Experimental Project:

Common Subsequence Algorithms

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CSCI-665 @ RIT, Winter 2016

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Abstract

Four algorithms that establish the length of a longest common subsequence (LCS) of two arbitrary strings are implemented and compared side by side. The preliminary submittal contains the implementations of the naive, memoized and bottom-up dynamic algorithms to calculate the length of the LCS, and a sample run of the main driver program that measures and plots the run time of select implementations. Not yet implemented are the Hirschberg algorithm, as well as the portions of each algorithm (except naive) dealing with LCS reconstruction.

Table of Contents

Abstract	1
	3
	4
1 Introduction	5

2	Naive Algorithm	6
3	Top-Down Memoized Algorithm	10
4	Bottom-Up Dynamic Programming Algorithm	11
5	Hirschberg Linear Space Dynamic Programming Algorithm	14
6	Appendix 1: Naive Algorithm Implementation	15
7	Appendix 2: Memoized Algorithm Implementation	17
8	Appendix 3: Bottom-Up DP Algorithm Implementation	19
9	Appendix 4: Hirschberg DP Algorithm Implementation	21
10	Appendix 5: Driver Program	22
11	Appendix 6: Plotter Program	25
12	Appendix 7: String Generator Program	27
13	Appendix 8: Performance Profiler Program	28
Re	eferences	30

• •

List of Figures

2.1	CPU time vs input string length: naive algorithm	8
2.2	CPU time vs recursion depth: naive algorithm	9
4.1	CPU time vs input string length: memoization vs dynamic	
	algorithms	12
4.2	CPU time for a long input string: dynamic algorithm	13

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List of Tables

Introduction

The implementation of all algorithms except Hirschberg's quadratic-time linear-space algorithm is based on (Cormen & al. 2009).

Naive Algorithm

The naive recursive solution is based on recursion (15.9) in (Cormen & al. 2009) repeated here for clarity.

$$c[i,j] = \begin{cases} 0, & \text{if } i = 0 \text{ or } j = 0, \\ c[i-1,j-1] + 1 & \text{if } i,j > 0 \text{ and } x_i = y_i, \\ max(c[i,j-1], c[i-1,j]) & \text{if } i,j > 0 \text{ and } x_i \neq y_j \end{cases}$$
 (2.1)

For the Python implementation, see listing in sec. 6.

Figure 2.1 shows a run time profiler results using the naive algorithm for strings of varying length and structure. For a feasible run time, only strings of length 10 and 15 were used. Both binary and character alphabets are compared.

Figure 2.2 shows a recursion depth profiler run using the naive algorithm

and character alphabet for strings of varying length and structure. For a feasible run time, only strings of length 10 and 15 were used.

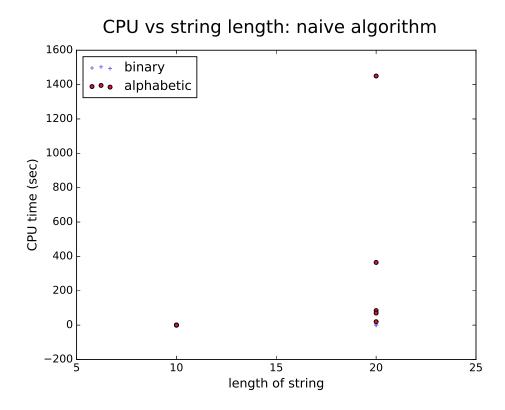


Figure 2.1: CPU time vs input string length: naive algorithm.

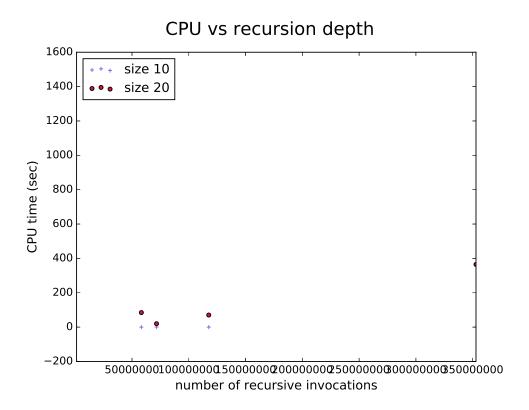


Figure 2.2: CPU time vs recursion depth: naive algorithm.

Top-Down Memoized

Algorithm

The memoized implementation uses top-down recursion essentially identical to the naive approach in sec. 6, except that performed computations are saved in a table. *TODO*: add pseudo-code.

For the Python implementation, see listing in sec. 7.

Bottom-Up Dynamic

Programming Algorithm

The DP implementation uses bottom-up iterative approach in Fig. 15.8 in (Cormen & al. 2009). *TODO*: add pseudo-code.

For the Python implementation, see listing in sec. 8.

Figure 4.1 shows a run time profiler output comparing the bottom-up dynamic vs top-down memoization algorithm performance for the same string input.

Figure 4.2 shows a run time profiler output for a string of size 10,000.

CPU vs input length: memoized vs dynamic

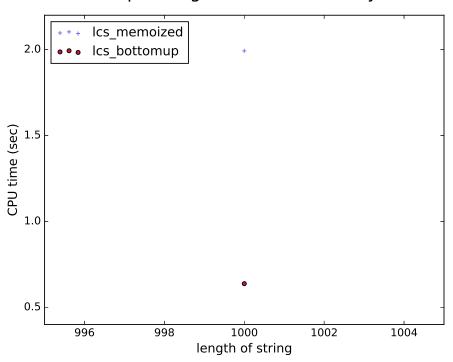


Figure 4.1: CPU time vs input string length: memoization vs dynamic algorithms.

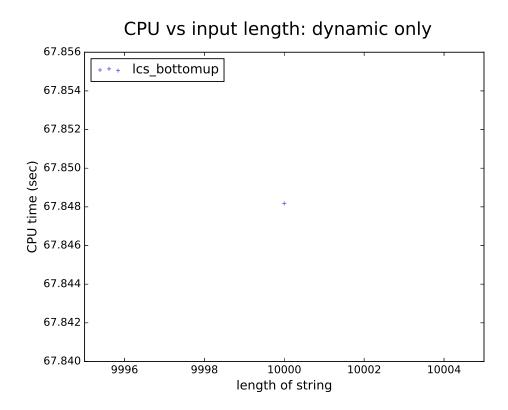


Figure 4.2: CPU time for a long input string: dynamic algorithm.

Hirschberg Linear Space Dynamic Programming Algorithm

TODO: This is a stub for a future section.

For the Python implementation, see listing in sec. 9.

Appendix 1: Naive Algorithm Implementation

Following is the implementation of the naive algorithm in sec. 2:

```
from profilers import len_recursion, time_profiler, registry
   from generate_string import strgen
18
19
   @time_profiler(repeat = 1)
21
  def lcs_naive(seq1, seq2):
       """Calls helper function to calculate an LCS."""
24
      return _lcs_naive(seq1, seq2, len(seq1)-1, \
25
                 len(seq2)-1, "")
26
29 @len_recursion
30 def _lcs_naive(seq1, seq2, i, j, lcs):
       """Naive recursive solution to LCS problem.
      See CLRS pp.392-393 for the recursive formula.
32
33
      Args:
          seq1 (string): a string sequence generated
                            by generate_string.strgen()
          seq2 (string): another random string sequence
                            like seq1
          i (int): index into seq1
39
          j (int): index into seq2
40
          lcs (string): an LCS string being built-up
      Returns:
          lcs: longest common subsequence (can be empty
```

```
string)
      11 11 11
45
46
      if i < 0 or j < 0:
47
         return lcs
48
      else:
49
         if seq1[i] == seq2[j]:
50
            52
         else:
53
            return max(_lcs_naive(seq1, seq2, i-1, j, \
54
                      lcs),
55
                   _lcs_naive(seq1, seq2, i, j-1, lcs), key=len)
56
```

 $/home/max/classes/16_spring/algorithms/project/pylib/naive.py$

Appendix 2: Memoized

Algorithm Implementation

Following is the implementation of the memoized dynamic programming algorithm in sec. 3:

```
16 from profilers import len_recursion, time_profiler, registry
  from generate_string import strgen
18
19
   @time_profiler(repeat=1)
20
   def lcs_memoized(seq1, seq2):
21
       """Calls helper function to calculate an LCS.
22
          seq1 (string): a random string sequence
                            generated by generate_string.strgen()
          seq2 (string): another random string sequence like seq1
27
          LCS length (int)
       11 11 11
      len1 = len(seq1)
32
      len2 = len(seq2)
33
34
      \# store length of LCS[i,j] in lcs_table
35
      lcs_table = [[None for j in range(len2)] \
36
                     for i in range(len1)]
      _lcs_memoized(seq1, seq2, len1-1, len2-1, \
                     lcs_table)
      return lcs_table[len1-1][len2-1]
```

```
@len_recursion
42
        _lcs_memoized(seq1, seq2, i, j, lcs_table):
43
       """Recursive solution with memoization to LCS problem.
44
       See CLRS ex. 15.4-3.
45
46
       Args:
47
          seq1 (string): a string sequence generated by
49
                             generate_string.strgen()
          seq2 (string): another random string sequence like seq1
50
          i (int): index into seq1
51
           j (int): index into seq2
52
          lcs_table (2D list): a matrix of LCS length for
53
                                 [i, j] prefix
54
55
       Returns:
       None: modifies in place LCS length table
57
58
       if i < 0 or j < 0:
59
          return 0
60
       else:
61
          if lcs_table[i][j] is not None:
62
63
              return lcs_table[i][j]
64
          else:
              if seq1[i] == seq2[j]:
65
                  val = 1 + \
66
                      _lcs_memoized(seq1, seq2, i-1, \
67
                                    j-1, lcs_table)
68
              else:
69
                  val = max(_lcs_memoized(seq1, seq2, \
70
                                    i-1, j, lcs_table),
71
                         _lcs_memoized(seq1, seq2, i, \
72
                                    j-1, lcs_table))
73
74
              lcs_table[i][j] = val
75
              return val
```

/home/max/classes/16_spring/algorithms/project/pylib/memoized.py

Appendix 3: Bottom-Up DP

Algorithm Implementation

Following is the implementation of the bottom-up dynamic programming algorithm in sec. 4:

```
17 from profilers import len_recursion, time_profiler, registry
  from generate_string import strgen
19
def lcs_bottomup(seq1, seq2):
    """Calls helper function to calculate an LCS.
21
22
23
       Args:
           seq1 (string): a random string sequence generated by
                           generate_string.strgen()
           seq2 (string): another random string sequence like seq1
       Returns:
           LCS length (int)
       len1 = len(seq1)
       len2 = len(seq2)
        \begin{tabular}{ll} \# \ store \ length \ of \ LCS[i,j] \ in \ lcs\_table \\ \end{tabular} 
       lcs_table = [[0 for j in range(len2+1)] \
36
       for i in range(len1+1)]
_lcs_bottomup(seq1, seq2, len1+1, len2+1,
                       lcs_table)
       return lcs_table[len1][len2]
```

```
# @to_profile
43
   def _lcs_bottomup(seq1, seq2, i, j, lcs_table):
44
       """Iterative bottom-up dynamic programming solution to
45
       LCS problem. See CLRS p.394.
46
47
48
       Args:
          seq1 (string): a string sequence generated by
50
                             generate_string.strgen()
          seq2 (string): another random string sequence
51
                             like seq1
52
          i (int): number of rows in LCS table
53
                             (=len(seq1) + 1)
54
          j (int): number of columns in LCS table
55
56
                             (=len(seq2) + 1)
          lcs_table (2D list): a matrix of LCS length for
                                [i-1, j-1] prefix
58
       Returns:
59
          None: modifies in place LCS length table
60
61
62
       for row in range(1, i):
63
          for col in range(1, j):
              if seq1[row-1] == seq2[col-1]:
65
                  lcs_table[row][col] = \
66
                     lcs_table[row-1][col-1] + 1
67
              elif lcs_table[row-1][col] \
68
                     >= lcs_table[row][col-1]:
69
                  lcs_table[row][col] = \
70
                         lcs_table[row-1][col]
71
              else:
72
                  lcs_table[row][col] = \
73
                         lcs_table[row][col-1]
```

/home/max/classes/16_spring/algorithms/project/pylib/dynamic.py

Appendix 4: Hirschberg DP Algorithm Implementation

TODO: this is a stub for a future section.

Appendix 5: Driver Program

Listing for the overall driver program:

```
23 import os.path, sys
24 import importlib
_{25} from plot import plot_scatter
26 from generate_string import strgen
28
   import naive
   import memoized
   import dynamic
   #import hirschberg
33
   # increase recursion limit
34
  sys.setrecursionlimit(100000)
35
  # set up directory refs
  CURDIR = os.path.abspath(os.path.curdir)
  FIGDIR = os.path.join(os.path.dirname(CURDIR),\
              'docs/source/figures')
40
  LOGDIR = os.path.join(CURDIR, 'logs')
41
42
   # alphabets
43
   ALPHAS = {'bin': ['0', '1']},
            'alpha': ['A','C','G','T']}
45
46
47
   if __name__ == "__main__":
48
       ## (1) for each algorithm
       # plot CPU time vs string length for several
       # inputs at each length
      # set legend by length of LCS or alphabet
str_lens = [10, 20]
```

```
title = 'CPU vs string length: naive algorithm'
       series = []
56
57
       set1 = {'x':[], 'y':[]}
58
       seq1_bin = [strgen(alphabet=ALPHAS['bin'], \
59
                  size=strlen) for i in range(5) \
60
                  for strlen in str lens]
61
       seq2_bin = [strgen(alphabet=ALPHAS['bin'], \
                  size=strlen) for i in range(5) \
63
                  for strlen in str_lens]
64
       for (str1, str2) in zip(seq1_bin, seq2_bin):
65
           algo_name, time_elapsed, lcs = \
66
                  naive.lcs_naive(str1, str2)
67
           set1['x'].append(len(str1))
68
           set1['y'].append(time_elapsed)
69
       series.append('binary')
70
71
       #store CPU vs length
72
       set2 = {'x':[], 'y':[]}
73
       set3 = {'x':[], 'y':[]}
74
       set4 = \{'x':[], 'y':[]\}
75
       seq1_alpha = [strgen(alphabet=ALPHAS['alpha'], \
76
                      size=strlen) for i in range(5) \
77
                      for strlen in str_lens]
78
       seq2_alpha = [strgen(alphabet=ALPHAS['alpha'], \
79
                      size=strlen) for i in range(5) \
80
                      for strlen in str_lens]
81
82
       for (str1, str2) in zip(seq1_alpha, seq2_alpha):
83
           strlen = len(str1)
84
           algo_name, time_elapsed, lcs = \
                  naive.lcs_naive(str1, str2)
86
           set2['x'].append(len(str1))
87
           set2['y'].append(time_elapsed)
           if strlen == 10:
89
              set3['x'].append(\
                      naive.registry['_lcs_naive'])
               set3['y'].append(time_elapsed)
           elif strlen == 20:
              set4['x'].append(\
94
                      naive.registry['_lcs_naive'])
95
       set4['y'].append(time_elapsed)
series.append('alphabetic')
96
97
       plot_scatter(set1, set2, series, title, \
98
              xlabel = 'length of string', \setminus
              ylabel = 'CPU time (sec)')
100
101
       print("-> done with naive algorithm runs")
102
103
       # (2) plot CPU time vs number of recursive
104
       # invocations for several lengths
105
106
       plot_scatter(set3, set4, ['size 10', 'size 20'], \
107
              title = 'CPU vs recursion depth', \
              xlabel = 'number of recursive invocations',\
108
              ylabel = 'CPU time (sec)')
109
110
       print("-> done with CPU vs recursion depth plots")
111
```

```
# (3) TODO: plot memory use vs string length for
113
       # several inputs at each length
114
115
116
       ## (4) algorithm comparison
117
       # for given string, plot CPU time for each
118
       # algorithm; do this for several strings
       # (vary length and alphabet)
120
       title = 'CPU vs input length: memoized vs dynamic'
121
       str1 = strgen(alphabet=ALPHAS['alpha'], size=1000)
122
       str2 = strgen(alphabet=ALPHAS['alpha'], size=1000)
123
124
       set1 = {'x':[], 'y':[]} #store CPU vs length: memoized
125
126
       labels = []
       algo_name, time_elapsed, lcs_len = \
              memoized.lcs_memoized(str1, str2)
128
       set1['x'].append(len(str1))
129
       set1['y'].append(time_elapsed)
130
       labels.append(algo_name)
131
132
       #store CPU vs length: bottom-up dynamic
133
       set2 = {'x':[], 'y':[]}
       algo_name, time_elapsed, lcs_len = \
135
              dynamic.lcs_bottomup(str1, str2)
136
       set2['x'].append(len(str1))
137
       set2['y'].append(time_elapsed)
138
       labels.append(algo_name)
139
       plot_scatter(set1, set2, labels, title, \
140
                  xlabel = 'length of string', \
141
                  ylabel = 'CPU time (sec)')
143
144
       print("-> done with CPU vs length plots")
145
146
       ## (5) show time for strlen == 40000
147
       title = 'CPU vs input length: dynamic only'
148
       str1 = strgen(alphabet=ALPHAS['alpha'], \
              size=10000)
150
       str2 = strgen(alphabet=ALPHAS['alpha'], \
151
              size=10000)
152
153
       set1 = {'x':[], 'y':[]} #store CPU vs length: dynamic
154
       labels = []
155
       algo_name, time_elapsed, lcs_len = \
156
              dynamic.lcs_bottomup(str1, str2)
157
       set1['x'].append(len(str1))
158
       set1['y'].append(time_elapsed)
159
       labels.append(algo_name)
160
161
       plot_scatter(set1, [], labels, title, \
162
                  xlabel = 'length of string',
163
                  ylabel = 'CPU time (sec)')
164
165
166
       print("-> done with plot for dynamic @ 10K long string")
167
```

Appendix 6: Plotter Program

Listing for the plotting routine:

```
import matplotlib.pyplot as plt
   import os.path
4 CURDIR = os.path.abspath(os.path.curdir)
   DOCDIR = os.path.join(os.path.dirname(CURDIR), \
              'docs/source/figures')
   def plot_scatter(set1, set2, labels, title, xlabel, ylabel):
       """Save 2D scatter plot of numplots sets of data.
9
10
      Arqs:
11
          set1 (dict): dict of coordinate lists; set1 =
12
                         \{'x':[list\ of\ x-coords],
13
                         'y':[list\ of\ y-coords]
          set2: ditto
          labels (list of strings): series labels
16
                                   for each data set
17
          title (string): plot title
18
          xlabel, ylabel (string): axes labels
19
20
      fig = plt.figure()
      axes = plt.gca()
23
      axes.set_xlim([set1['x'][0] - 5, \
24
              set1['x'][-1] + 5])
      plt.ticklabel_format(style='plain', axis='both', \
26
             useOffset = False)
27
      fig.suptitle(title, fontsize=20)
      plt.xlabel(xlabel, fontsize=14)
      plt.ylabel(ylabel, fontsize=14)
      if set2:
31
          plt.scatter(set1['x'], set1['y'], s=20,
```

```
c='mediumslateblue',
    marker='+', label = labels[0])
plt.scatter(set2['x'], set2['y'], s=20, c='crimson',
34
35
                     marker='o', label = labels[1])
36
        else:
37
            plt.scatter(set1['x'], set1['y'], s=20,
38
                     c='mediumslateblue',
marker='+', label = labels[0])
39
41
        plt.legend(loc='upper left')
42
        fname = '_'.join(title.replace(':','_').split()) + '.ps'
43
        fig.savefig(os.path.join(DOCDIR,fname))
44
45
        #plt.show()
```

 $/home/max/classes/16_spring/algorithms/project/pylib/plot.py$

Appendix 7: String Generator Program

Listing for the string generator routine:

```
#!/usr/bin/env python3
generate_string.py

Generate a string given alphabet and length of string.

Usage:
python3 generate_string.py
"""

__author__ = "Maksim Yegorov"
__date__ = "2016-04-06 Wed 08:06 PM"

from random import choice

def strgen(alphabet=['0', '1'], size=40000):
    """Generates string of characters from alphabet of given length."""
astring = ""
for i in range(size):
    astring += choice(alphabet)
return astring
```

/home/max/classes/16_spring/algorithms/project/pylib/generate_string.py

Appendix 8: Performance

Profiler Program

Listing for runtime, recursion depth and memory profilers:

```
19 import time, sys
20 from memory_profiler import profile as mem_profiler
21 from collections import defaultdict
   # keep track of recursive function calls
  registry = defaultdict(int)
24
26 def len_recursion(func):
      """Decorator that counts the number of function
27
      invocations.
28
      Args:
         func: decorated function
      Returns:
         decorated func
      Caveats:
34
      does not account for repeated runs!
      # count number of invocations
      def inner(*args, **kwargs):
          """Increments invocations and returns the
          callable unchanged."""
41
          registry[func.__name__] += 1
42
          return func(*args, **kwargs)
      return inner
```

```
def time_profiler(repeat = 1):
47
       """Decorator factory that times the function
48
       invocation. A function is timed over 'repeat' times
49
       and then runtime is averaged.
50
51
       Args:
52
          repeat (int): number of repeat runs to average
                             runtime over.
54
       Returns:
55
          decorated func
56
57
       def decorate(func):
58
           """Decorator.
59
60
          Args:
           func: decorated function
63
          def inner(*args, **kwargs):
64
              """Sets timer and returns the elapsed time
65
              and result of original function.
66
67
              Returns:
69
                  func.__name__, elapsed_time,
                     original_return_value (tuple)
70
              11 11 11
71
72
              start = time.perf_counter()
73
              for i in range(repeat):
74
                  return_val = func(*args, **kwargs)
75
              finish = time.perf_counter()
76
              elapsed = finish - start
77
78
              return (func.__name__, elapsed, return_val)
79
          return inner
80
       return decorate
```

/home/max/classes/16_spring/algorithms/project/pylib/profilers.py

References

Cormen & al., 2009. Introduction to Algorithms, Cambridge, Mass.: The MIT Press.