Hashing

- Average case for operations on a search tree:
 O(log N)
- Question: Can we do better at least for some of the operations?
- Goal: insertion, deletion and find in O(1)
- Solution: Hash table
- Hashing does not support operations that require ordering information: findMin, findMax and print in sorted order

Hash Table

 Hash Table: an array of fixed size (TableSize) containing the items

```
A[ 0 ], A[ 1 ], ..., A[ TableSize – 1 ]
```

- Item = key + data member.
- Search operations are performed on the keys.
- Each key is mapped into an integer between 0 and TableSize-1 and placed in the appropriate cell.
- Mapping is done using a hash function.
- Ideal hash function:
 - Easy to compute
 - Maps two distinct keys into different cells (impossible!).
- Good hash function: distributes the keys evenly among the cells.

Hash Table

Example:

$$h(John) = 3$$

 $h(Phil) = 4$

$$h(Dave) = 6$$

$$h(Mary) = 7$$

Problems:

- Choosing the hash function.
- Resolving collisions (when two keys hash to the same value).
- Deciding the table size.

0	
1	
2	
3	John 25000
4	Phil 31250
5	
6	Dave 27500
7	Mary 28200
8	
9	

Example:

```
h(key) = (key) \mod (TableSize)
```

- Good if keys are uniformly distributed.
- If TableSize = 10 and all keys ending in zero => bad choice
- Solution: TableSize = a prime number
- If key is a string => use the sum of ASCII values of the characters as input to h().

Example 1:

```
int hash( const string & key, int tableSize )
      int hashVal = 0;
      for( int i = 0; i < \text{key.length}(); i++)
         hashVal += \text{key}[i];
      return hashVal % tableSize;
If table size is large => keys are not distributed well.
Example: TableSize = 10,007 ( prime number)
         8 characters keys
         ASCII value of a character < 127
=> key values between 0 and 127*8 = 1,016
```

=> not an equitable distribution

Example 2:

```
int hash( const string & key, int tableSize )
{
    return ( key[ 0 ] + 27* key[ 1 ] + 729 * key [ 2 ]) %
        tableSize;
}
```

- Assumes key length >= 3
- Example: TableSize = 10,007 (prime number)
 Number of combinations of 3 letters => 26³ = 17,576
 English is not random => the number of different combination of 3 letters is 2,851
 If no collision => only 28 % of the table is used
 => not appropriate as a hash function if the size of the table is large

Example 3:

- Involves all characters in a key
- h(Key) = Key[KeySize 1] + Key[KeySize 2]*37 + ...+ Key[0]*37 KeySize -1
- It is based on Horner's rule for evaluating polynomials:

$$H = k_0 + 37*k_1 + 37^2*k_2 = ((k_2)*37 + k_1)*37 + k_0$$

Good distribution, but slow for large keys.

```
hashVal %= tableSize;
if( hashVal < 0 ) //overflow may introduce a negative number
hashVal += tableSize;
```

Complement of 2 representation:

4 bit signed integer: 1 sign bit | 3 bits

CC1: CC2:

+0:0000 + 0:0000 invert the bits then add 1

.

. $+7: 0111 \rightarrow 1000 \rightarrow 1001 \rightarrow -7$

+7: 0111 +7: 0111

-7: 1000 -8: 1000

. .

. .

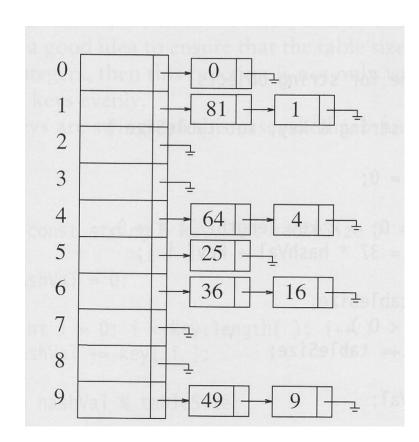
-0: 1111 -1: 1111

Collision Resolution

- Collision: given two keys k_1 and k_2 we have a collision at slot x if $h(k_1) = h(k_2) = x$
- How to deal with collisions?
- Solutions:
 - separate chaining
 - open addressing
- Procedure for finding an item with a key value K:
 - Compute the table location h(K).
 - Starting with slot h(K) locate the item containing key K using (if necessary) a collision resolution policy.

Separate Chaining

- Idea: keep a list of all elements that hash to the same value.
- Assumption: h(x) = x mod 10 and the keys are the first ten perfect squares.
- Find(k):
 - H(k) = x
 - Search list x.
- Insert(k):
 - H(k) = x
 - Search list x, if element is not on the list insert the element at the front.



Separate Chaining

```
template <typename HashedObj>
class HashTable
{
      public:
          explicit HashTable( int size = 101 );
          bool contains( const HashedObj & x ) const;
          void makeEmpty( );
          void insert( const HashedObj & x );
          void remove( const HashedObj & x );
      private:
          vector<list<HashedObj> > theLists; // The array of Lists
          int currentSize:
          void rehash( );
          int myhash( const hashedobj & x ) const;
};
int hash( const string & key );
int hash( int key );
```

myhash & makeEmpty

```
int myhash( const HashedObj & x ) const
{
             int hashVal = hash(x);
         hashVal %= theLists.size();
         if( hashVal < 0)
             hashVal += theLists.size ();
         return hashVal;
void makeEmpty( )
{
         for( int i = 0; i < theLists.size(); i++)
                 theLists[ i ].clear( );
}
```

contains & remove

```
bool contains( const HashedObj & x ) const
      const list<HashedObj> & whichList = theLists[ myhash( x ) ];
      return find( whichList.begin( ), whichList.end( ), x) != whichList.end( );
}
bool remove( const HashedObj & x )
      list<HashedObj> & whichList = theLists[ myhash( x ) ];
      list<HashedObj>::iterator itr = find( whichList.begin( ), whichList.end( ),
x );
      if( itr == whichList.end( ) )
         return false;
      whichList.erase( itr );
      --currentSize;
      return true;
```

Insert

```
bool insert( const HashedObj & x )
{
    list<HashedObj> & whichList = theLists[ myhash( x ) ];
    if( find( whichList.begin(), whichList.end(), x ) != whichList.end() )
        return false;

    whichList.push_back( x );

// Rehash
    if( ++currentSize > theLists.size( ) )
        rehash( );
    return true;
}
```

Running Time Analysis

- What is the running time for find?
- Find:
 - Time to evaluate the hash => O(1)
 - Time to traverse the list => O(?)
- Load factor λ = N/M

N = number of items; M = hash table size

- Unsuccessful search: T(N) is given by the number of average elements in the lists => O(N/M) => $O(\lambda)$
- Successful search: $T(N) = 1 + \lambda/2 => O(\lambda)$
- Find, insert and remove $=> O(\lambda)$
- Size of the table is not important only the load factor is.
- Suggestions:
 - Make the size of the table as large as the number of elements $=>\lambda=1$
 - Table size = prime number