

Chapter 4



Greedy Approach

روش حریصانه

The idea

- داده‌ها یکی یکی انتخاب می‌شوند.
- هر انتخاب باید **بهترین** باشد، بدون توجه به انتخاب‌های قبلی یا در آینده.
- اغلب برای حل مسائل بهینه‌سازی استفاده می‌شود.
- بهینگی راه حل ارایه شده توسط الگوریتم حریصانه باید اثبات شود.

مثال: پول خرد کردن

Coins



Amount owed: 36 cents

Step

1. Grab 25



2. Grab first 10



3. Reject second 10



4. Reject 5



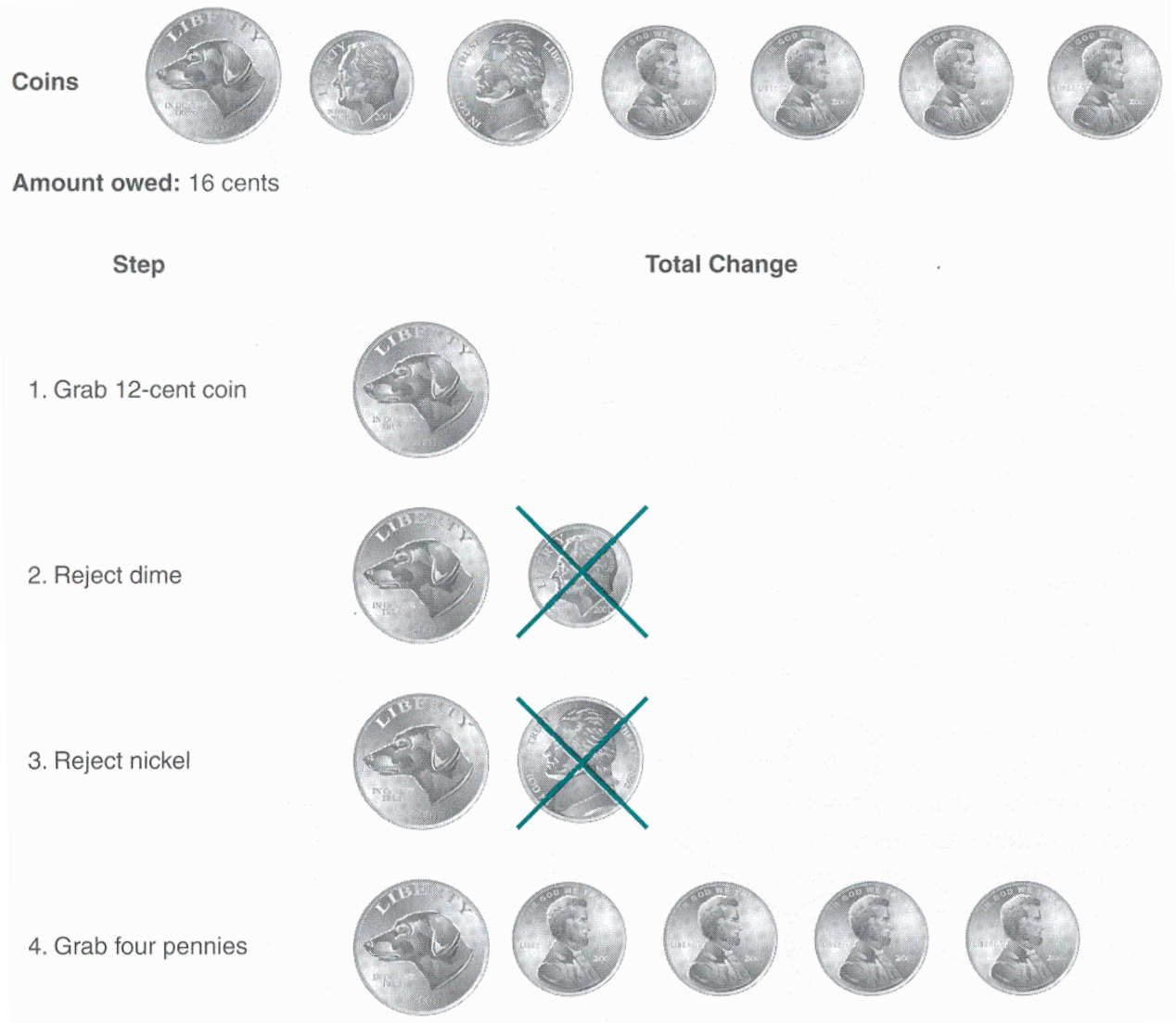
5. Grab 1



Figure 4.1 • A greedy algorithm for giving change.

U.S. coins (penny(1), nickel(5), dime(10), quarter(25), half dollar(50))

حالتی که روش
 حریصانه جواب
 بهینه را نمی‌دهد



if we include a 12-cent coin with the U.S. coins, the greedy algorithm does not always give an optimal solution.

Greedy solution: $16 = 12 + 1 + 1 + 1 + 1$
 Optimal solution: $16 = 10 + 5 + 1$

The algorithm

```
while ( there are more coins and the instance is not solved){  
    grab the largest remaining coin;  
                                     // selection procedure  
    If (adding the coin makes the change exceed the amount owed)  
        reject the coin;             // feasibility check  
    else  
        add the coin to the change;  
    If (the total value of the change equals the amount owed)  
                                     // solution check  
        the instance is solved;  
}
```

Basic components in Greedy approach

- A **selection procedure** chooses the next item to add to the set. The selection is performed according to a **greedy criterion** that satisfies some locally optimal consideration at the time.
- A **feasibility check** determines if the new set is feasible by checking whether it is possible to complete this set in such a way as to give a solution to the instance.
- A **solution check** determines whether the new set constitutes a solution to the instance.

Minimum spanning trees

درخت پوشای کمینه

تعریف درخت □

Definition

An *undirected graph* G consists of a finite set V whose members are called the vertices of G , together with a set E of pairs of vertices in V . These pairs are called the edges of G . We denote G by

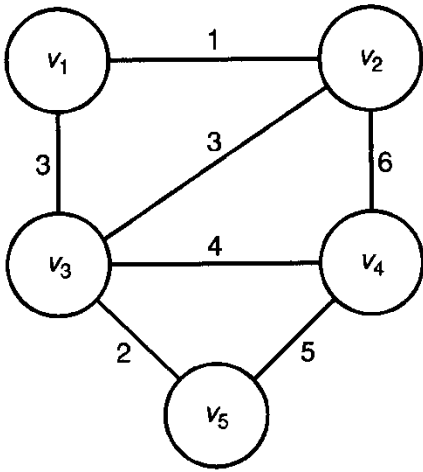
$$G = (V, E).$$

□ دو روش حل برای یافتن درخت پوشای کمینه:

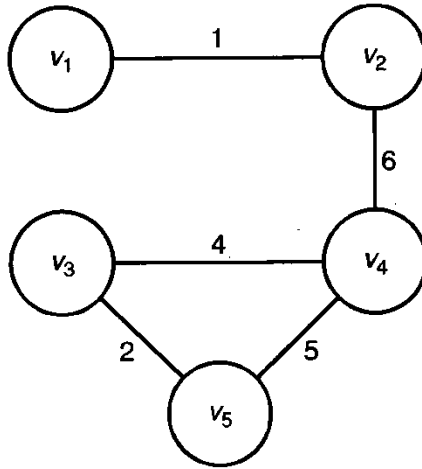
■ روش کراسکال (Kruskal)

■ روش پریم (Prim)

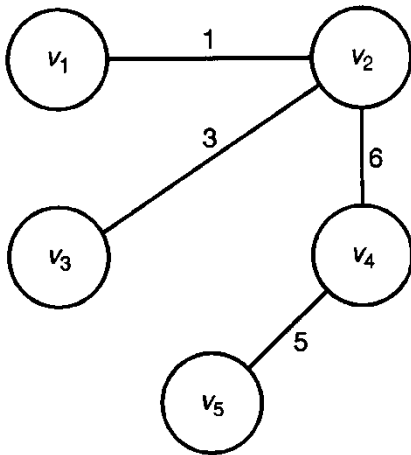
(a) A connected, weighted, undirected graph G .



(b) If (v_4, v_5) were removed from this subgraph, the graph would remain connected.



(c) A spanning tree for G .



(d) A minimum spanning tree for G .

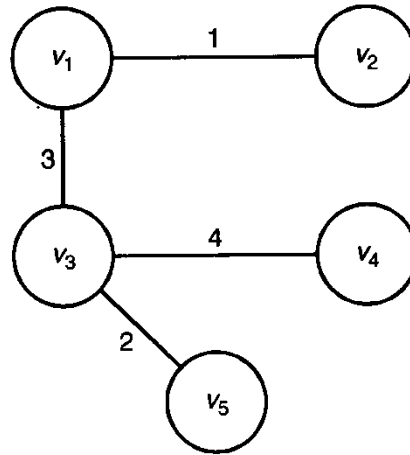
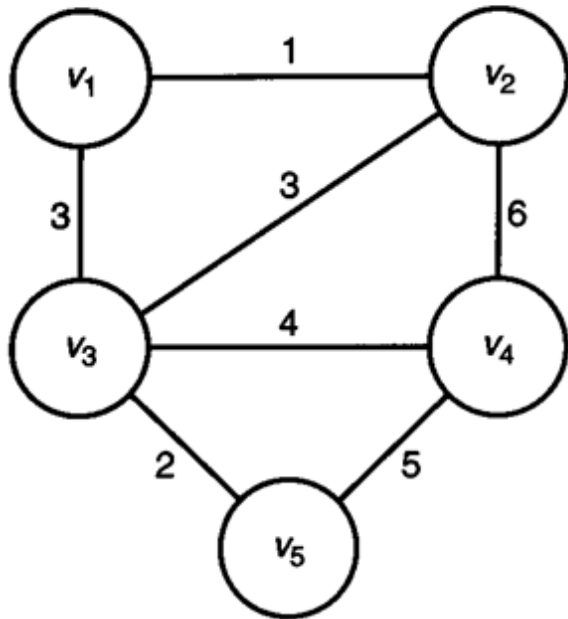


Figure 4.3 • A weighted graph and three subgraphs.

مثال:
گراف و چند درخت
متناظر با آن

An illustration (1)

Determine a minimum spanning tree.



1. Edges are sorted by weight.

(v_1, v_2) 1

(v_3, v_5) 2

(v_1, v_3) 3

(v_2, v_3) 3

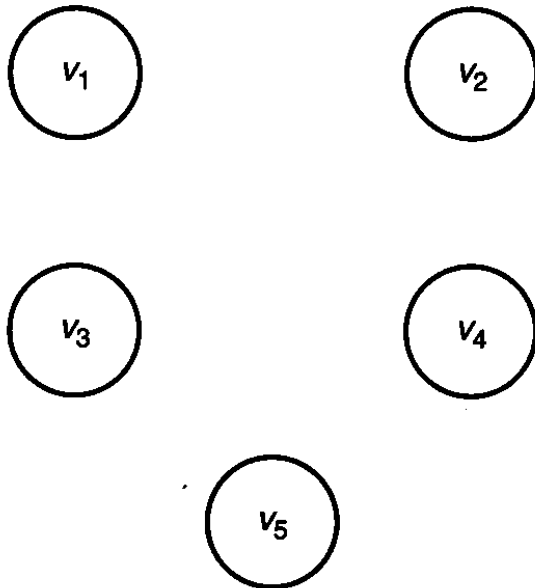
(v_3, v_4) 4

(v_4, v_5) 5

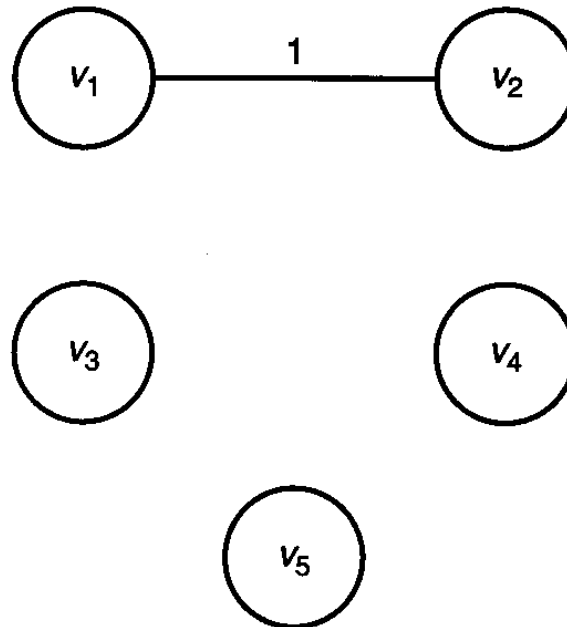
(v_2, v_4) 6

An illustration (2)

2. Disjoint set are created.

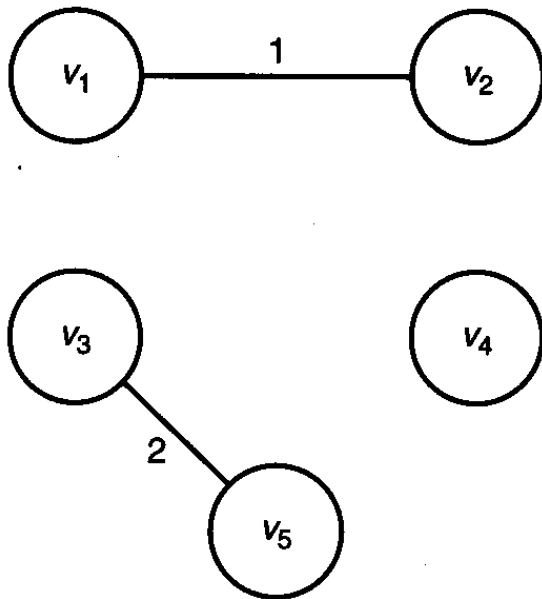


3. Edge (v_1, v_2) is selected.

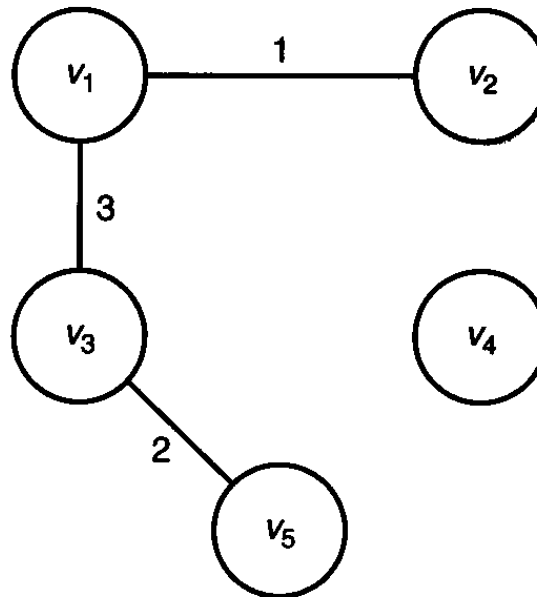


An illustration (3)

4. Edge (v_3, v_5) is selected.

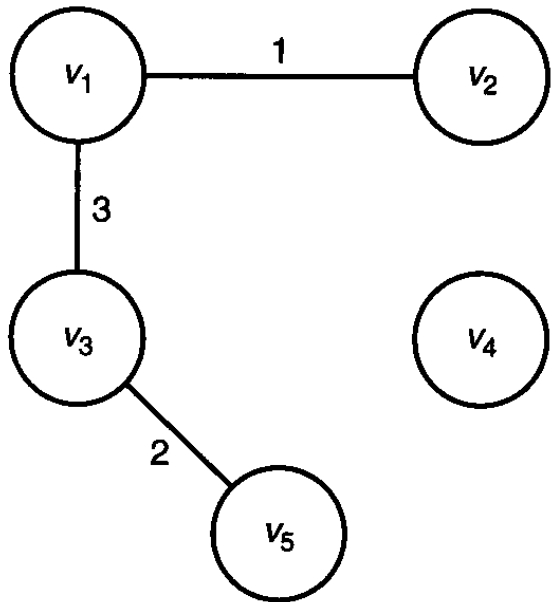


5. Edge (v_1, v_3) is selected.

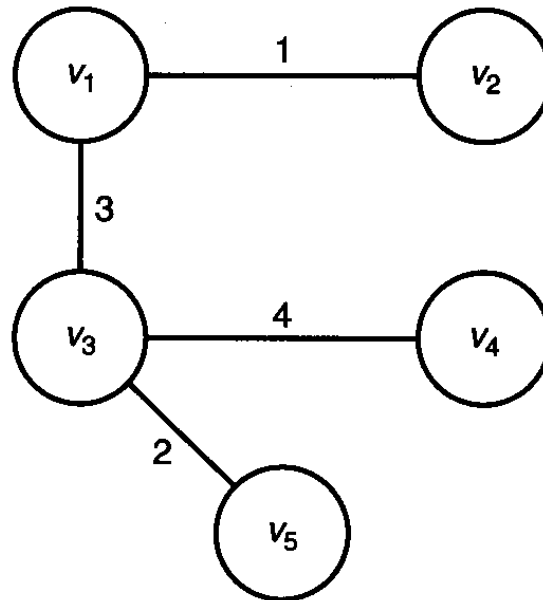


An illustration (4)

6. Edge (v_2, v_3) is selected.



7. Edge (v_3, v_4) is selected.



Data types and operations

□ Data types

- index i ;
- set_pointer p, q ;

□ Operations

- $initial(n)$ initializes n disjoint subsets, each of which contains exactly one of the indices between 1 and n .
- $p = find(i)$ makes p point to the set containing index i .
- $merge(p, q)$ merges the two sets, to which p and q point, into the set.
- $equal(p, q)$ returns true if p and q both point to the same set.

Kruskal's algorithm

```
 $F = \emptyset;$  // Initialize set of
// edges to empty.
create disjoint subsets of  $V$ , one for each
vertex and containing only that vertex;
sort the edges in  $E$  in nondecreasing order;
while (the instance is not solved){
    select next edge; // selection procedure
    if (the edge connects two vertices in // feasibility check
        disjoint subsets){
        merge the subsets;
        add the edge to  $F$ ;
    }
    if (all the subsets are merged) // solution check
        the instance is solved;
}
```

Kruskal's algorithm

```
void kruskal (int n, int m, set_of_edges E, set_of_edges& F) {  
    index i, j;  
    set_pointer p, q;  
    edge e;  
    Sort the m edges in E by weight in nondecreasing order;  
    F =  $\emptyset$ ;  
    initial (n); // Initialize n disjoint subsets.  
    while (number of edges in F is less than n - 1){  
        e = edge with least weight not yet considered;  
        i, j = indices of vertices connected by e;  
        p = find(i);  
        q = find(j);  
        if (! equal(p, q)){  
            merge(p, q);  
            add e to F;  
        }  
    }  
}
```

Worst-case time complexity

- Basic operation: a comparison instruction
- Input size: n , the number of vertices, and m , the number of edges
- Analysis:
 - The time to sort the edges: $W(m) \in \Theta(m \lg m)$
 - The time in the while loop: $W(m) \in \Theta(m \lg m)$
 - The time to initialize n disjoint sets: $T(n) \in \Theta(n)$
 - Overall: $W(m, n) \in \Theta(m \lg m)$
 - In the worst case every vertex can be connected to every other vertex $m = n(n-1)/2$,
Therefore $W(m, n) = \Theta(n^2 \lg n)$

Proof

- ▲ **Lemma 4.2** Let $G = (V, E)$ be a connected, weighted, undirected graph; let F be a promising subset of E ; and let e be an edge of minimum weight in $E - F$ such that $F \cup \{e\}$ has no simple cycles. Then $F \cup \{e\}$ is promising.
- **Theorem 4.2:** Kruskal's algorithm always produces a minimum spanning tree

Prim's algorithm

$F = \emptyset$ // Initialize set of edges to empty.

$Y = \{v_1\}$ // Initialize set of vertices to
// contain only the first one.

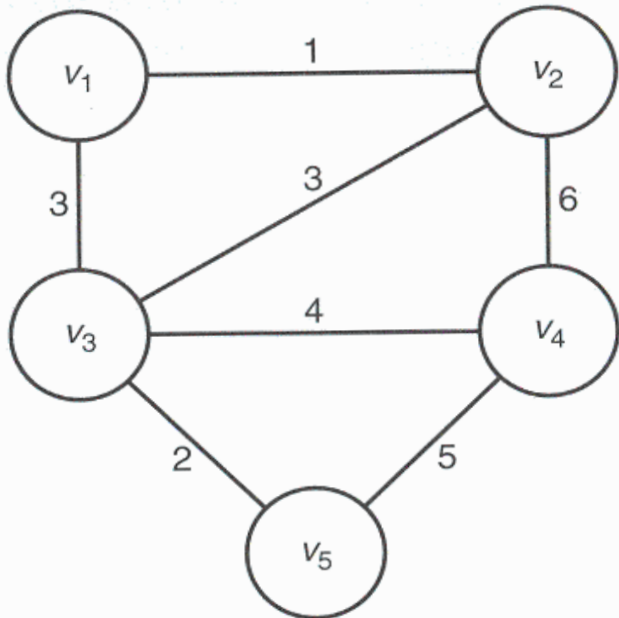
```
while (the instance is not solved){
    select a vertex in  $V - Y$  that is nearest to  $Y$ ;
        // selection procedure and feasibility check

    add the vertex to  $Y$ ;
    add the edge to  $F$ ;

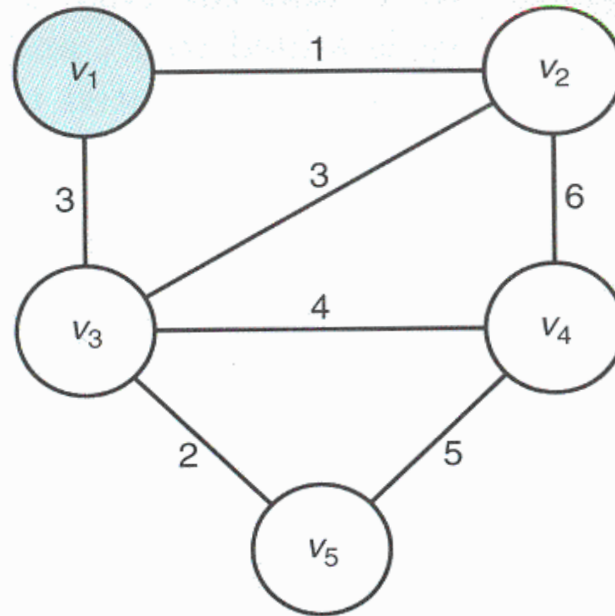
    if ( $Y == V$ )// solution check the instance is solved;
}
```

An example

Determine a minimum spanning tree.

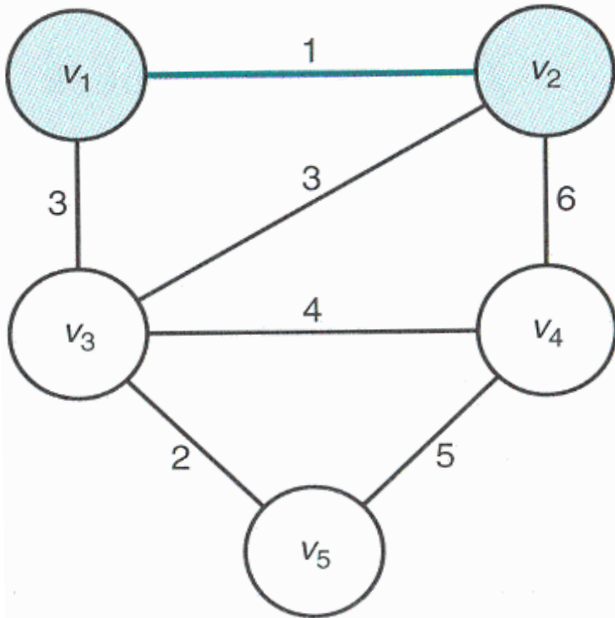


1. Vertex v_1 is selected first.

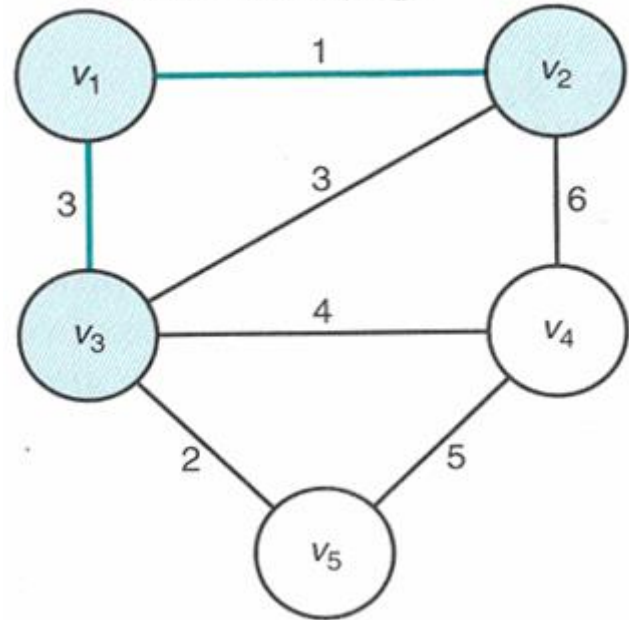


An example

2. Vertex v_2 is selected because it is nearest to $\{v_1\}$.

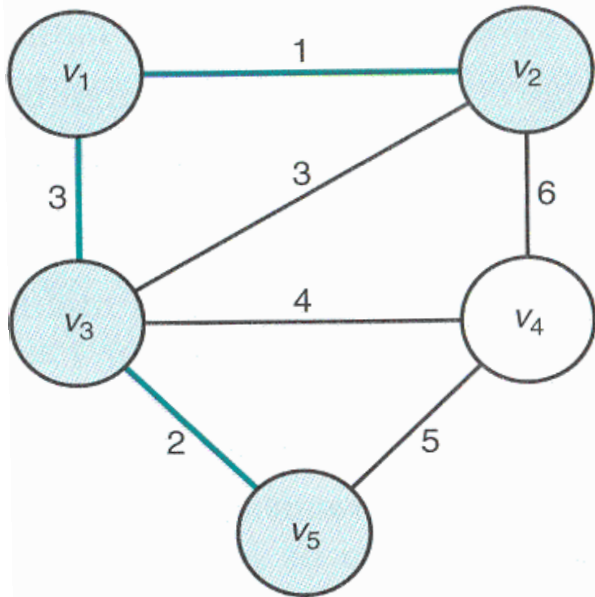


3. Vertex v_3 is selected because it is nearest to $\{v_1, v_2\}$.

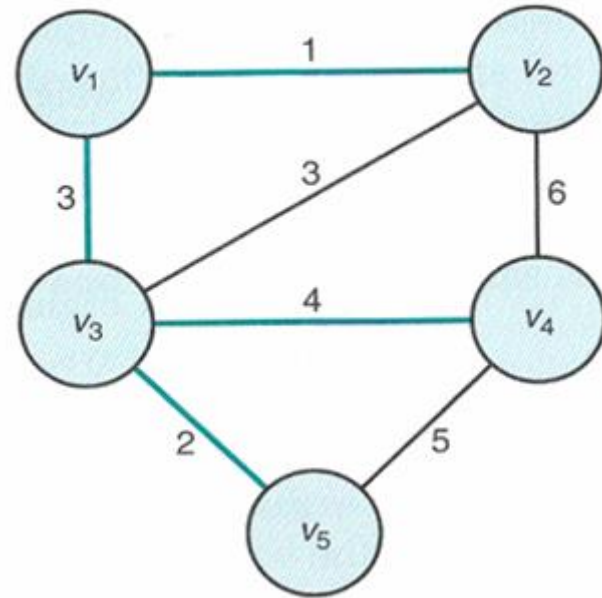


An example

4. Vertex v_2 is selected because it is nearest to $\{v_1, v_2, v_3\}$.



5. Vertex v_4 is selected.



Adjacency matrix

$$W[i][j] = \begin{cases} \text{weight on edge} & \text{if there is an edge between } v_i \text{ and } v_j \\ \infty & \text{if there is no edge between } v_i \text{ and } v_j \\ 0 & \text{if } i = j. \end{cases}$$

	1	2	3	4	5
1	0	1	3	∞	∞
2	1	0	3	6	∞
3	3	3	0	4	2
4	∞	6	4	0	5
5	∞	∞	2	5	0

Figure 4.5 • The array representation W of the graph in Figure 4.3(a).

Prim's algorithm

$nearest[i]$ = index of the vertex in Y nearest to v_i

$distance[i]$ = weight on edge between v_i and the vertex indexed by $nearest[i]$

Algorithm 4.1: Prim's Algorithm

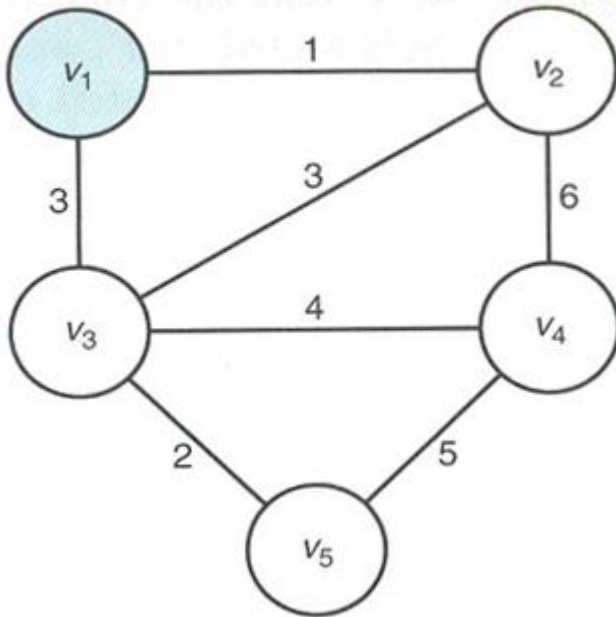
```
void prim (int  $n$ , const number  $W[][]$ , set_of_edges&  $F$ )
{
    index  $i$ ,  $v_{near}$ ;
    number  $min$ ;
    edge  $e$ ;
    index  $nearest[2..n]$ ;
    number  $distance[2..n]$ ;
     $F = \emptyset$ ;
    for ( $i = 2$ ;  $i \leq n$ ;  $i++$ ){
         $nearest[i] = 1$ ;           // For all vertices, initialize  $v_1$ 
         $distance[i] = W[1][i]$ ; // to be the nearest vertex in
    }
    ...
}
```

Prim's algorithm (cont'd)

```
repeat ( $n - 1$  times){ // Add all  $n - 1$  vertices to  $Y$ .
     $min = \infty$ ;
    for ( $i = 2; i \leq n; i++$ ) // Check each vertex for being nearest to  $Y$ .
        if ( $0 \leq distance[i] < min$ ){
             $min = distance[i]$ ;
             $vnear = i$ ;
        }
     $e =$  edge connecting vertices indexed by  $vnear$  and  $nearest[vnear]$ ;
    add  $e$  to  $F$ ;
     $distance[vnear] = -1$ ; // Add vertex indexed by
    for ( $i = 2; i \leq n; i++$ ) //  $vnear$  to  $Y$ .
        if ( $W[i][vnear] < distance[i]$ ){ // For each vertex not in
             $distance[i] = W[i][vnear]$ ; //  $Y$ , update its distance
             $nearest[i] = vnear$ ; // from  $Y$ .
        }
    }
}
```


An example

1. Vertex v_1 is selected first.



nearest

2	3	4	5
1	1	1	1

distance

2	3	4	5
1	3		

$F = \emptyset;$

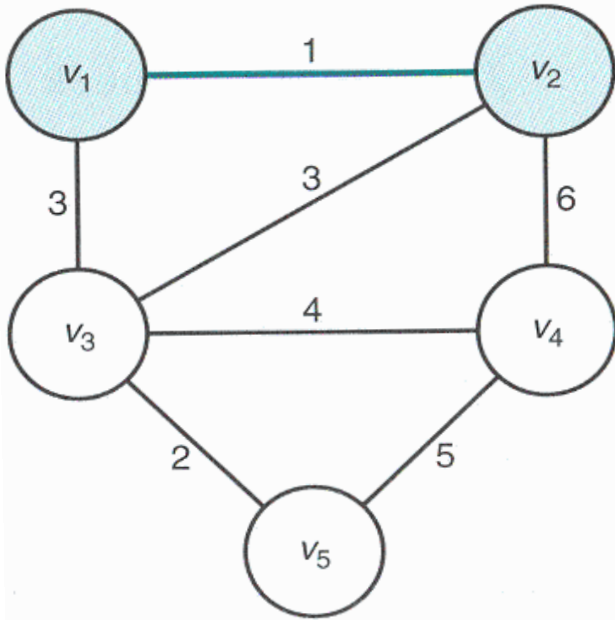
for ($i = 2; i \leq n; i++$) {

$nearest[i] = 1;$ // For all vertices, initialize v_1

$distance[i] = W[1][i];$ // to be the *nearest* vertex in

}

2. Vertex v_2 is selected because it is nearest to $\{v_1\}$.



nearest			
2	3	4	5
1	1	1	1
		2	

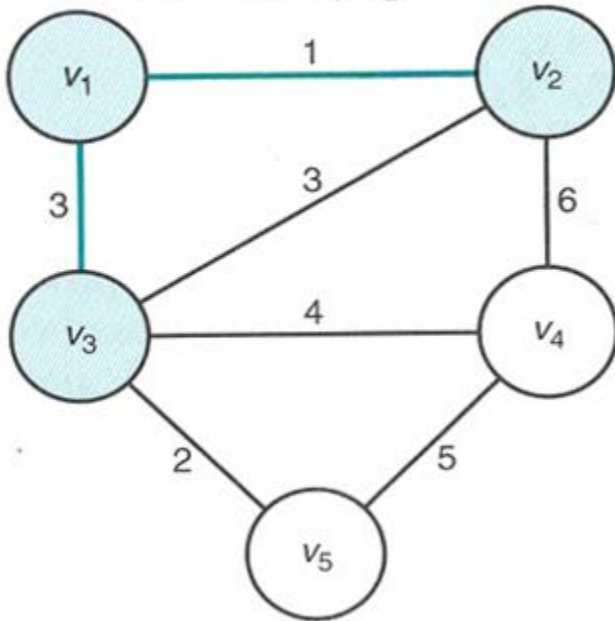
distance			
2	3	4	5
1	3		
-1		6	

```

repeat (n - 1 times){
    min = ∞;
    for (i = 2; i ≤ n; i++){
        if (0 ≤ distance[i] < min){
            min = distance[i];
            vnear = i;}
    e = edge connecting vertices indexed by vnear and nearest[vnear];
    add e to F;
    distance[vnear] = -1;
    for (i = 2; i ≤ n; i++){
        if (W[i][vnear] < distance[i]){
            distance[i] = W[i][vnear];
            nearest[i] = vnear;}
    }
}

```

3. Vertex v_3 is selected because it is nearest to $\{v_1, v_2\}$.



nearest				
2	3	4	5	
1	1	1	1	
		2		
		3	3	

distance				
2	3	4	5	
1	3			
-1		6		
	-1	4	2	

```

repeat (n - 1 times){
    min = ∞;
    for (i = 2; i ≤ n; i++){
        if (0 ≤ distance[i] < min){
            min = distance[i];
            vnear = i;}

```

e = edge connecting vertices indexed by $vnear$ and $nearest[vnear]$;
 add e to F ;

$distance[vnear] = -1$;

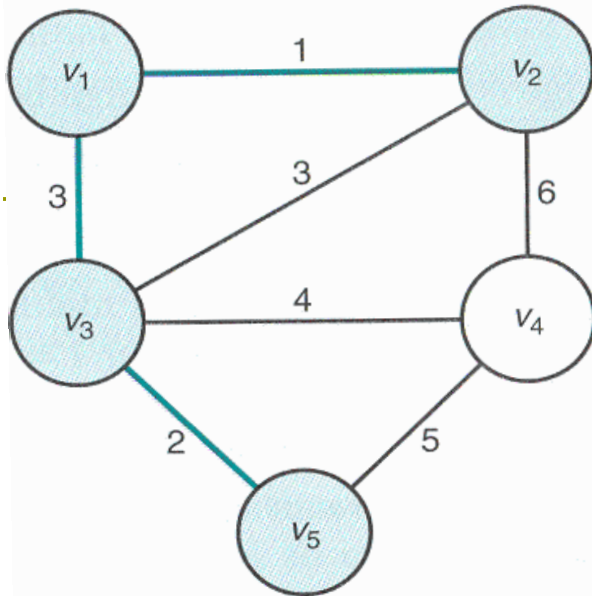
```

for (i = 2; i ≤ n; i++){
    if (W[i][vnear] < distance[i]){
        distance[i] = W[i][vnear];
        nearest[i] = vnear; }

```

}

4. Vertex v_2 is selected because it is nearest to $\{v_1, v_2, v_3\}$.



nearest			
2	3	4	5
1	1	1	1
		2	
		3	3

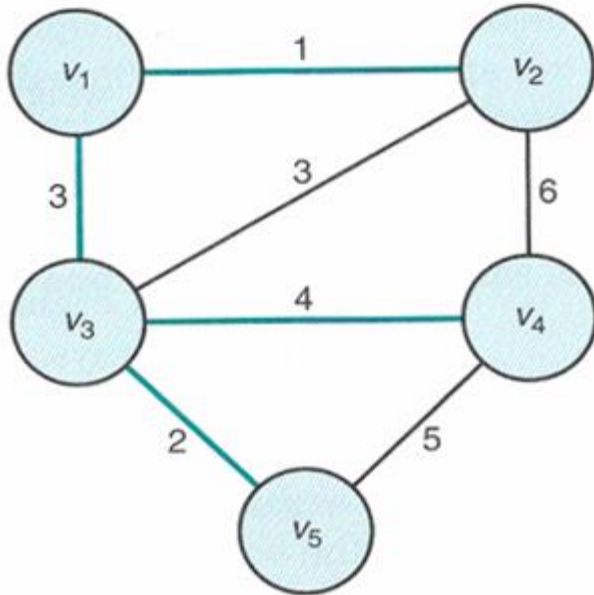
distance			
2	3	4	5
1	3		
-1		6	
	-1	4	2
			-1

```

repeat (n - 1 times){
    min = ∞;
    for (i = 2; i <= n; i++)
        if (0 ≤ distance[i] < min){
            min = distance[i];
            vnear = i;}
    e = edge connecting vertices indexed by vnear and nearest[vnear];
    add e to F;
    distance[vnear] = -1;
    for (i = 2; i <= n; i++)
        if (W[i][vnear] < distance[i]){
            distance[i] = W[i][vnear];
            nearest[i] = vnear; }
}

```

5. Vertex v_4 is selected.



nearest			
2	3	4	5
1	1	1	1
		2	
		3	3

distance			
2	3	4	5
1	3		
-1		6	
	-1	4	2
			-1
		-1	

```

repeat (n - 1 times){
    min = ∞;
    for (i = 2; i ≤ n; i++)
        if (0 ≤ distance[i] < min){
            min = distance[i];
            vnear = i;}
    e = edge connecting vertices indexed by vnear and nearest[vnear];
    add e to F;
    distance[vnear] = -1;
    for (i = 2; i ≤ n; i++)
        if (W[i][vnear] < distance[i]){
            distance[i] = W[i][vnear];
            nearest[i] = vnear;}
}

```

Every-case time complexity

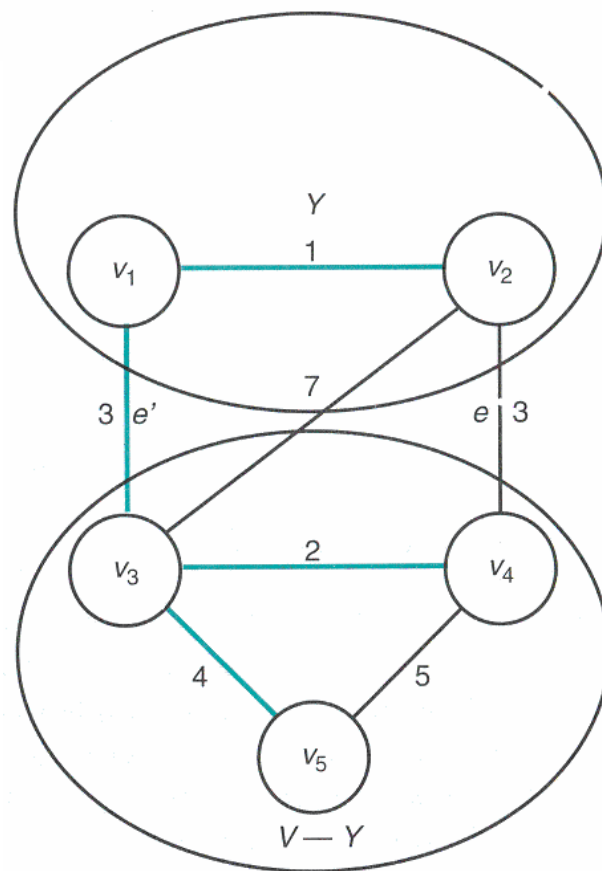
- Basic Operation: There are two loops, each with $n - 1$ iterations, inside the **repeat** loop. Executing the instructions inside each of them can be considered to be doing the basic operation once.
- Input size: n , the number of vertices
- Then:

$$T(n) = 2(n - 1)(n - 1) \in \Theta(n^2)$$

Proof

▲ **Lemma 4.1** Let $G = (V, E)$ be a connected, weighted, undirected graph; let F be a promising subset of E ; and let Y be the set of vertices connected by the edges in F . If e is an edge of minimum weight that connects a vertex in Y to a vertex in $V - Y$, then $F \cup \{e\}$ is promising.

- **Theorem 4.1: Prim's algorithm always produces a minimum spanning tree**



Comparing Prim's algorithm with Kruskal's algorithm

- Prim's algorithm: $T(n) \in \Theta(n^2)$
- Kruskal's algorithm:
 - $W(m, n) \in \Theta(m \lg m)$
- $n - 1 \leq m \leq n(n-1)/2$
- Sparse graph
 - *Kruskal's algorithm should be faster.*
- Highly connected
 - *Prim's algorithm should be faster.*

Dijkstra's algorithm

shortest paths from one source to all the others

الگوریتم دایکسترا

کوتاهترین مسیر از مبدا به همه راس‌ها

Dijkstra's algorithm

shortest paths from one source to all the others

$Y = \{v_1\};$

$F = \emptyset;$

while (the instance is not solved) {

 select a vertex v from $V - Y$, that has a shortest path from v_1 , using only vertices in Y as intermediates;

 //selection procedure and feasibility check

 add the new vertex v to Y ;

 add the edge (on the shortest path) that touches v to F ;

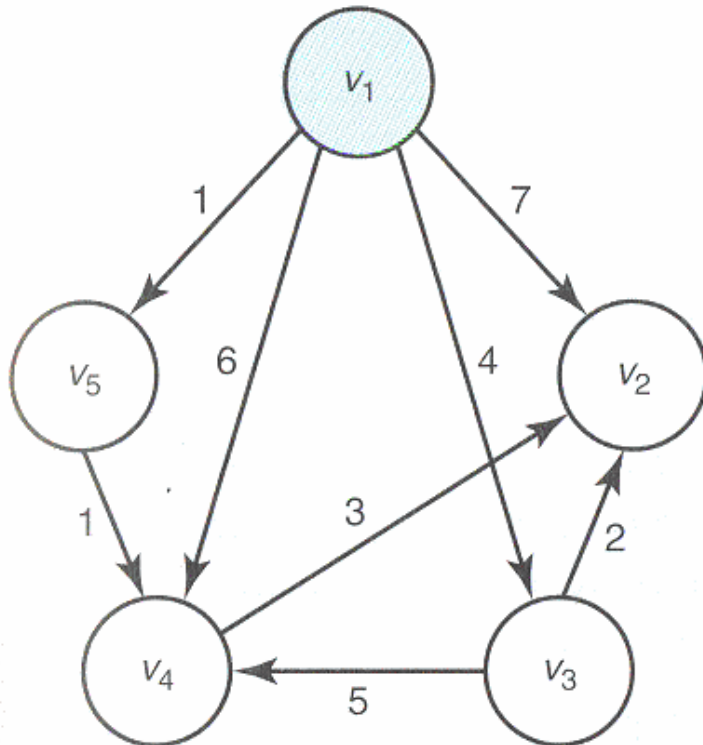
 if ($Y == V$) the instance is solved;

 // solution check

}

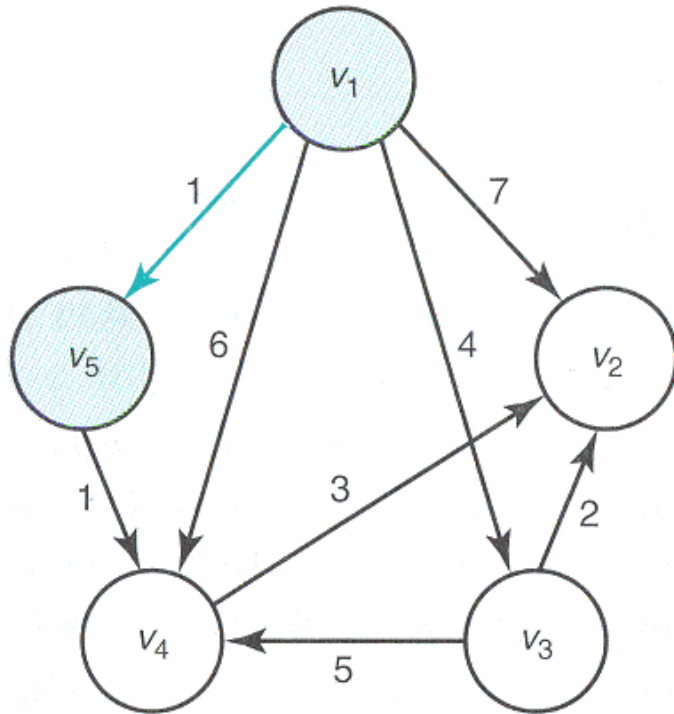
An example (1)

Compute shortest paths from v_1 .

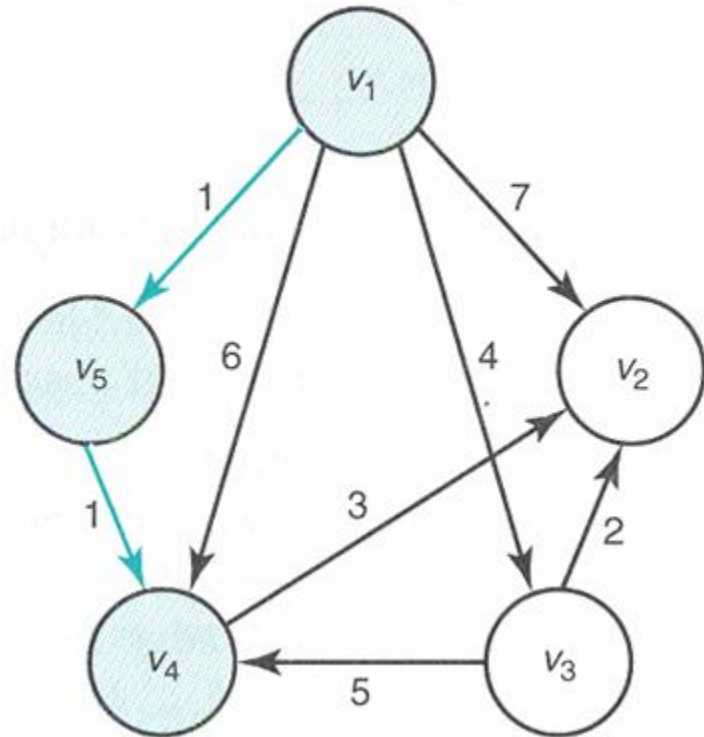


An example (2)

1. Vertex v_5 is selected because it is nearest to v_1 .

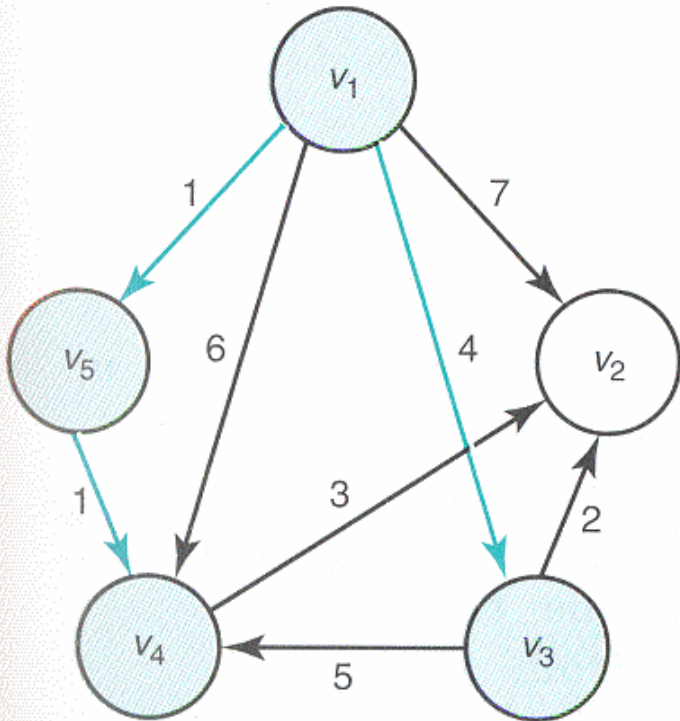


2. Vertex v_4 is selected because it has the shortest path from v_1 using only vertices in $\{v_5\}$ as intermediates.

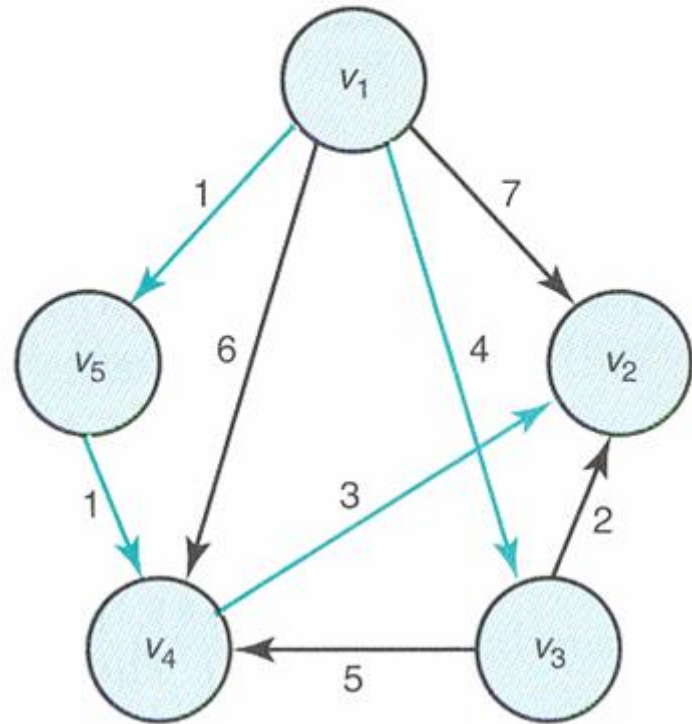


An example (3)

3. Vertex v_3 is selected because it has the shortest path from v_1 using only vertices in $\{v_4, v_5\}$ as intermediates.



4. The shortest path from v_1 to v_2 is $[v_1, v_5, v_4, v_2]$.



Auxiliary arrays

- $Touch[i]$ = index of vertex v in Y such that the edge $\langle v, v_i \rangle$ is the last edge on the current shortest path from v_1 to v_i using only vertices in Y as intermediates

آخرین گره برای دسترسی به گرهی که انتخاب نشده است

- $length[i]$ = length of the current shortest path from v_1 to v_i using only vertices in Y as intermediates

The algorithm (1)

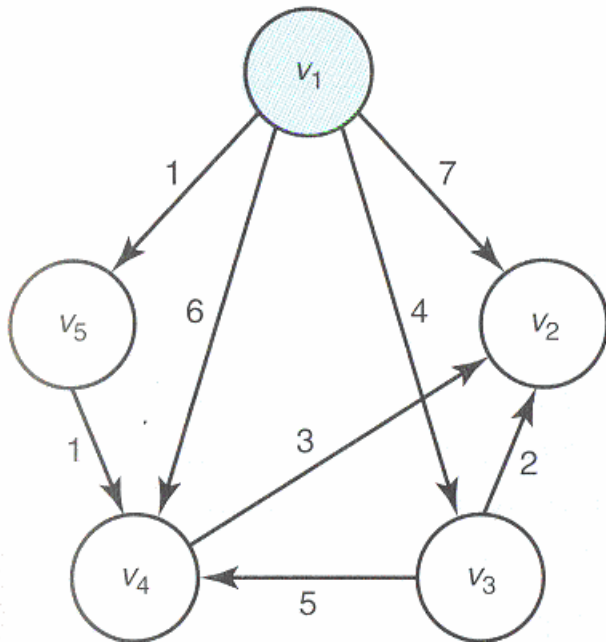
- Algorithm 4.3: Dijkstra's Algorithm

```
void dijkstra (int n, const number W[][i],  
set_of_edges&F) {  
    index i, vnear;  
    edge e;  
    index touch [2 .. n]; number length [2 .. n];  
    F =  $\emptyset$ ;  
    for (i = 2; i <= n; i++) {  
        touch [i] = 1;  
        length [i] = W[1] [i];  
    }  
}
```

The algorithm (2)

```
repeat ( $n - 1$  times){  
     $min = \infty$ ;  
    for ( $i = 2; i \leq n; i++$ )  
        if ( $0 \leq length[i] < min$ ) {  
             $min = length[i]$ ;  
             $vnear = i$ ;}  
     $e =$  edge from vertex indexed by  $touch[vnear]$  to  
    vertex indexed by  $vnear$ ;  
    add  $e$  to  $F$ ;  
    for ( $i = 2; i \leq n; i++$ )  
        if ( $length[vnear] + W[vnear][i] < length[i]$ ) {  
             $length[i] = length[vnear] + W[vnear][i]$ ;  
             $touch[i] = vnear$ ;}  
     $length[vnear] = -1$ ;  
}
```


Compute shortest paths from v_1 .

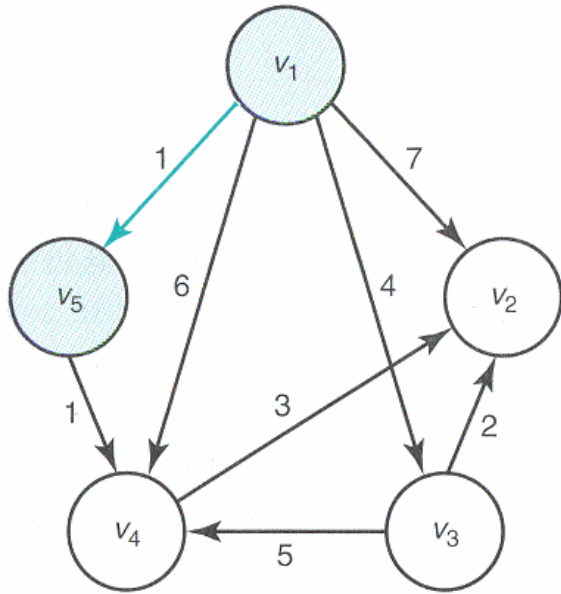


touch			
2	3	4	5
1	1	1	1

length			
2	3	4	5
7	4	6	1

```
for ( $i = 2; i \leq n; i++$ ) {  
     $touch[i] = 1;$   
     $length[i] = W[1][i];$   
}
```

1. Vertex v_5 is selected because it is nearest to v_1 .



le

touch			
2	3	4	5
1	1	1	1
		5	

length			
2	3	4	5
7	4	6	1
		2	-1

```

repeat ( $n - 1$  times){    $min = \infty$ ;
  for ( $i = 2; i \leq n; i++$ )
    if ( $0 \leq length[i] < min$ ) {
       $min = length[i]$ ;
       $vnear = i$ ;}

```

$e =$ **edge** from vertex indexed by $touch[vnear]$ to vertex indexed by $vnear$;
 add e to F ;

```

  for ( $i = 2; i \leq n; i++$ )
    if ( $length[vnear] + W[vnear][i] < length[i]$ ){
       $length[i] = length[vnear] + W[vnear][i]$ ;
       $touch[i] = vnear$ ;}

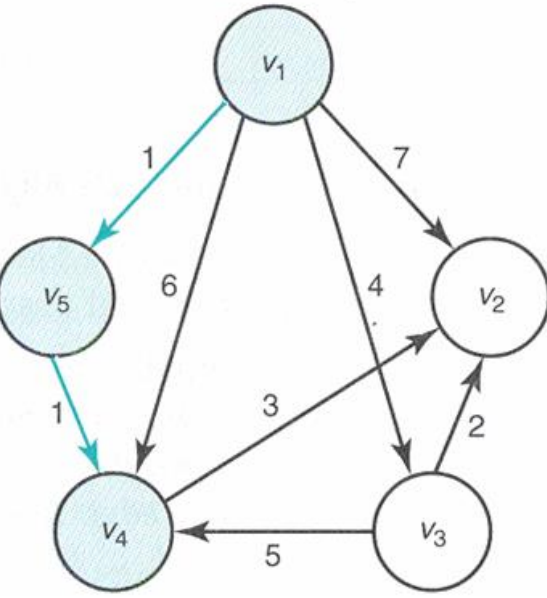
```

```

   $length[vnear] = -1$ ;}

```

2. Vertex v_4 is selected because it has the shortest path from v_1 using only vertices in $\{v_5\}$ as intermediates.



touch				
2	3	4	5	
1	1	1	1	
		5		
4				

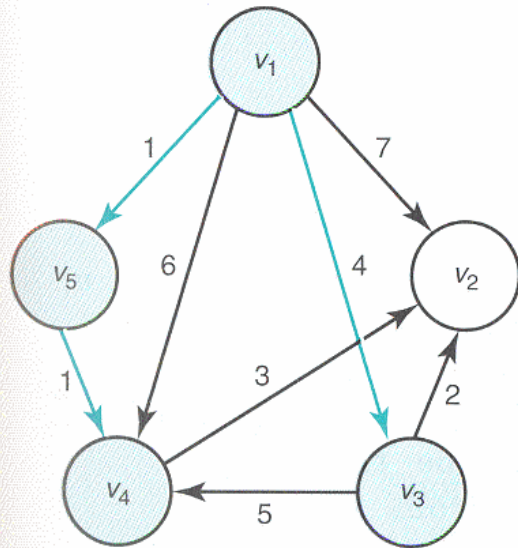
length				
2	3	4	5	
7	4	6	1	
		2	-1	
5		-1		

```

repeat ( $n - 1$  times){
     $min = \infty$ ;
    for ( $i = 2; i \leq n; i++$ )
        if ( $0 \leq length[i] < min$ ) {
             $min = length[i]$ ;
             $v_{near} = i$ ;
        }
     $e =$  edge from vertex indexed by  $touch[v_{near}]$  to vertex indexed by  $v_{near}$ ;
    add  $e$  to  $F$ ;
    for ( $i = 2; i \leq n; i++$ )
        if ( $length[v_{near}] + W[v_{near}][i] < length[i]$ ){
             $length[i] = length[v_{near}] + W[v_{near}][i]$ ;
             $touch[i] = v_{near}$ ;
        }
     $length[v_{near}] = -1$ ;

```

3. Vertex v_3 is selected because it has the shortest path from v_1 using only vertices in $\{v_4, v_5\}$ as intermediates.



ole

touch			
2	3	4	5
1	1	1	1
		5	
4			

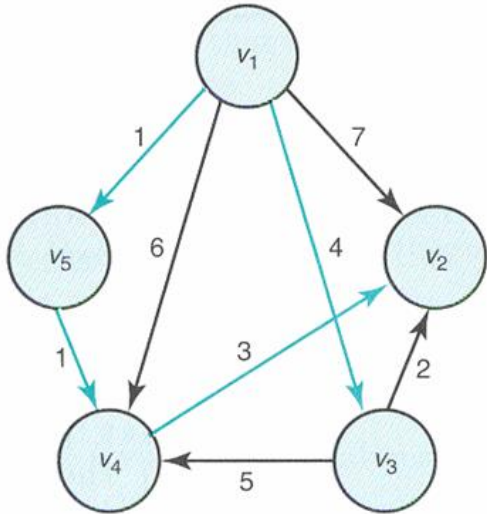
length			
2	3	4	5
7	4	6	1
		2	-1
5		-1	
	-1		

```

repeat ( $n - 1$  times){    $min = \infty$ ;
  for ( $i = 2; i \leq n; i++$ )
    if ( $0 \leq length[i] < min$ ) {
       $min = length[i]$ ;
       $vnear = i$ ;}
   $e =$  edge from vertex indexed by  $touch[vnear]$  to vertex indexed by  $vnear$ ;
  add  $e$  to  $F$ ;
  for ( $i = 2; i \leq n; i++$ )
    if ( $length[vnear] + W[vnear][i] < length[i]$ ){
       $length[i] = length[vnear] + W[vnear][i]$ ;
       $touch[i] = vnear$ ;}
   $length[vnear] = -1$ ;}

```

4. The shortest path from v_1 to v_2 is $[v_1, v_5, v_4, v_2]$.



ple

touch			
2	3	4	5
1	1	1	1
		5	
4			

length			
2	3	4	5
7	4	6	1
		2	-1
5		-1	
	-1		
-1			

```

repeat ( $n - 1$  times){    $min = \infty$ ;
  for ( $i = 2; i \leq n; i++$ )
    if ( $0 \leq length[i] < min$ ) {
       $min = length[i]$ ;
       $vnear = i$ ;}
   $e =$  edge from vertex indexed by  $touch[vnear]$  to vertex indexed by  $vnear$ ;
  add  $e$  to  $F$ ;
  for ( $i = 2; i \leq n; i++$ )
    if ( $length[vnear] + W[vnear][i] < length[i]$ ){
       $length[i] = length[vnear] + W[vnear][i]$ ;
       $touch[i] = vnear$ ;}
   $length[vnear] = -1$ ;}
  
```

Every-case time complexity

□ $T(n) = 2(n-1)^2 \in(n^2)$

Dijkstra vs Prim

```
repeat (n - 1 times){           min = ∞;
  for (i = 2; i <= n; i++)
    if ( 0 ≤ length [i] < min) {
      min = length [i];
      vnear = i;}
  e = edge from vertex indexed by touch [vnear] to vertex indexed by vnear;
  add e to F;
  for (i = 2; i <= n; i++)
    if (length [vnear] + W[vnear] [i] < length [i]){
      length[i] = length[vnear] + W[vnear][i];
      touch[i] = vnear;}
  length[vnear] = -1;}
```

```
repeat (n - 1 times){           min = ∞;
  for (i = 2; i <= n; i++)
    if (0 ≤ distance[i] < min){
      min = distance[i];
      vnear = i;}
  e = edge connecting vertices indexed by vnear and nearest[vnear];
  add e to F;
  distance[vnear] = -1;
  for (i = 2; i <= n; i++)
    if (W[i][vnear] < distance[i]){
      distance[i] = W[i] [vnear];
      nearest [i] = vnear; }
}
```

Scheduling

زمان بندی

□ دو نوع زمان بندی

۱- جمع زمانی که کارها منتظرند یا در حال سرویس گرفتن هستند کمینه شود.

- Minimize the total time in **waiting** and **being served** (time in the system)

۲- زمان بندی بامهلت

- scheduling with deadlines

زمان بندی نوع اول (بدون مهلت)

□ Compute the optimal scheduling:

$$t_1 = 5, t_2 = 10, \text{ and } t_3 = 4$$

three jobs and their service times

■ برای حل کارها را به ترتیب از کوچک به بزرگ مرتب می کنیم.

The algorithm

sort the jobs by service time in nondecreasing order;

while (the instance is not solved){

 schedule the next job; // selection procedure
 // feasibility check

 if (there are no more jobs) // solution check
 the instance is solved;

}

Proof

□ Time complexity: $W(n) \in \Theta(n \lg n)$

□ Theorem 4.3

The only schedule that minimizes the total time in the system is one that schedules jobs in nondecreasing order by service time

Multiple-server scheduling problem

- Server 1 serves jobs $1, (1+m), (1+2m), (1+3m), \dots$
- Server 2 serves jobs $2, (2+m), (2+2m), (3+3m), \dots$
- ...
- Server i serves jobs $i, (i+m), (i+2m), (i+3m), \dots$
- ...
- Server m serves jobs $m, (m+m), (m+2m), (m+3m), \dots$

Scheduling with deadlines

زمان‌بندی با مهلت

- در این نوع مساله زمان‌بندی، انجام هر کار، یک واحد زمان می‌برد تا به پایان برسد و یک مهلت (ضرب‌الاجل) و یک سود دارد.
- اگر کار قبل یا در مهلت شروع شود، سود حاصل می‌شود.

<i>Job</i>	<i>Deadline</i>	<i>Profit</i>
1	2	30
2	1	35
3	2	25
4	1	40

- هدف: به دست آوردن بیشترین منفعت

Some terms

- Feasible sequence
 - all the jobs in the sequence start by their deadlines
- Feasible set
 - there exists at least one feasible sequence for the jobs in this set.
- Optimal sequence
 - a feasible sequence with maximum total profit.

The algorithm

sort the jobs in nonincreasing order by profit;

$S = \emptyset$

```
while (the instance is not solved){  
    select next job; // selection procedure  
  
    if (S is feasible with this job added)  
        add this job to S;  
  
    if (there are no more jobs)  
        the instance is solved;  
}
```

The formal algorithm

- Algorithm 4.4: Scheduling with Deadlines

```
void schedule (int n, const int deadline [],  
sequence_of_integer& j)
```

```
{
```

```
    //sorted jobs in nonincreasing order by profit
```

```
    index i;
```

```
    sequence_of_integer K;
```

```
    J = [1];
```

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

```
    {
```

```
        K = J with i added according to nondecreasing  
        values of deadline[i];
```

```
        if (K is feasible)
```

```
            J = K;
```

```
    }
```

```
}
```


Example

<i>Job</i>	<i>Deadline</i>	<i>Profit</i>
1	3	40
2	1	35
3	1	30
4	3	25
5	1	20
6	3	15
7	2	10

Example

1. J is set to [1].
2. K is set to [2, 1] and is determined to be feasible.
 J is set to [2, 1] because K is feasible.
3. K is set to [2, 3, 1] and is rejected because it is not feasible.
4. K is set to [2, 1, 4] and is determined to be feasible.
 J is set to [2, 1, 4] because K is feasible.
5. K is set to [2, 5, 1, 4] and is rejected because it is not feasible.
6. K is set to [2, 1, 6, 4] and is rejected because it is not feasible.
7. K is set to [2, 7, 1, 4] and is rejected because it is not feasible.

The final value of J is [2, 1, 4].

Worst-case time complexity

- Basic operation: comparison instructions
- Input size: n , the number of jobs
- Time complexity:
 - time for sorting: $\Theta(n \lg n)$
 - comparisons in for- i loop:
$$\sum_{i=2}^n [(i-1) + i] = n^2 - 1 \in \Theta(n^2)$$
 - Overall: $W(n) \in \Theta(n^2)$

Theorem 4.4

- Algorithm 4.4 always produces an optimal set of jobs
- Proof: induction on the number of jobs n

Huffman code

کدهافمن

□ در فشرده‌سازی کاربرد دارد.

- File: ababcbbbc
- Encoding scheme:
 - a: 00, b: 01, c: 11
 - a: 10, b: 0, c: 11

Prefix codes

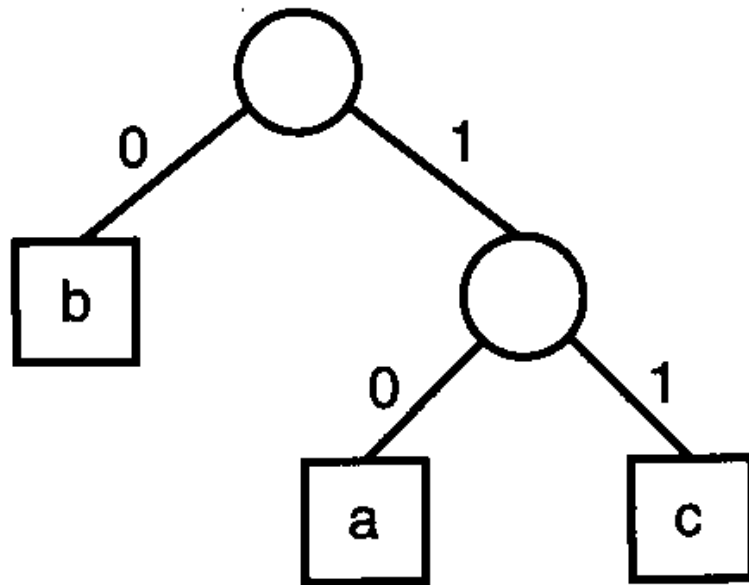


Figure 4.9 • Binary tree corresponding to Code 4.2.

An example

Character	Frequency	C1(Fixed-Length)	C2	C3 (Huffman)
<i>a</i>	16	000	10	00
<i>b</i>	5	001	11110	1110
<i>c</i>	12	010	1110	110
<i>d</i>	17	011	110	01
<i>e</i>	10	100	11111	1111
<i>f</i>	25	101	0	10

Table 4.1 • Three codes for the same file. C3 is optimal.

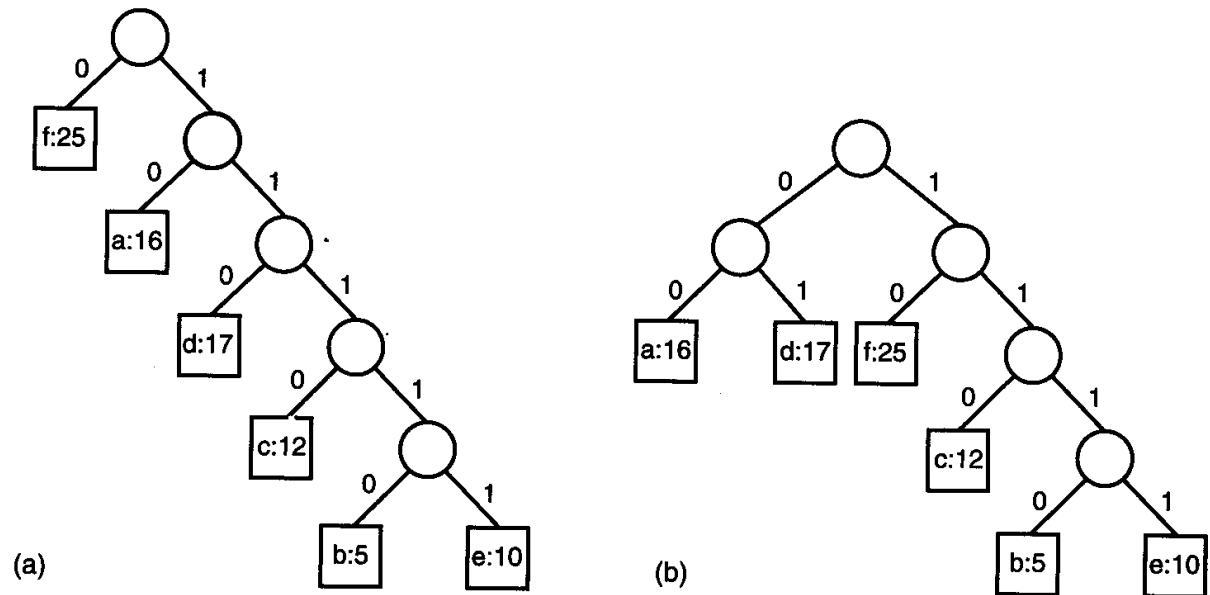


Figure 4.10 • The binary character code for Code C2 in Example 4.7 appears in (a), while the one for Code C3 (Huffman) appears in (b).

The number of bits taken to encode a file

$$\mathit{bits}(T) = \sum_{i=1}^n \mathit{frequency}(v_i) \mathit{depth}(v_i)$$

- Compute $\mathit{Bits}(C1)$, $\mathit{Bits}(C2)$, and $\mathit{Bits}(C3)$

An example of application (1)

b:5

e:10

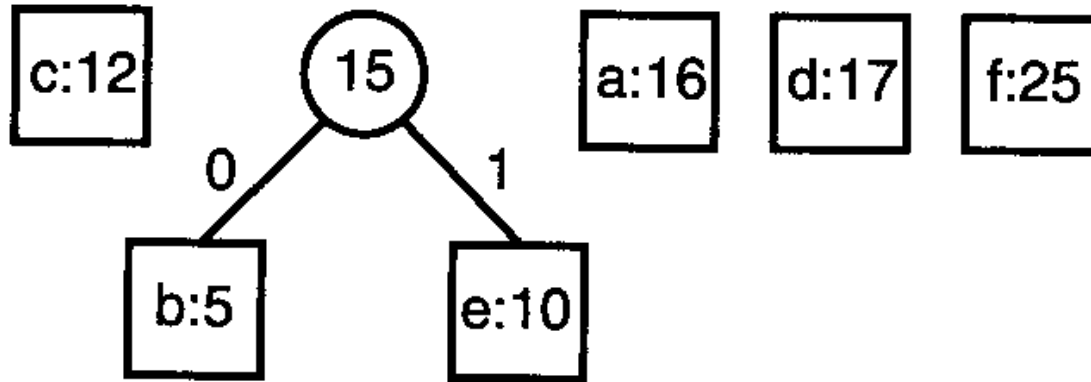
c:12

a:16

d:17

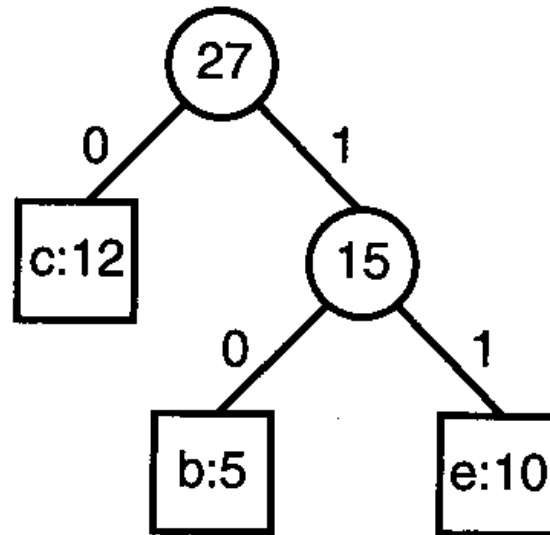
f:25

An example of application (2)



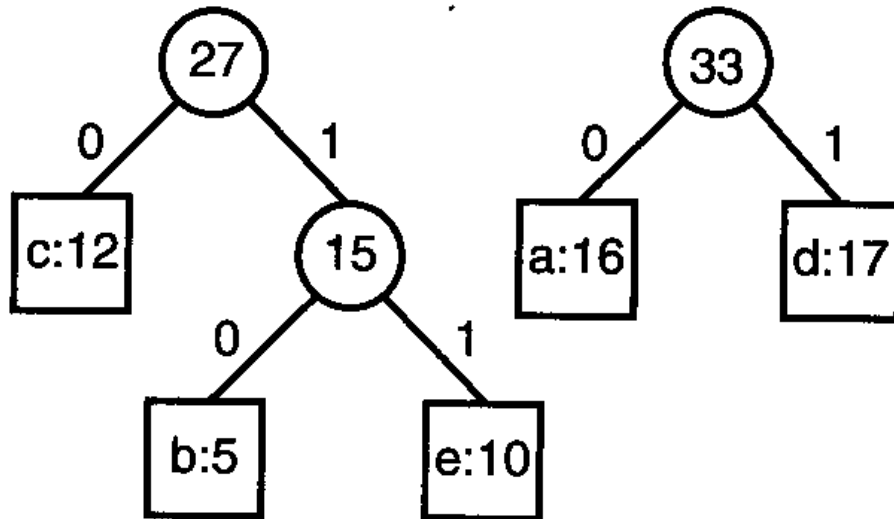
An example of application (3)

a:16 d:17 f:25

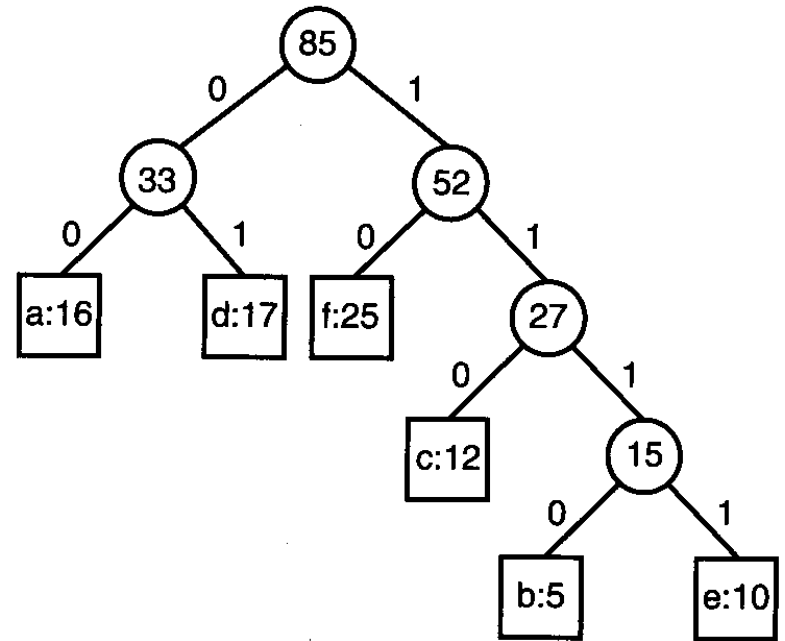
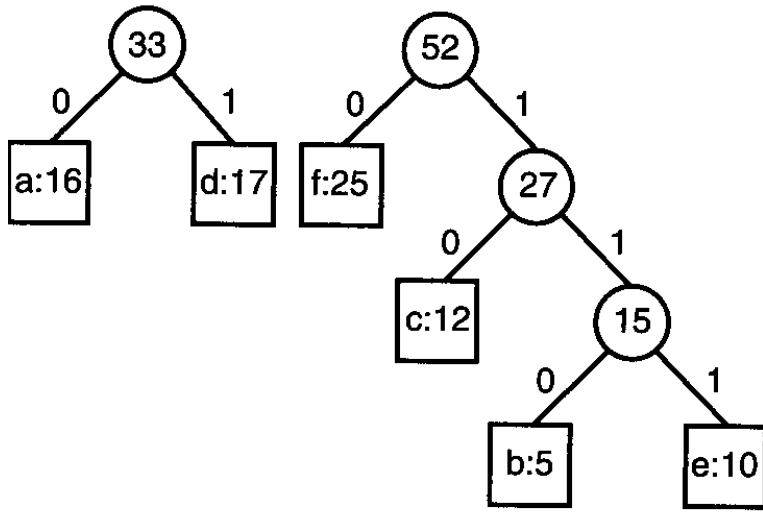


An example of application (4)

f:25



An example of application (5)



Priority queue *PQ*

- Arrange n pointers to `nodetype` records in *PQ* so that for each pointer p in *PQ*

$p \rightarrow symbol$ = a distinct character in the file

$p \rightarrow frequency$ = the frequency of that character in the file

$p \rightarrow left = p \rightarrow right = \text{NULL}$

for ($i = 1; i \leq n-1; i++$) { // There is no solution check; rather,
 $remove(PQ, p);$ // solution is obtained when $i = n - 1$.

$remove(PQ, q);$ // Selection procedure.

$r = \text{new nodetype};$ // There is no feasibility check.

$r \rightarrow left = p;$

$r \rightarrow right = q;$

$r \rightarrow frequency = p \rightarrow frequency + q \rightarrow frequency;$

$insert(PQ, r);$

}

$remove(PQ, r);$

return $r;$

Proof

- Lemma 4.4

The binary tree corresponding to an optimal binary prefix code is full. That is, every nonleaf has two children

- Theorem 4.5

Huffman's algorithm produces an optimal binary code

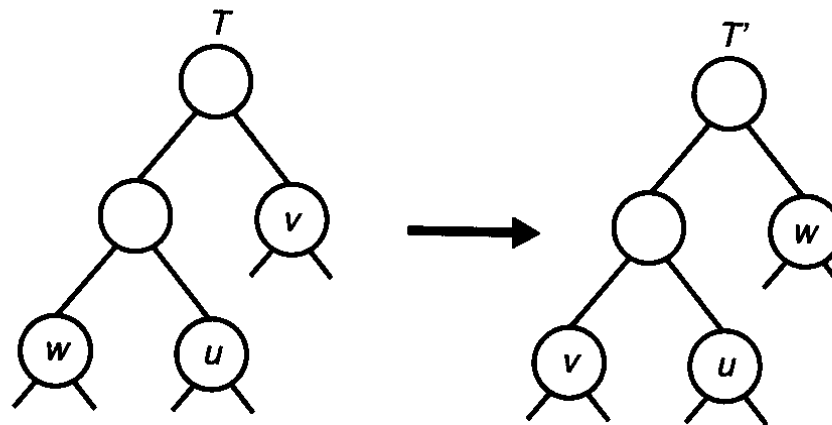


Figure 4.12 • The branches rooted at v and w are swapped.

The knapsack problem

مساله کوله پشته

□ The knapsack problem

■ Let

- $S = \{item_1, item_2, \dots, item_n\}$
- $w_i =$ weight of $item_i$
- $p_i =$ profit of $item_i$
- $W =$ maximum weight the knapsack can hold

□ کوله پشته کسری

□ The fractional knapsack problem

□ کوله پشته صفر و یک

□ The 0-1knapsack problem

روش حریصانه برای حل مساله کوله‌پشتی کسری

- Item 1 2 3
- Profit 50\$ 60\$ 140\$
- Weight 5 10 20
- Knapsack capacity = 30

□ مرتب‌سازی آیتم‌ها بر اساس منفعت وزنی

- Total profit in the previous example
 $\$50 + \$140 + (5/10)(\$60) = \220

روش حریصانه برای حل مساله کوله‌پشتی صفر و یک

□ The 0-1knapsack problem

■ Let

- $S = \{item_1, item_2, \dots, item_n\}$

- $w_i =$ weight of $item_i$

- $p_i =$ profit of $item_i$

- $W =$ maximum weight the knapsack can hold

■ Determine a subset A such that

$$\sum_{item_i \in A} p_i \text{ is maximized subject to } \sum_{item_i \in A} w_i \leq W$$

در مساله صفر و یک، حریصانه راه حل بهینه نمی دهد

□ $W = 30$

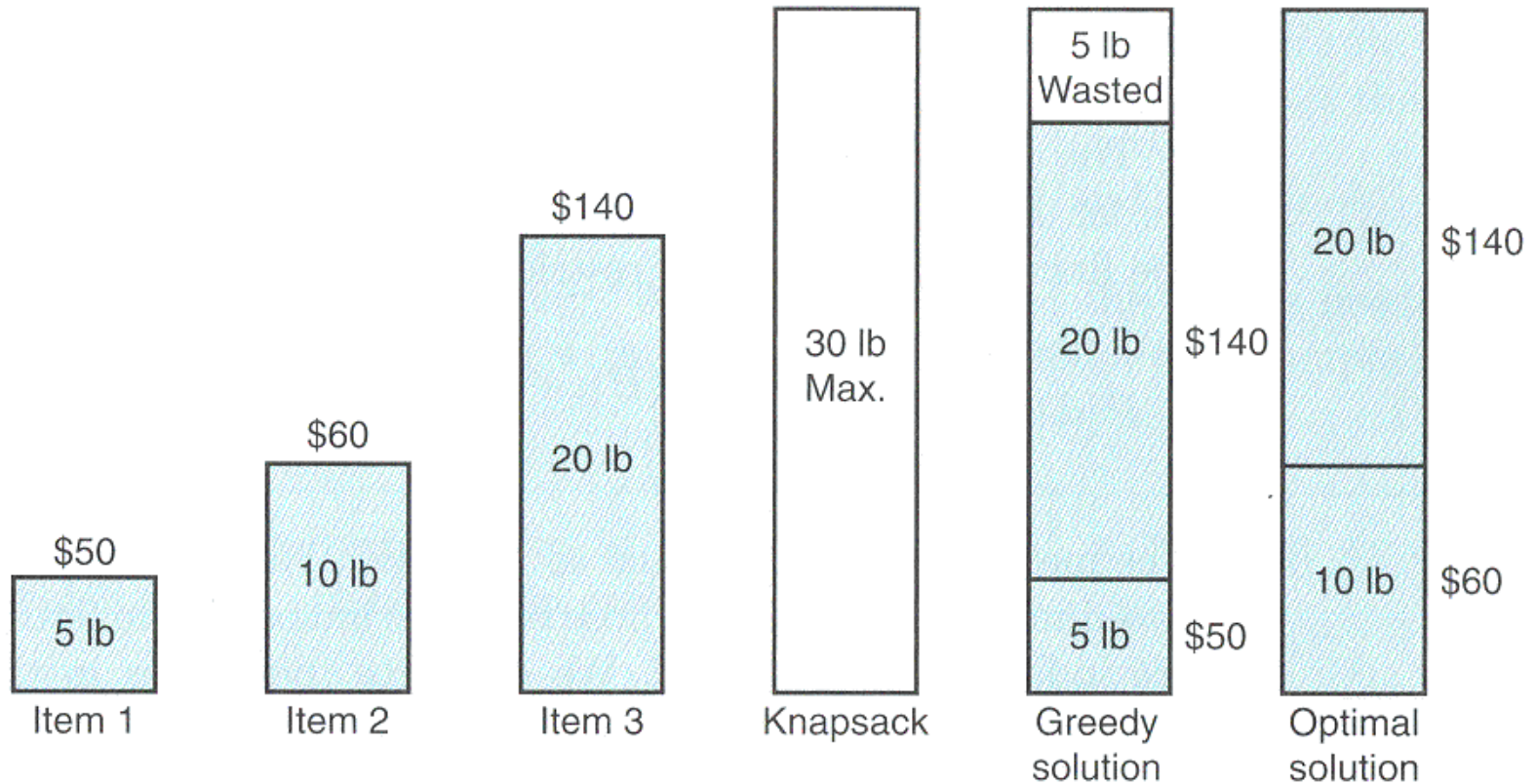


Figure 4.13 • A greedy solution and an optimal solution to the 0-1 Knapsack problem.