



Workload Management System for SPD Online filter

VII SPD Collaboration Meeting. 23.05.2024 Nikita Greben, MLIT

SPD Online Filter as a middleware software



«SPD OnLine filter» – hardware and software complex providing multi-stage high-throughput processing and filtering of data for SPD detector.

Data management system

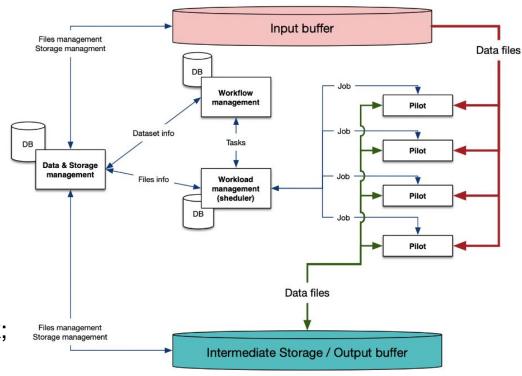
Data lifecycle support (data catalog, consistency check, cleanup, storage);

Workflow Management System:

Define and execute processing chains by generating the required number of computational tasks;

Workload management system:

- Create the required number of jobs to perform the task;
- Dispatch jobs to working nodes via pilots;
- Control job execution;
- Pilot control (identification of "dead" pilots);
- Efficient resource management;

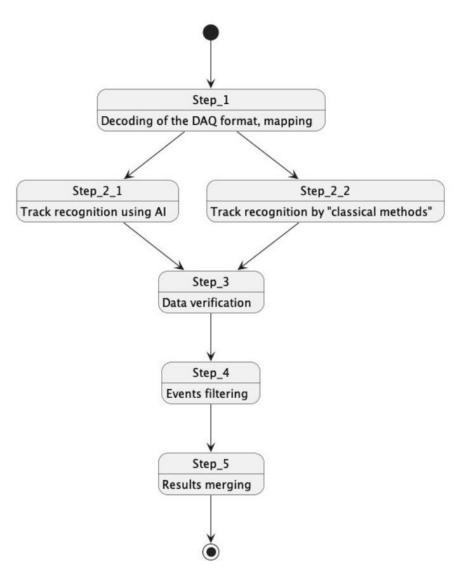


Architecture of SPD Online Filter

High-throughput computing



- ➤ HTC is defined as a type of computing that simultaneously executes numerous simple and computationally independent jobs to perform a data processing task.
- Since each data element can be processed simultaneously, this can be applied to data aggregated by a data acquisition system (DAQ).
- To ensure efficient utilization of computational resources, data processing should be multi-stage:
 - One stage of processing → task
 - Processing a block of data (file) → job



Task and job definition



- A task is a workload unit responsible for processing a block of homogeneous data dataset.
- A processing request is a set of input data, which may consist of multiple files, and a handler.
- > The completion criterion of the **task** is the processing of the data block.
- The **Workflow Management System** is responsible for defining and executing workflows, as well as defining a processing request, which is a **task**.

- > A job (payload) is a unit of work that processes a unit of data (file).
- The unit responsible for processing a single file in terms of workload is called a job.
- The **Workload Management System** (WMS) is responsible for generating jobs, sending them to compute nodes, and executing them.

Workload management system requirements



The key requirement - systems must meet the high-throughput paradigm.

- ❖ Task registration: formalized task description, including job options and required metadata registration.
- Jobs definition: generation of required number of jobs to perform task by controlled loading of available computing resources.
- Jobs execution management: continuous job state monitoring by communication with pilot, job retries in case of failures, job execution termination.

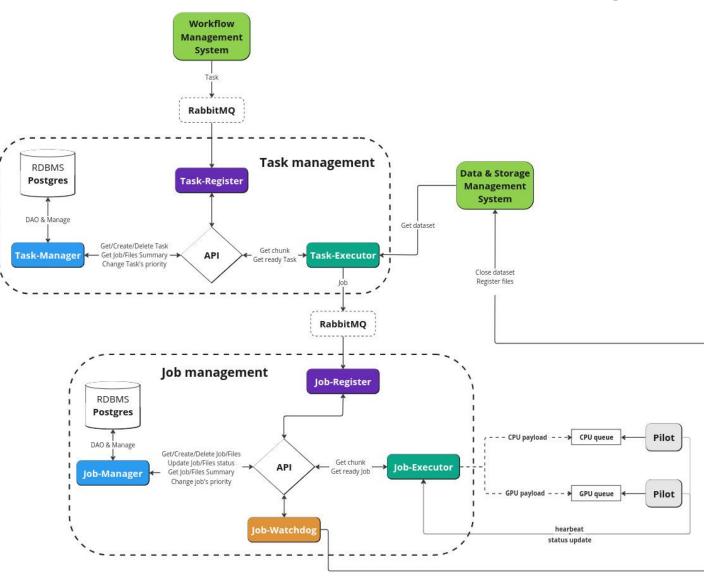


Forming jobs based on dataset contents, one file per one job

Architecture and functionality of Workload Management System



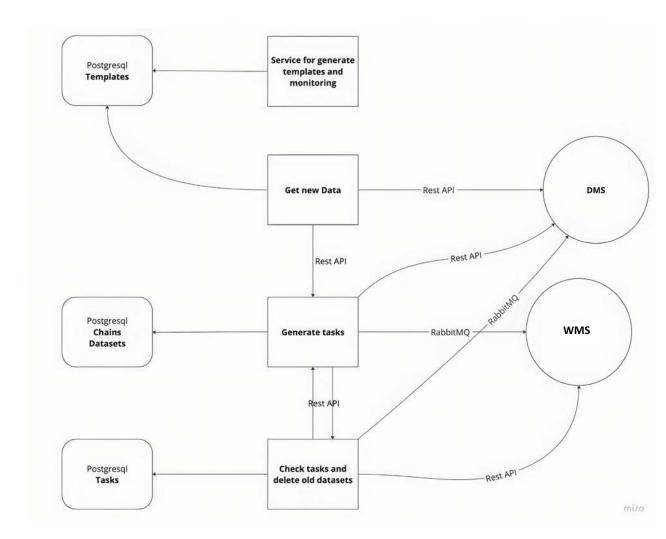
- task-manager implements both external and internal REST APIs. Responsible for registering tasks for processing, cancelling tasks, reporting on current output files and tasks in the system.
- task-executor responsible for forming jobs in the system by dataset contents.
- job-manager accountable for storing jobs and files metadata, as well as providing a REST API for the executed jobs.
- job-executor responsible for distribution of jobs to pilot applications, updating the status of jobs, registering output files and closing the dataset.
- pilot responsible for running jobs on compute nodes, organizing their execution, and communicating various information about their progress and status.



Interaction with the Workflow Management System



- Registration of a task for processing
 - WfMS passes the task description into message queue
- Summary of current intermediate properties of jobs/files in the system
 - Aggregated information about the status of each job/file for further decision making
- > Task cancellation
 - Based on the decision made on the WfMS (too many errors occuring) or operator side
- Change priority of a task
 - Control management



Interaction with the Data Management System

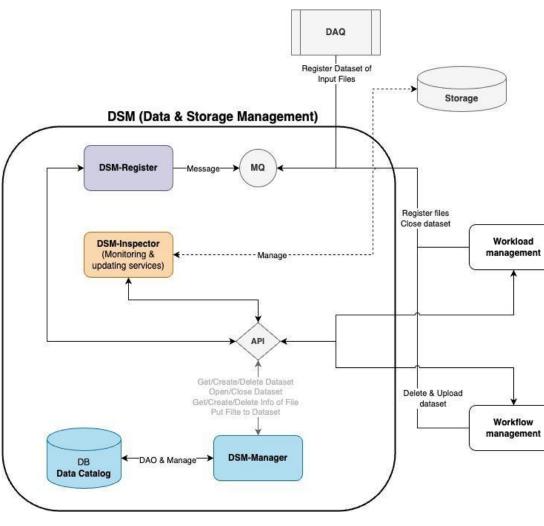


Routing Key	Msg	Algo
dataset.close	Dataset info Dataset UID File check list (file names)	Request the registered files in the dataset. If they match the checklist, set the status to CLOSED . Otherwise, return the messages back to the queue for deferred execution.
dataset.upload	Dataset UID	Marking dataset for uploading (TO_UPLOAD)
dataset.delete	Dataset UID	Marking dataset for deletion (TO_DELETE)

Signature and algorithm of message receiving gateways for the dsm-register service

Within a **Workload Management System**, there are several scenarios for interacting with the data management system:

- Obtain information about dataset contents for forming jobs from
 DSM-Manager (Data Catalog REST API)
- Register files in datasets after executing payload on compute
 node DSM-Register (Data Registration)
- Close dataset after cancellation or sufficient number of successfully processed files – DSM-Register



Architecture of Data Management

Database design

RDBMS - PostgreSQL 16

Tables:

- alembic_version managing and tracking database schema changes
- file_dat a directory specifying the output files and logs generated on the pilot
- job_dat jobs currently being processed in the system
- task_dat current tasks in the system

Extra mechanisms:

- Indexes on filter fields for optimization of operations
- Procedures task and job generation for test purposes
- Triggers rank update logic
- Decomposition single database per microservice







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public

pilot_dat

job_id

device_type

created_at

updated_at

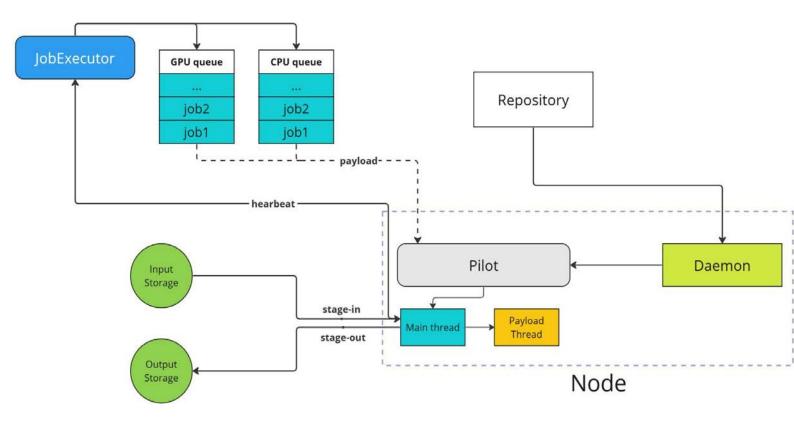


ER Diagram of the Workload Management System Database

Internal design of Pilot Agent

- The agent application is deployed on a compute node and consists of the following two components: a UNIX daemon and the pilot itself.
- The UNIX daemon's objective is to run the next pilot by downloading an up-to-date version from the repository.
 - Pilot itself is a multi-threaded Python application responsible for
 - Receiving and validating jobs from the message broker.
 - Downloading input files for the payload stage and uploading the result files to the output storage.
 - Launching a subprocess to execute a payload (decoding DAQ format, track recognition algorithm, etc.)
 - Keeping the upstream system informed of the current status of the payload and the pilot itself via heartbeat/status updates during each phase of pilot execution.





- Compute nodes differ only in the availability of specialized co-processors (GPUs) and are assigned to the appropriate message broker based on the computational needs of the job.
- Regardless of the presence of an error, when the pilot finishes, the UNIX daemon launches a new instance of the pilot.

Interaction with the Pilot Agent

NICA

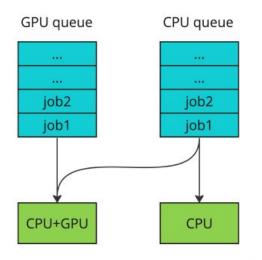
- Pilot has a series of prepossessing stages before running a job itself:
 - a. start logging
 - b. read configuration
 - c. getting a job from message queue
 - d. validation
- After those steps the Pilot launches another thread where it does
 - a. environment setup script
 - b. copying files locally from the input storage
 - c. starts execution of a job itself in a separate sub-process
 - d. analysis of the result of a job
 - e. copying output data and logs to storage
 - f. sends regular messages to **WMS**
 - g. cleaning up the local environment
- Pilot sends status-update message at any point of internal changes
- **WMS** may terminate the job if the corresponding task is cancelled or if an error occurs.
- A detailed job status model has been described
- Error codes introduced
- Pilot ran through all major stages of the job execution (DAG)
- > Pilot at this stage runs a script that does a basic hash compute
- > Further debugging needed

Two communication channels:

- HTTP (aiohttp)
- AMQP (message broker RabbitMQ)

Two types of nodes:

- Multi-CPU
- Multi-CPU + GPU



Tech stack



 Common ➤ Python 3.12 ➤ docker compose - running multi-container applications 	Frameworks ➤ aio-pika (RabbitMQ + asyncio) - asynchronous API with RabbitMQ ➤ FastAPI + uvicorn
 DB ➤ PostgreSQL - RDBMS ➤ Alembic (Migration) ➤ SQLAlchemy 2.0 ➤ asyncpg - Postgres DBAPI 	 Extra ➤ aiohttp - asynchronous HTTP client/server framework ➤ Pydantic - validate and serialize data schemes ➤ pytest-asyncio - test purposes

Key results



Design of services:

- Implemented a mechanism for declaring the data model in the database based on ORM and migration scripts;
- Designed and implemented a list of required REST API methods and their signatures;
- > Configured CD tools (build and deployment) on the JINR LIT infrastructure;
- Designed inter-service interaction scenarios;
- Redesigned Pilot internal architecture;

Prototype of services:

- > Run through all job execution state model, debugging interactions with the pilot;
- Job management subsystem is the most advanced: most interactions implemented and being tested, job-watchdog microservice is being developed;
- > Pilot is in active stage of development (Leonid Romanychev SPbU).

Next major steps



Task processing

- Implementing task-partitioning algorithm.
- Closing datasets for DSM.
- Execute the entire workchain set up on the level of WfMS.

Logging

- Currently, each microservice's logs are mapped to the host via a shared file system between Docker and the host.
- Ideally ELK (Elastic-Logstash-Kibana) stack to build a log analysis platform.

Configuration

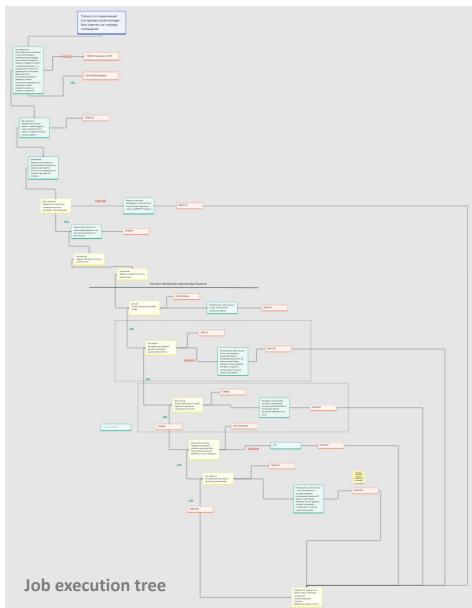
 Consider to centralize some of the shared configurations across multiple services.

Documentation

 Given the increasing complexity of the internal logic of the software, it is necessary to document each step of the development, for example: job state model could grow in complexity.

Metrics and monitoring

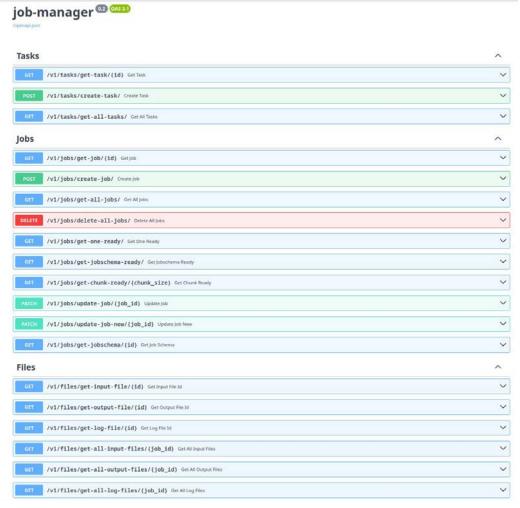
- For example, service query-per-second, API responsiveness, service latency etc.
- InfluxDB, Prometheus, Graphana.

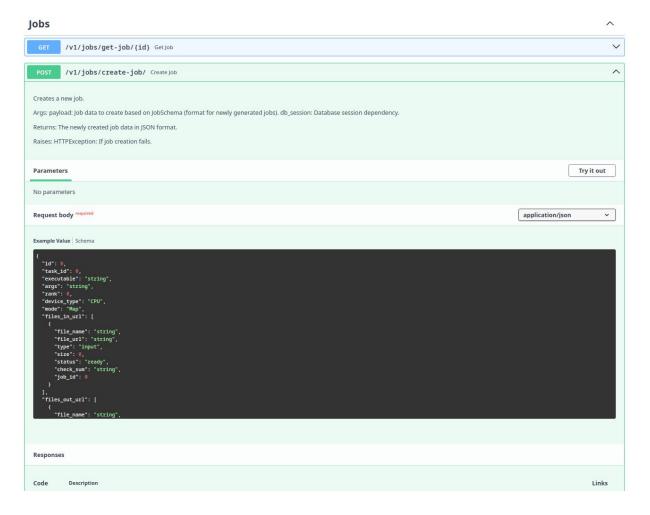


Prototyping Job-Manager (API)



- The chosen framework for building the service is FastAPI + Uvicorn asynchronous framework
- A basic set of CRUD operations on data in the form of REST API is developed.
- API description autogeneration according to OpenAPI 3.0 specification is implemented (available in Swagger UI at <server address>/docs)

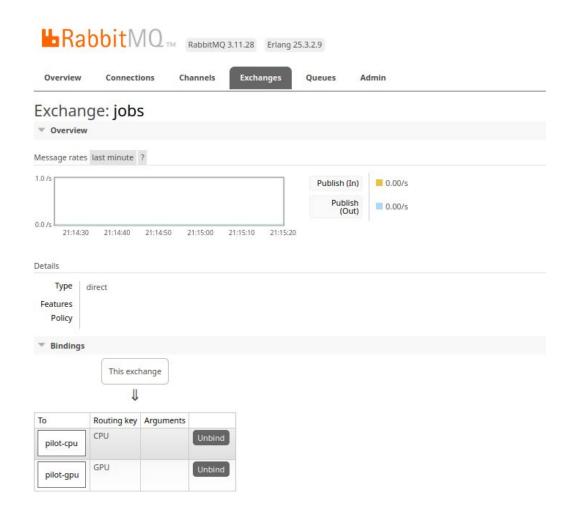


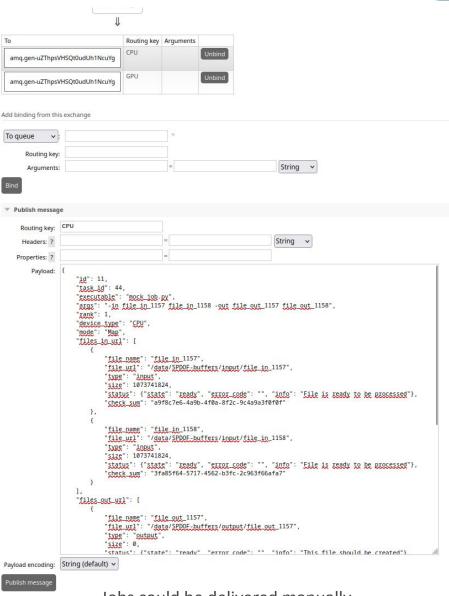


Prototyping Job-Executor - Pilot (RabbitMQ queues)



- RabbitMQ is selected as the message broker
- Queues are defined using the declarative notation of the aio-pika tool
- At the start of the application their unfolding is performed





Jobs could be delivered manually

Summary



We have designed the components of the **Workload Management System**, taking into account the characteristics and internal requirements of both the **WFMS** and **DSM** systems.

Our goal is to complete the prototyping phase and fully integrate with the application layer components of the «SPD On-Line Filter» platform.

Current plans:

- Run a simple data flow: create dataset, define a task, propagate through WMS, register files, close dataset
- Deploy pilot on multiple machines

Plans for the year:

- Defining and implementing obvious data processing pipelines;
- Debugging basic algorithms and external interfaces;
- Work out integration with application software and test on SPD-DAQ modelled data.



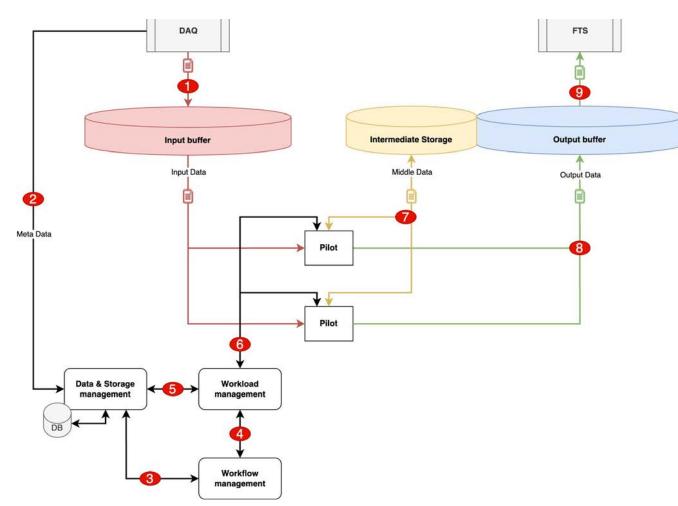
Thank you for your attention!

Dataflow and data processing concept



Main data streams:

- SPD DAQs, after dividing sensor signals into time blocks, send data to the SPD Online Filter input buffer as files of a consistent size.
- The workflow management system creates and deletes intermediate and final data sets
- The workload management system "populates" the data sets with information about the resulting files
- At each stage of data processing, pilots will read and write files to storage and create secondary data



Modularization: deploying and using own packages



Following tools are used

- Poetry
 - Particularly good at handling complex dependency trees and ensuring that the different modules can integrate with each other without version conflicts
- Python packages
 - separate GitLab repositories for each package
 - Poetry for packaging and dependency management
- ❖ Gitlab
 - Access Tokens used as kind of credentials for scripts and other tools
 - > CI/CD for automate testing and building



wms-schema is a package that contains a scheme for task and job data that is used in almost every other service

R&D



- Jobs scheduling (algo)
- Partitioning of a task
 - Imagine a multitasking operating system.
 - Each dataset represents a process, and each record within a dataset is like a thread within that process.
 - The algorithm acts as the operating system's scheduler, allocating processing time to threads based on their priority.
- Chunk size and rank/priority of a job as a basic control unit:

$$rank_{i+1} = \alpha \times x_i + \beta \times y_i + \gamma \times rank_i$$

```
x_i - aging, y_i - retries
```

```
Algorithm 1 Task Scheduling Algorithm
  Variables:
  global_queue - global queue with tasks
  dataset - array of datasets
  N – number of datasets
 rank_max - maximum task priority
 heap – binary heap storing maximum task priorities
 rank – array with task priorities
 Algorithm:
 1: initilize_datasets(dataset)
 2: build_heap(rank)
 3: while true do
     rank_max = heap.top()
     for r = 1 to rank_max do
       for i = 1 to N do
 6:
          if not dataset[i].chunk.empty() and rank[i] \geq r then
 7:
            await dataset[i].chunk.cur_item
 8:
            update(dataset[i].chunk - i.cur_item)
 9:
          else if dataset[i].chunk.empty() then
10:
            if dataset[i].chunk.cur_item then
11:
              dataset[i] = global\_queue.head()
12:
            end if
13:
            update(rank[i])
14:
            update(heap)
15:
          end if
16:
        end for
17:
     end for
18:
19: end while
```