

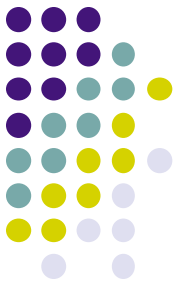
# COMP20003

## Algorithms and Data Structures Algorithms

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# Outline of the first few lectures

- Algorithms: general
- This subject: details
- ➔ ● Algorithm efficiency
- Computational complexity
- Data structures
  - Basic data structures
  - Algorithms on basic data structures
  - Complexity analysis of algorithms on basic ds's



# Revisit: What is an algorithm?

- A set of steps to accomplish a task.
- Computer algorithms must be:
  - Specific.
  - Correct.
  - Reasonably efficient.

# A small diversion: Which C standard?



- C standards:
  - ANSI C (C89)
  - C99 – “substantially” completely supported in gcc 4.5 (with `-std=C99` option on)
  - C11 (current C standard, from 2011) **gcc 4.8**
- On **nutmeg.eng.unimelb.edu.au**:
  - **gcc -v:**
  - **gcc version 4.4.7**



# C for gcc on nutmeg

- ANSI C with *some* of the features of C99, *e.g.*
  - Supported:
    - inline functions
    - long long int
  - Not supported:
    - Variable length arrays
    - Doesn't insist on explicit return type for function
- For all the new features in C99 see:

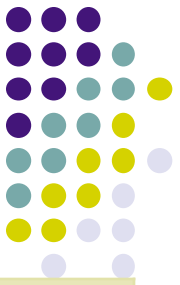
<http://www.open-std.org/jtc1/sc22/wg14/www/newinc9x.htm>



# Algorithms and Efficiency

- An algorithm must:
  - be accurate (to within the required tolerance).
  - compute in a “reasonable” amount of time.
- The most accurate algorithm in the world is no use if it takes forever to compute.
- We are particularly interested in efficiency as the size of the input grows.
- Why?

# Example algorithm: Compute Fibonacci numbers



- $F_n = F_{n-1} + F_{n-2}$ 
  - $F_0 = 0$
  - $F_1 = 1$

n	F(n-1)	F(n-2)	F(n)
0	-	-	0
1	-	-	1
2	1	0	1
3	1	1	2
4	2	1	3
5	3	2	5
6			8
7	8	5	13
8	13	8	21

- 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55....

# Fibonacci numbers

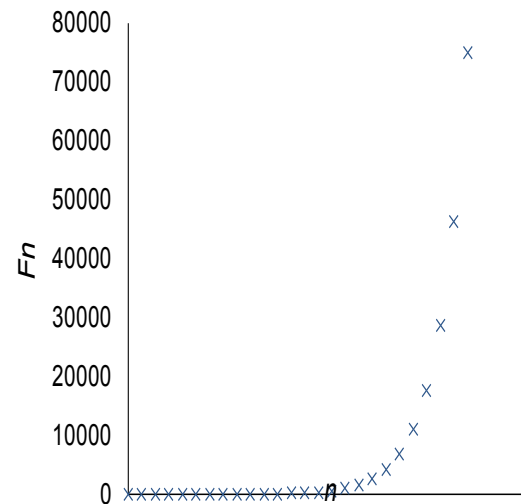


- Fibonacci numbers grow very quickly:

- $F_{10} = 55$

- $F_{15} = 610$

- $F_{20} = 6765$



Painting by Kobi, 19<sup>th</sup> c, public domain



# Fibonacci numbers



- The original problem that Fibonacci was investigating (1202):
  - How fast can rabbits breed under ideal circumstances?
  - <http://www.maths.surrey.ac.uk/hosted-sites/R.Knott/Fibonacci/fibnat.html#Rabbits>



Painting by Kobi, 19<sup>th</sup> c, public domain

# How to compute Fibonacci numbers?



- Given the definition of Fibonacci numbers:
  - $F_n = F_{n-1} + F_{n-2}$ 
    - $F_0 = 0$
    - $F_1 = 1$
- Does this suggest an easy way to calculate  $F_n$ ?

# Computing Fibonacci Numbers: Scaffolding



```
main()
{
    int n, ans;

    printf("Enter a number:\n");
    scanf("%d", &n);

    if(DEBUG)
        printf("%d\n", n)
    ans = fib(n);
    printf("Fibonacci of %d is %d\n", n, ans);
}
```



# Naïve Fibonacci algorithm

```
int  
fib (int n)  
{  
    if (n==0) return 0;  
    if (n==1) return 1;  
    return fib(n-1) + fib(n-2);  
}
```

Definition:

$$F_0 = 0$$

$$F_1 = 1$$

$$F_n = F_{n-1} + F_{n-2}$$

- Is the algorithm correct?



# Naïve Fibonacci algorithm

```
int
fib (int n)
{
    if (n==0) return 0;
    if (n==1) return 1;
    return fib(n-1) + fib(n-2);
}
```

- How long does it take to compute?



# Fibonacci computation

- Approach to estimating computation effort:
  - Count operations.
  - Count operations as a function of input size.
  - Count operations as a proxy for time.
- $T(n) =_{def}$  run time for input  $n$ .
- $T(n)$  = calculate as *number of operations* for input size  $n$ .
  - Portable between machines.
  - Can compare algorithms.



# Fibonacci computation

- Looking at  $T(n)$  as number of operations to calculate the  $n$ th Fibonacci number, then
  - $T(n) = T(n-1) + T(n-2) + 3$  (*operations*)
  - $T(1) = T(0) = 1$
- Example: unrolling the loop
  - $T(4) = T(3) + T(2) + 3$



# Fibonacci computation

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  - $T(4) = T(3) + T(2) + 3$
  - $= T(2) + T(1) + 3 + T(2) + 3$





# Fibonacci computation

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# Fibonacci computation

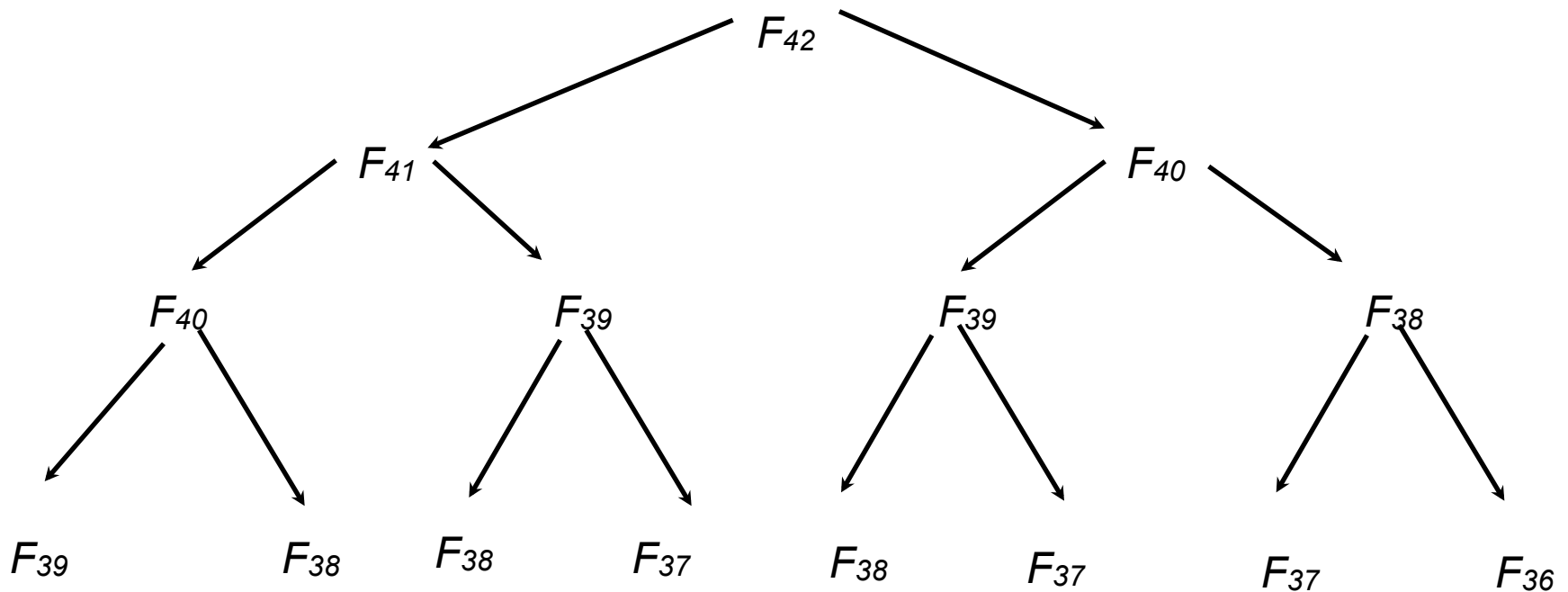
- Looking at  $T(n)$  as number of operations to calculate the  $n$ th Fibonacci number, then
  - $T(n) = T(n-1) + T(n-2) + 3$  (operations)
  - $T(1) = T(0) = 1$
- Example: unrolling the loop
  - $T(4) = T(3) + T(2) + 3$
  - $= T(2) + T(1) + 3 + T(2) + 3$
  - $= T(1) + T(0) + 3 + T(1) + 3 + T(1) + T(0) + 3$   
+ 3
  - $= 1+1+3+1+3+1+1+3+3 = 17$  operations



# Fibonacci computation

- Looking at  $T(n)$  as number of operations to calculate the  $n$ th Fibonacci number, then
  - $T(n) = T(n-1) + T(n-2) + 3$  (*operations*)
  - $T(1) = T(0) = 1$
- Example:
  - $T(5) = T(4) + T(3)$
  - $= 17 + T(2) + T(1) + 3$
  - $= 17 + 9$
  - $= 26$

# How many operations to calculate $F_{42}$ ?



Any obvious inefficiencies?



# Memoization

- Store previously computed values.

```
fib(n)
{
    int i;
    int fib[n+1];

    fib[0] = 0;
    fib[1] = 1;

    for(i = 2; i <= n; i++)
    {
        f[i] = f[i-1] + f[i-2];
    }
    return (fib[n]);
}/* how many operations to calculate fib(n)? */
```

# ...or without storing all the intermediate results



```
/* this one will work in C */
int fib(n)
{
    int result = 0;
    int preOldResult = 1; int oldResult = 1;

    if(n <= 0) return 0;
    if(n > 0 && n < 3) return 1;
    for (int i=3;i<n;i++)
    {
        result = preOldResult + oldResult;
        preOldResult = oldResult;
        oldResult = result;
    }
    return result;
}
```



# Counting operations

- Count of operations used as proxy for run time:
  - Advantages?
  - Caveats?
- How long does calculation of `fib(2)` take
  - Using the naïve algorithm?
  - Using memoization?
- Do we care how long things take for small input  $n$ ?

# Complexity analysis: general method



- Count operations for  $T(n)$  – “time” (number of ops) taken for input  $n$ .
- Note: best to identify the most expensive operation and count that operation.
  - e.g. count multiplications, ignore additions
- Note also that we can sometimes trade off space for time.



# Complexity analysis: a fine point for fib()



- Assumption: addition of two numbers takes constant time.
- True *if* both numbers can fit into one computer word: 32 bits, number  $< 2^{32}$ .
- But Fibonacci numbers get very large:
  - $F_n$  takes approximately  $0.694n$  bits, so
  - To fit in one word,  $n < 32/0.694 = 46$
  - $F_{50} = 12,586,269,025 > 12 \cdot 10^9 > 2^{33}$
  - Last operations take longer. Assumption not valid for large  $n$ .

# Closed form for Fibonacci numbers



- Binet's formula:

$$F_n = \frac{1}{\sqrt{5}} \left( \left( \frac{1 + \sqrt{5}}{2} \right)^n - \left( \frac{1 - \sqrt{5}}{2} \right)^n \right)$$

- For large  $n$ ,  $F_n \approx 2^{0.694n}$
- For more on Fibonacci numbers, see :  
<http://mathworld.wolfram.com/FibonacciNumber.html>



# Complexity analysis: Intuition

- Fortunately, most algorithms do not deal with such large numbers.
- Counting operations usually suffices.
- Always be aware of the assumptions:
  - What is the most expensive operation?
  - Are the operations really constant?
  - What are the inputs and outputs?

# Skiena: Algorithm Design Manual



- Chapter 1:
  - Algorithm correctness.
  - Example problems.
- Chapter 2:
  - Chapter 2.1: counting operations



## Next section

- Complexity analysis more formally.
- Big-O and related formalisms.