

COMP10001 Foundations of Computing Algorithms

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Lecture Outline

① Algorithm Fundamentals

② Algorithm Families

Example: Searching

- Search
 - looking for (the first instance of) particular value in a “collection”
 - specification:

```
def search(value, Numbers):
    # Inputs: Numbers = list of numbers
    #         value = number
    # Output: position of value in Numbers
    # or None if value is not in Numbers
```

- Examples:

```
>>> search(4, [3,1,4,2,5])
2
>>> search(7, [3,1,4,2,5])
None
```

Lecture Agenda

- Last lecture:
 - The Internet and HTML
- This lecture:
 - Properties and families of algorithms

What is an Algorithm?

- Definition: An algorithm is a set of steps for solving an instance of a particular problem type
- Computational desiderata of algorithms:
 - **Correctness**
 - an algorithm should terminate for every input with the correct output
 - incorrect algorithms can either: (a) terminate with the wrong output; or (b) not terminate
 - **Efficiency**
 - *runtime*: run as fast as possible
 - *storage*: require as little storage as possible

Linear Search: Algorithm

- Algorithm idea:
 - ① Initialise the index to the first element of the list
 - ② While the index points to a list element:
 - (a) If the value at the current list index is equal to the required value, terminate and return the index
 - (b) Else increment the index
 - ③ If the index has run off the end of the list, return None

Algorithmic Analysis: Linear Search

- Is it correct? How do we know?
- Is it run-time efficient? How efficient is it (best case vs. worst case vs. average)?
- Is it storage efficient? How efficient is it (best case vs. worst case vs. average)?

Algorithmic Analysis: Binary Search

- Is it correct? How do we know?
- Is it run-time efficient? How efficient is it (best case vs. worst case vs. average)?
- Is it storage efficient? How efficient is it (best case vs. worst case vs. average)?

Exact vs. Approximate Methods

- Exact approach: calculate the solution (set), with a guarantee of correctness, e.g.:
 - brute force
 - divide and conquer
- Approximate approaches: estimate the solution (set), ideally with an additional estimate of how “close” this is to the exact solution (set), e.g.
 - simulation
 - heuristic search
- Always use exact approaches where possible

Binary Search: Algorithm

Initialise a sub-list to the full list, and our index to the mid-point of the list

While the sub-list is non-empty:

- If the value at the current list index is smaller than the one wanted, continue to search over the right half of the current sub-list
- Else if the value at the current list index is larger than the one wanted, continue to search over the left half of the current sub-list
- Else if the value at the current list index is equal to the required value, return the current index

Lecture Outline

① Algorithm Fundamentals

② Algorithm Families

Brute-Force (aka “Generate and Test”)

- Assumptions
 - A candidate answer is easy to test
 - The set of candidate answers is ordered or can be generated exhaustively
- Strategy
 - Generate candidate answers and test them one by one until a solution is found
- Examples:
 - Linear search
 - Test whether a number is prime
 - Our solution for Coins and Denominations using loops

Coins: A Bruteforce Approach

```

1 '''Count the number of combinations of
2   (1,2,5,10,20) that sum to N
3   '''
4 answer = 0
5 for a in range(N+1):
6     for b in range(N//2+1):
7         for c in range(N//5+1):
8             for d in range(N//10+1):
9                 for e in range(N//20+1):
10                     if a+2*b+5*c+10*d+20*e == N:
11                         answer += 1

```

Binary Search: Divide and Conquer

```

def bsearch(val, nlist):
    return bs_rec(val, nlist, 0, len(nlist)-1)

def bs_rec(val, nlist, start, end):
    if start > end:
        return None
    mid = start+(end-start)//2
    if nlist[mid] == val:
        return mid
    elif nlist[mid] < val:
        return bs_rec(val, nlist, mid+1, end)
    else:
        return bs_rec(val, nlist, start, mid-1)

```

Divide & Conquer and Memoisation

```

fib = {}

def fib_fast(n):
    global fib
    if n < 2: return 1
    else:
        if n-1 not in fib:
            fib[n-1] = fib_fast(n-1)
        if n-2 not in fib:
            fib[n-2] = fib_fast(n-2)
        return fib[n-1] + fib[n-2]

```

Divide and Conquer

- Strategy:
 - Solve a smaller sub-problem
 - Extend the sub-solution to create the solution of the original problem
 - (Sounds like recursion, but can also be iterative.)
- Examples:
 - Binary search
 - Many sorting algorithms

Divide & Conquer and Memoisation

- Naive implementations of divide and conquer can lead to many repeated, identical function calls
- For example, the calculation of Fibonacci numbers

```

def fib(n):
    if n < 2:
        return 1
    else:
        return fib(n-1) + fib(n-2)

```

- F(9) calls F(8) & F(7); F(8) calls F(7) & F(6); ...
- “Memoisation” (i.e. storing the value for each element used, to avoid recalculating it), can lead to more efficient algorithms.

More Divide & Conquer

- A more interesting example of divide & conquer:

Given a list of integers, calculate the maximum sum of a contiguous sublist of elements in the list

```

>>> lst = [5, 3, -1]
>>> maxsubsum(lst)
8

```

- The brute-force solution simply calculates the sum of each (non-empty) sublist, and calculates the maximum among them

More Divide & Conquer

```
def sublist_sum_bf(lst):
    """Brute force."""
    max_so_far = (lst[0], 0, 0)

    for s in range(len(lst)):
        for e in range(s, len(lst)):
            subsum = sum(lst[s:e+1])
            if subsum > max_so_far[0]:
                max_so_far = (subsum, s, e)
    return(max_so_far)

print(sublist_sum_bf([3,-1,-2,2]))
```

More Divide & Conquer

```
def dc_maxsubsum(lst):
    assert len(lst) > 0
    max_sum_i = [lst[0]]
    # base case

    for i in range(1, len(lst)):
        c1 = lst[i]
        # start new

        c2 = max_sum_i[-1] + lst[i]
        # or extend

        max_sum_i.append(max(c1, c2))
        # now take the max of all these

    return max(max_sum_i)
```

Simulation

- Strategy:
 - Randomly generate a large amount of data to predict an overall trend
 - Use multiple runs to verify the stability of an answer
 - Used in applications where it is possible to describe individual properties of a system, but hard/impossible to capture the interactions between them
- Applications:
 - Weather forecasting
 - Movement of planets
 - Prediction of share markets

More Divide & Conquer

- The divide-and-conquer approach work as follows:
 - Assume $\text{maxsubsum}(i-1)$ is the maximum sum for the sublist ending at $i-1$ (inclusive)
 - The maximum sum for the sublist $\text{lst}[:i+1]$ is $\max(\text{lst}[i], \text{lst}[i] + \text{maxsubsum}(i-1))$
- The recursive version will have issues with the limit on recursion depth, so implement iteratively

More Divide & Conquer

- How run-time efficient are the respective implementations (brute-force vs. divide-and-conquer)? (best case vs. worst case vs. average)?
- How storage efficient are the respective implementations (brute-force vs. divide-and-conquer)? (best case vs. worst case vs. average)?

Simulation: A Game of Chance

- Gambling game:
 - You bet \$1, and roll two dice
 - If the total is between 8 and 11, you win \$2
 - If the total is 12, you win \$6
 - Otherwise, you lose
- Is it worth playing?
 - Start with a \$5 float and play to \$0 or \$20
 - How many games do you win on average?

Monte Carlo Simulation

- Method:
 - iteratively test a model using random numbers as inputs
 - problem is complex and/or involves uncertain parameters
 - a simulation typically has at least 10,000 evaluations
 - approximate solution to problem that is not (readily) analytically solvable
- Game of chance:
 - should a casino offer this game?

Lecture Summary

- What is an algorithm?
- What computational desiderata are associated with algorithms?
- What are “exact” and “approximate” methods? What are common examples of each?
- What are each of: brute-force, divide and conquer, simulation and heuristic search?

Heuristic Search

- Strategy:
 - Search via a cheap, approximate solution which works *reasonably* well *most* of the time ... but where there is no proof of how close to optimal the proposed solution is
- Examples:
 - For finding a closest neighbor many Location-based Services use Euclidean distances as they are easy to compute
 - There is no guarantee they are equal to road-network distances though
 - So whatever you find, is just “possibly” a good solution