

# Lecture 17: Matroids and the Greedy Algorithm

Michael Dinitz

October 26, 2021

601.433/633 Introduction to Algorithms

# Introduction

Last time: somewhat greedy algorithm (Prim's), extremely greedy algorithm (Kruskal's)

Question: when does greedy algorithm return optimal solution?

Want abstraction that includes MSTs, but also works for many other problems.

*Weighted Set System:*

- ▶ Universe  $\mathbf{U}$
- ▶ Collection  $\mathcal{I} \subseteq 2^{\mathbf{U}}$  (so  $\mathbf{I} \subseteq \mathbf{U}$  for all  $\mathbf{I} \in \mathcal{I}$ ). Called *independent sets*
- ▶ Weights  $\mathbf{w} : \mathbf{U} \rightarrow \mathbb{R}^+$

Problem: find *max weight* independent set

## MST as Weighted Set System

MST: weighted graph  $\mathbf{G} = (\mathbf{V}, \mathbf{E}, \mathbf{w})$ . Find MST.

Set system:

- ▶  $\mathbf{U} = \mathbf{E}$
- ▶  $\mathcal{I} = \{\mathbf{F} \subseteq \mathbf{E} : (\mathbf{V}, \mathbf{F}) \text{ a forest}\}$

What about weights? MST is minimize, but problem we defined is maximize.

- ▶ Let  $\bar{\mathbf{w}} > \mathbf{w}(\mathbf{e})$  for all  $\mathbf{e} \in \mathbf{E}$ , let  $\mathbf{w}'(\mathbf{e}) = \bar{\mathbf{w}} - \mathbf{w}(\mathbf{e})$  for all  $\mathbf{e} \in \mathbf{E}$

For any tree  $\mathbf{T}$ :

$$\mathbf{w}'(\mathbf{T}) = \sum_{\mathbf{e} \in \mathbf{T}} \mathbf{w}'(\mathbf{e}) = \sum_{\mathbf{e} \in \mathbf{T}} (\bar{\mathbf{w}} - \mathbf{w}(\mathbf{e})) = (\mathbf{n} - 1)\bar{\mathbf{w}} - \sum_{\mathbf{e} \in \mathbf{T}} \mathbf{w}(\mathbf{e})$$

So under weights  $\mathbf{w}'$ , max-weight IS = max-weight forest = max-weight tree = min-weight tree (weights  $\mathbf{w}$ )

- ▶ So finding max-weight forest = finding min spanning tree.

## Useful Properties of Forests

Let  $\mathbf{U} = \mathbf{E}$  and  $\mathcal{I} = \{\mathbf{F} \subseteq \mathbf{E} : (\mathbf{V}, \mathbf{F}) \text{ a forest}\}$

Useful properties:

1.  $\emptyset \in \mathcal{I}$
2. If  $\mathbf{F} \in \mathcal{I}$  and  $\mathbf{F}' \subseteq \mathbf{F}$ , then  $\mathbf{F}' \in \mathcal{I}$
3. *Augmentation Property*: If  $\mathbf{F}_1 \in \mathcal{I}$  and  $\mathbf{F}_2 \in \mathcal{I}$  with  $|\mathbf{F}_2| > |\mathbf{F}_1|$ , then there is some edge  $e \in \mathbf{F}_2 \setminus \mathbf{F}_1$  such that  $\mathbf{F}_1 \cup \{e\} \in \mathcal{I}$ .

### Proof Sketch that Forests have Augmentation Property.

Suppose false: no edge in  $\mathbf{F}_2 \setminus \mathbf{F}_1$  can be added to  $\mathbf{F}_1$ . Let  $c_1 = \#$  components in  $\mathbf{F}_1$ ,  $c_2 = \#$  components in  $\mathbf{F}_2$

$\implies$  every edge of  $\mathbf{F}_2$  has both endpoints in same component of  $\mathbf{F}_1$

$\implies$  every component of  $\mathbf{F}_2$  contained in component of  $\mathbf{F}_1 \implies c_2 \geq c_1$

But  $c_2 = n - |\mathbf{F}_2| < n - |\mathbf{F}_1| = c_1$ .

Contradiction. □

# Matroids

## Definition

$(\mathbf{U}, \mathcal{I})$  is a *matroid* if the following three properties hold:

1.  $\emptyset \in \mathcal{I}$ ,
2. If  $\mathbf{F} \in \mathcal{I}$  and  $\mathbf{F}' \subseteq \mathbf{F}$ , then  $\mathbf{F}' \in \mathcal{I}$ , and
3. If  $\mathbf{F}_1 \in \mathcal{I}$  and  $\mathbf{F}_2 \in \mathcal{I}$  with  $|\mathbf{F}_2| > |\mathbf{F}_1|$ , then there is some element  $\mathbf{e} \in \mathbf{F}_2 \setminus \mathbf{F}_1$  such that  $\mathbf{F}_1 \cup \{\mathbf{e}\} \in \mathcal{I}$ .

$(\mathbf{U}, \mathcal{I})$  is a *hereditary set system* if the first two properties hold.

Matroid theory: super interesting area of combinatorics! Surprising amount of structure.

Warmup: In any matroid, the maximal independent sets (called [bases](#)) have the same size (called the [rank](#) of the matroid).

# Examples of Matroids

- ▶ Forests in graphs
- ▶ Linearly independent vectors in vector space
  - ▶  $\mathbf{U}$  a finite set of vectors in  $\mathbb{R}^d$
  - ▶  $\mathcal{I} = \{\mathbf{F} \subseteq \mathbf{U} : \mathbf{F} \text{ linearly independent}\}$
  - ▶  $\emptyset$  linearly independent
  - ▶ If  $\mathbf{F}$  linearly independent and  $\mathbf{F}' \subseteq \mathbf{F}$ , then  $\mathbf{F}'$  linearly independent
  - ▶ Augmentation: if  $\mathbf{F}_1$  linearly independent,  $\mathbf{F}_2$  linearly independent, and  $|\mathbf{F}_2| > |\mathbf{F}_1| \implies \dim(\text{span}(\mathbf{F}_1)) = |\mathbf{F}_1| < |\mathbf{F}_2| = \dim(\text{span}(\mathbf{F}_2))$

Matroids: generalize both graph theory and linear algebra!

- ▶ Originally invented by Whitney as an attempt to generalize the concept of “linear independence”

# Representation

To do algorithms with matroids, need to figure out how they're represented.

Option 1: list all independent sets

- ▶ Too many of them!

What did we need for MST (Kruskal)?

**Independence Oracle:** algorithm which take  $\mathbf{F} \subseteq \mathbf{U}$ , returns YES if  $\mathbf{F} \in \mathcal{I}$ , NO if  $\mathbf{F} \notin \mathcal{I}$

For MST: “does  $\mathbf{F}$  have any cycles”? Independence oracle: DFS/BFS

We'll assume we have independence oracle.

# Greedy Algorithm

Kruskal, generalized to matroids (and max weight)!

$\mathbf{F} = \emptyset$

Sort  $\mathbf{U}$  by weight (largest to smallest)

For each  $\mathbf{u} \in \mathbf{U}$  in sorted order {

    If  $\mathbf{F} \cup \{\mathbf{u}\} \in \mathcal{I}$ , add  $\mathbf{u}$  to  $\mathbf{F}$

}

Return  $\mathbf{F}$



# Correctness

## Theorem

Let  $\mathbf{F}$  be independent set returned by greedy. Then  $\mathbf{w}(\mathbf{F}) \geq \mathbf{w}(\mathbf{F}')$  for all  $\mathbf{F}' \in \mathcal{I}$ .

- ▶  $\mathbf{F} = \{\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_r\}$ , where  $\mathbf{w}(\mathbf{f}_i) \geq \mathbf{w}(\mathbf{f}_{i+1})$  for all  $i$  (order added by greedy)
- ▶  $\mathbf{F}' = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_r\}$  where  $\mathbf{w}(\mathbf{e}_i) \geq \mathbf{w}(\mathbf{e}_{i+1})$  for all  $i$

**Claim:**  $\mathbf{w}(\mathbf{f}_i) \geq \mathbf{w}(\mathbf{e}_i)$  for all  $i$ .

**Proof:** Suppose false, let  $j$  smallest integer such that  $\mathbf{w}(\mathbf{f}_j) < \mathbf{w}(\mathbf{e}_j)$ .

Let  $\mathbf{F}_1 = \{\mathbf{f}_1, \dots, \mathbf{f}_{j-1}\}$  and let  $\mathbf{F}_2 = \{\mathbf{e}_1, \dots, \mathbf{e}_j\}$

$|\mathbf{F}_2| > |\mathbf{F}_1|$ , so by augmentation there is some  $\mathbf{e}_z \in \mathbf{F}_2 \setminus \mathbf{F}_1$  such that  $\mathbf{F}_1 \cup \{\mathbf{e}_z\} \in \mathcal{I}$ .

$$\mathbf{w}(\mathbf{e}_z) \geq \mathbf{w}(\mathbf{e}_j) > \mathbf{w}(\mathbf{f}_j)$$

Contradiction! Greedy would add  $\mathbf{e}_z$  next, not  $\mathbf{f}_j$ .

## Converse

So greedy works on matroids. Amazing fact: if greedy works, set system is a matroid!

### Theorem

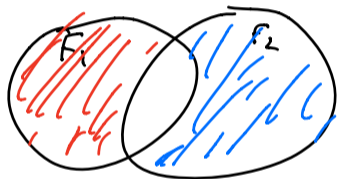
*Let  $(\mathbf{U}, \mathcal{I})$  be an hereditary set system. If for every weighting  $\mathbf{w} : \mathbf{U} \rightarrow \mathbb{R}_{\geq 0}$  the greedy algorithm returns a maximum weight independent set, then  $(\mathbf{U}, \mathcal{I})$  is a matroid.*

So for hereditary set systems, matroids exactly characterize when the greedy algorithm works!

# Proof

Contradiction. Suppose false  $\implies (\mathbf{U}, \mathcal{I})$  hereditary but not matroid.

$\implies \exists \mathbf{F}_1, \mathbf{F}_2 \in \mathcal{I}$  such that  $|\mathbf{F}_1| < |\mathbf{F}_2|$  but  $\mathbf{F}_1 \cup \{\mathbf{e}\} \notin \mathcal{I}$  for all  $\mathbf{e} \in \mathbf{F}_2 \setminus \mathbf{F}_1$



Easy facts:

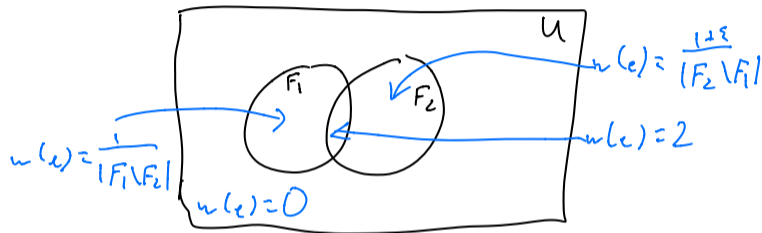
1.  $|\mathbf{F}_2 \setminus \mathbf{F}_1| > |\mathbf{F}_1 \setminus \mathbf{F}_2|$
2.  $|\mathbf{F}_2 \setminus \mathbf{F}_1| \geq 1$
3.  $|\mathbf{F}_1 \setminus \mathbf{F}_2| \geq 1$  (hereditary)

$\implies \exists \epsilon > 0$  such that  $0 < (1 + \epsilon)|\mathbf{F}_1 \setminus \mathbf{F}_2| < |\mathbf{F}_2 \setminus \mathbf{F}_1|$

$$\implies \frac{1}{|\mathbf{F}_1 \setminus \mathbf{F}_2|} > \frac{1 + \epsilon}{|\mathbf{F}_2 \setminus \mathbf{F}_1|}$$

## Proof (cont'd)

Use fact that  $\frac{1}{|F_1 \setminus F_2|} > \frac{1+\epsilon}{|F_2 \setminus F_1|}$  to define weights.



Greedy:

- ▶ Adds all of  $F_1 \cap F_2$
- ▶ Adds all of  $F_1 \setminus F_2$
- ▶ Can't add any of  $F_2 \setminus F_1$

$$\begin{aligned}w(\text{greedy}) &= 2|F_1 \cap F_2| + |F_1 \setminus F_2| \frac{1}{|F_1 \setminus F_2|} \\ &= 2|F_1 \cap F_2| + 1\end{aligned}$$

$$\begin{aligned}w(F_2) &= 2|F_1 \cap F_2| + |F_2 \setminus F_1| \frac{1+\epsilon}{|F_2 \setminus F_1|} \\ &= 2|F_1 \cap F_2| + 1 + \epsilon\end{aligned}$$

Greedy not optimal: contradiction!