro simulation. Clearly, to exploit scale separation fully, at higher timescale-separation numbers the temporal gearing should increase.

The temporal gearing in HMM is the ratio: $T_{\text{macro}}/T_{\text{micro}}$, i.e., $g = S$. Depending, though, on the error that is acceptable or the accuracy desired, this particular relationship between temporal gearing and timescale separation may not be appropriate. In this respect, the SHMM (and also the proposed multi-step SHMM) has an advantage. Here, the temporal gearing is simply the ratio of the two clock time scales, $g = t/\tau$. This means that any functional dependence of the temporal gearing on the timescale separation number can be implemented.

In this paper, we choose a straightforward functional relationship between temporal gearing and timescale separation, as follows: when there is no scale separation, the macro clock should approach the micro clock (i.e. $g = 1$ at $S \leq 1$); when there is pure scale separation ($S \geq 10$), the temporal gearing should increase proportionally with the degree of scale separation, (i.e., $g \propto S$):

$$g(S) = \begin{cases} 1, & \text{for } S \leq 1, \\ kS, & \text{for } S \geq 10, \end{cases} \quad (2)$$

where an increase in the constant of proportionality $k$ increases computational saving, at the expense of accuracy. For values of $S$ in the “mixed” regime ($1 < S < 10$) we choose the minimum-order polynomial that enforces continuity of value and slope of $g$ at $S=1$ and $S=10$:

$$g(S) = a_1 S^3 + a_2 S^2 + a_3 S + a_4$$

for $1 < S < 10$,

$$\quad (3)$$

where the coefficients $a_n$ depend on the specific value of $k$. Figure 3 shows a plot of temporal gearing $g$ against $S$, for $k=0.25$ (note, this is a more conservative temporal gearing as compared to HMM, which implies $k = 1$).

Figure 3: Temporal gearing, $g$, against the timescale-separation number, $S$; $k=0.25$.

2.1. Multi-step SHMM

For high degrees of temporal scale separation ($S \gg 1$), the microscopic state is relaxed to the macroscopic state in a quasi-steady manner. This means, the closer the micro solution is to its locally (constrained) relaxed state the more accurate the result. In this scenario, for a given macro time ($T$), it is more accurate to have $M$ micro steps constrained to the one macro state at $T$ (as it is in HMM), than it is to have a converged micro state over a range of $M$-previous macro states (as is the case in SHMM).

On the other hand, when $S$ is in the mixed regime, having the micro state converged over a range of previous macro states (as in SHMM), reflects the true micro state more accurately, as it is not locally relaxed at each macro instant (as in HMM).
The multi-step SHMM we propose seeks to combine the best of both approaches over the full range of $S$. As per SHMM, the micro time step, $\Delta t$, is held constant, and is set by stability/accuracy considerations. In SHMM the macro time step can be set directly by the gearing desired (as determined by equations (2)-(3), for example):

$$\Delta' t = \Delta t \cdot g,$$

(4)

Here, though, differently to SHMM, we allow the number of micro steps per coupling instance, $N$, to be greater than 1 (as illustrated in Figure 1c). The macro time step is then set according to:

$$\Delta'' t = \Delta t \cdot g \cdot N,$$

(5)

under the proviso that it does not exceed the maximum macro time step allowable $\Delta_{\text{max}}$ (based on stability/accuracy considerations).

The most accurate/efficient number of multisteps, $N$, will depend upon the temporal scale separation, $S$. Given that for pure timescale separation, a larger $N$ provides a more accurate solution (as discussed above), we set $N$ to be maximum for $S \geq 10$. For non-timescale-separated conditions ($S \leq 1$), the macro and micro time steps should be synchronous, (i.e. $N=1$). For values of $S$ in the “mixed” regime ($1 < S < 10$), we choose the minimum-order polynomial that enforces continuity of value and slope of $N$ at $S=1$ and $S=10$:

$$N(S) = \begin{cases} 
1, & \text{for } S \leq 1, \\
b_1S^3 + b_2S^2 + b_3S + b_4, & \text{for } 1 < S < 10, \\
N_{\text{max}}, & \text{for } S \geq 10,
\end{cases}$$

(6)

where

$$N_{\text{max}} = \frac{\Delta_{\text{max}}}{\Delta t \cdot g},$$

(7)

and the coefficients $b_n$ are dependent on $N_{\text{max}}$.

3. Test cases: transient one-dimensional shear flows

In timescale-separated simulations there are three sources of error: error due to the micro solution; error due to the macro solution; and error due to the time-step coupling scheme (e.g. SHMM). We restrict our attention to the latter.

The test cases in this paper are time-dependent flows between parallel plates. Here, the macro model calculates the pressure gradient or the channel-wall velocity, and the micro model calculates the spatial distribution of streamwise velocity (parallel to the walls). For these cases, the macro calculation is trivial, exact and known a priori. This is convenient, as it allows us to eliminate macro solver error in our evaluation of the alternative time-stepping schemes, and since there is no error in the constraints applied to the micro solver, all error is either from the micro model (which is deliberately kept negligibly small) or error due to the temporal coupling scheme. In all simulations we specify a maximum macro time step $\Delta_{\text{max}} = T_{\text{macro}}/5$, which in practice would be set by macroscopic stability/accuracy requirements.

3.1. Oscillatory Poiseuille Flow

The flow between two parallel plates resulting from an oscillating pressure gradient is a canonical transient case in fluid dynamics. If this is a locally near-thermodynamic equilibrium flow \footnote{Note, timescale separation can be independent, as it is here, of the state of local thermodynamic equilibrium} an appropriate micro model is the one-dimensional Navier-Stokes momentum equation:

$$\frac{\partial u}{\partial t} = v \frac{\partial^2 u}{\partial y^2} + \Pi(t),$$

(8)

where $v$ is the kinematic viscosity, $y$ is a direction perpendicular to the walls, and $u$ is the flow-velocity component parallel to the wall. The micro model is constrained through the time-dependent streamwise pressure gradient, $\Pi(t)$. The characteristic microscopic time scale, $T_{\text{micro}}$ is taken as the startup time: the time taken, after a suddenly imposed and constantly applied pressure gradient, for the mean velocity to reach 95% of its steady-state value.

The macro model solver, in this example, is the model that calculates the pressure gradient, simply:

$$\Pi = \hat{\Pi} \sin(\omega t),$$

(9)

where $\omega$ is the radial frequency of oscillation and $t$ is the macroscopic time. The characteristic macroscopic time scale is chosen to be the quarter-wave period:

$$T_{\text{macro}} = \frac{\pi}{2\omega},$$

(10)

and thus the timescale separation number at a particular frequency is:

$$S = \frac{T_{\text{macro}}}{T_{\text{micro}}} = \frac{2\omega T_{\text{micro}}}{\pi}.$$
We consider a range of oscillation frequencies, spanning timescale separation numbers from $S=1$ to $50$. The temporal gearing for a given $S$ is as described by equations (2) and (3), and as shown in Figure 3 ($k=0.25$). For the multi-step SHMM the number of micro time steps per coupling instance ($N$) is calculated using equation (6); $N$ is plotted in Figure 4 as a function of $S$. In evaluating equation (6) and (7), we take $\Delta t_{\text{max}} = T_{\text{macro}}/5$.

![Figure 4](image)

**Figure 4:** Number of micro timesteps per coupling instance, $N$, against the timescale separation number, $S$.

For evaluating the accuracy of SHMM and the multi-step SHMM, a classical analytical solution to the full problem (equations 8 and 9) is used:

$$u(y) = \Re \left\{ e^{i \omega t} \hat{U}_{\text{wall}} \right\},$$  

where

$$A_1(y, \eta) = \left( 1 - \frac{\cosh[\eta(y-h/2)(i+1)]}{\cosh[\eta h/(i+2)]} \right),$$  

and $\eta = \sqrt{i \omega / \nu}$, with $h$ the channel height.

Figure 5 compares the accuracy of the SHMM and the multi-step SHMM. For all values of $S$ considered, the multi-step method is more accurate (for the same gearing); for $S \gg 1$ the multi-step approach is over $8 \times$ more accurate. Furthermore, the multi-step HMM involves approximately $1600 \times$ fewer macroscopic calculations and instances of coupling (though, in this illustrative example, the macroscopic calculation has negligible cost). Note, direct comparison with the conventional HMM cannot be made, since the temporal gearing of HMM differs from that used in Figure 5. However, for indicative purposes, for HMM (with $g = S$), the error for $S \geq 5$ is approximately $5\%$; the error is greater for $S < 5$.

![Figure 5](image)

**Figure 5:** Maximum error ($\%$) (over time) of mean velocity (over $y$) against $S$. S-HMM (dashed line) against multi-step S-HMM (solid line).

### 3.2. Oscillatory Couette Flow

Another canonical transient flow in fluid dynamics is that between two plates generated by the oscillatory transverse motion of one wall relative to the other. In this case, the macro model provides the constraint on the micro solution in the form of an imposed wall velocity, i.e.:

$$u(0) = \hat{U}_{\text{wall}} \sin(\omega t),$$  

where $\hat{U}_{\text{wall}}$ is the amplitude of the moving wall (at $y=0$). For evaluating the accuracy of SHMM and the multi-step SHMM, a classical analytical solution to the full problem (equations 8 and 14) is used:

$$u(y) = \Im \left\{ e^{i \omega t} \hat{U}_{\text{wall}} A_2(y, \xi) \right\},$$  

where

$$A_2(y, \xi) = \left[ \cosh(\xi y) - \coth(\xi h) \cosh(\xi y) \right],$$  

and $\xi = \sqrt{i \omega / \nu}$. Figure 6 compares the accuracy of the two methods. Again, for all values of $S$ considered, the multi-step method is more accurate (for the same gearing); for $S \gg 1$ the multi-step approach is over $8 \times$ more accurate, and this is achieved using over a thousand times fewer macroscopic calculations and coupling instances than in the SHMM.

### 3.3. Non-equilibrium Oscillatory Poiseuille Flow

At atmospheric pressures, gas flow through micro/nano-channels can be highly non-equilibrium (Reese et al., 2003). The final example considered here, is isothermal oscillatory Poiseuille flow...
Figure 6: Maximum error (%) (over time) of mean velocity (over y) against $S$. S-HMM (dashed line) against multi-step S-HMM (solid line).

(as in §3.1) of a monatomic gas in rarefied/non-equilibrium conditions.

The degree of departure from local thermodynamic equilibrium in a gas (the state of rarefaction), is expressed by the Knudsen number:

$$Kn = \frac{\lambda}{h},$$  

where $\lambda$ is the mean distance between molecular collisions, and $h$ is the characteristic macroscopic length scale; here, this is the channel height.

The micro model chosen here is the BGK Boltzmann equation:

$$\frac{\partial f}{\partial t} + \hat{v} \frac{\partial f}{\partial y} + \frac{\partial f}{\partial \hat{u}} F_x(t) = -\frac{(f - f_{eq})}{\xi}$$  

(18)

where $f$ is the molecular distribution function, $f_{eq}$ is the equilibrium distribution, $\xi$ is the molecular relaxation time, $F_x(t)$ is the time-dependent body force (equivalent to an oscillatory pressure gradient) and $\hat{u}, \hat{v}$ are the $x$- and $y$-components of molecular velocity, respectively. The molecular relaxation time for the BGK model is given by $\xi = \mu/p$, where $\mu$ is the dynamic viscosity and $p$ is the gas pressure.

We obtained solutions for the micro model (18) using a discrete velocity method (DVM) similar to that used by Valougeorgis (1988). This numerical scheme, which uses Gaussian quadrature to integrate in velocity space, has been tested extensively against a host of problems in rarefied gas dynamics and has proved to be both accurate and highly efficient (see, e.g., Naris and Valougeorgis, 2005; Valougeorgis, 1988; Valougeorgis and Naris, 2003).

As in previous sections, the macro model solver for the oscillatory driver is trivial. Simply:

$$F_x(t) = \hat{F}_x \sin(\omega t).$$  

(19)

All parameters relating to the SHMM and multi-step SHMM are identical to those in §3.1.

Figure 7 shows the time evolution of mass flow rate resulting from the oscillatory forcing (19) over a selected period of the cycle; for this simulation $Kn = 1$ and the timescale separation $S=20$. The mass flow rate is normalised with the Navier-Stokes prediction of mass-flow-rate amplitude (notice that the mass-flow-rate amplitude predicted by the full BGK solution is over five times that of the Navier-Stokes equations). The SHMM is compared to the multi-step SHMM; both solutions produce a good prediction with a temporal gearing of five (in this case, effectively a factor of five computational saving over the full BGK solution). However, the multi-step SHMM is still significantly closer to the full BGK solution, both in terms of amplitude and phase. As was the case in section §3.1, the multistep SHMM obtains this better prediction, with far fewer macroscopic calculations and coupling instances. In this case, where $S \gg 1$, an equivalent HMM solution would be as accurate as the multi-step SHMM (for a gearing of $g=S$).

Figure 8 shows the time evolution of mass flow rate resulting from a higher frequency forcing; for this simulation $Kn = 1$ and the timescale separation $S=1.5$. In this case, both the SHMM and multi-scale SHMM are barely distinguishable from the full BGK solution. This is to be expected, as the temporal gearing in this case is $g \approx 1$. A HMM solution is
also plotted to demonstrate that, clearly, fully relaxing the micro solution to the macroscopic constraints is inaccurate at low degrees of scale separation.

4. Summary and Conclusions

A multi-step version of the seamless heterogeneous multiscale method (E et al., 2009) has been proposed. It shares the main advantages of SHMM over HMM, e.g.:

1. the micro-solver does not require re-initialisation at every macro time step;
2. the temporal gearing can be controlled in a straightforward manner by adjusting the ratio of the micro and macro clocks;
3. it can be used in both timescale-separated and non-timescale-separated conditions, thus appropriate for an adaptive scheme with time-varying scale separation;

while retaining the accuracy and macroscopic efficiency of HMM:

1. when timescales are purely separated, multi-step SHMM provides a more accurate result;
2. and in such conditions, far fewer coupling instances and macroscopic calculations are required.

The multi-step scheme can thus be viewed as an amalgam of HMM and SHMM: a combination of the best features of both methods.

We have presented results from three transient shear flow problems. In each case, the macro model computation is trivial (i.e. the evaluation of a sine function). This has allowed us to focus on error arising from the particular time-stepping scheme. Two micro models have been considered: a Navier-Stokes micro model and a Boltzmann BGK micro model. In each of the three examples, the multi-step SHMM demonstrated significantly greater accuracy than the SHMM, and at a fraction of the number of coupling instances and macroscopic computations.

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