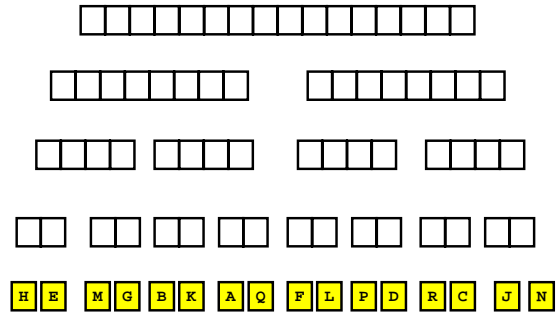


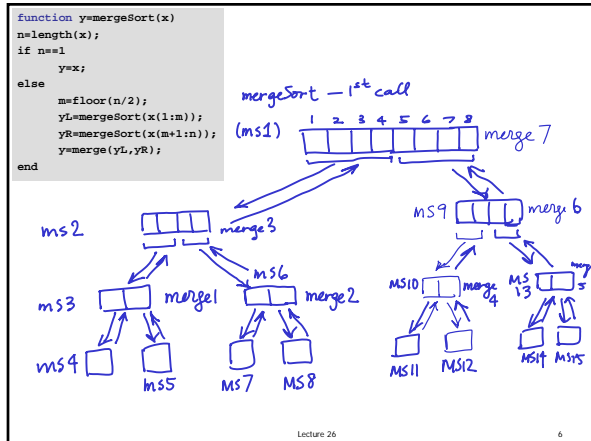
- Previous Lecture:
 - “Divide and conquer” strategies
 - Binary search
 - Merge sort
- Today’s Lecture:
 - “Divide and conquer” strategies (cont’d)—recursion
 - Merge sort
 - Removing a character (e.g., the blank) from a string
 - Tiling (subdividing) a triangle, e.g., Sierpinski Triangle
 - Some efficiency considerations
- Announcements
 - Project 6 due Dec 2nd at 11pm
 - CS1112 final will be 12/10 (Fri) 9am in Barton indoor field West. Email Randy Hess (rbhess@cs.cornell.edu) your entire exam schedule if you have a conflict. We must have this information by Thursday (12/2).

Merge sort is a “divide-and-conquer” strategy



Lecture 26

2



Lecture 26

6

How do merge sort, insertion sort, and bubble sort compare?

- Insertion sort and bubble sort are similar
 - Both involve a series of comparisons and swaps
 - Both involve nested loops
- Merge sort uses recursion

Lecture 26

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```
function x = insertSort(x)
% Sort vector x in ascending order with insertion sort
n = length(x);
for i = 1:n-1
% Sort x(1:i+1) given that x(1:i) is sorted
j = i;
need2swap = x(j+1) < x(j);
while need2swap
% swap x(j+1) and x(j)
temp = x(j);
x(j) = x(j+1);
x(j+1) = temp;
j = j-1;
need2swap = j > 0 && x(j+1) < x(j);
end
end
```

Insertion sort is more efficient than bubble sort on average—fewer comparisons (Lecture 24)

Lecture 26

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How do merge sort and insertion sort compare?

- Insertion sort: (worst case) makes i comparisons to insert an element in a sorted array of i elements. For an array of length N :
_____ for big N
- Merge sort: _____
- Insertion sort is done *in-place*; merge sort (recursion) requires much more memory

Lecture 26

10

```
function y = mergeSort(x)
% x is a vector. y is a vector
% consisting of the values in x
% sorted from smallest to largest.
```

```
n = length(x);
if n==1
    y = x;
else
    m = floor(n/2);
    yL = mergeSort(x(1:m));
    yR = mergeSort(x(m+1:n));
    y = merge(yL,yR);
end
```

All the comparisons between vector values are done in merge

Lecture 26

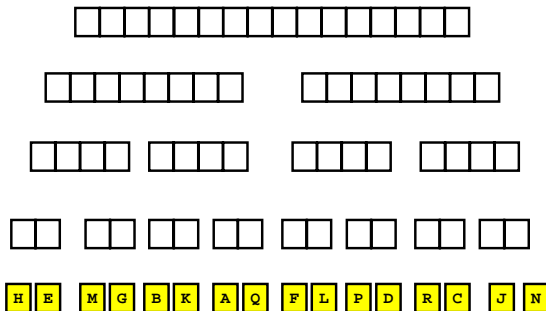
12

```
function z = merge(x,y)
nx = length(x); ny = length(y);
z = zeros(1, nx+ny);
ix = 1; iy = 1; iz = 1;
while ix<=nx && iy<=ny
    if x(ix) <= y(iy)
        z(iz) = x(ix); ix=ix+1; iz=iz+1;
    else
        z(iz) = y(iy); iy=iy+1; iz=iz+1;
    end
end
while ix<=nx % copy remaining x-values
    z(iz) = x(ix); ix=ix+1; iz=iz+1;
end
while iy<=ny % copy remaining y-values
    z(iz) = y(iy); iy=iy+1; iz=iz+1;
end
```

Lecture 26

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Merge sort: $\log_2(N)$ "levels"; N comparisons each level



Lecture 26

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How to choose??

- Depends on application
- Merge sort is especially good for sorting **large data set** (but watch out for memory usage)
- Insertion sort is "order N^2 " at **worst case**, but what about an **average case**? If the application requires that you *maintain* a sorted array, insertion sort may be a good choice

Lecture 26

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Why not just use Matlab's sort function?

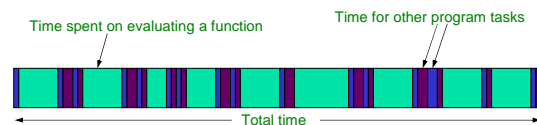
- **Flexibility**
- E.g., to maintain a sorted list, just write the code for insertion sort
- E.g., sort strings or other complicated structures
- Sort according to some criterion set out in a function file
 - Observe that we have the comparison $x(j+1) < x(j)$
 - The comparison can be a function that returns a **boolean** value
- Can combine different sort/search algorithms for specific problem

Lecture 26

18

Expensive function evaluations

- Consider the execution of a program that is dominated by multiple calls to an expensive-to-evaluate function (e.g., climate simulation models)



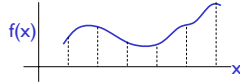
- Can try to improve efficiency by dealing with the expensive function evaluations

Lecture 26

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Dealing with expensive function evaluations

- Can the function code be improved?
- Can we do fewer function evaluations?
- Can we **pre-compute** and **store** specific function values so that during the main program execution the program can just **look up** the values?
 - Consider function $f(x)$. If there are many function calls and few distinct values of x , can get substantial speedup
 - Only speeds up main program execution—it still takes time to do the pre-computation



Lecture 26

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What are some issues and potential problems with the “table look-up” strategy?

x	$f(x)$
1	1.01
2	2.67
3	5.71
4	9.12
5	7.98
:	:

Pre-calculate and store these values (e.g., in a vector H)

- Accuracy—need a “dense grid” to get high accuracy
→ significant memory usage
- If an exact x -value is not found, need some kind of approximation
- Incur searching cost if the x -values are not simple indices
- Feasible in high dimensions (multiple dependent variables)?

To be continued in this week's lab.

Lecture 26

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Example: removing all occurrences of a character

- Can solve using iteration—check one character (one component of the vector) at a time
- Can solve using **recursion**
 - E.g., **remove** all the blanks in string s
Same as **remove blank** in $s(1)$
and **remove blanks** in $s(2:\text{length}(s))$

Lecture 26

24

```
function s = removeChar(c, s)
% Return string s with character c removed

if length(s)==0 % Base case: nothing to do
    return
else
    if s(1)~=c

    else

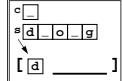
    end
end
```

Lecture 26

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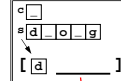
```
function s = removeChar(c, s)
if length(s)==0
    return
else
    if s(1)~=c
        s = [s(1) removeChar(c, s(2:length(s)))];
    else
        s = removeChar(c, s(2:length(s)));
    end
end
```

s [d _ o _ g]
 c []

removeChar - 1st call

```
function s = removeChar(c, s)
if length(s)==0
    return
else
    if s(1)~=c
        s = [s(1) removeChar(c, s(2:length(s)))];
    else
        s = removeChar(c, s(2:length(s)));
    end
end
```

s [d _ o _ g]
 c []

removeChar - 1st callremoveChar - 2nd call