

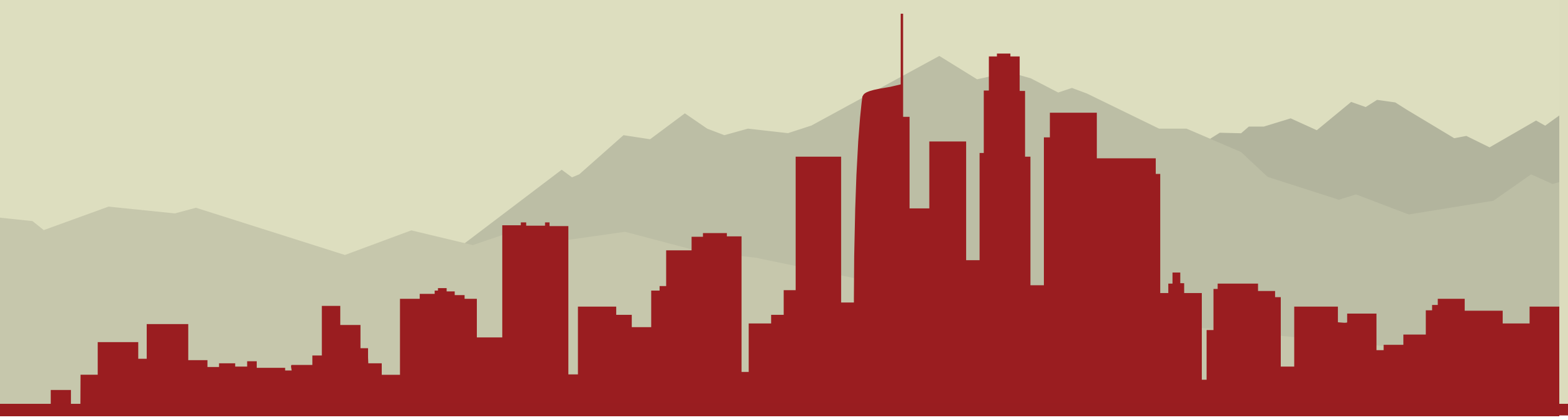
Simulation of Multi-surface Plasticity in the Discontinuous-Mesh GPU-powered Wave Propagation Code AWP

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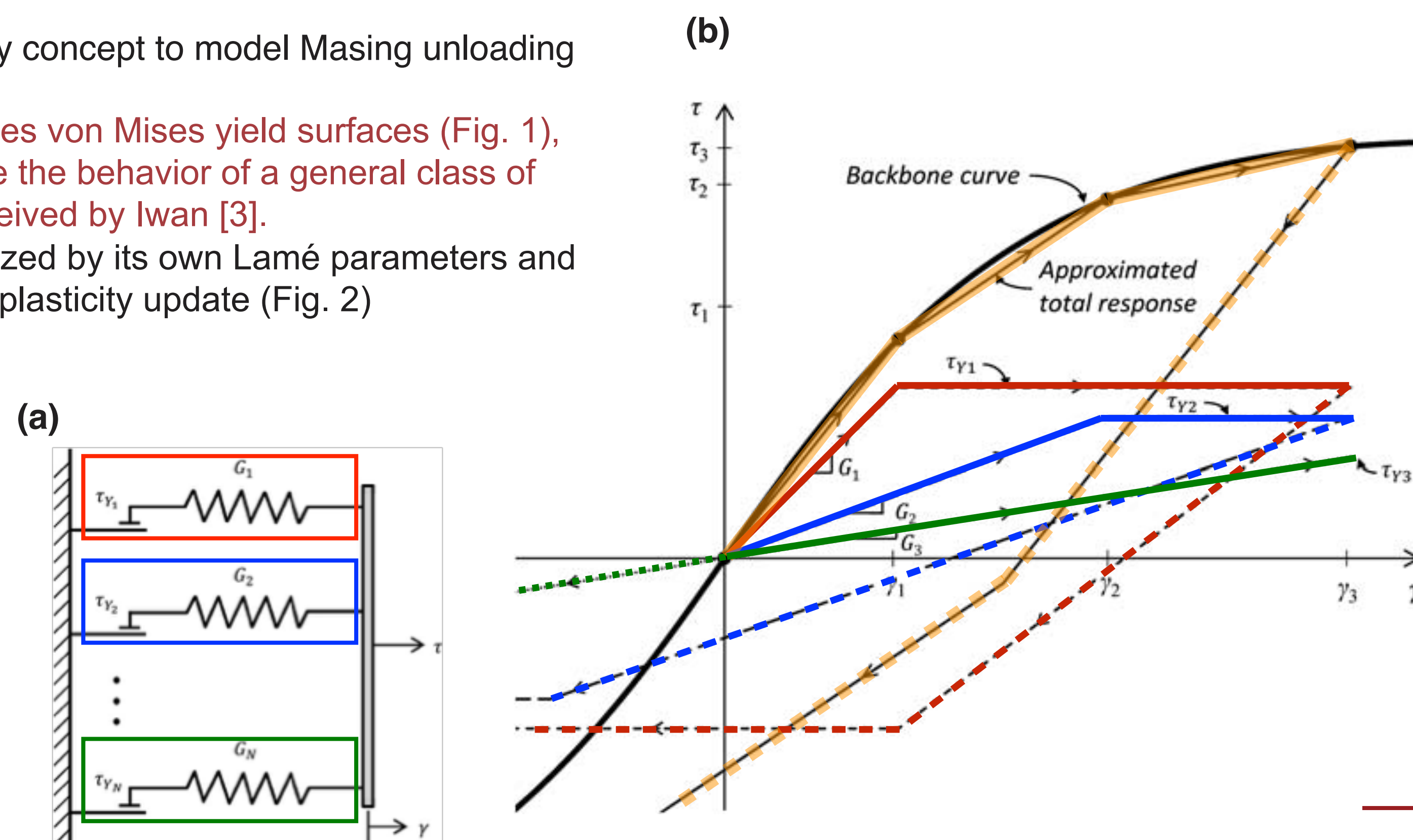
1. Motivation and Scope

- Wave propagation simulations and vertical arrays observations demonstrate *nonlinear coupling* between source, path and site effects which determines strong ground motions [1].
- Predicting this *shallow crust nonlinearity* is the focus of a technical activity group (TAG) within SCEC.
- One of the TAG's goals is the development, verification and validation of new wave propagation codes which accurately model the shear modulus degradation in soils and which are scalable and efficient enough to resolve a large computational domain.
- We have added support for Iwan-type nonlinearity in the finite difference wave propagation code AWP.
- Iwan nonlinearity was first implemented in the CPU code of AWP and verified against independent codes under previous SCEC awards [2].
- Iwan-type nonlinearity has now been implemented in the GPU-powered, discontinuous mesh (DM) version AWP-GPU-DM.

2. Overlay Concept

- AWP-GPU-DM uses the overlay concept to model Masing unloading and reloading behavior.
- It tracks a series of parallel-series von Mises yield surfaces (Fig. 1), which in combination reproduce the behavior of a general class of material models originally conceived by Iwan [3].
- Each yield surface is characterized by its own Lamé parameters and requires a separate stress and plasticity update (Fig. 2)

Figure 1. (a) Parallel-series configuration of spring-slider arrangement in 1D Iwan model. (b) Stress-strain behavior of 3 elasto-plastic elements [modified from 4].



3. Implementation Details

- We have developed CUDA kernel functions which perform stress and J2 plasticity updates and compute the overlay velocity field (Fig. 2).
- To conserve GPU memory, material properties pertaining to each surface are re-calculated on demand, rather than permanently stored.

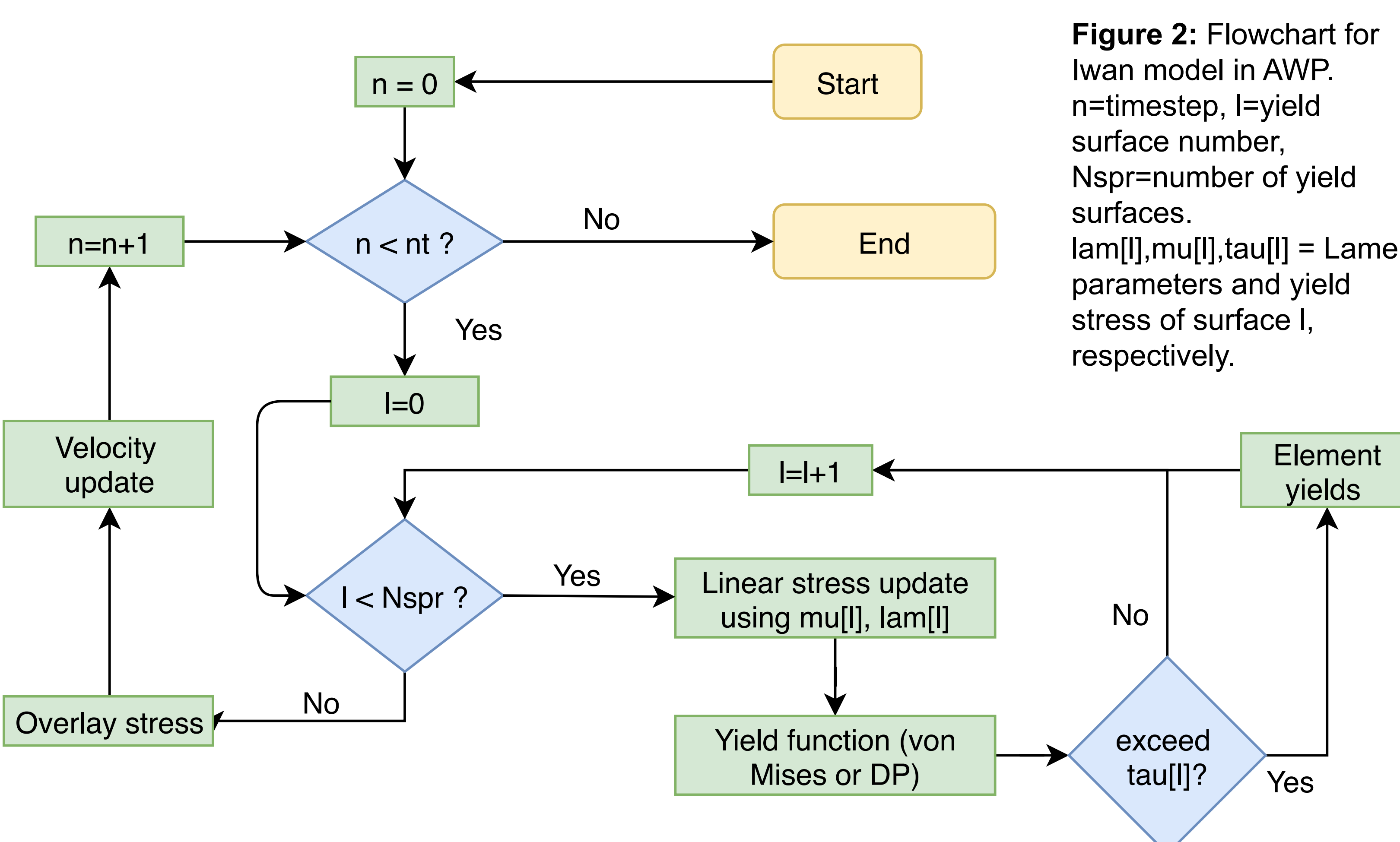


Figure 2: Flowchart for Iwan model in AWP. n =timestep, l =yield surface number, $Nspr$ =number of yield surfaces. $\text{lam}[l], \mu[l], \tau[l]$ = Lamé parameters and yield stress of surface l , respectively.

4. Code Verification

- Verification benchmark involves a horizontally layered soil column reflecting conditions at Kik-net site KSRH10 (also used in the PRENOLIN benchmark) [5].
- Plane strain conditions are specified by enabling periodic boundary conditions at horizontal domain boundaries and defining source as plane wave entering at domain bottom (recently implemented in AWP-GPU-DM).
- A linear simulation using 10 elements was first carried out to verify correctness of Lamé parameter initialization, multi-surface stress updates and computation of overlay velocity (Fig. 3).
- Next the multi-surface J2 plasticity routines, initial stresses and yield strength of each surface were verified using a fully nonlinear computation with 10 yield surfaces (Fig. 4).
- Minor differences between the CPU and GPU versions of AWP with Iwan nonlinearity arise from different optimization strategies which affect the order of interpolations on the staggered grid.
- Linear and nonlinear results obtained with the Iwan model are consistent with those obtained using AWP-CPU-IWAN, which has been verified against the 1D and 2D versions of the Noah [6] code [2].

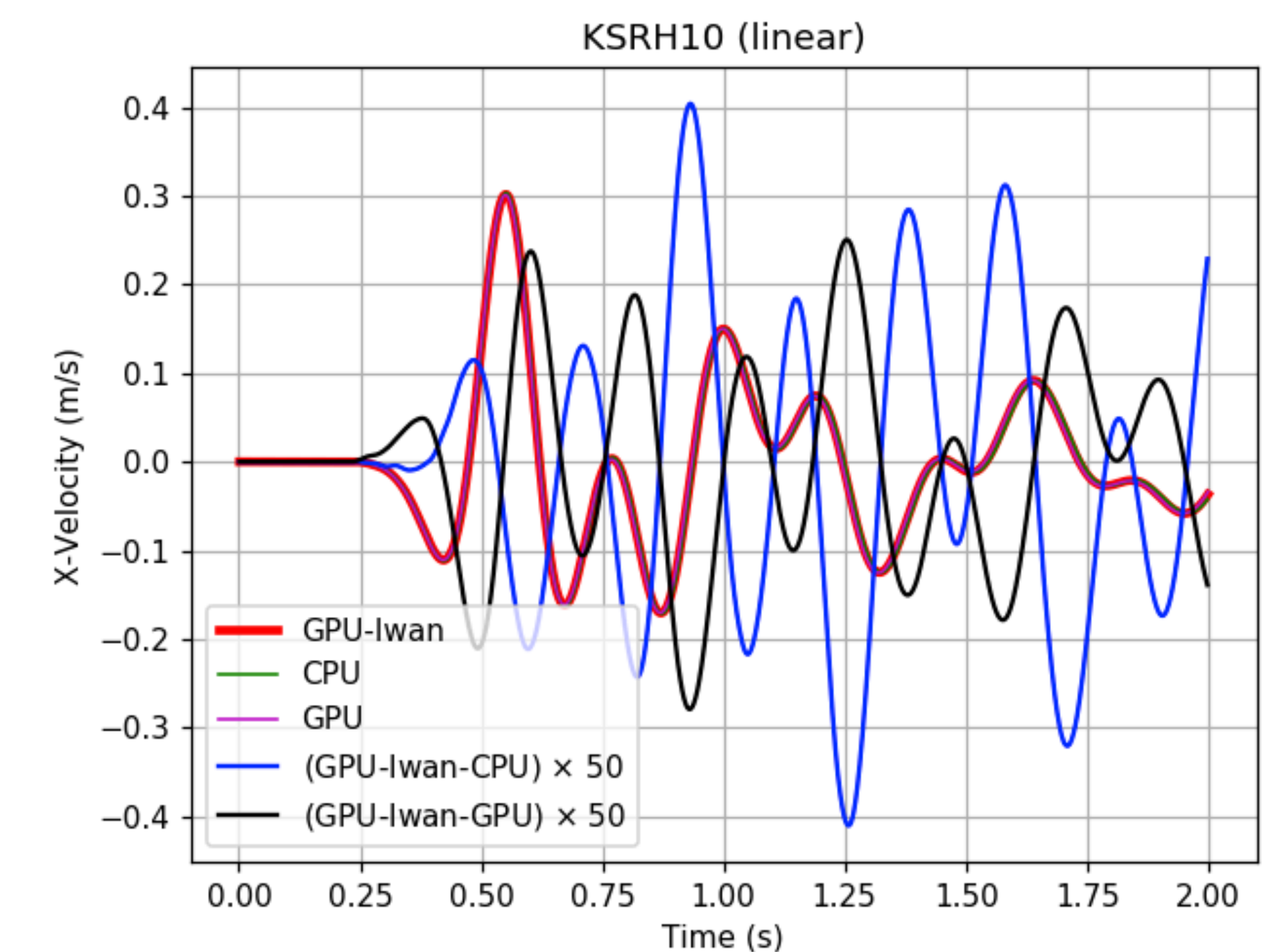


Figure 3: Surface velocity time series obtained from GPU code with a linear multi-surface simulation and from linear single-surface simulation using the CPU and GPU versions of AWP. Blue and black lines show differences between multi-surface linear and single-surface results, inflated by a factor of 50.

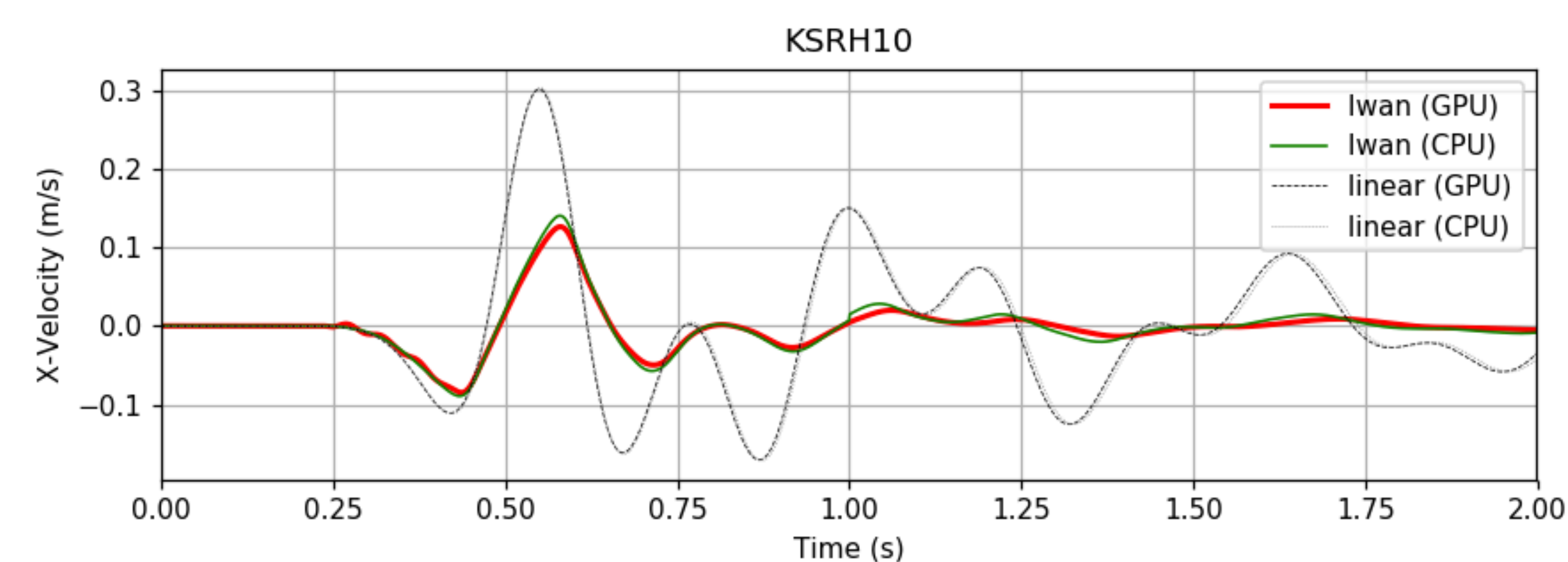


Figure 4: Surface velocity time series at Kik-net site KSRH10 obtained from Iwan-type nonlinear simulation using the GPU and CPU versions of AWP. The linear solution is shown for reference. Nonlinear predictions using the Iwan model were obtained using 10 yield surfaces.

5. Parallel Efficiency

- We have benchmarked the parallel efficiency of AWP-GPU-DM with Iwan nonlinearity on OLCF Summit
- The benchmark was carried out using a DM with 3 different mesh sizes and 10 yield surfaces in the uppermost grid.
- We measured a parallel efficiency of 94.7% on 16,384 Volta V100 GPUs (Fig. 5, Table 1)

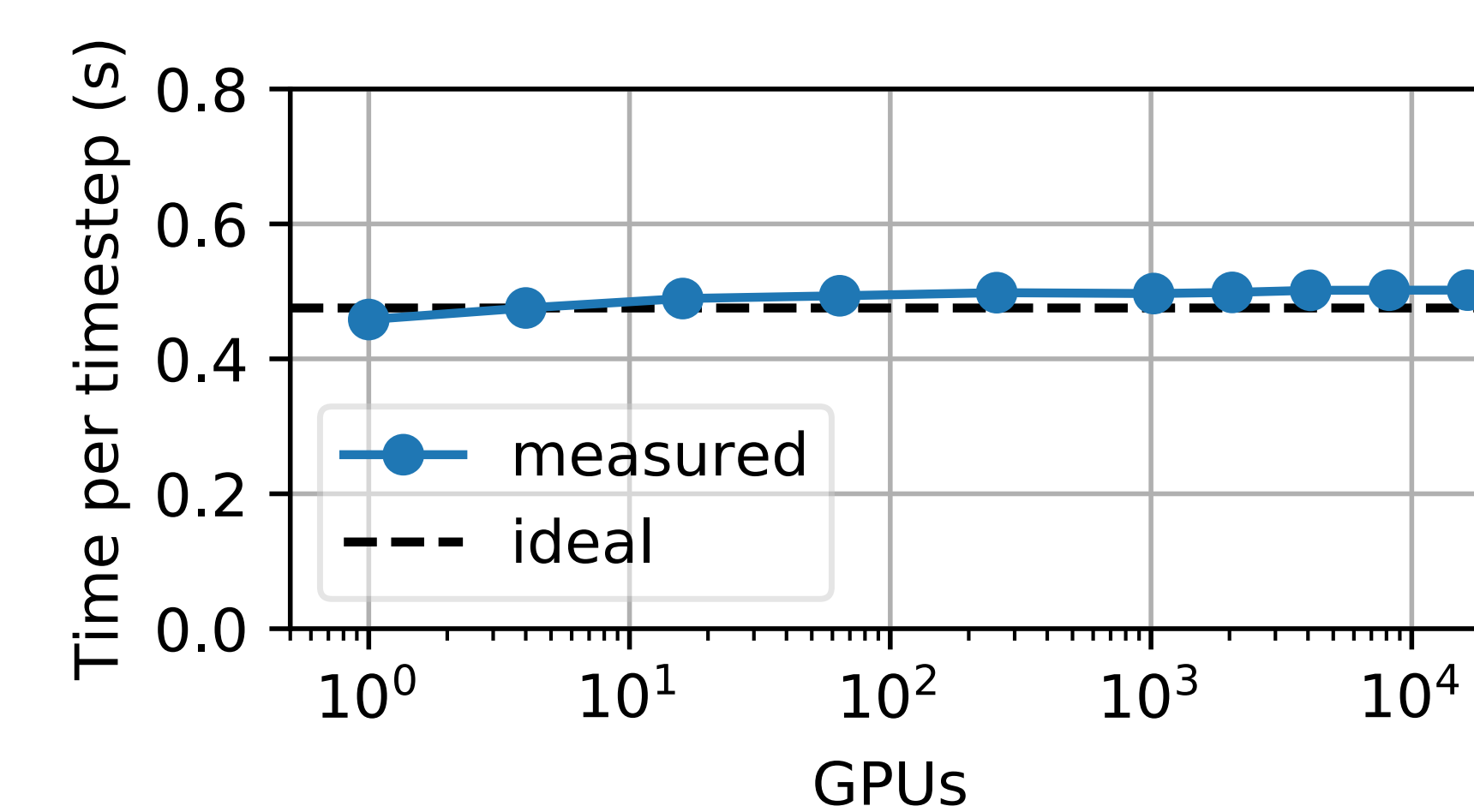


Figure 5: Weak scaling of AWP-GPU-DM with Iwan nonlinearity on OLCF Summit.

| GPUs | Time per Timestep | Efficiency (%) |
|--------|-------------------|----------------|
| 1 | 0.4583 | 103.7 |
| 4 | 0.4754 | 100.0 |
| 16 | 0.4894 | 97.1 |
| 64 | 0.4935 | 96.3 |
| 256 | 0.4981 | 95.4 |
| 1,024 | 0.4968 | 95.7 |
| 2,048 | 0.4985 | 95.4 |
| 4,096 | 0.5016 | 94.8 |
| 8,192 | 0.5018 | 94.7 |
| 16,384 | 0.5019 | 94.7 |

Table 1: Time per timestep and parallel efficiency from AWP-GPU-DM weak scaling test with Iwan nonlinearity on Summit.

6. Summary and Outlook

- Multi-surface nonlinearity using the Iwan model was implemented in both the CPU and GPU versions of the AWP finite difference code.
- The correctness of the algorithm was verified by comparing solutions of 1D and 2D benchmarks against reference waveforms computed using Noah [2,6].
- CUDA kernels required for the Iwan model in the GPU code were verified by comparison against the CPU version of AWP.
- The high computational density of Iwan plasticity results in very good weak scaling performance on several 1,000 GPUs.

Selected References

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