

Parallel Edit Distance

Parallel Programming for Machine Learning

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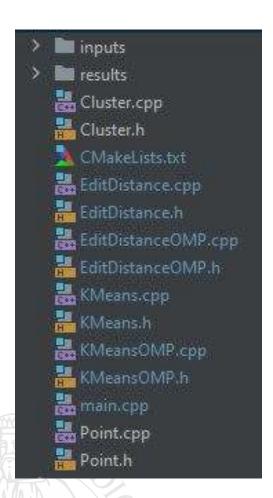
Introduction

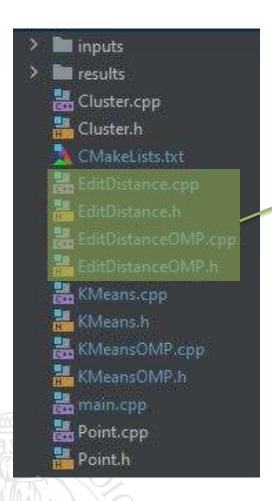
Edit Distance, also known as Levenshtein Distance, is a measure of similarity between two strings A and B. It considers the minimum number of **basic operations** between characters (insertion, substitution, deletion) needed to transform string A into string B. It was chosen to evaluate each basic operation with the same impact, so each of them increases the Edit Distance value by 1.

There are three **sequential** implementations of Edit Distance and two **parallel** implementations. The purpose of this work is to observe the **speedup** obtained with the parallel versions compared to the sequential ones.

The programming language is C++ (MinGW compiler, CLion IDE), and parallelization was done with **OpenMP**.

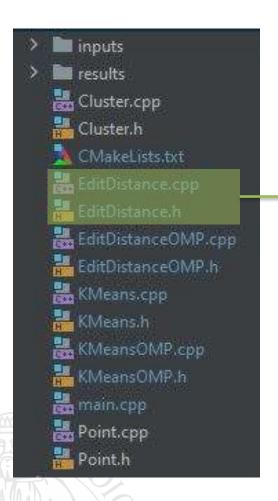
All experiments were conducted on the ssh server "papavero.dinfo.unifi.it".





Files included in this project. The remaining ones are related to another project with a common directory.

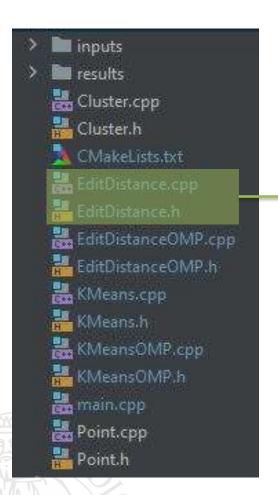




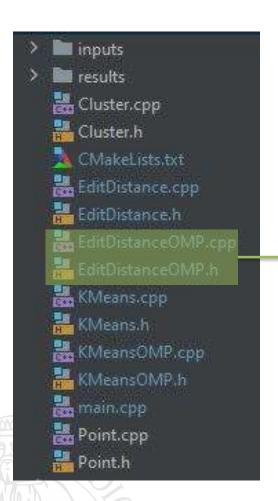
Files containing the sequential implementations of Edit Distance: Full Matrix, Skew Diagonal, and Matrix Row.

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Code Structure

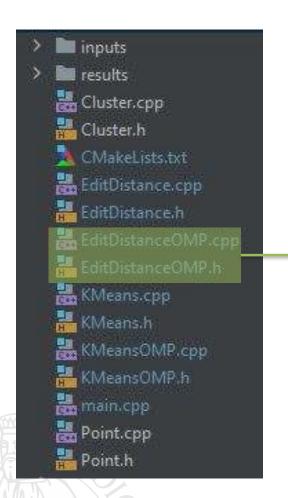


```
int levenshteinDistFM(const std::string& word1, const std::string& word2, int length1, int length2);
int levenshteinDistSD(const std::string& word1, const std::string& word2, int length1, int length2);
int levenshteinDistMR(const std::string& word1, const std::string& word2, int length1, int length2);
```



Files containing the two parallelized versions of Edit Distance using as base the Skew Diagonal approach.





Sequential Algorithm: Full Matrix

```
int levenshteinDistFM(const std::string& word1, const std::string& word2, int length1, int length2) {
   int N = length1 + 1;
   std::vector<int> distMatrix( m N * M);
   if (length2 == 0) {
       distMatrix[i] = i;
       distMatrix[j * M] = j;
           int substitutionCost = (word1[i - 1] == word2[j - 1]) ? 0 : 1;
           distMatrix[i * N + j] = std::min(std::min(distMatrix[(i - 1) * M + j] + 1, //deletion cost
                                                distMatrix[i * M + j - 1] + 1), //insertion cost
                                       distMatrix[(i - 1) * M + j - 1] + substitutionCost); //substitution cost
   int result = distMatrix[length1 * N + length2];
   return result:
```

Edit Distance matrix computation.

Sequential Algorithm: Skew Diagonal

```
int dMIN = 1 - M;
int dMAX = N - 1;
for (int d = dMIN; d <= dMAX; d ++) {
    int iMIN = std::max(d, 1);
    int iMAX = std::min(M + d, N - 1);
    for (int i = iMIN; i <= iMAX; i ++) {
        if (word1[i - 1] != word2[j - 1]) {
            distMatrix[i * N + j] = std::min(std::min(distMatrix[(i - 1) * M + j],
                                                      distMatrix[i * M + j - 1])
                                             distMatrix[(i - 1) * M + j - 1]) + 1;
            distMatrix[i * N + j] = distMatrix[(i - 1) * N + j - 1];
    if (d == -1) {
        d += 2
int result = distMatrix[length1 * N + length2];
return result;
```

```
egin{pmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1m} \ a_{21} & a_{22} & a_{23} & \dots & a_{2m} \ a_{51} & a_{32} & a_{33} & \dots & a_{3m} \ \vdots & \vdots & \vdots & \vdots & \vdots \ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nm} \end{pmatrix}
```

Sequential Algorithm: Matrix Row

```
nt levenshteinDistMR(const std::string& word1, const std::string& word2, int length1, int length2) {
  std::vector<int> prevRow( n: length2 + 1, value 0);
  if (length2 == 0) {
      prevRow[j] = j;
      currRow[0] = i + 1;
          int deletionCost = prevRow[j + 1] + 1;
          int insertionCost = currRow[j] + 1;
          int substitutionCost;
          if (word1[i] == word2[j]) {
              substitutionCost = prevRow[j];
              substitutionCost = prevRow[j] + 1;
          currRow[j + 1] = std::min(std::min(deletionCost, insertionCost), substitutionCost);
  return currRow[length2];
```

i-th row of the matrix based on the (i-1)-th row.

Parallelized Skew Diagonal

Single parallel overhead

```
for (int d = dMIN; d <= dMAX; d++) {
   int iMIN = std::max(d, 1);
   int iMAX = std::min(M + d, N - 1);
   for (int i = iMIN; i <= iMAX; i++) {
       if (word1[i - 1] != word2[j - 1]) {
           distMatrix[i * N + j] = std::min(std::min(distMatrix[(i - 1) * M + j],
                                                      distMatrix[i * M + j - 1]),
                                             distMatrix[(i - 1) * M + j - 1]) + 1;
       } else {
           distMatrix[i * N + j] = distMatrix[(i - 1) * N + j - 1];
   if (d == -1) {
```

Parallelized Skew Diagonal

Multiple parallel overhead

```
for (int d = dMIN; d <= dMAX; d++) {
   int iMIN = std::max(d, 1);
   int iMAX = std::min(M + d, N - 1);
   shared(distMatrix, word1, word2) num_threads(threads)
   for (int i = iMIN; i <= iMAX; i++) {
       if (word1[i - 1] != word2[j - 1]) {
           distMatrix[i * N + j] = std::min(std::min(distMatrix[(i - 1) * M + j],
                                                      distMatrix[i * M + j - 1]),
                                             distMatrix[(i - 1) * M + j - 1]) + 1;
       } else {
           distMatrix[i * N + j] = distMatrix[(i - 1) * N + j - 1];
   if (d == -1) {
```



```
Inputs
results
📇 Cluster.cpp
Cluster.h
丸 CMakeLists.txt
Edit Distance.cpp
EditDistance.h
EditDistanceOMP.cpp
🚟 EditDistanceOMP.h .
KMeans.cpp
📇 KMeans.h
# KMeansOMP.cpp
KMeansOMP.h
Point.cpp
```

H Point.h

```
std::string random_string(std::size_t length, int seed) {....}
louble testStringSearchFMTime(const std::string& word1, const std::string& word2, bool repeat) {...}
louble testStringSearchSDTime(const std::string& word1, const std::string& word2, bool repeat) {...}
double testStringSearchMRTime(const std::string& word1, const std::string& word2, bool repeat) {...}
<code>louble testStringSearchSD_OMPTime(const std::string& word1, const std::string& word2, int nThreads, bool repeat) {...}</code>
louble testStringSearchSD_OMP2Time(const std::string& word1, const std::string& word2, int nThreads, bool repeat) {...}
std::vector<double> compareTimeStringSearch(const std::string& word1, const std::string& word2, int nThreads,
                                            bool repeat = true) {...}
ouble compareTimeStringSearchSD(const std::string& word1, const std::string& word2, int nThreads,
```



Comparison between the various approaches, 16 threads

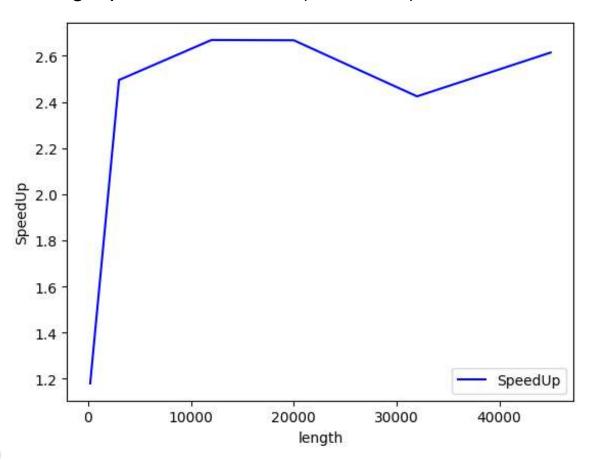
	Full Matrix	Skew Diagonal	Matrix Row
P1, len = 5000	3. 2071	3. 70324	2. 31112
P2, len = 5000	2. 85164	3. 29279	2. 05497
P1, len = 15000	3. 3498	5. 15474	2. 43479
P2, len = 15000	3. 18759	4. 90512	2. 31688
P1, len = 28000	3. 63548	6. 48565	2. 58662
P2, len = 28000	3. 52203	6. 28326	2. 5059



Comparison between the various approaches, 4 threads

	Full Matrix	Skew Diagonal	Matrix Row
P1, len = 5000	1. 94831	2. 22286	1. 3185
P2, len = 5000	2. 02005	2. 3047	1. 36705
P1, len = 15000	1. 85594	2. 8919	1. 34012
P2, len = 15000	1. 8461	2. 87656	1. 33301
P1, len = 28000	1. 84884	3. 40718	1. 33429
P2, len = 28000	1. 86523	3. 43739	1. 34611

Varying string lengths, speedup calculated between Full Matrix and Skew Diagonal with single parallel overhead (8 threads)



Test 3Varying the number of threads, with strings length of 20,000.

