



# Using ML clustering tools to improve data transfer management operations



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The **GRID** computing paradigms adopted by the main HEP experiments are based on the **distribution** of experimental data on computer resources located all over the world

All the **data transfer** processes are tracked in **log files** produced by the various services involved and such log files represent a source of **information** which is largely **underutilized**

An approach based on **unsupervised ML** techniques is used to automatically process information stored in **log files** with the aim of grouping the error messages and speed up the procedures for **detecting errors** and **solving problems**

## Clustering pipeline

### Step 1: text pre-processing

- Lowercase transformation and punctuation/stopword stripping
- Tokenization
- URL split

### Step 2: vectorization

- Transformation of the pre-processed text into numeric information to map each message to a point in a vectorial subspace (embedding)
- *word2vec* language model adopted

### Step 3: clustering

- k-means++ algorithm: intuitive approach and good performance in a wide range of applications
- The number of clusters *k* is selected at each clustering stage based on a grid search and geometrical criteria:
  - Within cluster Sum of Squared Errors (elbow method)

$$WSSE(\text{dist}, k) = \sum_{j=1}^k \sum_{x_i \in C_j} \text{dist}(x_i - \bar{x}_j)$$

$x_i$  generic data point

$\bar{x}_j$  centroid of a generic cluster  $C_j$

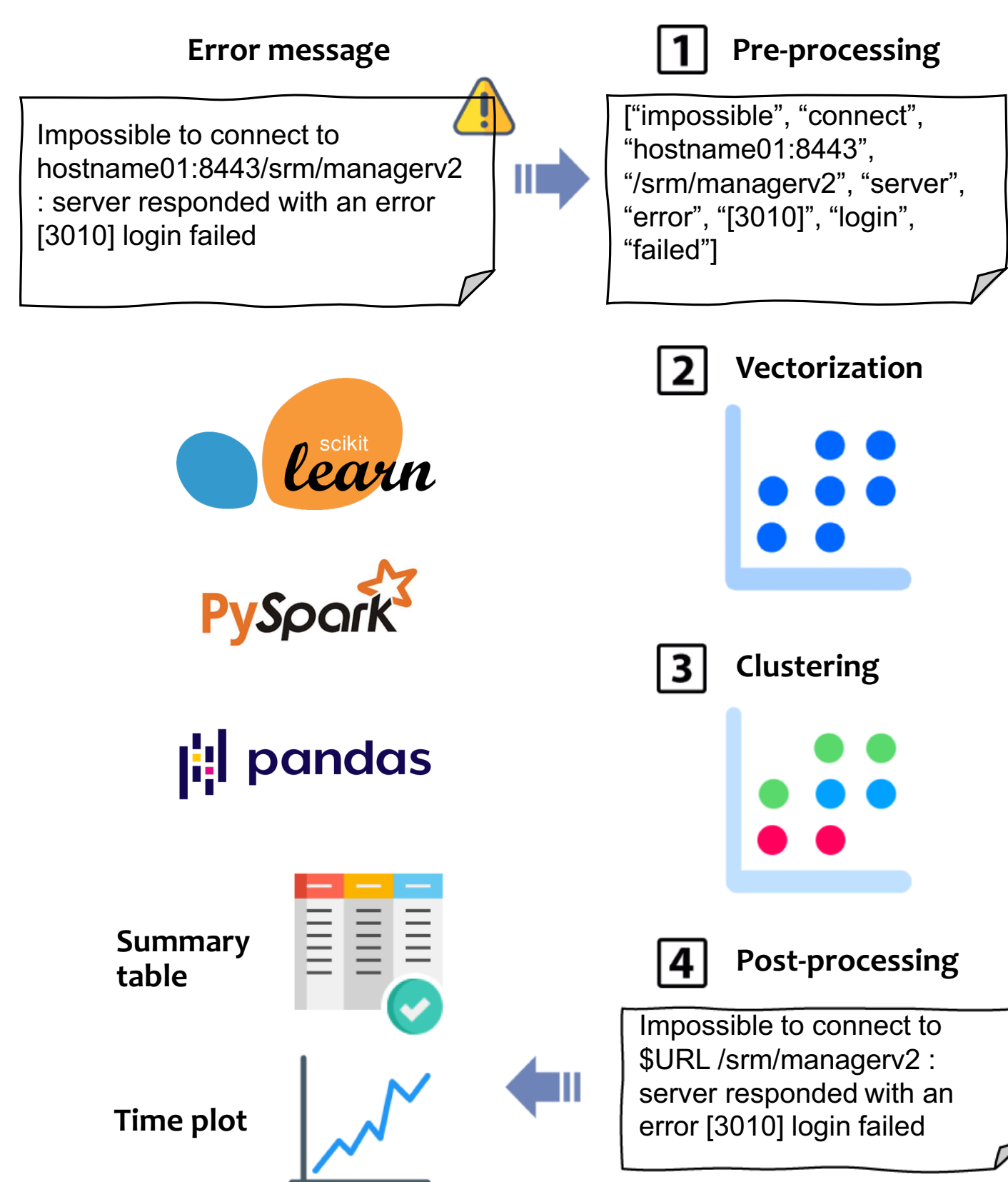
- Average Silhouette Width (max)

$$ASW(\text{dist}, k) = \frac{1}{n} \sum_{i=1}^n \frac{b_i - \bar{a}_i}{\max(\bar{a}_i, b_i)}$$

$\bar{a}_i$  distance of  $x_i$  from same cluster points  
 $b_i$  distance of  $x_i$  from other clusters points

### Step 4: post-processing

- Organize results according to the use case



## Analysis of File Transfer Service (FTS) errors

ID	cluster size	# strings	# patterns	message	n	%	source route	destination route
0	819405	117	14	destination overwrite sm-fts err communication error on send err [setlrmnm] [SRM] srm/manager/2 cgi-goop running on \$ADDRESS reports error installing context gas major status authentication failed gas minor status error chain globus_gsi_status_err handshake problems globus_gsi_callback_module could not verify credential globus_gsi_callback_module the certificate has been revoked serial number = 1 [continues]	85545	10.44%	Site-1	Site-4
6	9673	347	60	source sm-fts_err_err communication error on send err [setlrmnm] [SRM] srm/manager/2 cgi-goop running on \$ADDRESS reports error installing context gas major status authentication failed gas minor status error chain globus_gsi_status_err handshake problems globus_gsi_callback_module could not verify credential globus_gsi_callback_module the certificate has been revoked serial number = 1 [continues]	1838	19.00%	Site-22	Site-46
3	34183	1568	1537	error reported from sm-fts_err_err [setlrmnm] [SRM] srm/manager/2 cgi-goop running on \$ADDRESS reports error installing context gas major status authentication failed gas minor status error chain globus_gsi_status_err handshake problems globus_gsi_callback_module could not verify credential globus_gsi_callback_module the certificate has been revoked serial number = 1 [continues]	9333	27.30%	Site-12	Site-17

### Cluster Description (tabular format)

Analysis of the error messages in FTS log files (1 day)  
The three most frequent triplets of <pattern>-<source>-<destination> reported in descending order for each cluster

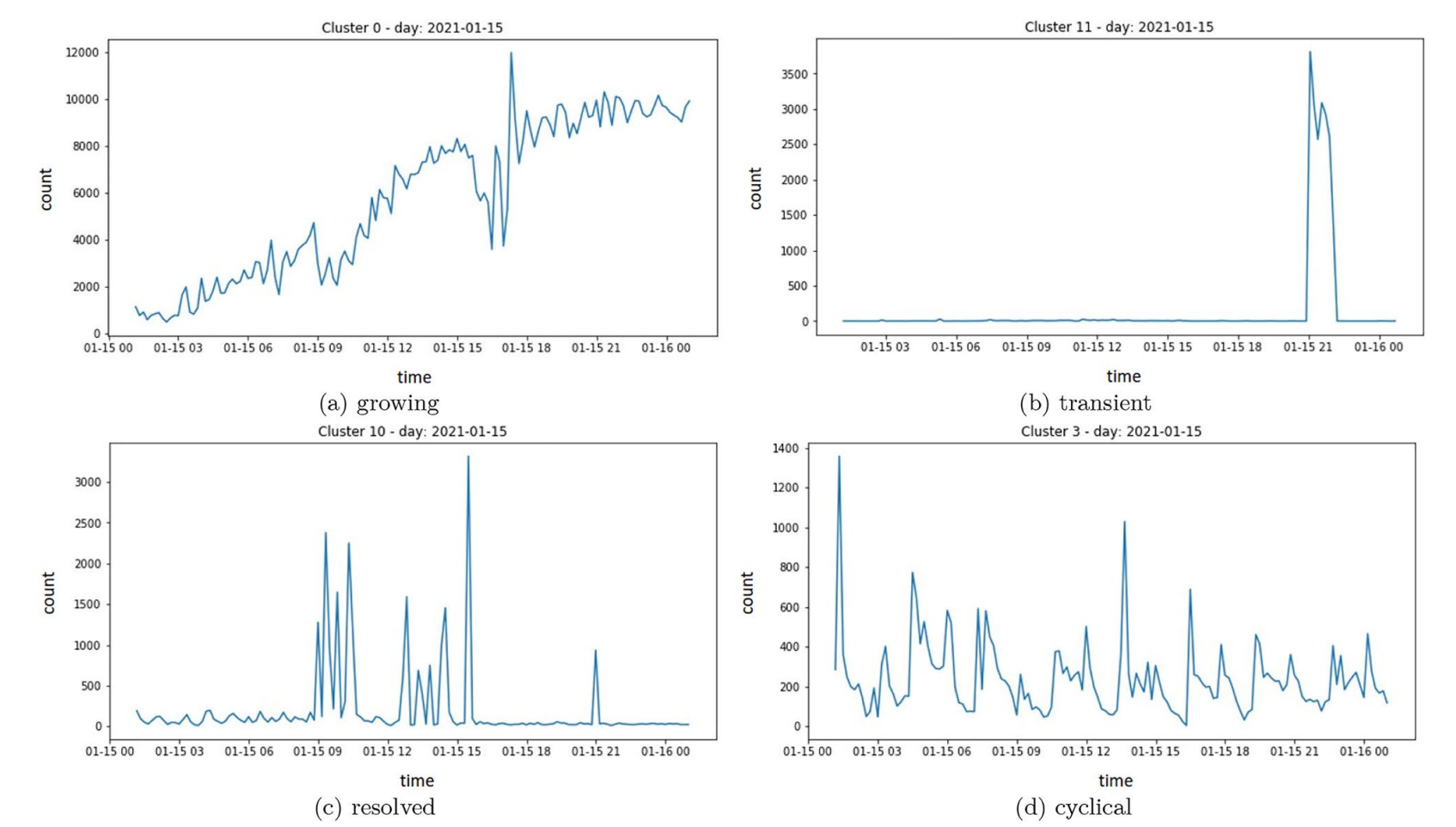
### Precious insights for spotting:

- what type of errors and where they occur
- amount of errors, both absolute and relative to the error group

### Time Series

Temporal trend of the number of errors generated by each cluster

- escalating or cyclical failures → require immediate actions
- transient or in resolution



N. clusters	ASW	WSSE	Perfect match	Fuzzy match	Partial match	False positives	False negatives
15	0.89	17107	7	3	2	3	1

Summary of the cross-check between clusters and incidents reported in GGUS. Most of the groups discovered are linked to reported issues, with only 3 false positives and 1 false negative

Computing and Software for Big Science (2022) 6:16 <https://doi.org/10.1007/s41781-022-00089-z>

Extensive testing using incidents reported in GGUS as a benchmark (17 days window)

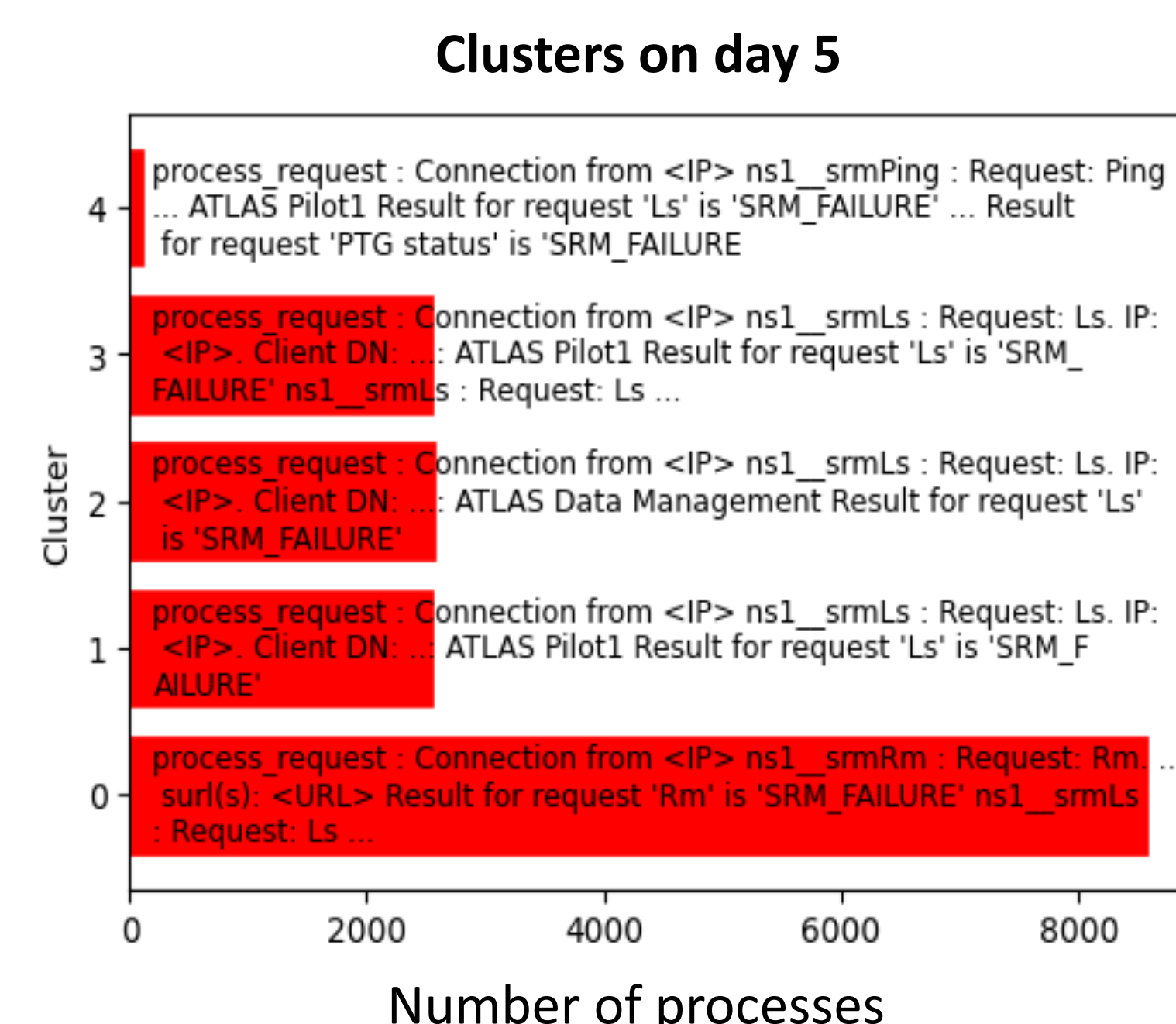
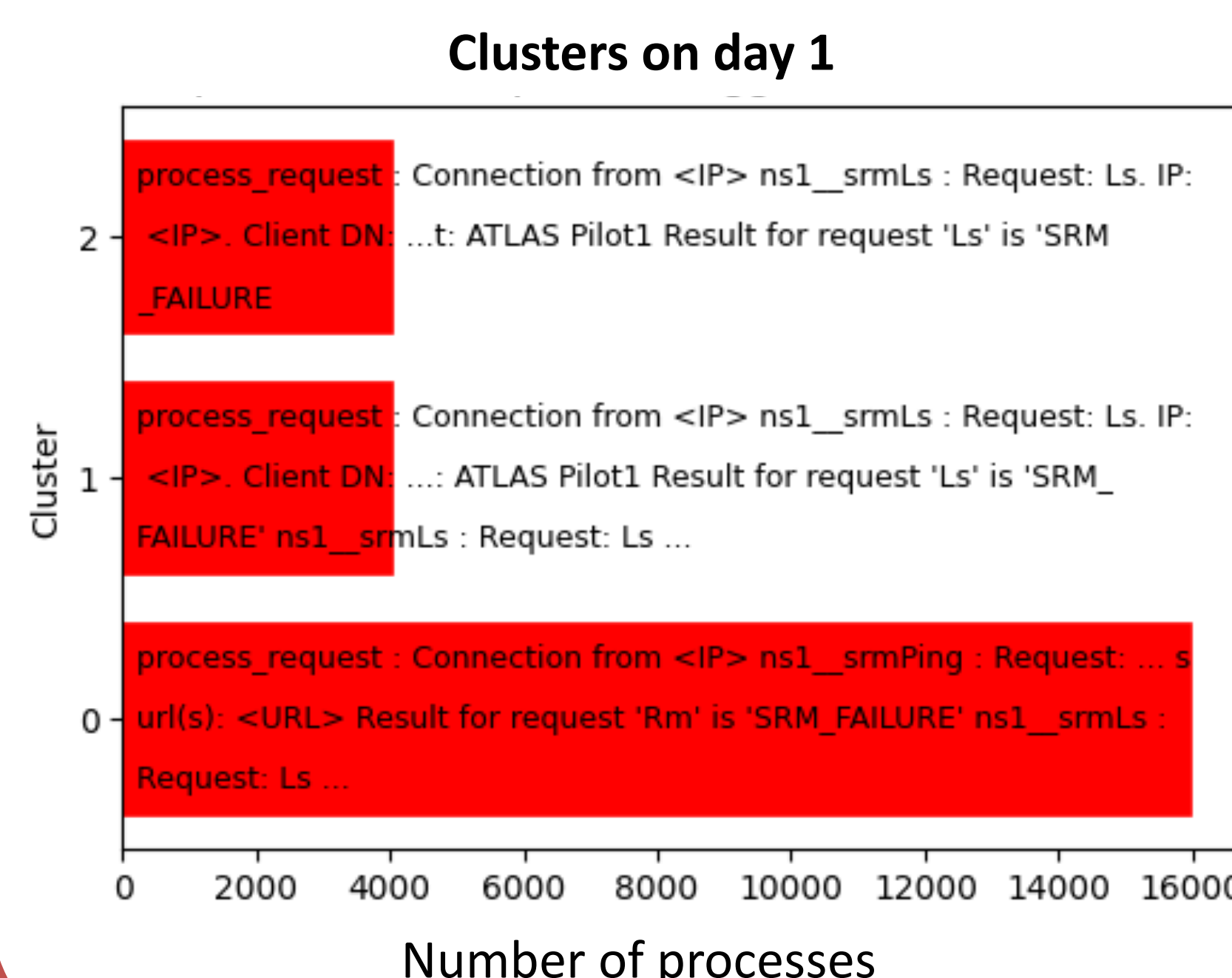
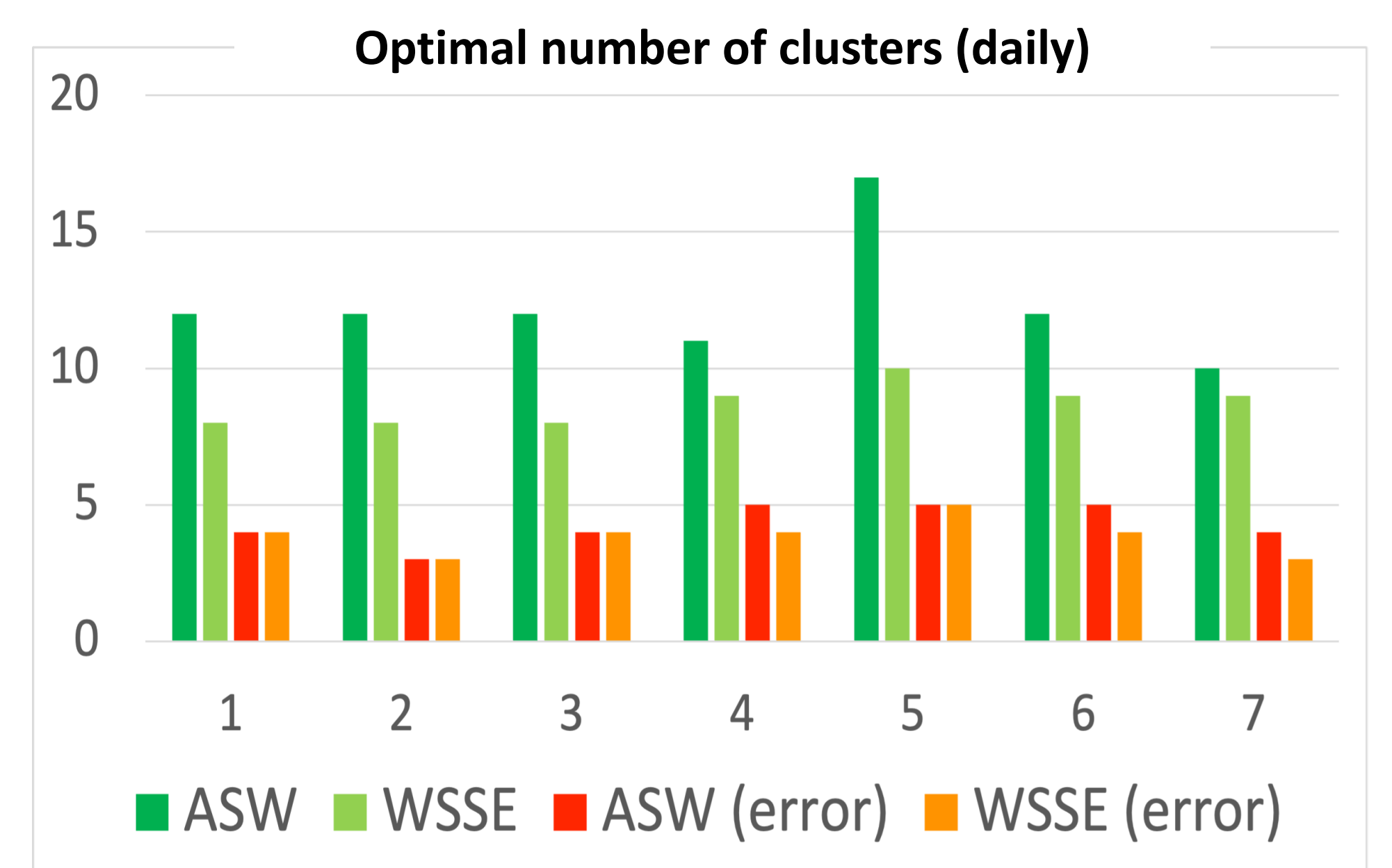
- overlapping between discovered clusters and the reported issues

## Analysis of StoRM log files

Analysis performed on a 7 days interval Dataset (daily log files) from INFN-CNAF StoRM system

Messages of a single log file are grouped by StoRM process (separately for explicit error messages) and then passed to the clustering pipeline

Higher values of the optimal number of clusters may give a hint of an anomaly (day 5 in this example)



The variation of the daily number of clusters and clustered error messages could give more information on the cause of the problem