4190.101 **Discrete Mathematics**

Chapter 8 Advanced Counting Techniques

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Chapter Summary

- Applications of Recurrence Relations
- Solving Linear Recurrence Relations
 - Homogeneous Recurrence Relations
 - Nonhomogeneous Recurrence Relations
- Divide-and-Conquer Algorithms and Recurrence Relations
- Generating Functions
- Inclusion-Exclusion
- Applications of Inclusion-Exclusion

Applications of Recurrence Relations

Section 8.1

Section Summary

- Applications of Recurrence Relations
 - Fibonacci Numbers
 - The Tower of Hanoi
 - Counting Problems

Recurrence Relations (recalling definitions from Chapter 2)

- **Definition:** A recurrence relation for the sequence $\{a_n\}$ is an equation that expresses a_n in terms of one or more of the previous terms of the sequence, namely, a_0 , a_1 , ..., a_{n-1} , for all integers n with $n \ge n_0$, where n_0 is a non-negative integer.
- A sequence is called a solution of a recurrence relation if its terms satisfy the recurrence relation.
- The initial conditions for a sequence specify the terms that precede the first term where the recurrence relation takes effect.

Rabbits and the Fiobonacci Numbers

- Example: A young pair of rabbits (one of each gender) is placed on an island. A pair of rabbits does not breed until they are 2 months old. After they are 2 months old, each pair of rabbits produces another pair each month. Find a recurrence relation for the number of pairs of rabbits on the island after *n* months, assuming that rabbits never die.
- This is the original problem considered by Leonardo Pisano (Fibonacci) in the thirteenth century.

Rabbits and the Fiobonacci Numbers

Modeling the Population Growth of Rabbits on an Island

Reproducing pairs (at least two months old)	Young pairs (less than two months old)	Month	Reproducing pairs	Young pairs	Total pairs
		1	0	1	1
		2	0	1	1
0 40	0° 40	3	1	1	2
0 40	e to et to	4	1	2	3
0 40 0 40 0 40 0 40	040040	5	2	3	5
0 40 00 40 00 40 00 40 00 00 00 00 00 00	040040	6	3	5	8
	at a at a				

Rabbits and the Fiobonacci Numbers

- **Solution**: Let f_n be the number of pairs of rabbits after n months.
 - There are is f_1 = 1 pairs of rabbits on the island at the end of the first month.
 - We also have f_2 = 1 because the pair does not breed during the first month.
 - To find the number of pairs on the island after n months, add the number on the island after the previous month, f_{n-1} , and the number of newborn pairs, which equals f_{n-2} , because each newborn pair comes from a pair at least two months old.
- Consequently the sequence $\{f_n\}$ satisfies the recurrence relation $f_n = f_{n-1} + f_{n-2}$ for $n \ge 3$ with the initial conditions $f_1 = 1$ and $f_2 = 1$.
- The number of pairs of rabbits on the island after n months is given by the n-th Fibonacci number

The Tower of Hanoi

- The puzzle consisting of three pegs on a board with disks of different sizes. Initially all of the disks are on the first peg in order of size, with the largest on the bottom.
- Rules: You are allowed to move the disks one at a time from one peg to another as long as a larger disk is never placed on a smaller.
- Goal: Using allowable moves, end up with all the disks on the second peg in order of size with largest on the bottom.

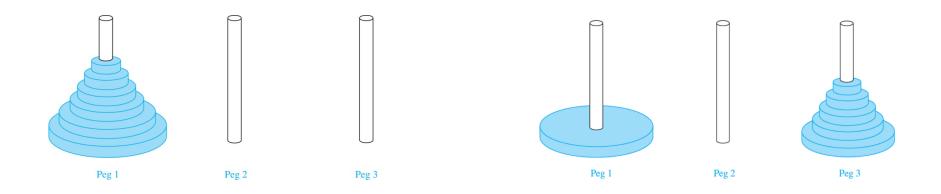
Peg 2

Peg 3

Peg 1

The Tower of Hanoi

- Let H_n denote the number of moves needed to solve the Tower of Hanoi problem with n disks. Set up a recurrence relation for the sequence.
- **Solution**: Begin with *n* disks on peg 1.
 - We can transfer the top n-1 disks to peg 3 using H_{n-1} moves.
 - Then, we use one move to transfer the largest disk to peg 2.
 - We can transfer the n-1 disks on peg 3 to peg 2 using H_{n-1} additional moves.
 - Therefore, $H_n = 2H_{n-1} + 1$.



The Tower of Hanoi

• We can use an iterative approach to solve this recurrence relation by repeatedly expressing H_n in terms of the previous terms of the sequence.

```
H_{n} = 2H_{n-1} + 1

= 2(2H_{n-2} + 1) + 1 = 2^{2}H_{n-2} + 2 + 1

= 2^{2}(2H_{n-3} + 1) + 2 + 1 = 2^{3}H_{n-3} + 2^{2} + 2 + 1

\vdots

= 2^{n-1}H_{1} + 2^{n-2} + 2^{n-3} + \dots + 2 + 1

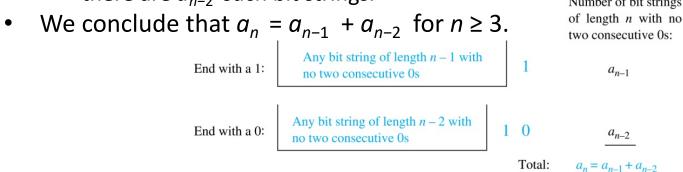
= 2^{n-1} + 2^{n-2} + 2^{n-3} + \dots + 2 + 1 because H_{1} = 1

= 2^{n} - 1 using the formula for the sum of the terms of a geometric series
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- There was a myth created with the puzzle. Monks in a tower in Hanoi are transferring 64 gold disks from one peg to another following the rules of the puzzle. They move one disk each day. When the puzzle is finished, the world will end.
- Using this formula for the 64 gold disks of the myth, 2^{64} −1 = 18,446, 744,073, 709,551,615 days are needed to solve the puzzle (more than 500 billion years).

Counting Bit Strings

- Example 3: Find a recurrence relation and give initial conditions for the number of bit strings of length n without two consecutive 0s. How many such bit strings are there of length five?
- **Solution**: Let a_n denote the number of bit strings of length n without two consecutive 0s. To obtain a recurrence relation for $\{a_n\}$ the number of bit strings of length n that do not have two consecutive 0s is the number of bit strings ending with a 0 plus the number of such bit strings ending with a 1.
- Now assume that $n \ge 3$.
 - The bit strings of length n ending with 1 without two consecutive 0s are the bit strings of length n-1 with no two consecutive 0s with a 1 at the end. Hence, there are a_{n-1} such bit strings.
 - The bit strings of length n ending with 0 without two consecutive 0s are the bit strings of length n -2 with no two consecutive 0s with 10 at the end. Hence, there are a_{n-2} such bit strings.



Counting Bit Strings

- The initial conditions are:
 - $-a_1 = 2$, since both the bit strings 0 and 1 do not have consecutive 0s.
 - $-a_2$ = 3, since the bit strings 01, 10, and 11 do not have consecutive 0s, while 00 does.
- To obtain a_5 , we use the recurrence relation three times to find that:

$$-a_3 = a_2 + a_1 = 3 + 2 = 5$$

$$-a_4 = a_3 + a_2 = 5 + 3 = 8$$

$$-a_5 = a_4 + a_3 = 8 + 5 = 13$$

Note that $\{a_n\}$ satisfies the same recurrence relation as the Fibonacci sequence. Since $a_1 = f_3$ and $a_2 = f_4$, we conclude that $a_n = f_{n+2}$.

Counting Bit Strings

- **Example**: Find a recurrence relation for C_n , the number of ways to parenthesize the product of n+1 numbers, $x_0 \cdot x_1 \cdot x_2 \cdot \cdots \cdot x_n$, to specify the order of multiplication.
 - $((x_0\cdot x_1)\cdot x_2)\cdot x_3$, $(x_0\cdot (x_1\cdot x_2))\cdot x_3$, $(x_0\cdot x_1)\cdot (x_2\cdot x_3)$, $x_0\cdot ((x_1\cdot x_2)\cdot x_3)$, $x_0\cdot (x_1\cdot (x_2\cdot x_3))$, ...
- **Solution**: Note that however parentheses are inserted in $x_0 \cdot x_1 \cdot x_2 \cdot \cdots \cdot x_n$, one "·" operator remains outside all parentheses. This final operator appears between two of the n+1 numbers, say x_k and x_{k+1} . Since there are C_k ways to insert parentheses in the product $x_0 \cdot x_1 \cdot x_2 \cdot \cdots \cdot x_k$ and C_{n-k-1} ways to insert parentheses in the product $x_{k+1} \cdot x_{k+2} \cdot \cdots \cdot x_n$, we have

$$C_n = C_0 C_{n-1} + C_1 C_{n-2} + \dots + C_{n-2} C_1 + C_{n-1} C_0$$
$$= \sum_{k=0}^{n-1} C_k C_{n-k-1}$$

• The initial conditions are $C_0 = 1$ and $C_1 = 1$.

The sequence $\{C_n\}$ is the sequence of **Catalan Numbers**. This recurrence relation can be solved using the method of generating functions; see Exercise 41 in Section 8.4.

Solving Linear Recurrence Relations

Section 8.2

Section Summary

- Linear Homogeneous Recurrence Relations
- Solving Linear Homogeneous Recurrence Relations with Constant Coefficients.
- Solving Linear Nonhomogeneous Recurrence Relations with Constant Coefficients.

Linear Homogeneous Recurrence Relations

- **Definition:** A linear homogeneous recurrence relation of degree k with constant coefficients is a recurrence relation of the form $a_n = c_1 a_{n-1} + c_2 a_{n-2} + \dots + c_k a_{n-k}$ where c_1, c_2, \dots, c_k are real numbers, and $c_k \neq 0$
 - It is *linear* because the right-hand side is a sum of the previous terms of the sequence each multiplied by a function of *n*.
 - It is *homogeneous* because no terms occur that are not multiples of the a_j s. Each coefficient is a constant.
 - The *degree* is k because a_n is expressed in terms of the previous k terms of the sequence.

By strong induction, a sequence satisfying such a recurrence relation is uniquely determined by the recurrence relation and the k initial conditions $a_0 = C_1$, $a_0 = C_1$, ..., $a_{k-1} = C_{k-1}$.

Examples of Linear Homogeneous Recurrence Relations

•
$$P_n = (1.11)P_{n-1}$$

Linear homogeneous recurrence relation of degree one

•
$$f_n = f_{n-1} + f_{n-2}$$

Linear homogeneous recurrence relation of degree two

•
$$a_n = a_{n-1} + a_{n-2}^2$$

Not linear

•
$$H_n = 2H_{n-1} + 1$$

Not homogeneous

•
$$B_n = nB_{n-1}$$

Coefficients are not constants

Solving Linear Homogeneous Recurrence Relations

- The basic approach is to look for solutions of the form $a_n = r^n$, where r is a constant.
- Note that $a_n = r^n$ is a solution to the recurrence relation $a_n = c_1 a_{n-1} + c_2 a_{n-2} + \cdots + c_k a_{n-k}$ if and only if $r^n = c_1 r^{n-1} + c_2 r^{n-2} + \cdots + c_k r^{n-k}$.
- Algebraic manipulation yields the *characteristic* equation: $r^k c_1 r^{k-1} c_2 r^{k-2} \cdots c_{k-1} r c_k = 0$.
- The sequence $\{a_n\}$ with $a_n = r^n$ is a solution if and only if r is a solution to the characteristic equation.
- The solutions to the characteristic equation are called the characteristic roots of the recurrence relation. The roots are used to give an explicit formula for all the solutions of the recurrence relation.

Solving Linear Homogeneous Recurrence Relations of Degree Two

• **Theorem 1**: Let c_1 and c_2 be real numbers. Suppose that $r^2 - c_1 r - c_2 = 0$ has two distinct roots r_1 and r_2 . Then the sequence $\{a_n\}$ is a solution to the recurrence relation $a_n = c_1 a_{n-1} + c_2 a_{n-2}$ if and only if

$$a_n = \alpha r_1^n + \alpha_2 r_2^n$$

for n = 0, 1, 2, ..., where α_1 and α_2 are constants.

Using Theorem 1

- **Example**: What is the solution to the recurrence relation $a_n = a_{n-1} + 2a_{n-2}$ with $a_0 = 2$ and $a_1 = 7$?
- **Solution**: The characteristic equation is $r^2 r 2 = 0$. Its roots are r = 2 and r = -1. Therefore, $\{a_n\}$ is a solution to the recurrence relation if and only if $a_n = \alpha_1 2^n + \alpha_2 (-1)^n$, for some constants α_1 and α_2 .
- To find the constants α_1 and α_2 , note that $a_0 = 2 = \alpha_1 + \alpha_2$ and $a_1 = 7 = \alpha_1 2 + \alpha_2 (-1)$.
- Solving these equations, we find $\alpha_1 = 3$ and $\alpha_2 = -1$.
- Hence, the solution is the sequence $\{a_n\}$ with $a_n = 3 \cdot 2^n (-1)^n$.

An Explicit Formula for the Fibonacci Numbers

- We can use Theorem 1 to find an explicit formula for the Fibonacci numbers. The sequence of Fibonacci numbers satisfies the recurrence relation $f_n = f_{n-1} + f_{n-2}$ with the initial conditions: $f_0 = 0$ and $f_1 = 1$.
- **Solution**: The roots of the characteristic equation $r^2 r 1 = 0$ are

$$r_1 = \frac{1+\sqrt{5}}{2}$$

$$r_2 = \frac{1 - \sqrt{5}}{2}$$

Fibonacci Numbers

Therefore by Theorem 1

$$f_n = \alpha_1 \left(\frac{1 + \sqrt{5}}{2} \right)^n + \alpha_2 \left(\frac{1 - \sqrt{5}}{2} \right)^n$$

for some constants α_1 and α_2 .

• Using the initial conditions $f_0 = 0$ and $f_1 = 1$, we have

$$f_0 = \alpha_1 + \alpha_2 = 0$$

$$f_1 = \alpha_1 \left(\frac{1 + \sqrt{5}}{2} \right) + \alpha_2 \left(\frac{1 - \sqrt{5}}{2} \right) = 1$$

• Solving, we obtain $\alpha_1 = \frac{1}{\sqrt{5}}$, $\alpha_2 = -\frac{1}{\sqrt{5}}$.

• Hence,
$$f_n = \frac{1}{\sqrt{5}} \left(\frac{1+\sqrt{5}}{2} \right)^n - \frac{1}{\sqrt{5}} \left(\frac{1-\sqrt{5}}{2} \right)^n$$

The Solution When There is a Repeated Root

• **Theorem 2**: Let c_1 and c_2 be real numbers with $c_2 \neq 0$. Suppose that $r^2 - c_1 r - c_2 = 0$ has one repeated root r_0 . Then the sequence $\{a_n\}$ is a solution to the recurrence relation $a_n = c_1 a_{n-1} + c_2 a_{n-2}$ if and only if

$$a_n = \alpha r_0^n + \alpha_2 n r_0^n$$

for n = 0,1,2,..., where α_1 and α_2 are constants.

Using Theorem 2

- **Example**: What is the solution to the recurrence relation $a_n = 6a_{n-1} 9a_{n-2}$ with $a_0 = 1$ and $a_1 = 6$?
- **Solution**: The characteristic equation is $r^2 6r + 9 = 0$.
- The only root is r = 3. Therefore, $\{a_n\}$ is a solution to the recurrence relation if and only if

$$a_n = \alpha_1 3^n + \alpha_2 n(3)^n$$

where α_1 and α_2 are constants.

- To find the constants α_1 and α_2 , note that $a_0 = 1 = \alpha_1$ and $a_1 = 6 = \alpha_1 \cdot 3 + \alpha_2 \cdot 3$.
- Solving, we find that $\alpha_1 = 1$ and $\alpha_2 = 1$.
- Hence, $a_n = 3^n + n3^n$.

Solving Linear Homogeneous Recurrence Relations of Arbitrary Degree

- This theorem can be used to solve linear homogeneous recurrence relations with constant coefficients of any degree when the characteristic equation has distinct roots.
- Theorem 3: Let c_1 , c_2 ,..., c_k be real numbers. Suppose that the characteristic equation

$$r^k - c_1 r^{k-1} - \dots - c_k = 0$$

has k distinct roots r_1 , r_2 , ..., r_k . Then a sequence $\{a_n\}$ is a solution of the recurrence relation

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \dots + c_k a_{n-k}$$

if and only if

$$a_n = \alpha_1 r_1^n + \alpha_2 r_2^n + \dots + \alpha_k r_k^n$$

for n = 0, 1, 2, ..., where $\alpha_1, \alpha_2, ..., \alpha_k$ are constants.

The General Case with Repeated Roots Allowed

• **Theorem 4**: Let $c_1, c_2, ..., c_k$ be real numbers. Suppose that the characteristic equation

$$r^k - c_1 r^{k-1} - \cdots - c_k = 0$$

has t distinct roots $r_1, r_2, ..., r_t$ with multiplicities $m_1, m_2, ..., m_t$, respectively so that $m_i \ge 1$ for i = 0, 1, ..., t and $m_1 + m_2 + ... + m_t = k$. Then a sequence $\{a_n\}$ is a solution of the recurrence relation

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \dots + c_k a_{n-k}$$

if and only if

$$a_{n} = (\alpha_{1,0} + \alpha_{1,1}n + \dots + \alpha_{1,m_{1}-1}n^{m_{1}-1})r_{1}^{n}$$

$$+(\alpha_{2,0} + \alpha_{2,1}n + \dots + \alpha_{2,m_{2}-1}n^{m_{2}-1})r_{2}^{n}$$

$$+\dots + (\alpha_{t,0} + \alpha_{t,1}n + \dots + \alpha_{t,m_{t}-1}n^{m_{t}-1})r_{t}^{n}$$

for n = 0, 1, 2, ..., where $\alpha_{i,j}$ are constants for $1 \le i \le t$ and $0 \le j \le m_{i-1}$.

Linear Nonhomogeneous Recurrence Relations with Constant Coefficients

 Definition: A linear nonhomogeneous recurrence relation with constant coefficients is a recurrence relation of the form:

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \dots + c_k a_{n-k} + F(n)$$

where c_1 , c_2 ,, c_k are real numbers, and F(n) is a function not identically zero depending only on n.

The recurrence relation

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \dots + c_k a_{n-k}$$

is called the associated homogeneous recurrence relation.

Linear Nonhomogeneous Recurrence Relations with Constant Coefficients

 The following are linear nonhomogeneous recurrence relations with constant coefficients:

$$a_n = a_{n-1} + 2^n$$
,
 $a_n = a_{n-1} + a_{n-2} + n^2 + n + 1$,
 $a_n = 3a_{n-1} + n3^n$,
 $a_n = a_{n-1} + a_{n-2} + a_{n-3} + n!$
where the following are the associated linear homogeneous recurrence relations, respectively:

$$a_n = a_{n-1}$$
,
 $a_n = a_{n-1} + a_{n-2}$,
 $a_n = 3a_{n-1}$,
 $a_n = a_{n-1} + a_{n-2} + a_{n-3}$

Solving Linear Nonhomogeneous Recurrence Relations with Constant Coefficients

• **Theorem 5**: If $\{a_n^{(p)}\}$ is a particular solution of the nonhomogeneous linear recurrence relation with constant coefficients

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \cdots + c_k a_{n-k} + F(n)_n$$

then every solution is of the form $\{a_n^{(p)} + a_n^{(h)}\}$, where $\{a_n^{(h)}\}$ is a solution of the associated homogeneous recurrence relation

$$a_n = c_1 a_{n-1} + c_2 a_{n-2} + \cdots + c_k a_{n-k}$$

Solving Linear Nonhomogeneous Recurrence Relations with Constant Coefficients

- **Example**: Find all solutions of the recurrence relation $a_n = 3a_{n-1} + 2n$.
- **Solution**: The associated linear homogeneous equation is $a_n = 3a_{n-1}$. Its solutions are $a_n^{(h)} = \alpha 3^n$, where α is a constant.
- Because F(n) = 2n is a polynomial in n of degree one, to find a particular solution we might try a linear function in n, say $p_n = cn + d$, where c and d are constants. Suppose that $p_n = cn + d$ is such a solution.
- Then $a_n = 3a_{n-1} + 2n$ becomes cn + d = 3(c(n-1) + d) + 2n.
- Simplifying yields (2 + 2c)n + (2d 3c) = 0. It follows that cn + d is a solution if and only if 2 + 2c = 0 and 2d 3c = 0. Therefore, cn + d is a solution if and only if c = -1 and d = -3/2.
- Consequently, $a_n^{(p)} = -n 3/2$ is a particular solution.

Solving Linear Nonhomogeneous Recurrence Relations with Constant Coefficients

- **Example**: Find all solutions of the recurrence relation $a_n = 3a_{n-1} + 2n$.
- **Solution**: The associated linear homogeneous equation is $a_n = 3a_{n-1}$. Its solutions are $a_n^{(h)} = \alpha 3^n$, where α is a constant.
- By Theorem 5, all solutions are of the form $a_n = a_n^{(p)} + a_n^{(h)} = -n 3/2 + \alpha 3^n$, where α is a constant.
- Let's find the solution with $a_1 = 3$. Then $3 = -1 3/2 + 3 \alpha$, and $\alpha = 11/6$.
- Hence, the solution is $a_n = -n 3/2 + (11/6)3^n$.

Divide-and-Conquer Algorithms and Recurrence Relations

Section 8.3

Section Summary

- Divide-and-Conquer Algorithms and Recurrence Relations
- Examples
 - Binary Search
 - Merge Sort
 - Fast Multiplication of Integers
- Master Theorem
- Closest Pair of Points (not covered yet in these slides)

Divide-and-Conquer Algorithmic Paradigm

• **Definition**: A divide-and-conquer algorithm works by first dividing a problem into one or more instances of the same problem of smaller size and then conquering the problem using the solutions of the smaller problems to find a solution of the original problem.

Examples:

- Binary search, covered in Chapters 3 and 5: It works by comparing the element to be located to the middle element. The original list is then split into two lists and the search continues recursively in the appropriate sublist.
- Merge sort, covered in Chapter 5: A list is split into two approximately equal sized sublists, each recursively sorted by merge sort. Sorting is done by successively merging pairs of lists.

Divide-and-Conquer Recurrence Relations

- Suppose that a recursive algorithm divides a problem of size n into a subproblems.
- Assume each subproblem is of size n/b.
- Suppose g(n) extra operations are needed in the conquer step.
- Then f(n), the number of operations to solve a problem of size n, satisfies the following recurrence relation:

$$f(n) = af(n/b) + g(n)$$

• This is called a divide-and-conquer recurrence relation.

Example: Binary Search

- Binary search reduces the search for an element in a sequence of size n to the search in a sequence of size n/2. Two comparisons are needed to implement this reduction;
 - One to decide whether to search the upper or lower half of the sequence and
 - The other to determine if the sequence has elements.
- Hence, if f(n) is the number of comparisons required to search for an element in a sequence of size n, then

$$f(n) = f(n/2) + 2$$

when *n* is even.

Example: Merge Sort

- The merge sort algorithm splits a list of n
 (assuming n is even) items to be sorted into two
 lists with n/2 items. It uses fewer than n
 comparisons to merge the two sorted lists.
- Hence, the number of comparisons required to sort a sequence of size n, is no more than than M(n) where

$$M(n) = 2M(n/2) + n.$$

Example: Fast Multiplication of Integers

- An algorithm for the fast multiplication of two 2n-bit integers (assuming n is even) first splits each of the 2n-bit integers into two blocks, each of n bits.
- Suppose that a and b are integers with binary expansions of length 2n. Let

$$a = (a_{2n-1}a_{2n-2} \dots a_1a_0)_2$$
 and $b = (b_{2n-1}b_{2n-2} \dots b_1b_0)_2$.

• Let $a = 2^n A_1 + A_0$, $b = 2^n B_1 + B_0$, where

$$A_1 = (a_{2n-1} \dots a_{n+1} a_n)_2$$
, $A_0 = (a_{n-1} \dots a_1 a_0)_2$,

$$B_1 = (b_{2n-1} \dots b_{n+1} b_n)_2$$
, $B_0 = (b_{n-1} \dots b_1 b_0)_2$.

The algorithm is based on the fact that ab can be rewritten as:

$$ab = (2^{2n} + 2^n)A_1B_1 + 2^n(A_1 - A_0)(B_0 - B_1) + (2^n + 1)A_0B_0.$$

- This identity shows that the multiplication of two 2*n*-bit integers can be carried out using three multiplications of *n*-bit integers, together with additions, subtractions, and shifts.
- Hence, if f(n) is the total number of operations needed to multiply two n-bit integers, then

$$f(2n) = 3f(n) + Cn$$

where *Cn* represents the total number of bit operations; the additions, subtractions and shifts that are a constant multiple of *n*-bit operations.

Estimating the Size of Divide-and-Conquer Functions

 Theorem 1: Let f be an increasing function that satisfies the recurrence relation

$$f(n) = af\left(\frac{n}{b}\right) + c$$

whenever n is divisible by b, where $a \ge 1$, b is an integer greater than 1, and c is a positive real number.

Then

$$f(n) = \begin{cases} O(n^{\log_b a}) & \text{if } a > 1\\ O(\log n) & \text{if } a = 1 \end{cases}$$

• Furthermore, when $n = b^k$ and $a \neq 1$, where k is a positive integer,

$$f(n) = C_1 n^{\log_b a} + C_2$$

where $C_1 = f(1) + c/(a-1)$ and 2 = -c/(a-1).

Complexity of Binary Search

- **Binary Search Example**: Give a big-O estimate for the number of comparisons used by a binary search.
- **Solution**: Since the number of comparisons used by binary search is f(n) = f(n/2) + 2 where n is even, by Theorem 1, it follows that f(n) is $O(\log n)$.

Estimating the Size of Divide-andconquer Functions

 Theorem 2. Master Theorem: Let f be an increasing function that satisfies the recurrence relation

$$f(n) = af\left(\frac{n}{b}\right) + cn^d$$

whenever $n=b^k$, where k is a positive integer greater than 1, and c and d are real numbers with c positive and d nonnegative. Then

$$f(n) = \begin{cases} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Complexity of Merge Sort

- Merge Sort example: Give a big-O estimate for the number of comparisons used by merge sort.
- **Solution**: Since the number of comparisons used by merge sort to sort a list of n elements is less than M(n) where M(n) = 2M(n/2) + n, by the master theorem M(n) is $O(n \log n)$.

Complexity of Fast Integer Multiplication Algorithm

- Integer Multiplication Example: Give a big-O estimate for the number of bit operations used needed to multiply two n-bit integers using the fast multiplication algorithm.
- **Solution**: We have shown that f(n) = 3f(n/2) + Cn, when n is even, where f(n) is the number of bit operations needed to multiply two n-bit integers. Hence by the master theorem with a = 3, b = 2, c = C, and d = 1 (so that we have the case where $a > b^d$), it follows that f(n) is $O(n^{\log 3})$.
- Note that $\log 3 \approx 1.6$. Therefore the fast multiplication algorithm is a substantial improvement over the conventional algorithm that uses $O(n^2)$ bit operations.

Inclusion-Exclusion

Section 8.5

Section Summary

- The Principle of Inclusion-Exclusion
- Examples

Principle of Inclusion-Exclusion

 In Section 2.2, we developed the following formula for the number of elements in the union of two finite sets:

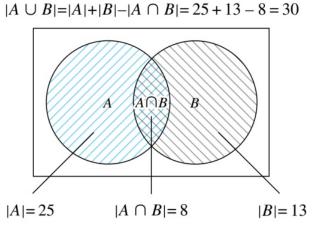
$$|A \cup B| = |A| + |B| - |A \cap B|$$

 We will generalize this formula to finite sets of any size.

Two Finite Sets

• Example: In a discrete math class every student is a major in computer science (CS) or math or both. The number of students having CS as a major (possibly along with math) is 25; the number of students having math as a major (possibly along with CS) is 13; and the number of students majoring in both CS and math is 8. How many students are in the class?

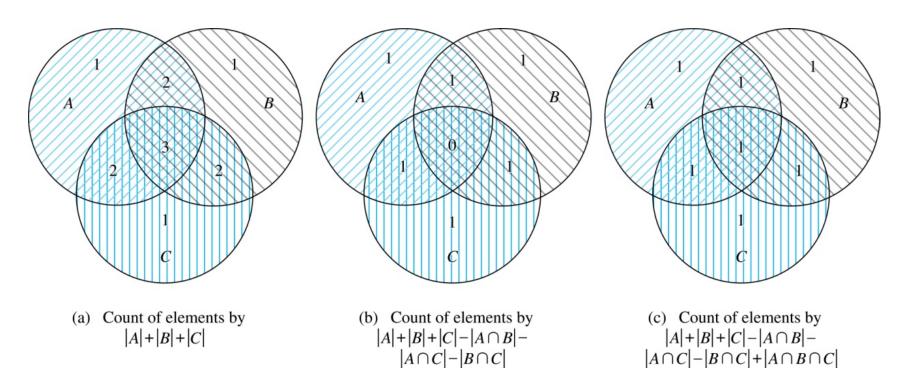
• Solution: $|A \cup B| = |A| + |B| - |A \cap B|$ = 25 + 13 -8 = 30



Three Finite Sets

$$|A \cup B \cup C| =$$

 $|A| + |B| + |C| - |A \cap B| - |A \cap C| - |B \cap C| + |A \cap B \cap C|$



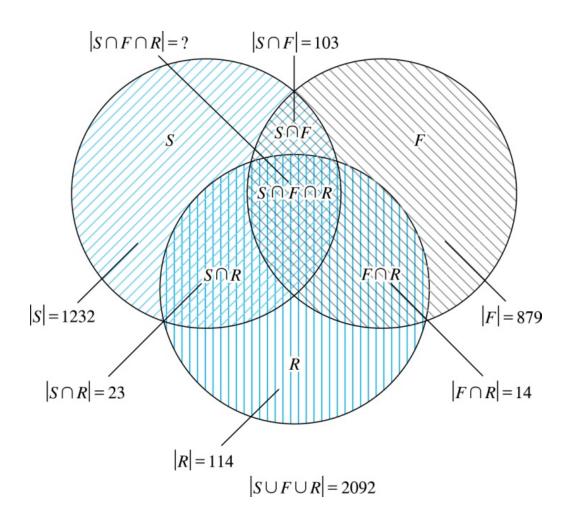
Three Finite Sets

- Example: 1232 students have taken a course in Spanish, 879 have taken a course in French, and 114 have taken a course in Russian. Further, 103 have taken courses in both Spanish and French, 23 have taken courses in both Spanish and Russian, and 14 have taken courses in both French and Russian. If 2092 students have taken a course in at least one of Spanish, French and Russian, how many students have taken a course in all 3 languages.
- **Solution**: Let *S* be the set of students who have taken a course in Spanish, *F* the set of students who have taken a course in French, and *R* the set of students who have taken a course in Russian. Then, |S| = 1232, |F| = 879, |R| = 114, $|S \cap F| = 103$, $|S \cap R| = 23$, $|F \cap R| = 14$, and $|S \cup F \cup R| = 23$.

Using the equation

 $|S \cup F \cup R| = |S| + |F| + |R| - |S \cap F| - |S \cap R| - |F \cap R| + |S \cap F \cap R|$, we obtain 2092 = 1232 + 879 + 114 -103 -23 -14 + $|S \cap F \cap R|$. Solving for $|S \cap F \cap R|$ yields 7.

Illustration of Three Finite Set Example



The Principle of Inclusion-Exclusion

• Theorem 1. The Principle of Inclusion-Exclusion: Let $A_1, A_2, ..., A_n$ be finite sets. Then:

$$|A_1 \cup A_2 \cup \cdots \cup A_n| =$$

$$\sum_{1 \le i \le n} |A_i| - \sum_{1 \le i \le j \le n} |A_i \cap A_j| +$$

$$\sum_{1 \le i \le j \le k \le n} |A_i \cap A_j \cap A_k| - \dots + (-1)^{n+1} |A_1 \cap A_2 \cap \dots \cap A_n|$$

The Principle of Inclusion-Exclusion

- Proof: An element in the union is counted exactly once in the right-hand side of the equation.
 Consider an element a that is a member of r of the sets A₁,..., A_n where 1≤ r ≤ n.
 - It is counted C(r,1) times by $\sum |A_i|$
 - It is counted C(r,2) times by $\sum |A_i \cap A_j|$
 - In general, it is counted C(r,m) times by the summation of m of the sets A_i .

The Principle of Inclusion-Exclusion

- Thus the element is counted exactly $C(r,1) C(r,2) + C(r,3) \cdots + (-1)^{r+1} C(r,r)$
 - times by the right hand side of the equation.
- By Corollary 2 of Section 6.4, we have $C(r,0) C(r,1) + C(r,2) \cdots + (-1)^{r+1} C(r,r) = 0.$
- Hence,

$$1 = C(r,0) = C(r,1) - C(r,2) + \cdots + (-1)^{r+1} C(r,r).$$

Applications of Inclusion-Exclusion

Section 8.6

Section Summary

- Counting Onto-Functions
- Derangements

The Number of Onto Functions

- **Example**: How many onto functions are there from a set with six elements to a set with three elements?
- **Solution**: Suppose that the elements in the codomain are b_1 , b_2 , and b_3 . Let P_1 , P_2 , and P_3 be the properties that b_1 , b_2 , and b_3 are not in the range of the function, respectively. The function is onto if none of the properties P_1 , P_2 , and P_3 hold.
- By the inclusion-exclusion principle the number of onto functions from a set with six elements to a set with three elements is $N [N(P_1) + N(P_2) + N(P_3)] + [N(P_1P_2) + N(P_1P_3) + N(P_2P_3)] N(P_1P_2P_3)$
 - Here the total number of functions from a set with six elements to one with three elements is $N = 3^6$.
 - The number of functions that do not have in the range is $N(P_1) = 2^6$. Similarly, $N(P_2) = N(P_3) = 2^6$.
 - Note that $N(P_1P_2) = N(P_1P_3) = N(P_2P_3) = 1$ and $N(P_1P_2P_3) = 0$.
- Hence, the number of onto functions from a set with six elements to a set with three elements is:

$$3^6 - 3 \cdot 2^6 + 3 = 729 - 192 + 3 = 540$$

The Number of Onto Functions

Theorem 1: Let m and n be positive integers with m ≥
 n. Then there are

$$n^m - C(n,1)(n-1)^m + C(n,2)(n-2)^m - \cdots + (-1)^{n-1}C(n,n-1) \cdot 1^m$$
 onto functions from a set with m elements to a set with n elements.

• Proof follows from the principle of inclusion-exclusion (see Exercise 27).

Derangements

- **Definition**: A *derangement* is a permutation of objects that leaves no object in the original position.
- Example: The permutation of 21453 is a derangement of 12345 because no number is left in its original position. But 21543 is not a derangement of 12345, because 4 is in its original position.

Derangements

 Theorem 2: The number of derangements of a set with n elements is

$$D_n = n! \left[1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \dots + (-1)^n \frac{1}{n!} \right].$$

 Proof follows from the principle of inclusionexclusion (see text)

Derangements

- The Hatcheck Problem: A new employee checks the hats of n people at restaurant, forgetting to put claim check numbers on the hats. When customers return for their hats, the checker gives them back hats chosen at random from the remaining hats. What is the probability that no one receives the correct hat?
- **Solution**: The answer is the number of ways the hats can be arranged so that there is no hat in its original position divided by n!, the number of permutations of n hats.

Remark: It can be shown that the probability of a derangement approaches 1/e as n grows without bound.

$$\frac{D_n}{n!} = \left[1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \dots + (-1)^n \frac{1}{n!}\right]$$

TABLE 1 The Probability of a Derangement.						
n	2	3	4	5	6	7
$D_n/n!$	0.50000	0.33333	0.37500	0.36667	0.36806	0.36786