

STCF Core Software Status

Teng LI on behalf of the STCF core software development team

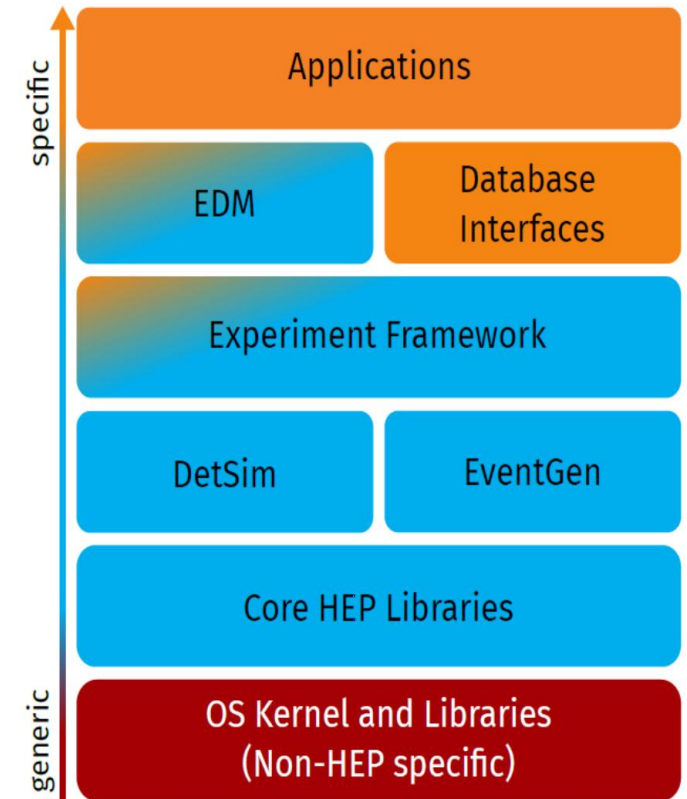
Shandong University

2025-7-4

2025年超级陶粲装置研讨会 湘潭

Main R&D Challenges for STCF Software

- ❖ Main R&D challenges and innovations for STCF core software
 - The amount of data requires much more advanced performance
 - Relying on pure single-threaded CPU resource to process **hundreds of PB** of data is hardly realistic
 - Parallel computing, as well as heterogeneous resources, need to be considered to overcome the challenges.
 - The core software needs to provide ready-to-use development and run time environment
 - Adoption of common software developed for future colliders
 - OSCAR is developed partially based on Key4hep, including EDM based on podio, geometry based on DD4hep etc.
 - Better support of ML-based applications



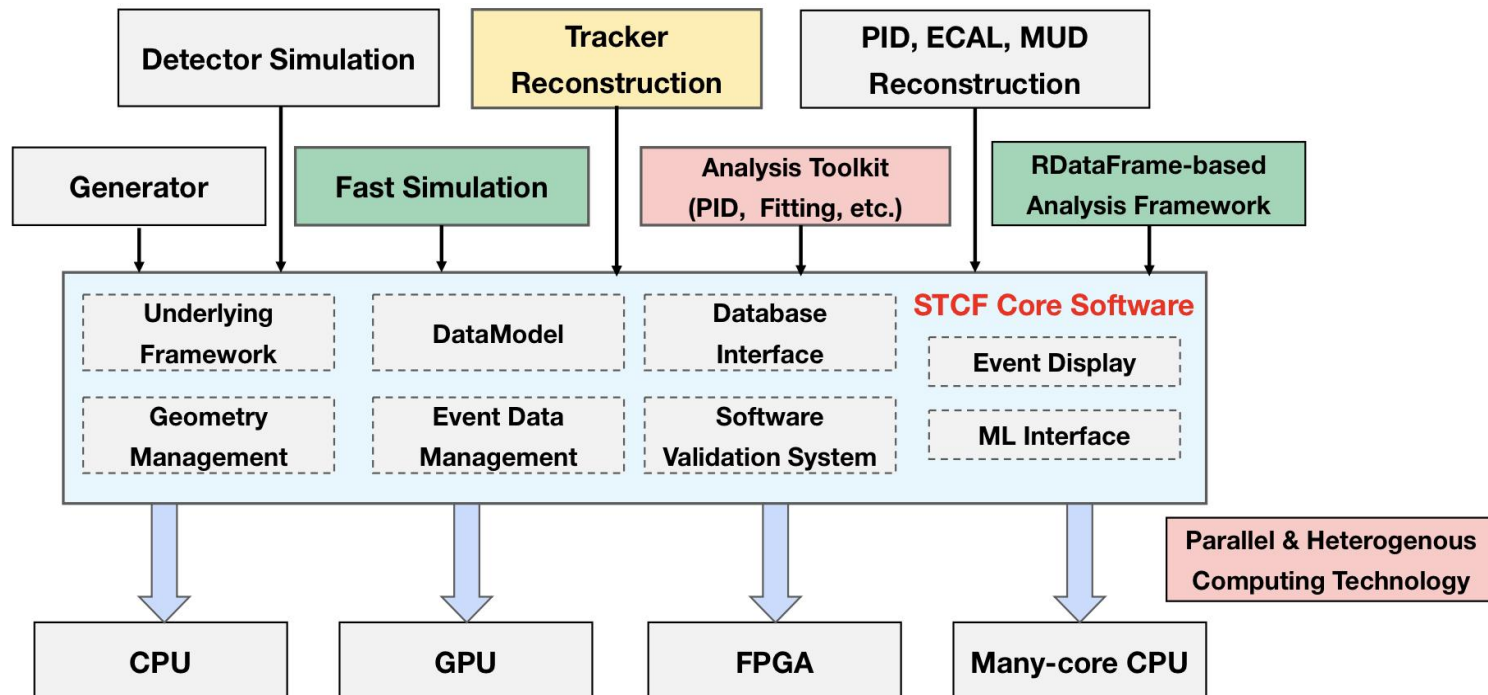
Key4hep
Thomas Madlener,
Epiphany Conference 2021

Overview of STCF Core Software

❖ The task of the STCF core software

- To fulfill official offline data processing tasks, i.e. detector simulation, digitization, calibration and reconstruction
- Provide a common platform for users to perform data analysis

❖ Overview of STCF core software



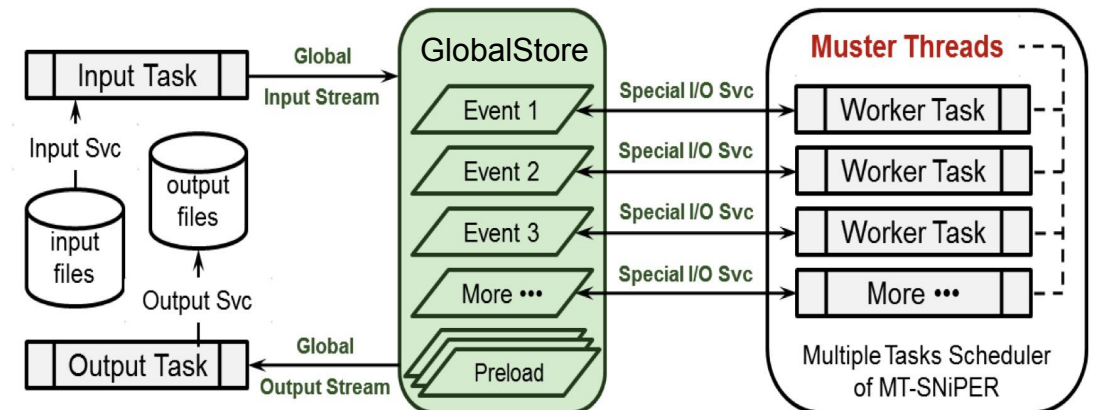
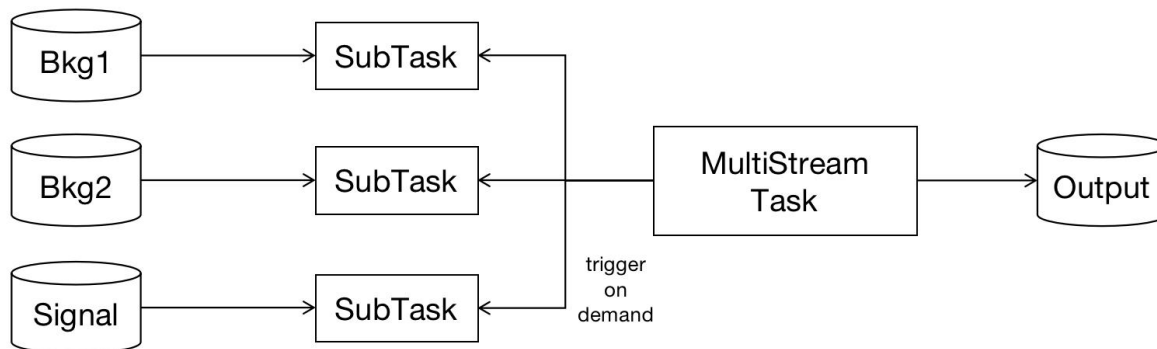
- The underlying framework
- Event data management
- Detector description and conditions data management
- Event display
- Support of ML, parallel computing, and heterogeneous computing
- Software and physics validation
- Software build, installation and distribution

Recent Progress of OSCAR

- ❖ Several new releases since FTCT2025
 - 2.6.0, 2.6.1, **2.6.2 (current release)** and a few pre-releases
 - Most functionalities in place and stable, supporting MC production and physics studies
 - Supporting physics studies for the TDR as the first priority
- ❖ Major updates in 2025:
 - Great optimization of disk consumption and running speed of simulation and reconstruction jobs
 - Release of fast simulation software package
 - Release of Global-PID based on weighted combined likelihood method
 - Release of ACTS-based tracking
 - Major updates of EventDisplay
 - Lots of optimizations and fixes for various physics simulation studies
- ❖ Dedicated tutorial in 2025 Feb. Many analyzers get involved

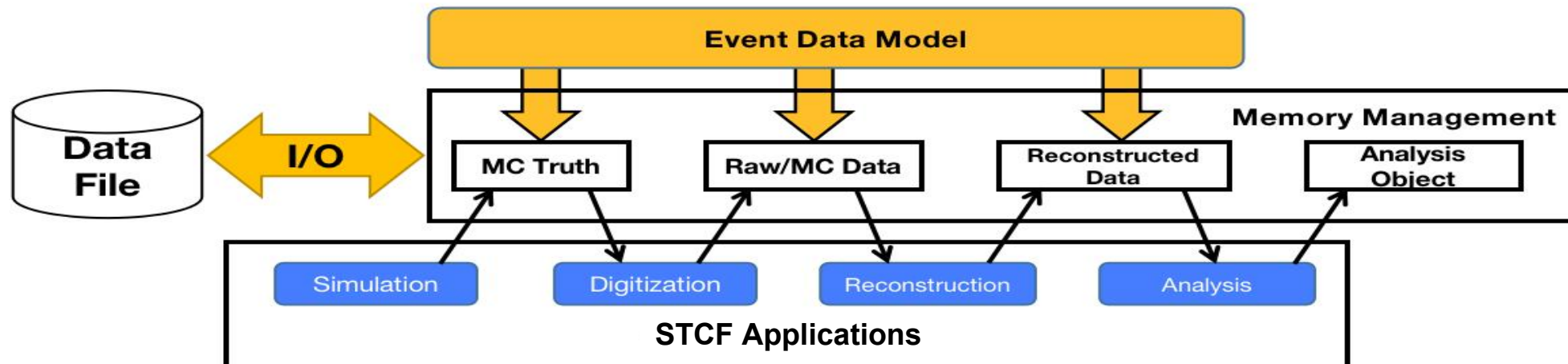
Underlying Framework: SNIpER

- ❖ The underlying framework builds the skeleton of OSCAR
 - Provide basic functionalities of **event loop control, algorithm scheduling, thread management, user interface, job configuration, logging** etc. (Like Gaudi for BOSS)
- ❖ OSCAR adopts SNIpER as the underlying framework
 - **Lightweighted, efficient and highly extendable**
 - Developed since 2012, maintained by **10+ developers** from IHEP, SDU, etc.
 - Adopted by JUNO (neutrino), LHAASO (cosmic ray), nEXO (neutrinoless double beta decay) and HERD (dark matter)
- ❖ Recent updates
 - Better support for inter- and intra- event level parallism

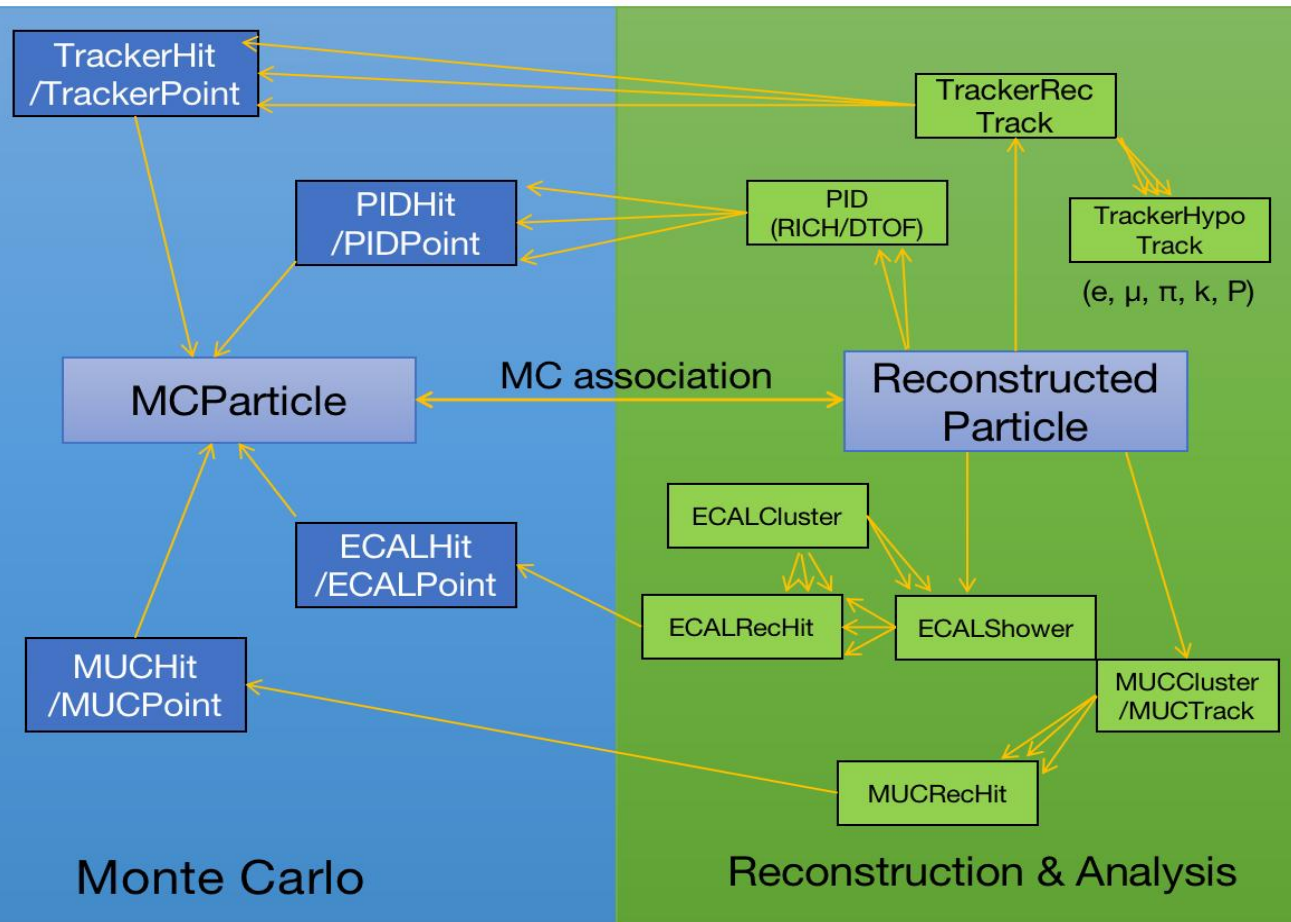


Event Data Management

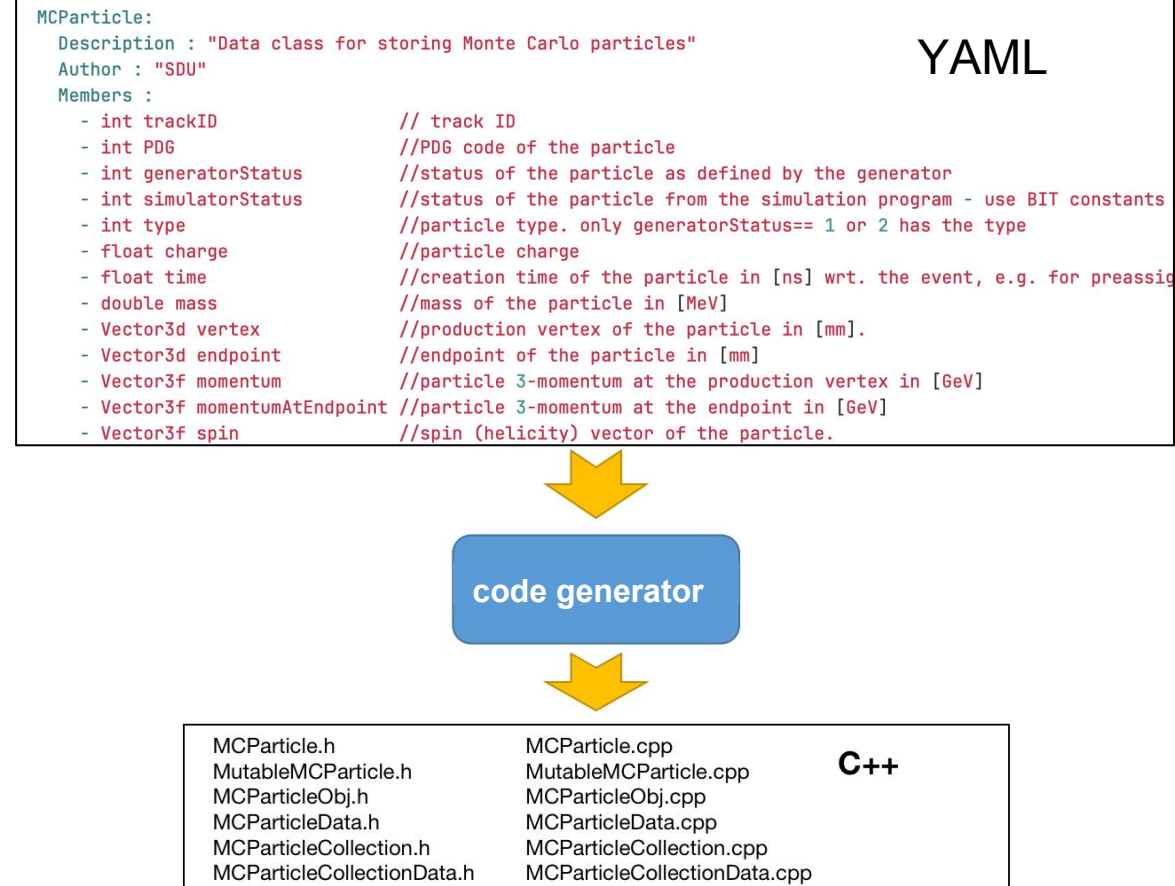
- ❖ Event data management is the most crucial part of the framework
 - Provide tools to define the Event Data Model (EDM)
 - The definition of physics event data (MC particles, hits, readouts, tracks, clusters, reconstructed particles),
 - Construct relationship between data objects (e.g. which particle makes these hits? Which hits are used to fit a track, etc.)
 - Provide automated memory management and data I/O functionalities
 - Provide backward and forward compatibility, very important for the long running of STCF.
 - Guarantee thread-safety, and provide high performance for MT applications



Event Data Model and of OSCAR



EDM classes defined in OSCAR

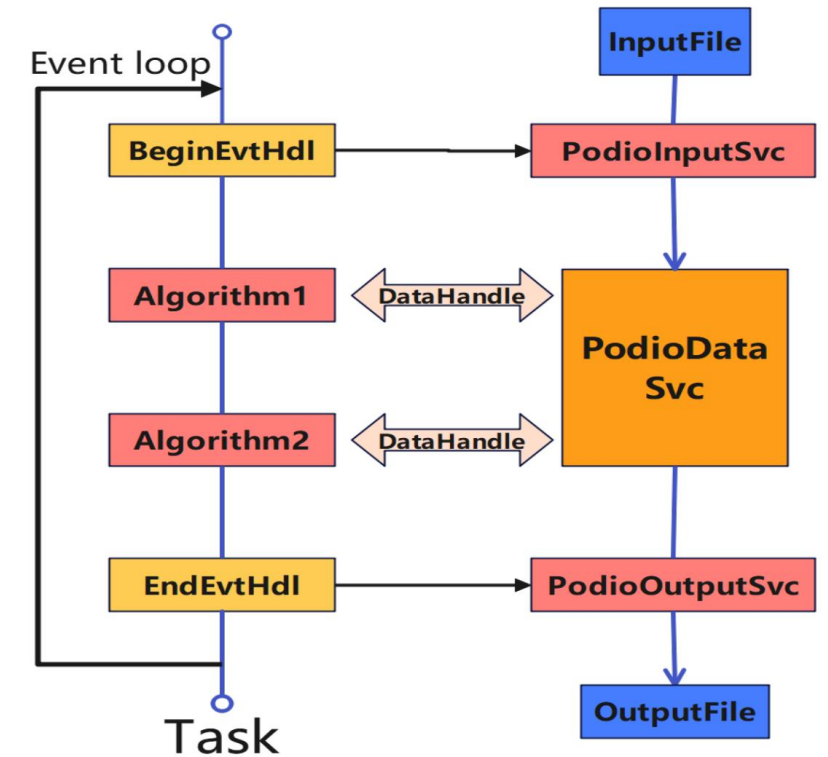
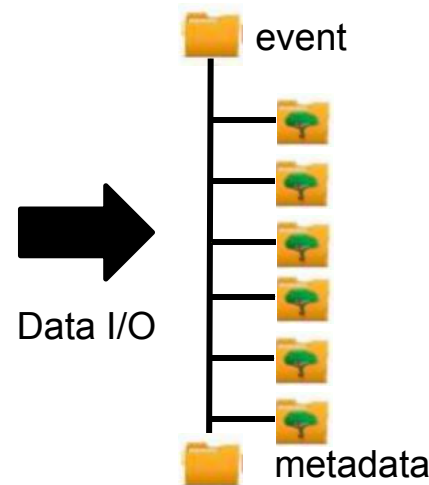
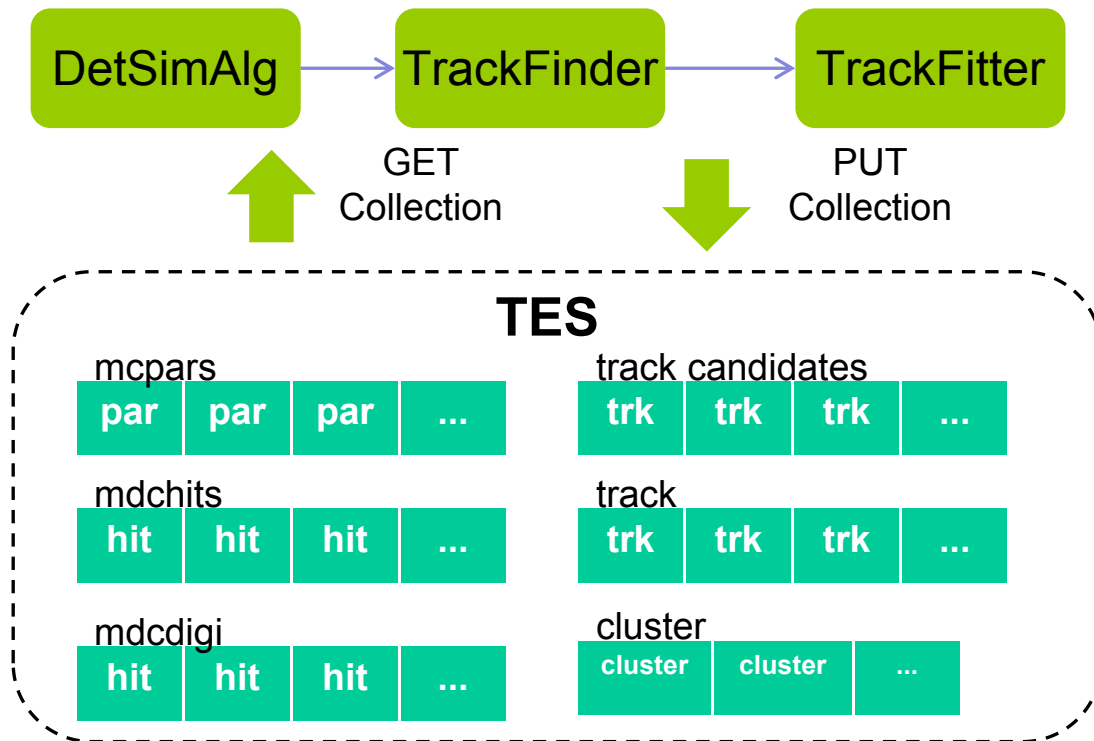


Based on YAML definition, generate EDM C++ code accordingly

Transient Event Store and Data I/O

❖ **Transient Event Store** (TES) is where EDM objects are stored in memory

- TES in OSCAR is developed based on podio::EventStore
- Being migrated to podio::Frame (code mostly ready)
- Support both serial and parallel applied software

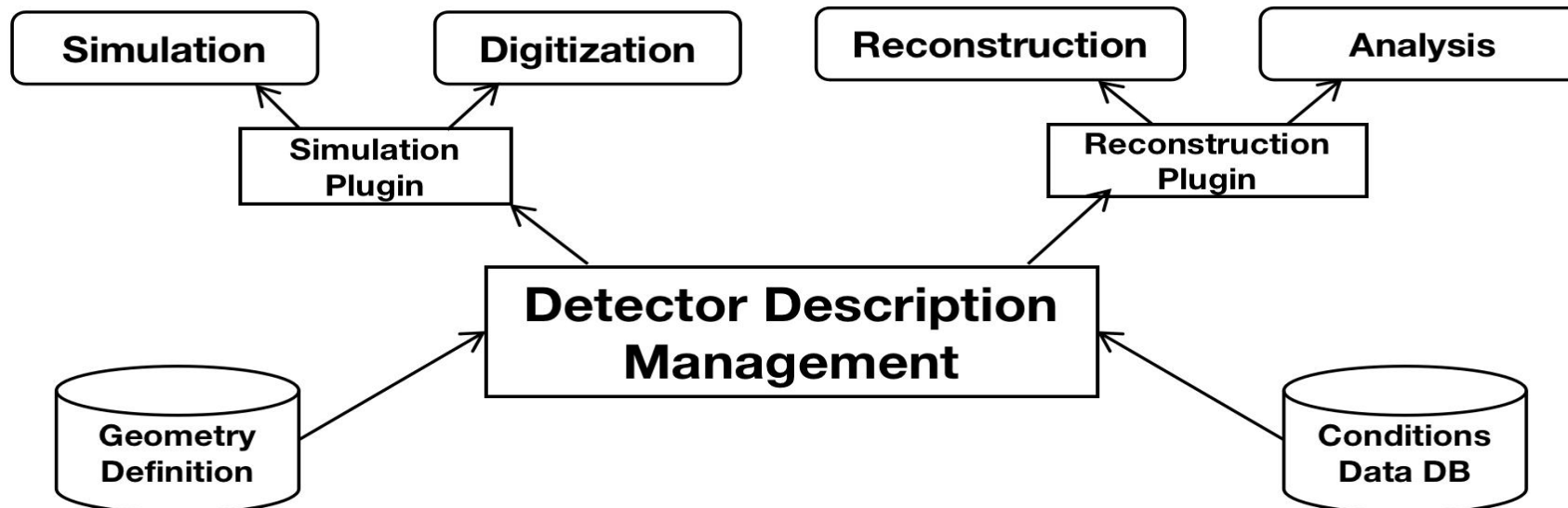


Implementation of TES and data I/O

- PodioDataSvc
- PodioInputSvc
- PodioOutputSvc

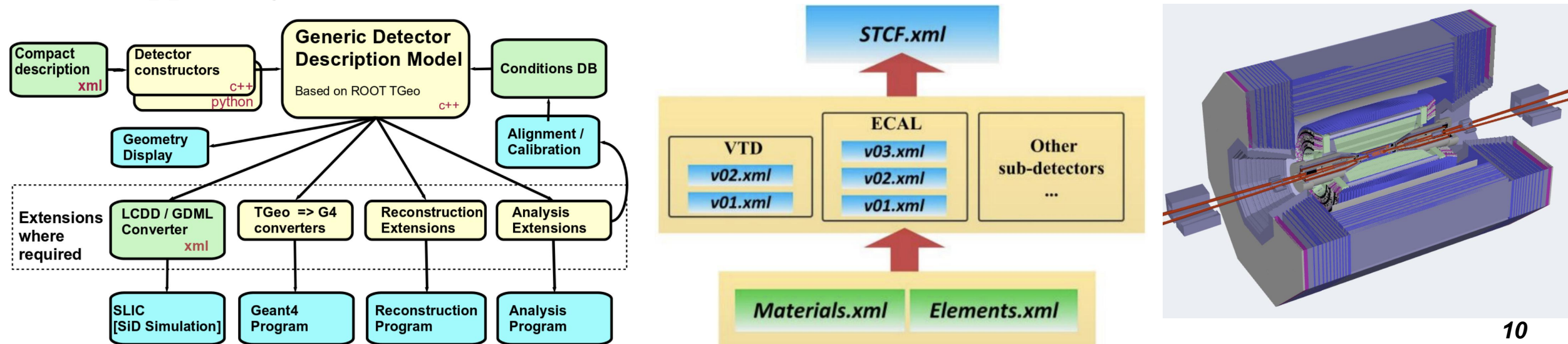
Detector Description Management: Requirements

- ❖ A powerful detector description management system is necessary across the full offline data processing workflow
 - Provide simple method for **geometry description definition**
 - Provide **consistent detector description** for all applications
 - Provide **geometry conversion** for different applications, and versioning management
 - Provide interface for **conditions data and detector alignment**
 - Provide simple and **ready-to-use interfaces** for applications



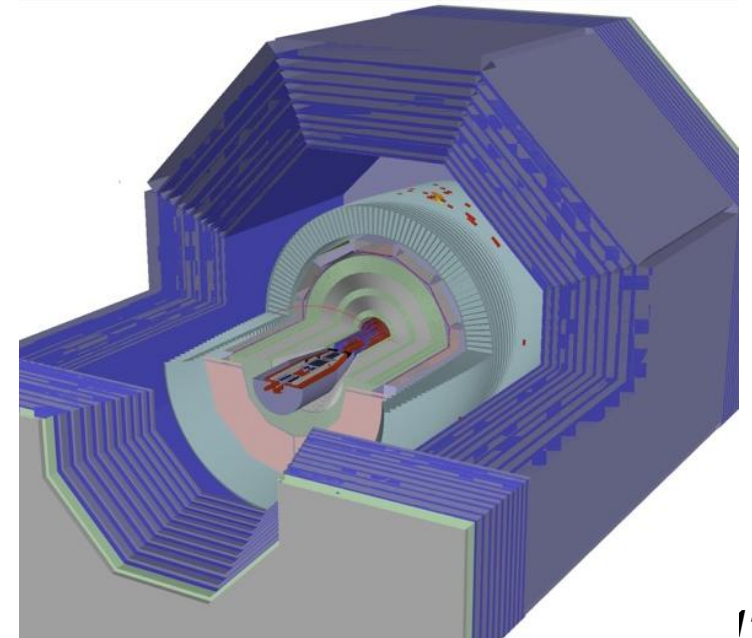
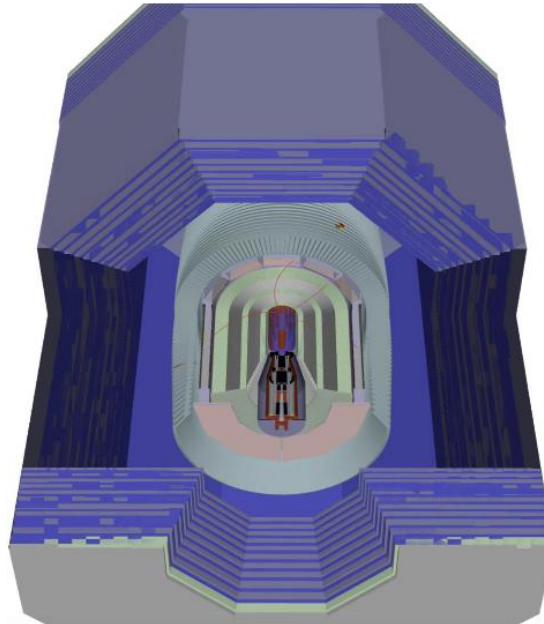
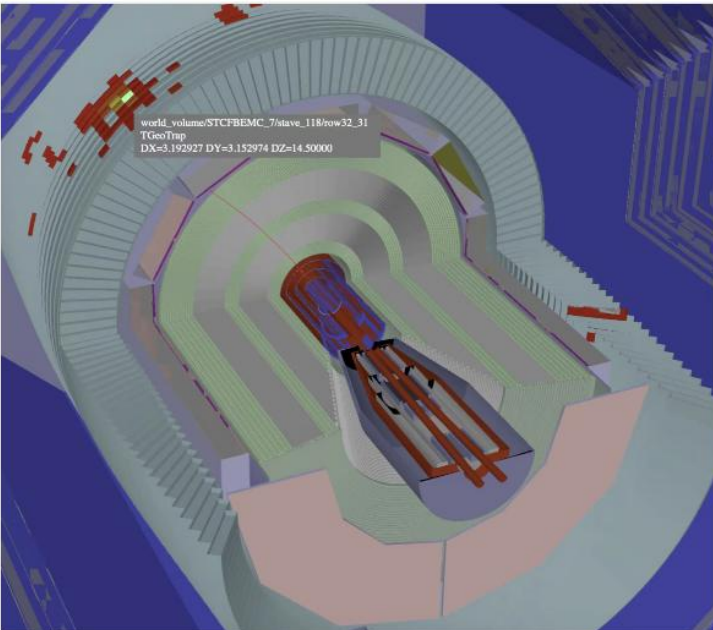
Geometry Management System

- ❖ Geometry Management System (GMS) in OSCAR is based on DD4hep
- ❖ Single source of detector information for detector description, simulation reconstruction and event display
 - Complete geometry defined with XML files and C++ parser
 - Various plugins for applications
 - Interface for alignment and conditions data
- ❖ Full detector defined and stably used, now being further refined (e.g. implementing supporting structures)



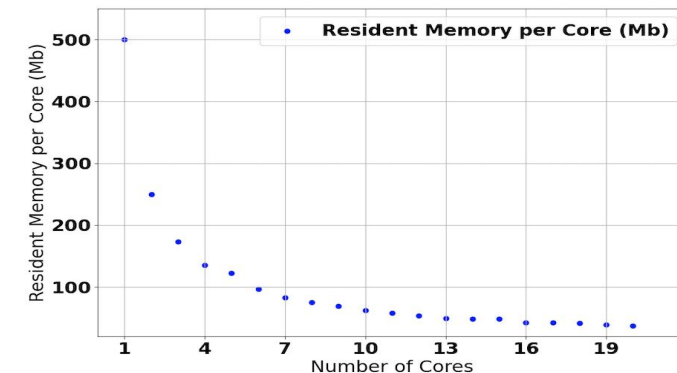
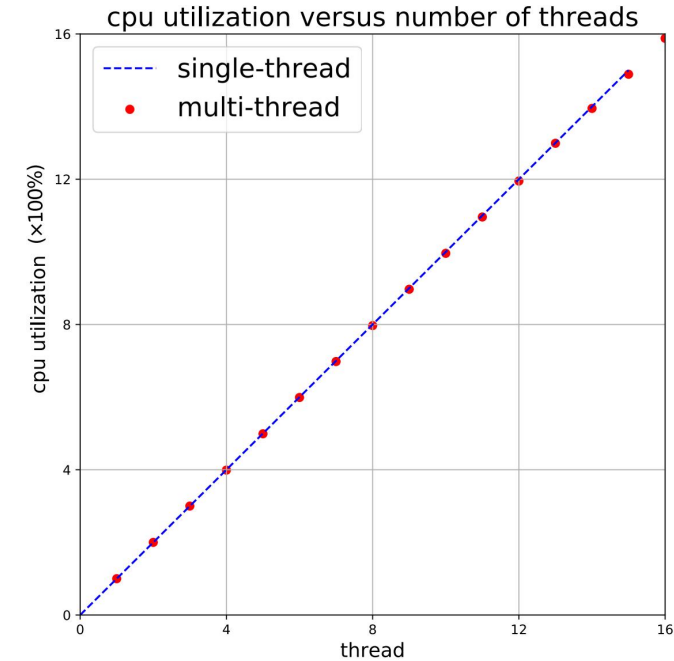
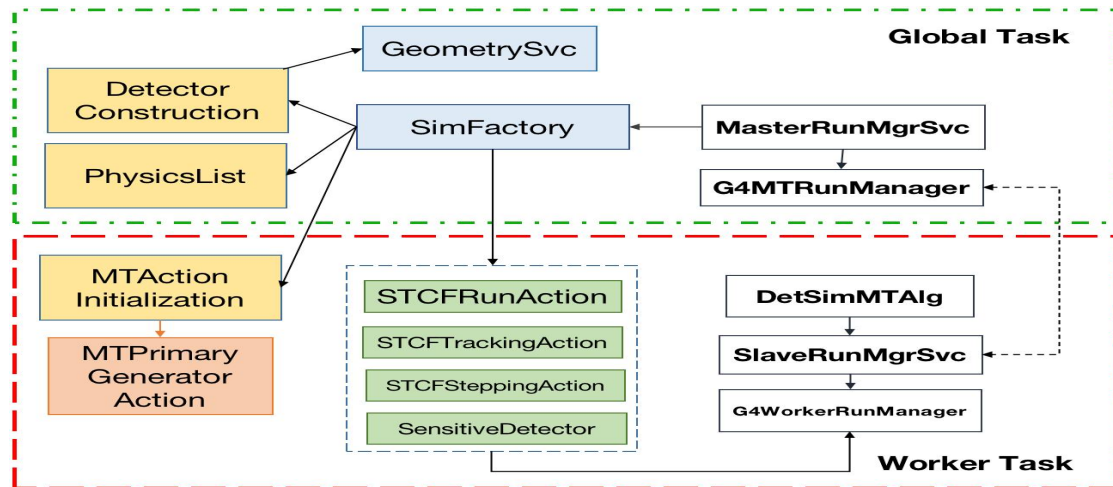
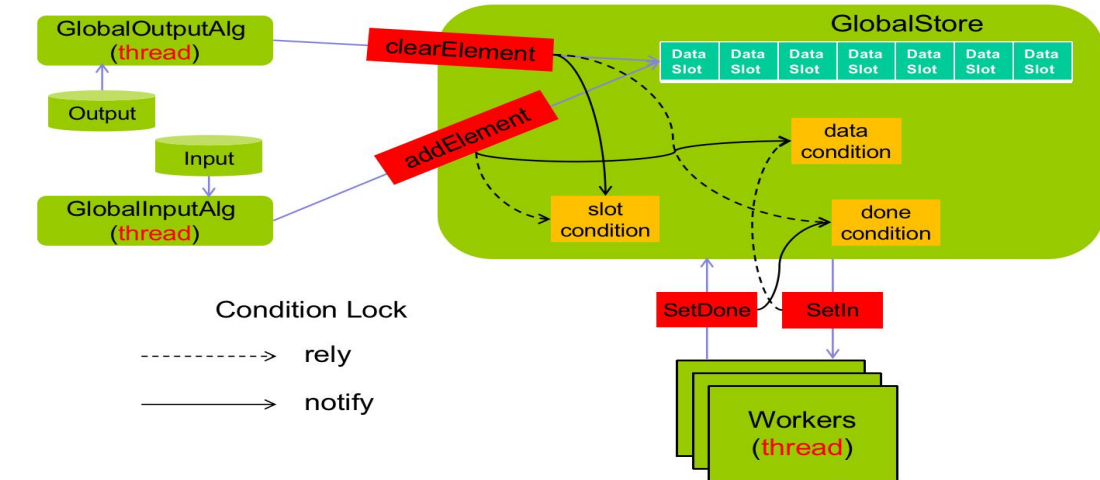
Detector and Event Display

- ❖ A common geometry and event display system is being developed
 - Based on Web3D technology and the open-source JSRoot framework
 - 3D engine and graphic library based on Three.JS
 - Geometry information from detector description from DD4hep (XML), and event data from podio
 - Major updates in 2025, now supporting the latest event data format (display of tracker and ECAL hits, tracks and showers are supported)



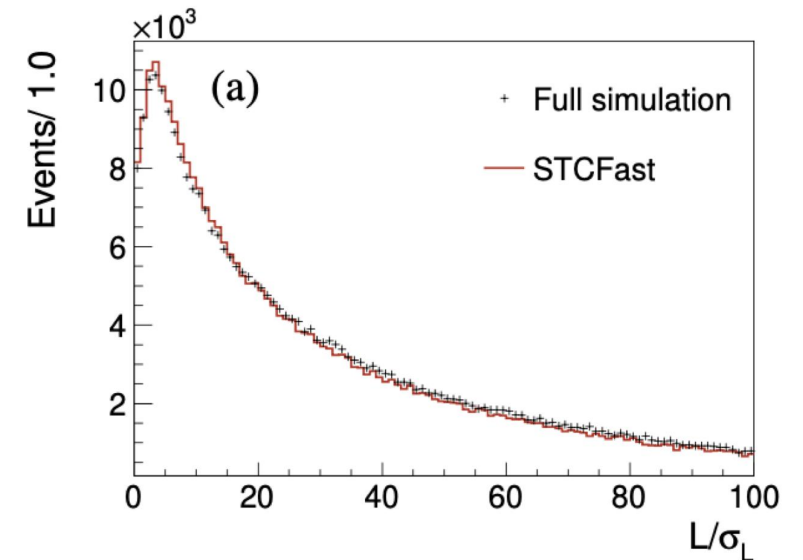
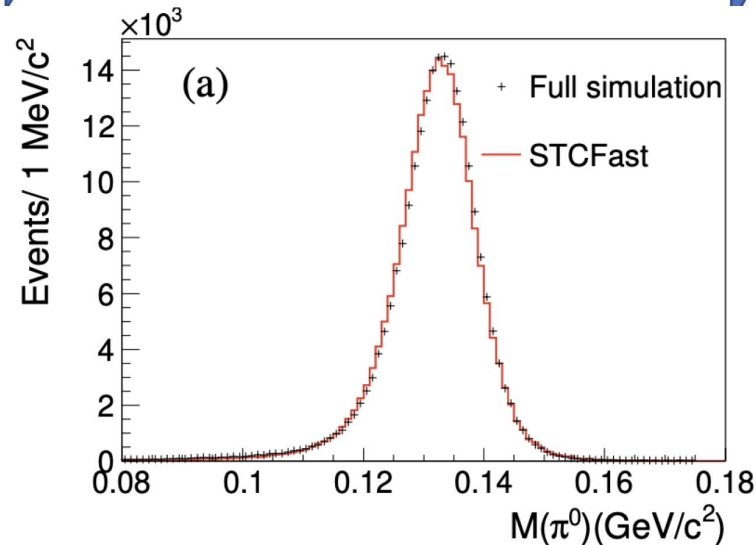
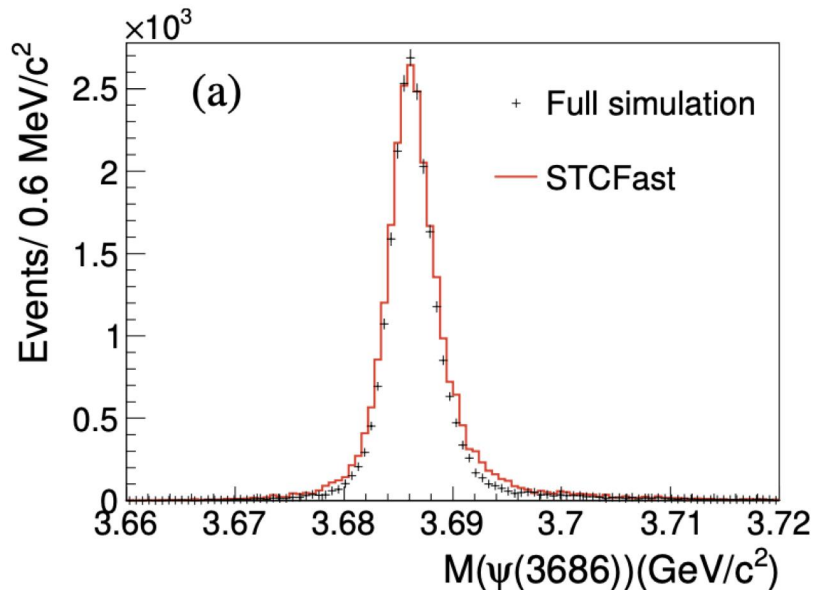
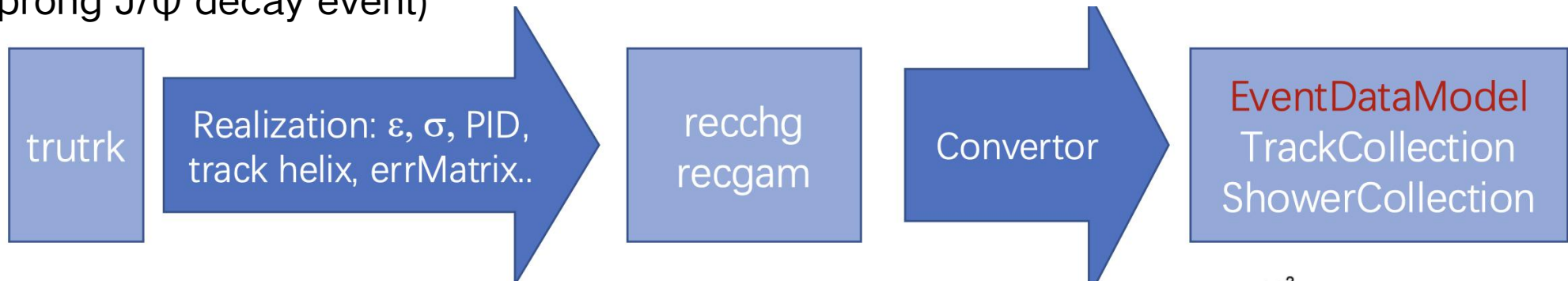
Parallelized Data Processing

- ❖ Parallelized detector simulation and reconstruction applications are implemented
 - Basic performance tests show promising scalability



Fast Simulation Framework

- ❖ The fast simulation framework is now integrated with OSCAR (more features being developed)
- ❖ Flexible for different detecting response and friendly for physics sensitivity study
- ❖ Much faster and less disk storage consuming compared to full simulation (~2ms per event, ~2kb storage for a 4prong J/ ψ decay event)



Machine Learning Support

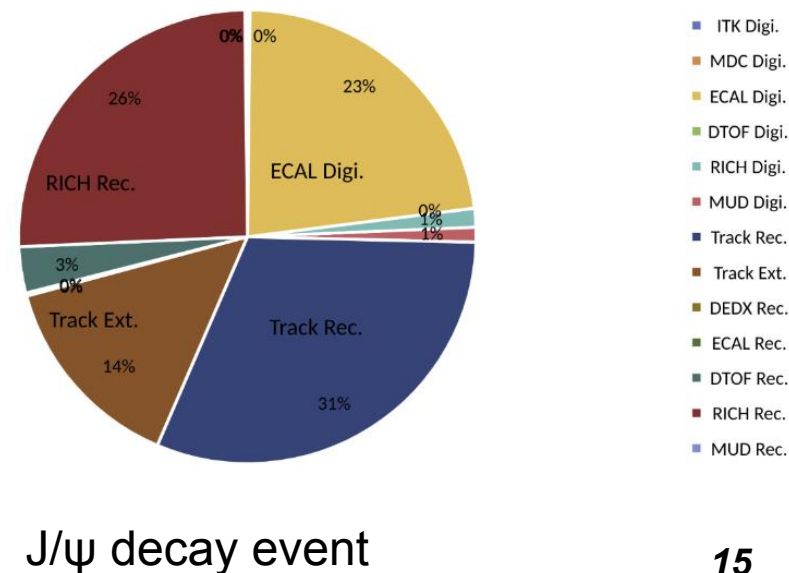
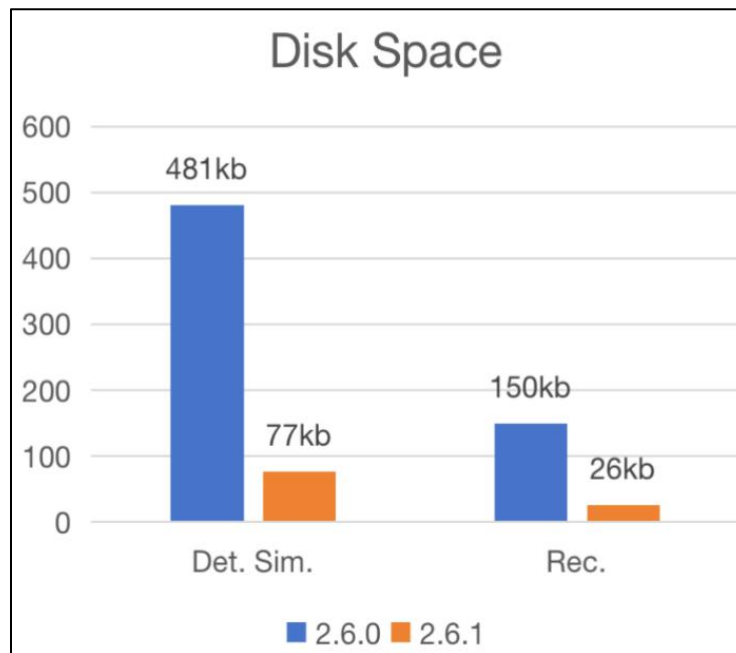
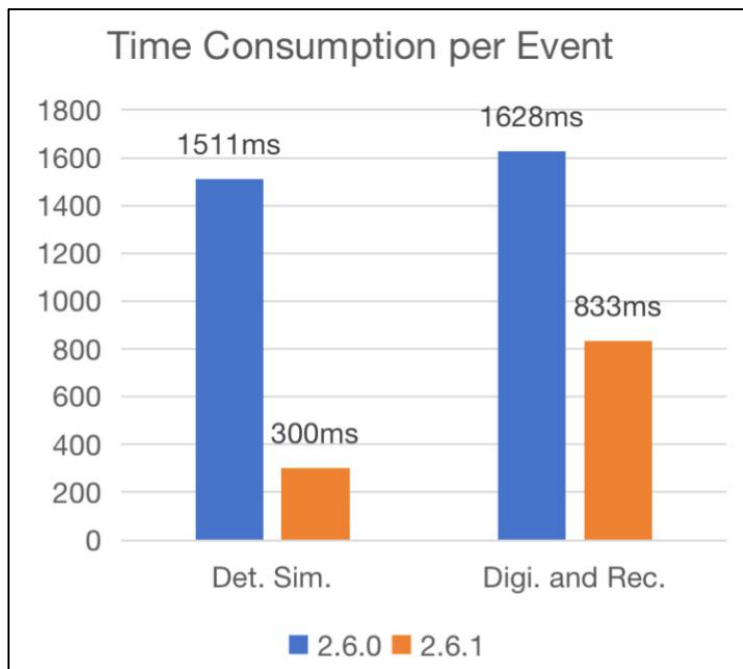
- ❖ Various applications in OSCAR (being) developed using ML techniques. Integrating trained model, with the data processing chain properly is vital
- ❖ ONNX Runtime is provided for ML model inference
 - Convert ML models to common middleware representation and embedded into OSCAR offline data processing
 - Deep-learning framework agnostic
 - Support inference on both GPU and CPU
 - Now being used in DTOF CNN-based PID algorithm
 - Being applied to GNN-based noise filtering

```
bool OrtInferenceAlg::initialize() {  
  
    m_env = std::make_shared<Ort::Env>(ORT_LOGGING_LEVEL_WARNING, "ENV");  
    m_session_options = std::make_shared<Ort::SessionOptions>();  
    m_session_options->SetIntraOpNumThreads(m_intra_op_nthreads);  
    m_session_options->SetInterOpNumThreads(m_inter_op_nthreads);  
  
    m_session = std::make_shared<Ort::Session>(*m_env, m_model_file.c_str(), *m_session_options);  
}
```

```
Ort::MemoryInfo info("Cpu", OrtDeviceAllocator, 0, OrtMemTypeDefault);  
  
auto input_tensor = Ort::Value::CreateTensor(info,  
                                              inputs.data(),  
                                              inputs.size(),  
                                              dims.data(),  
                                              dims.size());  
  
std::vector<Ort::Value> input_tensors;  
input_tensors.push_back(std::move(input_tensor));  
  
auto output_tensors = m_session->Run(Ort::RunOptions{ nullptr },  
                                     m_input_node_names.data(),  
                                     input_tensors.data(),  
                                     input_tensors.size(),  
                                     m_output_node_names.data(),  
                                     m_output_node_names.size());  
  
for (int i = 0; i < output_tensors.size(); ++i) {  
    LogInfo << "[" << i << "]"  
            << " output name: " << m_output_node_names[i]  
            << " results (first 10 elements): "  
            << std::endl;  
    const auto& output_tensor = output_tensors[i];  
    const float* v_output = output_tensor.GetTensorData<float>();  
  
    for (int j = 0; j < 10; ++j) {  
        LogInfo << "[" << i << "]" << "[" << j << "]"  
                << v_output[j]  
                << std::endl;  
    }  
}
```

Software Optimization

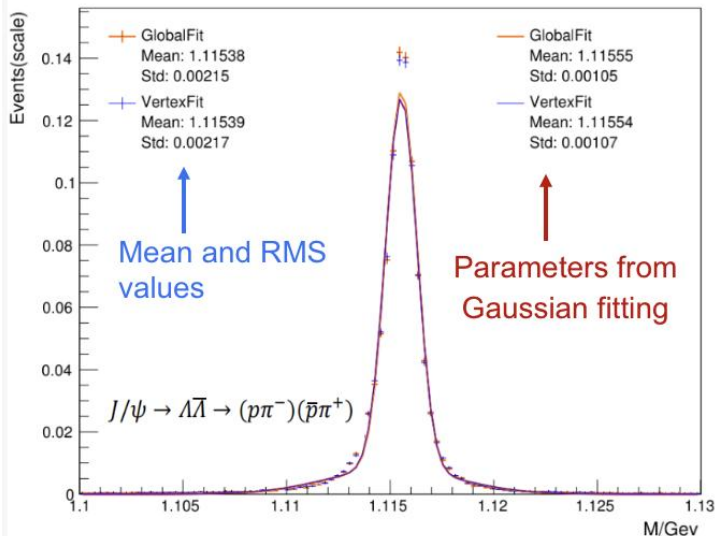
- ❖ Towards massive data production for the TDR, OSCAR is greatly optimized, in terms of the execution speed and output data volume
 - Lots of optimization of simulation, digitization and reconstruction algorithms (further optimization is still being performed)
 - Optimization of event data model (removal of redundant information, using more efficient data types, etc.)
 - Performance is comparable to BESIII now



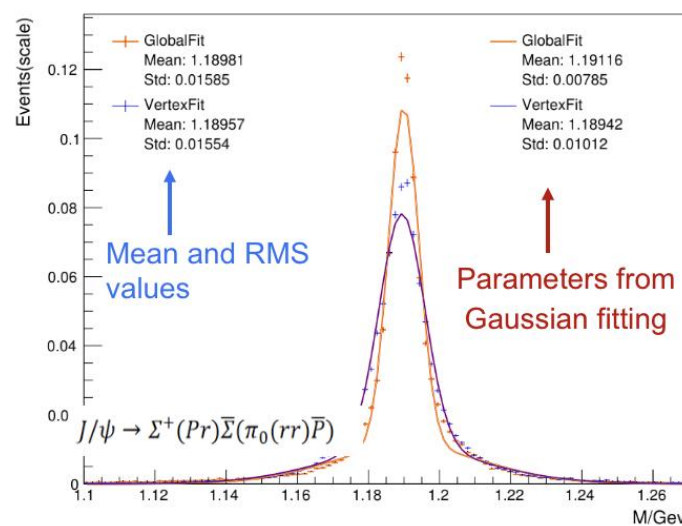
Data Analysis

- ❖ GlobalFit package designed for STCF based on the tree fitting algorithm of Belle II, showing better performance than VertexFit imported from BESIII

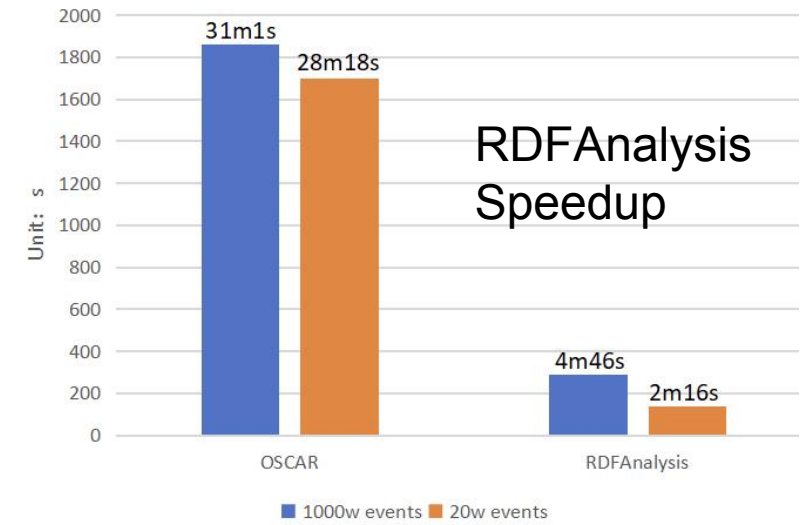
Invariant mass of Λ



Invariant mass of Σ^+



running time comparison



- ❖ RDataFrame-based analysis framework keeps being enriched and tested
 - Physical analysis results are consistent using $J/\Psi \rightarrow \Lambda \bar{\Lambda}$
 - Running speed significantly improved using parallel computing technique

Other Development Activities

- ❖ New features and updates of OSCAR being developed:
 - **Analysis Event Data Model** is being developed
 - Skimmed data on full reconstruction EDM (like BESIII DST)
 - Greatly simplify analysis, and reduce disk storage burden
 - Initial design was done
 - Software deployment based on **Spack**
 - As a multi-platform package manager that builds and installs multiple versions and configurations of software, allowing flexible management of various external libraries
 - **AlmaLinux9** support
 - Now OSCAR runs in CentOS7 similarity containers, updating to el9 is being performed
 - Fast calorimeter simulation based on **GAN**
 - Can greatly reduce computation resource consumption for MC production

Summary

- ❖ We introduced the basic design and functionalities of STCF core software
- ❖ Based on the core components, STCF full simulation and reconstruction chain has been established
- ❖ A dedicated OSCAR tutorial is performed during 2025 February
 - Including how to simulate and reconstruct data, how to perform data analysis, and how to develop new algorithms in OSCAR
 - A lot of physics analyzers are now involved, using OSCAR to perform physics studies
 - OSCAR has been improved greatly since then (thanks a lot to all the feedbacks)
 - TDR preparation has begun based on OSCAR
- ❖ We have been continuously improving the core software
 - To improve the software and physics performance