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A complete system integration of stream-based IP flow-record querier

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ABSTRACT

Short summary of the contents in English. . .

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ACRONYMS

IPFIX Internet Protocol Flow Information Export

HDF Hierarchical Data Format

LALR Look-Ahead LR Parser

PLY Python Lex-Yacc



Part I

INTRODUCTION

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TRAFFIC MEASUREMENT APPROACHES

- 1.1 CAPTURING PACKETS
- 1.2 CAPTURING FLOWS
- 1.3 REMOTE MONITORING
- 1.4 REMOTE METERING



FLOW EXPORT PROTOCOLS

- 2.1 NETFLOW
- 2.2 IPFIX
- 2.3 SFLOW



LANGUAGES AND TOOLS

- 3.1 SQL-BASED QUERY LANGUAGES
- 3.1.1 *NetFlow exports as relational DBMS*
- 3.1.2 Data Stream Management System
- 3.1.3 Gigascope
- 3.1.4 Tribeca
- 3.2 FILTERING LANGUAGES
- 3.2.1 *flow-tools*
- 3.2.2 *nfdump*
- 3.3 PROCEDURAL LANGUAGES
- 3.3.1 FlowScan
- 3.3.2 Clustering NetFlow Exports
- 3.3.3 SiLK Analysis Suite



LEGAL CONSIDERATION



Part II

STATE OF THE ART

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Flowy [2][3] is the first prototype implementation of a stream-based flow record query language [4][1][5]. The query language allows to describe patterns in flow-records in a declarative and orthogonal fashion, making it easy to read and flexible enough to describe complex relationships among a given set of flows.

5.1 PYTHON FRAMEWORK

Flowy is written in Python. The framework is subdivided into two main modules: the validator module and the execution module. The validator module is used for syntax checking and interconnecting of all the stages of the processing pipeline and the execution module is used to perform actions at each stage of the runtime operation.

5.1.1 PyTables and PLY

Flowy uses PyTables [6] to store the flow-records. PyTables is built on top of the Hierarchical Data Format (HDF) library and can exploit the hierarchical nature of the flow-records to efficiently handle large amounts of flow data. In addition, Flowy uses Python Lex-Yacc (PLY) for generating a Look-Ahead LR Parser (LALR) parser and providing extensive input validation, error reporting and validation on the execution module.

5.1.2 Records

Flow-records are the principal unit of data exchange throughout Flowy's processing pipeline. The prototype implementation allows the Record class to be dynamically generated allowing future implementations to easily plug in support for Internet Protocol Flow Information Export (IPFIX) or even newer versions of NetFlow [7] exports. The Record class handles the reading of the flow-records and saving them in PyTables for future processing.

5.2 PROCESSING PIPELINE

The pipeline consists of a number of independent processing elements that are connected to one another using UNIX-based pipes. Each element receives the content from the previous pipe, performs an

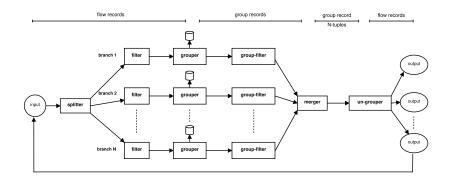


Figure 1: Flowy: Processing Pipeline [1]

operation and pushes it to the next element in the pipeline. Figure 1 shows an overview of the processing pipeline. The flow record attributes used in this pipeline exactly correlate with the attributes defines in the IPFIX Information Model specified in RFC 5102 [8]. A complete description on the semantics of each element in the pipeline can be found in [4]

5.2.1 Splitter

The splitter takes the flow-records data as input in the flow-tools compatible format. It is responsible to duplicate the input data out to several branches without any processing whatsoever. This allows each of the branches to have an identical copy of the flow data to process it independently.

The splitter module handles the duplication of the Record instances to separate branches. Instead of duplicating each flow-record to every branch (as specified in the specification), the implementation follows a pragmatic approach by filtering the records beforehand against all the defined filter rules to determine which branches a flow-record might end up in and saves this information in a record-mask tuple of boolean flags. The go(...) method in the Splitter class then iterates over all the (record, record-mask) pairs to dispatch the records to corresponding branches marked by their masks using the split(...) method. The class uses branch names to branch objects mapping to achieve the dispatch.

5.2.2 Filter

The filter performs *absolute* filtering on the input flow-records data. The flow-records that pass the filtering criterion are forwarded to the grouper, the rest of the flow-records are dropped. The filter compares separate fields of a flow-record against either a constant value or a

value on a different field of the *same* flow-record. The filter cannot *relatively* compare two different incoming flow-records

The filter module handles the filtering stage of the pipeline. Since in the implementation the filtering stage occurs before the splitting stage, a single Filter class instance suffices for all the branches. Within the filter module, each filtering statement is converted into a Rule class instance, against which the flow-records are matched. The Rule instances are constructed using the (branch mask, logical operator, arguments) tuple. After matching the records against the rules, the record's branch mask is set and is then used by the splitter to dispatch the records to the filtered branches.

5.2.3 Grouper

The grouper performs aggregation of the input flow-records data. It consists of a number of rule modules that correspond to a specific subgroup. A flow-record in order to be a part of the group should be a part of at-least one subgroup. A flow-record can be a part of multiple subgroups within a group. In addition a flow-record cannot be part of multiple groups. The grouping rules can be either absolute or relative. The newly formed groups which are passed on to the group filter can also contain meta-information about the flow-records contained within the group using the aggregate clause defined as part of the grouper query.

The grouper module handles the grouping of flow-records data. The Group class instance contains group-record's field information required for absolute filtering. It also contains the first and last records of the group required for relative filtering of the group-records. The AggrOp class instance handles the aggregation of group-records. The allowed aggregation operations are defined in aggr_operators module. Custom-defined aggregation operations are also supported using -aggr-import command line argument.

5.2.4 Group-Filter

The group-filter performs *absolute* filtering on the input group-records data. The group-records that pass the filtering criterion are forwarded to the merger, the rest of the group-records are dropped. The group-filter compares separate fields (or aggregated fields) of a flow-record against either a constant value or a value on a different field of the *same* flow-record. The group-filter cannot *relatively* compare two different incoming group-records

The groupfilter module handles the filtering of group-records. The GroupFilter class within the module iterates over the flow-records within the group and applies filtering rules across them. The filtering rules reuse the Rule class from the filter module. The flow-records

are then added to the time index and stored in a pytables file for further processing. For groups that do *not* have a group-filter defined for them, run through a AcceptGroupFilter class instance.

The timeindex module handles the mapping of the time intervals to the flow-records. The time index is used by the merger stage to learn about the records that satisfy the Allen relations. The add(...) method in the TimeIndex class is used to add new records to the time index. The get_interval_records(...) method on the other hand is used to retrieve records within a particular time interval.

5.2.5 Merger

The merger performs relative filtering on the N-tuples of groups formed from the N stream of groups passed on from the group-filter as input. The merger rule module consists of a number of a submodules, where the output of the merger is the set difference of the output of the first submodule with the union of the output of the rest of the submodules. The relative filtering on the groups are applied to express timing and concurrency constraints using Allen interval algebra [9]

The merger module handles the merging of stream of groups passed as input. It is implemented as a nested branch loop organized in an alphabetical order where every branch is a separate for-loop over its records. During iteration, each branch loop executes the rules that matches the arguments defined in the group record tuple and subsequently passes them to the lower level for further processing. The Merger class represents the highest level branch loop and as such it must iterate over all of its records since it does not have any rules to impose restrictions on the possible records. The MergerBranch on the other hand represents an ordinary branch loop with rules.

5.2.6 Ungrouper

The ungrouper unwraps the tuples of group-records into individual flow-records, ordered by their timestamps. The duplicate flow-records appearing from several group-records are eliminated and are sent as output only once.

The ungrouper module handles the unwrapping of the group-records. The generation of flow-records can also be suppressed using -no-records-ungroup command line option. The Ungrouper class instance is initialized using a merger file and an explicit export order.





FLOWY: APPLICATIONS

- 8.1 IPV6 TRANSITION FAILURE IDENTIFICATION
- 8.2 CYBERMETRICS: USER IDENTIFICATION
- 8.3 APPLICATION IDENTIFICATION USING FLOW SIGNATURES
- 8.4 TCP LEVEL SPAM DETECTION



Part III MOTIVATION



Part IV

WORK PLAN

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DESIGN



IMPLEMENTATION



PERFORMANCE EVALUATION





Part V

IMPLEMENTATION AND EVALUATION

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DESIGN



IMPLEMENTATION



PERFORMANCE EVALUATION



FUTURE WORK





Part VI

APPENDIX





APPENDIX

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DECLARATION	
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