Dijkstra -> stacks and recursion NOTE: Introduction to Algorithms has a more generalized and detailed overview of techniques

Recursion and Trees

- · a function that calls itself
- · requires a termination condition
- · recursion and trees are tightly connected
- mathematical recurrences and recursion in programs are connected as well
- · additional items:
 - divide and conquer
 - dynamic programming
 - tree traversal
 - · depth-first (and breadth-first) search

Recursive Algorithms

· solves a problem by solving one or more smaller instances of the problem

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

- (the mathematical concept/definition of) recurrence relations are recursively defined functions
- any recursively defined function can be expressed using loops and any computation using loops can be defined recursively
- · Euclid's algorithm

```
int gcd(int m, int n) {
   if (n == 0) return m;
   return gcd(n, m % n);
}
```

evaluate prefix expressions (example of a recursive descent parser)

```
char *a; int i;
int eval() {
   int x = 0;
   while (a[i] == ' ') {
```

```
i++;
}
if (a[i] == '+') {
    i++;
    return eval() + eval();
}
if (a[i] == '*') {
    i++;
    return eval() * eval();
}
while ((a[i] >= '0') && (a[i] <= '9')) {
    x = 10 * x + (a[i++] - '0');
}
return x;
}</pre>
```

- · the definition of linked data structures is inherently recursive
- · linked list recursive functions:

```
int count(link x) {
    if (x == 0) {
        return 0;
    }
    return 1 + count(x->next);
}
void traverse(link h, void visit(link)) {
    if (h == 0) {
        return;
    }
    visit(h);
    traverse(h->next, visit);
}
void traverseR(link h, void visit(link)) {
    if (h == 0) {
        return;
    traverseR(h->next, visit);
    visit(h);
void remove(link& x, Item v) {
    while (x != 0 \&\& x->item == v) {
        link t = x;
        x = x->next;
        delete t;
    }
    if (x != 0) {
        remove(x->next, v);
    }
}
```

Divide and Conquer

- · uses two recursive calls, each split to roughly half of the input
- · divide and conquer to find the max

```
Item max(Item a[], int l, int r) {
    if (l == r) {
        return a[l];
    }
    int m = (l + r) / 2;
    Item u = max(a, l, m);
    Item v = max(a, m+1, r);
    if (u > v) {
        return u;
    } else {
        return v;
    }
}
```

· Towers of Hanoi:

```
void hanoi(int N, int d) {
    if (N == 0) {
        return;
    }
    hanoi(N-1, -d);
    shift(N, d);
    hanoi(N-1, -d);
}
```

draw a ruler (recursive divide and conquer):

```
void rule(int l, int r, int h) {
   int m = (l+r)/2;
   if (h > 0) {
      rule(l, m, h-1);
      mark(m, h);
      rule(m, r, h-1);
   }
}
```

draw a ruler (non-recursive):

```
void rule(int l, int r, int h) {
   for (int t = 1, j = 1; t <= h; j += j, t++) {
     for (int i = 0; l+j+i <= r; i += j+j) {
        mark(l+j+i, t);
   }</pre>
```

}

- the ruler drawing problem can be used to demonstrate a number of relationships
 - n bit numbers
 - effects of preorder vs postorder and bottom-up order
 - · fractals
 - combine and conquer
- · properties:
 - a recursive function that divides a problem of size n into two independent non-empty parts that it solves recursively calls itself less than n times
 - recursive Towers of Hanoi produces a solution with 2 ^ n 1 moves
- must be aware of the stack during recursion -> both the size and the elements stored in each frame
- additional related topics:
 - · divide into parts of varying size
 - divide into n parts
 - divide into overlapping parts
 - do varying amounts of work in the non-recursive part of the algorithm

Dynamic Programming

- divide and conquer's sub-problems are independent
- when the sub-problems are not independent the solutions become more complicated because direct recursive approaches take excessive time
- bottom-up dynamic programming:
 - mathematical recurrences work with integers recursively
 - these functions can be solved by computing all of the function values in order starting with the smallest and computing the value at each step using values from previous steps
 - (possibly uses an array to store previous values and access them)
- top-down dynamic programming:
 - simpler corollary to bottom-up dynamic programming (at same cost or less cost)
 - the recursive program saves each computed value as its last action and checks for the computed value to avoid re-computation as its first action
 - the running time is reduced to linear running time
 - · sometimes called memoization
- dynamic Fibonacci:

```
int F(int i) {
    static int knownF[maxN];
    if (knownF[i] != 0) {
        return knownF[i];
    }
    int t = i;
    if (i < 0) {
        return 0;
    }
    if (i > 1) {
        t = F(i-1) + F(i-2);
    }
    return knownF[i] = t;
}
```

· recursive knapsack:

```
int knap(int cap) {
    int i, space, max, t;
    for (i = 0, max = 0; i < N; i++) {
        if ((space = cap-items[i].size) >= 0) {
            if ((t = knap(space) + items[i].val) > max) {
                max = t;
            }
        }
    }
    return max;
}
```

· non-recursive knapsack:

```
int knap(int M) {
    int i, space, max, maxi = 0, t;
    if (maxKnown[M] != unknown) return maxKnown[M];
    for (i = 0, max = 0; i < N; i++)
        if ((space = M-items[i].size) >= 0)
            if ((t = knap(space) + items[i].val) > max) {
                max = t;
                 maxi = i;
            }
        }
    }
    maxKnown[M] = max; itemKnown[M] = items[maxi];
    return max;
}
```

- properties:
 - dynamic programming reduces the running time of a recursive function to be at most the time required to evaluate the function for all arguments less than or equal to the given argument, treating a recursive call as constant

- top-down vs bottom-up
 - top-down -> save known values
 - bottom-up -> precompute known values
 - prefer top-down because:
 - · it is a mechanical transformation of a natural problem solution
 - · the ordering of sub-problem computation takes care of itself
 - · may not need to compute all of the answers to all of the sub-problems

Additional Notes

- · See Introduction to Algorithms for a more detailed discussion
- NOTE: dynamic programming is characterized by:
 - · the overlap between sub-problems
 - optimal sub-structure (an optimal solution can be constructed efficiently from optimal solutions of its sub-problems)
 - (non-overlapping sub-problems leads to divide and conquer solutions)
- · top-down:
 - solve the top of the input tree first in terms of the sub-problems recursively and store solutions to sub-problems in a table (memoization)

```
return_type memoization(p, n) {
    let r[0...n] be a new array
    for (i = 0 \text{ to } n) {
        r[i] = some invalid value;
    return memoize_aux(p, n, r);
}
return_type memoize_aux(p, n, r) {
    if (r[n] is defined (i.e. greater than 0 or a valid value)) {
        return r[n]:
    } else {
        // this is just generic to signal the action of computing smaller values
        // and storing them/returning them
        q = -infinity; // (i.e. invalid value)
        for (i = 1 \text{ to } n) {
            q = recursively calculate q
        r[n] = q;
        return q;
    }
```

- bottom-up:
 - start by solving sub-problems first and then solve larger and larger problems (using a tabular form (tabulation))

```
return_type bottom_up(p, n) {
    let r[0...n] be a new array
    r[0] = 0 // base case
    for (j = 1 to n) {
        q = -infinity // (i.e. invalid value)
        for (i = 1 to j) { // working from sub-problems upward
            calculate q
        }
        r[j] = q;
    }
    return r[n]
}
```

Additional Notes

factorial

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

possible non-recursive implementation

```
unsigned int factorial(unsigned int n) {
   int t;
   int i;
   for (t = 1, i = 1; i <= n; i++) {
        t *= i;
   }
   return t;
}</pre>
```

better

```
unsigned int factorial(unsigned int n) {
   int a = 1;
   for (int i = n; i > 1; i--) {
        a *= i;
   }
   return a;
}
```

power

```
double power(double x, unsigned int n) {
   if (n == 0) {
      return 1.0;
   } else {
      return x * power(x, n - 1);
   }
}
```

· non-recursive

```
double power(double x, unsigned int n) {
    double a = 1.0;
    for (int i = n; i > 1; i--) {
        a *= x;
    }
    return a;
}
```

tail recursion

just a glorified loop

```
void tail(int i) {
   if (i > 0) {
      std::cout << i << " ";
      tail(i - 1);
   }
}</pre>
```

· iterative equivalent

```
void iterativeTail(int i) {
    for (; i > 0; i--) {
        std::cout << i << " ";
    }
}</pre>
```

nontail recursion

· the call stack can store information without having to explicitly store it

```
void reverse() {
    char ch;
    std::cin.get(ch);
    if (ch != '\n') {
        reverse();
        std::cout.put(ch);
    }
}
```

· non-recursive

```
void iterativeReverse() {
    char stack[80];

int ch;
    int i = 0;
    while (ch = (std::cin.get() && ch != '\n')) {
        stack[top] = ch;
        i++;
    }
    for (; i >=0; i--) {
        std::cout.put(stack[i]);
    }
}
```

with separate stack

```
void nonRecursiveReverse() {
    int ch;
    while (ch = (std::cin.get() && ch != '\n')) {
        stack.push(ch);
    }
    while (!stack.empty()) {
        std::cout.put(stack.pop());
    }
}
```

indirect recursion

 when one function calls another function that in turn calls the first (at some point in the ensuing call stack)

```
receive() -> decode() -> store() -> receive() -> decode() -> store()
```

nested recursion

 function defined in terms of itself with the additional definition of parameters in terms of original function

backtracking

- recursively travel all possible paths to find the correct solution
- · relates to trying all possible permutations
- · generically:
 - · from starting point
 - · test for accept or reject and return if either
 - · while solution has not been found
 - · recursively try each possibility in current path
 - if no solution, return to starting point
 - · if no solution return false
- · backtracking is really just depth-first search on an implicit graph (or tree)

```
void backtrack(candidate) {
    if (reject(instance, candidate)) {
        return;
    }
    if (accept(instance, candidate)) {
            return output(instance, candidate);
    }
    element s = first(instance, candidate);
    while (s != null_case) {
            backtrack(s);
            s = next(instance, s)l
    }
}
```

· recursive

```
boolean solve(Node n) {
   if n is a leaf node {
      if the leaf is a goal node, return true
      else return false
   } else {
      for each child c of n {
        if solve(c) succeeds, return true
    }
}
```

```
return false
}
```

· non-recursive

NOTE: review recursive descent parser here

linear recursion

- · general outline:
 - if n == 0
 - · return constant
 - else return function(n 1)

binary recursion

- · general outline:
 - if n,m == base case:
 - · return constant
 - else return function(start, n, n + m / 2) + function(start, n + m / 2, m)

multiple recursion

- generalized from binary recursion
- much like dynamic program, multiple possible configurations,

general design

- · test for base case
- recur

eliminating tail recursion

• iterate or explicitly use a stack