Laser additive manufacturing Thermal Field Prediction (LTFP)

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- * Result and Discussion
- Summary and Future Work



Background



❖ <u>L</u>aser <u>A</u>dditive <u>M</u>etal <u>M</u>anufacturing (LAMM):

Involves a set of metal additive manufacturing processes

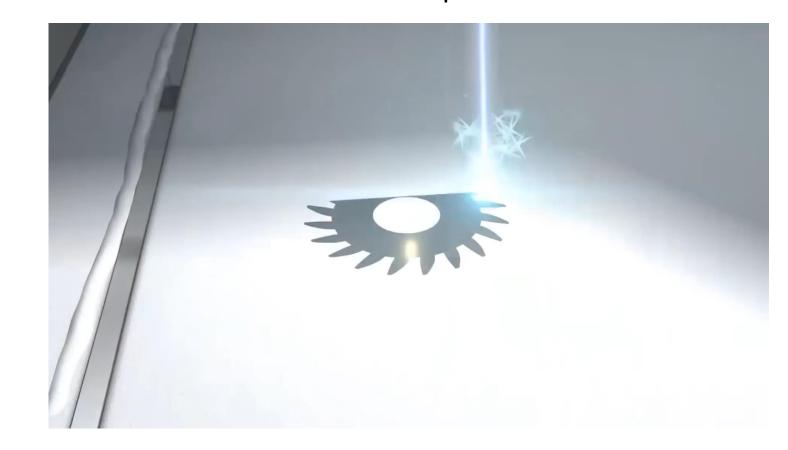
> Harness high power density of lasers to melt and fuse the metal powder and build the

structure layer by layer

Unprecedented geometry flexibility & rapid prototyping capability

Laser Beam Scans Pattern:

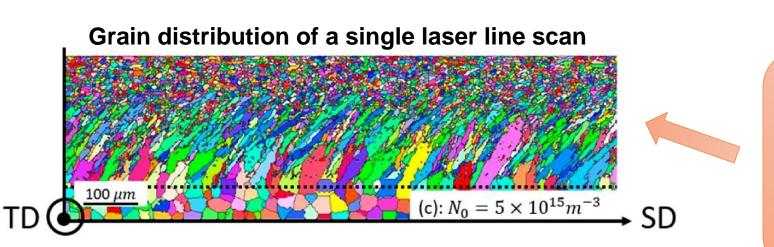
Various scan patterns are developed trying to improve product quality



Background



- Different scan pattern leads to different temperature distribution in the material, leading to different grain growth
- Grain growth is difficult to monitor during the printing
 - > Simulation is essential to revealing the grain formation process
 - Numerical models can be used to improve the scan path design by predicting the grain distribution before part is actually printed





Scan path design

Numerical model

- Residual stress concentration
- Micro-defects
- Deformation of the printed part
- Grain formation

Xiao, Zhen, et al. "Recent progress on microstructure manipulation of aluminium alloys manufactured via laser powder bed fusion." *Virtual and Physical Prototyping* 18.1 (2023): e2125880.

Motivation



- Thermal field is essential to the prediction of grain growth and other phenomena during LAMM
- ❖ Cellular Automata (CA) model
 - > Relays on precomputed thermal field
 - Lack of interaction with CA model
 - > Low flexibility
- ❖ Laser scan Thermal Field Prediction (LTFP)
 - Interact with the CA model
 - Various boundary conditions and domain increment
 - High efficiency (Parallelization)

Grain growth prediction workflow

Grain distribution Laser heat source

Thermal model

Cellular Automata (predict grain growth)

Method



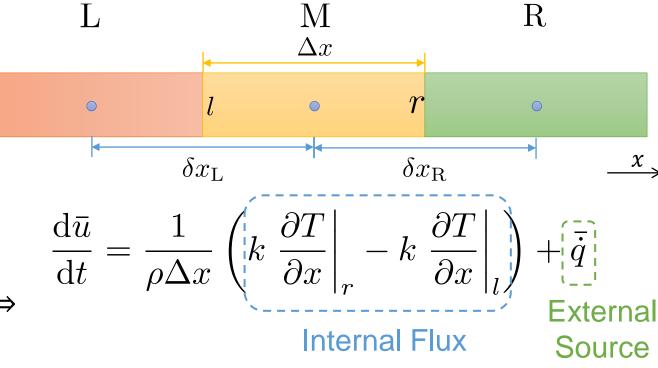
- ❖ <u>Finite Volume Method</u> (FVM):
 - > Computational domain is divided into non-overlapping control volumes
 - > The state is stored in each cell, typically the mean value
- Diffusion Equation:

> 1D Case:
$$\frac{\partial u}{\partial t} = \frac{k}{\rho} \frac{\partial^2 T}{\partial x^2} + \dot{q}$$

 $\triangleright u(T)$: specific internal energy dependent on temperature

$$\int_{l}^{r} \frac{\partial u}{\partial t} \, dx = \int_{l}^{r} \frac{k}{\rho} \frac{\partial^{2} T}{\partial x^{2}} \, dx + \int_{l}^{r} \dot{q} \, dx \quad \Rightarrow$$

> Define:
$$\bar{X} \equiv \frac{1}{\Delta x} \int_{x}^{x} X \, dx$$



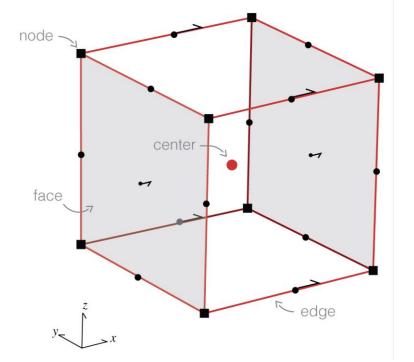
$$ightharpoonup$$
 Define: $ar{X} \equiv rac{1}{\Delta x} \int_{l}^{r} X \; \mathrm{d}x$ $\qquad rac{\mathrm{d}ar{u}}{\mathrm{d}t} = rac{1}{
ho\Delta x} \left[rac{k_r \left(T_\mathrm{R} - T_\mathrm{M}
ight)}{\delta x_\mathrm{R}} - rac{k_l \left(T_\mathrm{M} - T_\mathrm{L}
ight)}{\delta x_\mathrm{L}}
ight] + ar{q}$

Method



- ❖ <u>Finite Volume Method</u> (FVM) in 3D:
- Governing equation: $\frac{\partial u}{\partial t} = \frac{k}{\rho} \nabla^2 T + \dot{q}$
- \clubsuit Also define: $\bar{X} \equiv \frac{1}{V} \int_{C} X \ dV$
- \bigstar Recall 1D case: $\frac{\mathrm{d} \bar{u}}{\mathrm{d} t} = \frac{1}{\rho \Delta x} \left(k \left. \frac{\partial T}{\partial x} \right| k \left. \frac{\partial T}{\partial x} \right|_{L} \right) + \bar{q}$
- Similarly, we have (with central discretization):

$$\frac{\mathrm{d}\bar{u}}{\mathrm{d}t} = \frac{1}{\rho\Delta V} \left[\frac{k_{x^+} \left(T_{x^+} - T_{\mathrm{M}} \right)}{\delta x_+} \Delta y \Delta z + \frac{k_{x^-} \left(T_{x^-} - T_{\mathrm{M}} \right)}{\delta x_-} \Delta y \Delta z \right] + \frac{k_{y^+} \left(T_{y^+} - T_{\mathrm{M}} \right)}{\delta y_+} \Delta z \Delta x + \frac{k_{y^-} \left(T_{y^-} - T_{\mathrm{M}} \right)}{\delta y_-} \Delta z \Delta x + \frac{k_{z^-} \left(T_{z^-} - T_{\mathrm{M}} \right)}{\delta y_-} \Delta z \Delta x + \frac{k_{z^+} \left(T_{z^+} - T_{\mathrm{M}} \right)}{\delta z_+} \Delta x \Delta y + \frac{k_{z^-} \left(T_{z^-} - T_{\mathrm{M}} \right)}{\delta z_-} \Delta x \Delta y} \right] + \bar{q} \implies \frac{\mathrm{d}\bar{u}}{\mathrm{d}t} = \cdots \equiv f(T)$$



Source: https://row1.ca/pixels-and-their-neighbors

$$\implies \frac{\mathrm{d}\bar{u}}{\mathrm{d}t} = \dots \equiv f(T)$$

Method



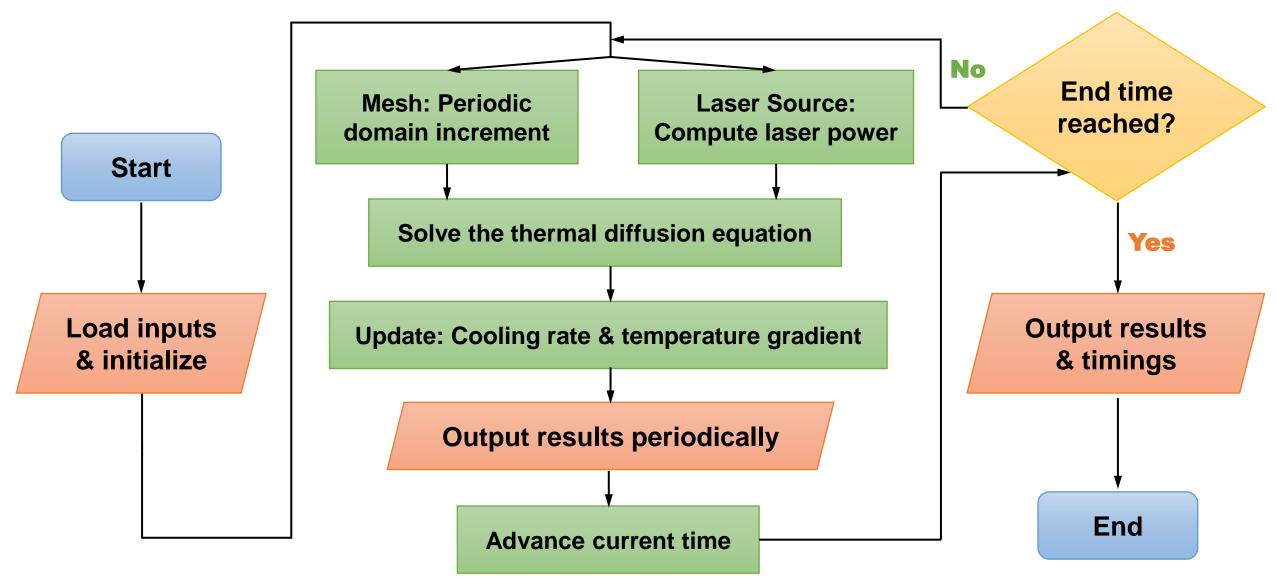
- ❖ Time stepping: $\frac{\mathrm{d}\bar{u}}{\mathrm{d}t} = \cdots \equiv f(T)$ ➤ Forward Euler (FE): $\frac{\mathrm{d}\bar{u}}{\mathrm{d}t} = f(T) \implies \left[\bar{u}_{i,j,k}^{(n+1)} = \bar{u}_{i,j,k}^{(n)} + \Delta t_{i,j,k}^{(n)} f\left(T_{i,j,k}^{(n)}\right)\right]$

- > Von Neumann stability Criterion: $\Delta t_{i,j,k}^{(n)} \leq \min \left| \frac{(\Delta x)^2}{6k^{(n)}/\rho}, \frac{(\Delta y)^2}{6k^{(n)}/\rho}, \frac{(\Delta z)^2}{6k^{(n)}/\rho} \right|$
- ❖ Cooling rate and temperature gradient are needed for the grain growth model:

$$-\frac{\partial T^{(n)}}{\partial t} = \frac{T^{(n-1)} - T^{(n)}}{\Delta t}, \quad \nabla T^{(n)} = \left(\frac{T_{i+1,j,k}^{(n)} - T_{i-1,j,k}^{(n)}}{2\Delta x}, \frac{T_{i,j+1}^{(n)} - T_{i,j-1}^{(n)}}{2\Delta y}, \frac{T_{i,j,k+1}^{(n)} - T_{i,j,k-1}^{(n)}}{2\Delta z}\right)$$

Flowchart





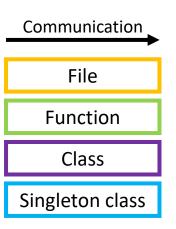
Code Algorithm

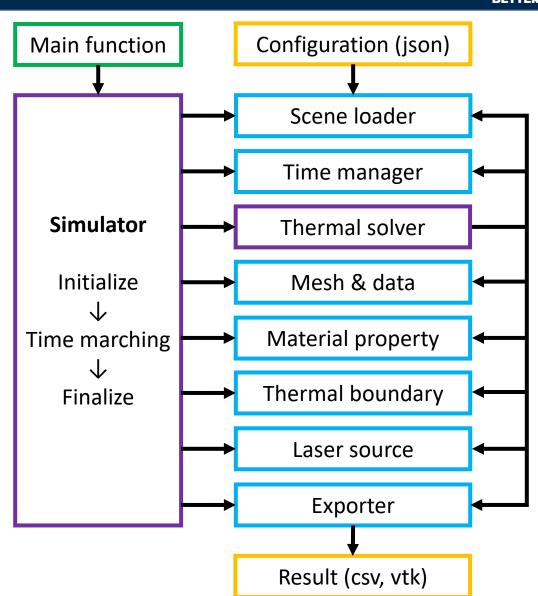


Object-oriented design and programming using



- All modules of simulation are written as class objects
- Make most modules singleton to provide global access
- Modules can be reassembled differently to meet different simulation needs
- Uses <u>inherited class</u> + <u>singleton manager</u> to keep track of different types of objects in the same module and provide unified interface



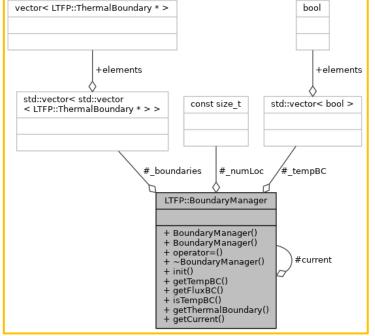


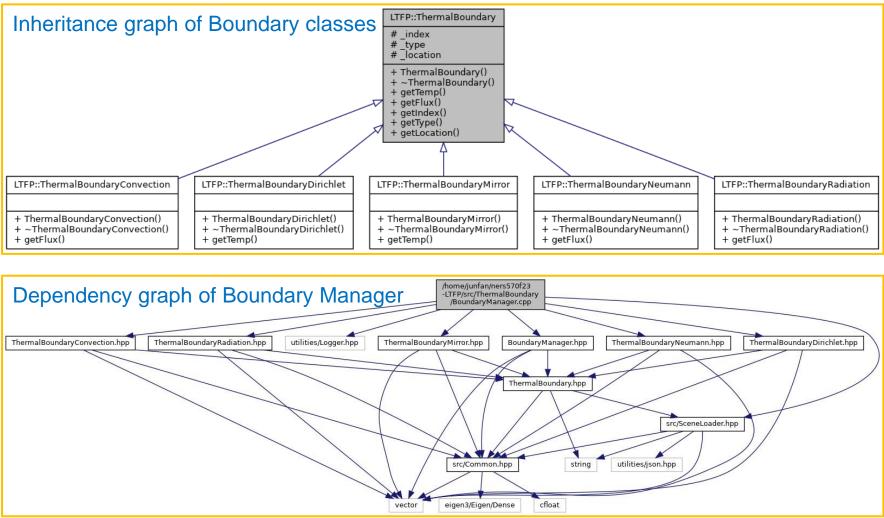
Code Algorithm



Structure of thermal boundary module (diagrams generated by Doxygen)

UML diagram of Boundary Manager object | vector < LTFP::ThermalBoundary * > | bool |





Simulation result – 1D diffusion



1D diffusion test

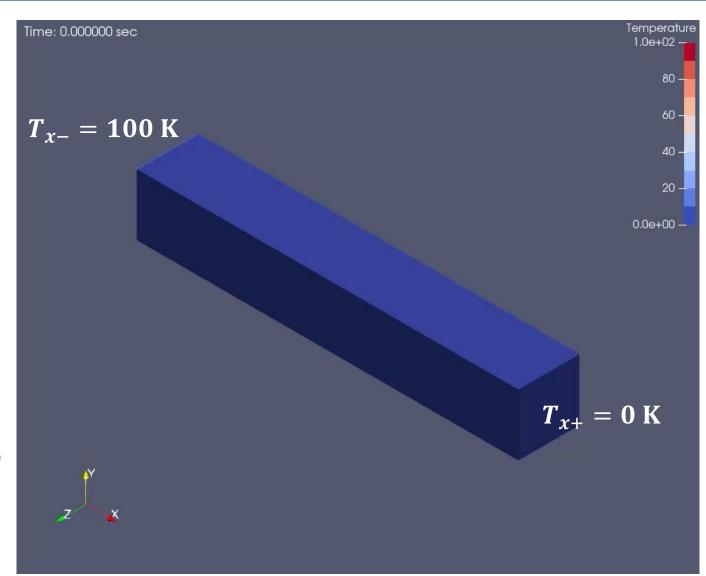
- Constant diffusion coefficient
- ❖ Initial temperature: 0 K
- Dirichlet boundary at

$$> x + : 0 K$$

$$> x^-: 100 \text{ K}$$

➤ Analytical solution:

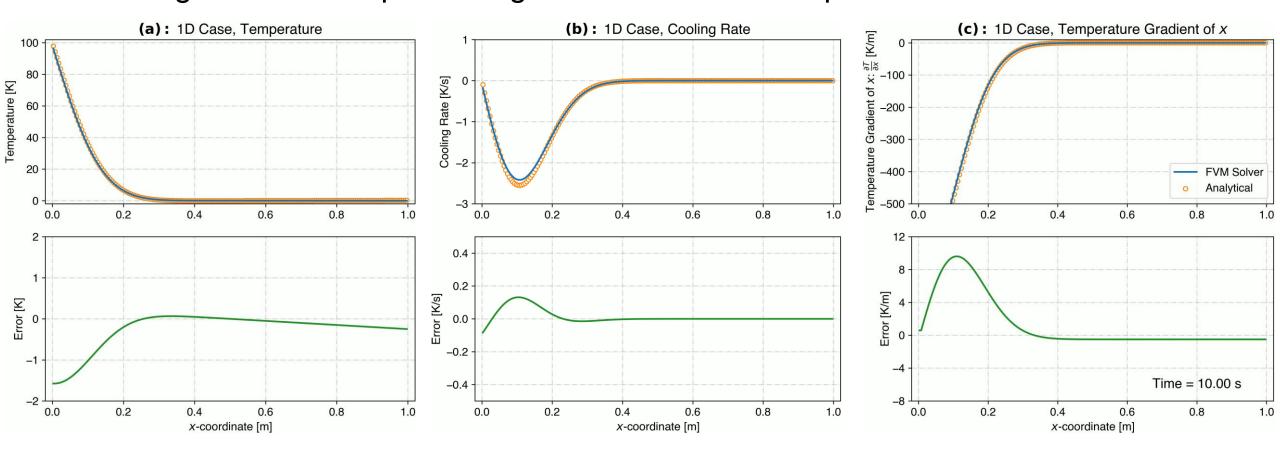
$$T^*(t,x) = \sum_{n=1}^{+\infty} \left\{ \frac{200}{n\pi} \left[\frac{(-1)^n}{n\pi} - 1 \right] \sin(n\pi x) \right.$$
$$\times e^{-\frac{k}{c_p \rho} (n\pi)^2 t} \right\} + 100 - 100x$$



Simulation result – 1D diffusion



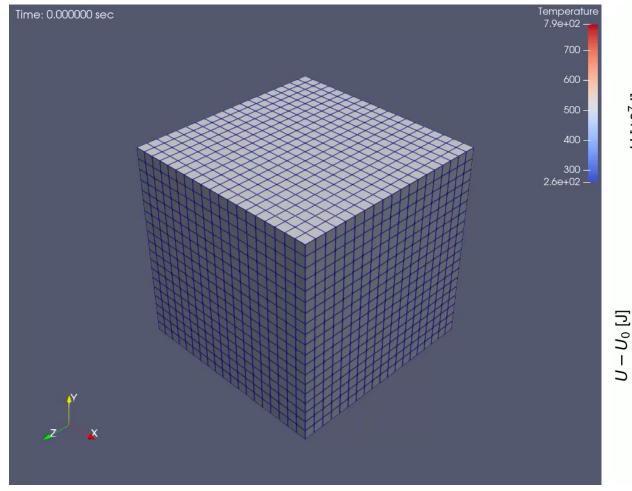
- ❖ Temperature, cooling rate and temperature gradient with respective to the x-axis (diffusion direction)
- Cooling rate and temperature gradient behave as expected

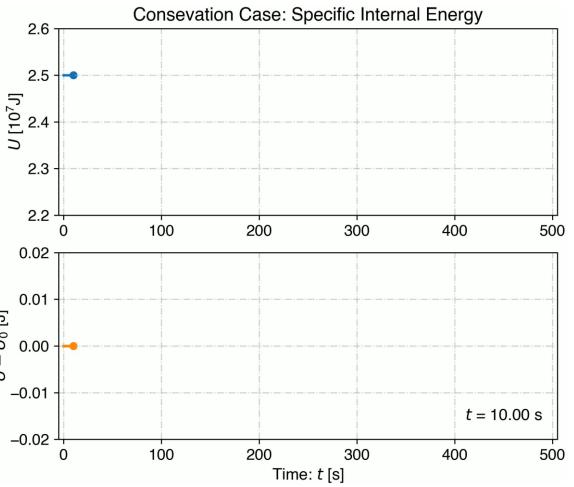


Simulation result - Conservation



- Conservation test case with Neumann boundary on all surfaces. All flux is summed up to zero.
- Applied temperature-dependent thermal conductivity and specific heat.



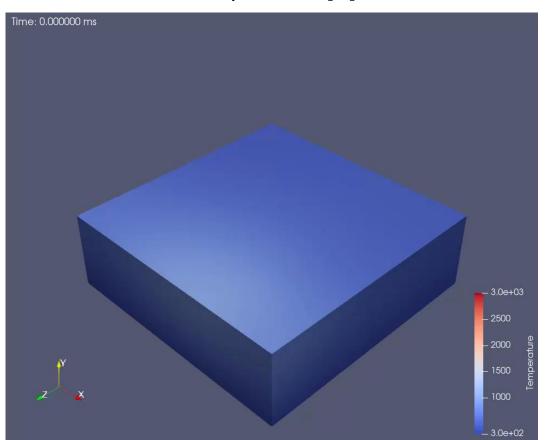


Simulation result – Multi-layer scan test

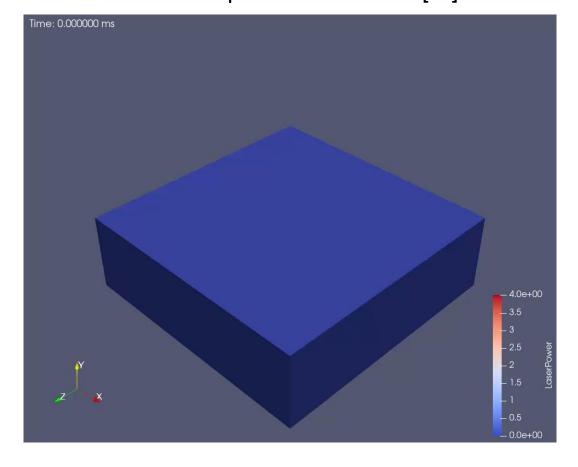


Multi-layer scan on a block steel with domain increment

Temperature [K]



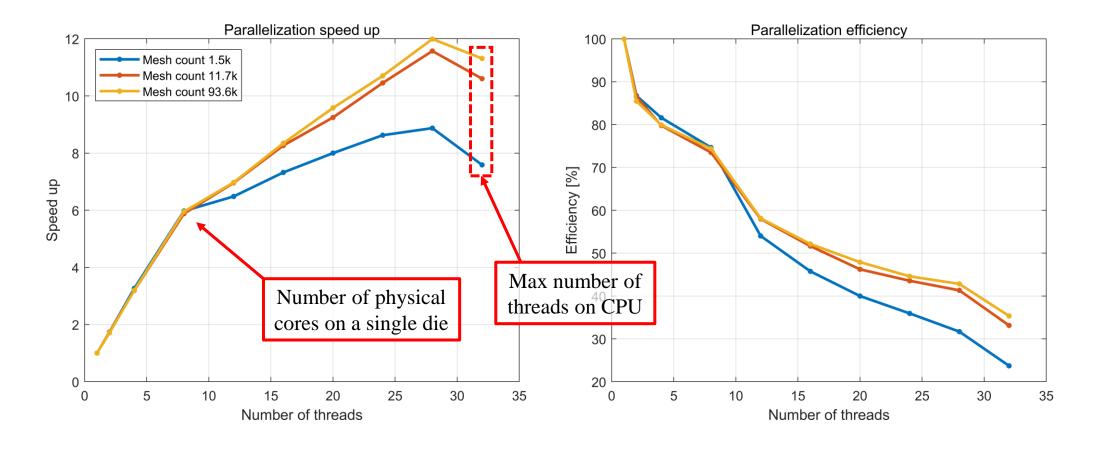
Laser power distribution [W]



Parallelization performance



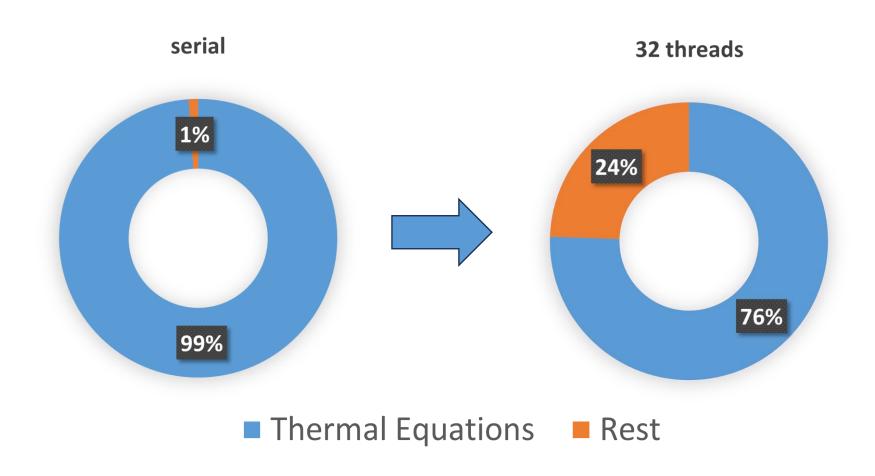
- Loops are parallelized with OpenMP
- ❖ The multilayer scan case is timed in WSL running on Ryzen 7950X, 16c32t, @5.4GHz



Parallelization performance



❖ Time consumption of solving thermal equations is greatly reduced after parallelization, but still takes a significant part of comptutation



Documentation

Vector3

LTFP::ThermalBoundary Class Reference

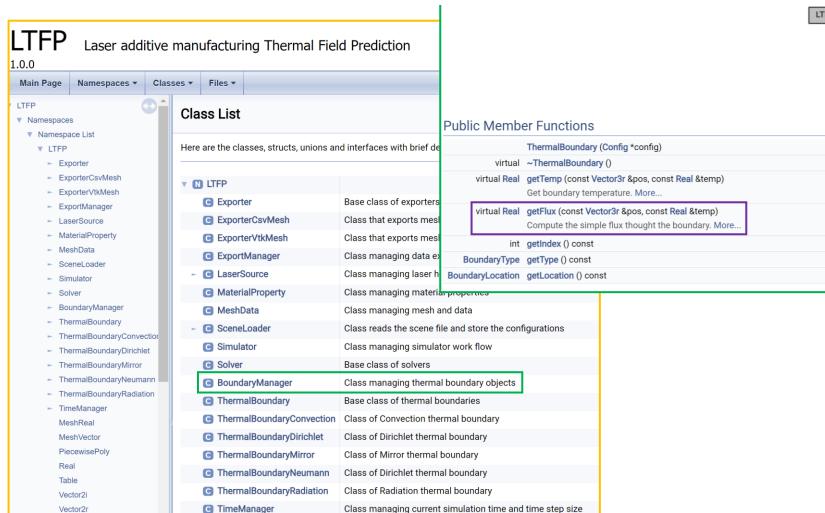
Base class of thermal boundaries. More..

#include <ThermalBoundary.hpp>

Inheritance diagram for LTFP::ThermalBoundary:



Documentation is generated using Doxygen



LTFP::ThermalBoundaryConvection LTFP::ThermalBoundaryDirichlet LTFP::ThermalBoundaryMirror LTFP::ThermalBoundaryNeumann LTFP::ThermalBoundaryRadiation [legend]

Member Function Documentation

• getFlux()

 $\textbf{Real} \ \texttt{LTFP::} Thermal Boundary:: getFlux \ (\ const \ \textbf{Vector3r} \ \& \ \ \textbf{pos,}$

const **Real** &

temp

Compute the simple flux thought the boundary.

Parameters

pos Position of the boundary neighboring cell

temp Temperature of boundary neighboring cell

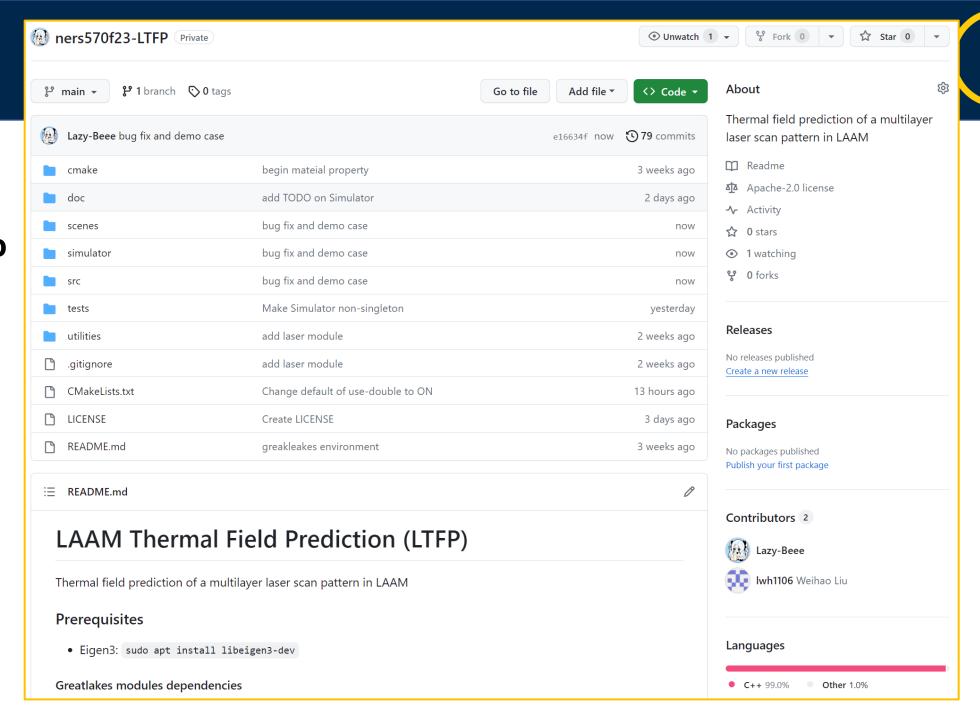
Returns

Flux though the boundary

Reimplemented in LTFP::ThermalBoundaryRadiation, LTFP::ThermalBoundaryRadiation

GitHub

Code is managed on GitHub



WORK BETTER

Summary



- Designed algorithm and completed coding of LTFP
 - Using the object-oriented programming language
 - Combined with parallelization
- Preliminarily tested the model
 - 1D case
 - Energy-conserved case
 - Functional test
- Generated git repo and documentation

Future work



Work in progress

- ❖ Add more advanced and stable **solvers**: RK2, RK4, ...
- ❖ Analysis and improve <u>parallelization</u> performance
- More <u>analysis</u> on generated thermal field and different scan patterns

Future work

- Validation using experiment data
- Integrate LTFP into the grain prediction model
- Uneven domain increment
- Z-ordered data storage

