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**A**  
***HPC Project***  
***Report On***

Generic Compression

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# **ABSTRACT**

In today's environment, data is created at a breakneck speed. As a result, data compression is quite useful. In this project, we're using run length encoding. Run-length encoding (RLE) is a very simple method of lossless data compression in which data runs (that is, sequences in which the same data value appears in many consecutive data elements) are saved as a single data value and count instead of the original run. This is especially handy when dealing with data that comprises a large number of such runs.

Simple visual graphics, such as icons, line drawings, and animations, are good examples. It's not recommended for files with few runs because it might significantly increase the file size. Additionally, when dealing with huge data sets, parallel compression will save a significant amount of time.

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## **1. PROBLEM STATEMENT**

To create a parallel approach to conduct 'run length encoding' on several core GPUs at the same time.

RLE is **a basic form of data compression that converts consecutive identical values into a code consisting of the character and the number marking the length of the run.** The more similar values there are, the more values can be compressed.

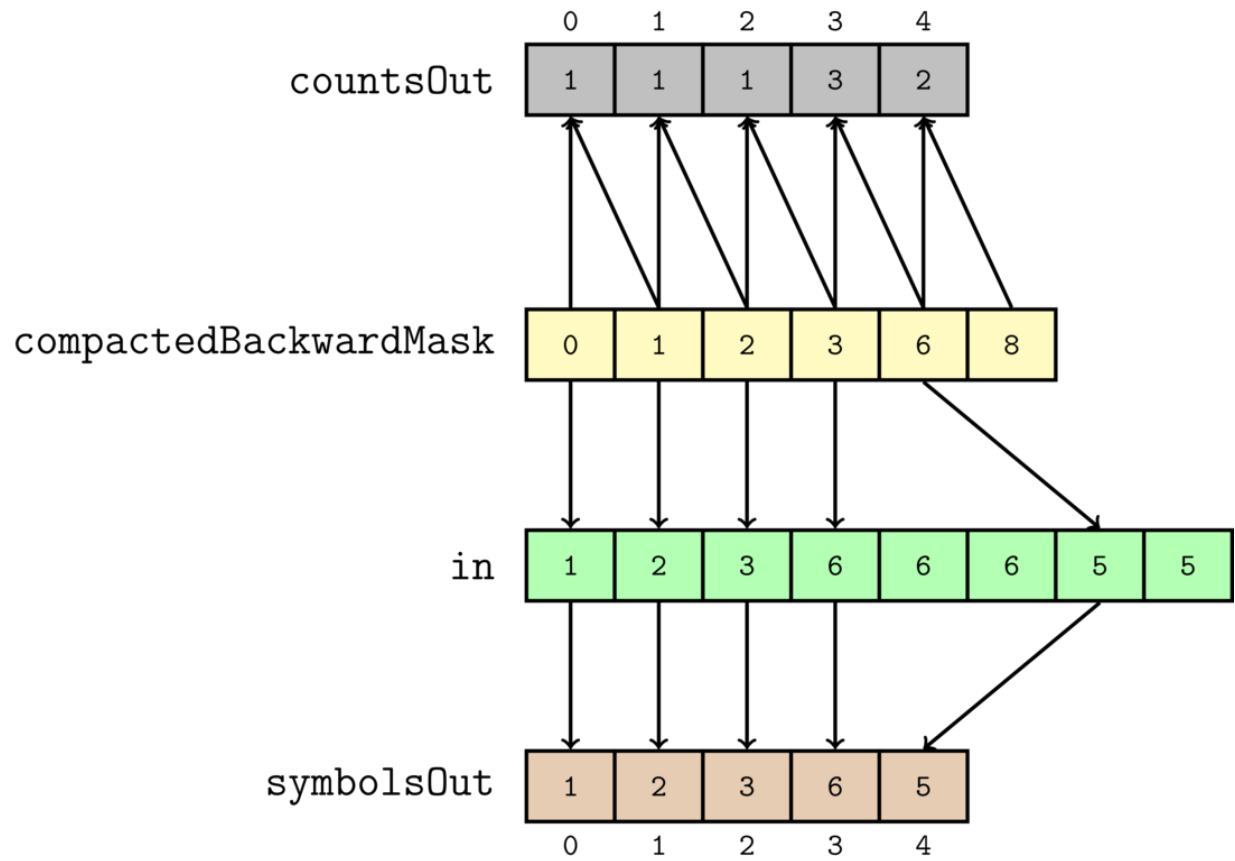
Run-length encoding (RLE) is a form of lossless data compression in which runs of data (sequences in which the same data value occurs in many consecutive data elements) are **stored as a single data value and count**, rather than as the original run. This is most useful on data that contains many such runs.

## **2. DOMAIN**

High Performance Computing.

We'll write a technique to execute run length encoding on many core GPUs concurrently.

### 3.METHODOLOGY



## **4.SERIAL ALGORITHM**

```
class RunLengthEncoding
{
    // Perform Run Length Encoding (RLE) data compression algorithm
    // on String str
    public static String encode(String str)
    {
        // stores output String
        String encoding = "";
        int count;

        for (int i = 0; i < str.length(); i++)
        {
            // count occurrences of character at index i
            count = 1;
            while (i + 1 < str.length() && str.charAt(i) == str.charAt(i+1)) {
                count++;
                i++;
            }

            // append current character and its count to the result
            encoding += String.valueOf(count) + str.charAt(i);
        }

        return encoding;
    }
}
```

## **5. PARALLEL ALGORITHM**

// Backward Masks of all elements are calculated with this program.

```
__global__ void maskKernel(int *g_in, int* g_backwardMask, int n) {  
    for (int i : hemi::grid_stride_range(0, n)) {  
  
        if (i == 0)  
            g_backwardMask[i] =  
                1;  
  
        else {  
            g_backwardMask[i] = (g_in[i] != g_in[i - 1]);  
        }  
  
    }  
}
```

```
__global__ void compactKernel(int* g_scannedBackwardMask,  
                             int* g_compactedBackwardMask,  
                             int* g_totalRuns,  
                             int n) {  
  
    for (int i : hemi::grid_stride_range(0, n)) {  
  
        if (i == (n - 1)) {  
            g_compactedBackwardMask[g_scannedBackwardMask[i]]  
                = i + 1;  
            *g_totalRuns = g_scannedBackwardMask[i];  
        }  
  
        if (i == 0) {  
            g_compactedBackwardMask[0]
```



```

        = 0;
    }
    else if (g_scannedBackwardMask[i] !=
            g_scannedBackwardMask[i - 1]) {
        g_compactedBackwardMask[g_scannedBackwardMask[i] - 1]
        = i;
    }
}

```

### //Final Compression Algorithm

```

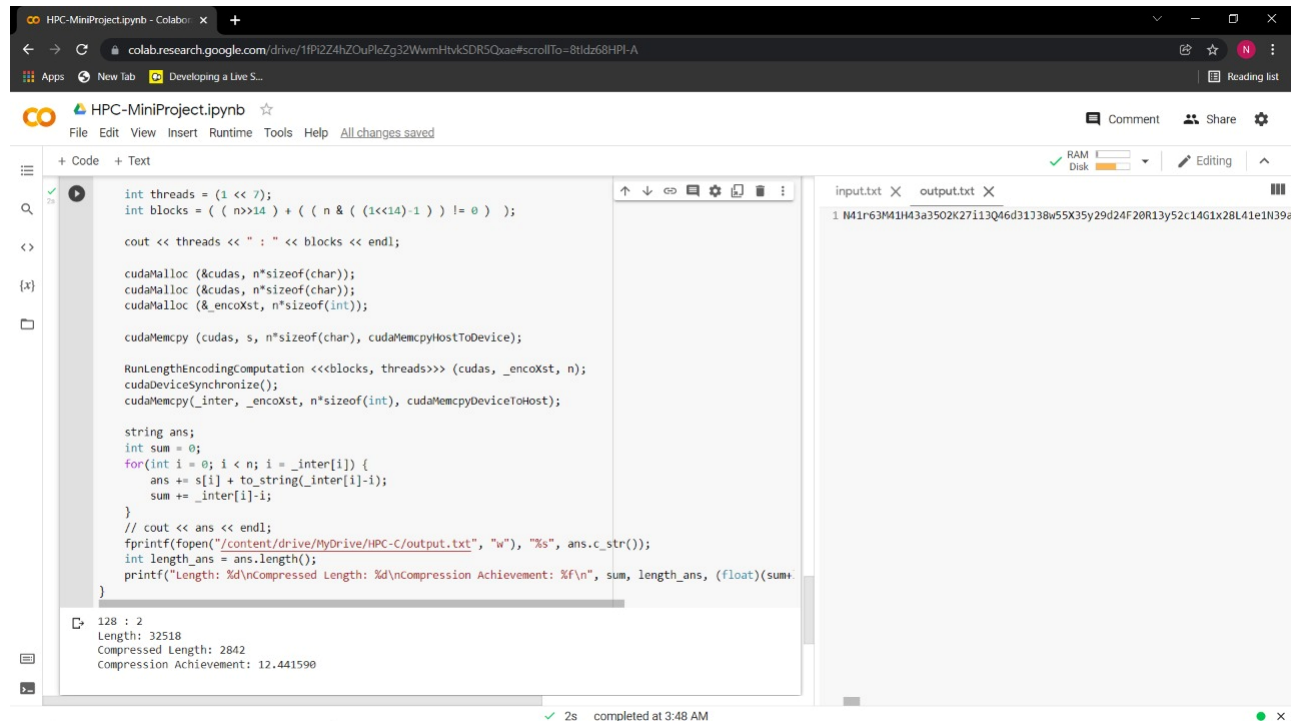
__global__ void scatterKernel(
    int* g_compactedBackwardMask,
    int* g_totalRuns,
    int* g_in,
    int* g_symbolsOut,
    int* g_countsOut) {
    int n = *g_totalRuns;

    for (int i : hemi::grid_stride_range(0, n)) {
        int a = g_compactedBackwardMask[i];
        int b = g_compactedBackwardMask[i + 1];

        g_symbolsOut[i] =
            g_in[a]; g_countsOut[i]
            = b - a;
    }
}

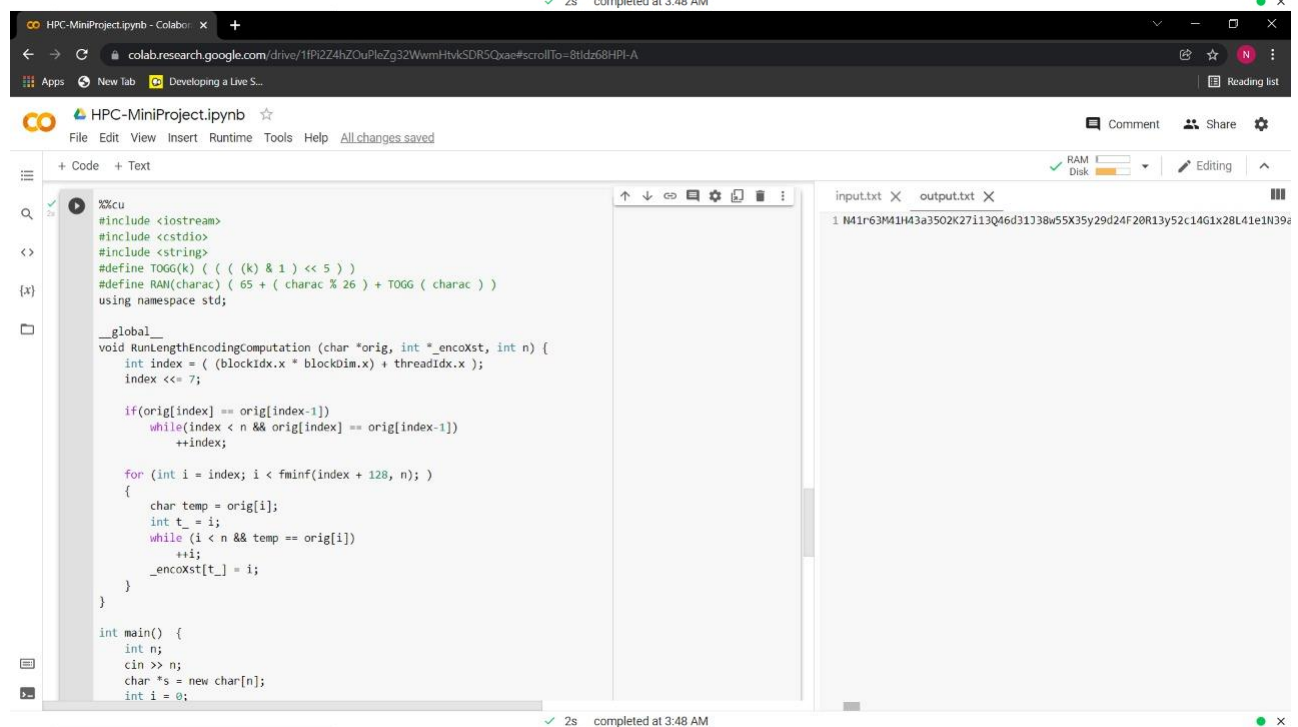
```

## 6. Output



The screenshot shows a Jupyter Notebook interface with a C++ code cell. The code defines a function `RunLengthEncodingComputation` that takes a character array `s`, an integer `n`, and an integer `encoxst`. It uses CUDA-like memory management and a loop to process the array. The output of the code is displayed in a text box:

```
128 : 2
Length: 32518
Compressed Length: 2842
Compression Achievement: 12.441590
```



The screenshot shows a Jupyter Notebook interface with a C++ code cell. The code defines a function `RunLengthEncodingComputation` that takes a character array `orig`, an integer `encoxst`, and an integer `n`. It uses a loop to process the array and calculate the compressed length. The output of the code is displayed in a text box:

```
1 H41r63M41H43a3502K27113Q46d31J38w55X35y29d24F20R13y52c14G1x28L41e1N39z
```

## **7.CONCLUSION**

As a result, we were able to successfully implement the run length encoding technique on a multi-core GPU.

