CSCI 104
Log Structured Merge Trees

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Series Summation Review

- Let n = 1 + 2 + 4 + … + 2^k = $\sum_{i=0}^{k} 2^i$. What is n?
  – n = 2^{k+1}-1
- What is $\log_2(1) + \log_2(2) + \log_2(4) + \log_2(8)+\ldots+ \log_2(2^k)$ = 0 + 1 + 2 + 3+… + k = $\sum_{i=0}^{k} i$
  – $O(k^2)$

Arithmetic series:

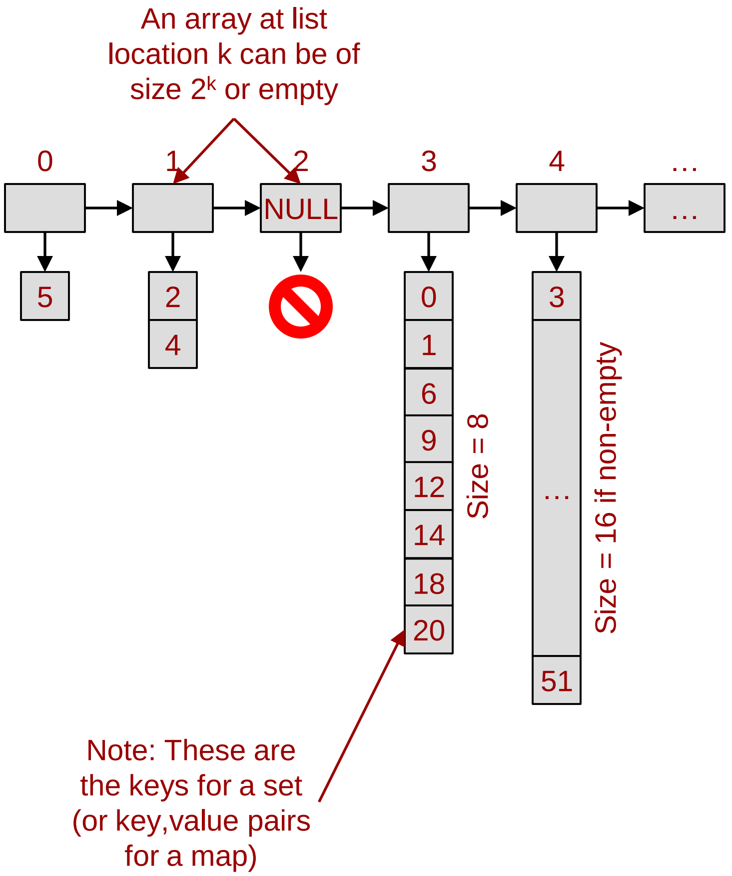
$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2} = \theta(n^2)$$

Geometric series

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

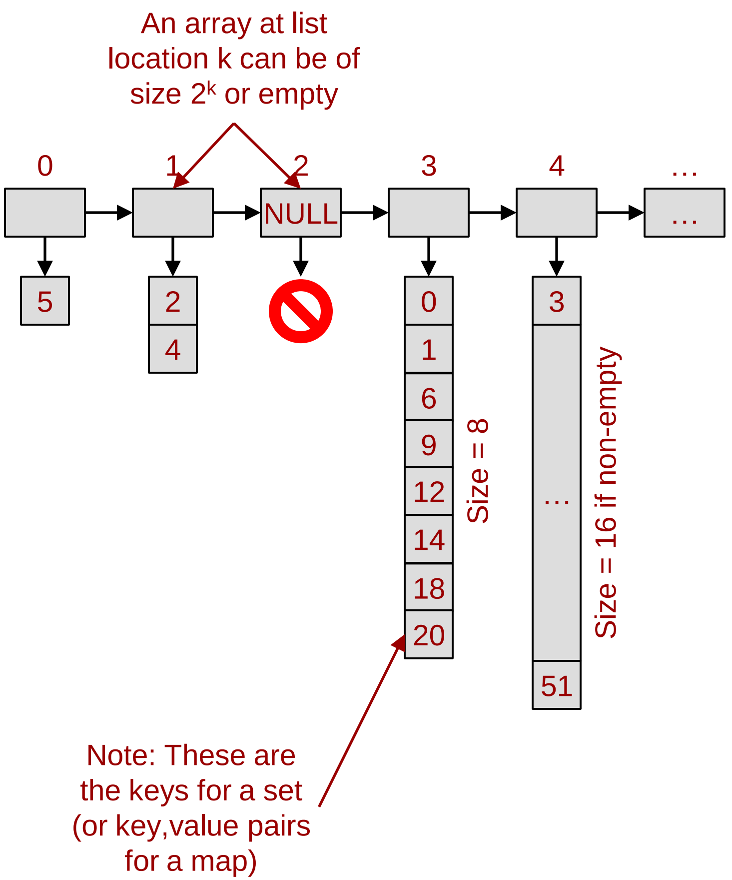
Merge Trees Overview

- 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



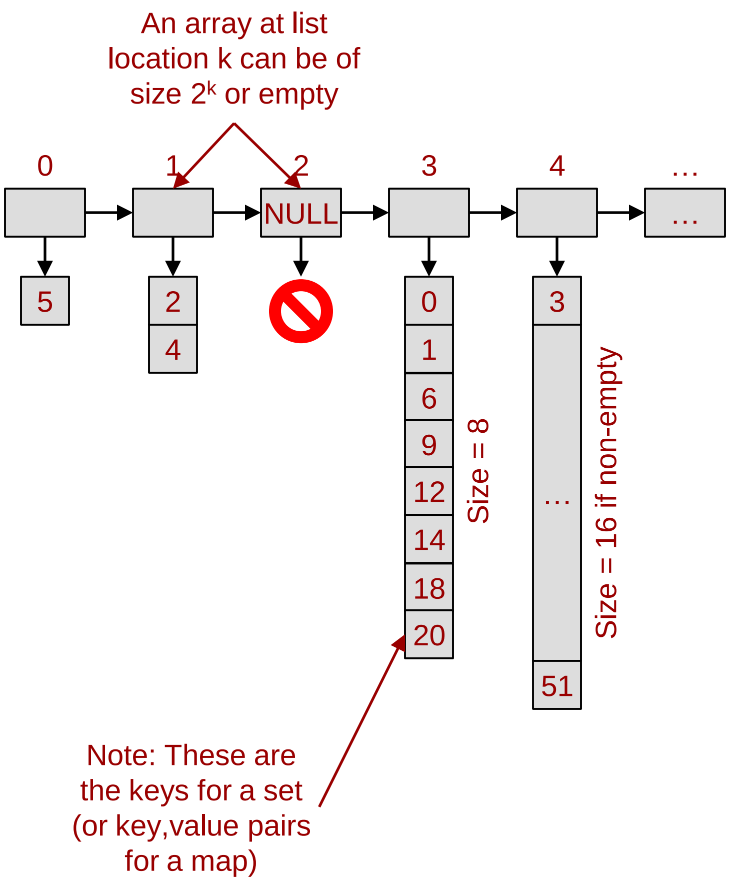
Merge Trees Size

- Define…
  - n as the # of keys in the entire structure
  - k as the size of the list (i.e. positions in the list)
- Given k, what is n?
  - Let n = 1 + 2 + 4 + … + 2^k = $\sum_{i=0}^{k} 2^i$. What is n?
- $n=2^k+1$



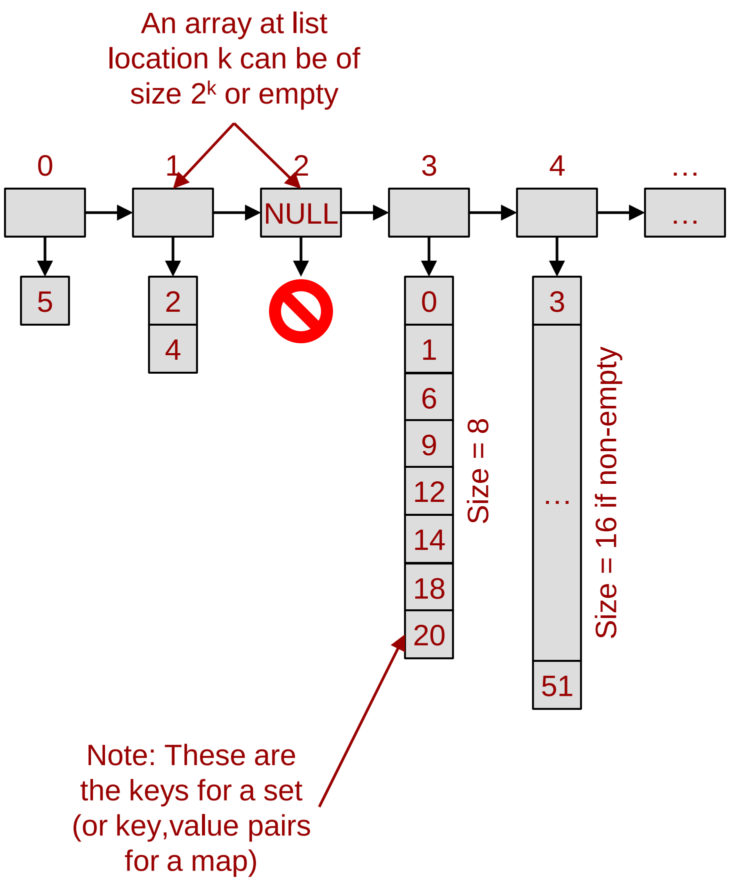
Merge Trees Find Operation

- 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



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
- $T(n) = \log_2(1) + \log_2(2) + \ldots \log_2(2^k)$
- $= 0 + 1 + 2 + \ldots + k = \sum_{i=0}^{k} i$
  $= O(k^2)$
- 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 $O(\log_2(n)^2)$

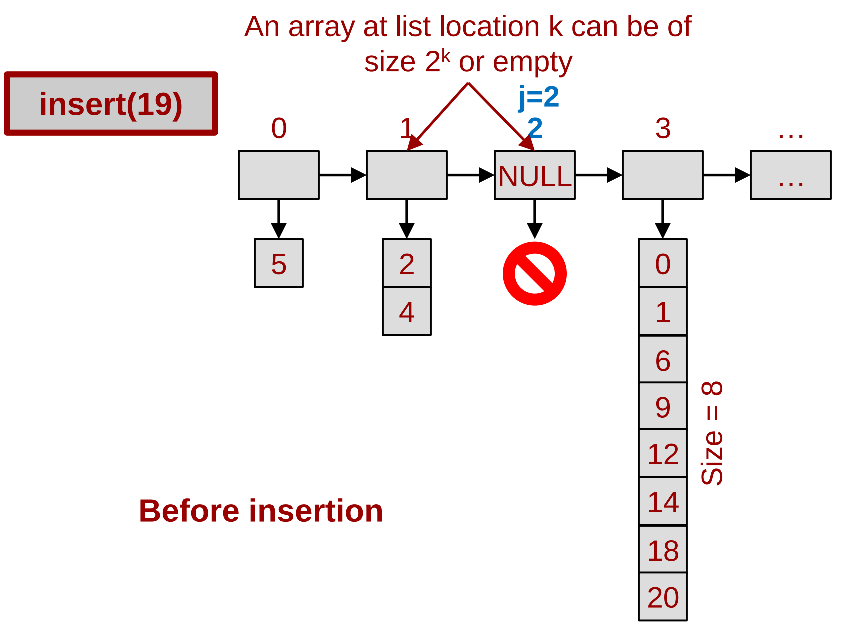


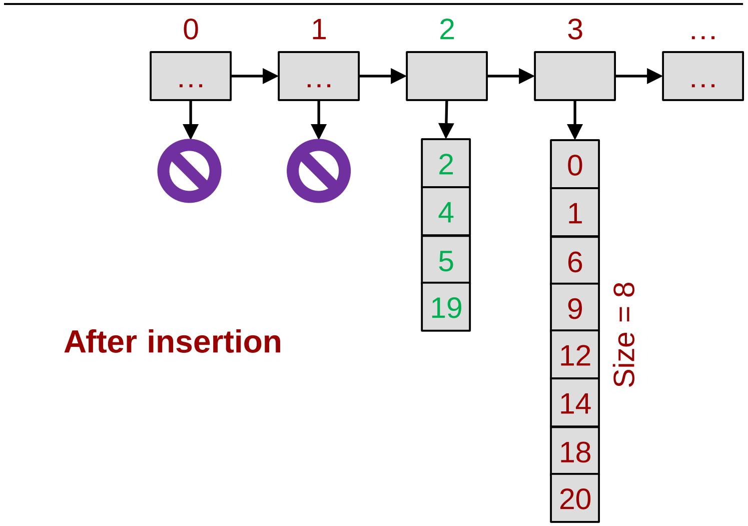
Improving Find's Runtime

- 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

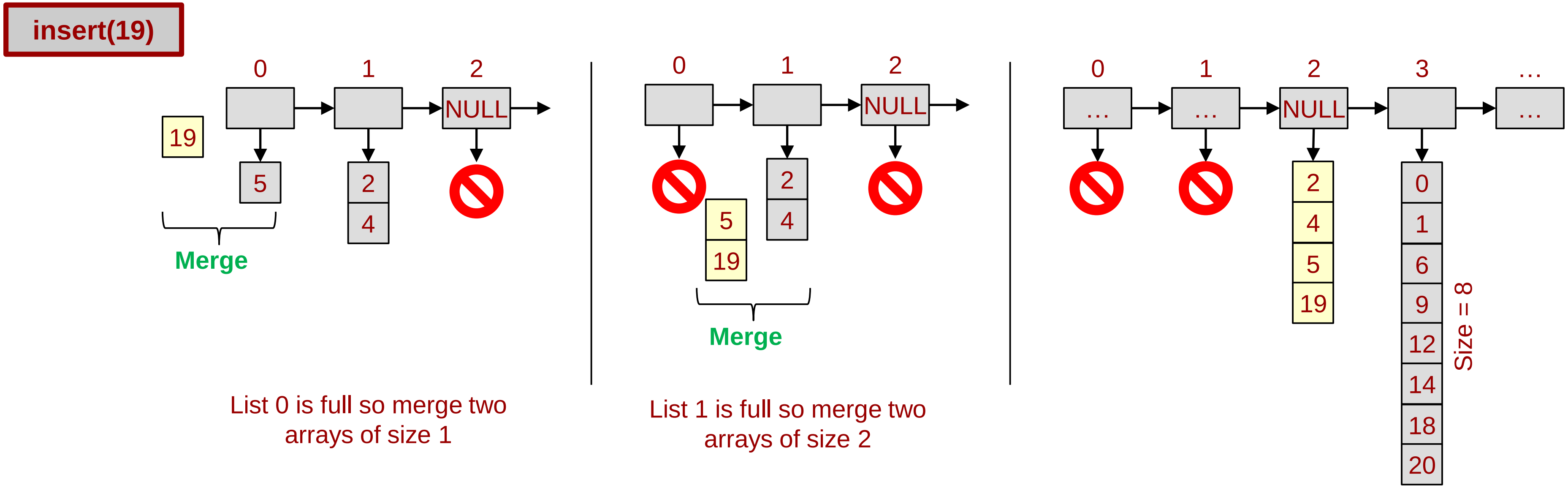
- Let j be the smallest integer such that array j is empty (first empty slot in the list of arrays)
- An insertion will cause
  - Location j's array to become filled
  - Locations 0 through j-1 to become empty



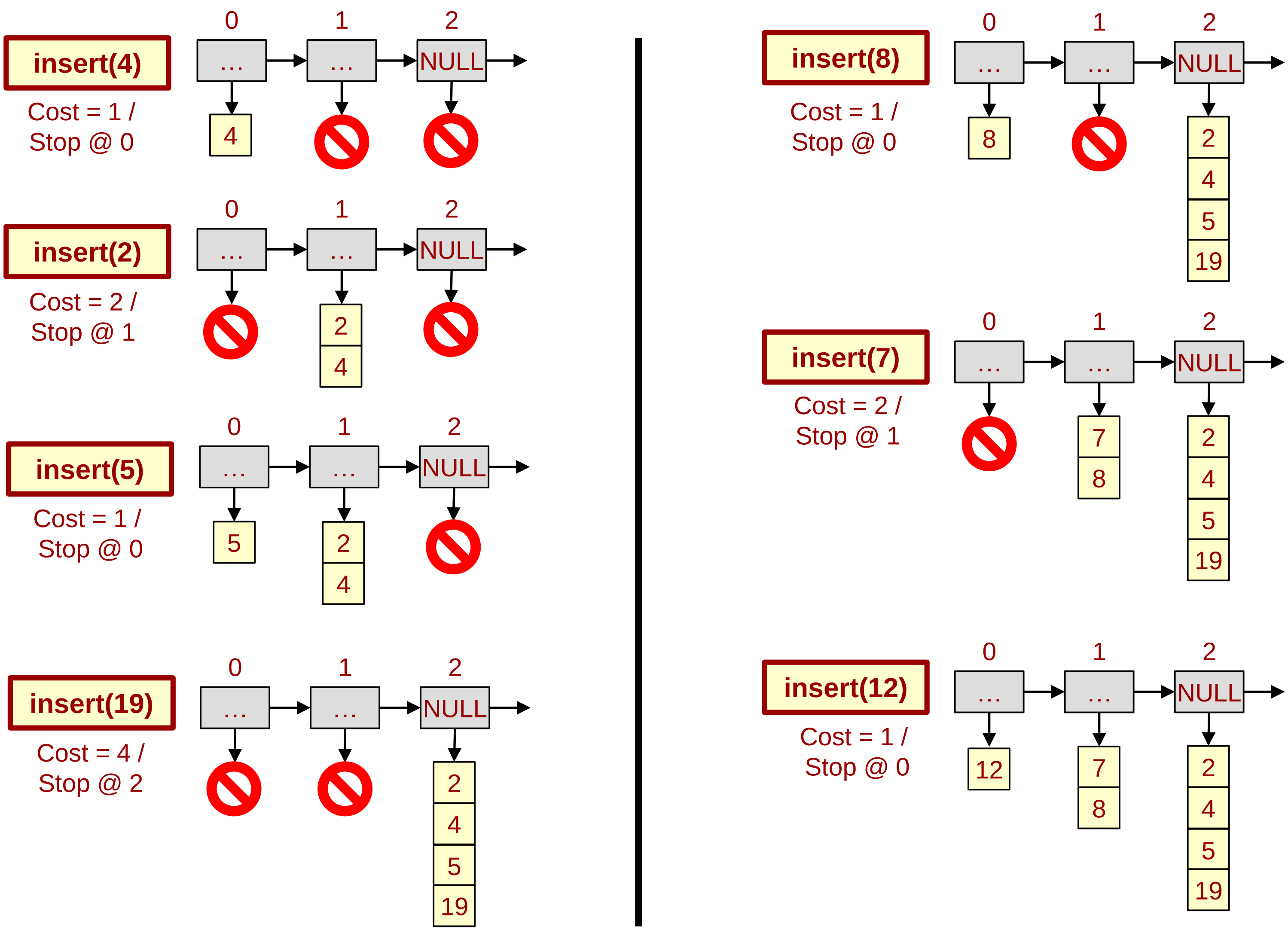


Insertion Algorithm

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

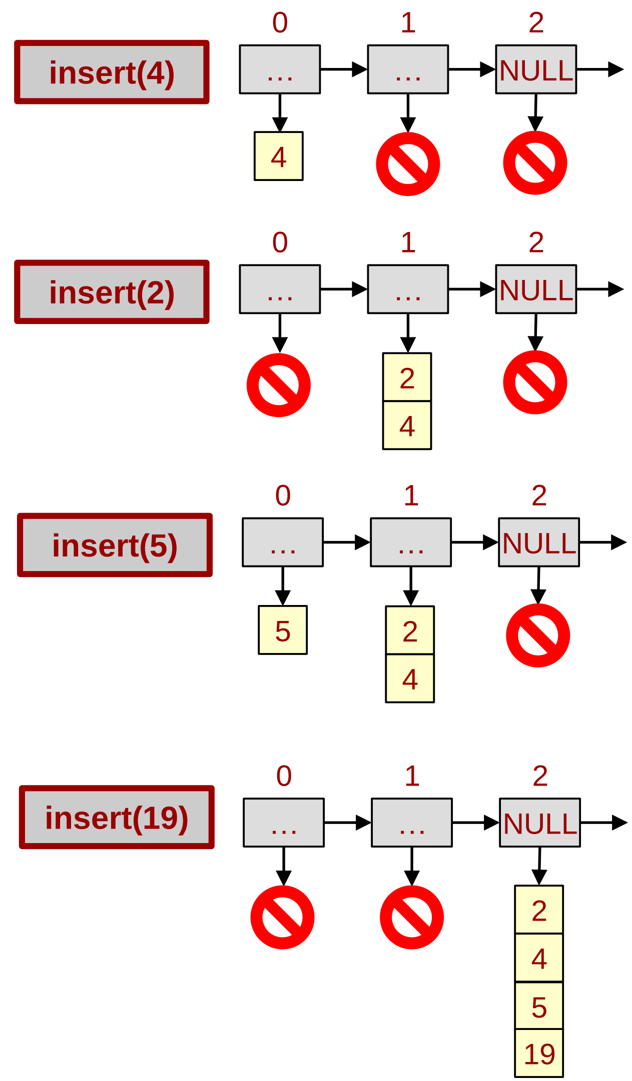


Insert Examples



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 arrays of size m is $\Theta(m)$
  - So the total cost of all the merges is $1 + 2 + 4 + 8 + \ldots + n = 2*n-1 = \Theta(n) = \Theta(2^k)$
- 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

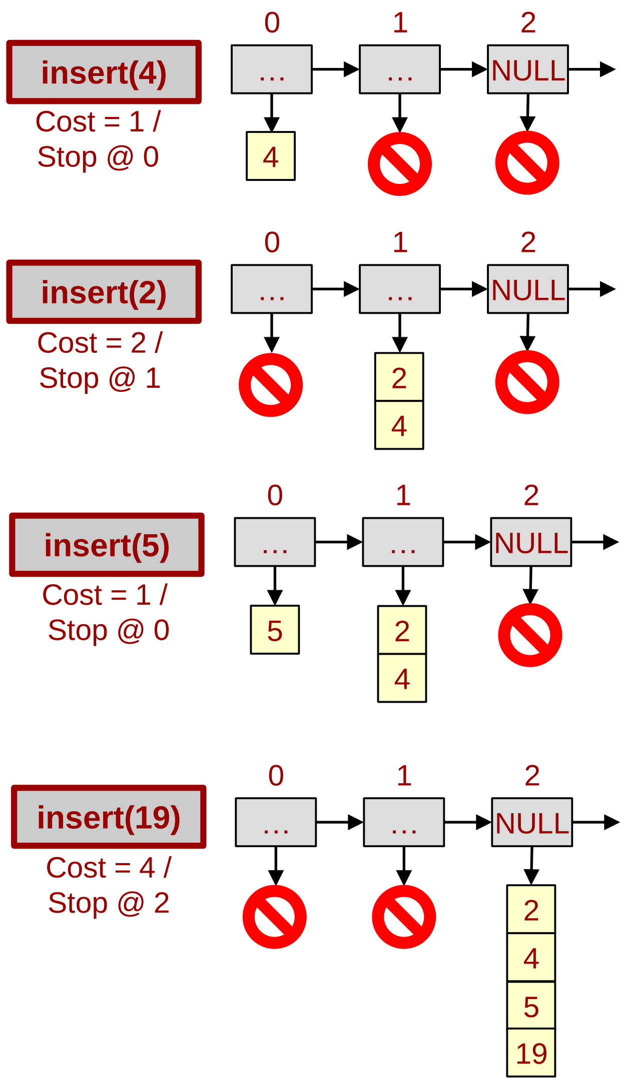


Total Cost for N insertions

- Total cost of n=16 insertions:
  - 1+2+1+4+1+2+1+8+1+2+1+4+1+2+1+16
- =1*n/2 + 2*n/4 + 4*n/8 + 8*n/16 + n
- =n/2 + n/2 + n/2 + n/2 + n
- $=n/2*\log_2(n) + n$
- Amortized cost = Total cost / n operations
  - $\log_2(n)/2 + 1 = O(\log_2(n))$

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 0^{th} position will be free with probability ½ (p=0.5)
  - We will stop at the 1^{st} position with probability ¼ (p=0.25)
  - We will stop at the 2^{nd} position with probability 1/8 (p=0.125)
  - We will stop at the k^{th} position with probability $1/2^k = 2^{-k}$
- 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