



# Chapter 3

# Vectors

# 3 vectors

A vector (field, array) manages a fixed number of elements of a uniform type.

Accessing an element via an integer index (the position in Vector)

Access effort for all elements is constant



## Methods

Hashing: Datenorganisationsform

Sorting: Data reorganization methods

## 3.1 Dictionary

A *dictionary* is a data structure in which the individual elements each consist of a *key* and an *information part* (info).

Supports only the simple operations of insert, delete and search provided.

examples are

Dictionaries, name lists, inventory lists, etc.

Vectors are well suited to creating dictionaries

Structure

key0	key1	key2	...	after-1
info0	info1	info2		info-1

# Examples

dictionary

(detail)

Key	Information
computable	predictable
computation	calculation
compute	(calculate
computer	calculator

Inventory list

Key	Information
1	CPU
4	Screen
17	Keyboard
25	Maus



# The operation

Construct

Creating an empty dictionary

IsEmpty

Query for empty dictionary

Insert

Inserting an element

Delete

Delete an element

LookUp

Query an element via its key and deliver it  
Information

# Klasse Dictionary

```
class Dictionary {  
    private:  
        node { KeyType Key; InfoType Info; }; node *Dict;  
  
        int NumberElements;  
  
    public:  
        Dictionary(int max) {  
            Dict = new node[max];  
            NumberElements = 0;  
        }  
        ~Dictionary() { delete[] Dict; } void  
        Insert(KeyType k, InfoType I); void Delete(KeyType  
        k);  
        InfoType LookUp(KeyType k);  
};
```

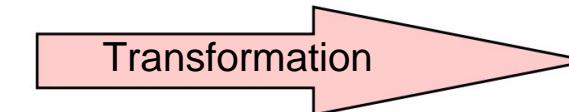
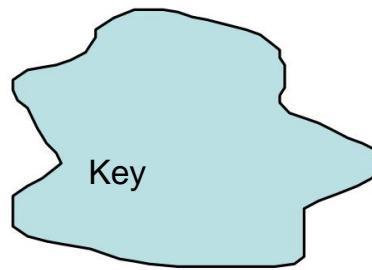


## 3.2 Hashing

*Hashing* is a method of directly assigning elements in a table address by replacing their key values with arithmetic Transformations can be translated directly into table addresses (indexes).

In other words, the index (the position in the table) of a Elements is calculated from the key value of the element itself.

Key  $\ddot{y}$  Address (index)



Table



# Hashing, general

## Approach

Set K of n keys  $\{k_0, k_1, \dots, k_{n-1}\}$

*Hash table T* of size m

Boundary condition:  $n \gg m$

(Number of possible key values much larger than places in the table)

## Transformation

*Hash function h*

$$h: K \rightarrow \{0, 1, \dots, m-1\}$$

For each j, the key  $k_j$  should be stored at location  $h(k_j)$  in the table. The value  $h(K)$  is called the hash value of K.

Choice of hash function (actually) arbitrary!

# Hash function

## Desired properties

Easy and quick to calculate.

Elements are evenly distributed throughout the table.

All positions in the table have equal probability calculated.

### Examples

#### Modulo education, key mod m

You should make sure that m is a prime number, otherwise it can happen

Key transformations (strings) lead to cluster formations (accumulations) through common divisors.

#### Bitstring-Interpretation

Parts of the binary representation of the key value are used as a hash value used.

#### Transformation tables

# Example hashing

Inventory list

List size 7

Hash function  $h(k) = k \bmod 7$

Records

Key	Information
1	CPU
4	Screen
17 25	Keyboard
	Maus

1 against 7 = 1



17 against 7 = 3



4 against 7 = 4



0	
1	1 CPU
2	17 Keyboard
3	4 Screen
4	
5 6	

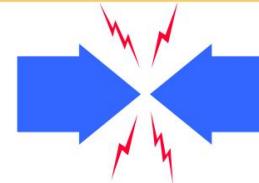
$25 \bmod 7 = 4$  → Position in the table already occupied

*collision*



# collision

A *collision* occurs when two or more key values are mapped (*hashed*) to the same table position .



This means that for 2 keys  $k_i$  ,  $k_j$  , with  $k_i \neq k_j$  ,  
 $h(k_i) = h(k_j)$  .

However, this situation is to be expected since there are many more possible keys than table positions ( $n \gg m$  as a boundary condition).

The hash function  $h$  is generally not injective, i.e. it does not follow from  $h(x) = h(y)$ . necessarily  $x=y$ .

# Collision treatment

A collision triggers necessary *collision handling*.



Ie, the colliding key value must have a  
Alternative location can be found.

Collision handling measures are usually quite complex and affect the  
efficiency of hashing processes.

the collision path is defined by the fallback locations of all  
elements that have been "hashed" to the same location

We now consider 2 simple collision treatments

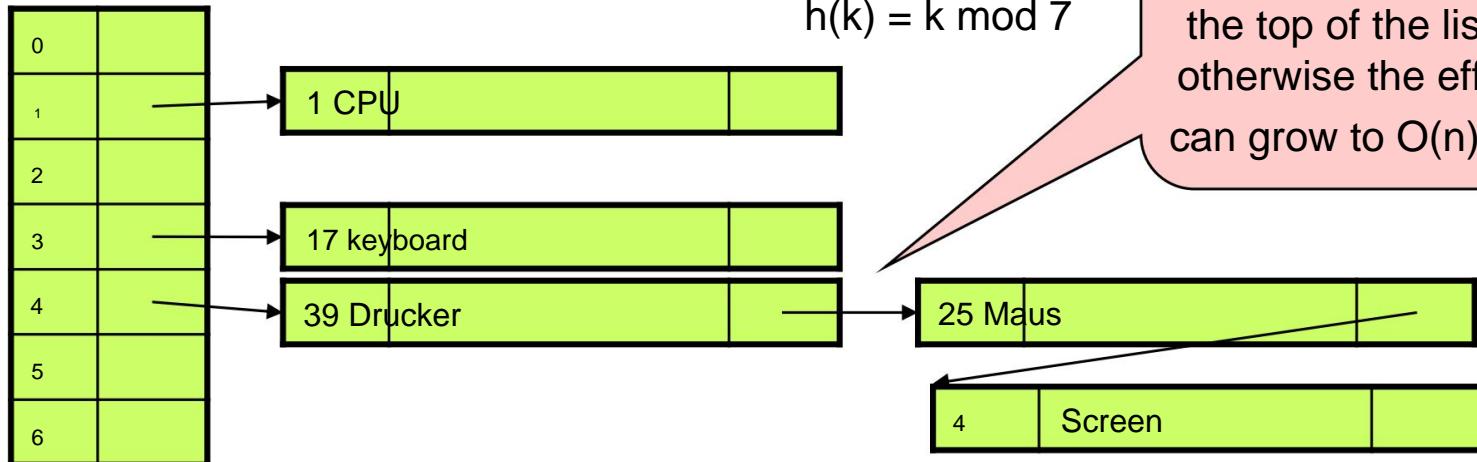
*Separate Chaining und Double Hashing*

### 3.2.1 Separate Chaining

With separate (simple) chaining, collisions are handled accomplished with linear lists.

Colliding key values are stored in a linear overflow chain (list) starting from the original hashed table space.

Example:



The entry of the element (39, printer) leads to an extension of the overflow chain.

### 3.2.2 Double Hashing

With *double hashing*, in the event of a collision, a second one of  $h$  independent hash function used to determine an alternative position. Overflow to position  $a_0 = h(k)$

Determination of an alternative position with collision function  $g : K \rightarrow \{1, \dots, m-1\}$ , e.g.  $a_{i+1} = (a_i + g(k)) \bmod m$

Example

$$h(k) = k \bmod 7$$

$$g(k) = \text{last digit of } k \text{ times 3}$$

$$a_0 = 25 \bmod 7 = 4$$

$$a_1 = (4 + (5 \cdot 3)) \bmod 7 = 5$$

$$a_0 = 39 \bmod 7 = 4$$

$$a_1 = (4 + (9 \cdot 3)) \bmod 7 = 3$$

$$a_2 = (3 + (3 \cdot 3)) \bmod 7 = 2$$

 Collision

 K

 K

0	
1	1 CPU
2	39 Drucker
3	17 keyboard
4	4 Screen
5	25 Maus
6	



# Linear Probing

Special case of double hashing

Collision handling: The next free space is used

When you reach the end of the table, you look at the beginning of the table further

Hashfunktionen:  $a_0 = h(k)$ ,  $a_{i+1} = (a_i + g(k)) \bmod m$

$$g(k) = 1$$

Example

25 against 7 = 4



(4+1) against 7 = 5

39 against 7 = 4



(4+1) against 7 = 5

(5+1) against 7 = 6

0	
1	1 CPU
2	
3	17 keyboard
4	4 Screen
5	25 Maus
6	39 Drucker

# Searching and deleting in hash tables

## Seek

As long as a target address is occupied by an element and the searched key value has not yet been found, the search must continue using the collision path.

Separate Chaining: Tracking the linear list

Double Hashing: Finding all possible locations of the collision function.

## Delete

Separate Chaining: Removing the list item

Double hashing: Positions where an element was deleted may need to be marked with a special value (“free again”) so that the corresponding collision path for the remaining elements is not interrupted.



# Characteristics

Generally for hashing processes

- + Effort when accessing an element is constant in the best case,  $O(1)$ , simple evaluation of the hash function.
- Collision handling is complex, a full hash table leads to many collisions, so (rule of thumb) never fill more than 70%.
- No access in sort order possible.

Separate chaining +

dynamic data structure, any number of elements.

- Storage space intensive (additional pointer variable).

Double hashing +

optimal storage space utilization.

- Static data structure, table size specified.
- Collision handling complex, “clear again” marking when deleting.

# Analyse Hashing

Data management

- Insert and delete supported

Data volume

- limited

- depending on the size of the existing hash table

Models

- Main memory oriented

- Support for simple operations, no range queries, no sort order



# Analyse Hashing (2)

Storage space	O(1)
Constructor	O(1)
access	O(1)
Insert	O(1)
Delete	O(1)

valid only for  
pure  
Hash process  
without collision  
handling

Please note: The effort involved in hashing depends heavily on the Collision method, often goes towards  $O(n)$  in the case of collisions

### 3.3 Dynamic hashing methods

Dynamic hashing procedures try in the event of collisions  
expand the original hash table and create

overflow areas (e.g. linear lists)

Enlargement of the primary area (original table)

Requires hash function modification

Approach: Families of hash functions

$h_1 : K \rightarrow \text{addr}_1$

...

$h_n : K \rightarrow \text{addr}_n$

Where  $|\text{addr}_1| < |\text{addr}_2| < \dots < |\text{addr}_n|$

| Goal: Switching from  $\text{addr}_i$  to  $\text{addr}_{i+1}$  requires minimal reorganization of the hash table (i.e.  
minimizing restoring of data)

# Hash functions scheme

Simple approach

$$|addr+1| = 2 * |addr|$$

$h(x)$  is a hash function

Let  $h_i(x)$  be the  $i$  “least significant bits” of  $h(x)$

Therefore applies

$$h_{i+1}(x) = h_i(x) \text{ or}$$

$$h_{i+1}(x) = h_i(x) + 2^i$$

Example:

$$\text{addr}_2 = 0 \dots 3,$$

$$\text{addr}_3 = 0 \dots 7,$$

$$h(x) = x$$

4 address changes

Value $h(x)$		binary	$h_2(x)$	$h_3(x)$
7	7	00111 11=3		111=7
8	8	01000 00=0 000=0		
9	9	01001 01=1		001=1
10	10	01010 10=2 010=2		
13	13	01101 01=1		101=5
14	14	01110 10=2 110=6		
23	23	10111 11=3		111=7
26	26	11010 10=2 010=2		

# *Dynamic hashing methods*

## „Two-Disk-Access“ Prinzip

Finding a key with a maximum of 2 disk accesses

## Hash process for external storage media

### *Linear Hashing*

Directoryless hashing method

Primary areas, overflow areas

### *Extendible Hashing*

Directory with binary expansion

Primary areas

### *Bounded Index Size Extendible Hashing*

Limited size directory

Binary expansion of blocks

### 3.3.1 Linear Hashing

W. Litwin 1980

#### Organization of elements in blocks (buckets)

Analogous to B-trees

Block size  $b$ , ie up to  $b$  can be in one address in the table

Elements are saved (=block)

#### Idea

When a block overflows (=  $b+1$ st element of a block inserted) (splitting), the primary and overflow areas are expanded by one block each

Primary scope by adding a block to the end of the hash table

Overflow area through linear list for individual blocks

Round number  $d$  stores the number of splits per block, there are simultaneously blocks that were split  $d$  times and blocks that were split  $d+1$  times

# *Splitting and searching*

## Splitting is carried out in rounds

A round ends when all the original N blocks (for round R)  
have been split

Blocks 0 to NextToSplit-1 are already split

Next block to split is defined by index NextToSplit

Current lap number is defined by d

## Search

Search for value x

```
a = hd(x); if(a < NextToSplit) a = hd+1(x);
```

d=2, NextToSplit=1, b=3

000					2=000010
01	8	17			8=001000
10	25	34	50	2	17=010001
11					25=011001
100	28				28=011100
					34=100010
					50=110010

# Insert

## Insert

Search block (hd or hd+1)

If block full

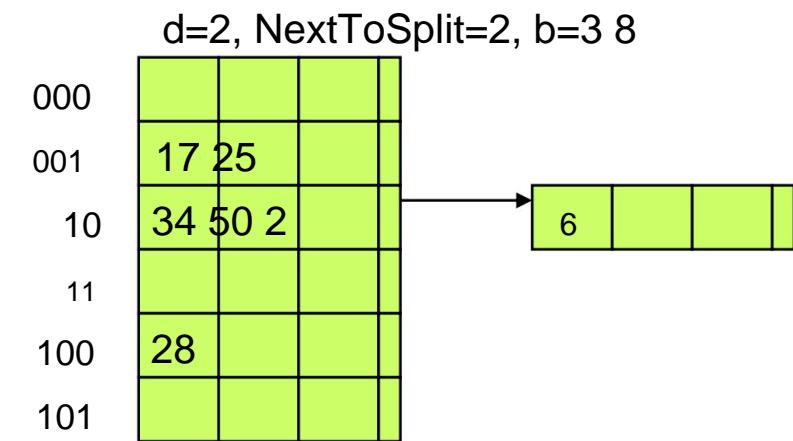
Add element into an overflow block

New element in the linear list

Split block NextToSplit and increase NextToSplit by 1

Example: Inserting 6 (=000110)

				$d=2$ , NextToSplit=1, b=3
000	8 8=001000	17 25 17=010001	2=000010	
01	34 50	2 25=011001		
10	28=011100	34=100010		
11		50=110010		
100	28			



# Example: linear hashing

Current  $hi(x) = h3(x)$  Insert: 16, 13, 21, 37 d=3,  
 NextToSplit=0, b=3 d=3, NextToSplit=0, b=3 8

000	8	16	17	25	17	25	34	50	2
001	50	2							
010									
011									
100	28								
101	5								
110	55								
111									

2=000010 28=011100  
 5=000101 34=100010  
 8=001000 50=110010  
 17=010001 55=110111  
 25=011001

000	34			
001				
010				
011				
100	28			
101	5	13	21	
110	55			
111				

16=010000  
 13=001101  
 21=010101  
 37=100101

Overflow block

0000	25			
001	34	50	2	
010				
011				
100	28			
101	5	13	21	
110	55			
111	8			

37				
37				

# Remarks

The round number d is increased once the splitting is done, ie

```
if(NextToSplit == 2d) { d++; NextToSplit=0; }
```

In practice, no linear address spaces but rather splitting them

primary area to multiple hash files per disk

Allocation system via administration blocks

### 3.3.2 Extendible Hashing

Hash process with index

R. Fagin, J. Nievergelt, N. Pippenger and HR Strong, 1979

Idea

If primary area becomes too small, enlarge by doubling

Primary area is managed via Index T

If a block overflows, split this block and manage the block  
by doubling the index

Index is much smaller than file, so doubling it is much cheaper

No overflow areas

# Structure and search

## Characteristics

Each index entry references exactly a data block

Each data block has  $\geq b$  entries and is made up of exactly  $2^k$  with  $k$  out of  $N_0$   
 Index entries referenced

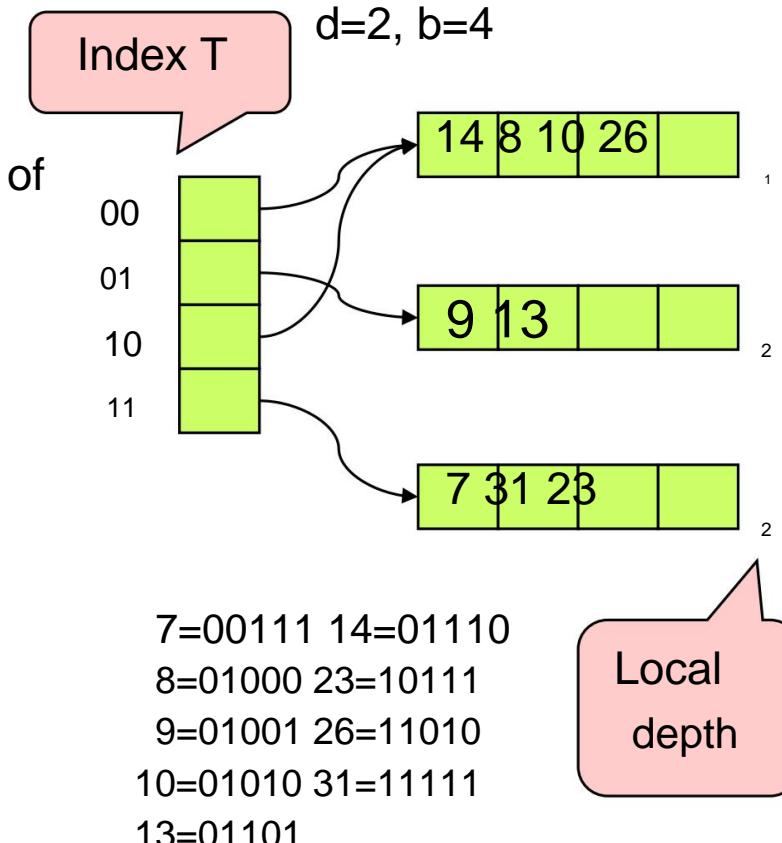
The number of index entries that contain the Referencing block is known locally in the block

Managed by a local depth  $t \leq d$

(Opposite global depth  $d$  for the entire hash table)

Number of referencing the block

Index entries corresponds to  $2^{d-t}$



## Search

Element  $x$  im Block  $T[hd(x)]$

# Insert

Two cases must be distinguished if a block overflows

- $t < d$ : multiple index entries reference this block
- $t = d$ : one index entry references this block

Fall 1:  $t < d$

e.g. B. Insert 6

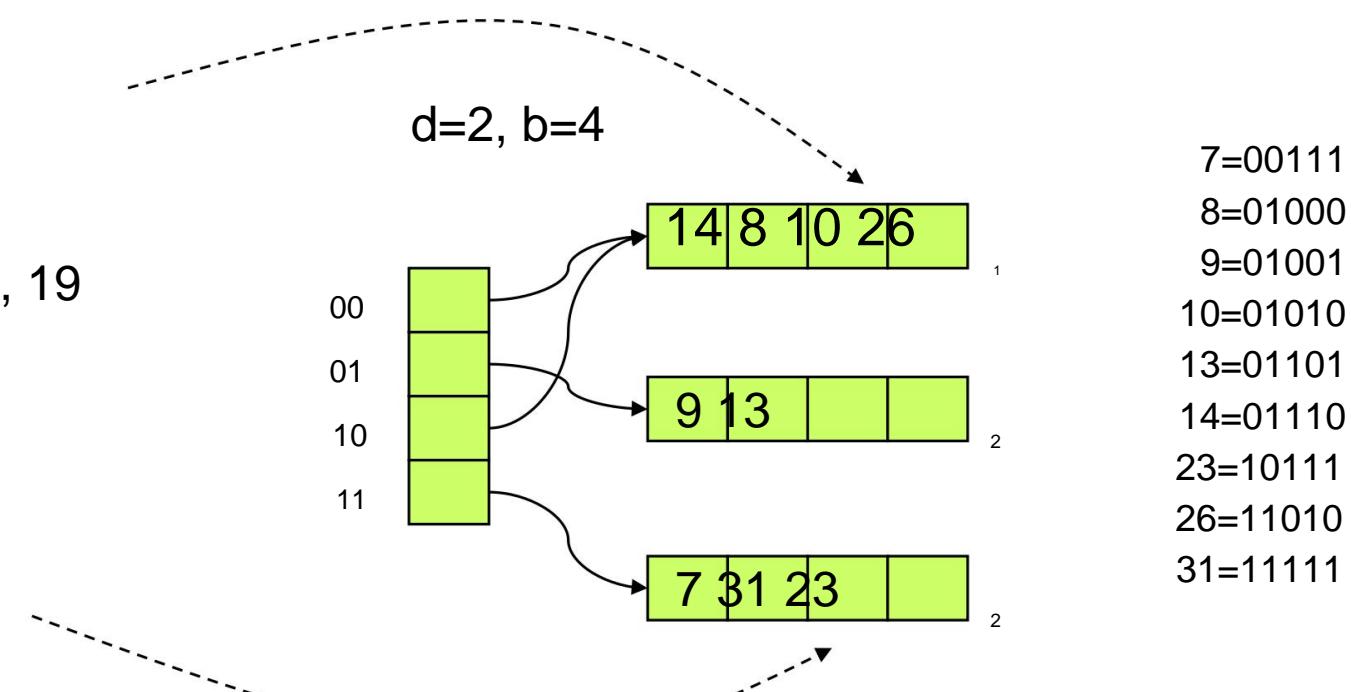
6=00110

Fall 2:  $t = d$

e.g. B. Insert 15, 19

15=01111

19=10011



# Insert case 1: $t < d$

Try to do the split without index expansion

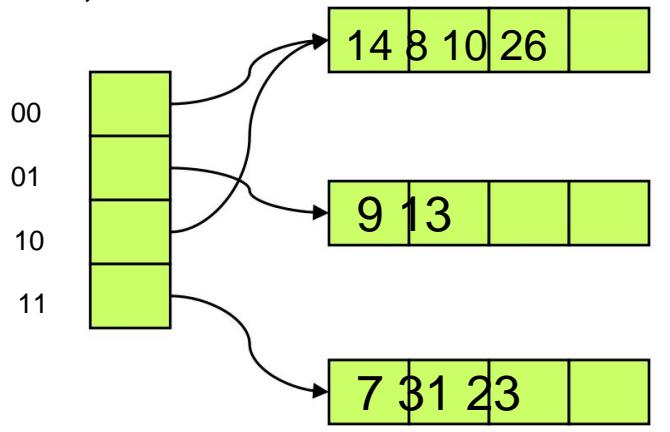
Request new data block

Splitting the data of the overflowing block according to  $ht+1$   
 $t = t + 1$  for old and new data block

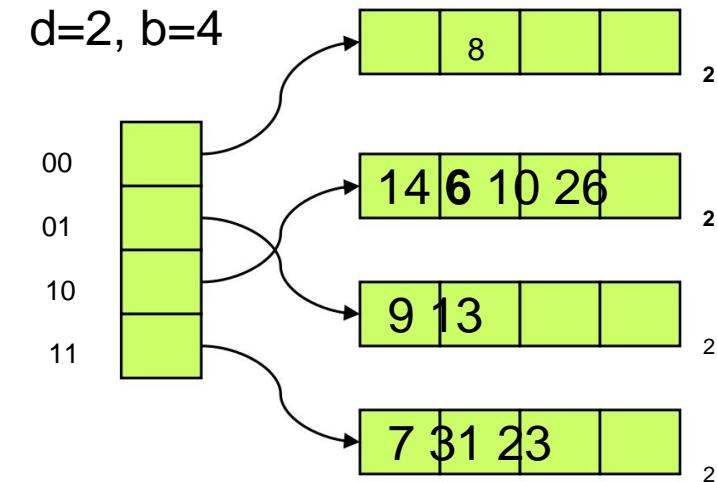
If split leads to overflow again, repeat

Example: Inserting 6 (=00110)

$d=2, b=4$



$d=2, b=4$



# Insert case 2: $t = d$

Requires index expansion (doubling the index)

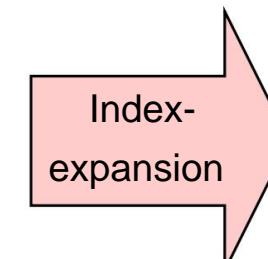
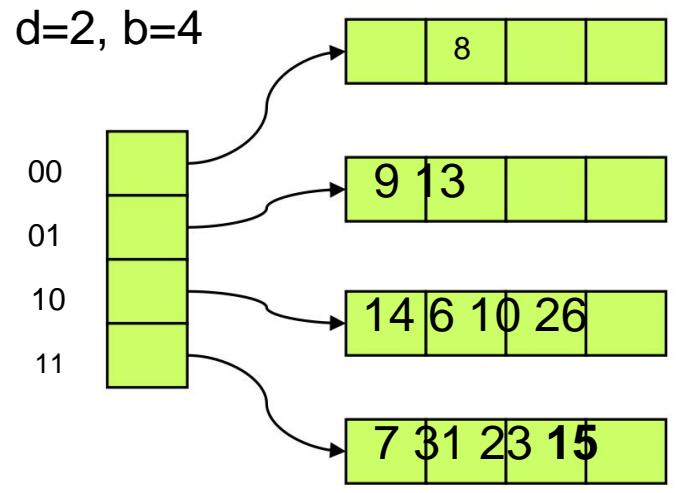
Request new storage area for  $2d$  additional references

For every  $T[x]$ , for which  $x \geq d$   
 $= d+1$

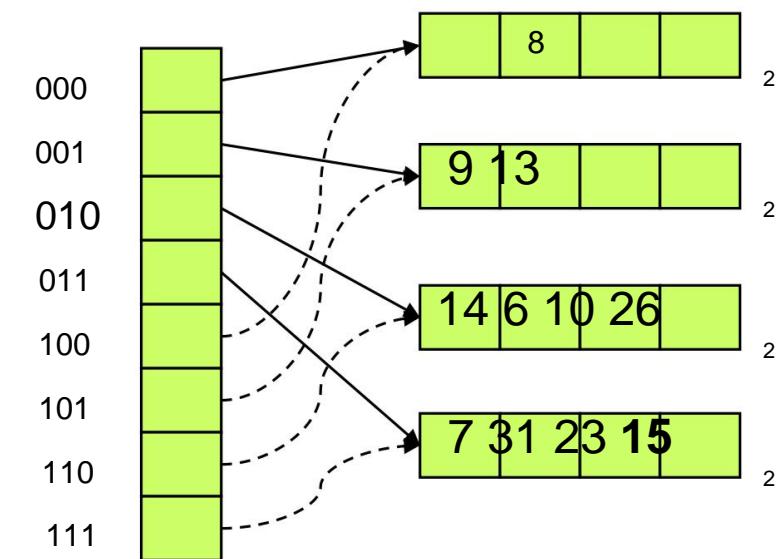
: Indexeintrag  $T[x] = T[x-2^d]$

Then return to case 1

Example: Inserting 15 and 19



$d=3, b=4$



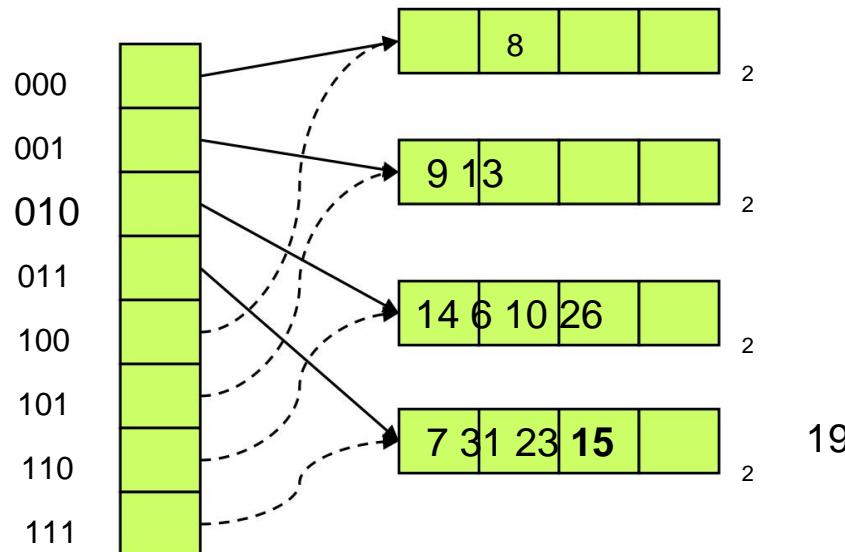


# Insert case 2: $t = d$ (2)

Now case 1

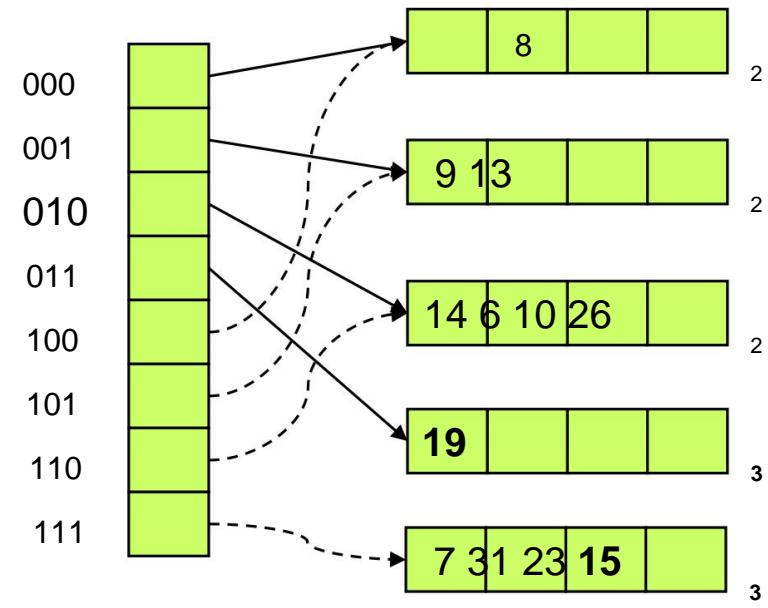
$7=00111$   
 $15=01111$   
 $19=10011$   
 $23=10111$   
 $31=11111$

$d=3, b=4$



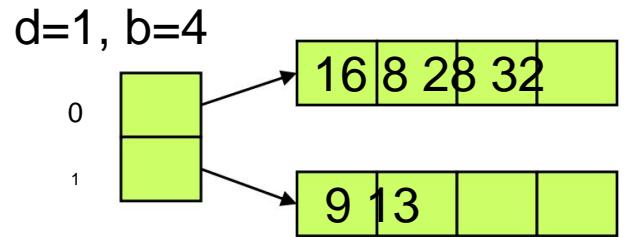
19

$d=3, b=4$



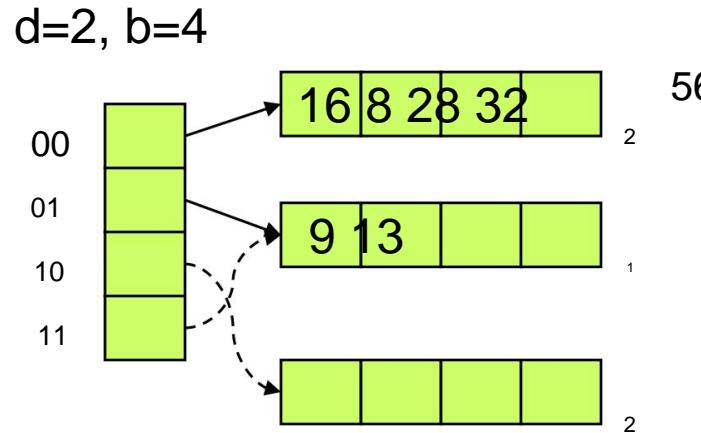
# *Index growth problem*

Index expansion can fail, multiple doubling

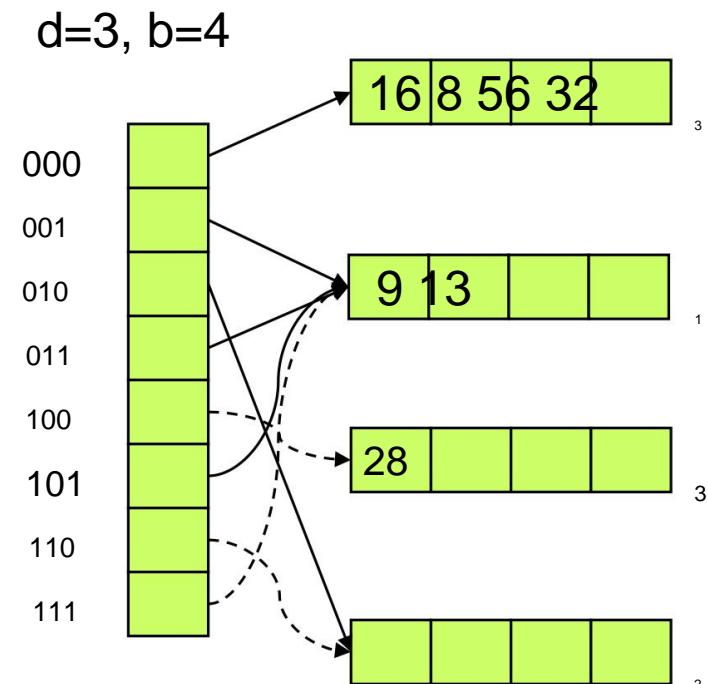


$8=001000$   
 $9=001001$   
 $13=001101$   
 $16=010000$   
 $28=011100$   
 $32=100000$   
 $56=111000$

Insert 56



1. Indexexpansion



2. Indexexpansion

# Comments

Deleting empty blocks is not easy

- Merging with their split buddies (“split images”)

- Analogous for index reduction

If index can be kept in main memory, only external access is necessary

In the worst case, exponential index growth

- Typical for clustered data (data is not evenly distributed)

- Therefore storage space requirement for index is not acceptable in the worst case

- Index usually in main memory

Index blocking problem on disk

No guaranteed minimum occupancy

# Analyse

## Analysis complex

It can be shown that to store a file with  $n$  records and a block size of  $b$  records, Extendible Hashing requires approximately  $1.44 * (n/b)$  blocks

On average, the directory has  $n+1/b/b$  for uniformly distributed data sets  
Entries

Big advantage if the directory fits in the main memory, only disk access to one data set is necessary

### 3.3.3 Bounded Index Size Extendible Hashing

approach

Primary area (analogous to Extendible Hashing)

Limited size index

Binary expansion of data areas

(1 block, 2, 4, ... blocks)

Attempts to address the inability of extendible hashing to accommodate the index on disk with a special index structure



# Structure

Index consists of x areas

A range contains y entries of the form  $\langle z, \text{ptr} \rangle$  where

$\text{ptr}$  is an address on a disk block

$z$  gives the number of duplications of the corresponding data areas

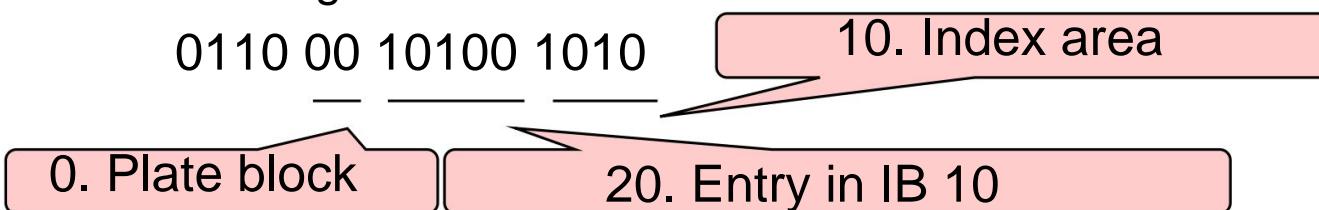
i.e.  $\text{ptr}$  is the address of a data area with  $2z$  disk blocks

x and y are powers of 2

Hash values are interpreted zoned, e.g.

16 index areas ( $x=16$ ), 32 entries per index area ( $y=32$ ), 4 blocks in each data area

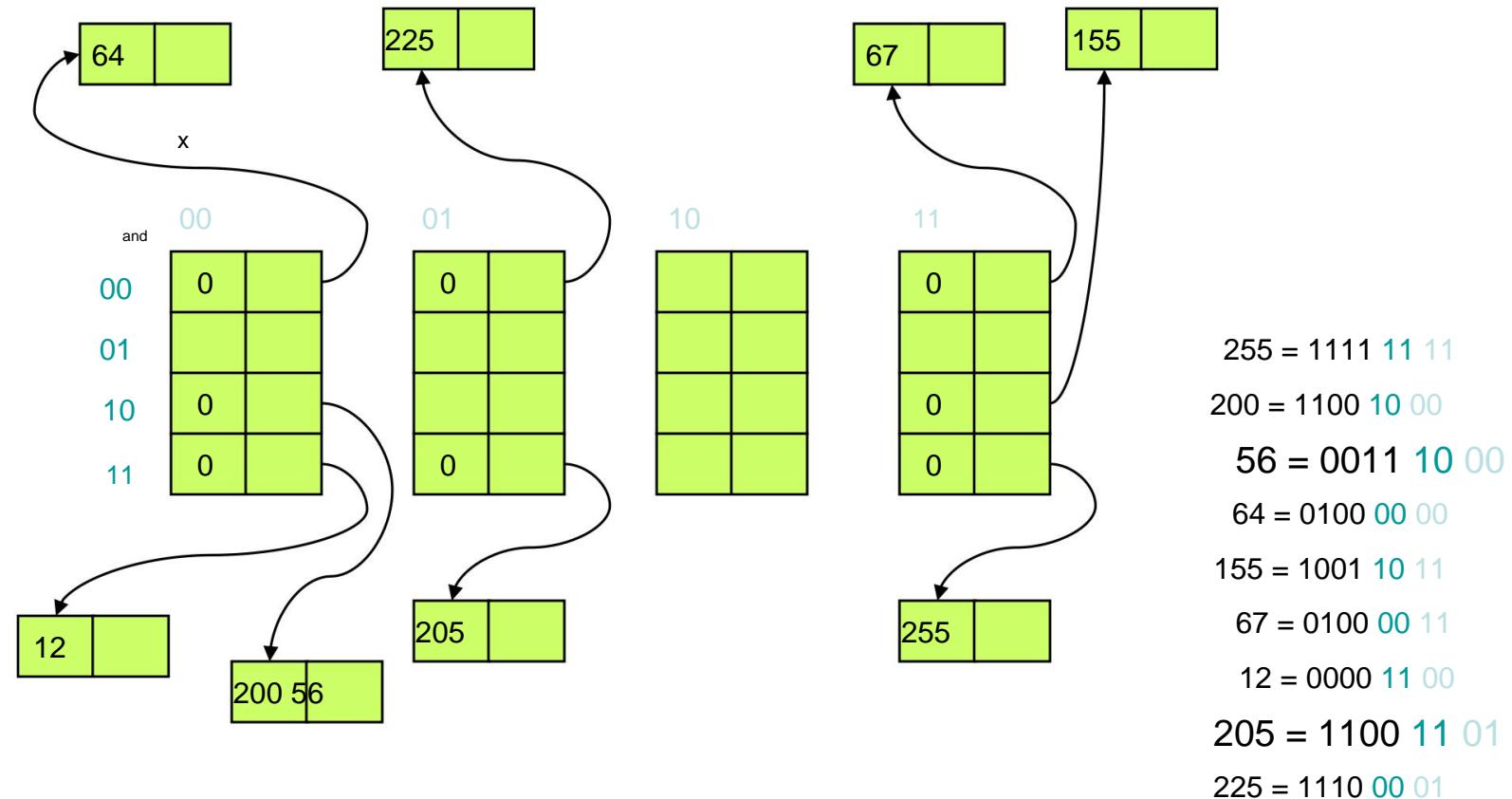
Hash signature for the 0th block referenced by the 20th entry in index range 10





# Example BISEH

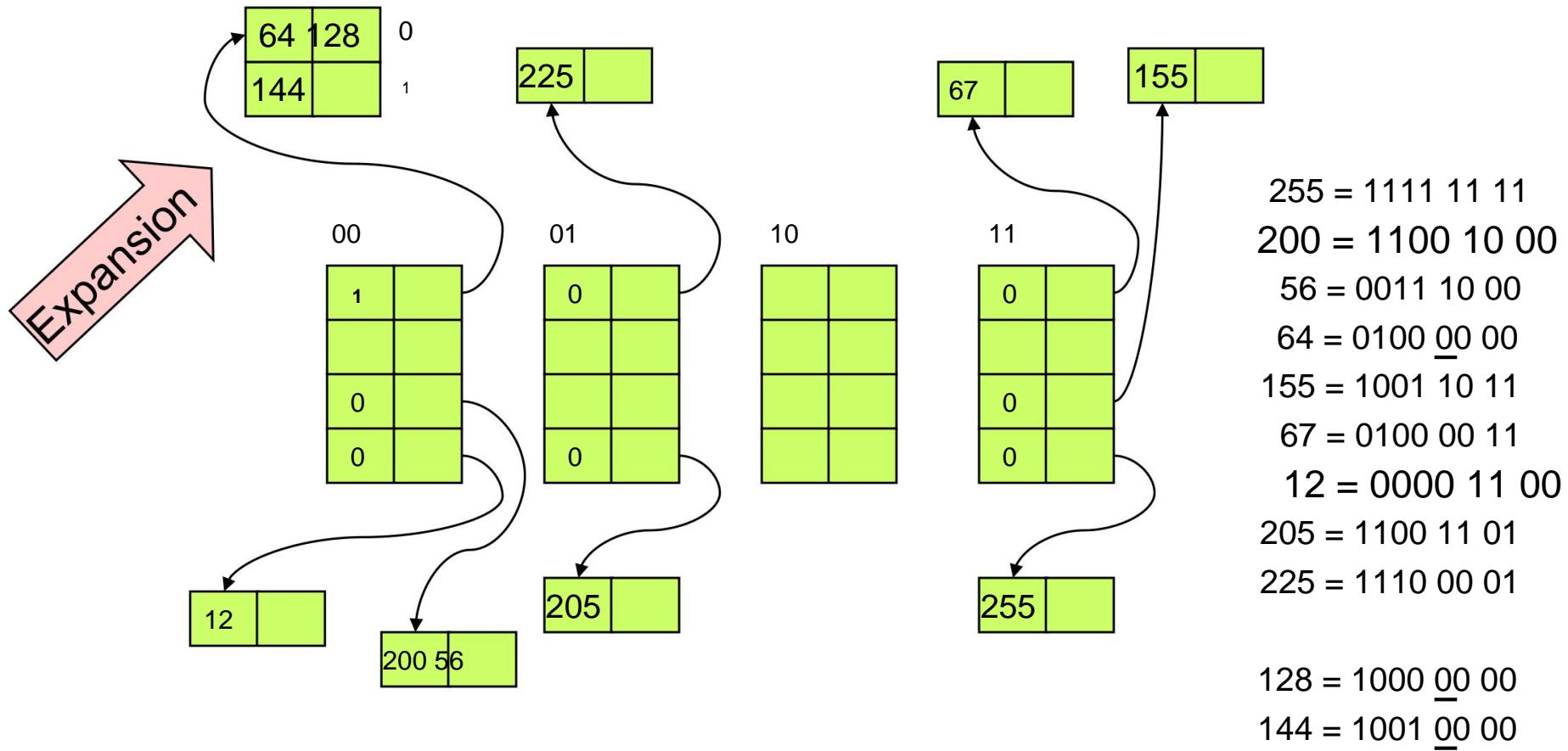
b=2 (block size), x=4 (index areas), y=4 (entries per IB)





# Example BISEH (2)

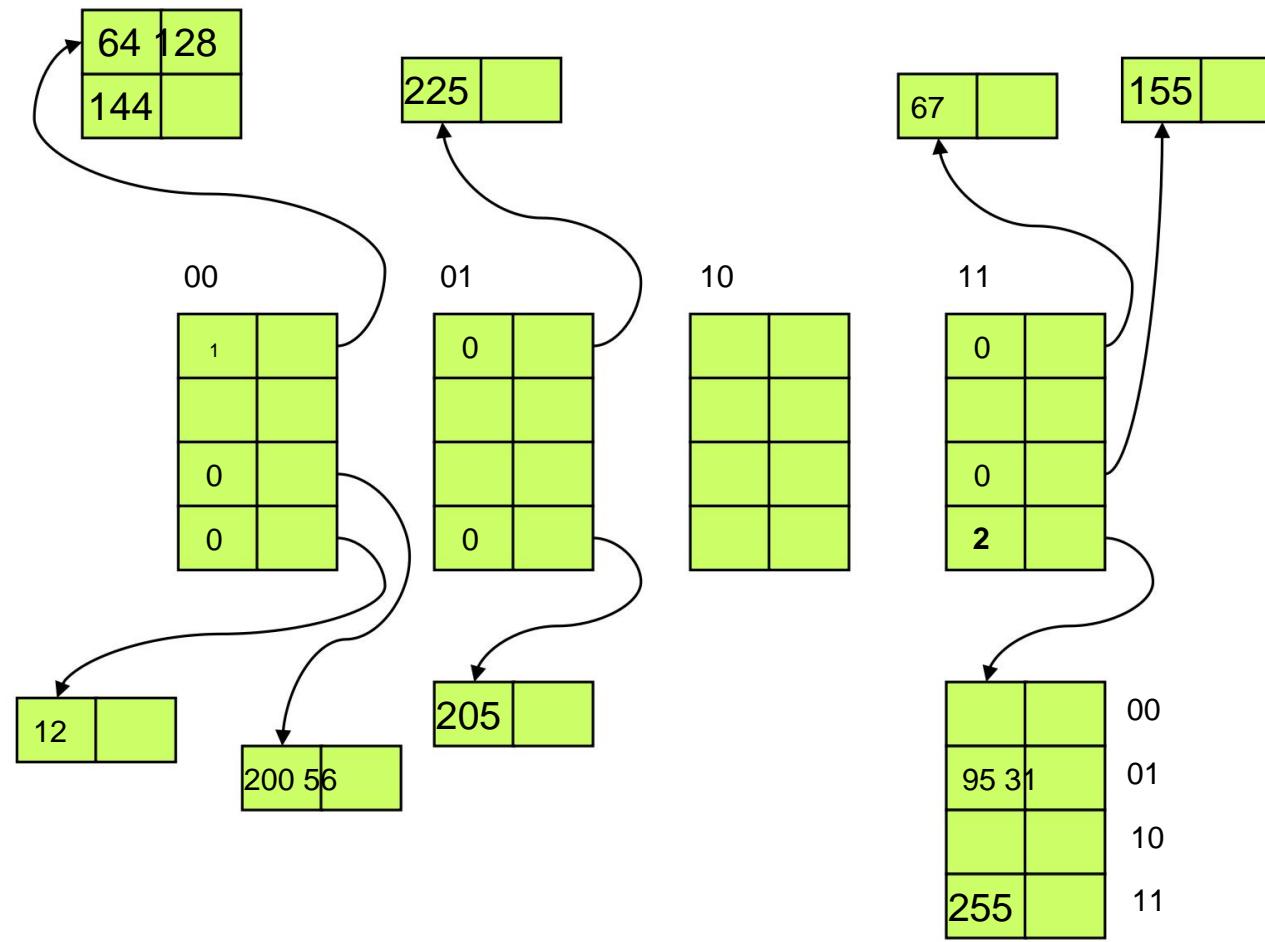
Insertion of 128 and 144, successful in first doubling attempt





# Example BISEH (3)

Insertion of 95 and 31, only successful in the second doubling attempt



$255 = 1111\_\underline{11}11$   
 $200 = 1100\ 10\ 00\ 56$   
 $= 0011\ 10\ 00\ 64 =$   
 $0100\ 00\ 00\ 155 =$   
 $1001\ 10\ 11\ 67 =$   
 $0100\ 00\ 11\ 12 =$   
 $0000\ 11\ 00\ 205 =$   
 $1100\ 11\ 01\ 225 =$   
 $1110\ 00\ 01\ 128 =$   
 $1000\ 00\ 00\ 144 =$   
 $1001\ 00\ 00$   
  
 $95 = 0101\ \underline{11}\ 11$   
 $31 = 0001\ \underline{11}\ 11$

# *General comparison of hash to Index procedure*



Points to note:

Costs of periodic reorganization

Insertion and deletion frequency

Average Case versus Worst Case Aufwand

Expected query types

Hashing is generally better for exact key queries

Index structures (e.g. search trees) are preferable for range queries

### 3.4 Sorting method

Sorting elements (values, data records, etc.) is one of *the* most important and complex problems in IT practice.

There are hundreds of sorting algorithms in a wide variety of variants for a wide variety of applications.

Main memory - external memory

Sequential - Parallel

In situ - Ex situ

Stable - Unstable

...

Sorting methods are among the most extensively analyzed and refined algorithms in computer science.

# Criteria for selecting Sorting method



universität  
wien

## Main memory - external memory

The algorithm can only store data in main memory or files  
Sort (files) or tapes on the external storage (disk, tape station)?

## In situ - Ex situ

Does the sorting process make do with the original data area (in situ) or  
does it need additional storage space (ex situ)?

## Worst Case - Average Case

Is the (asymptotic) runtime behavior of the sorting process always the same or can it “degenerate” (become faster or slower) in special cases?

## Stable - Unstable

If multiple records have the same key value, will the original order be retained after sorting?

## Sorting method

### Task A

vector of size  $n$  of the form  $V[0], V[1], \dots, V[n-1]$  is to be sorted.

Lexicographic order on the elements.

After sorting, the following applies:  $V[0] \leq V[1] \leq \dots \leq V[n-1]$ .

### Elementary (Simple) Procedures (Selection)

Selection Sort

Bubble Sort

### Refined (“smarter”) procedures (selection)

Quicksort

Mergesort

Heapsort

# Class Vector

```
class Vector {  
    int *a, size; public:  
  
    Vector(int max) { a = new int[max]; size = max; }  
    ~Vector() { delete[] a; } void  
    Selectionsort(); void  
    Bubblesort(); void  
    Quicksort(); void  
    Mergesort(); int Length();  
    private: void  
    quicksort(int,  
              int); void mergesort(int, int, int*);  
    void swap(int&, int&); };
```

### 3.4.1 Selection Sort

#### Selection sort or minimum search algorithm

Find the smallest element in the vector and swap it with the element in the first place. Repeat this process for all unsorted elements.

#### Swap

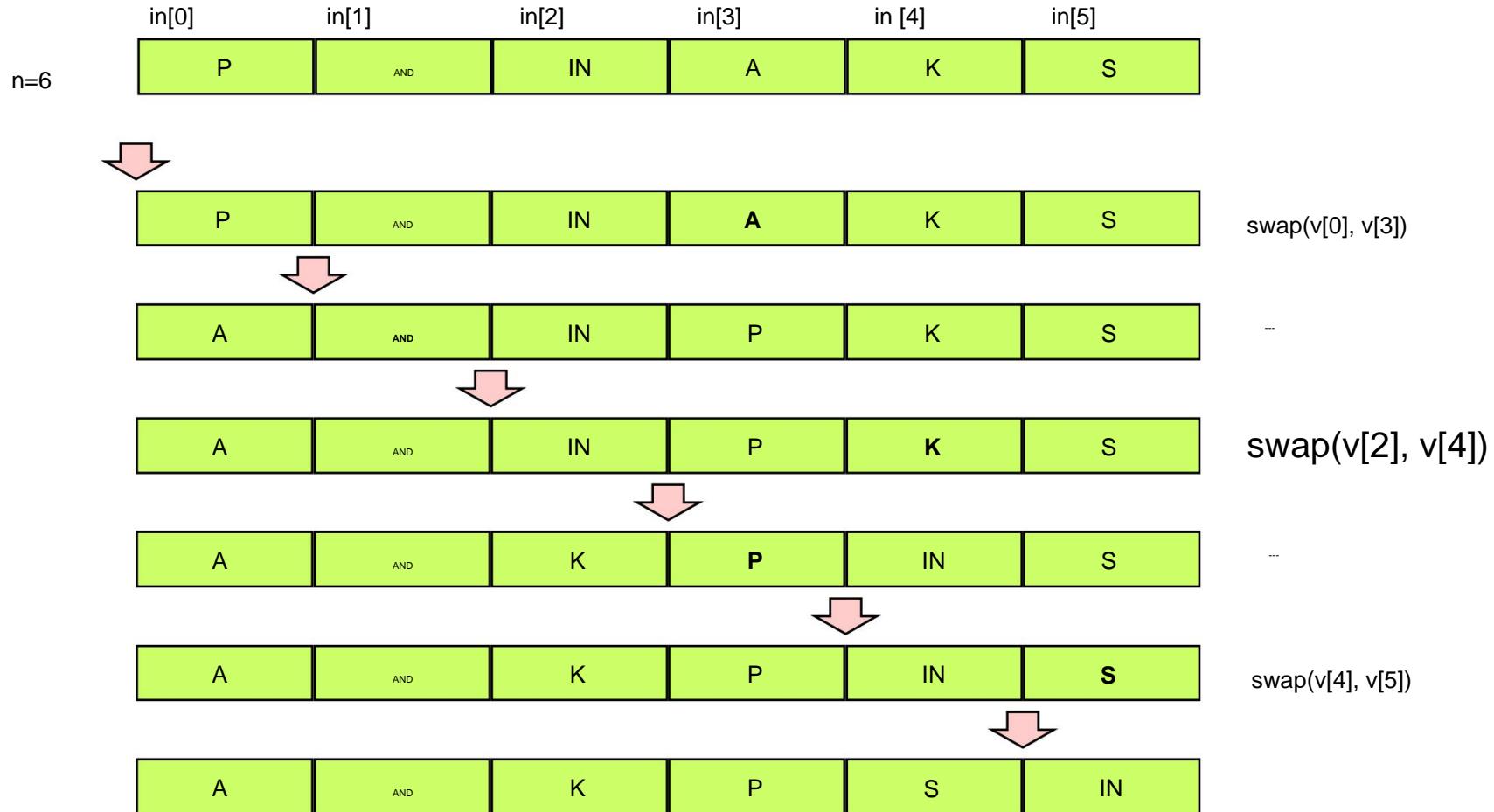
Swapping two elements (*int*) is a central process in sorting.

```
void swap(int &x, int &y) {  
    int help = x;  
    x = y;  
    y = help;  
}
```



# Selection sort, example

## Example



# Selection sort algorithm

## algorithm

```
void Vector::Selectionsort() { int n =  
    Length(); int i, j, min;  
    for(i = 0; i < n; i++)  
    { min = i; for(j = i+1; j < n; j++)  
        if(a[j] < a[min]) min = j; swap(a[min],  
        a[i]);  
    }  
}
```

# *Selection sort, properties*

## Characteristics

Main memory algorithm

In situ algorithm

sorts in its own data area, only needs a temporary variable,  
constant disk space consumption

Expense

general  $O(n^2)$

### 3.4.2 Bubble Sort

#### Bubble sort algorithm

Looping through the vector repeatedly.

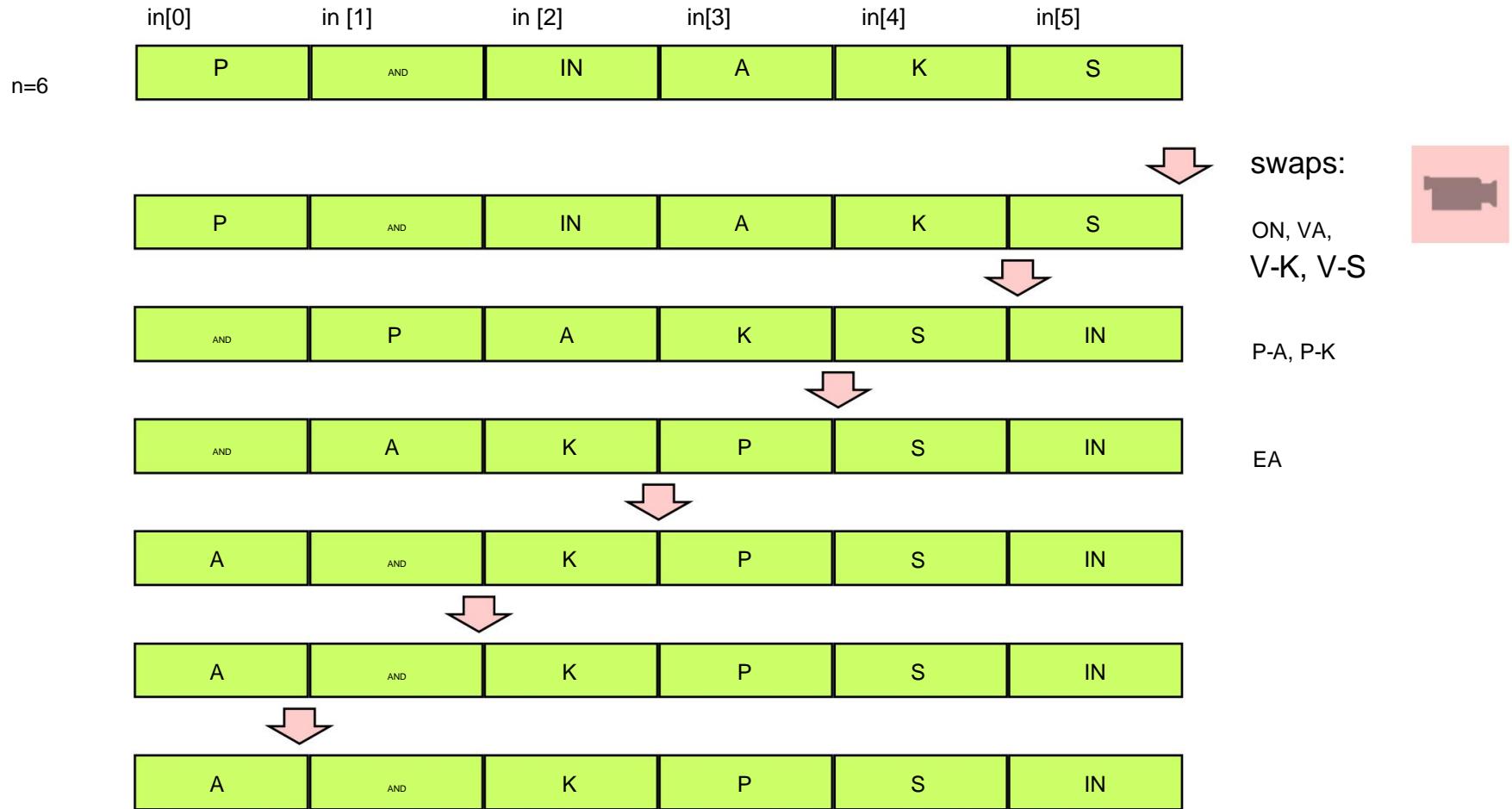
If 2 adjacent elements are out of order (larger before smaller), swap them.

This means that in the first run the largest element is placed last , in the second run the second largest is put in the second to last place, etc. The elements rise like bubbles.



# Bubble sort, example

## Example



# Bubble sort, algorithm

## algorithm

```
void Vector::Bubblesort() {  
    int n = Length(); int i, j;  
    for(i = n-1; i  
        >= 1; i--)  
        for(j = 1; j <= i; j++) if(a[j-1] > a[j])  
            swap(a[j-1], a[j]);  
}
```

# Bubble Sort, properties

## Characteristics

Main memory algorithm

In situ algorithm

sorts in its own data area, only needs a temporary variable,  
constant disk space consumption

Expense

general  $O(n^2)$

### 3.4.3 Quicksort

#### Quicksort (Hoare 1962) algorithm

The vector is defined in relation to a freely selectable pivot element

Rearrange the vector elements divided into 2 parts so that all elements to the left of the pivot element are smaller and all elements to the right are larger.

The two partial vectors are recreated independently of each other using quicksort (recursively) sorted.

divide-and-conquer Paradigm

The pivot element is now in its final position. Therefore, only  $n-1$  elements (in the two parts) remain to be sorted (reducing the size of the problem)

Based on the “divide-and-conquer” approach:

- Break down a problem into smaller, non-overlapping sub-problems that are (often) solved recursively. •

Subproblems here: numbers that are smaller than the pivot element and numbers that are larger than the pivot element

## Reorder pivot element

Select any element as pivot element

(in example left element in the vector, also random selection or median of 3 random numbers common)

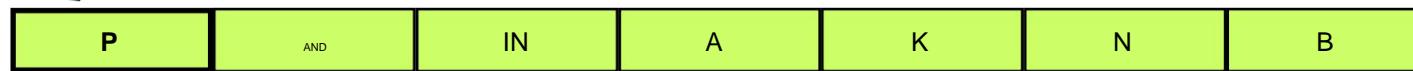
Pivot element 'sink into' the vector from the left or right

i.e. swap

sequentially from the left with all the smaller elements and from the right with all the larger ones.

In case of blocking, ie smaller element on the left and larger element on the right, swap these two elements

Pivotelement

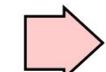
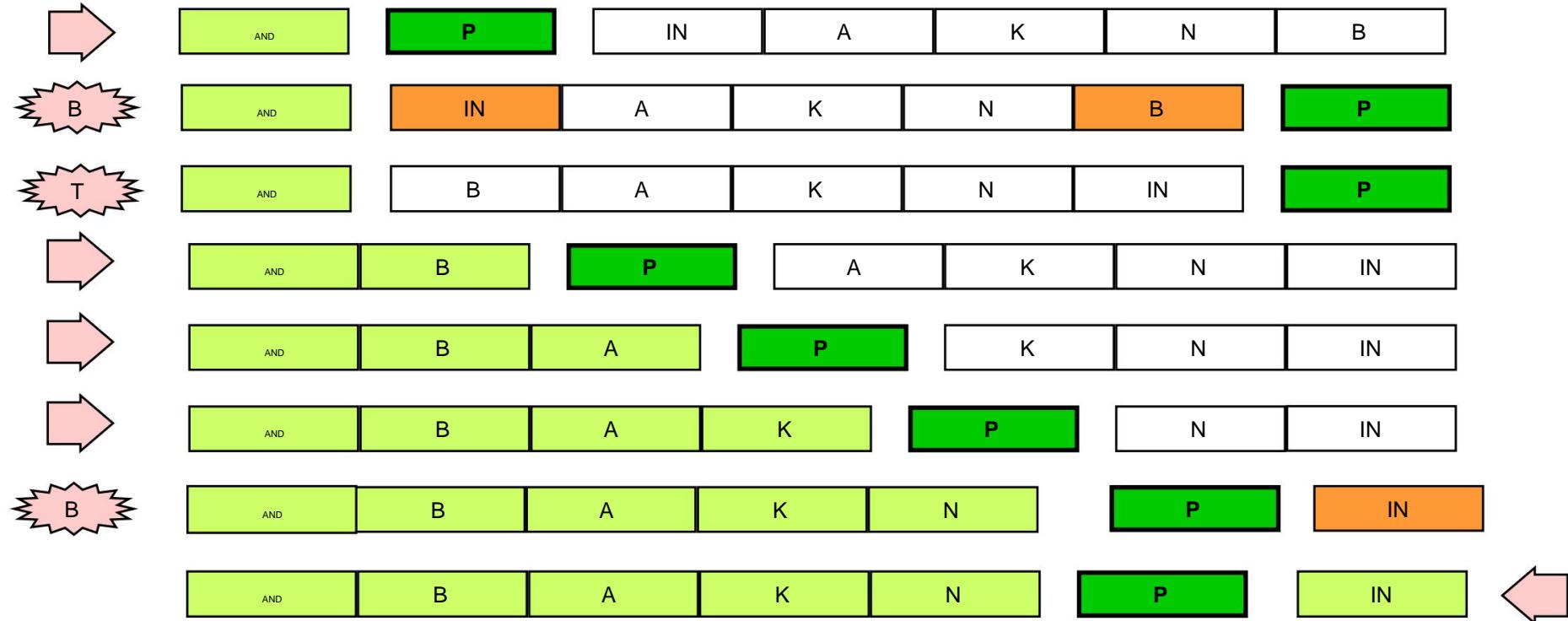


sink in from the left

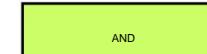
sink in from the right



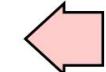
# Reorder pivot element (2)



sink in from the left



rearranged elements



sink in from the right



blocking elements



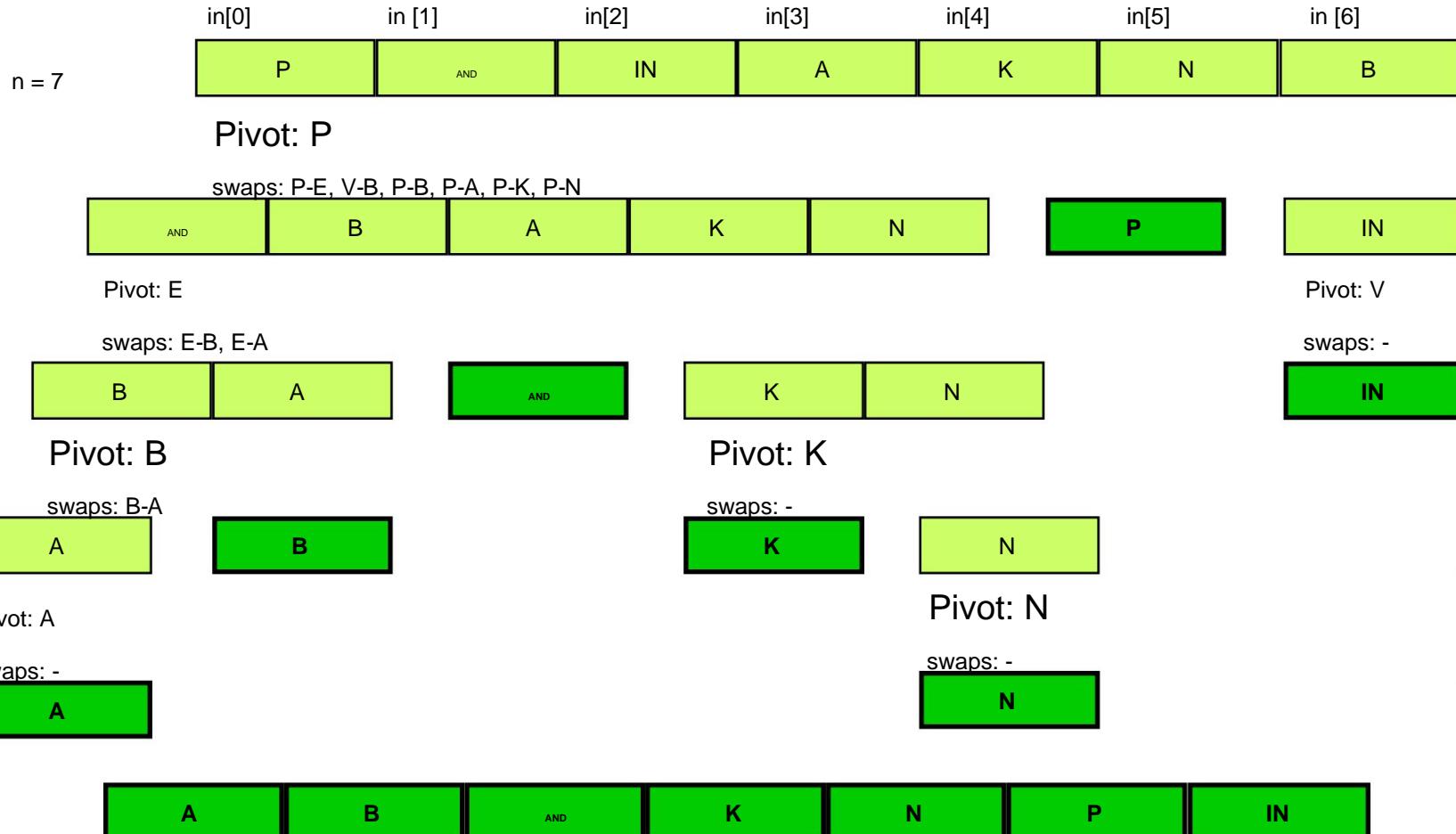
Blocking and swapping



Pivotelement



# Quicksort, recursion



# Quicksort, algorithm

```
void Vector::Quicksort() {
    quicksort(0, Length()-1);

} void Vector::quicksort(int l, int r) {
    int i, j; int pivot; if(r > l) {

        pivot = a[r]; i = l-1; j = r; for(;;) { while(a[+
        +i] < pivot);
            while(a[--j] > pivot) if (j == l)
                break; if(i >= j) break; swap(a[i], a[j]);

        } swap(a[i], a[r]);
        quicksort(l, i-1);
        quicksort(i+1, r);
    }
}
```

```
void Vector::quicksort(int l, int r) {  
    int i, j; int pivot;  
    if(r > l) {  
        pivot = a[r]; i = l-1; j = r;  
        for(;;) {  
            while(a[++i] < pivot);  
            while(a[--j] > pivot) if (j == l) break;  
            if(i >= j) break;  
            swap(a[i], a[j]);  
        }  
        swap(a[i], a[r]);  
        quicksort(l, i-1);  
        quicksort(i+1, r);  
    }  
}
```

3 5 2 6 1 7 4

3 1 2 6 5 7 4

```
void Vector::quicksort(int l, int r) {  
    int i, j; int pivot;  
    if(r > l) {  
        pivot = a[r]; i = l-1; j = r;  
        for(;;) {  
            while(a[++i] < pivot);  
            while(a[--j] > pivot) if (j == l) break;  
            if(i >= j) break;  
            swap(a[i], a[j]);  
        }  
        swap(a[i], a[r]);  
        quicksort(l, i-1);  
        quicksort(i+1, r);  
    }  
}
```

7 6 5 4 3 2 1

```
void Vector::quicksort(int l, int r) {
    int i, j; int pivot;
    if(r > l) {
        pivot = a[r]; i = l-1; j = r;
        for(;;) {
            while(a[++i] < pivot);
            while(a[--j] > pivot) if (j == l) break;
            if(i >= j) break;
            swap(a[i], a[j]);
        }
        swap(a[i], a[r]);
        quicksort(l, i-1);
        quicksort(i+1, r);
    }
}
```

5 6 7 3 2 1 4



```
void Vector::quicksort(int l, int r) {
    int i, j; int pivot;
    if(r > l) {
        pivot = a[r]; i = l-1; j = r;
        for(;;) {
            while(a[++i] < pivot);
            while(a[--j] > pivot) if (j == l) break;
            if(i >= j) break;
            swap(a[i], a[j]);
        }
        swap(a[i], a[r]);
        quicksort(l, i-1);
        quicksort(i+1, r);
    }
}
```

i can never be larger than r because pivot =a[r]  
if (j==l) then have

- all elements at positions k < ir a value > pivot
- i == r

if (i >= j) then have

- all elements at positions k < i have a value >= pivot
- the element at position i has a value >= pivot
- all elements at positions k > i have a value <= pivot

# Quicksort, properties

Main memory algorithm

In situ algorithm

sorts in its own data area, only needs temporary auxiliary variables  
(Stack), constant memory consumption

Expense

On average  $O(n \cdot \log(n))$

Can degenerate to  $O(n^2)$ .

Example: Vector is pre-sorted and the first or is always used as the pivot element  
last vector element selected

Sensitive to careless choice of pivot element

Stability is not easily achieved

Recursive algorithm

Complex to implement if recursion is not available

### 3.4.4 Mergesort

#### Mergesort (???, 1938) algorithm

The vector to be sorted is recursively divided into partial vectors of half length until the (trivial, scalar) vectors consist of a single element

Afterwards, 2 partial vectors each become a vector that is twice as large merged (sorted)

When merging, the first two elements of the vectors are successively compared and the smaller one is transferred to the new vector until all Elements in the new vector are stored sorted

The merging is repeated until there are 2 vectors in the last phase  
Length  $n/2$  merge into a sorted vector of length  $n$

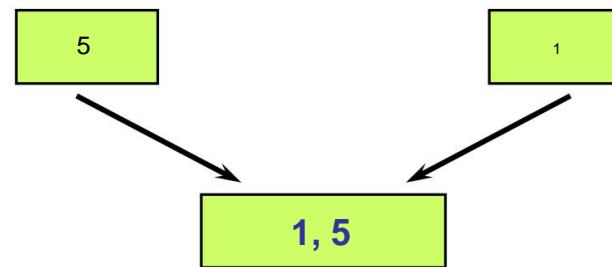
“divide-and-conquer” approach



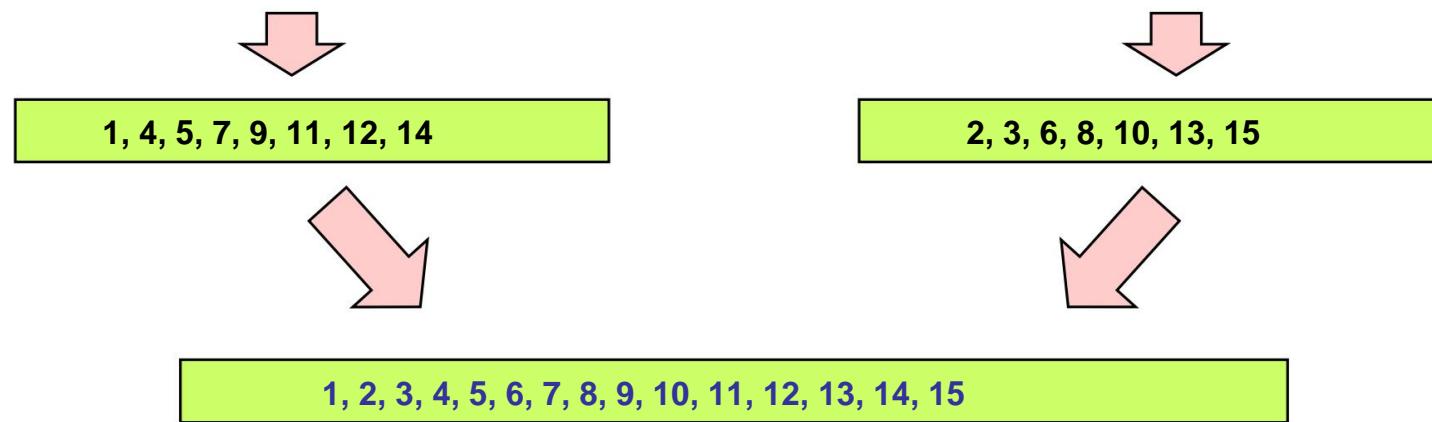
# Merging

## Merging two vectors

trivialer Fall



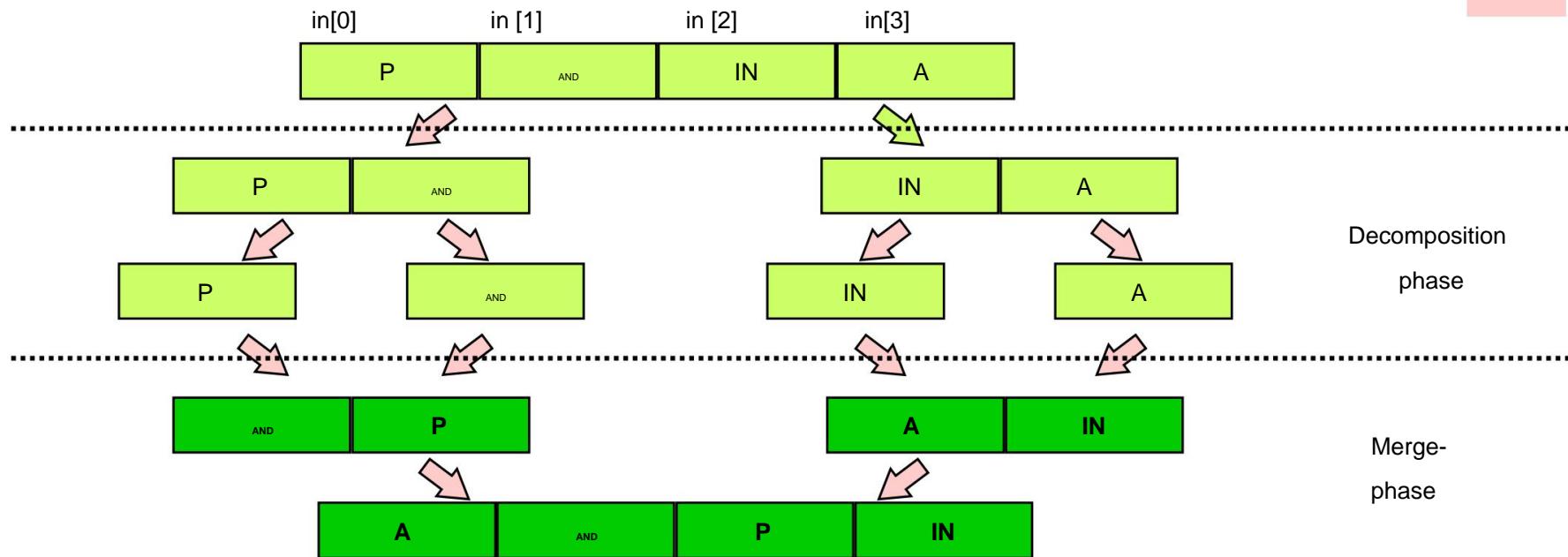
general case



# Mergesort, Recursion



n=4



# Class Vector

```
class Vector {  
    int *a, size; public:  
  
    Vector(int max) { a = new int[max]; size = max; }  
    ~Vector() { delete[] a; } void  
    Selectionsort(); void  
    Bubblesort(); void  
    Quicksort(); void  
    Mergesort(); int Length();  
    private: void  
    quicksort(int,  
              int); void mergesort(int, int, int*);  
    void swap(int&, int&); };
```

# Merge sort, algorithm (naive)

```

void Vector::Mergesort() {
  if (size>1) { int
    m=size/2; Vector
    left(m); Vector
    right(size-m); for (int i=0; i<m;
    ++i) left.a[i]=a[i]; for (int i=0; i<m; ++i) right.a[i]=a[i+m];

    left.Mergesort();
    right.Mergesort();

    int li=0; int
    ri=0; for (int
    i=0; i<size; ++i)
      if (li>=left.size) a[i]=right.a[ri++]; else if (ri>=right.size)
      a[i]=left.a[li++]; else a[i]=(right.a[ri]<left.a[li])?

      right.a[ri++]:left.a[li++];

  }
}

```

covers special cases?



# Merge sort, algorithm (optimized)

```

void Vector::Mergesort() {
    int *buf=new int[Length()]; mergesort(0,
    Length()-1, buf); delete[] buf;
}

void Vector::mergesort(int left, int right, int* buf) {
    if(right > left) { int i, j, k,
        m; m = (right+left)/2;
        mergesort(left, m, buf);
        mergesort(m+1, right, buf);
                                buf is temporary vector
        for(i=m+1; i>left; --i) buf[i-1] = a[i-1]; for(j=m; j<right; ++j)
        buf[right+m-j] = a[j+1]; // reverse order

        for(k=left; k<=right; ++k)
                                // i = left, j = right
            a[k]=(buf[i]<buf[j])?buf[i++]:buf[j--];
    }
}

```

# Merge sort, algorithm (optimized)

```
void Vector::Mergesort() {
    int *buf=new int[Length()]; mergesort(0,
    Length()-1, buf); delete[] buf;
}
```

```
void Vector::mergesort(int left, int right, int*
buf) { if(right > left) { int i, j, k,
m; m = (right+left)/2;
mergesort(left, m,
buf); mergesort(m+1, right,
buf);
```

**for(i=m+1; i>left; --i) buf[i-1] = a[i-1]; for(j=m; j<right; ++j)  
 buf[right+m-j] = a[j+1]; for(k=left; k<=right; ++k)**

**a[k]=(buf[i]<buf[j])?buf[i++]:buf[j--];**

}

```
void Vector::Mergesort() {
    if (size>1) {
        int m=size/2;
        Vector left(m);
        Vector right(size-m);
        for (int i=0; i<m; ++i) left.a[i]=a[i];
        for (int i=0; i<m; ++i) right.a[i]=a[i+m];

        left.Mergesort();
        right.Mergesort();

        int li=0;
        int ri=0;
        for (int i=0; i<size; ++i)
            if (li>=left.size) a[i]=right.a[ri++];
            else if (ri>=right.size) a[i]=left.a[li++];
            else a[i]=(right.a[ri]<left.a[li])?
                right.a[ri++]:left.a[li++];
    }
}
```

Correct?

Sonderfälle abdecken



# Merge sort, algorithm (optimized)

```
void Vector::Mergesort() {  
    int *buf=new int[Length()]; mergesort(0,  
    Length()-1, buf); delete[] buf;
```

```
}
```

```
void Vector::mergesort(int left, int right, int* buf) {
```

```
    if(right > left) { int i, j, k,  
        m; m = (right+left)/2;  
        mergesort(left, m, buf);  
        mergesort(m+1, right, buf);
```

possible overflow

```
    for(i=m+1; i>left; --i) buf[i-1] = a[i-1]; for(j=m; j<right; ++j)  
        buf[right+m-j] = a[j+1]; for(k=left; k<=right; ++k)
```

```
        a[k]=(buf[i]<buf[j])?buf[i++]:buf[j--];
```

```
}
```

```
}
```

# Merge sort, properties

## Characteristics

Main memory algorithm, but also as an external memory algorithm usable.

External sorting: instead of partial phase, predetermined data division is often used, or only divided until partial vector (file) can be sorted in main memory.

Stable sorting process

Mergen must be implemented correctly

Exsitu algorithm needs

a second equally large auxiliary vector for restoring

Expense

General  $O(n^* \log(n))$  **Master theorem!**

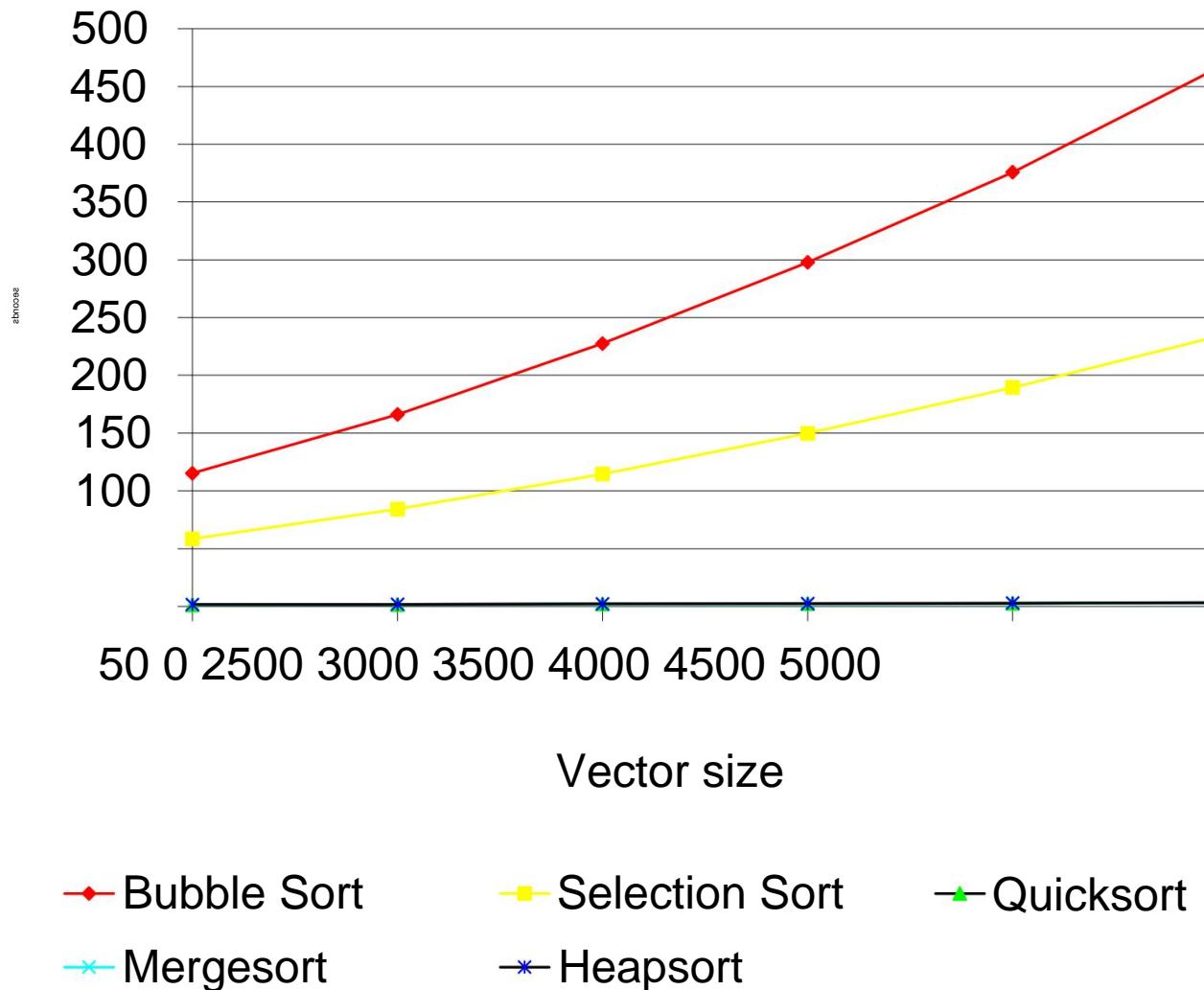
Takes double space Recursive

algorithm

Complex to implement if recursion is not available.



# Runtime comparison of the procedures



### 3.4.5 Comparison sort procedure

These well-known procedures are called *comparative Sorting method* (Comparison sort)

The order of the elements is determined by comparing the elements with each other

Only comparisons of the form  $x_i < x_j$ ,  $x_i \geq x_j$ , ... applied

These algorithms show an order of  $O(n \log n)$  as the cheapest running time , which means that we have so far been able to determine the lower limit as  $\Omega(n \log n)$ .

Conjecture: The problem of sorting by comparing the Elements can be described with  $\Omega(n \log n)$ .

If valid  $\Omega$  Algorithmic gap closed since  $\Omega(P) = O(A)$

# Decision tree

Consideration of comparative sorting methods

## Decision trees

The decision tree contains all possible comparisons of a  
Sorting process by any sequence of a given one  
sort length

The leaf nodes represent all possible permutations of the input  
sequence

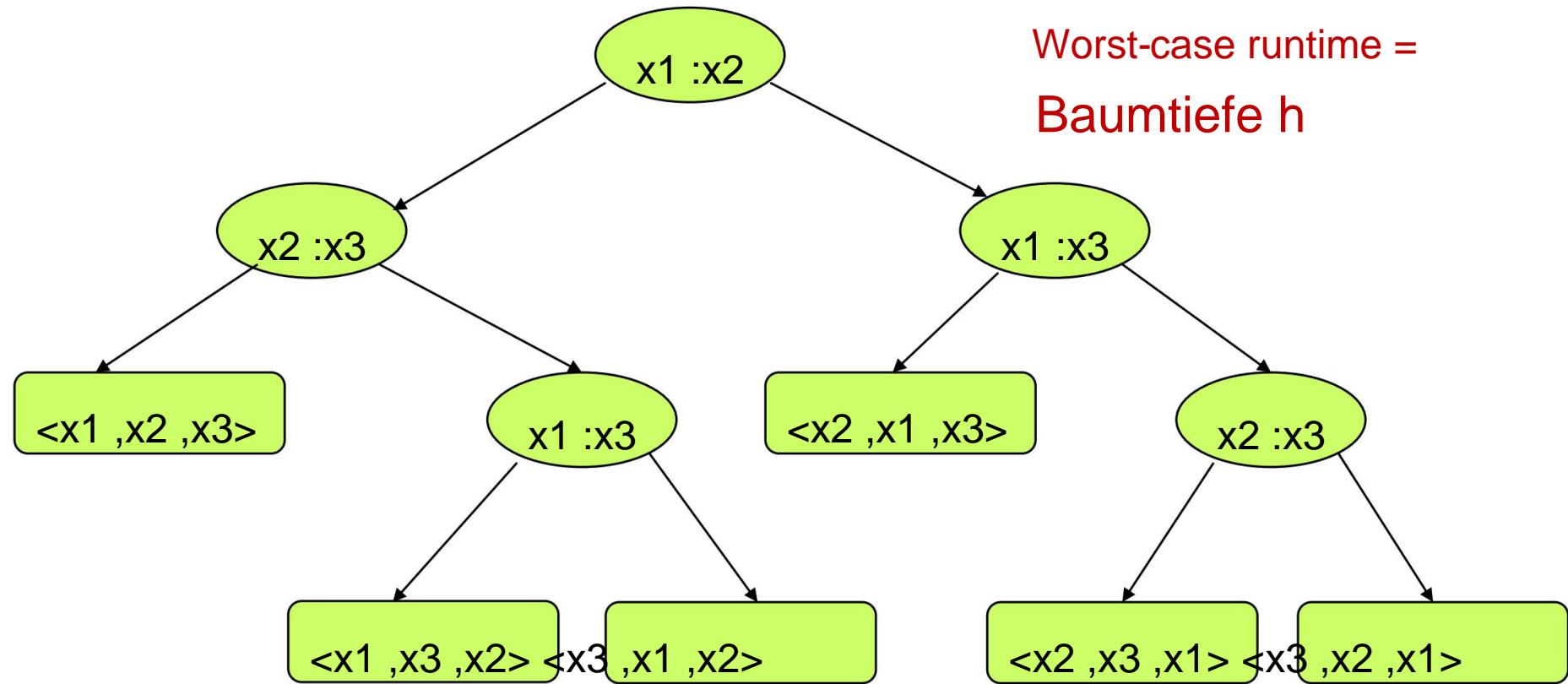
The paths from the root to the leaves define the necessary ones  
Compare to achieve the corresponding permutations



# Example decision tree

Decision tree for 3 elements ie  $3! = 6$

possible permutations of the input elements  $\langle x_1, x_2, x_3 \rangle$  Left successor relation  $x:y$  fulfilled, right successor not fulfilled



# Proof: Lower limit for the worst case

We ignore runtime of swapping, administration operations, etc.

The number of leaves in the decision tree is  $n!$

The number of leaves in a binary tree of depth  $h$  is  $\geq 2^h$

therefore

$$2^h \geq n!$$

$$h \lg( !) \geq n$$

*Stirling approximation*  $n! \approx \sqrt{\pi n} \frac{n^n}{in} + \frac{\pi}{4} \left(\frac{1}{n}\right)^n$

$$h \geq \lg\left(\frac{n^n}{in}\right) = n \lg n - \lg$$

$$h = \Theta(n \lg n)$$

i.e. Algorithmic Gap for Comparative Sorting Methods closed

## 3.5 Sorting in linear time

Sorting methods that do NOT rely on comparing two values can sort an input sequence in linear time

Reach order  $O(n)$

### Examples

Counting Sort

Sort Radix

Bucket Sort

### 3.5.1 Counting Sort

1954 First used by Harold H. Seward MIT thesis “Information Sorting in the Application of Electronic Digital Computers to Business Operations” as the basis for Radix Sort.

1956 First published by EH Fried

1960 under the name Mathsor by W. Feurzig.



# Counting Sort

Condition

The  $n$  elements to be sorted are integer values in the range 0 to  $k$

Usually  $k > n$  applies

If  **$k$  is of order  $n$**  ( $k = O(n)$ ) the effort for this is

Sortierverfahren Counting Sort  $O(n)$

Approach

An exact permutation is given

Each element is simply assigned to the position of its value in a

Target vector set

A	5	8	4	2	7	1	6	3
---	---	---	---	---	---	---	---	---

B	1	2	3	4	5	6	7	8
---	---	---	---	---	---	---	---	---

# Counting Sort (2)

extension

The elements come from the range  $[1..k]$  and there are no duplicate values

Question: how can we extend the first approach?

General case

The elements are from the range  $[1..k]$ , duplicates are allowed. Idea of counting sort

For each element  $x$ , determine the number of elements smaller than  $x$  in the vector, use this information to place the element  $x$  at its position in the resulting vector



# Counting sort algorithm

Input vector A, result vector B, auxiliary vector **C** (C has size k)

```
for(i=1; i<=k; i++)
```

```
    C[i] = 0;
```

```
for(j=1; j<=length(A); j++)
```

```
    C[A[j]] = C[A[j]] + 1;
```

```
for(i=2; i<=k; i++)
```

```
    C[i] = C[i] + C[i-1];
```

```
for (j=length(A); j>=1; j--) {
```

```
    B[C[A[j]]] = A[j];
```

```
    C[A[j]] = C[A[j]] - 1;
```

```
}
```

1: C[i] contains the number of elements equal to i,  $i = 1, 2, \dots, k$

2: C[i] contains the number of Elements less than or equal to i

3: Each element is placed in its correct position

If all elements  $A[j]$  are different, the correct position is in  $C[A[j]]$ , since elements can be equal the Value  $C[A[j]]$  decreased by 1



# Counting sort example

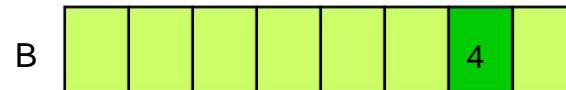
1:



2:



3:



1st pass

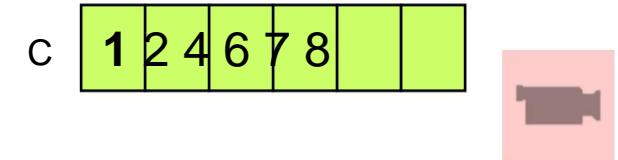


```
for (j=length(A); j>=1; j--)
    B[C[A[j]]] = A[j];
    C[A[j]] = C[A[j]] - 1;
```

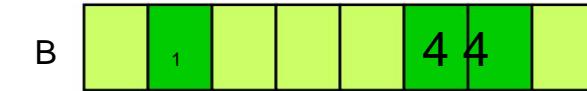
3:



2. D.



3:



3. D.



3:



End

# Counting Sort Analyse

```

for(i=1; i<=k; i++)
  C[i] = 0;
for(j=1; j<=length(A); j++)
  C[A[j]] = C[A[j]] + 1;
for(i=2; i<=k; i++)
  C[i] = C[i] + C[i-1];
for (j=length(A); j>=1; j-) {
  B[C[A[j]]] = A[j];
  C[A[j]] = C[A[j]] - 1;
}
  
```

OK)

O(*n*)

OK)

O(*n*)

It follows that the total effort is  $O(n + k)$ .

In practice we very often have  $k = O(n)$ , which means we have a real overhead of  $O(n)$ .

Note that there was no comparison between values, but rather the values of the elements were used for placement (see hashing)

Counting Sort is a stable sorting method

Guaranteed by final loop working from back to front

No insitu method, 3 vectors of size  $O(n)$  necessary

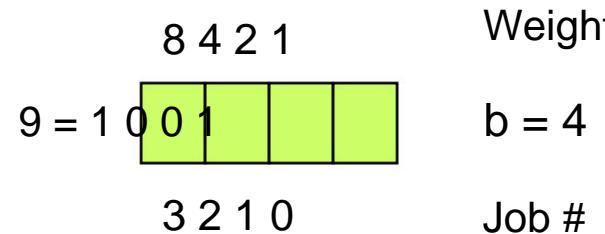
## 3.5.2 Radix Sort

Radixsort looks at the structure of the key values

The key values of length  $b$  are in a base number system

$M$  represented ( $M$  is the radix)

e.g.  $M=2$ , the keys are represented in binary,  $b = 4$



Keys are sorted by comparing their individual bits at the same bit position

The principle can also be extended to alphanumeric strings

# Radix sort algorithm

General radix sort algorithm

**for(index i runs over all b places) {**

    sort key values with a (stable)

        Sorting procedure regarding location i

}

Direction of index run, stability

Left to right

d.h. **for(int i=b-1; i>=0; i--) { ... }**

Leads to *Binary Quicksort (Radix Exchange Sort)*

Right to left and stable sorting process

d.h. **for(int i=0; i<b; i++) { ... }**

Leads to the *LSD Radix Sort (Straight Radix Sort)*

# Binary quicksort

Very old procedure

In 1929, for the first time for machine sorting of punch cards.

1954 H.H. Seward [MIT R-232 (1954), p25-28 ]

independent and detailed: P.Hildebrandt J. Schwartz . H. Isbitz, H. Rising,  
[JACM 6

(1959), 156-163]

1 year before Quicksort



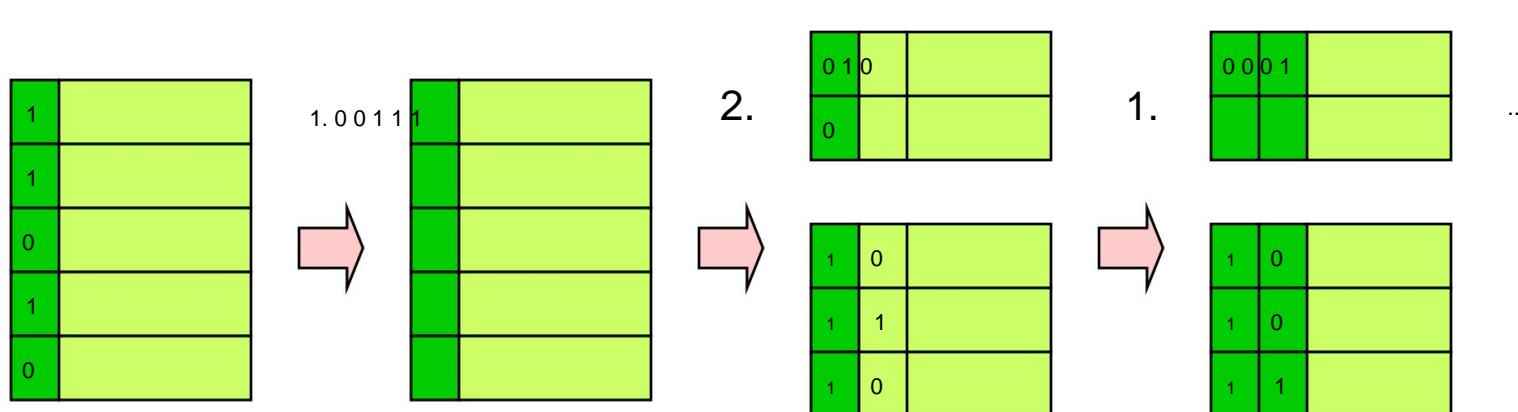
# Binary quicksort

## Binary quicksort algorithm

Look at spots from left to right

1. Sort data sets based on the bit under consideration
2. Divide the data sets into M independent groups, ordered by size, and sort the M groups recursively, ignoring the bits already under consideration

## Example





# Binary quicksort partition

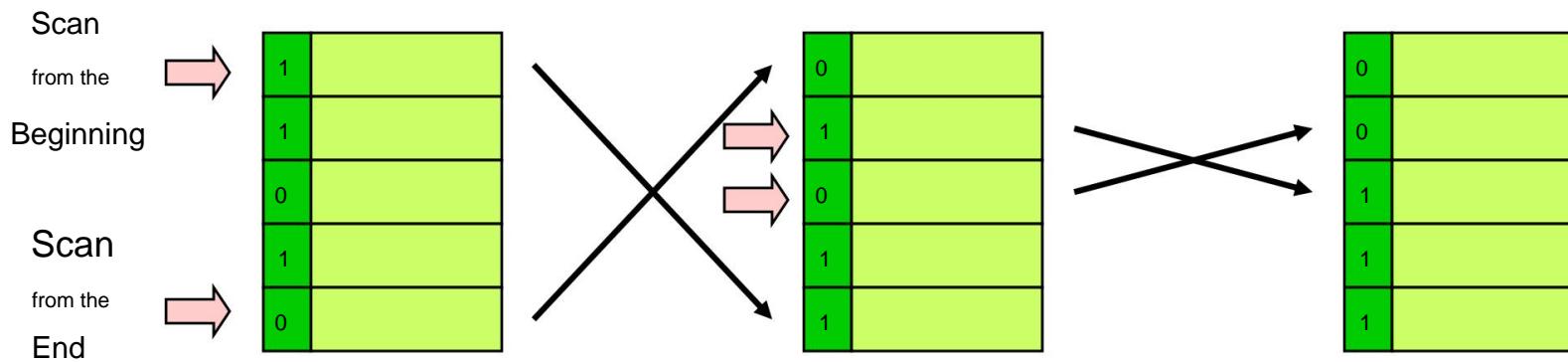
Splitting the data sets using an in-situ approach similar to that of Partitioning in quicksort

**do {**

**scan top-down to find key starting with 1; scan bottom-up to  
    find key starting with 0; exchange keys;**

**} while (not all indices visited)**

## Example

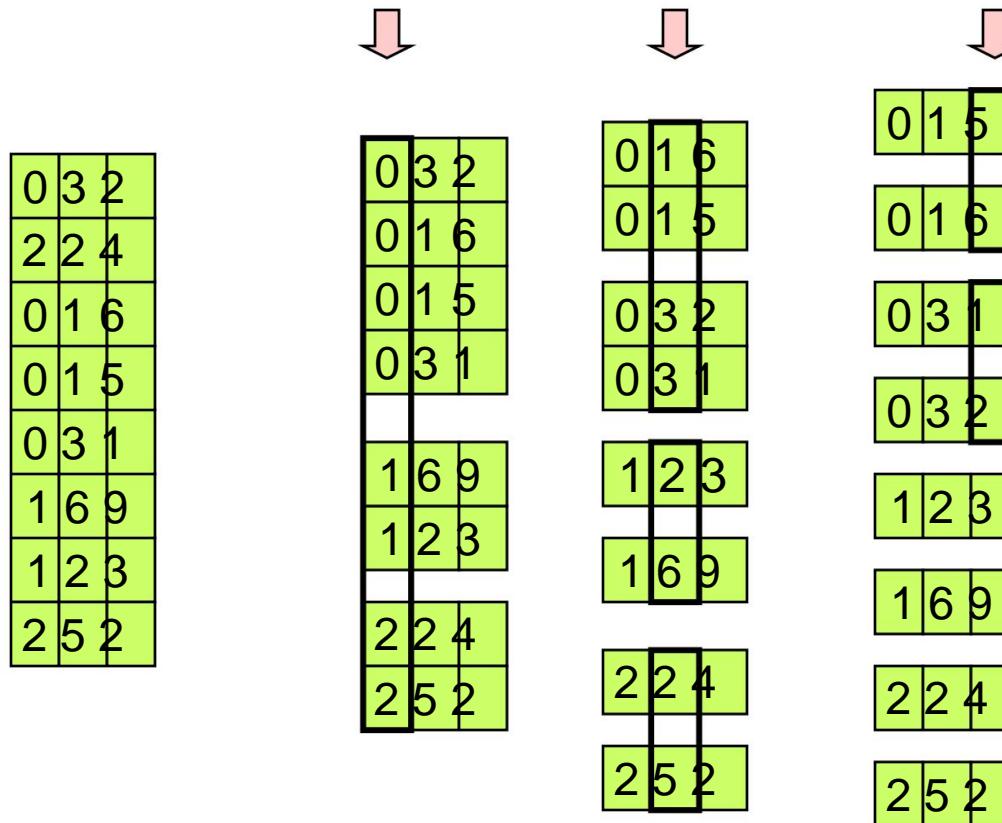


# Analogue: “Decimal” quicksort for Decimal numbers



Number of groups M is 10

Digit values 0 to 9



# Binary quicksort properties

Binary quicksort uses on average approximately  $N \log N$

bit comparisons  $\log N$  (encoding the numerical value) = usually constant factor  
 $N$  = Number of values to sort

Well suited for small numbers

Requires relatively less memory than other linears  
Sorting method

It is an unstable sorting method

Similar to quicksort only for positive integers

# LSD-Radixsort

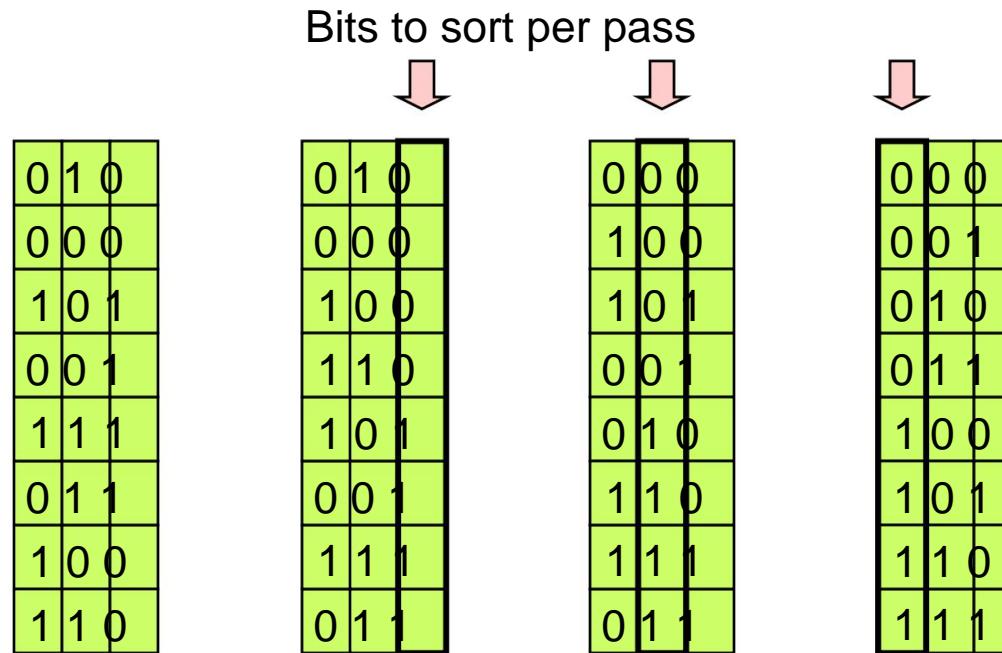
## LSD radix sort algorithm

Look at bits from right to left

In binary quicksort from left to right!

LSD ... “least significant digit”

Sort data sets stably (!) based on the bit under consideration



# What does “stable” mean?

With a **stable** sorting process, the relative order of the same keys (= value under consideration) is not changed

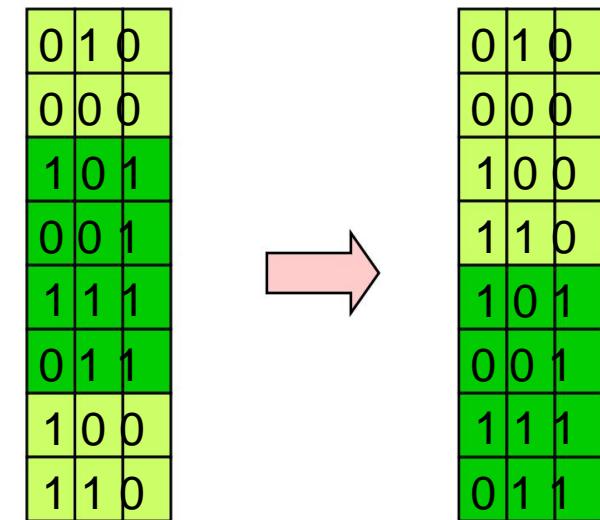
## Example

First pass of last example

The relative order of the Keys that end with 0 remain unchanged, the same applies to keys that end with 1

Stability guarantees correctness algorithm

Possible procedure: Counting Sort



# What does “stable” mean?

With a stable sorting process, the relative order of identical keys is not changed

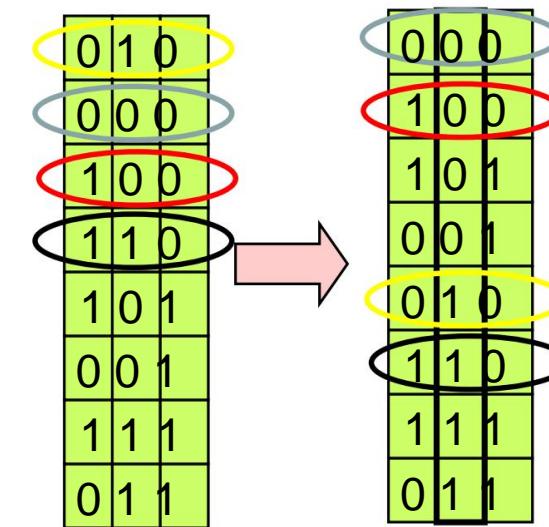
## Example

First pass of last example

The relative order of the Keys that end with 0 remain unchanged, the same applies to keys that end with 1

Stability guarantees correctness algorithm

Possible procedure: Counting Sort



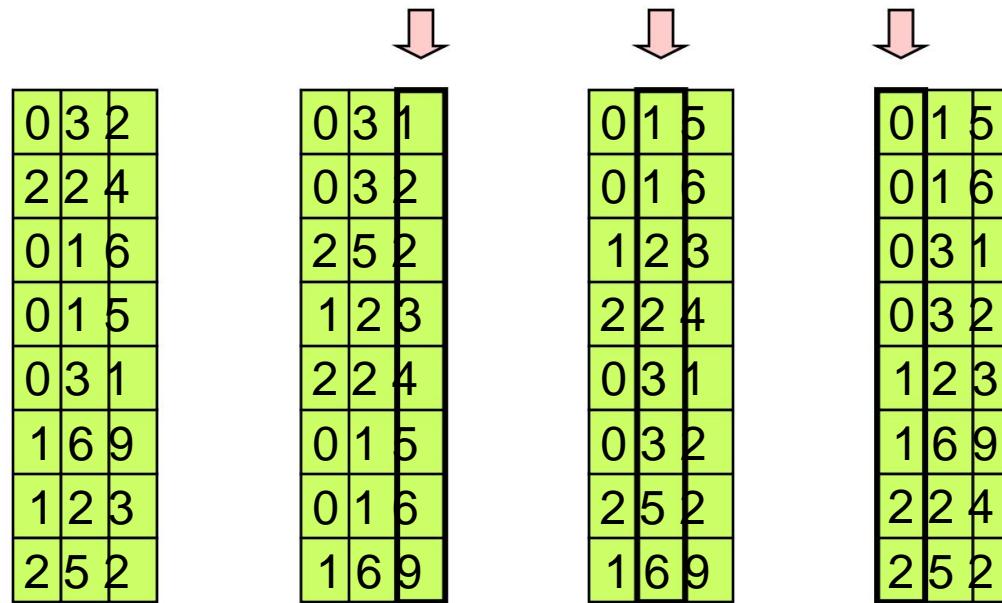
Relative order of blue and red remains unchanged because they have the same value in the 2nd bit



# LSD radix sort for decimal numbers

Number of groups M is 10

Digits 0 to 9





# Analysis of Radix Sort

Sorting  $n$  numbers, key length  $b$

```
for(Index i runs over all b positions) {  
    sort n key values with a (stable)  
        Sorting procedure regarding location i  
}
```

When using a sorting method with a runtime of  $O(n)$  (such as partitioning in Radix Exchange Sort, Counting Sort in Straight Radix Sort) is therefore the order of Radix Sort  $O(bn)$

If you look at  $b$  as a constant you get  $O(n)$

For a numerical range of  $m$  values, representation on the computer with  $b = \log(m)$   
However, since the range of values on the computer is limited, approach  $b$  is consistently acceptable

No in-situ procedure, double storage space required (for that stable sorting processes)

In practice, this leads to the use of  $O(n \log n)$  methods

### 3.5.3 Bucket Sort

Assumption about the  $n$  input elements that they are uniform are distributed over the interval  $[0,1)$ .

#### Bucket sort algorithm

Divide the interval  $[0,1)$  into  $n$  equal-sized sub-intervals (buckets) and distribute the  $n$  input elements among the buckets

Since the elements are evenly distributed, only a few elements fall into the same bucket

Sort the items per bucket using a different sorting method (e.g.  
B. Selection Sort)

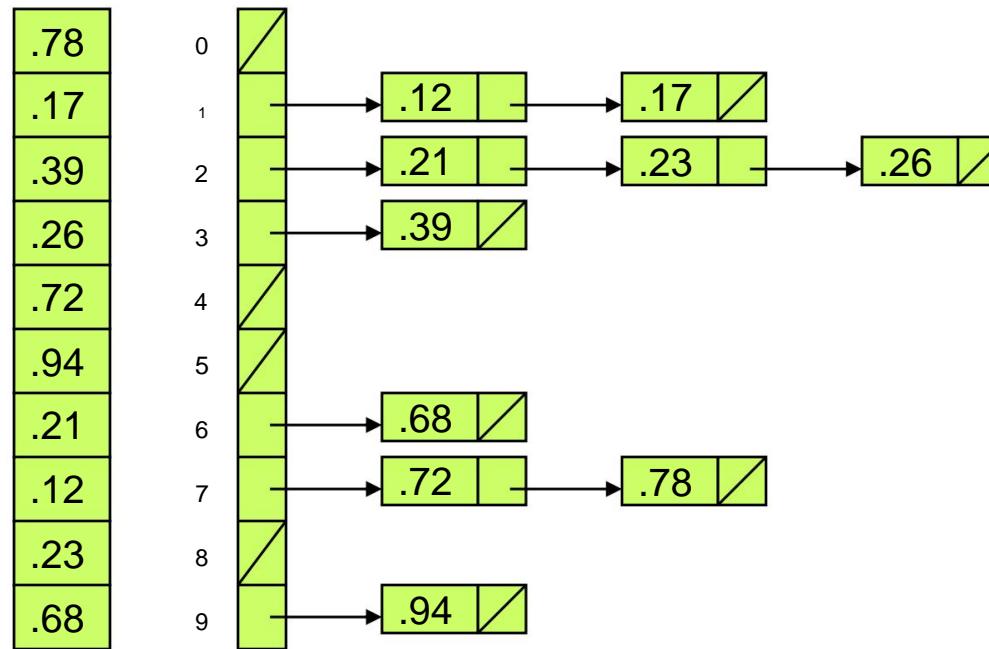
Visit the buckets along the interval and output the sorted items

# Bucket sort example

Input vector A of size 10

Bucket Vector B manages elements via linear lists

Bucket B[i] contains the values of the interval  $[i/10, (i+1)/10)$



# Bucket sort algorithm

**n = length(A);**

**for( $i=1$ ;  $i \leq n$ ;  $i++$ ) insert  $A[i]$**

**into list  $B[\lfloor n * A[i] \rfloor]$ ;**

**for( $i=0$ ;  $i < n$ ;  $i++$ )**

**sort list  $B[i]$  with insertion sort;**

**Concatenate the lists  $B[0], B[1], \dots, B[n-1]$**



# Analyse Bucket Sort

All statements except sort are of order  $O(n)$

## Analysis of sorting

$n_i$  denotes the random number of elements per bucket

Selection Sort runs in  $O(n^2)$  time, i.e. the expected time for sorting of a bucket is  $E[O(n^2)] = O(E[n^2])$ , since all buckets have to be sorted

$$\sum_{i=0}^{n-1} O(\eta) \cdot \sum_{j=i+1}^{n-1} n_i O(\eta)$$

To calculate the expression

Determine distribution of the random variable

Assume that the probability of an element falling into a bucket is  $p = 1/n$  (e.g. elements in  $[0,1]$  equally distributed)  $n_i$  follows the binomial distribution  $E[n_i] =$

$$n \cdot p = 1 \text{ and } \text{Var}[n_i] = n \cdot p \cdot (1-p) = 1 - 1/n$$

$$E[n_i^2] = \text{Var}[n_i] + E^2[n_i] = 1 - 1/n + 1/2 = 2 - 1/n = \bar{\eta}(1)$$

It follows that the effort for bucket sort is below Assumption ( ) is

# What do we take with us?

Dictionary

Hashing

Static and dynamic procedures

sort by

Classic methods  $O(n^2)$  and  $O(n \log n)$

Linear methods  $O(n)$