



- A randomized algorithm is an algorithm that employs a degree of randomness as part of its logic.
 - The algorithm typically uses uniformly random bits as an auxiliary input to guide its behaviour, in the hope of achieving good performance in the average case over all possible choices of random bits.
- The performance of a randomized algorithm is a random variable determined by the random bits.
 - The worst-case performance is typically bad with a very small probability but the average performance can be good.
- Two categories: Las Vegas algorithm and Monte Carlo algorithm

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Randomized Algorithm (2/2)

- Las Vegas algorithm
 - A Las Vegas algorithm is a randomized algorithm that always gives correct results
- Monte Carlo algorithm
 - A Monte Carlo algorithm is a randomized algorithm whose output may be incorrect with a certain (typically small) probability

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 Given an unsorted list where half of the elements have a key k1 and the other half have a key k2, find an element in the list with key k1.

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An Example (2/5) Las Vegas algorithm Algorithm findKey(L, k1) Input: list L, key k1 Output: an element in L with key k1 { repeat randomly select e∈L; until key(e)=k1; return e; }



Analysis:

- Probability of success: 1
- The number of iterations varies and can be arbitrarily large, but the expected number of iterations is:

$$\lim_{n\to\infty} \sum_{i=0}^n \frac{i}{2^i} = 2$$

• The expected time complexity is O(1)

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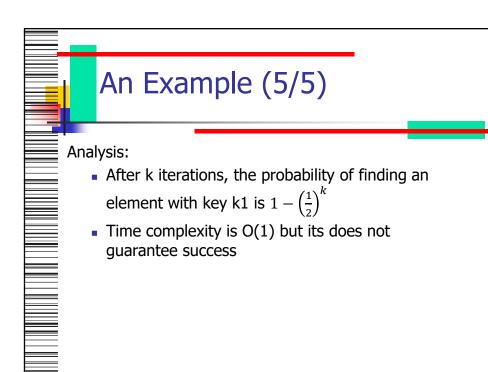
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An Example (4/5)

Monte Carlo algorithm

```
Algorithm findKey(L, k1)
Input: list L, key k1
Output: an element in L with key k1
{
    i=0;
    repeat
        randomly select e∈L;
        i++;
    until key(e)=k1 or i=m;
    return e;
}
```

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- Given an integer k and n elements x₁, x₂, ..., x_n, taken from a total order, find the k-th smallest element in this set.
- Of course, we can sort the set in O(n log n) time and then index the k-th element.

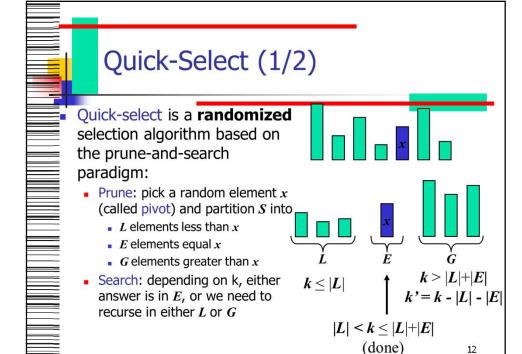
k=3 $7 \ 4 \ 9 \ \underline{6} \ 2 \rightarrow 2 \ 4 \ \underline{6} \ 7 \ 9$

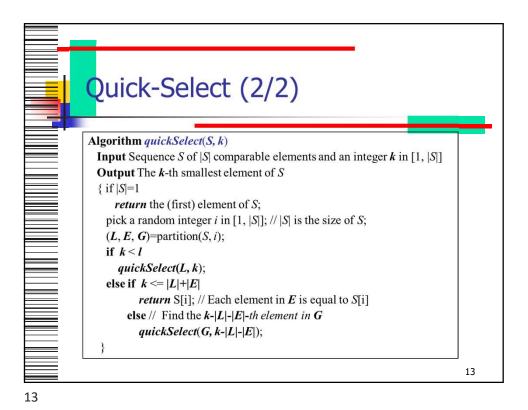
Can we solve the selection problem faster?

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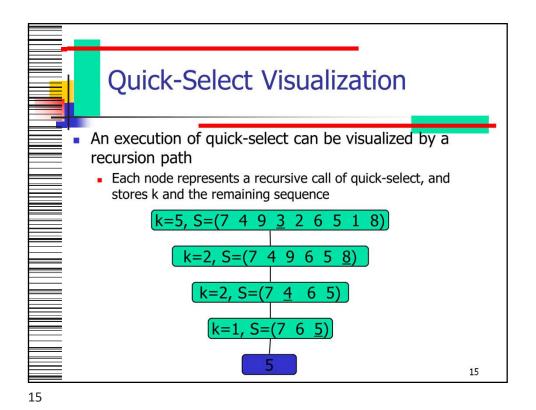
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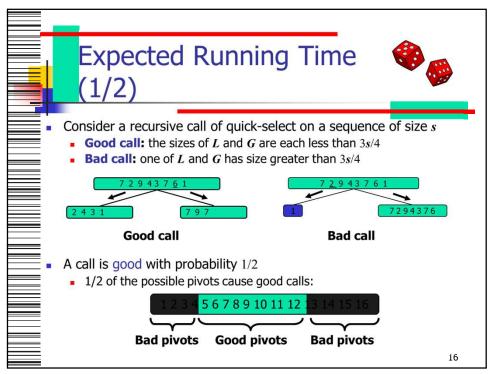
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Partition Algorithm partition(S, i)**Input** sequence S, index i of the piv We partition an input Output subsequences L, E, G of the sequence as in the quick-sort elements of S less than, equal to, algorithm: or greater than the pivot, resp. We remove, in turn, each { L, E, G = empty sequences; element y from S and x = S[i];We insert y into L, E or G, while $(\neg S.isEmpty())$ depending on the result of $\{ y = S.remove(S.first()); \}$ the comparison with the if (y < x)Each insertion and removal is L.insertLast(y); at the beginning or at the else if (y=x)end of a sequence, and E.insertLast(y); hence takes O(1) time else //y > xThus, the partition step of G.insertLast(y); } quick-select takes O(n) time return L, E, G;

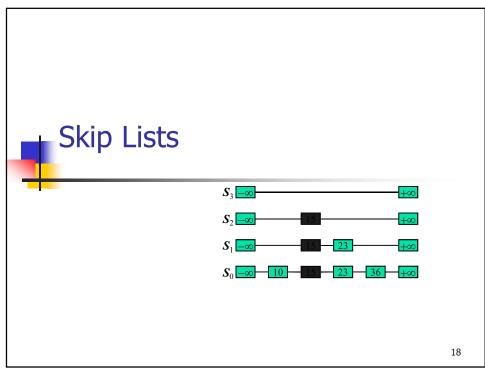


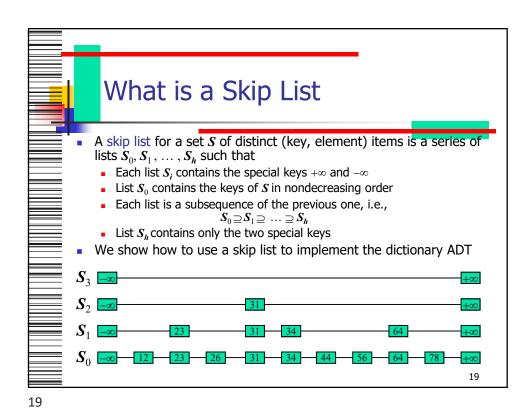


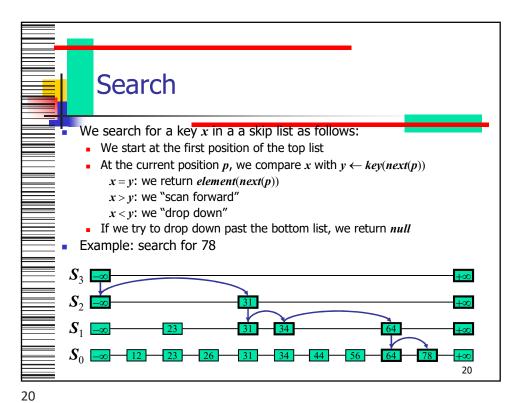
Expected Running Time (2/2)

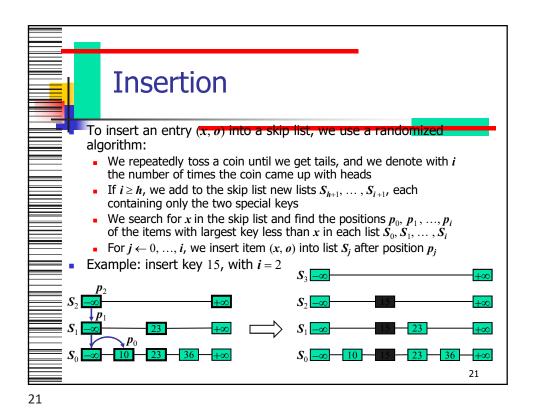


- Probabilistic Fact #1: The expected number of coin tosses required in order to get one head is two
- Probabilistic Fact #2: Expectation is a linear function:
 - E(X+Y) = E(X) + E(Y)
 - E(cX) = cE(X)
- Let T(n) denote the expected running time of quick-select.
- By Fact #2,
 - $T(n) \le T(3n/4) + bn^*$ (expected # of calls before a good call)
- By Fact #1,
 - $T(n) \le T(3n/4) + 2bn$
- That is, T(n) is a geometric series:
 - $T(n) \le 2bn + 2b(3/4)n + 2b(3/4)^2n + 2b(3/4)^3n + \dots$
- So T(n) is O(n).
- We can solve the selection problem in O(n) expected time.







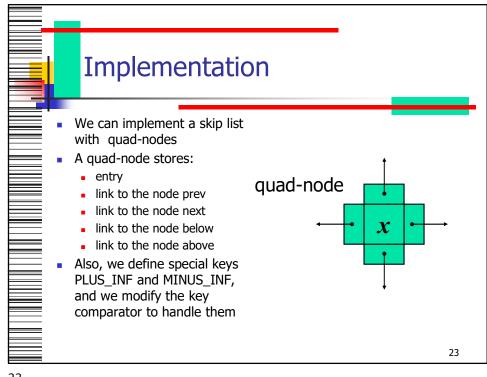


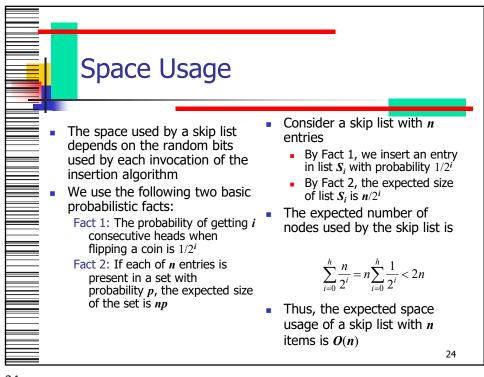
Deletion

To remove an entry with key x from a skip list, we proceed as follows:

• We search for x in the skip list and find the positions $p_0, p_1, ..., p_i$ of the items with key x, where position p_j is in list S_j • We remove positions $p_0, p_1, ..., p_i$ from the lists $S_0, S_1, ..., S_i$ • We remove all but one list containing only the two special keys

• Example: remove key 34 $S_0 = S_1 = S_2 = S_2 = S_3 = S_$







Height

- The running time of the search an insertion algorithms is affected by the height h of the skip list
- We show that with high probability, a skip list with nitems has height $O(\log n)$
- We use the following additional probabilistic fact:
 - Fact 3: If each of *n* events has probability p, the probability that at least one event occurs is at most np

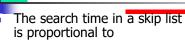
- Consider a skip list with n entires
 - By Fact 1, we insert an entry in list S_i with probability $1/2^i$
 - By Fact 3, the probability that list S_i has at least one item is at most $n/2^i$
- By picking $i = 3\log n$, we have that the probability that $S_{3\log n}$ has at least one entry is at most

$$n/2^{3\log n} = n/n^3 = 1/n^2$$

Thus a skip list with n entries has height at most $3\log n$ with probability at least $1 - 1/n^2$

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Search and Update Times



- the number of drop-down steps, plus
- the number of scan-forward steps
- The drop-down steps are bounded by the height of the skip list and thus are $O(\log n)$ with high probability
- To analyze the scan-forward steps, we use yet another probabilistic fact:

Fact 4: The expected number of coin tosses required in order to get tails is 2

- When we scan forward in a list, the destination key does not belong to a higher list
 - A scan-forward step is associated with a former coin toss that gave tails
- By Fact 4, in each list the expected number of scanforward steps is 2
- Thus, the expected number of scan-forward steps is $O(\log n)$
- We conclude that a search in a skip list takes $O(\log n)$ expected time
- The analysis of insertion and deletion gives similar results

