## **Application of High-performance Visual Analysis Methods to Laser Wakefield Particle Acceleration Data**

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Background

Laser WakeField Particle Accelerators (LWFAs) promise to be a compact source of high-energy electron beams and radiation. Similar to the wake of a boat, the radiation pressure of a laser pulse fired into a plasma displaces the electrons in the plasma, and with the space-charge restoring force of the ions, this displacement drives a wave (wake) in the plasma. Electrons can be trapped and accelerated by the longitudinal field of the wake, forming electron bunches of high energy. LWFAs have been shown to generate electric fields thousands of times stronger than those in conventional particle accelerators, accelerating particles to high energies of 1GeV within 3cm compared to ~50m in traditional electromagnetic accelerators.

**Simulations** are performed over 2D and 3D domains using the VORPAL code. Due to the large amount of particles required to achieve accurate simulation results, it is not possible to simulate the entire plasma at once. Simulations are therefore restricted to a window that covers only a subset of the plasma in x direction in the vicinity of the beam. The simulation code moves the window along the local x axis over the course of the simulation. Each simulation produces a set of files for the particle and field data (at typically 40-100 timesteps) with the following main characteristics:

- Particle data: scattered data with: x,y,z (particle location), px, py, pz (particle momentum), id (particle identifier) • No. of particles per timestep:  $\sim 0.4*10^6 - 30*10^6$  (in 2D) and  $\sim 80*10^6 - 120*10^6$  (in 3D)
- Total size: • Field data: electric field, magnetic field, and RhoJ (defined on a regular grid) Resolution:
  - Total size:

System Design

To accelerate data mining operations, we use FastBit, a stateof-the-art data management technology for indexing and searching. We use FastBit to perform data subsetting/selection and to compute conditional histograms. We implemented these operations using FastBit directly in the file-reader stage of the processing pipeline in Vislt, a production-quality, parallel capable visual analysis system (see Figure a). The conditional histograms serve as basis for the visual presentation of data vis-a-vis histogram-based parallel coordinates.

In contrast to earlier work, we employ a histogram-based parallel coordinates rendering for both context and focus views of large, complex data. Via FastBit, we can recompute conditional histograms fast, thus enabling support for fast data selection and smooth drill-down into finer level of detail in very large datasets. As a further improvement, we also support adaptively binned (equal-weight) histograms. As illustrated in Figure b, adaptively binned histograms are especially useful for display of low-level-of-detail views where the number of bins per variable is much smaller than the number of pixels per parallel axis. -10

Our system uses two main types of data selection that are implemented by FastBit: i) multi-variate thresholding, and ii) identifierbased selection. Multivariate thresholding is used for defining "interesting" data subsets. ID-based selection is the basis for tracing of particles over time. We parallelize computations over the temporal domain to accelerate, e.g., particle tracking.

## **Beam Analysis**

In order to gain a deeper understanding of the acceleration process, we need to address complex questions such as: i) Which particles become accelerated? ii) How are particles accelerated? and iii) How is the beam of highly accelerated particles formed and how does it evolve? To identify those particles that were accelerated, we first perform a selection of particles at a given timestep when the beam has already formed. By tracing the selected particles over time (using ID-queries) we can effectively analyze the temporal behavior of the beam. By refining selections based on information from different timesteps, we are then able to identify characteristic substructures of a beam. An example analysis of a 3D LWFA dataset is shown below.

Figure a) shows a parallel coordinates plot of timestep t=12 of the 3D dataset with particles with x-direction momentum  $px>2*10^9$ as context (gray) and particles satisfying the condition (px>4.856\*1010) &&(x>5.649\*10-4) as focus (red). Figure b) shows a volume rendering of the plasma density and the focus particles (red). Clearly, the selected particles form a compact beam in the first wake period following the laser pulse. Figure c) shows the traces of the selected beam at t=9 to t=14 colored according to px, from low (blue) to high (red). The transition from blue to red indicates that particles undergo constant accelerated over time.





~3.5GB - >70GB (in 2D) and ~200GB - >2TB (in 3D)





Figure b: Histogram-based parallel coordinates using 32x32 uniform binning (left) and adaptive binning (right)



