

# **CMPSC 465**

## **Data Structures and Algorithms**

### **Spring 2022**

---

Instructor: Chunhao Wang

# Greedy algorithms

---

# Greedy algorithms

---

**Matroid, Task Scheduling**  
**(Cormen et al. 16.4, 16.5)**

Very abstract!

“Computer Science is a science of **abstraction** — creating the right model for a problem and devising the appropriate mechanizable techniques to solve it.”

— Alfred Aho

# Matroid

**Matroid** is a combinatorial structure

Many problems for which a greedy approach provides optimal solution can be formulated as some problems involve matroids

A more abstract view of graph vs. matroid

Graph  $G = (V, E)$

1.  $V$ : finite nonempty set
2.  $E$ : a collection of subsets of  $V$  (or  $E \subseteq \mathcal{P}(V)$ )  
each  $e \in E$  has two elements of  $V$   
called an *edge*

**Matroid**  $M = (S, \mathcal{I})$

1.  $S$ : finite nonempty set
2.  $\mathcal{I} \subseteq \mathcal{P}(S)$  s.t.
  - if  $A \subseteq B$  and  $B \in \mathcal{I}$   
then  $A \in \mathcal{I}$  (Hereditary property)
  - if  $A, B \in \mathcal{I}$  and  $|A| < |B|$   
then  $\exists x \in B - A$  s.t.  
 $A \cup \{x\} \in \mathcal{I}$  (Exchange property)

For a matroid  $M = (S, \mathcal{I})$ , each  $A \in \mathcal{I}$  is called an **independent subset**

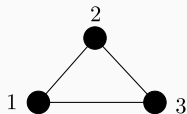
# Graphic Matroid

Given undirected  $G = (V, E)$ , construct **graphic matroid**  $M_G = (S, \mathcal{I})$  via

- $S = E$
- $\mathcal{I} = \{A \subseteq E : A \text{ is acyclic}\}$   
 $A$  is a forest

Why  $M_G$  is a matroid?

- Hereditary: a subset of forest is still a forest
- Exchange: demonstrate by example



$$S = \{(1, 2), (2, 3), (3, 1)\}$$

$$\mathcal{I} = \{\emptyset, \{(1, 2)\}, \{(2, 3)\}, \{(1, 3)\}, \{(1, 2), (2, 3)\}, \\ \{(1, 3), (1, 2)\}, \{(1, 3), (2, 3)\}\}$$

$$\text{Say } A = \{(2, 3)\}, B = \{(1, 3), (1, 2)\}$$

$$x \in B - A, \text{ for example, } x = (1, 3)$$

$$\text{then } A \cup \{x\} = \{(2, 3), (1, 3)\} \subseteq \mathcal{I}$$

# Connection to spanning tree

## Definition

For all  $A \in \mathcal{I}$ ,  $x \in S$  is an **extension** of  $A$  if  $A \cup \{x\} \in \mathcal{I}$

## Definition

$A \in \mathcal{I}$  is **maximal** if it has no extension

## Theorem

*All maximal  $A \in \mathcal{I}$  have the same size*

## Proof.

Suppose  $A, B \in \mathcal{I}$  are both maximal, but  $|B| > |A|$ . Then by exchange property, there exists an  $x \in B - A$  s.t.  $A \cup \{x\} \in \mathcal{I}$ , which is a contradiction of  $A$  being maximal □

For connected undirected  $G$ , every maximal independent subset of  $M_G$  must be a tree with  $|V| - 1$  edges. Hence it is a spanning tree



# Weighted matroid

## Definition

A **weighted matroid**  $M = (S, \mathcal{I})$  is one that has a strictly positive weight  $w(x)$  for all  $x \in S$ . The weight function  $w$  extends to  $\mathcal{I}$  as for all  $A \in \mathcal{I}$ :

$$w(A) = \sum_{a \in A} w(a)$$

**Note:** for graphic matroids, weight of  $M_G$  is corresponding to edge weights

# Optimization problem for matroids

## Problem (Maximum-weighted Independent Subset)

*Given a weighted matroid  $M$ , the goal is to find the maximum-weighted independent subset of  $M$*

**Remark:** because weights are positive, it always helps to find a subset as large as possible

**Application:** MST of  $G \rightarrow$  max-weighted independent subset of  $M_G$  via

- $G$  with  $w(e) \rightarrow M_G$  with  $w'(e) = c - w(e)$  where  $c$  is a constant larger than the largest  $w(e)$
- For  $M_G$ ,  $w'(e)$  are positive
- For max-weighted independent subset  $A$   
 $w'(A) = (|V| - 1)c - w(A)$ , so  $w(A)$  is minimized

Hence a max-weighted indep. subset of  $M_G$  corresponds to an MST of  $G$

# Pseudocode for finding max-weighted independent subset

```
1 def GREEDY( $M = (S, \mathcal{I})$ , weights  $w$ ):  
2   Set  $A := \{\}$ ;  
3   Sort  $S$  in decreasing order of  $w$  ; //  $O(n \log n)$   
4   for  $x \in S$ :  
5     if  $A \cup \{x\} \in \mathcal{I}$ :  
6        $A := A \cup \{x\}$ ;  
7   return  $A$ ;
```

**Proof of correctness:** Cormen et al. 16.4

**Running time:** let  $n = |S|$

Assume checking if  $A \cup \{x\} \in \mathcal{I}$  takes  $O(f(n))$ . Lines 5-6 takes  $O(n \cdot f(n))$

**Total running time:**  $O(n \log n + n \cdot f(n))$

# Application: task scheduling

## Problem (Task scheduling)

### Setup:

- $n$  unit-time tasks  $a_1, \dots, a_n$
- $d_1, \dots, d_n$  deadlines for each task,  $1 \leq d_i \leq n$
- $w_1, \dots, w_n > 0$  penalties if  $a_i$  is not completed by  $d_i$

**Goal:** Find a **schedule** (i.e., permutation of tasks) that minimizes the penalties incurred

	task	a	b	c	d
Example:	deadline	1	1	4	2
	penalty	5	10	1	3

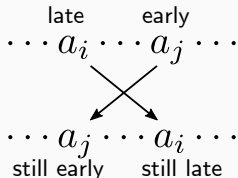
A (non-optimal) schedule	b	a	c	d	penalty: 8
	✓	✗	✓	✗	

# The canonical form of a schedule

## Definition

In a schedule, a task is **early** if it finishes before its deadline; a task is **late** if it finishes after its deadline

We can transfer any schedule into the **early-first** form, i.e., early tasks before late ones



## Definition

A schedule is in the **canonical form** if it's early-first and its early tasks are ordered by increasing deadlines

We can transfer any schedule into its canonical form

# Finding optimal schedule using matroid

How to find an optimal schedule?

1. Optimizing over tasks in the canonical form:
  - 1.1 Find a set  $A$  of tasks that are early
  - 1.2 Sort the tasks of  $A$  in increasing deadlines
  - 1.3 Add late tasks in any order
2. Minimize penalties of late tasks  $\equiv$  maximize penalties of early tasks

Modeled by a matroid  $M = (S, \mathcal{I})$ , where

$$S = \{a_1, \dots, a_n\}$$

$$\mathcal{I} = \{A \subseteq S : \exists \text{ a way to schedule the tasks in } A \text{ s.t. no task is late}\}$$

$w$  : penalty

Finding an optimal schedule  $\equiv$  finding max-weighted indep. subset of  $M$