## COMP26120 Lab 5 Report

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1 What makes a problem hard for Dynamic Programming?

## 1.1 Hypothesis

In hypothesis, the capacity of the knapsack is an aspect that will make a knapsack problem hard for dynamic programming. More specifically, if the number of items is constant and each item's weight and value also remain constant, as the capacity of the knapsack increases, the amount of time for computing an optimal combination of items in the knapsack using dynamic programming will increase. The reason for this hypothesis is that dynamic programming uses a matrix of size (n+1)\*(w+1) to store the solutions to the subproblems, where each cell represents a subproblem. To fill in all (n+1)\*(w+1) cells in the matrix, and each cell requires constant time to compute, the time complexity for dynamic programming is O(n\*w) where w is the knapsack capacity, and n is the number of items.

## 1.2 Design

The experiment involves the following steps:

- 1. We generate ten .txt files using kp\_generate.py (in Appendix B.1), each with 1000 items. When generating these ten files, we set the upper bound and the knapsack capacity to 100 and 200, respectively, and each item's weight and value are randomly generated. We save the ten .txt files with knapsack capacity set to 200 in a folder named 200\_capacity\_input\_tests.
- 2. We make another nine copies of this folder, with knapsack capacity change to 400, 600, 800, 1000, 1200, 1400, 1600, 1800, and 2000, respectively (in Appendix A.1).
- 3. We run the ten files in the folder named 200\_capacity\_input\_tests using a shell script called custom\_test.sh (in Appendix B.2). We record the time taken to run each file in a CSV file (in Appendix A.2), which is measured using the UNIX time command. We repeat the third step for the folders with knapsack capacity set to 400, 600, 800, 1000, 1200, 1400, 1600, 1800, and 2000 respectively. We record all data in the same CSV file.
- 4. We use Pandas to read the CSV file and NumPy to calculate the total time to run all ten files with knapsack capacity = 200. The total time is divided by 10 to find the average time to run each of the ten files. It represents the time to compute an optimal combination of items in the knapsack given knapsack capacity = 200. We repeat the fourth step to calculate the time taken to compute an optimal combination of items in the knapsack given knapsack capacity = 400, 600, 800, 1000, 1200, 1400, 1600, 1800, and 2000, respectively.
- 5. We use pyplot to plot a resulting graph for all average time values with their corresponding knapsack capacity, with knapsack capacity on the x-axis and average time (measured in milliseconds) on the y-axis. A line of best fit is generated in the form of y = m\*x + c using NumPy's polyfit function, in which m is

the gradient of the line, and c is a constant value. Consequently, in this experiment, the only independent variable is the knapsack capacity, the controlled variable is the number of items and the weight and value of each item, and the dependent variable is the time taken to compute an optimal combination of items in the knapsack.

We generate the weight and the value of each item randomly to remove bias because an uneven distribution of weight and value for each item in this experiment might cause bias. We use a large sample size by generating ten files with 1000 items in each file for a specific knapsack capacity to reduce uncertainty and obtain more precise results estimates.

### 1.3 Results

We use Matplotlib and NumPy to generate the resulting graph in Figure 1. The graph shows how capacity affects the difficulty of the knapsack problem for dynamic programming, measured by the time taken (y-axis) to find a solution for 1000 items with a specific knapsack capacity (x-axis). NumPy's polyfit function generates the line of best fit of all data points in the graph with the formula: y = 0.24\*x+9.73, where 0.24 is the line gradient, and 9.73 is a constant value. The line of best fit with gradient value = 0.24 clearly shows a linear relationship between the knapsack capacity and the time to find a solution. More specifically, given the same number of items and for each item, its weight and value remain constant; as the capacity of the knapsack increases, the time taken for computing an optimal combination of items in the knapsack using dynamic programming will increase.

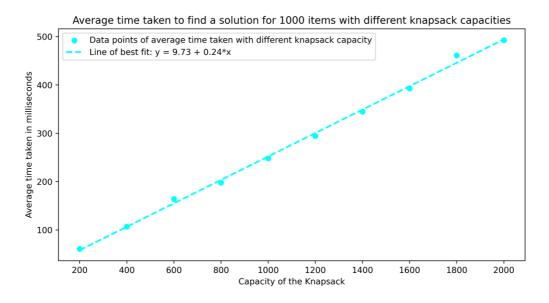


Figure 1: Time taken to find a solution for 1000 items with different knapsack capacities.

#### 1.4 Discussion

The result in the graph confirms the hypothesis that knapsack capacity is an aspect that will make a knapsack problem hard for dynamic programming. There are three perspectives to discuss the correctness and truthfulness of the hypothesis:

- 1. The line of best fit with gradient = 0.24 shows the linear relationship between knapsack capacity and the time taken to compute an optimal combination of items in the knapsack using dynamic programming. Given that the number of items stays constant, and their weight and value remain constant for each item, the time taken will increase as the knapsack capacity increases.
- 2. All data points are the actual measurements of average time taken with different knapsack capacities, and no data point deviates from the line of the best fit. Thus, the actual measurements also support the hypothesis. Furthermore, when performing the actual measurements, bias and uncertainty are eliminated, which means the actual measurements have high accuracy that supports the truthfulness of the hypothesis.
- 3. Theoretically, the time complexity of dynamic programming in solving the knapsack problem is  $O(n^*w)$ . Since in this experiment, n (number of items) is constant, the time complexity of dynamic programming should only be affected by w (knapsack capacity), which should be O(w). It represents a linear relationship between the time to compute an optimal solution and the knapsack capacity.

In conclusion, the accurate actual measurements, the line of best fit generated by the actual measurements, and the time complexity theory of dynamic programming all confirm that the hypothesis is correct.

#### 1.5 Data Statement

In terms of the data statements,

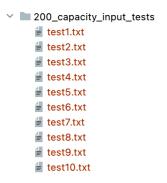
- 1. All the test files generated by kp\_generate.py that are with knapsack capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively are shown in Appendix A.1
- 2. The raw data for time taken to run 10 different sets of items with the same knapsack capacity. This is repeated 10 times, with capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively is shown in Appendix A.2.
- 3. The raw data of the average time taken to compute an optimal solution with knapsack capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively is shown in Appendix A.3.
- 4. The script of kp generate.py is shown in Appendix B.1.
- 5. The shell script of custom test.sh is shown in Appendix B.2.
- 6. The command lines to generating 10 sets of items, each set has 1000 items are shown in Appendix B.3.
- 7. The command line to run all tests files using custom\_test.sh is shown in Appendix B.4.

# **Appendix**

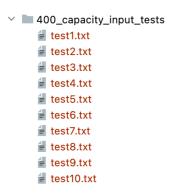
## A Raw Data for the experiment

A.1 All test files generated by kp.generate.py with knapsack capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively

10 test files of 1000 items with knapsack capacity = 200, in each test file the weight and value are randomly generated for each item



10 test files of 1000 items with knapsack capacity = 400, in each test file the weight and value are randomly generated for each item



10 test files of 1000 items with knapsack capacity = 600, in each test file the weight and value are randomly generated for each item

```
    600_capacity_input_tests
    test1.txt
    test2.txt
    test3.txt
    test4.txt
    test5.txt
    test6.txt
    test7.txt
    test8.txt
    test9.txt
    test9.txt
```

10 test files of 1000 items with knapsack capacity = 800, in each test file the weight and value are randomly generated for each item

10 test files of 1000 items with knapsack capacity = 1000, in each test file the weight and value are randomly generated for each item

```
1000_capacity_input_tests
test1.txt
test2.txt
test3.txt
test4.txt
test5.txt
test6.txt
test7.txt
test8.txt
test9.txt
test10.txt
```

10 test files of 1000 items with knapsack capacity = 1200, in each test file the weight and value are randomly generated for each item

```
1200_capacity_input_tests
test1.txt
test2.txt
test3.txt
test4.txt
test5.txt
test6.txt
test7.txt
test8.txt
test9.txt
test9.txt
```

10 test files of 1000 items with knapsack capacity = 1400, in each test file the weight and value are randomly generated for each item

```
→ 1400_capacity_input_tests

test1.txt

test2.txt

test3.txt

test4.txt

test5.txt

test6.txt

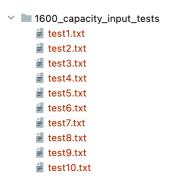
test7.txt

test8.txt

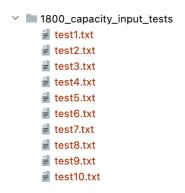
test9.txt

test10.txt
```

10 test files of 1000 items with knapsack capacity = 1600, in each test file the weight and value are randomly generated for each item



10 test files of 1000 items with knapsack capacity = 1800, in each test file the weight and value are randomly generated for each item



10 test files of 1000 items with knapsack capacity = 2000, in each test file the weight and value are randomly generated for each item

```
2000_capacity_input_tests
test1.txt
test2.txt
test3.txt
test4.txt
test5.txt
test6.txt
test7.txt
test8.txt
test9.txt
test10.txt
```

A.2 Time taken to run 10 different sets of items with the same knapsack capacity. This is repeated 10 times, with capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively.

	Time in
Capacity of knapsack	milliseconds
20	0.093
20	0.056
20	0.058
20	0.056
20	0.057
20	0.056
20	0.056
20	0.061
20	0.058
20	0.056
40	0.105
40	0.105
40	0.108
40	0.105
40	0.107
40	0.106
40	0.106
40	0.105
40	0.111
40	0.109
60	0.163
60	0.157
60	0.166
60	0.25
60	0.157
60	0.151
60	0.15
60	0.149

600	0.148	
600	0.149	
800	0.196	
800	0.195	
800	0.199	
800	0.2	
800	0.197	
800	0.197	
800	0.196	
800	0.2	
800	0.201	
800	0.198	
1000	0.248	
1000	0.246	
1000	0.249	
1000	0.251	
1000	0.255	
1000	0.253	
1000	0.25	
1000	0.244	
1000	0.242	
1000	0.243	
1200	0.291	
1200	0.29	
1200	0.293	
1200	0.295	
1200	0.3	
1200	0.292	
1200	0.295	
1200	0.295	
1200	0.298	
1200	0.295	
1400	0.347	

1400	0.347	
1400	0.347	
1400	0.341	
1400	0.344	
1400	0.344	
1400	0.343	
1400	0.343	
1400	0.349	
1400	0.341	
1600	0.391	
1600	0.396	
1600	0.389	
1600	0.394	
1600	0.393	
1600	0.389	
1600	0.401	
1600	0.391	
1600	0.391	
1600	0.392	
1800	0.441	
1800	0.446	
1800	0.439	
1800	0.443	
1800	0.455	
1800	0.521	
1800	0.487	
1800	0.475	
1800	0.449	
1800	0.455	
2000	0.5	
2000	0.501	
2000	0.503	
2000	0.489	

2000	0.485	
2000	0.489	
2000	0.486	
2000	0.498	
2000	0.483	
2000	0.49	

A.3 Average time taken to compute an optimal solution with knapsack capacity set to 200, 400, 600, 800, 1000, 1200, 1400, 1600, 1800, 2000 respectively.

```
Capacity of the Knapsack Time in milliseconds
                 200.0
                                   0.0607
                 400.0
                                   0.1067
                 600.0
                                   0.1640
                 800.0
                                   0.1979
                1000.0
                                   0.2481
                1200.0
                                   0.2944
                1400.0
                                   0.3446
                1600.0
                                   0.3927
                1800.0
                                   0.4611
                2000.0
                                   0.4924
```

# **B Shell Scripts for this experiment**

## B.1 kp\_generate.py

```
import random
import sys

n = int(sys.argv[1])
c = sys.argv[2]
b = int(sys.argv[3])
name = sys.argv[4]

fileName = name
fileObj = open(fileName,'w')
fileObj.write(str(n) + "\n")

for i in range(1, n + 1):
    weight = random.randint(1, b)
    profit = random.randint(1, b)
    fileObj.write(str(i) + " " + str(profit) + " " + str(weight) + "\n")
fileObj.write(str(c))
```

### B.2 custom\_test.sh

```
function timeout() { perl -e 'alarm shift; exec @ARGV' "$@"; }
```

```
function for c {
 echo "c/$1"
function for_java {
 echo "java -cp java comp26120.$1 kp"
function for python {
 echo "python3 python/$1 kp.py"
function get {
 if [[ $1 == "c" ]]; then for c $2; fi
  if [[ $1 == "java" ]]; then for_java $2; fi
  if [[ $1 == "python" ]]; then for python $2; fi
}
lang="c"
if [ $# -eq 1 ]
then
  lang=$1
fi
if [[ "$lang" != "c" && "$lang" != "java" && "$lang" != "python" ]]
 echo "Supply either c, java, or python as the language"
 exit
fi
algs=(dp)
inputs=( 'test1.txt' 'test2.txt' 'test3.txt' 'test4.txt' 'test5.txt'
'test6.txt' 'test7.txt' 'test8.txt' 'test9.txt' 'test10.txt')
declare -A times=(
 ['test1.txt']=20
 ['test2.txt']=20
 ['test3.txt']=20
 ['test4.txt']=20
 ['test5.txt']=20
 ['test6.txt']=20
 ['test7.txt']=20
 ['test8.txt']=20
 ['test9.txt']=20
  ['test10.txt']=20
Sizes=( 200 400 600 800 1000 1200 1400 1600 1800 2000 )
  for SIZE in ${Sizes[@]}
  do
    for FILE in ${inputs[@]}
      LIMIT="${times[$FILE]}"
      echo "Running on $FILE for $LIMIT seconds"
      echo
      echo "\begin{tabular}{||1||1||1||} \hline"
      echo "Algorithm & Optimal Value & Time Taken & Result \\\\ \hline"
        for alg in ${algs[@]}
        do
        RUN=$(get $lang $alg)
        TIME=$({ time timeout ${LIMIT}s ${RUN}}
data/${SIZE}_capacity_input_tests/$FILE > ${alg}_${FILE}_out ; } 2>&1 |
grep real | grep -o '[0-9].*')
        LAST=$(grep -o '\(Current best solution\|value\)=[0-9]*'
${alg}_${FILE}_out | tail -1)
        VALUE=\overline{\$} (echo \$LAST \mid sed -n -e 's/.*=//g' -e 'p')
```

```
# remove everything after "m" (including "m")
       minutes=${TIME%m*}
       seconds=${TIME#*m}
       seconds=${seconds%*s} # remove everything after "s" (including
"s")
       minutes=$(($minutes*60))
       total seconds=$(echo "$minutes + $seconds" | bc)
       printf -v alg %-10.10s $alg
       printf -v VALUE %-25.25s "$VALUE"
       echo "$alg & $VALUE & ${TIME} & $CORRECT \\\\"
       echo $SIZE, $total seconds>> data/data python.csv
 done
 done
 done
 echo "\hline \end{tabular}"
 echo
 echo
```

## B.3 Generating 10 sets of items, each set has 1000 items

Command Line

```
python3 kp_generate.py 1000 200 100 test1.txt
python3 kp_generate.py 1000 200 100 test2.txt
python3 kp_generate.py 1000 200 100 test3.txt
python3 kp_generate.py 1000 200 100 test4.txt
python3 kp_generate.py 1000 200 100 test5.txt
python3 kp_generate.py 1000 200 100 test6.txt
python3 kp_generate.py 1000 200 100 test6.txt
python3 kp_generate.py 1000 200 100 test7.txt
python3 kp_generate.py 1000 200 100 test8.txt
python3 kp_generate.py 1000 200 100 test9.txt
python3 kp_generate.py 1000 200 100 test9.txt
```

### B.4 Run all tests files using custom test.sh

Command Line

bash custom test.sh python