



# GREEDY ALGORITHMS, PART 2: THEORY OF GREEDY ALGORITHMS

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## Outline

1. Subset systems
2. Optimal caching
3. Shortest path

- A *subset system*  $S = (U, \mathcal{V})$  has
  - universe of elements  $U$
  - a set of feasible sets  $\mathcal{V}$ , closed under inclusion (=independent)
- For weights  $w_u \geq 0, u \in U$  we have the *optimization problem*

$$S^* = \operatorname{argmax}_{S \in \mathcal{V}} W(S) \quad (1)$$

to find the feasible set  $S^*$  of largest total weight

$$W(S^*) = \sum_{e \in S^*} w_e.$$

## Natural greedy algorithm

```

 $S := \emptyset$ 
while  $U \neq \emptyset$  do
  select  $u \in U$  s.t.  $w_u$  maximal  $O(n)$ 
   $U := U \setminus \{u\}$ 
  if  $S \cup \{u\} \in \mathcal{V}$  then  $\parallel$   $\{ O(n) \}$ 
     $S := S \cup \{u\}$ 
  end if
end while
return  $S$   $\Rightarrow O(n^2)$ 

```

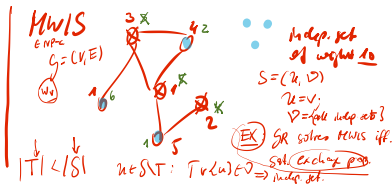
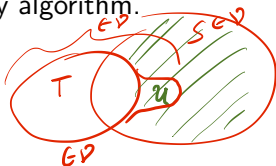
Handwritten annotations: Red circles around  $\emptyset$ ,  $w_u$ ,  $U \setminus \{u\}$ , and  $S$ . Red circles and brackets group the inner loop steps, with  $O(n)$  written next to each group. A large red bracket on the right groups the entire loop body with  $O(n)$ . A red arrow points from the loop body to  $\Rightarrow O(n^2)$ .

- A subset system has the exchange property (EP) if for all  $S, T \in \mathcal{V}$  s.t.  $|S| > |T|$  there is a  $u \in S \setminus T$  s.t.  $T \cup \{u\} \in \mathcal{V}$ .
- An independent subset system that satisfies EP is a matroid.
- Main result:

*The greedy algorithm solves the corresponding optimization problem of an independent subset system  $S = (U, \mathcal{V})$  iff  $S$  is a matroid.*

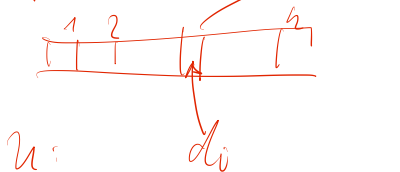
(for any weights)

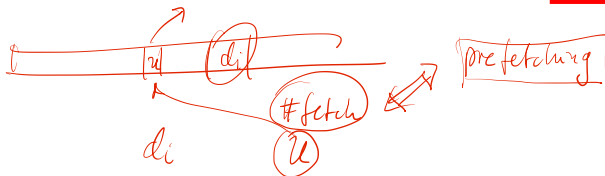
Proof:  $\rightarrow$ : exchange argument;  $\leftarrow$ : weights that fool the greedy algorithm.



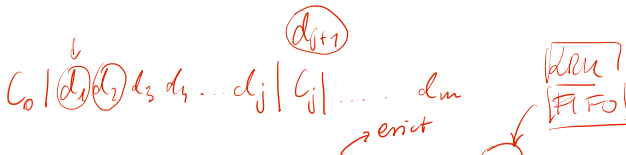
- Take a universe  $U$  of  $n$  elements, a cache of size  $k$ , initially some arbitrary items in cache.
- Consider a sequence of references  $d_1, d_2, \dots, d_m$ ,  $d_i \in U$  to the cache.
- To process  $d_i$ : either cache hit, or cache miss.
- On miss: fetch  $d_i$  and evict another element.

# fetches



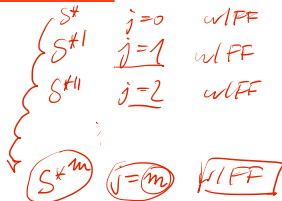


- In principle: we could take action at any time.
- A *reduced schedule*: acts only if  $d$  is required but not in cache.
- Every schedule can be made reduced, by lazy evaluation: postpone acts until needed.



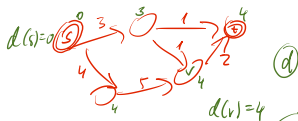
- Belady (1966): farthest in future schedule  $FF$  is optimal.
- **Claim:** If schedules  $S$  and  $FF$  agree on prefix  $j$  then there is  $S'$  that agrees on prefix  $j+1$  and does not cost more.
- This allows us to transform optimal schedule  $S^*$  to  $FF$ .

**Proof:** (more complicated) exchange argument.



SSSP  $\Leftarrow$  Shortest st-path.

APSP (all pairs shortest path)  $\Rightarrow$   $n \times$  SSSP



- Directed graph  $G = (V, A)$ , with distances  $d_a = d_{(u,v)} \geq 0$  on arcs  $a = (u, v) \in A$ .  $\in \mathbb{Z}_+$
- Source vertex  $v \in V$   $d(v) = \text{dist}(s, v)$
- Find: shortest distance  $\text{dist}(s, v)$  from  $s$  to all vertices  $v$  (single-source shortest path)  
(Store it in array  $d$ , initially  $d(s) = 0$  and  $d(v) = \infty, v \in V \setminus \{s\}$ ).

- 1) Shortest paths ✓
  - 2) Maximum flow ✓
  - 3) Max. matchings ✓
- } tP

Big 3



- To **relax** an edge  $(u, v)$ :

if  $d(v) > d(u) + d_{uv}$ :  $d(v) := d(u) + d_{uv}$

- To **relax** a vertex  $v$ :

for  $w \in N^+(v)$ :  $\text{relax}((v, w))$



- Key insight** (Dijkstra):

(Greedy) Relaxing vertices in order of non-increasing current distances  $d$  is optimal.

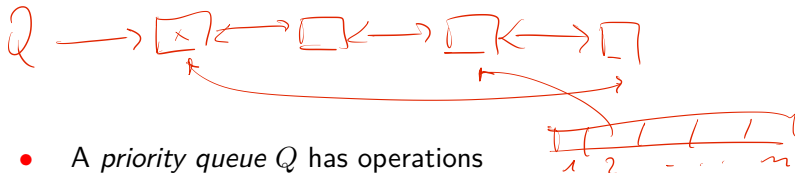
$n$  relaxations

Once

$O(n)$  all vertices **unvisited**; set **initial** distances  
 $n \times$  while there's a <sup>unvisited</sup> **unvisited** vertex  
 (a) • select unvisited vertex  $v$  of minimal  $d(v)$  (\*)  
 relax( $v$ ) and mark  $v$  as **visited** (current)  
 end while  
 $O(\delta^+(v)) \rightsquigarrow O(\sum_{\text{relax}} \delta^+(v)) = O(m)$   
 $d(v) = \text{dist}(s, v)$   
 Greedy  
 delete

- **Requires:** data structure that finds the next vertex quickly.
- Otherwise: what's the complexity

$$O(m^2 + m) \stackrel{m \leq n^2}{=} O(n^2) \quad \leftarrow \text{Dijkstra}$$



- A priority queue  $Q$  has operations

$Q := \emptyset$  create an empty priority queue

$\text{push}(Q, (e, k))$  add element  $e$  with key  $k$

$\text{deletemin}(Q)$  remove and return element of smallest key

$\text{update}((e, k))$  set key of element  $e$  to  $k$

$m > n$  connected

Dijkstra

$$O\left(\underbrace{n \cdot \text{deletemin}}_{\substack{n \log n \\ \text{out}}} + \underbrace{n \cdot \text{insert}}_{\text{out}} + \underbrace{m \cdot \text{update}}_{\substack{O(1) \\ \text{Fib heap}}}\right)$$

Sort  $n$  numbers:

$n \times \text{insert}$   
 $n \times \text{deletemin}$

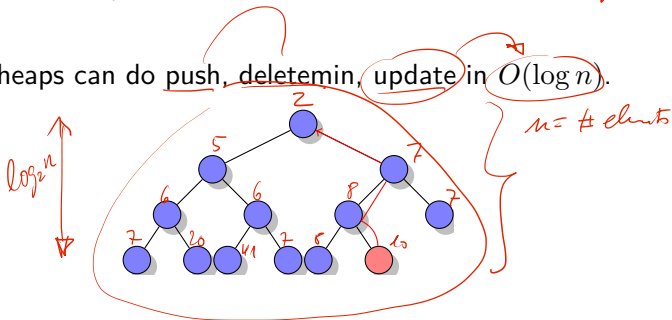
$\Omega(n \log n)$

$O(n(\text{insert} + \text{deletemin}))$   
 $\log n$

$O(n \log n + m)$

$$O(n \log n + m \log n) = O(\underline{(n+m)} \cdot \log n)$$

- Binary heaps can do push, deletemin, update in  $O(\log n)$ .



```

 $d_s := 0; d_v := \infty, \forall v \in V \setminus \{s\}$ 
 $\text{visited}(v) := \text{false}, \forall v \in V$ 
 $Q := \emptyset; \text{insert}(Q, (s, 0))$ 
while  $Q \neq \emptyset$  do
     $v := \text{deletemin}(Q)$ 
     $\text{visited}(v) := \text{true}$ 
    for  $u \in N^+(v) \mid \text{notvisited}(u)$  do
        if  $d_u = \infty$  then
             $d_u := d_v + d_{vu}$ 
             $\text{insert}(Q, (u, d_u))$ 
        else if  $d_v + d_{vu} < d_u$ 
             $d_u := d_v + d_{vu}$ 
             $\text{update}(Q, (u, d_u))$ 
        end if
    end for
end while

```

*Handwritten notes:*

- $\rightarrow n \times \text{deletion}$  (pointing to  $\text{deletemin}(Q)$ )
- $(*) d(v) = \text{dist}(s, v)$  (next to  $\text{visited}(v) := \text{true}$ )
- $n \times \text{insert}$  (next to  $\text{insert}(Q, (u, d_u))$ )
- $m \times \text{update}$  (next to  $\text{update}(Q, (u, d_u))$ )