Pyrate v2: a software system for data transformation, event reconstruction and analysis at the SABRE experiment

Federico Scutti

Swinburne University of Technology, John St, Hawthorn VIC 3122, Melbourne, AU

E-mail: fscutti@swin.edu.au

Abstract. The Pyrate system has been previously developed and recently updated by the SABRE experiment for direct dark matter detection. This framework, developed in Python, provides a dynamic, versatile, and memory-efficient approach to data format transformations, object reconstruction and data analysis in particle physics. Pyrate relies on a blackboard design pattern where object dependencies are dynamically evaluated throughout a run where a central control unit manages root nodes. The framework intends to improve the user experience, portability and scalability of offline software systems currently available in the particle physics community, with particular attention to medium to small-scale experiments.

1. Introduction

The Pyrate software system [1, 2] has been developed and is being used by the SABRE experiment for dark matter direct detection [3]. This software framework enables the transformation of data formats, reconstruction of events and data analysis at SABRE and has been developed to suit the need of small-scale experiments. The system is designed to make it easy to maintain, modular, and stable against many workflows for a small team of developers.

Pyrate is written entirely in Python [4], and provides all the necessary functionalities typically implemented in particle physics workflows by other larger and more complex systems [5, 6]. The software is currently hosted on the local Bitbucket repository of the SABRE collaboration. Experiment-specific algorithms have been implemented to transform the custom binary files of the SABRE DAQ system output into ROOT ntuples [7], to augment ROOT ntuples with higher-level event-reconstruction variables, and to achieve data analysis and plotting. The first version of the system Pyrate v1 has been previously presented [1, 2]. This document outlines the second version of the software, Pyrate v2, including updates to the core class structure. The following two sections will be dedicated to the structure and essential components of the system, and then illustrate the behaviour of the software at runtime.

2. The structure of Pyrate v2

Pyrate v2, as its previous version, implements a *blackboard design pattern* [8]. Different algorithms cooperate toward the computation of an object and share data using a Store. The control element in the pattern, the Run, creates dependencies between objects at runtime which are arranged in a directed acyclic graph. The main elements of the system and their relationships

are illustrated in the diagram in Figure 1. In the following, a summary of their essential functions is given.

- **Configuration** : Pyrate jobs are configured using YAML [9] files. A primary configuration file is used to specify input objects (Readers), output ones (Writers) and intermediate objects to be handled by a run instance and how many run instances will be managed in a single job. In the configuration files, each object O_i references other objects $O_j, ..., O_n$ as their inputs and declares its output variables, which other objects can use as inputs in turn.
- Job : The Job class creates and launches each run. In parsing configuration files, this class retrieves all input objects and creates run instances for each input separately. The Job class contains facilities for handling parallel tasks on generic batch systems, where different runs are executed in parallel.
- Run : The Run class is the control element in the blackboard design pattern. It is responsible for creating an instance of the blackboard, the **Store**, and creating one or more directed acyclic graphs where each head node corresponds to an output object declared in the configuration files. In the graph, each node corresponds to an object, and edges point to their dependencies as found in the configuration. The execution order of the algorithms computing the objects does not need to be explicitly declared in the configuration files. The Run resolves the computation by starting from the head node of the graph and recursively calling further dependencies until all objects necessary for the computation of the depending object are found in the **Store**. Circular dependencies are naturally prevented by using a directed acyclic graph which is implemented with the NetworkX python package [10]. All objects are re-evaluated by the Run each time new events are loaded from input files by a Reader object corresponding to the lowest nodes in the dependency graph.
- Store : The Store is the blackboard of the system. It is instantiated by the Run and implements
 the get(object_name) and put(object_name) methods to retrieve and store information.
 Pyrate algorithms call these functions during execution using a private Store instance. The
 Store is partitioned into two elements called the *Permanent* and *Transient* stores, where
 objects are never cleared or cleared after each event, respectively, during the execution of a
 program.
- Algorithm : All objects in Pyrate, including input Readers and output Writers, inherit from this class. The user is responsible for implementing three methods: initialise, execute and finalise. The initialise method is used to initialise basic algorithm data structures. The execute method is launched after updating the current event information, allowing the algorithm to access event-based data. The finalise method finalises the computation of the object after the event loop is over.
- Reader : A Reader is an object dedicated to reading a stream of files. If a Reader is used to provide event-based information, it inherits from both the Input and the Algorithm base classes, otherwise only from the latter. Different Readers are defined depending on different input file stream data formats. Each Reader inheriting from Input implements a get_event() function from its parent virtual class, which is dedicated to retrieving event information from the input file and putting it on the store. Event-based readers maintain an event_id identifier for the read event. The get_event() function is called by the Run for every event-loop iteration and before launching calls for evaluating objects in the directed acyclic graph. If the event can already be considered built into the input, then only one type of Reader is used. Otherwise, a special type of Reader is defined, called the EventBuilder, which is composed of several Readers managing different data streams.
- Input : The Input class is a virtual one, declaring methods to manage the retrieval of eventbased information from an input file. These methods are to be implemented by the derived Reader classes.

- Writer : A Writer is an object inheriting from an Algorithm and dedicated to writing its output to a file. It holds an instance of the OutputFileHandler class managing the state of output files. Since Writer objects are dedicated to the computation of the highest-level output type, they are the head node of the object dependency graph.
- OutputFileHandler : This class is dedicated to managing the state of all output files in a Run. It is instantiated by the Run and passed to all Writers which retrieve files via its get_file() method.

3. Runtime behaviour

A run collection is configured using a job configuration file, referencing object configurations. Based on the job file, a Job instance associates a single input to a Run instance which is first set up and then executed. The following two sections outline these two steps in detail.

3.1. Run setup

During the setup stage of the Run, instances are created for all objects declared in the configuration passed by the Job. In addition to this, an instance is created of the Store, the OutputFileHandler and the directed acyclic graph representing dependencies, using the NetworkX package [10]. Each graph node is then filled with object instances and connected to other nodes based on the input/output dependencies declared in the configuration. The head node of a graph always corresponds to Writer objects, while Readers occupy the end of the dependency chain. In addition to this, to each node, a boolean flag is added (was_called), indicating the execution status of an object within the current iteration, which will be updated when the three algorithmic states run and when events are read.

3.2. Run execution

A Run execution, qualitatively sketched in Figure 2, is performed in three stages, called *initialise*, *execute* and *finalise*, corresponding to the three methods implemented by all objects. The Run class also implements methods corresponding to these three states and calls them in order within the launch() function. Each method is responsible for evaluating the entire object dependency chain in multiple iterations where multiple evaluations are necessary during the *execute* method, the so-called event_loop(). Each iteration of the event_loop() proceeds by first calling the get_event() method of the Reader, expected to put event-based information on the Store, and then by evaluating the object dependency chain.

The loop() function evaluates the object dependency chain where head node objects are called in turn using the call(object_name) method. This recursive function checks the was_called status of the node for a given iteration. A loop over the node dependencies is performed if was_called==False. Dependencies are evaluated with call(dependency_name) executions in turn. This recursive strategy reaches the lowest level in the dependency chain, *i.e.* the Reader. At this point, fundamental event-based information has previously been computed by the get_event() function, which makes its output available on the Store. This allows, in turn, execution of the Reader dependent objects, and so forth, up to the highest head node level. After each iteration in this evaluation strategy, the was_called flag is reset, and a new evaluation is ready to proceed.

The recursive nature of the evaluation strategy described above allows for defining a mechanism where only a subset of dependency objects has to be evaluated under certain event conditions. For data analysis workflows, this allows one to handle event selection efficiently, where one does not need to evaluate all selection criteria downstream of a list if previous criteria fail. For those workflows not implementing a selection of events, an option is provided



Figure 1. Class structure of the Pyrate core.

for instructing the Run to arrange objects in an ordered list and execute call(object_name) according to that order.



Figure 2. Qualitative representation of the different stages of the execution of Pyrate. Objects are arranged in a directed acyclic graph and are evaluated in the three stages: *initialise*, *execute* and *finalise*. A Pyrate output corresponds to the graph head node.

4. Conclusions

The Pyrate v2 system has been described as corresponding to the second version of the software system developed and used by the SABRE collaboration for data transformations, event reconstruction and data analysis in particle physics. The Pyrate system currently supports custom binary input files and ROOT input and outputs. In the future, the addition of an extended variety of formats is foreseen, including HDF5 [12] and Parquet [13].

References

- Scutti F 2023 Pyrate: a novel system for data transformations, reconstruction and analysis J. Phys.: Conf. Ser. 2438 012061.
- [2] Scutti F 2022 Pyrate DOI:10.5281/zenodo.6257646.
- [3] Bignell L et al 2020 SABRE and the Stawell Underground Physics Laboratory Dark Matter Research at the Australian National University EPJ Web Conf. vol 232 (EDP Sciences) p 6.
- [4] Van Rossum G and Drake F 2009 Python 3 Reference Manual (CreateSpace).
- [5] Barrand G et al 2001 GAUDI A software architecture and framework for building HEP data processing applications Comp. Phys. Comm. 140 45-55.
- [6] Zou J et al 2015 SNiPER: an offline software framework for noncollider physics experiments J. Phys.: Conf. Series 664 p 072053.
- [7] Brun R and Rademakers F 1997 ROOT An Object Oriented Data Analysis Framework Nucl. Inst. & Meth. in Phys. Res. A 389 81-86.
- [8] Corkill D 1991 Blackboard Systems AI Expert 9 40–47.
- [9] Ingerson B, Evans C and Ben-Kiki O 2001 Yet Another Markup Language (YAML) 1.0.
- [10] Aric A et al Exploring network structure, dynamics, and function using NetworkX Proceedings of the 7th Python in Science Conference (SciPy2008) 2008 pp 11-15
- [11] Gamma E, Helm R, Johnson R and Vlissides J 1994 Design Patterns: Elements of Reusable Object-Oriented Software (Addison Wesley) pp 107.
- [12] Koranne S 2011 Hierarchical data format 5: HDF5 Handbook of Open Source Tools (Springer) pp 191-200.
- [13] Vohra D 2016 Apache Parquet Practical Hadoop Ecosystem (Apress) pp 325-335.