

Adaptive Log Compression for Massive Log Data

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Motivation

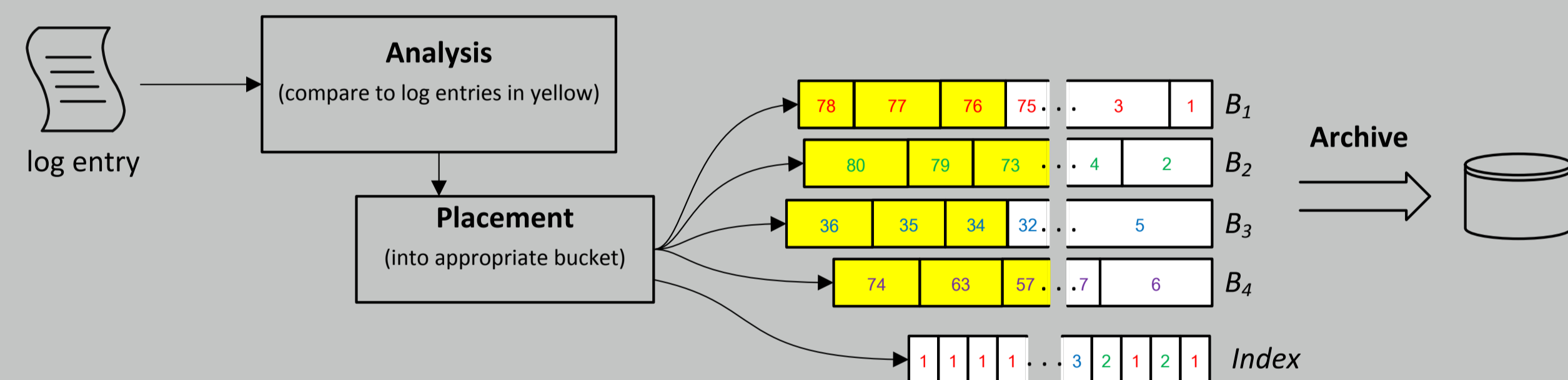
```
[2d] Mar 19 22:41:29 nid00456 #1#-ng[442]: STATS: dropped 0
[2d] Mar 19 22:41:29 nid00528 #1#-ng[451]: STATS: dropped 0
[2d] Mar 19 22:41:29 nid00960 #1#-ng[442]: STATS: dropped 0
[1e] Mar 19 22:41:29 #2# dhcpcd: DHCPACK on 10.1.100.183 to 00:0f:cb:9e:bd:40 via eth3
[1e] Mar 19 22:41:29 #2# dhcpcd: DHCPREQUEST for 10.1.100.183 from 00:0f:cb:9e:bd:40 via eth3
[2d] Mar 19 22:41:30 nid00268 #1#-ng[442]: STATS: dropped 0
[2d] Mar 19 22:41:30 nid00279 #1#-ng[442]: STATS: dropped 0
[2d] Mar 19 22:41:30 nid00339 #1#-ng[442]: STATS: dropped 0
```

- Log data is **humongous** and often stored for an extended period of time.
- Cost to store log data is proportional to the number of bytes the data occupies.
- Log entries are often **heterogeneous**, with varying patterns over time
- Compressors such as *bzip2* and *gzip* are most **effective** when compressing **predictable data**.

Definition (Adaptive Log Compression)

For a log data D with n log entries e_1, \dots, e_n (sorted on arrival timestamps), a budget g , produce g disjoint log buckets B_1, \dots, B_g , such that $\forall i \in [1, n], \exists j \in [1, g], e_i \in B_j$, and $\forall x, y \in [1, g], B_x \cap B_y = \emptyset$. Each bucket also stores log entries in sorted order. Given any compression method Z , $Z(D)$ is the compressed output with size (in bytes) $|Z(D)|$. The objective is to maximize $|Z(D)| - \sum_{j=1}^g |Z(B_j)|$.

System Overview



- Analysis:** A new log entry is compared to the m most recent entries in B_1, \dots, B_g using a similarity function.
- Placement:** The log entry is placed into the bucket B_k which resulted in the **best similarity score** during the **analysis** phase. The index k is saved to an index log, so the log can be reassembled when unarchived.
- Archive:** Each bucket B_1, \dots, B_g and the **Index** is compressed separately and in parallel using a compressor such as *bzip2* or *gzip* before being saved on a hard disk.

Test Data

- Red Storm is a system log from the supercomputer **Red Storm**.
- Apache log is an Apache web server log for the website **MesoWest**.

Table: Real world data

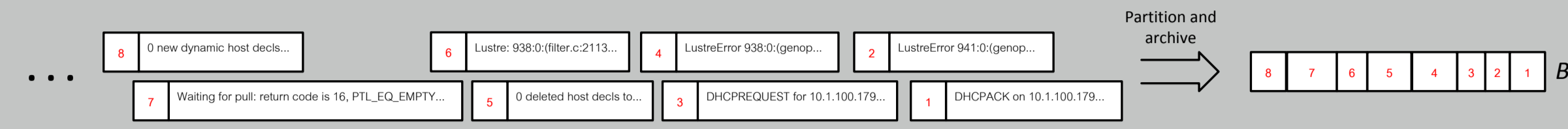
Source	Uncompressed (MB)	<i>bzip2</i> (MB)	<i>gzip</i> (MB)	log entries	med entry length
Red Storm	32,596	664	1080	219,096,168	99
Apache log	7501	296	533	26,568,851	277

References

- [1] K. Beyer, P. J. Haas, B. Reinwald, Y. Sismanis, and R. Gemulla. On synopsis for distinct-value estimation under multiset operations. In *Proceedings of the 2007 ACM SIGMOD international conference on Management of data*, pages 199–210. ACM, 2007.

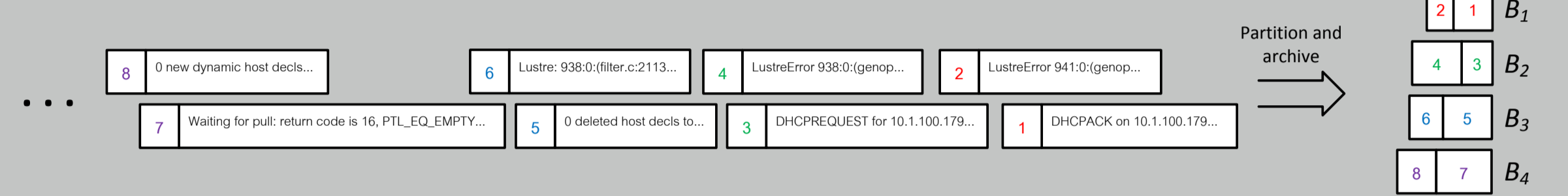
Similarity Functions

Figure: Centralized



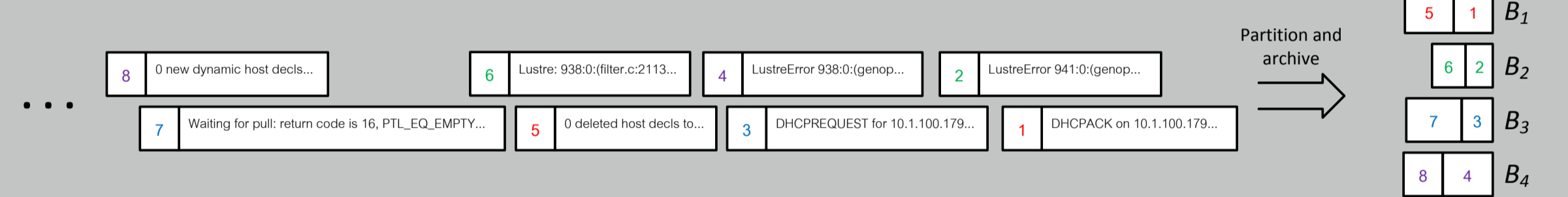
Centralized places each entry in a single archive file in chronological order. This is the method applied currently in industry when compressing logs.

Figure: Segmentation with bucket budget $g = 4$ and $n = 8$



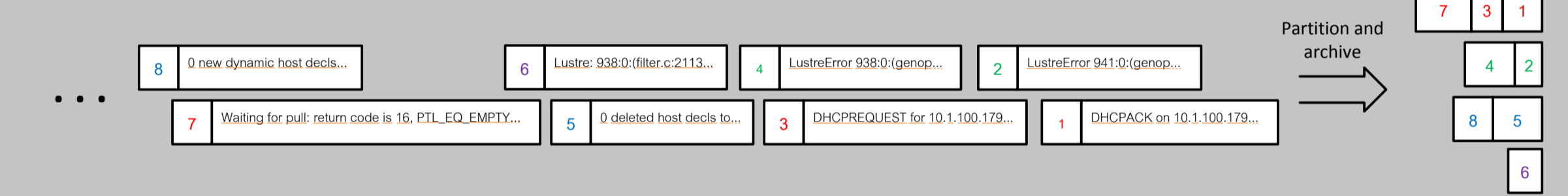
Segmentation only considers the total number of log entries n , and the bucket budget g . Each bucket will be given a continuous chunk of the log with an equal number of log entries. B_1 will contain $e_1, \dots, e_n/g$, B_2 will contain $e_{n/g+1}, \dots, e_{2n/g}$, etc.

Figure: Round Robin with bucket budget $g = 4$



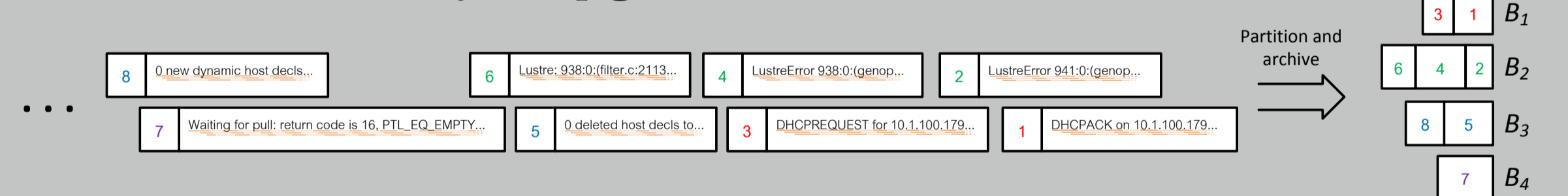
Round Robin inserts the i^{th} log entry into the bucket B_k , where $k = i \bmod g$.

Figure: Adaptive Word with bucket budget $g = 4$



Adaptive Word inserts each word of an incoming log entry into a set W . The incoming log entry will be inserted into the bucket B_k which has the largest **average** Jaccard index when compared with the log entries in B_k within the sliding window.

Figure: Adaptive q -gram with bucket budget $g = 4$ and $q = 3$



Adaptive q -gram splits an incoming log entry into q -grams, which are inserted into a set W . The incoming log entry will be inserted into bucket B_k which has the largest **average** Jaccard index when compared with the log entries in B_k within the sliding window. Computing the Jaccard index between sets which contain a large number of q -grams can be very expensive. In order to make **Adaptive q -gram** practical, the Jaccard index was **estimated** using a KMV synopsis[1]. Estimating the Jaccard this way did not degrade the final archive compression ratio.

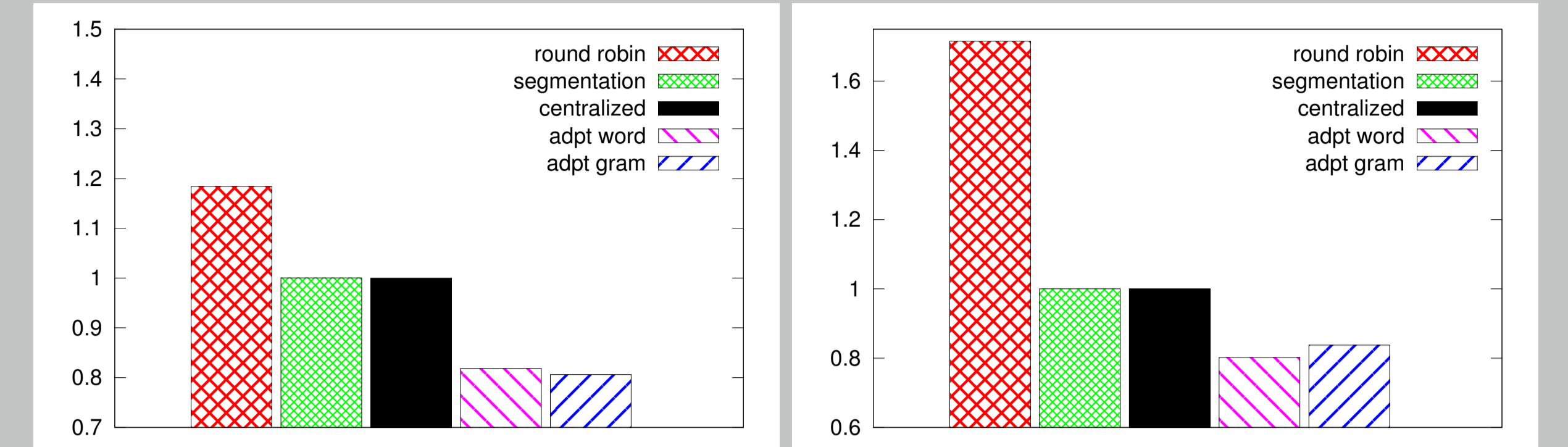
Experimental Setup

We use **centralized** as a reference for comparing the effectiveness of the other methods. The output size of any method is shown as a **ratio to the output size of centralized**.

Table: default settings for the similarity functions. The value is ignored if not applicable.

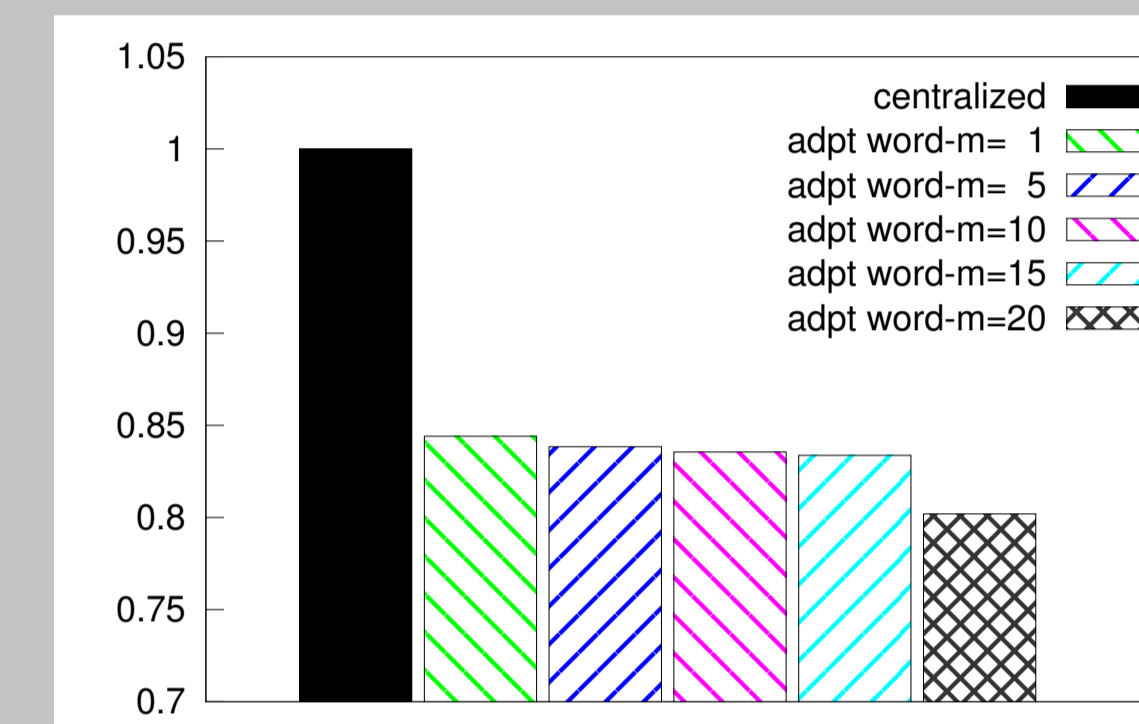
variable	value
bucket budget	$g = 32$
sliding window size	$m = 10$
q -gram length	$q = 6$
synopsis size	$k = 60$

Results

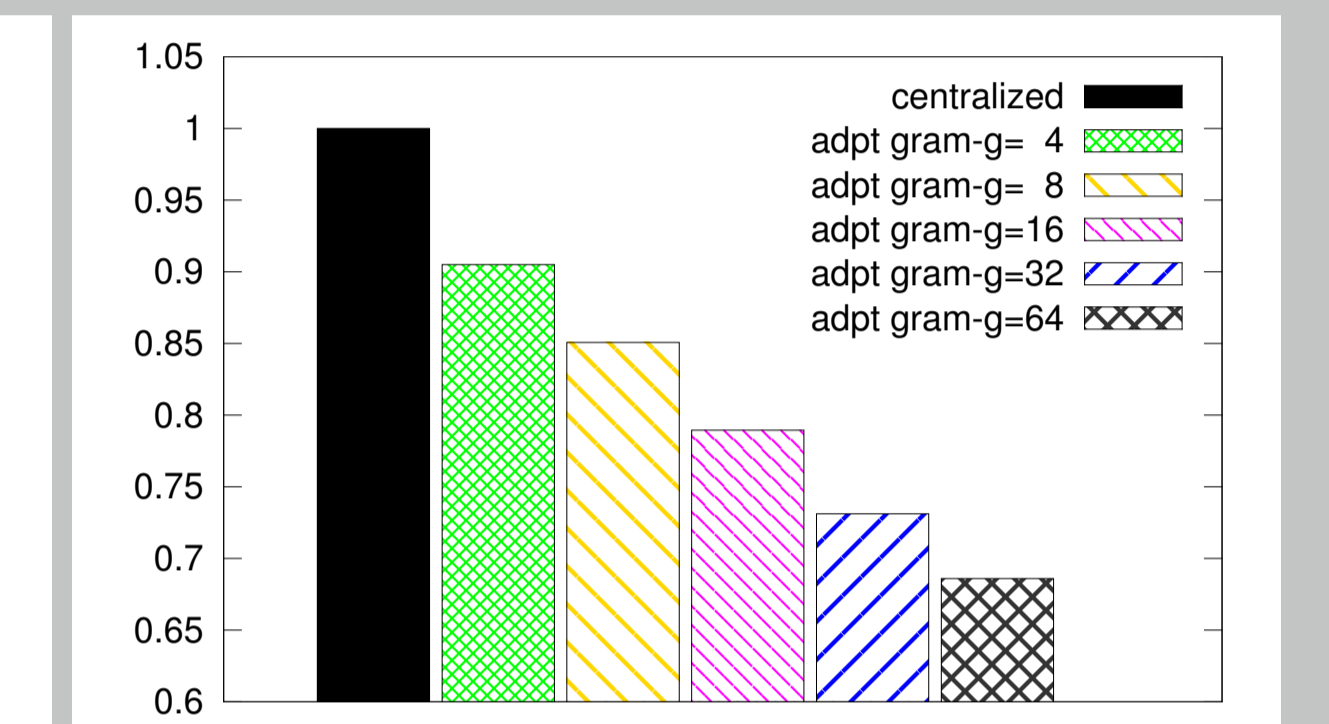


(a) Effectiveness of similarity functions on Apache log using the *bzip2* compressor when compared to centralized.

(b) Effectiveness of similarity functions on Red Storm using the *gzip* compressor when compared to centralized.



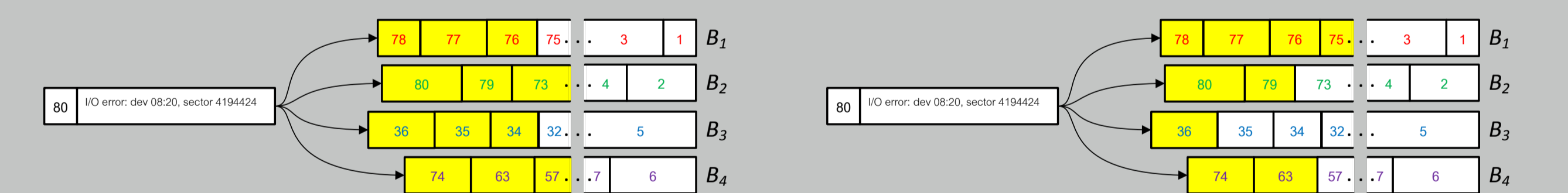
(c) Effectiveness of the adaptive word similarity function on Red Storm using the *gzip* compressor while varying the sliding window depth m when compared to centralized.



(d) Effectiveness of the adaptive q -gram similarity function on Apache log using the *bzip2* compressor while varying the number of buckets g when compared to centralized.

- Naive log distribution can significantly **decrease** the compression effectiveness, as shown by **round robin** in figure (a) and (b).
- adaptive word** improves the compression ratio on **Red Storm** by up to **20%**. see figure (c).
- adaptive q -gram** improves the compression ratio on **Apache log** by up to **30%**. see figure (d).
- The number of buckets **positively** correlates with the compression ratio. see figure (d).

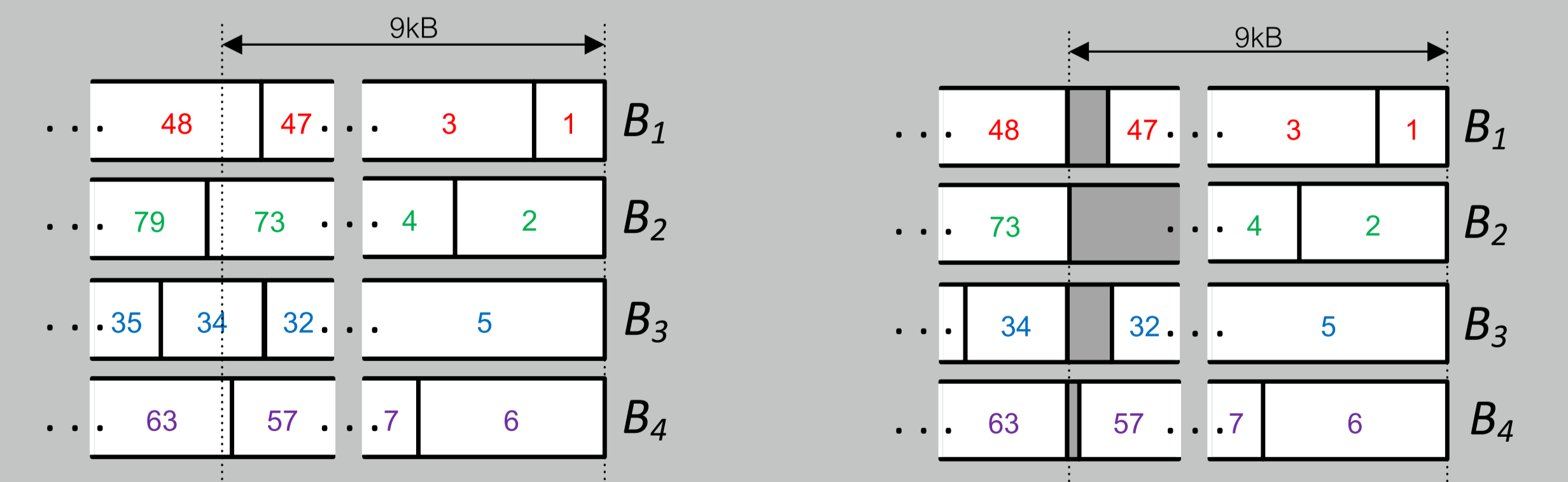
Future Work



(a) current method, **static** sliding window depth

(b) proposed method, **dynamic** sliding window depth

- Sliding window depth m should be dynamic.
 - Smaller when bucket contents is nearly **identical**.
 - Larger when bucket contents are more **varied**.



(a) current method, compression blocks are **not** aligned

(b) proposed method, compression blocks are aligned

- Compression utilities compress data in small compression blocks.
 - Maximum block size for *bzip2* and *gzip* is **9KB**.
 - Compression tables are **not shared** between blocks.
- Straddling log entries between two compression blocks can **decrease** compression **effectiveness**.
- Avoiding straddling log entries **isolates** archive **corruption**.