

- Previous Lecture:
 - Linear search, binary search
 - Insertion sort
 - (Reading: Bubble Sort)
- Today's Lecture:
 - Merge Sort
 - What's next?
- Announcements
 - P6 due Thursday at 11pm
 - Final exam: Dec 7th 2-4:30pm, Rockefeller Hall
 - Last names beginning with A-N: Room 201
 - Last names beginning with O-Z: Room 203

Linear search and binary search

- **Linear search**
 - "Effort" is **linearly proportional** to n , the size of the search space (e.g., the length of the vector)
 - Can represent effort by the number of comparisons against the search target done during the search
- **Binary search**
 - Effort is proportional to $\log_2(n)$ where n is the size of the search space
- Saving of $\log_2(n)$ over n is significant when n is large! But binary search requires sorted vector

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Binary search is efficient, but we need to sort the vector in the first place so that we can use binary search

- Many different algorithms out there...
- We saw insertion sort (and read about bubble sort)
- Let's look at **merge sort**
- An example of the "divide and conquer" approach using recursion

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Which task is "easier," **sort a length 1000 array** or **merge* two length 500 sorted arrays into one?**

A. Sort

B. Merge

*Merge two sorted arrays so that the resultant array is sorted

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Motivation: merging is an easier job than sorting!

If I have two helpers, I'd...

- Give each helper half the array to sort
- Then I get back the sorted subarrays and **merge** them.

What if those two helpers each had two sub-helpers?

And the sub-helpers each had two sub-sub-helpers? And...

Subdivide the sorting task

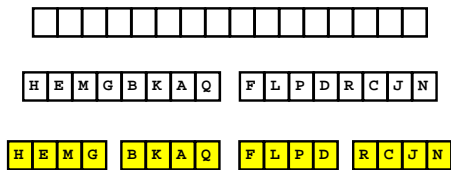
H E M G B K A Q F L P D R C J N

H E M G B K A Q F L P D R C J N

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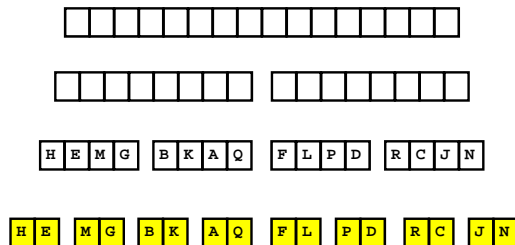
Subdivide again



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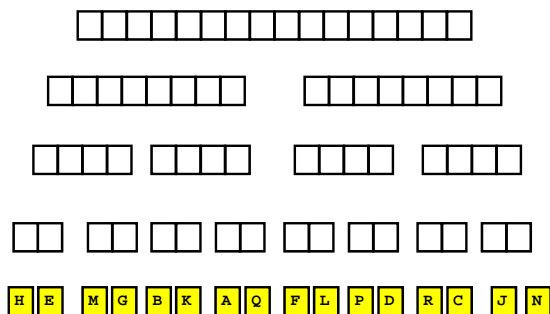
And again



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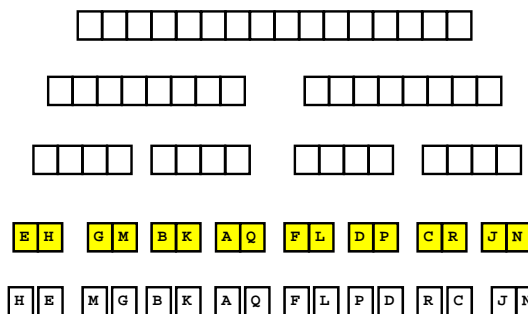
And one last time



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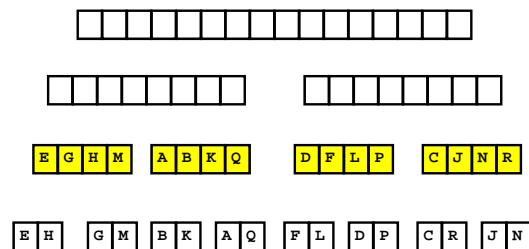
Now merge



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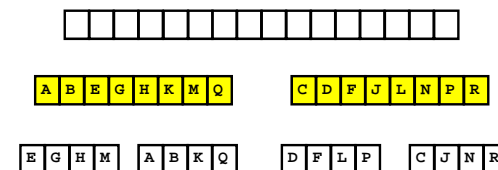
And merge again



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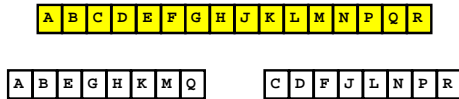
And again



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And one last time



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```
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 = mergeSortL(x(1:m));
    yR = mergeSortR(x(m+1:n));
    y = merge(yL,yR);
end
```

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The central sub-problem is the **merging** of two sorted arrays into one single sorted array

12 33 35 45

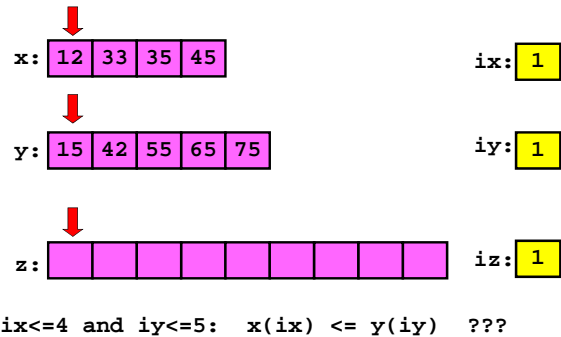
15 42 55 65 75

12 15 33 35 42 45 55 65 75

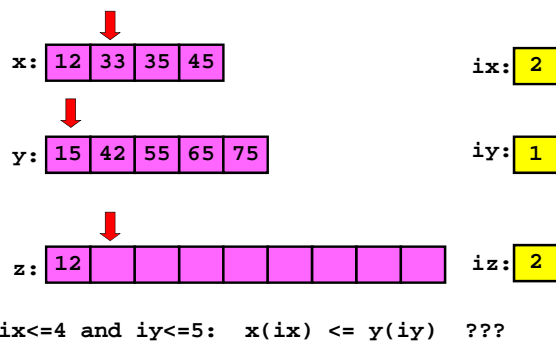
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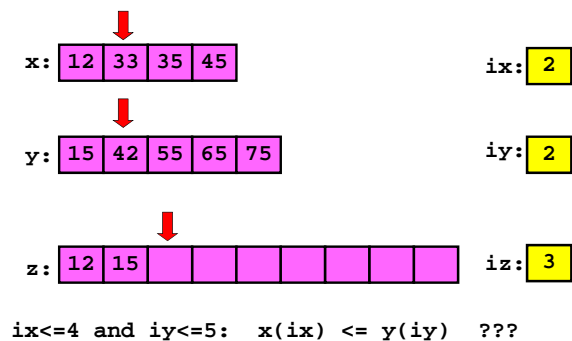
Merge

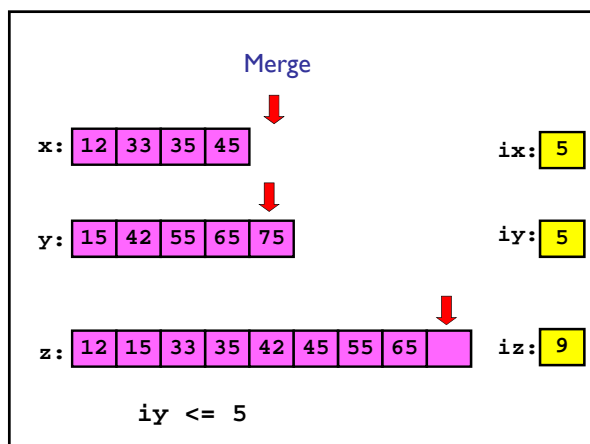
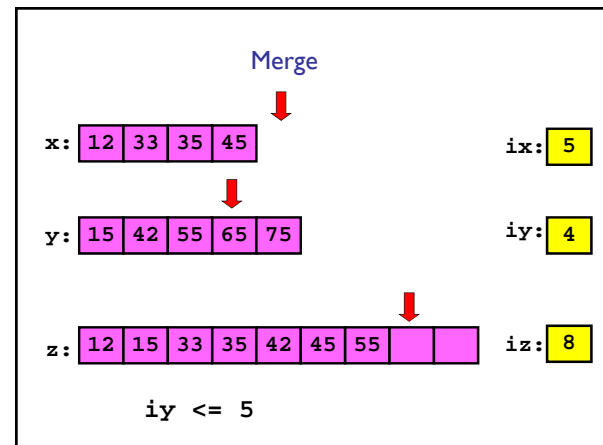
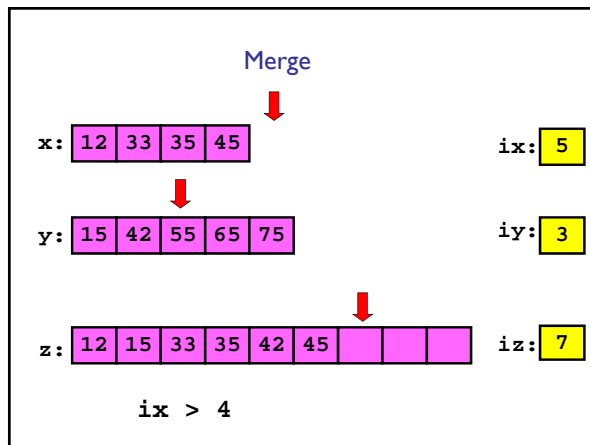
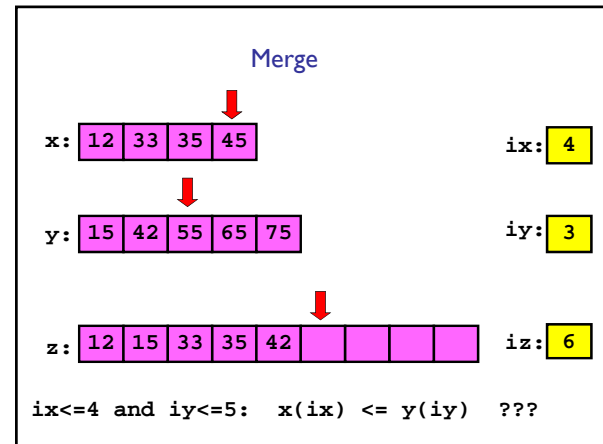
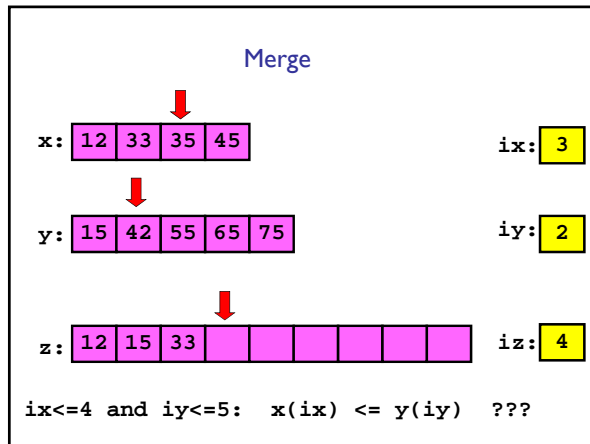


Merge



Merge





```
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

    end
    % Deal with remaining values in x or y
```

```
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
end
```

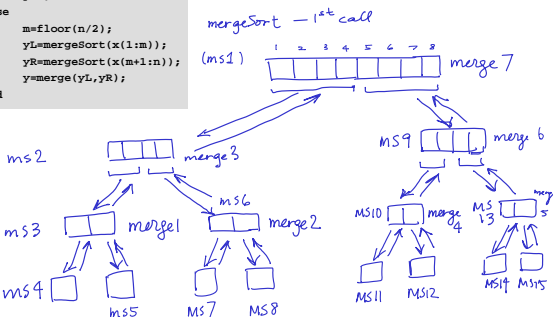
```
function y = mergeSort(x)
% x is a vector. y is a vector
% consisting of the values in x
% sorted from smallest to largest.
```

```
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
```

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```
function y=mergeSort(x)
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
```



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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

See [InsertionSort.m](#)

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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;
    while j>0 && x(j+1)<x(j)
        % swap x(j+1) and x(j)
        temp= x(j);
        x(j)= x(j+1);
        x(j+1)= temp;

        j= j-1;
    end
end
```

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How do merge sort and insertion sort compare?

- Insertion sort: (worst case) makes k comparisons to insert an element in a sorted array of k 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

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```
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

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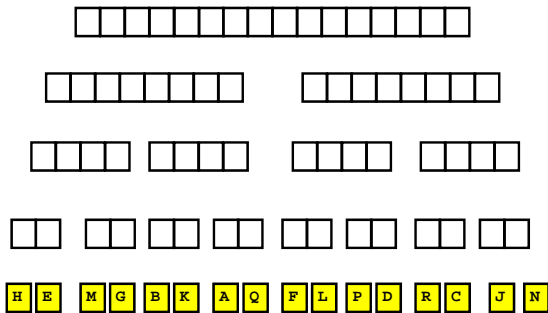
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```
function z = merge(x,y)
nx = length(x); ny = length(y);
z = zeros(1, nx+ny);
ix = 1; iy = 1; iz = 1;
while ix<=nx && iy<=ny
    if x(ix) <= y(iy)
        z(iz) = x(ix); ix=ix+1; iz=iz+1;
    else
        z(iz) = y(iy); iy=iy+1; iz=iz+1;
    end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

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



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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

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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

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We've reached the end of CS1112... now what?

- Continue practicing your problem solving—problem decomposition—skills, in programming and other arenas!
- Interested in further study?
 - ENGRD/CS 2110 Object-oriented programming and data structure

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ENGRD/CS 2110 OOP and Data Structures

- Learn new programming concepts and further explores those you've seen in CS1112
 - OOP, program design and development
 - Recursion
 - Complex data structures and related algorithms
- Taught in **Java**
- Optional **CS 2111** meets 1 hr/week; additional practice with OOP, Java, and other course topics
- During break, check out this website:
<http://www.cs.cornell.edu/courses/CS1130/2015sp/>

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We've reached the end of CS1112... now what?

- Continue practicing your problem solving—problem decomposition—skills, in programming and other arenas!
- Interested in further study?
 - ENGRD/CS 2110 Object-oriented programming and data structure
 - Short courses in **Python (CS 1133)**, **C++ (CS 2024)**, ..., etc.
 - More general CS courses: **CS 2800 Discrete structures**, **CS 2850 Networks**

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What we learned...

- Develop/implement **algorithms** for problems
- Develop programming skills
 - Design, implement, document, test, and debug
- Programming “tool bag”
 - Functions for reducing redundancy
 - Control flow (if-else; loops)
 - Recursion
 - Data structures
 - Graphics
 - File handling

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What we learned... (cont'd)

- Applications and concepts
 - Image processing
 - Object-oriented programming
 - Sorting and searching—you should know the algorithms covered
 - Divide-and-conquer strategies
 - Approximation and error
 - Simulation
 - Computational effort and efficiency

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Computing gives us *insight* into a problem

- Computing is not about getting one answer!
- We build models and write programs so that we can “play” with the models and programs, learning—gaining insights—as we vary the parameters and assumptions
- Good models require domain-specific knowledge (and experience)
- Good programs ...
 - are modular and cleanly organized
 - are well-documented
 - use appropriate data structures and algorithms
 - are reasonably efficient in time and memory

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Final Exam

- Dec 7, 2-4:30pm, Rockefeller Hall 201 (A-N), 203 (O-Z)
- Covers entire course; some emphasis on material after Prelim 2
- Closed-book exam, no calculators
- Bring student ID card
- Check for announcements on webpage:
 - Study break office/consulting hours
 - Review session time and location
 - Review questions
 - List of potentially useful functions

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