

Recursion and Dynamic Programming

Biostatistics 615/815

Lecture 5

Last Lecture

- Principles for analysis of algorithms
 - Empirical Analysis
 - Theoretical Analysis
- Common relationships between inputs and running time
- Described two simple search algorithms

Recursive refers to ...

- A function that is part of its own definition

e.g. $Factorial(N) = \begin{cases} N \cdot Factorial(N-1) & \text{if } N > 0 \\ 1 & \text{if } N = 0 \end{cases}$

- A program that calls itself

Key Applications of Recursion

- Dynamic Programming
 - Related to Markov processes in Statistics
- Divide-and-Conquer Algorithms
- Tree Processing

Recursive Function in C

```
int factorial (int N)
{
    if (N == 0)
        return 1;
    else
        return N * factorial(N - 1);
}
```

Key Features of Recursions

- Simple solution for a few cases
- Recursive definition for other values
 - Computation of large N depends on smaller N
- Can be naturally expressed in a function that calls itself
 - Loops are sometimes an alternative

A Typical Recursion: Euclid's Algorithm

- Algorithm for finding greatest common divisor of two integers a and b
 - If a divides b
 - $\text{GCD}(a,b)$ is a
 - Otherwise, find the largest integer t such that
 - $at + r = b$
 - $\text{GCD}(a,b) = \text{GCD}(r,a)$

Euclid's Algorithm in C

```
int gcd (int a, int b)
{
    if (a == 0)
        return b;

    return gcd(b % a, a);
}
```


Evaluating $\text{GCD}(4458, 2099)$

$\text{gcd}(2099, 4458)$

$\text{gcd}(350, 2099)$

$\text{gcd}(349, 350)$

$\text{gcd}(1, 349)$

$\text{gcd}(0, 1)$

Divide-And-Conquer Algorithms

- Common class of recursive functions
- Common feature
 - Process input
 - Divide input in smaller portions
 - Recursive call(s) process at least one portion
- Recursion may sometimes occur before input is processed

Recursive Binary Search

```
int search(int a[], int value, int start, int stop)
{
    // Search failed
    if (start > stop)
        return -1;

    // Find midpoint
    int mid = (start + stop) / 2;

    // Compare midpoint to value
    if (value == a[mid]) return mid;

    // Reduce input in half!!!
    if (value < a[mid])
        return search(a, start, mid - 1);
    else
        return search(a, mid + 1, stop);
}
```

Recursive Maximum

```
int Maximum(int a[], int start, int stop)
{
    int left, right;

    // Maximum of one element
    if (start == stop)
        return a[start];

    left = Maximum(a, start, (start + stop) / 2);
    right = Maximum(a, (start + stop) / 2 + 1, stop);

    // Reduce input in half!!!
    if (left > right)
        return left;
    else
        return right;
}
```

An inefficient recursion

- Consider the Fibonacci numbers...

$$Fibonacci(N) = \begin{cases} 0 & \text{if } N = 0 \\ 1 & \text{if } N = 1 \\ Fibonacci(N-1) + Fibonacci(N-2) & \end{cases}$$

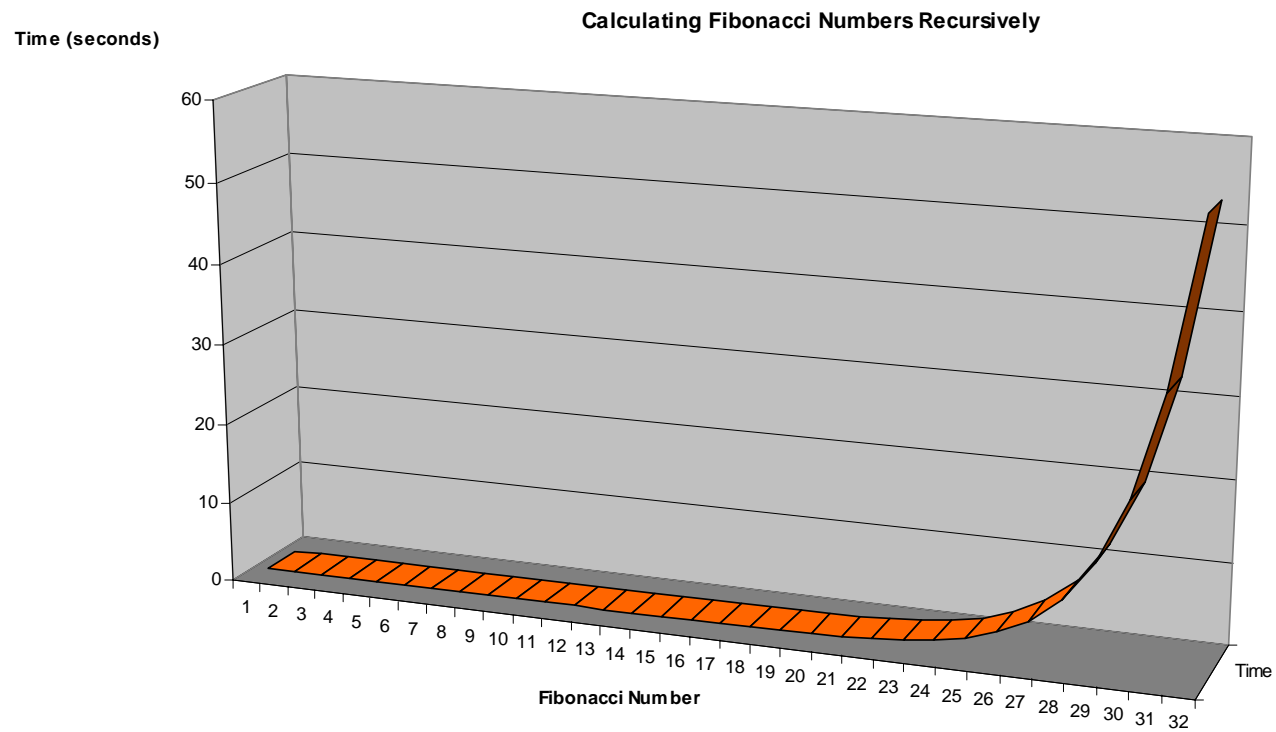
Fibonacci Numbers

```
int Fibonacci(int i)
{
    // Simple cases first
    if (i == 0)
        return 0;

    if (i == 1)
        return 1;

    return Fibonacci(i - 1) + Fibonacci(i - 2);
}
```

Terribly Slow!



Faster Alternatives

- Certain quantities are recalculated
 - Far too many times!
- Need to avoid recalculation...
 - Ideally, calculate each unique quantity once.

Dynamic Programming

- A technique for avoiding recomputation
- Can make exponential running times ...
- ... become linear!

Bottom-Up Dynamic Programming

- Evaluate function starting with smallest possible argument value
 - Stepping through possible values, gradually increase argument value
- Store all computed values in an array
- As larger arguments evaluated, precomputed values for smaller arguments can be retrieved

Fibonacci Numbers in C

```
int Fibonacci(int i)
{
    int fib[LARGE_NUMBER], j;

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

    for (j = 2; j <= i; j++)
        fib[j] = fib[j - 1] + fib[j - 2];

    return fib[i];
}
```

Fibonacci With Dynamic Memory

```
int Fibonacci(int i)
{
    int * fib, j, result;

    if (i < 2) return i;

    fib = malloc(sizeof(int) * (i + 1));

    fib[0] = 0; fib[1] = 1;
    for (j = 2; j <= i; j++)
        fib[j] = fib[j - 1] + fib[j - 2];

    result = fib[i];
    free(fib);

    return result;
}
```

Top-Down Dynamic Programming

- Save each computed value as final action of recursive function
- Check if pre-computed value exists as the first action

Fibonacci Numbers

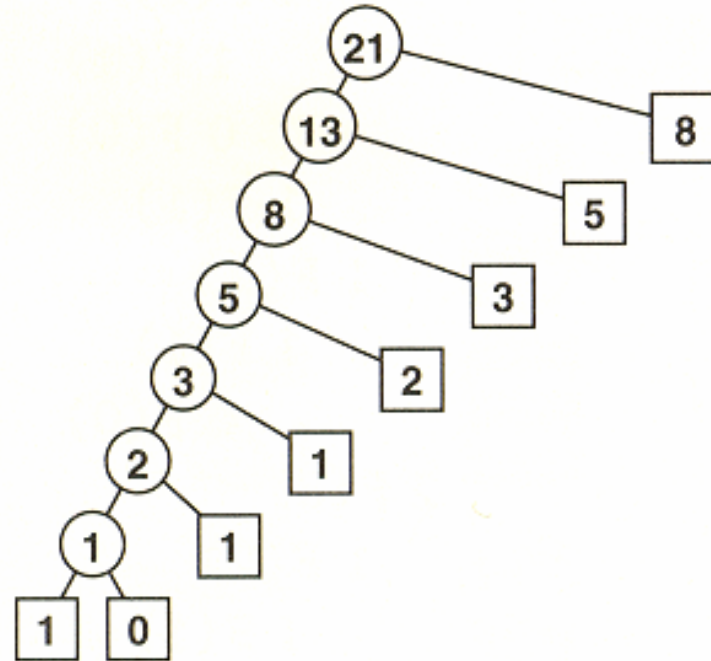
// Note: saveF should be a global array initialized all zeros

```
int Fibonacci(int i)
{
    // Simple cases first
    if (saveF[i] > 0)
        return saveF[i];

    if (i <= 1)
        return i;

    // Recursion
    saveF[i] = Fibonacci(i - 1) + Fibonacci(i - 2);
    return saveF[i];
}
```

Much less recursion now...



Dynamic Programming

Top-down vs. Bottom-up

- In bottom-up programming, programmer has to do the thinking by selecting values to calculate and order of calculation
- In top-down programming, recursive structure of original code is preserved, but unnecessary recalculation is avoided.

Examples of Useful Settings for Dynamic Programming

- Calculating Binomial Coefficients
- Evaluating Poisson-Binomial Distribution

Binomial Coefficients

- The number of subsets with k elements from a set of size N

$$\binom{N}{k} = \binom{N-1}{k} + \binom{N-1}{k-1}$$

$$\binom{N}{0} = \binom{N}{N} = 1$$

Bottom-Up Implementation in C

```
int Choose(int N, int k)
{
    int i, j, M[MAX_N][MAX_N];

    for (i = 1; i <= N; i++)
    {
        M[i][0] = M[i][i] = 1;

        for (j = 1; j < i; j++)
            M[i][j] = M[i - 1][j - 1] + M[i - 1][j];
    }

    return M[N][k];
}
```

Top-Down Implementation (I)

```
#define MAX_N 30
```

```
int choices[MAX_N][MAX_N];
```

```
void InitChoose()
```

```
{
```

```
    int i, j;
```

```
    for (i = 0; i < MAX_N; i++)
```

```
        for (j = 0; j < MAX_N; j++)
```

```
            choices[i][j] = 0;
```

```
}
```

Top-Down Implementation (II)

```
int Choose(int N, int k)
{
    // Check if this is an easy case
    if (N == k || k == 0)
        return 1;

    // Or a previously examined case
    if (choices[N][k] > 0)
        return choices[N][k];

    // If neither of the above helps, use recursion
    choices[N][k] = Choose(N - 1, k - 1) + Choose(N - 1, k);

    return choices[N][k];
}
```

Poisson-Binomial Distribution

- X_1, X_2, \dots, X_n are Bernoulli random variables
- Probability of success is p_k for X_k
- $\sum_k X_k$ has Poisson-Binomial Distribution

Recursive Formulation

$$P_1(0) = 1 - p_1$$

$$P_1(1) = p_1$$

$$P_j(0) = (1 - p_j)P_{j-1}(0)$$

$$P_j(j) = p_j P_{j-1}(j-1)$$

$$P_j(i) = p_j P_{j-1}(i-1) + (1 - p_j)P_{j-1}(i)$$

Summary

- Recursive functions
 - Arise very often in statistics
- Dynamic programming
 - Bottom-up Dynamic Programming
 - Top-down Dynamic Programming
- Dynamic program is an essential tool for statistical programming

Good Programming Practices

- Today's examples used global variables
 - Variables declared outside a function
 - Accessible throughout the program
- In general, these should be used sparingly
- Two alternatives are:
 - Using **static** variables that are setup on the first call
 - Using C++ to group a function and its data

Function with Built-In Initialization

```
int Choose(int N, int k)
{
    static int valid = 0, choices[MAX_N][MAX_N], i, j;

    // Check if we need to initialize data
    if (valid == 0)
    {
        for (i = 0; i < MAX_N; i++)
            for (j = 0; j < MAX_N; j++)
                choices[i][j] = 0;
        valid = 1;
    }

    // Check if this is an easy case
    if (N == k || k == 0) return 1;

    // Or a previously examined case
    if (choices[N][k] > 0) return choices[N][k];

    // If neither of the above helps, use recursion
    choices[N][k] = Choose(N - 1, k - 1) + Choose(N - 1, k);

    return choices[N][k];
}
```

C++ Declarations (for .h file)

```
class Choose
{
public:
    // This function is called to initialize data
    // Whenever a variable of type Choose is created
    Choose();

    // This function is called to calculate N_choose_k
    int Evaluate(int N, int k);

private:
    int choices[MAX_N][MAX_N];
};
```

C++ Code (for .cpp file)

```
Choose::Choose()
{
    for (int i = 0; i < MAX_N; i++)
        for (int j = 0; j < MAX_N; j++)
            choices[i][j] = 0;
}

int Choose::Evaluate()
{
    // Check if this is an easy case
    if (N == k || k == 0) return 1;

    // Or a previously examined case
    if (choices[N][k] > 0) return choices[N][k];

    // If neither of the above helps, use recursion
    choices[N][k] = Choose(N - 1, k - 1) + Choose(N - 1, k);

    return choices[N][k];
}
```

Using C++ Code

```
#include "Choose.h"
```

```
int main()
```

```
{
```

```
    Choose choices;
```

```
    // Evaluate 20_choose_10
```

```
    choices.Evaluate(20, 10);
```

```
    // Evaluate 30_choose_10
```

```
    choices.Evaluate(30, 10);
```

```
    return 0;
```

```
}
```

Reading

- Sedgewick, Chapters 5.1 – 5.3