



# ساختمان داده‌ها و الگوریتم‌ها

## فصل دوم

# فهرست مطالب

- ❖ مقدمه‌ای بر الگوریتم‌ها و مفاهیم پایه
- ❖ معرفی پیچیدگی زمانی و حافظه‌ای و روشهای تحلیل مسائل
- ❖ معرفی ساختمان داده‌های مقدماتی و الگوریتم‌های وابسته به آنها
  - آرایه
  - صف
  - پشته
  - لیست پیوندی
- ❖ تئوری درخت و گراف و الگوریتم‌های مرتبط
- ❖ الگوریتم‌های مرتب‌سازی و تحلیل پیچیدگی مربوط به آنها
- ❖ مباحث تکمیلی در ساختمان داده‌ها

# سوال

- اگر دو الگوریتم داشته باشیم که **یک هدف** را انجام دهند، چگونه باید فهمید که **کدام بهتر عمل می کنند؟**
- **معیار ارزیابی** چه پارامترهایی می تواند باشد؟

# Space Complexity

$$S(P) = C + S_p(I)$$

- Fixed Space Requirements (C)  
Independent of the characteristics of the inputs and outputs
  - instruction space
  - space for simple variables, fixed-size structured variable, constants
- Variable Space Requirements ( $S_p(I)$ )  
depend on the instance characteristic I
  - number, size, values of inputs and outputs associated with I
  - recursive stack space, formal parameters, local variables, return address

# Example

## Program 1.9: Simple arithmetic function

```
float abc(float a, float b, float c)
{
    return a + b + b * c + (a + b - c) / (a + b) + 4.00;
}
```


$$S_{abc}(I) = 0$$

---

## Program 1.10: Iterative function for summing a list of numbers

```
float sum(float list[ ], int n)
{
    float tempsum = 0;
    int i;
    for (i = 0; i < n; i++)
        tempsum += list[i];
    return tempsum;
}
```


$$S_{sum}(I) = 0$$

## Example (Cont.)

**\*Program 1.11: Recursive function** for summing a list of numbers (p.20)

```
float rsum(float list[ ], int n)
{
    if (n > 0)
        return rsum(list, n-1) + list[n-1];
    return 0;
}
```

$$S_{\text{rsum}}(I) = S_{\text{rsum}}(n) = 6n$$

Space needed for one recursive call of Program 1.11

Type	Name	Number of bytes
parameter: float *	list	2
parameter: integer	n	2
return address:(used internally)		2(unless a far address)
TOTAL per recursive call		6

# Time Complexity

$$T(P) = C + T_p(I)$$

- Compile time (C)  
independent of instance characteristics
- run (execution) time  $T_p$

- Definition

A *program step* is a syntactically or semantically meaningful program segment whose execution time is independent of the instance characteristics.

- Example

- $abc = a + b + b * c + (a + b - c) / (a + b) + 4.0$
- $abc = a + b + c$

# Methods to compute the step count

- Introduce **variable count** into programs
- **Tabular method**
  - Determine the total number of steps contributed by each statement  
 $\text{step per execution} \times \text{frequency}$
  - add up the contribution of all statements



# Variable Count Method (Example)

\*Program 1.12: Program 1.10 with count statements

$2n + 3$  Steps

```
float sum(float list[ ], int n)
{
    float tempsum = 0; count++; /* for assignment */
    int i;
    for (i = 0; i < n; i++) {
        count++; /*for the for loop */
        tempsum += list[i]; count++; /* for assignment */
    }
    count++; /* last execution of for */
    return tempsum;
    count++; /* for return */
}
```

# Variable Count Method (Example Cont.)

\*Program 1.13: Simplified version of Program 1.12

```
float sum(float list[ ], int n)
{
    float tempsum = 0;
    int i;
    for (i = 0; i < n; i++)
        count += 2;
    count += 3;
    return 0;
}
```



**2n + 3 Steps**

# Variable Count Method (Example Cont.)

\*Program 1.14: Program 1.11 with count statements added

```
float rsum(float list[ ], int n)
{
    count++;    /*for if conditional */
    if (n > 0) {
        count++; /* for return and rsum invocation */
        return rsum(list, n-1) + list[n-1];
    }
    count++;
    return 0;
}
```



**2n + 2 Steps**

# Variable Count Method (Example Cont.)

## \*Program 1.15: Matrix addition

```
void add( int a[ ][MAX_SIZE], int b[ ][MAX_SIZE],  
          int c [ ][MAX_SIZE], int rows, int cols)  
{  
    int i, j;  
    for (i = 0; i < rows; i++)  
        for (j= 0; j < cols; j++)  
            c[i][j] = a[i][j] +b[i][j];  
}
```

# Variable Count Method (Example Cont.)

\*Program 1.16: Matrix addition with count statements (p.25)

```
void add(int a[ ][MAX_SIZE], int b[ ][MAX_SIZE],
        int c[ ][MAX_SIZE], int row, int cols )
{
    int i, j;
    for (i = 0; i < rows; i++){
        count++; /* for i for loop */
        for (j = 0; j < cols; j++) {
            count++; /* for j for loop */
            c[i][j] = a[i][j] + b[i][j];
            count++; /* for assignment statement */
        }
        count++; /* last time of j for loop */
    }
    count++; /* last time of i for loop */
}
```

$2rows * cols + 2 rows + 1$

# Variable Count Method (Example Cont.)

## \*Program 1.17: Simplification of Program 1.16

```
void add(int a[ ][MAX_SIZE], int b [ ][MAX_SIZE],  
        int c[ ][MAX_SIZE], int rows, int cols)  
{  
    int i, j;  
    for( i = 0; i < rows; i++)  
    {  
        for (j = 0; j < cols; j++)  
            count += 2;  
        count += 2;  
    }  
    count++;  
}
```

$2rows * cols + 2 rows + 1$

Suggestion: **Interchange** the  
loops when rows >> cols

# Tabular Method (Example)

Iterative function to sum a list of numbers

steps/execution

Statement	s/e	Frequency	Total steps
float sum(float list[ ], int n)	0	0	0
{	0	0	0
float tempsum = 0;	1	1	1
int i;	0	0	0
for(i=0; i <n; i++)	1	n+1	n+1
tempsum += list[i];	1	n	n
return tempsum;	1	1	1
}	0	0	0
Total			2n+3

# Tabular Method (Example Cont.)

Step count table for recursive summing function

Statement	s/e	Frequency	Total steps
float rsum(float list[ ], int n)	0	0	0
{	0	0	0
if (n)	1	$n+1$	$n+1$
return rsum(list, n-1)+list[n-1];	1	$n$	$n$
return list[0];	1	1	1
}	0	0	0
Total			$2n+2$



# Tabular Method (Example Cont.)

## Step count table for matrix addition

Statement	s/e	Frequency	Total steps
Void add (int a[ ][MAX_SIZE] . . . )	0	0	0
{	0	0	0
int i, j;	0	0	0
for (i = 0; i < row; i++)	1	rows+1	rows+1
for (j=0; j< cols; j++)	1	rows • (cols+1)	rows • cols+rows
c[i][j] = a[i][j] + b[i][j];	1	rows • cols	rows • cols
}	0	0	0
Total			2rows • cols+2rows+1



# Home Work 1

Trade off between time and space complexity

# Asymptotic Notation

- Determining step counts help us to **compare the time complexities of two programs** and to **predict the growth in run time** as the instance characteristics change.
- But determining exact step counts could be very **difficult**. Since the notion of a step count is itself **inexact**, it may be worth the effort to compute the exact step counts.
- How does the algorithm behave as the problem size gets very large?

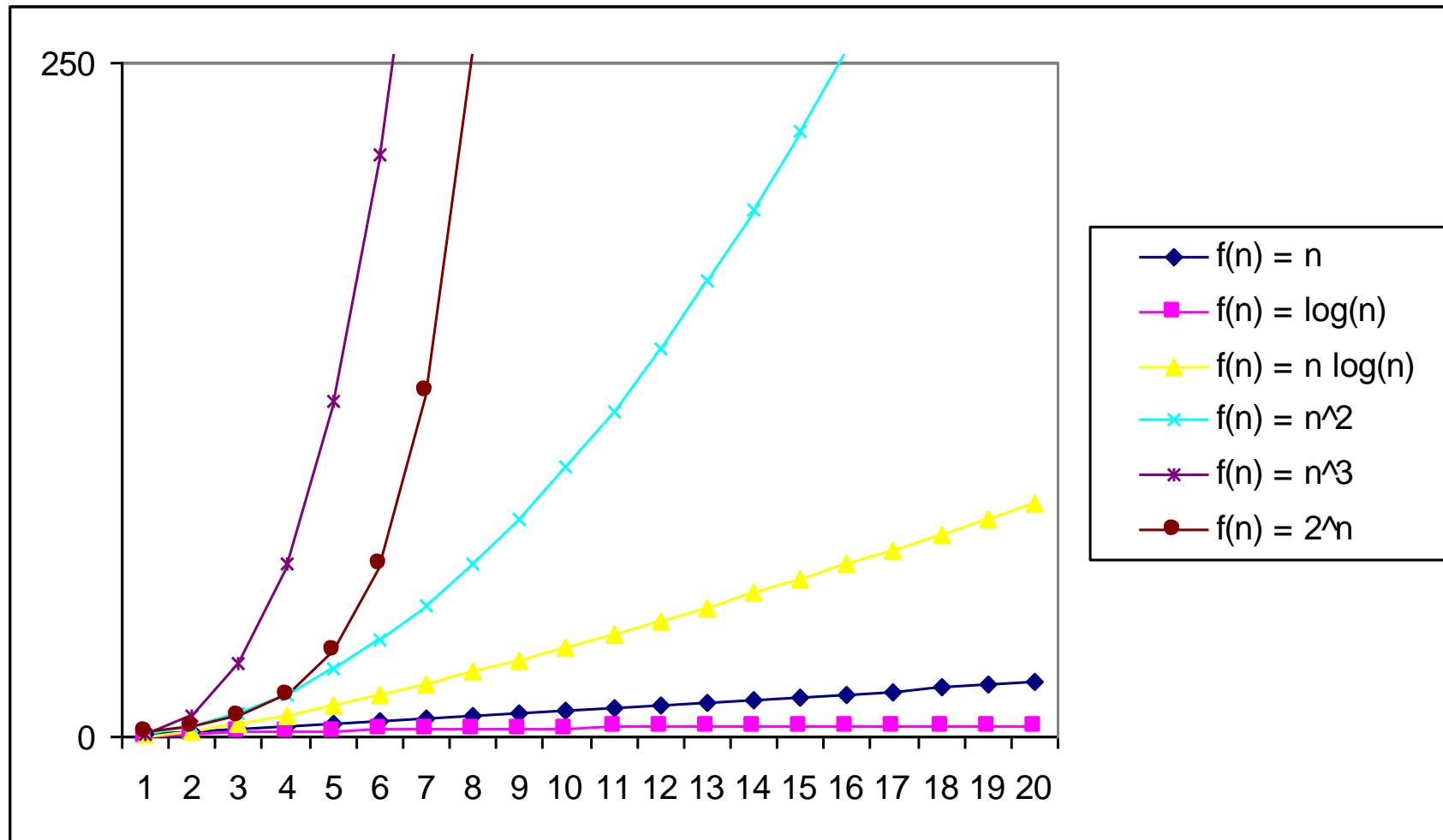
# Asymptotic Notation ( $O$ )

- Definition (Big Oh)
- $f(n) = O(g(n))$  iff there exist positive constants  $c$  and  $n_0$  such that  $f(n) \leq cg(n)$  for all  $n$ ,  $n \geq n_0$ .
- Examples
  - $3n+2=O(n)$  /\*  $3n+2 \leq 4n$  for  $n \geq 2$  \*/
  - $3n+3=O(n)$  /\*  $3n+3 \leq 4n$  for  $n \geq 3$  \*/
  - $100n+6=O(n)$  /\*  $100n+6 \leq 101n$  for  $n \geq 10$  \*/
  - $10n^2+4n+2=O(n^2)$  /\*  $10n^2+4n+2 \leq 11n^2$  for  $n \geq 5$  \*/
  - $6 \cdot 2^n + n^2 = O(2^n)$  /\*  $6 \cdot 2^n + n^2 \leq 7 \cdot 2^n$  for  $n \geq 4$  \*/

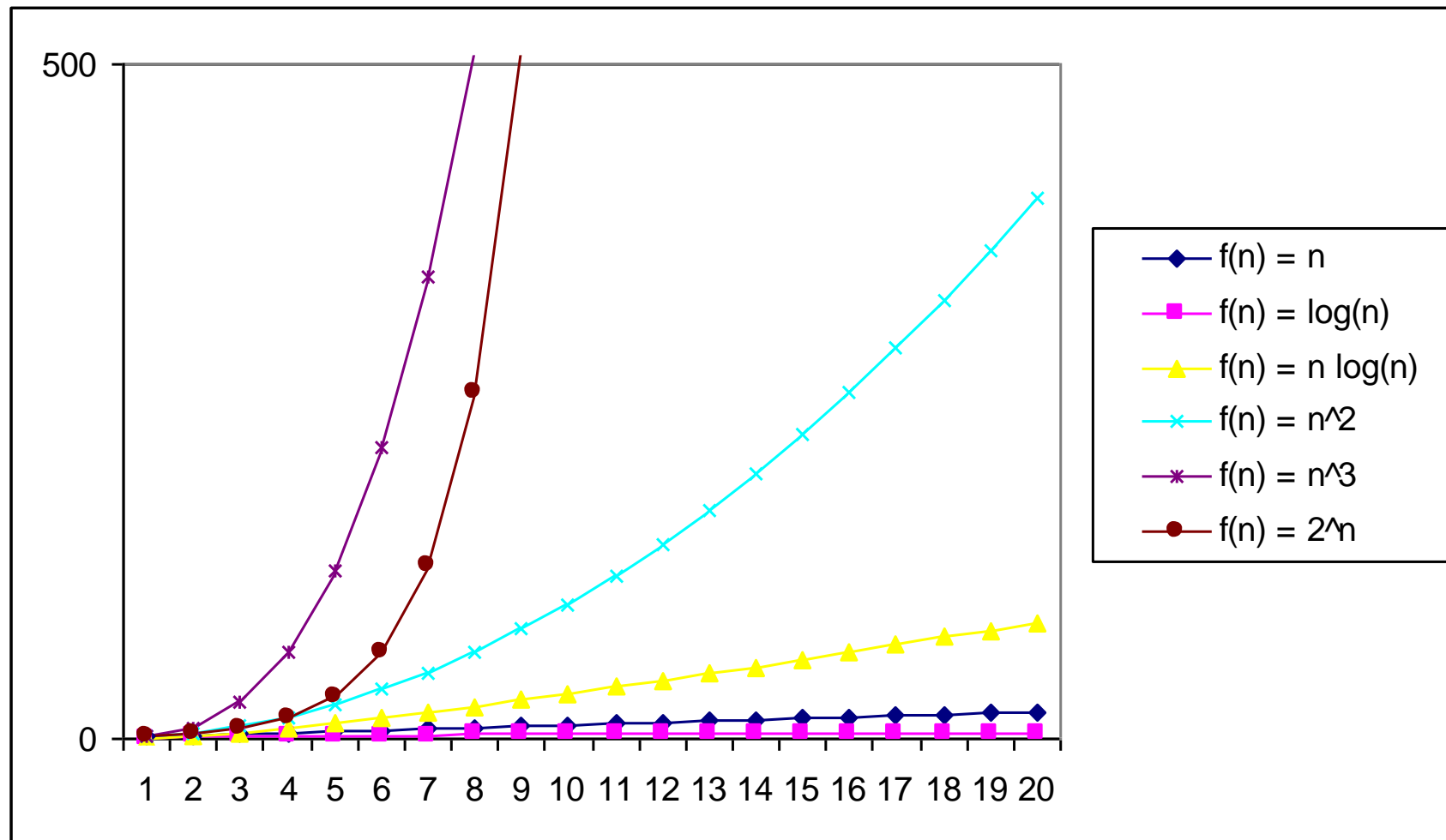
# Asymptotic Notation Cont. (O)

- $O(1)$ : constant
- $O(n)$ : linear
- $O(n^2)$ : quadratic
- $O(n^3)$ : cubic
- $O(2^n)$ : exponential
- $O(\log n)$  logarithmic
- $O(n \log n)$  log linear

# Practical Complexity

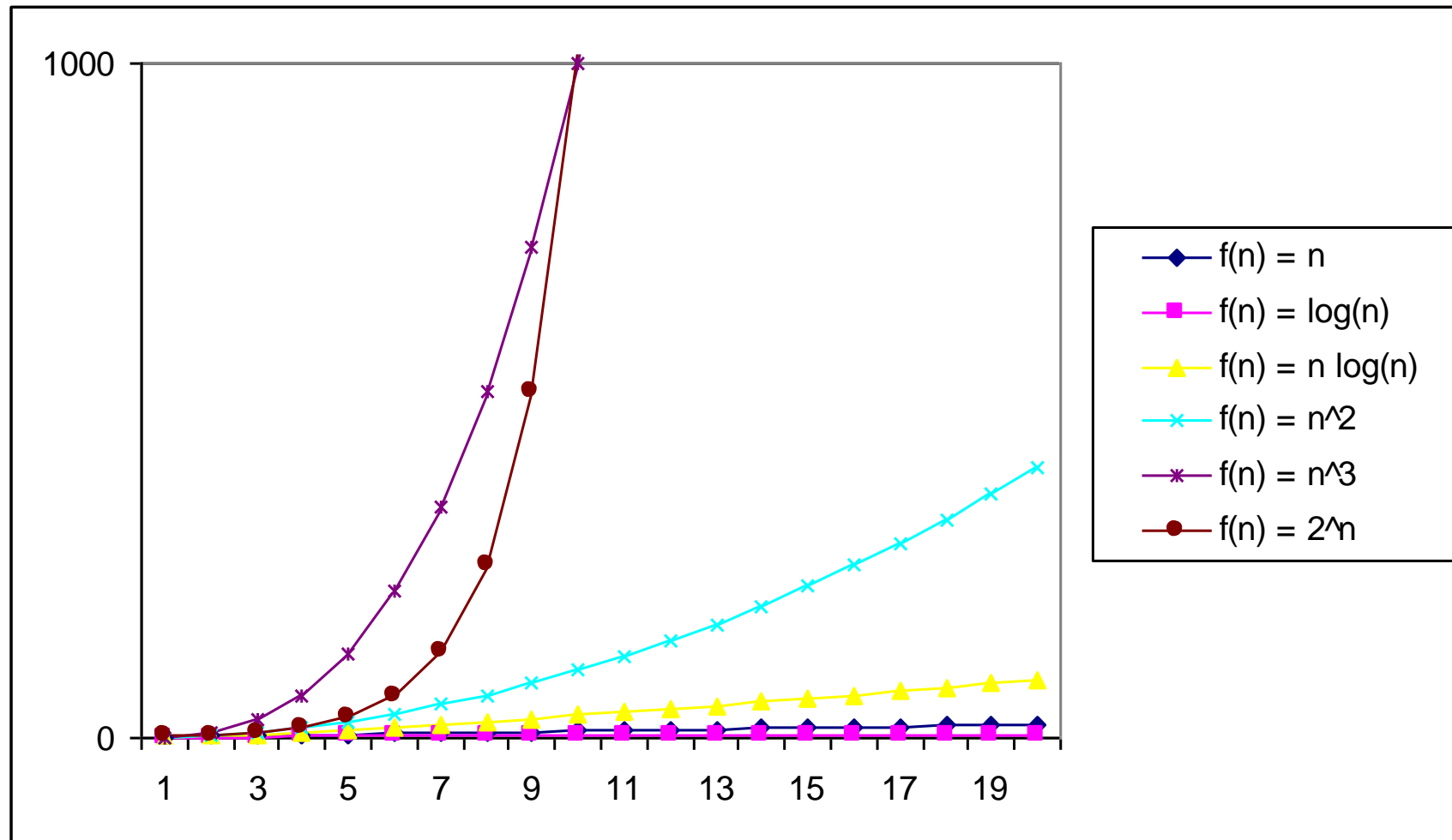


# Practical Complexity Cont.

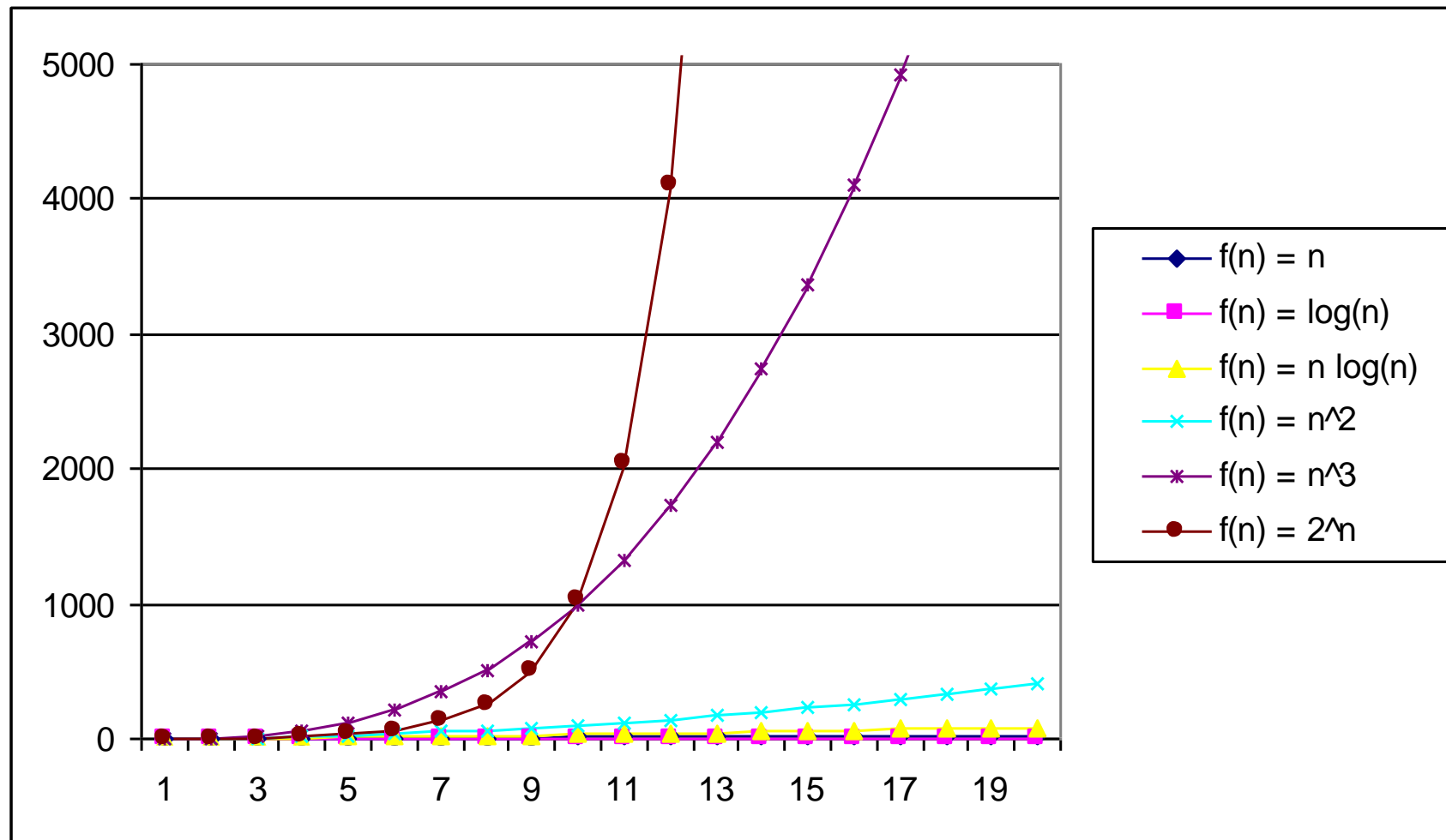




# Practical Complexity Cont.



# Practical Complexity Cont.



# Asymptotic Notation (Cont.)

- Definition: [Omega]  $f(n) = \Omega(g(n))$  iff there exist positive constants  $c$  and  $n_0$  such that  $f(n) \geq cg(n)$  for all  $n$ ,  $n \geq n_0$ .
- Example:
  - $3n + 2 = \Omega(n)$
  - $100n + 6 = \Omega(n)$
  - $10n^2 + 4n + 2 = \Omega(n^2)$

# Asymptotic Notation (Cont.)

Definition:  $f(n) = \Theta(g(n))$  iff there exist positive constants  $c_1$ ,  $c_2$ , and  $n_0$  such that  $c_1g(n) \leq f(n) \leq c_2g(n)$  for all  $n$ ,  $n \geq n_0$ .

# Asymptotic Notation (Cont.)

Theorem 1.2: If  $f(n) = a_m n^m + \dots + a_1 n + a_0$ , then  $f(n) = O(n^m)$ .

Proof:

$$\begin{aligned} f(n) &\leq \sum_{i=0}^m |a_i| n^i \\ &\leq n^m \sum_{i=0}^m |a_i| n^{i-m} \\ &\leq n^m \sum_{i=0}^m |a_i| \end{aligned} \quad \text{for } n \geq 1$$

So,  $f(n) = O(n^m)$

# Asymptotic Notation (Cont.)

Theorem 1.3: If  $f(n) = a_m n^m + \dots + a_1 n + a_0$  and  $a_m > 0$ , then  $f(n) = \Omega(n^m)$ .

Theorem 1.4: If  $f(n) = a_m n^m + \dots + a_1 n + a_0$  and  $a_m > 0$ , then  $f(n) = \Theta(n^m)$ .

Time for $f(n)$ instructions on a $10^9$ instr/sec computer							
$n$	$f(n)=n$	$f(n)=\log_2 n$	$f(n)=n^2$	$f(n)=n^3$	$f(n)=n^4$	$f(n)=n^{10}$	$f(n)=2^n$
10	.01 $\mu$ s	.03 $\mu$ s	.1 $\mu$ s	1 $\mu$ s	10 $\mu$ s	10sec	1 $\mu$ s
20	.02 $\mu$ s	.09 $\mu$ s	.4 $\mu$ s	8 $\mu$ s	160 $\mu$ s	2.84hr	1ms
30	.03 $\mu$ s	.15 $\mu$ s	.9 $\mu$ s	27 $\mu$ s	810 $\mu$ s	6.83d	1sec
40	.04 $\mu$ s	.21 $\mu$ s	1.6 $\mu$ s	64 $\mu$ s	2.56ms	121.36d	18.3min
50	.05 $\mu$ s	.28 $\mu$ s	2.5 $\mu$ s	125 $\mu$ s	6.25ms	3.1yr	13d
100	.10 $\mu$ s	.66 $\mu$ s	10 $\mu$ s	1ms	100ms	3171yr	$4 \times 10^{13}$ yr
1,000	1.00 $\mu$ s	9.96 $\mu$ s	1ms	1sec	16.67min	$3.17 \times 10^{13}$ yr	$32 \times 10^{283}$ yr
10,000	10.00 $\mu$ s	130.03 $\mu$ s	100ms	16.67min	115.7d	$3.17 \times 10^{23}$ yr	
100,000	100.00 $\mu$ s	1.66ms	10sec	11.57d	3171yr	$3.17 \times 10^{33}$ yr	
1,000,000	1.00ms	19.92ms	16.67min	31.71yr	$3.17 \times 10^7$ yr	$3.17 \times 10^{43}$ yr	

$\mu$ s = microsecond =  $10^{-6}$  seconds

ms = millisecond =  $10^{-3}$  seconds

sec = seconds

min = minutes

hr = hours

d = days

yr = years

# Home Work 2