

CSCI 104

Log Structured Merge Trees

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Series Summation Review

- Let $n = 1 + 2 + 4 + \dots + 2^k = \sum_{i=0}^k 2^i$. What is n ?
 - $n = 2^{k+1} - 1$
- What is $\log_2(1) + \log_2(2) + \log_2(4) + \log_2(8) + \dots + \log_2(2^k)$
 $= 0 + 1 + 2 + 3 + \dots + k = \sum_{i=0}^k i$
 - $O(k^2)$
- So then what if $k = \log(n)$ as in:
 $\log_2(1) + \log_2(2) + \log_2(4) + \log_2(8) + \dots + \log_2(2^{\log(n)})$
 - $O(\log^2 n)$

Arithmetic series:

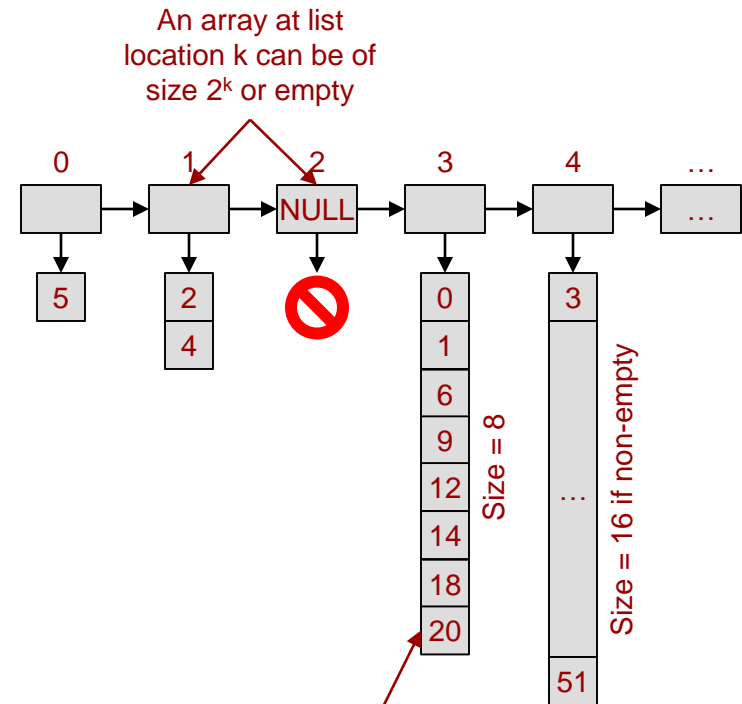
$$\sum_{i=1}^n i = \frac{n(n+1)}{2} = \theta(n^2)$$

Geometric series

$$\sum_{i=1}^n c^i = \frac{c^{n+1} - 1}{c - 1} = \theta(c^n)$$

Merge Trees Overview

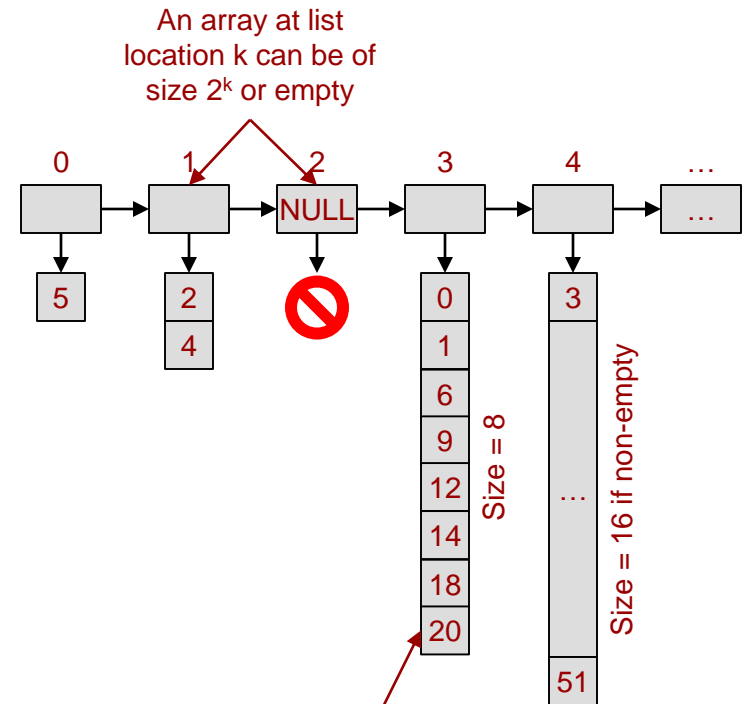
- Consider a list of (pointers to) arrays with the following constraints
 - Each array is sorted *though no ordering constraints exist between arrays*
 - The array at list index k is of exactly size 2^k or empty



Note: These are the keys for a set (or key,value pairs for a map)

Merge Trees Size

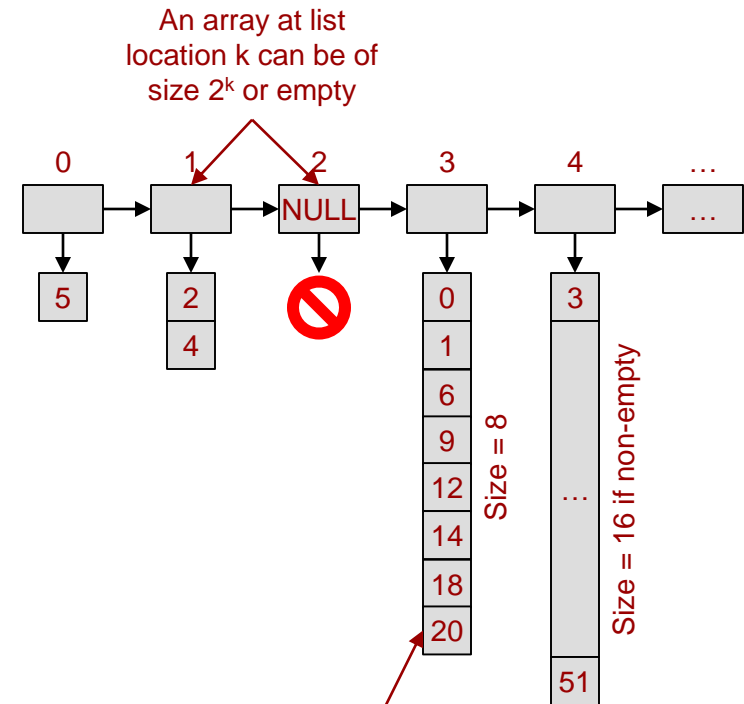
- Define...
 - n as the # of keys in the entire structure
 - k as the size of the list (i.e. positions in the list)
- Given list of size k, how many total values, n, may be stored?
 - Let $n = 1 + 2 + 4 + \dots + 2^{k-1} = \sum_{i=0}^{k-1} 2^i$.
 What is n?
- $n=2^k-1$



Note: These are the keys for a set (or key,value pairs for a map)

Merge Trees Find Operation

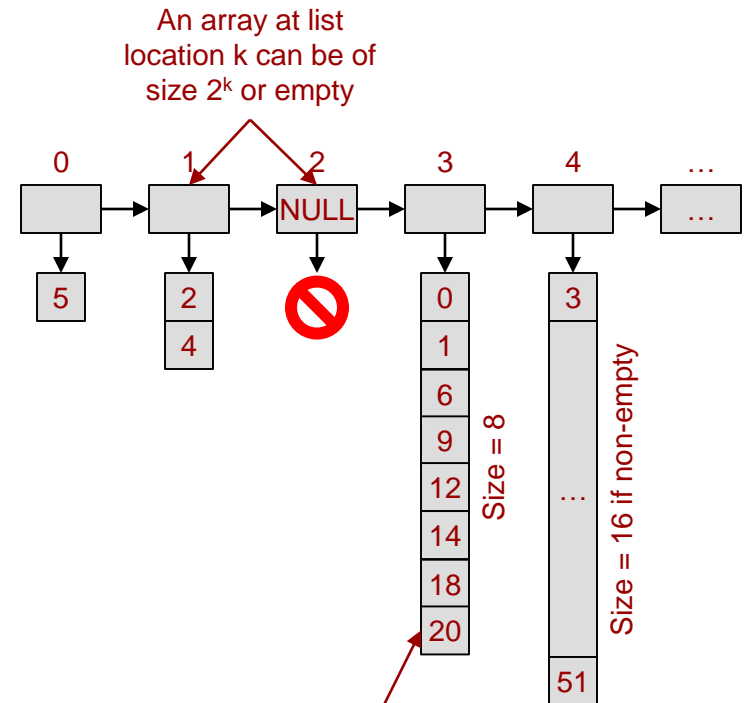
- To find an element (or check if it exists)
- Iterate through the arrays in order (i.e. start with array at list position 0, then the array at list position 1, etc.)
 - In each array perform a binary search
- If you reach the end of the list of arrays without finding the value it does not exist in the set/map



Note: These are the keys for a set (or key,value pairs for a map)

Find Runtime

- What is the worst case runtime of find?
 - When the item is not present which requires, a binary search is performed on each list
- $T(n) = \log_2(1) + \log_2(2) + \dots + \log_2(2^{k-1})$
- $= 0 + 1 + 2 + \dots + k-1 = \sum_{i=0}^{k-1} i$
- $= O(k^2)$
- But let's put that in terms of the number of elements in the structure (i.e. n)
 - Recall $k = \log_2(n+1)$
- So find is $O(\log_2(n)^2)$

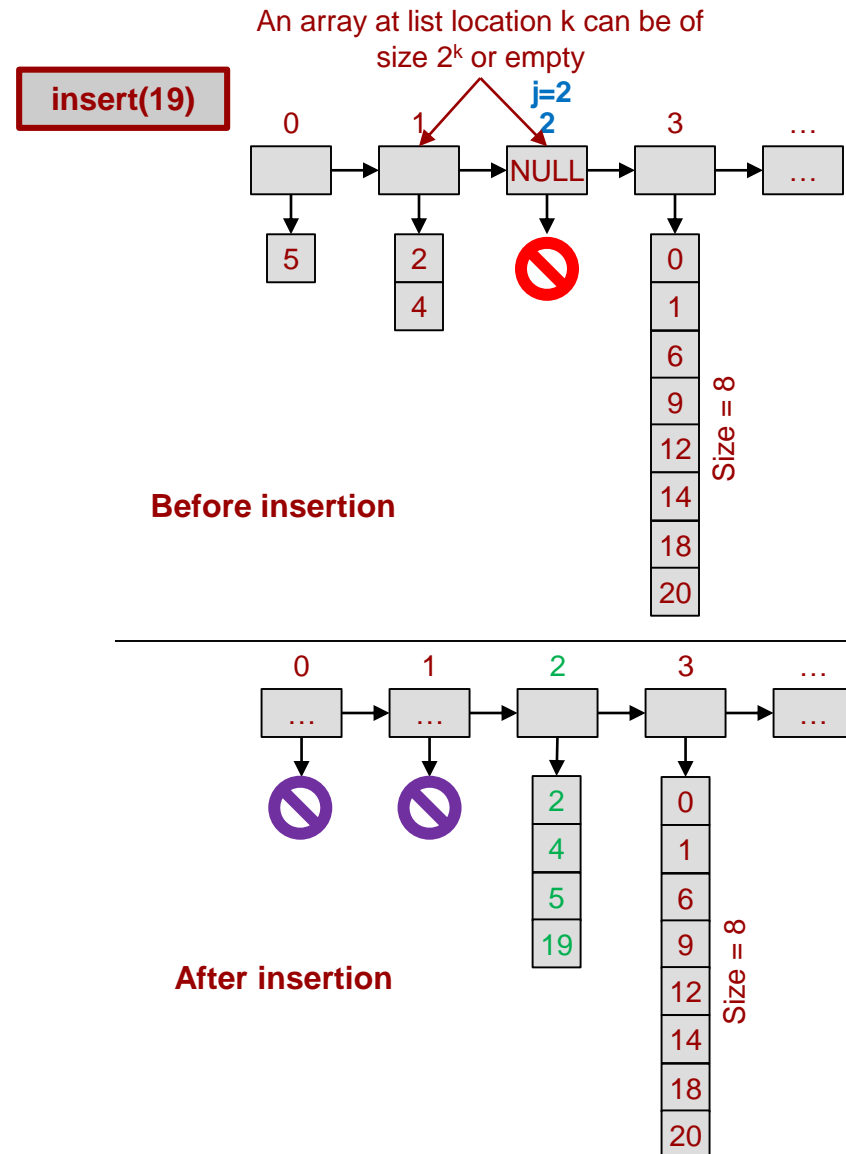


Improving Find's Runtime

- While we might be okay with $[\log(n)]^2$, how might we improve the find runtime in the general case?
 - Hint: I would be willing to pay $O(1)$ to know if a key is not in a particular array without having to perform find
- A Bloom filter could be maintained alongside each array and allow us to skip performing a binary search in an array

Insertion Algorithm

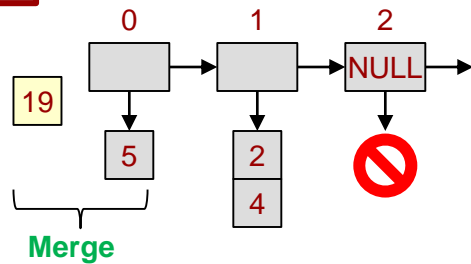
- Let j be the smallest integer such that array j is empty (first empty slot in the list of arrays)
- An insertion will cause
 - Location j 's array to become filled
 - Locations 0 through $j-1$ to become empty



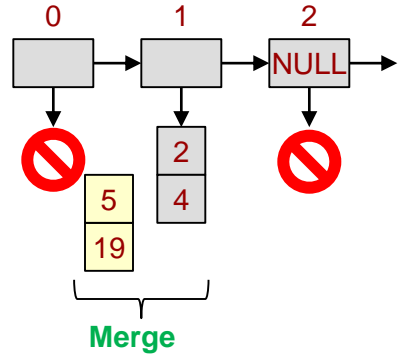
Insertion Algorithm

- Starting at array 0, iteratively merge the previously merged array with the next, stopping when an empty location is encountered

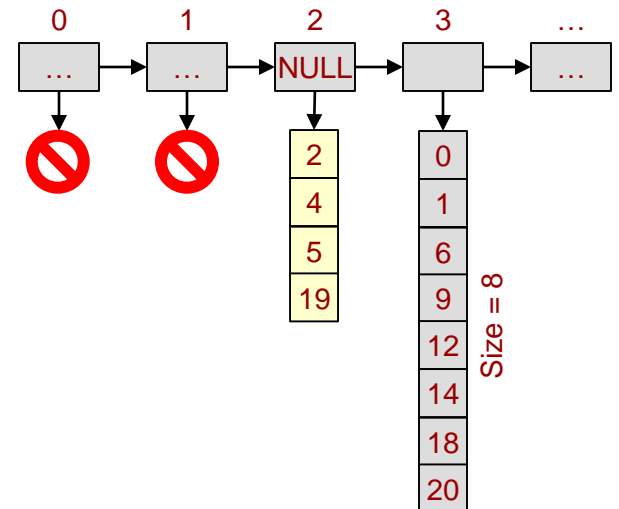
insert(19)



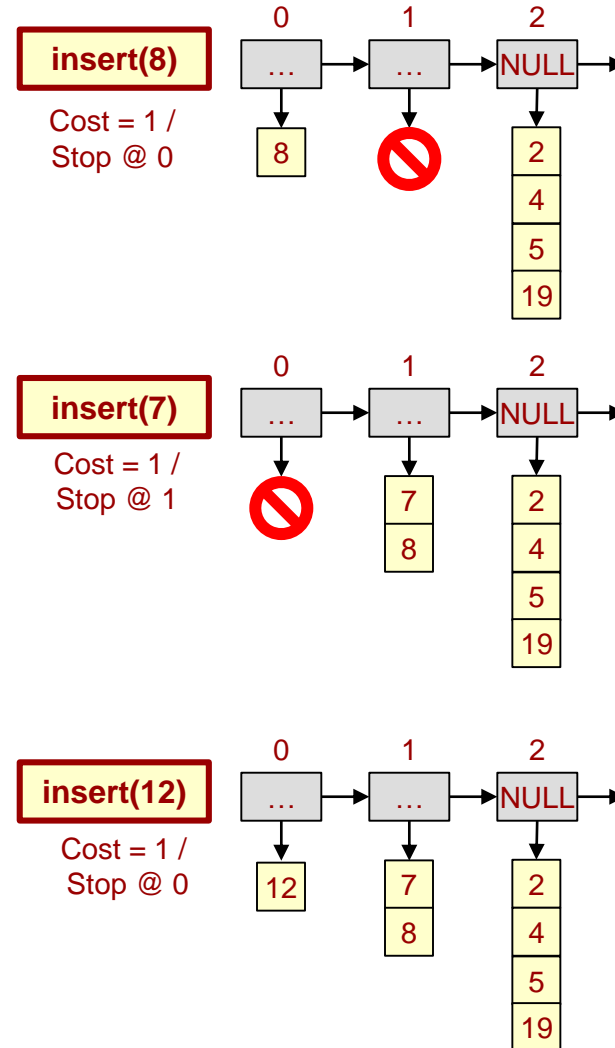
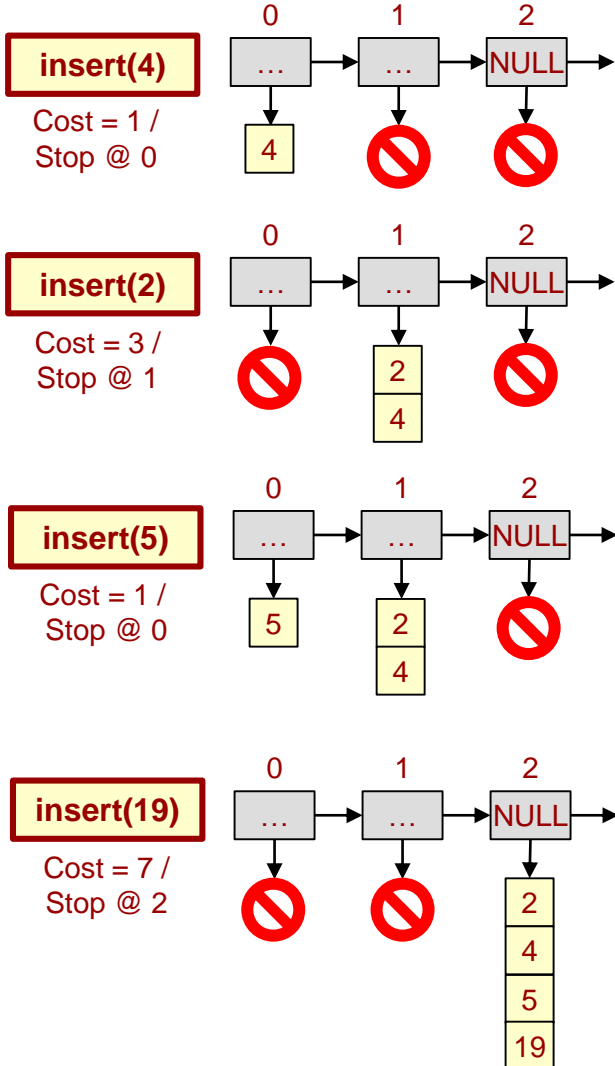
List 0 is full so merge two arrays of size 1



List 1 is full so merge two arrays of size 2

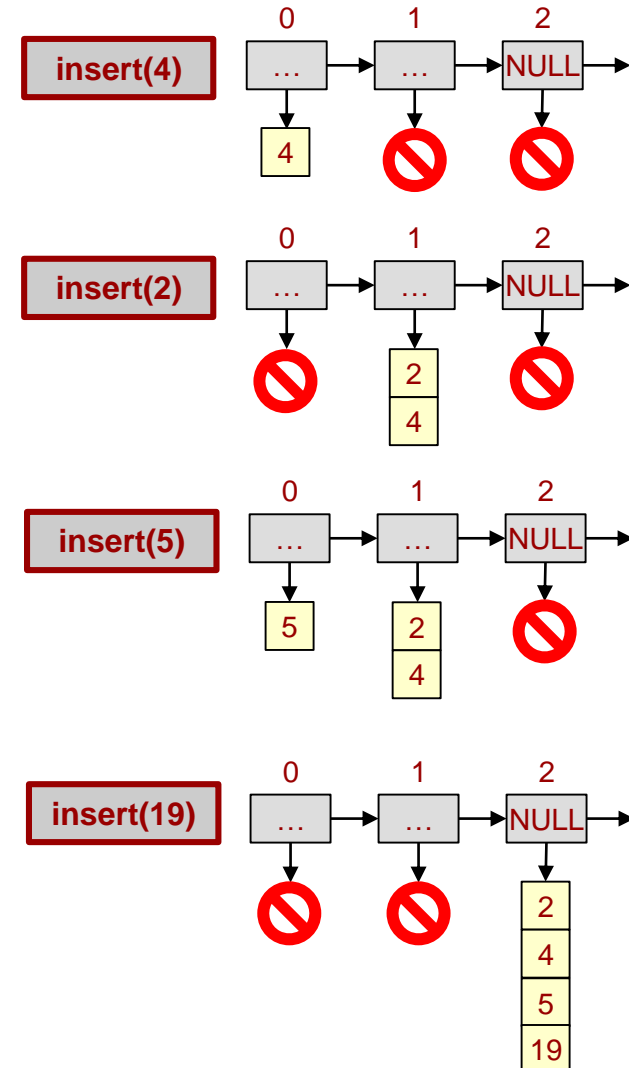


Insert Examples



Insertion Runtime: First Look

- Best case?
 - First list is empty and allows direct insertion in $O(1)$
- Worst case?
 - All list entries (arrays) are full so we have to merge at each location
 - In this case we will end with an array of size $n=2^k$ in position k
 - Also recall merging two sorted arrays of size $m/2$ is $\Theta(m)$
 - So the total cost of all the merges is $1 + 2 + 4 + 8 + \dots + 2^k = \Theta(2^{k+1}) = \Theta(n)$
- But if the worst case occurs how soon can it occur again?
 - It seems the costs vary from one insert to the next
 - This is a good place to use amortized analysis

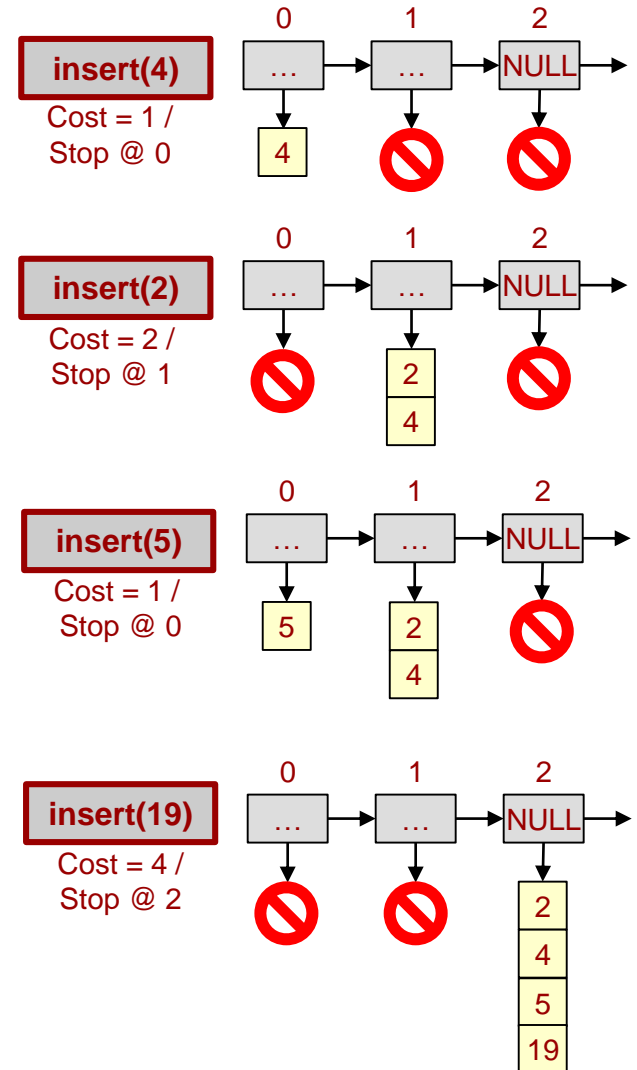


Total Cost for N insertions

- Reminder: Insert stopping at location k requires $1+2+4+\dots+2^{k-1}+2^k = 2^{k+1}-1 = O(2^{k+1})$ merge steps
- Total cost of n=16 insertions:
 - Stop at: 0,1,0,2,0,1,0,3,0,1,0,2,0,1,0,4
 - Cost: $2^1+2^2+2^1+2^3+2^1+2^2+2^1+2^4+2^1+2^2+2^1+2^3+2^1+2^2+2^1+2^5$
- $=2^1*n/2 + 2^2*n/4 + 2^3*n/8 + 2^4*n/16 + 2^5*1$
- $=n + n + n + n + 2*n$
- $=n*\log_2(n) + 2n$
- Amortized cost = Total cost / n operations
 - $\log_2(n) + 2 = O(\log_2(n))$

Amortized Analysis of Insert

- We have said when you end (place an array) in position k you have to do $O(2^{k+1})$ work for all the merges
- How often do we end in position k
 - The 0th position will be free with probability $\frac{1}{2}$ ($p=0.5$)
 - We will stop at the 1st position with probability $\frac{1}{4}$ ($p=0.25$)
 - We will stop at the 2nd position with probability $\frac{1}{8}$ ($p=0.125$)
 - We will stop at the kth position with probability $\frac{1}{2^{k+1}} = 2^{-(k+1)}$
- So we pay 2^{k+1} with probability $2^{-(k+1)}$
- Suppose we have n items in the structure (i.e. max k is $\log_2 n$) what is the expected cost of inserting a new element
 - $\sum_{k=0}^{\log(n)} 2^{k+1} 2^{-(k+1)} = \sum_{k=0}^{\log(n)} 1 = \log(n)$



Summary

- Variants of **log structured merge trees** have found popular usage in industry
 - Starting array size might be fairly large (size of memory of a single server)
 - Large arrays (from merging) are stored on disk
- Pros:
 - Ease of implementation
 - Sequential access of arrays helps lower its constant factors
- Operations:
 - Find = $\log^2(n)$
 - Insert = Amortized $\log(n)$
 - Remove = often not considered/supported
- More reading:
 - <http://www.benstopford.com/2015/02/14/log-structured-merge-trees/>