Accelerating Large-scale Phase Field Simulation with GPU

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Background



Micro-structures in Materials



Micro-structures: meso-scale morphological patterns





Micro-structures in Materials



Fatigue Failure











相场模型

Given a free energy functional $E(\phi)$, the gradient flow in L^2 :

 $\frac{\partial \phi}{\partial t} = -\frac{\partial E(\phi)}{\partial \phi};$

or the gradient flow in H^{-1} :

 $\frac{\partial \phi}{\partial t} = \Delta \frac{\delta E(\phi)}{\delta \phi}.$

It is easy to see that they satisfy, respectively the following energy laws:

$$\frac{d}{dt}E(\phi) = -\|\frac{\delta E(\phi)}{\delta \phi}\|^2,$$

 $\frac{d}{dt}E(\phi) = -\|\nabla \frac{\delta E(\phi)}{\delta \phi}\|^2.$

or





Phase Field Model



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or

$$\frac{d}{dt}E(\phi) = -\|\nabla\frac{\delta E(\phi)}{\delta\phi}\|^2$$



Explicit time marching- small time step



Allen-Cahn (AC) equation



Takashi Shimokawabe et. al. , SC2011. 4×10^9 cells, TSUBAME 2.0.



Martin Bauer et. al., SC2015. 8×10^9 cells. SuperMUC, Hornet and JUQUEEN.



Tomohiro Takaki et. al. Acta Materialia, 2016. 4 ×10⁹ cells, TSUBAME 2.5.

Energy stability



To fix the idea, we let $E(\phi) = \int_{\Omega} [\frac{1}{2} |\nabla \phi|^2 + \frac{1}{\eta^2} F(\phi)] dx$, where $F(\phi)$ is a general nonlinear free energy, and consider the gradient flow in H^{-1} :

$$\phi_t = \nabla \cdot \nabla \frac{\delta E}{\delta \phi}, \qquad \partial_n w|_{\partial \Omega} = 0;$$
 $w = \frac{\delta E}{\delta \phi} = -\Delta \phi \qquad + \frac{1}{\eta^2} f(\phi), \qquad \partial_n \phi|_{\partial \Omega} = 0,$

which satisfies the energy law:

$$\frac{\partial}{\partial t} \int_{\Omega} \left(\frac{1}{2} |\nabla \phi|^2 + \frac{1}{\eta^2} F(\phi) \right) = - \int_{\Omega} |\nabla (-\Delta \phi + \frac{1}{\eta^2} f(\phi))|^2.$$

Goal: Design simple, efficient and accurate numerical schemes that satisfy a discrete energy law.

Large Scale Phase Field Simulations



AC equation, explicit time marching

Small time step-size Integration scheme design, easy Stencil computing performance ~ 25% peak

Large scale simulation ~10 billion cells

CH equation, implicit time marching

Large time step-size Integration scheme design, hard Multi-level preconditioner-solver performance < 10% peak

Large scale simulation ~ 0.1 billion cells

The limited resolution in 3D simulations(CH) constitutes bottlenecks in validating predictions based on the phase field approach.

Accurate large-time-step marching scheme, scalability, efficiency

Compute intensive large time step algorithm





$$u_t = Lu + N(u, t).$$



Stable large time step-size

Exact integration & proper splitting of L and N High order accuracy

Multi-step, prediction-correction, Runge-Kutta

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Second order ETD scheme



$$\begin{cases} \frac{\partial u}{\partial t} = \nabla \cdot \left(M(u) \nabla (-\gamma \Delta u + u^3 - u) \right), & \mathbf{x} \in \Omega, t \in [t_0, t_0 + T], \\ u|_{t=t_0} = u_0, & \mathbf{x} \in \Omega, \end{cases}$$

$$\frac{\partial u}{\partial t} = -C(\gamma \Delta^2 u) - \nabla \cdot \left((M(u) - C)\nabla(\gamma \Delta u) \right) + \nabla \cdot \left(M(u)\nabla(u^3 - u) \right).$$

 $C \ge C - M((\mathbf{U})_{i,j}) \ge 0 \qquad C = \max_{i,j} M((\mathbf{U})_{i,j}).$

$$P_y^{-1} \textcircled{O} P_x^{-1} \textcircled{O} \eta^{\star} = P_y^{-1} \textcircled{O} P_x^{-1} \textcircled{O} \eta_n \odot (e^{\odot})^{-H_1 \Delta t} - P_y^{-1} \textcircled{O} P_x^{-1} \textcircled{O} (Lf(\eta_n) - Lk_1 \eta_n) \odot S_{r,s}^{1}$$

$$P_{y}^{-1} \widehat{\otimes} P_{x}^{-1} \widehat{\otimes} \eta_{n+1} = P_{y}^{-1} \widehat{\otimes} P_{x}^{-1} \widehat{\otimes} \eta_{n} \odot (e^{\odot})^{-H_{2}\Delta t} - P_{y}^{-1} \widehat{\otimes} P_{x}^{-1} \widehat{\otimes} (\frac{Lf(\eta_{n}) - Lk_{2}\eta_{n}}{2}) \odot S_{r,s}^{2} - P_{y}^{-1} \widehat{\otimes} P_{x}^{-1} \widehat{\otimes} (\frac{Lf(\eta_{\star}) - Lk_{2}\eta_{\star}}{2}) \odot S_{r,s}^{2}$$

$$(H_1)_{ij} = -L[\epsilon(d_i^x + d_j^y) - k_1], \ (H_2)_{ij} = -L[\epsilon(d_i^x + d_j^y) - k_2], \ S_{r,s}^i = \frac{1 - (e^{\odot})^{-H_i \Delta t}}{H_i}$$

Unconditionally Energy Stable



Time Integration Accuracy



High order accuracy in time is important for simulating coarsening dynamics with large-time-step schemes.

Time step-size can be 10-100X than 1st order implicit schemes; > 4 orders of magnitude larger than explicit Euler scheme.

Extensive numerical experiments can be found in



semi-implicit Euler

2nd order cETD

"Fast and accurate algorithms for simulating coarsening dynamics of Cahn–Hilliard equations", Computational Materials Science, 108 (2015), pp 272-282.

Example



$$F_{\rm c} = \int_{V} \mathrm{d}\mathbf{r} \left[\frac{1}{2} \alpha |\nabla c|^{2} + \frac{1}{2} \beta |\nabla \eta|^{2} + f(c, \eta) \right],$$

$$\frac{\partial c\left(\mathbf{r},t\right)}{\partial t} = M\nabla^{2}\frac{\delta F}{\delta c\left(\mathbf{r},t\right)},$$
$$\frac{\partial \eta\left(\mathbf{r},t\right)}{\partial t} = -L\frac{\delta F}{\delta \eta\left(\mathbf{r},t\right)},$$



$$f(c, \eta) = \frac{1}{2}A(c - c')^2 - \frac{1}{2}B\eta^2 - \frac{1}{4}C\eta^4 + \frac{1}{6}D\eta^6 + \frac{1}{2}Gc\eta^2$$



Fig. 1. Specific free energy vs composition curves for both the ordered and disordered solid solutions calculated according to equation (3) with $\eta = \eta_0(c)$ and A = 62.5, B = 15.0, C = 12.0, D = 6.25 and G = 25.0. See text for explanations.





$$\begin{split} & \frac{\partial \eta}{\partial t} = -L \begin{pmatrix} h'(\eta) \left(F^{\mathsf{v}}(c_{\mathsf{v}},c_{\mathsf{i}}) - F^{\mathsf{m}}(c_{\mathsf{v}},c_{\mathsf{i}}) \right) - k_{\eta}^{2} \nabla^{2} \eta - \nabla \left(k_{\eta}^{2}\right) \cdot \nabla \eta - \\ & \frac{\partial \left((\nabla \eta)^{2} \left(k_{\eta}\right) \left(\left(k_{\eta\varphi}^{-}\right) \left(-\frac{\eta_{y}}{\left(\eta_{x}\right)^{2} + \left(\eta_{y}\right)^{2}} \right) + \left(k_{\eta\varphi}^{-}\right) \left(\frac{\eta_{x}\eta_{z}}{\left(\nabla \eta\right)^{2} \sqrt{\left(\eta_{x}\right)^{2} + \left(\eta_{y}\right)^{2}}} \right) \right) \right) \\ & \frac{\partial \chi}{\partial t} \\ = -L \begin{pmatrix} \frac{\partial \left((\nabla \eta)^{2} \left(k_{\eta}\right) \left(\left(k_{\eta\varphi}^{-}\right) \left(\frac{\eta_{x}}{\left(\eta_{x}\right)^{2} + \left(\eta_{y}\right)^{2}} \right) + \left(k_{\eta\varphi}^{-}\right) \left(\frac{\eta_{y}\eta_{z}}{\left(\nabla \eta\right)^{2} \sqrt{\left(\eta_{x}\right)^{2} + \left(\eta_{y}\right)^{2}}} \right) \right) \right) \\ & \frac{\partial \chi}{\partial t} \\ = \frac{\partial \left((\nabla \eta)^{2} \left(k_{\eta}\right) \left(\left(k_{\eta\varphi}^{-}\right) \left(-\frac{\sqrt{\left(\eta_{x}\right)^{2} + \left(\eta_{y}\right)^{2}}}{\left(\nabla \eta\right)^{2}} \right) \right) \right) \\ & \frac{\partial \chi}{\partial z} \end{split}$$

$$\begin{split} F^{m}(c_{v},c_{i}) &= F_{A} + E_{v}^{f}c_{v} + E_{i}^{f}c_{i} + k_{B}T[c_{v}\ln(c_{v}) + c_{i}\ln(c_{i}) + (1 \\ &- c_{v} - c_{i})\ln(1 - c_{v} - c_{i})] \\ F^{v}(c_{v},c_{i}) &= k_{B}T((c_{v} - 1)^{2} + c_{i}^{2}) \end{split}$$



Fig. 2. An illustration of five representative planes in an hcp lattice.

$$\begin{split} &\frac{\partial c_{\mathbf{v}}}{\partial t} = \nabla \cdot \left(\left(M_{\mathbf{v}} \right) \nabla \left(\left(1 - h(\eta) \right) \frac{\partial \left(F^{\mathbf{m}}(c_{\mathbf{v}}, c_{\mathbf{i}}) \right)}{\partial c_{\mathbf{v}}} + h(\eta) \frac{\partial \left(F^{\mathbf{v}}(c_{\mathbf{v}}, c_{\mathbf{i}}) \right)}{\partial c_{\mathbf{v}}} \right) \right) - \gamma_{0} \eta^{2} c_{\mathbf{v}} c_{\mathbf{i}} \\ &\frac{\partial c_{\mathbf{i}}}{\partial t} = \nabla \cdot \left(\left(M_{\mathbf{i}} \right) \nabla \left(\left(1 - h(\eta) \right) \frac{\partial \left(F^{\mathbf{m}}(c_{\mathbf{v}}, c_{\mathbf{i}}) \right)}{\partial c_{\mathbf{i}}} + h(\eta) \frac{\partial \left(F^{\mathbf{v}}(c_{\mathbf{v}}, c_{\mathbf{i}}) \right)}{\partial c_{\mathbf{i}}} \right) \right) - \gamma_{0} \eta^{2} c_{\mathbf{v}} c_{\mathbf{i}} \end{split}$$



Localization



ETD



M. Hochbruck and A. Ostermann, "Exponential integrators," Acta Numerica, vol. 19, pp. 209–286, 2010.





Efficient direct subdomain integration based on FD spatial discretization

subdomain coupling techniques overlapping BC & discretization

large time step-size, stable and accurate, compute intensive

GPU Acceleration



MPI Communication



26 adjacent subdomains SD 3 recv_buf Twice per step (3)SD 2 send_buf SD 2 nx-4 recv buf (1)(2)SD 1 SD 0 SD 1 recv_buf send_buf send_buf

3-round scheme

Simulation setup



P100-PCIe-12GB: 4.7T=4812.8GFlops; 540GB

subdomain: 768*768*384=0.2109G Points,

216 subdomain = 45G points; 20,000~50,000 time steps, average step size ~ 10,000X vs. explicit

Subdomain divided into 192*192*192 blocks when calculating matrix exponentials

~ perform 32 tensor dot production simultaneously

2.45TFlops/step

Performance



Between subdomain: 73ms (pack,copy,MPI)

Tensor dot production: <u>2.42T@3.19T/sec, 759ms</u>, ~ 66% peak

Stencil & pointwise: 47ms

Overall performance: DP 2,787GFlops ~ 58% peak, ~880ms/step

Explicit FD scheme

Stencil: 12.8GFlops/step @ 40% peak ~6.2ms/step

10,000 steps= 62 sec

ETD is 70X faster!

Other Platforms



Sunway TaihuLight





40,960 SW26010 many-core processors; 260 cores, divided into 4 core groups (CGs), 1 MPE + 64 CPEs

8GB main memory for each CG

64KB SPM for each CPE

MPI recommended among CGs DMA available SPM \longleftrightarrow main memory

Performance Analysis





DGEMM: 457.2 and 408.5 GFlops, 60% and 53% peak Aggregate DMA BW in T and SP: ~ 22GB/s Overall : 316.1 to 324.5 Gflops, 41%-42% peak

Summary







A promising algorithm for a variety of architectures Large time step, scalable, compute intensive

Idea applicable to other stiff evolution equations fluid dynamics, structure-fluid interaction...

Thank you!