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Find an efficient implementation of a dynamic collection of elements with unique keys
Supported Operations: Insert, Search and Delete
The keys belong to a universal group of keys, $U=\{1 \ldots M\}$.
Direct Address Tables - An array of size m. An Element with key $i$ is mapped to cell i.
Time-complexity: $\mathrm{O}(1)$ $\qquad$
What might be the problem?
If the range of the keys is much larger than the number of elements a lot
$\qquad$ of space is wasted
might be too large to be practical. $\qquad$
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## Hash Tables

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In a hash table, we allocate an array of size $m$, which is much smaller than $|\mathrm{U}|$ (the set of keys).
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We use a hash function $h()$ to determine the entry of each $\qquad$
key.
What property should a good hash function have?
The crucial point: the hash function should "spread" the keys of U equally among all the entries of the array. $\qquad$
An example of a hash function:

$$
h(k)=k \bmod m
$$

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## Hash choices

- Assume we want to hash strings for a country club managing software, each string contains full name (e.g. "Parker Anthony").
- Suggested hash function : the ascii number of the first symbol in the string (e.g. h("Parker Anthony") $=$ ascii of ' $h$ ' $=80$.
- Is this a good hash function?
- No, since it is likely that many families register together. This means that we will have many collisions even with few entries.
-We would like to have something "random looking".
-The exact meaning of this will be explained next week, however, it is important to think of this point when choosing a hash function.
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## How to choose hash functions

Can we choose a hash function that will spread the keys uniformly between the cells?
Unfortunately, we don't know in advance what keys will be received, so there can always be a 'bad input'.
We can try to find a hash function that usually maps a small number of keys to the same cell.
Two common methods:

- The division method
- The multiplication method


## The division method

- Denote by $m$ the size of the table.

The division method is : $h(k)=k \bmod m$ $\qquad$

- What is the value of $m$ we should choose?
- For example:
$|\mathrm{U}|=2000$, if we want each search to take three operations (on the average), then choose a number close to 2000/3 $\approx 666$.
- For technical reasons it is preferable to choose m as a prime (for a prime $\mathrm{p} \mathrm{Z}_{\mathrm{p}}$ is a field).
- In our example we choose $\mathrm{m}=701$.


## The multiplication method

- The multiplication method:
- Multiply k by a constant $0<\mathrm{A}<1$.
- Take the fractional part of $k A$ $\qquad$
- Multiply by m.
- Formally, $\quad h(k)=\lfloor m(k A \bmod 1)\rfloor$
- The multiplication method does not depends as much on $m$ since A helps randomizing the hash function.
- Which values of A are good choices? Which are bad choices?


## The multiplication method

- A bad choice of A, example:
- if $\mathrm{m}=100$ and $\mathrm{A}=1 / 3$, then
- for $\mathrm{k}=10, \mathrm{~h}(\mathrm{k})=33$,
- for $\mathrm{k}=11, \mathrm{~h}(\mathrm{k})=66$,
- And for $\mathrm{k}=12, \mathrm{~h}(\mathrm{k})=99$.
- This is not a good choice of A, since we'll have only three values of $h(k)$...
- The optimal choice of A depends on the keys themselves.
- Knuth claims that $A \approx(\sqrt{5}-1) / 2=0.6180339887 \ldots$ is likely to be a good choice.

What if keys are not numbers?

- The hash functions we showed only work for numbers.
- When keys are not numbers, we should first convert them to numbers.
- How can we convert a string into a number?
- A string can be treated as a number in base 256 .
- Each character is a digit between 0 and 255.
- The string "key" will be translated to
$\left((\text { int })^{\prime} k^{\prime}\right) \times 256^{2}+\left((\mathrm{int})^{\prime} e^{\prime}\right) \times 256^{1}+\left((\mathrm{int})^{\prime} y^{\prime}\right) \times 256^{0}$


## Translating long strings to numbers

- The disadvantage of the conversion is:
- A long string creates a large number.
- Strings longer than 4 characters would exceed the capacity of a 32 bit integer.
- How can this problem be solved when using the division method?
- We can write the integer value of "word" as $(((\mathrm{w} * 256+\mathrm{o}) * 256+\mathrm{r}) * 256+\mathrm{d})$

When using the division method the following facts can be used: $-(\mathrm{a}+\mathrm{b}) \bmod \mathrm{n}=((a \bmod \mathrm{n})+\mathrm{b}) \bmod \mathrm{n}$
$-\left(a^{*} b\right) \bmod n=((a \bmod n) * b) \bmod n$. $\qquad$
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## Translating long strings to numbers

- The expression we reach is:
- ((((( $\mathrm{w} * 256+0) \bmod \mathrm{m}) * 256)+\mathrm{r}) \bmod \mathrm{m}) * 256+\mathrm{d}) \bmod \mathrm{m}$
- Using the properties of mod, we get the simple alg.:
int hash(String s, int m)
int $h=s[0]$
for ( i=1 ; i<s.length ; i++) $h=((h * 256)+s[i]))$ mod m
return h $\qquad$
- Notice that $h$ is always smaller than $m$.
- This will also improve the performance of the algorithm.


## The Birthdays Paradox

- There are 28 people in the room. Assuming that birthdays are distributed uniformly over the year, how many pairs of people who share birthdays will you expect?
- Surprisingly, the expected number of such pairs is approximately 1.
- Analysis: There are n people. The probability that two people share birthdays is $1 / 365$.
- The number of pairs is $\binom{n}{2}=\frac{n(n-1)}{2}$
- So - the expected number of pairs with the same birthday is $\binom{n}{2}=\frac{n(n-1)}{2 * 365}$


## Collisions

- Can several keys have the same entry?
- Yes, since $U$ is much larger than $m$.
- Collision - several elements are mapped into the same cell.
- Similar to the birthdays paradox analysis.
- When are collisions more likely to happen?
- When the hash table is almost full.

We define the "load factor" as $\alpha=n / m$.

- n - the number of keys in the hash table
- $m$ - the size of the table
- How can we solve collisions?


## Chaining

- There are two approaches to handle collisions:
- Chaining.
- Open Addressing
- Chaining:
- Each entry in the table is a linked list.
- The linked list holds all the keys that are mapped to this entry.
- Search operation on a hash table which applies chaining takes $O(1+\alpha)$ time.


## Chaining

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- This complexity is calculated under the assumption of uniform hashing.
Notice that in the chaining method, the load factor may be greater than one.
Analysis:
Insert - O(1)
Unsuccessful search - calculating the hash value - $\mathrm{O}(1)$,
Scanning the linked-list - expected $\mathrm{O}(\alpha)$. Total: $\mathrm{O}(1+\alpha)$.
Successful search - Let x be the i'th element inserted to the hash table.
When searching for x , the expected number of scanned elements is $(\mathrm{i}-1) / \mathrm{m}$.
For any element - the expected number of scanned elements is $\mathrm{O}(\alpha)$. Total: $\mathrm{O}(1+\alpha)$
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## Open addressing

- In this method, the table itself holds all the keys.
- We change the hash function to receive two parameters:
- The first is the key.
- The second is the probe number.
- We first try to locate $h(k, 0)$ in the table.
- If it fails we try to locate $h(k, 1)$ in the table, and so on.
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## Open addressing

- It is required that $\{\mathrm{h}(\mathrm{k}, 0), \ldots, \mathrm{h}(\mathrm{k}, \mathrm{m}-1)\}$ will be a permutation of $\{0, . ., m-1\}$.
- After m-1 probes we'll definitely find a place to locate k (unless the table is full).
- Notice that here, the load factor must be smaller than one.
- There is a problem with deleting keys. What is it? How can it be solved?


## Open addressing

- While searching key i and reaching an empty slot, we don't know if:
- The key i doesn't exist in the table.
- Or, key i does exist in the table but at the time key i was inserted this slot was occupied, and we should continue our search.

What functions can we use to implement open addressing?

- We will discuss two ways :
- linear probing
- double hashing $\qquad$
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## Open addressing

- Linear probing - $h(k, i)=(h(k)+i) \bmod m$
- The problem: primary clustering.
- If several consecutive slots are occupied, the next free slot has high probability of being occupied.
- Search time increases when large clusters are created.
- The reason for the primary clustering stems from the fact that there are only m different probe sequences.
- How can we change the hash function to avoid clusters?


## Open addressing

- Double hashing -

$$
h(k, i)=\left(h_{l}(k)+i h_{2}(k)\right) \bmod m
$$

- Better than linear probing.
- It's best if for all $k$ we have that $h_{2}(k)$ does not have a common divisor with $m$.
$-w^{3}$ hat way different probe sequences!
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## Performance

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- Insertion and unsuccessful search : $1 /(1-\alpha)$ probes on average.
- Intuition: We always check at least one cell. In probability $\alpha$ this cell is occupied. In probability $\approx \alpha^{2}(n / m *(n-1) / m)$ the next cell is also occupied
- Expected probes number $\approx 1+\alpha+\alpha^{2}+\alpha^{3}+\ldots=1 /(1-\alpha)$.
- A successful search: $(1 / \alpha) \ln (1 /(1-\alpha))$ probes on average
- Idea: When searching for $x$, the probes number is the same as when the element was inserted (depends on the number of elements that were inserted before x ). How to get to the formula look in CLR
- For example:
- If the table is $50 \%$ full then a search will take about 1.4 probes on average.
- If the table $90 \%$ full then the search will take about 2.6 probes on average.
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## Example for Open Addressing

- Lets look at an example:
- First assume we use linear probing
- insert the numbers: $4,28,6,38,26$ with $\mathrm{m}=11$
and $h(k)=k \bmod m$.
[] [ ] [ ] [ ] [4] [38] [28] [6] [26] [ ] [ ]
$\begin{array}{lllllllllll}0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10\end{array}$
- Note the clustering effect on 26 !


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## Example for Open Addressing

- Now lets try double hashing, again $m=11$
- Use:
$h_{1}(k)=k \bmod m$
$h_{2}(k)=1+k \bmod (m-1)$.
$h(k)=\left(h_{1}(k)+i * h_{2}(k)\right) \bmod m$ $\qquad$
- Insert numbers: 4, 28, 6, 38, 26
[26] [ ] [6] [ ] [4] [38] [28] [ ] [ ] [ ] [ ]
$\begin{array}{llllllllll}0 & 1 & 2 & 3 & 4 & 5 & 6 & 8 & 9 & 10\end{array}$ $\qquad$
- The clustering effect disapeard.
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## When should hash tables be used

- Hash tables are very useful for implementing dictionaries if we don't have an order on the elements, or we have order but we need only the standard operations.
- On the other hand, hash tables are less useful if we have order and we need more than just the standard operations.
- For example, last(), or iterator over all elements, which is problematic if the load factor is very low.
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## When should hash tables be used

- We should have a good estimate of the number of elements we need to store
- For example, the huji has about 30,000 students each year, but still it is a dynamic d.b.
- Re-hashing: If we don't know a-priori the number of elements, we might need to perform re-hashing, increasing the size of the table and re-assigning all elements.

