Compression and Decompression Techniques for Data over the Cloud Servers

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ABSTRACT

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| **Aims:** This work aims to present an adaptive lossless compression-decompression technique for different data types, such as text, image, audio, and video files. The primary objective is to optimize cloud storage by maximize space savings while maintaining data confidentiality and reducing data loss.**Study design:** Comparative analysis of experimental research of various compression methods**Place and Duration of Study:** The study was perform in computational research environment of Department of Computer Science Dept. of Computer Science, Babasaheb Bhimrao Ambedkar University, Lucknow, India, using data sets. **Methodology:** The proposed method used search-based compression techniques using linear, binary, and interpolation search, to get effective lossless compression. The method performance was computed using a dataset of different file formats, observing key parameters, including compression ratio, decompression time and space savings. The results were presented through graphs and table comparing with existing methods.**Results:** Proposed method shows that the method provides a higher compression ratio and faster decompression time compared to existing method. It effectively reduces storage space on cloud servers without reducing data integrity. Interpolation-based compression perform well than linear and binary search approaches, achieving up to 85% compression for audio and video files and 79% for text-based files while maintaining information integrity.**Conclusion:** The proposed method presents an effective efficient and scalable lossless compression method works on different file formats. Its adaptability makes it a practical solution for cloud storage optimization, improving storage efficiency, retrieval speed, and cost-effectiveness. Comparative analysis confirms its superiority over existing techniques, highlighting its potential for large-scale cloud applications. |

*Keywords: Cloud storage; Information retrieval, Search algorithms, Data compression, Data decompression.*

1. INTRODUCTION

From the literature, it is seen that there is tremendous rise in the growth of multimedia contents; therefore, it is essential to manage the cloud storage handling large-scale multimedia contents. The compression approach of data provides significant idea for optimizing the cloud storage and transmission efficiency with data integrity. Some traditional compression techniques, such as Huffman coding, Lempel-Ziv, and Shannon-Fano (SF), improve the cloud storage. The modern cloud storage gives more flexible and scalable methods that work with wide variety of file types, including text, image, audio and video, while ensuring minimal delays and cost.

Several studies had proposed many techniques to boost up the cloud storage, such as Hadoop Distributed File System (HDFS) improved cloud storage, but it lacks to provide compression method for all file types [1]. Classical compression methods were unfit for cloud data, especially related to medical records, as it requires full data recovery. Lossless method like Lempel–Ziv–Welch (LZW) had utilized for medical data, making sure for complete data retrieval with optimized space [2]. Hadoop Distributed File System (HDFS) reduces similar data but lack in strong compression method. ID-based encryption had been proposed; still its computational complexity was challenging task for big datasets. Self-managed cloud storage improved scheduling but does not work well on compression [3]. Techniques such as Boneh-Lynn-Shacham (BLS) cryptography and Automatic Blocker Protocol (ABP) raise security but lack on storage reduction [4]. Multi-keyword searchable encryption with Locality Sensitive Hashing (LSH) provides easy document access, but its ability to data compression remains untested. Additionally, Convergent Encryption (CE) and Proof of Ownership (PoW) techniques worked deduplication [5] but are ineffective on data compression. Many studies have analyzed the performance of used in compression techniques, such as Huffman, LZ77, and BID, based on compression ratio, speed, and storage efficiency. The methods remain standard in data compression, modern cloud environments that help in all file formats, minimizing cost [6-7].

**Table 1. Compression techniques for text, image, audio and video files**

|  |  |  |  |
| --- | --- | --- | --- |
| **Text/Doc/Pdf** | **Image** | **Audio** | **Video** |
| 7zip, Huffman Coding, Zip, Gzip, Bzip2, Shannon-Fano Coding, Arithmetic Coding, Lempel-Ziv Coding (LZ), Lempel-Ziv-Welsh (LZW), Golomb Coding. | EZW, DCT, Wavelet Transform, Run-length Encoding, Predictive Coding and DPCM, Area Image Compression, DEFLATE, Chain Codes, Adaptive Dictionary Algorithms, Entropy Encoding, Decreasing the Color Space, Transform coding, Chroma Subsampling, Fractal Compression | LPAC (Lossless predictive audio compression), ALAC (Apple lossless audio codec), Huffman shift coding, Intrinsic mode function, WMA (Windows media audio lossless), NERO AAC CODEC, FAAC (Freeware advance audio coder), IEEE 1857.2. | JPEG,MPEG,H.261,H.263, MPEG-1, MPEG-2, MPEG-4, MPEG-7, MPEG-21 and H.264. |

The latest literature highlights the significant gaps in compression methods for cloud storage. It is observed that various data compression techniques are described and some of the techniques are listed in the following Table 1.While deep learning-based models[8] had improved efficiency using high computational resources, making unfit for real-time applications. Adaptive algorithms, like as dictionary-based techniques [9], were adaptable but less optimize for multimedia contents. Graph entropy based compression methods had shown good results in less data redundancy, yet implementation on cloud server remains limited [10]. Hybrid encryption-compression techniques combine security with compression, but lead to increased processing cost. Furthermore, domain-specific approaches, such as medical image compression and genomic data compression, had provided high efficiency but are not for multi-format cloud storage [11]. An efficient data compression and storage method had been utilized by Regressive Probabilistic Key Encryption (RPKE) for data encryption and Lempel-Ziv-77-Huffman coding (LZ77-HM) for compressing huge datasets [12]. This method gave significant improvements in compression ratio and rate, but computational overhead with RPKE limit scalability in real-time applications. SCOPe, a unified pipeline focused on optimizing data storage and less costs in cloud environments is also used [13]. With combination of data placement, compression scheme and access pattern-aware data partitioning, SCOPe presented significant cost savings. However, the complexity and accurate access may create challenges for dynamic storage scenarios. A power-efficient image storage solution combines image compression with Super-Resolution Generative Adversarial Networks (SRGAN) which reduces image sizes and reconstructs after retrieval [14]. SRGANs may provide increased computational demands during image reconstruction. A detailed review of several distinct data compression strategies was conducted, including data quality-based, coding-based and application-based and data type-based techniques [15]. The important references provide ideas to create models, methods, tools and security approaches that make software creation easier [16-18] . Existing methods sometimes lacks in compressing bigger text files on handheld devices [19], improving lossless compression over variety of file formats [20], and choosing suitable method for text data [21]. Image compression methods are usually restrict to some formats like BMP [22], while generic file compression struggles domain-specific algorithms and restrict for resource-constrained scenarios [23]. Also, video compression methods like adaptive fuzzy inference systems use evaluation against common codecs to handle impact on quality and efficiency [24]. Resolving above gaps is important for forming more adaptable and effective compression method over various applications.

The above-mentioned limitations demonstrate the requirement for adaptive compression methods that emphasize cloud storage across all sorts of data formats while ensuring minimal computational cost. The present work demonstrates a compression-decompression method which shall increase cloud storage, retrieval speed and good performance. Focusing on storage issues such as delays and computational cost, the presented approach is a flexible technique that works well with modern cloud servers. The research work presented in this paper fills the above-mentioned gaps by introducing flexible, efficient and resource-aware model, ensuring improved cloud storage management for various data applications.

2. material and methods

By compressing a file, its size can be minimized. A smaller file size makes it simpler to manage large multimedia content. By compressing the data, unnecessary information can be eliminated using tools like video conferencing, file storage, and printers. The suggested method makes use of search-based data compression, even if adaptive modeling might not be the best compression methodology for natural language texts in full-text content retrieval scenarios. Data is compressed using search methods like linear, binary, and interpolation searches. Compressing file types like .pdf, .jpg, .txt, .doc, audio and video before storing them on the cloud is a good idea. Once the information is stored there, it can be simply accessed. Initially, the compressed information decompresses and is then obtained by the authorized client.

The following compression techniques are used and elaborated below.

## **Linear Search**

## In order to apply the linear search approach, a file must first be converted into binary format. Then, the algorithm searches for 0 and 1 values.

***Algorithm 1:*** *To compress using Linear Search*

***Input:*** *Select file of format .txt, .pdf, .doc, .jpeg*

*begin*

*file \_format=pdf||doc||txt||jpg*

*data = content*

*(For Compression)*

*bin\_value = ' '.join(format(ord(i), '08b') for i in str(data))*

*list.sort(bin\_value)*

*linear\_Sort()*

*rank\_zero=[n for n, val in enumerate(m) if val==0]*

*zero = len([0 for x in t if x==0])*

*rank\_one = [n for n,val in enumerate (m) if val==1]*

*one = len([0 for x in t if x==1])*

*file\_size= sys.getsizeof (compressed\_content)byte)*

*end*

1. **Binary Search**

Binary Search is an effective technique for narrowing your choices from an ordered set of options. It continuously bisects the portion of the list that can contain an item until the item's position is collapsed. The binary search algorithm is shown below:

***Algorithm 2:*** *To compress using Binary Search*

***Input:*** *Select file of format txt, pdf, doc, jpeg*

*begin*

*file \_format=.pdf||.doc||.txt||.jpg*

*data = content*

*(For Compression)*

*bin\_value = ' '.join(format(ord(i), '08b') for i in str(data))*

*list.sort(bin\_value)*

*binary\_search(arr,low=none,high=none)*

*if arr[high]==0:*

*return 0*

*if arr[low]==1:*

*return high-low+1*

*return binary\_sort(arr,low,mid)+binary\_sort(arr,mid+1,high)*

*result=binary\_search(arr)*

*No\_of\_ ones= result*

*a=l-result*

*No\_of\_zeros=a*

*print(file is compressesed)*

*end*

1. **Interpolation Search**

When the values in a sorted array are evenly distributed, a new technique called the interpolation search may be applied. In addition to being quicker and more accurate than binary search, interpolation search also avoids becoming caught in local minima. The algorithm for interpolation search is given below:

***Algorithm 3:*** *To compress using Interpolation Search*

***Input:*** *Select file of format txt, pdf, doc, jpeg*

*begin*

*file \_format=.pdf||.doc||.txt||.jpg*

*data = content*

*(For Compression)*

*def interpolation\_Search(A, result):*

*while A[high] != A[low] and A[left] <= target <= A[high]:*

*cen = low + (result - A[low]) \* (high -low) // (A[high] - A[low])*

*if result == A[cen]*

*return cen*

*elif result < A[cen]:*

*right = cen - 1*

*else:*

*low = cen + 1*

*if result == A[low]:*

*return low*

*index = interpolationSearch(A, key)*

*if index != -1:*

*no\_of\_one = [l for l,value in enumerate (lst) if value==1]*

*no\_of\_zero = [i for l,value in enumerate (lst) if value==0]*

*print(file is compressed)*

*end*

Method used to put back the compressed data in its original form is Decompression. There are several compression methods stated and each required decompression for data. The following algorithm is used to decompress any file using linear, binary, and interpolation search for compression.

***Algorithm 5:*** *To decompress file*

***Input:*** *To decompress, choose the input*

*file.path = filedialog.askopenfilename()*

*name=Path(path).stem*

*file\_extension = pathlib.Path(path).suffix*

*new\_path=shutil.copy (path1,'C:/Users/Hp/Desktop/Download')*

*print("File is downloaded")*

*end*

Audio video in digital form, have high quality and large file size. This makes necessary to compress audio and video file to make more space. Below is the ZAV-compression algorithm helps in compression and decompression of audio video file without damaging the quality of file.

***Algorithm 6:*** *To compress audio and video file*

***Input:*** *Select file format of mp3, mp4*

*begin*

*fileout = "compressed\_file"*

*(For Compression)*

*with open(filein, mode="rb") as fin, open(fileout, mode="wb") as fout:*

*content = fin.read()*

*compress\_file=zlib.compress(content, zlib.Z\_COMPRESSION)*

*print (f"Compressed size: {sys.getsizeof(compressed\_file)} bytes")*

*(For Decompression)*

*with open(fileout, mode="rb") as fin:*

*content = fin.read()*

 *Decompressed\_file = zlib.decompress (content)*

*print(f"Decompressed size: {sys.getsizeof(compressed\_data)} bytes")*

*end*

3. results and discussion

In this section, the proposed searching method’s performance is evaluated. There are three searching techniques Linear, Binary, Interpolation used on different file format such as text,.doc., pdf, .jpg, audio and video.

The dataset used is consist of file types including text (.txt, .pdf, .doc), images (.jpg), audio (.mp3), and video (.mp4), ranging 5,10,20,25 MBs, representing real-world application. Dataset is consists of both structured and unstructured data such as reports, academic papers, text and PDFs. Images in the dataset including high-resolution photographs and compressed web images. Audio files are of different bitrates and formats and video files include standard-definition (SD), high-definition (HD), and ultra-high-definition (UHD) formats. The dataset are designed to cover practical cloud storage environments.

The study evaluates the efficiency using different key parameters.

1. **Compression Ratio (CR):** Data is compressed effective can be measured by finding compression ratio. The percentage file size reduced after compression, higher percent gives better efficiency.

CR = $\frac{Uncompressed file bits}{Compressed file bits}$

1. **Compression Time (CT):** Time required compressing the file using proposed method. Usually measured in millisecond or seconds.
2. **Decompression Time (DT)**: Time required decompressing the file using proposed method. Usually measured in millisecond or seconds.

In the Table 2., the results of linear, binary and interpolation search compression are summarized, showing the effectiveness of compression and the time required for compression and decompression in terms of different file size and likewise in Table 3., the audio and video file's compression ratio, compression and decompression time are shown. Table 2., shows that the interpolation based search method gave high compression ratio with lower compression and decompression time in compare to linear and binary based search technique. Text, Pdfs and document files compression ratios are reduced to 79%, image files achieves upto 83% without losing data integrity. From Table 3., it is shown that audio and video files are reduced up to 85% and 83% after compression method. From both the tables, it is clear which suggested strategy offers a superior compression rate for cloud storage while keeping quality. Low compression and decompression time helps to access files faster, which improves the overall performance of cloud server. Key strength of these techniques is adaptability over multiple file formats, which makes suitable for real world application. Some techniques compromise the quality of data while proposed method ensures lossless compression. Also, reduced file size transfer quicker with minimum latency in storage and retrieval process. This advantage of proposed method shows the practical and scalable quality for handling large files over cloud storage. The results for various files compression and decompression time are displayed in Figure 1 and Figure 2

through graphs.

**Fig. 1. Compression time of different sizes of file types with different search methods**

**Fig. 2. Decompression time of different sizes of file types with different search methods**

**Table 2. Computation of compression and decompression for document and image files**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **File format** | **File Size (MB)** | **Compression ratio (%)** | **Compression time (sec)** | **Decompression time (sec)** |
| **Linear** | **Binary** | **Interpolation** | **Linear** | **Binary** | **Interpolation** | **Linear** | **Binary** | **Interpolation** |
| **.txt** | 5 | 45 | 47 | 49 | 35 | 22 | 18 | 33 | 21 | 16 |
| 10 | 46 | 48 | 51 | 38 | 23 | 19 | 35 | 23 | 17 |
| 15 | 48 | 50 | 53 | 43 | 25 | 23 | 41 | 23 | 20 |
| 20 | 50 | 56 | 54 | 47 | 29 | 25 | 43 | 26 | 22 |
| 25 | 53 | 58 | 57 | 49 | 32 | 27 | 46 | 30 | 24 |
| **.doc** | 5 | 54 | 56 | 59 | 30 | 21 | 15 | 28 | 19 | 14 |
| 10 | 57 | 58 | 62 | 32 | 23 | 18 | 30 | 20 | 17 |
| 15 | 61 | 66 | 68 | 36 | 26 | 19 | 35 | 24 | 19 |
| 20 | 63 | 69 | 73 | 39 | 28 | 22 | 37 | 26 | 20 |
| 25 | 66 | 71 | 76 | 41 | 30 | 25 | 39 | 30 | 23 |
| **.pdf** | 5 | 51 | 53 | 56 | 5 | 4 | 2 | 5 | 4 | 2 |
| 10 | 62 | 64 | 65 | 8 | 6 | 4 | 7 | 5 | 3 |
| 15 | 65 | 68 | 72 | 10 | 8 | 5 | 8 | 7 | 4 |
| 20 | 67 | 70 | 76 | 13 | 9 | 7 | 12 | 8 | 6 |
| 25 | 69 | 75 | 79 | 18 | 15 | 10 | 15 | 11 | 8 |
| **.jpg** | 5 | 65 | 67 | 69 | 5 | 4 | 2 | 5 | 3 | 2 |
| 10 | 68 | 72 | 75 | 7 | 5 | 3 | 6 | 5 | 2 |
| 15 | 70 | 73 | 78 | 9 | 8 | 7 | 7 | 7 | 6 |
| 20 | 75 | 77 | 80 | 11 | 10 | 8 | 9 | 9 | 6 |
| 25 | 79 | 81 | 83 | 13 | 12 | 10 | 12 | 11 | 9 |

**Table 3 Computation of compression and decompression time for audio-visual files using ZAV compression method**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **File format** | **File size (MB)** | **Compression ratio (%)** | **Compression time (sec)** | **Decompression time (sec)** |
| **audio** | 5 | 75 | 8 | 6 |
| 10 | 78 | 12 | 11 |
| 15 | 81 | 15 | 14 |
| 20 | 83 | 20 | 18 |
| 25 | 85 | 24 | 21 |
| **video** | 5 | 70 | 15 | 17 |
| 10 | 74 | 19 | 23 |
| 15 | 76 | 22 | 27 |
| 20 | 81 | 26 | 29 |
| 25 | 83 | 31 | 34 |

**Table 4.Comparison between different metrics of different techniques**

|  |  |  |  |
| --- | --- | --- | --- |
| **File format** | **File size (MB)** | **Existing methods** | **Proposed method (Interpolation search)** |
| **Compression Ratio (%)** | **Compression Time(sec)** | **Decompression Time(sec)** | **Compression Ratio (%)** | **Compression Time(sec)** | **Decompression Time(sec)** |
| **text [19]** | 1.879(SF) | 38.10 | 0.062 | - | 46.0 | 0.055 | 0.052 |
| 3.132(SF) | 38.02 | 0.101 | - | 42.0 | 0.089 | 0.082 |
| 3.871(SF) | 33.67 | 0.127 | - | 41.0 | 0.119 | 0.111 |
| **avg.** |  | **36.59** | **0.096** | - | **43.0** | **0.087** | **0.081** |
| **doc [20]** | 0.031(SF) | 44.5 | 0.000 | - | 46.3 | 0.010 | 0.013 |
| 0.048(SF) | 67.0 | 0.015 | - | 73.0 | 0.014 | 0.017 |
| 0.210(SF) | 77.7 | 0.031 | - | 92.9 | 0.029 | 0.052 |
| **avg.** |  | **63.0** | **0.015** |  | **70.73** | **0.017** | **0.027** |
| **pdf [21]** | 0.039(SF) | 59.2 | 12.600 | 12.876 | 63.3 | 10.738 | 11.580 |
| 0.071(SF) | 62.6 | 42.940 | 32.845 | 68.6 | 36.858 | 31.345 |
| 0.118(SF) | 63.7 | 153.810 | 114.180 | 64.5 | 95.755 | 93.328 |
| **avg.** |  | **61.8** | **69.7** | **53.3** | **65.4** | **37.783** | **45.417** |
| **image [22]** | 2.7(FC) | 85.62 | 0.078 | - | 89.0 | 0.051 | 0.047 |
| 4.8(FC) | 84.50 | 0.156 | - | 90.0 | 0.137 | 0.127 |
|  **[23]** | 8.47(Deflate) | 1.404 | - | - | 63.0 | 0.474 | 0.432 |
| 9.97(Deflate) | 2.714 | - | - | 57.0 | 0.587 | 0.524 |
| **avg.** |  | **43.55** | **0.058** |  | **74.75** | **0.312** | **0.282** |
|  |  | **ZAV Compression Method** |
| **audio[20]** | 1.738(SF) | 94.0 | 0.109 | - | 95.0 | 0.890 | 0.170 |
|  | 4.790(SF) | 76.0 | 0.250 | - | 87.0 | 0.200 | 0.240 |
|  | 9.580(SF) | 94.1 | 0.515 | - | 97.0 | 0.450 | 0.430 |
| **avg.** |  | **88** | **0.29** |  | **93.0** | **0.513** | **0.280** |
| **video[24]** | 1.468(IFC) | 27.68 | - | - | 67.12 | 5.810 | 5.180 |
| 2.600(IFC) | 25.30 | - | - | 78.63 | 6.890 | 6.590 |
| 4.418(IFC) | 26.04 | - | - | 81.02 | 8.458 | 8.106 |
| **avg.** |  | **26.34** | **-** | **-** | **75.59** | **7.052** | **6.625** |

Table 4., is a comparison table between existing algorithms to proposed method. The proposed method includes interpolation search method to compare. For text, pdf, doc and audio file the existing algorithm i.e., shannon fano algorithm is used, for image file two different method: deflate and fibonacci code (FC) based method is used and for video inter-frame-based compression (IFC) is used against the proposed method. For text, doc, pdf existing method shows avg. 36.59%,63%, 61.8% average compression ratio whereas method improve to 43%,70.73%, 65.4% faster processing speed. For image files of size 2.7MB and 4.8MB files fibonacci method is used and for 8.4MB and 9.9MB deflate method is used but interpolation method again outperform by 74.75% compression ratio speed. While in audio and video file achieved 93.0% and 75.59% compression ratio, giving good performance while securing playback quality. Major advantage of the proposed method is its versatility which makes it suitable for cloud storage diversity. Thus this is ideal method for service provider to minimize storage and keeping data quality and accessibility. Comparative analysis shows the continuous outperformance of proposed method in terms of speed, latency and storage efficiency. Reduced file size not only improve cloud management but also enhance data transfers in less time in cloud based environment.

4. Conclusion

This work examined the compression algorithms for text, pdf, doc, images, audio and video files. A file must first be chosen before being compressed and sent over the cloud server. Similarly, for decompression, the uploaded file is first converted to its original state before being downloaded. Linear search, binary search and interpolation search are used for compressing the data. Based on the experimental findings, every search is conducted using files of various sizes and file types. In the above search strategies, compression ratio, compression and decompression time are compared to one another. The outcomes demonstrate that the suggested plan provides the optimum search for files in minimum computation time. However, a study of compression ratio revealed that interpolation search is the most effective in compression of above-mentioned techniques. While the other two binary and linear, are equally effective. Further the interpolation is compared with existing technique mentioned in the literature. In result, interpolation again gives the most effective results, when considering the time from techniques used during compression and decompression and compression ratio. The strength of this research is adaptation on different file types and quality to reduce storage problem without data loss. Unless earlier studies that work on format-specific compression, this work presents a compression model that enhances cloud storage efficiency. Moreover, the direct comparison with existing techniques further substantiates its effectiveness in real-world applications. Future work is to optimization strategy to reduce the computational complexity of the proposed technique, making it more scalable for large-scale cloud environments. Integrating with machine learning models to can create most efficient compression technique which further enhance speed. Another improvement is the extension of this method to ensuring both compression efficiency and data security. This work delivers storage optimization by giving lossless and efficiency compression method that balances space savings, speed, and scalability. In result it motivates future research into smart compression method that can further refine data storage and retrieval processes in cloud server.

**COMPETING INTERESTS DISCLAIMER:**

Authors have declared that they have no known competing financial interests OR non-financial interests OR personal relationships that could have appeared to influence the work reported in this paper.

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