

Probing Strategies• Linear Probing

- primary clustering
- long runs of filled slots

$$h(k, i) = (h(k, 0) + i) \bmod m$$

• Quadratic Probing

$$h(k, i) = (h(k, 0) + i^2) \bmod m$$

• Double hashing

$$h(k, i) = (h_1(k) + i h_2(k)) \bmod m$$

→ excellent method by statistical studies

Usually pick $m = 2^r$ and $h_2(k)$ odd

Analysis of open addressing

* Assumption of uniform hashing

Each key is equally likely to have any one of the $m!$ permutations as its probe sequence independent of other keys

Theorem: $E[\# \text{ probes}] \leq \frac{1}{1-\alpha}$ if $\alpha < 1$
 (ie) $n < m$

Pf: (unsuccessful search)

1 probe is always necessary

with prob n/m , collision \Rightarrow 2nd probe necessary

" " $n-1/m-1$, " \Rightarrow 3rd probe necessary

" " $n-2/m-2$, " \Rightarrow 4th probe necessary

Check $\frac{n-i}{m-i} \leq \frac{n}{m} = \alpha$ for $i = 1, 2, \dots, m-1$

\therefore expected # of probes

$$= 1 + \frac{n}{m} \left(1 + \frac{n-1}{m-1} \left(1 + \frac{n-2}{m-2} \left(1 + \frac{n-3}{m-3} \left(1 + \frac{n-4}{m-4} \left(\dots \right) \right) \right) \right) \right) \right)$$

$$\leq 1 + \alpha (1 + \alpha (1 + \alpha (\dots (1 + \alpha) \dots)))$$

$$\leq 1 + \alpha + \alpha^2 + \alpha^3 + \dots = \frac{1}{1-\alpha}$$

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if $\alpha < 1$ is a constant,

$$\Rightarrow O(1) \text{ probes}$$

Ex: Table 50% full $\Rightarrow \alpha = 1/2 \Rightarrow 2 \text{ probes}$

90% full $\Rightarrow \alpha = 9/10 \Rightarrow 10 \text{ probes}$

→ Some Background on Probability Theory (1.3.4)

→ Sample space S

* outcomes of an experiment

* can be finite or infinite

→ probability space

= Sample space + prob fn

→ event E is a subset of S , $E \subseteq S$

$$* \Pr(\text{event } E) = \sum_{\omega \in E} \Pr(\omega)$$

→ Two events A & B are independent if

$$\Pr(A|B) = \Pr(A)$$

$$\approx \Pr(A \cap B) = \Pr(A) \Pr(B)$$

Random Variables

(4)

$$X : S \rightarrow \mathbb{R}$$

$$E(X) = \sum_x x \Pr(X=x)$$

$$= \sum_x \Pr(X \geq x)$$

→ Linearity of Expectation

$$Z = X + Y$$

$$\rightarrow E(Z) = E(X) + E(Y)$$

If X & Y are independent,

$$E(XY) = E(X)E(Y)$$

Remark

Hash functions

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We have seen compression maps till now
where we assume that the keys come from an
underlying $V = \{0, 1, 2, \dots, m-1\}$

How do we encode general keys to numbers?

→ hash code