

# Matroids

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## What's a Matroid

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- Linear Matroid

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- Matroid Optimization Problem

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- Finding Minimum Spanning Tree with GreedyBasis

- Scheduling with Deadlines

## What's a Matroid

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1. Definition of Matroid
2. Independent Systems
3. Independent Set, Basis, Rank

We say  $(S, \mathcal{I})$  is a **matroid** if  $S$  is a finite *ground* set and  $\mathcal{I}$  is a collection of subsets of  $S$  such that the following properties are satisfied:

1. **Non-emptiness:** The empty set is in  $\mathcal{I}$ . (Thus,  $\mathcal{I}$  is not itself empty.)
2. **Heredity:** If  $\mathcal{I}$  contains a subset  $X \subseteq S$ , then  $\forall Y \subset X$ ,  $Y \in \mathcal{I}$ .
3. **Exchange:** If  $X$  and  $Y$  are two sets in  $\mathcal{I}$  where  $|X| > |Y|$ , then  $\exists x \in X \setminus Y$ , such that  $Y \cup x \in \mathcal{I}$ .

$(S, \mathcal{I})$  is called an **Independent System** if it satisfies the first two properties: non-emptiness and heredity properties.

A matroid is an independent system satisfying the exchange property.

1. Every  $X \in \mathcal{I}$  is typically called an **independent set**.
2. An independent set  $X \in \mathcal{I}$  is called a **basis** of the matroid if  $\nexists Y \in \mathcal{I}$  such that  $X \subset Y$ , i.e., a basis is a maximal independent set of  $\mathcal{I}$ .
3. Every basis of a matroid  $(S, \mathcal{I})$  has the same size, and we denote this size as the **rank** of this matroid.

It is easy to use the exchange property to prove the uniqueness of the size of bases in a matroid by contradiction:

Suppose a matroid  $(S, \mathcal{I})$  has two maximal independent sets  $X$  and  $Y$  with sizes  $|X| > |Y|$ , then there must exist some element  $x \in X \setminus Y$  such that  $Y \cup x \in \mathcal{I}$ . Then  $Y$  is not a basis of  $(S, \mathcal{I})$ . A contradiction.

## Examples

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1. Linear Matroid
2. Uniform Matroid
3. Graphic Matroid



Let  $M$  be an arbitrary  $n \times m$  matrix,  $S$  be set of all  $m$  columns in  $M$ . Let  $\mathcal{I}$  be a collection of all the linearly independent subsets of  $S$ . Then  $(S, \mathcal{I})$  is a matroid.

(We say a subset  $X \subseteq S$  is a linearly independent subset if all the columns in  $X$  are linearly independent.)

A subset  $X \subseteq \{1, 2, \dots, n\}$  is independent if and only if  $|X| \leq k$ . Any subset of  $\{1, 2, \dots, n\}$  of size  $k$  is a basis of  $U_{k,n}$ .

Let  $G = (V, E)$  is an undirected graph. Let  $S = E$  and  $\mathcal{I}$  be the collection of all sets of edges that induce a forest of  $G$ .  $(S, \mathcal{I})$  is the graphic matroid  $M(G)$  of  $G$ . Every spanning tree of  $G$  is a basis of  $M(G)$ .

## Applications of Matroids

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1. Matroid Optimization Problem
2. A Greedy Method Solving Matroid Optimization Problem
3. Find Minimum Spanning Tree with Greedy Method
4. Scheduling with Deadlines

A matroid  $(S, \mathcal{I})$  is **weighted** if every element  $e \in S$  is weighted non-negatively by a weight function  $w : S \rightarrow \mathbb{R}_{\geq 0}$ .

The **matroid optimization problem** is finding a basis with maximum total weight.

For example, if  $M(G)$  is a graphic matroid for a undirected graph  $G = (V, E)$ , then the problem is finding a maximum spanning tree of  $G$ .

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GREEDYBASIS( $S[1..n], \mathcal{I}, w[1..n]$ ):  
  sort  $S$  in decreasing order of weight  $w$   
   $G \leftarrow \emptyset$   
  for  $i \leftarrow 1$  to  $n$   
    if  $G \cup \{S[i]\} \in \mathcal{I}$   
       $G \leftarrow G \cup \{S[i]\}$   
  Return  $G$ 
```

### Theorem (3.1)

*For any matroid  $M = (S, \mathcal{I})$  and any weight function  $w : S \rightarrow \mathbb{R}^+$ ,  $\text{GreedyBasis}(S, \mathcal{I}, w)$  returns a maximum-weight basis of  $M$ .*

Firstly, we introduce the greedy choice property:

Lemma (3.2)

*Let  $x$  be the first element in the weight decreasing order list s.t.  $\{x\} \in \mathcal{I}$ , then there is a basis  $A$  containing  $x$ .*

**Proof:** Suppose otherwise there exists another basis  $B$  with maximum total weight not containing  $x$ . We construct a subset  $A$  starting from  $A = \{x\}$ , according to the exchange property, when  $|A| < |B|$ , we could always find an element  $y \in B - A$  make  $A \cup \{y\} \in \mathcal{I}$ . Let  $A = A \cup \{y\}$  and repeat this step until we finally get a basis  $|A| = |B|$ . Since  $w(x) \geq w(y), \forall y \in B$ , we have  $w(A) \geq w(B)$ . □



Now we show the optimal substructure property:

### Lemma (3.3)

*Let  $x$  be the first element chosen by GreedyBasis. We define a matroid  $M' = (S', \mathcal{I}')$  by:*

$$\begin{aligned} S' &= \{y \in S \mid \{x, y\} \in \mathcal{I}\} \\ \mathcal{I}' &= \{B \subseteq S' \mid B \cup \{x\} \in \mathcal{I}\} \end{aligned}$$

*Then finding a maximal total weight basis containing  $x$  in  $M$  reduces finding a maximal total weight basis in  $M'$ .*

## Theorem (3.4)

*For any subset system  $\mathcal{S}$  that is not a matroid, there is a weight function  $w$  such that  $\text{GreedyBasis}(\mathcal{S}, w)$  does not return a maximum-weight set in  $\mathcal{S}$ .*

We already know that every basis of a graphic matroid  $M(G)$  is a spanning tree of  $G = (V, E)$ , and  $\text{GreedyBasis}(S, \mathcal{I}, w)$  gives us a basis with maximum total weight w.r.t. weight function  $w$ .

Let  $w' : E \rightarrow \mathbb{R}$  be the original edge weight function for  $G$ .

Firstly, we make every edge has non-negative weight by setting  $w''(e) = w'(e) - \min_{e' \in E} \{w'(e')\}$ . After that, we construct  $w : E \rightarrow \mathbb{R}$  by assigning  $w(e) = (\sum_{e \in E} w''(e)) - w''(e)$ .

$\text{GreedyBasis}(S, \mathcal{I}, w)$  gives us a minimum spanning tree of  $G$ .  
(Now  $\text{GreedyBasis}$  actually works as Kruskal's algorithm.)

Suppose you have a list  $T$  of  $n$  tasks  $[t_1, \dots, t_n]$  to do, each of them needs a full day to finish, a list  $D = [d_1, \dots, d_n]$  of deadlines for those tasks, and a list  $P = [p_1, \dots, p_n]$  of overdue penalties for  $T$ .

A schedule is a permutation  $\pi$  of  $\{1, 2, \dots, n\}$  which is an order you will work on the tasks. The question is how to find a schedule  $\pi$  such that the total penalty is minimized, i.e.

$$\min_{\pi} \sum_{i=1}^n P[i] \cdot [\pi[i] > D[i]]$$

The cost of a schedule is determined by the subset of tasks that are on time.

Let  $\pi$  be a schedule.

Task  $t_i$  is late in  $\pi$  if it finishes after its deadline  $d_i$ , otherwise, it is early.

$\pi$  is in early-first form if all early tasks precede late tasks.

$\pi$  is in canonical form if it is early-first and all early tasks are scheduled in non-decreasing order of deadlines.

**Every schedule  $\pi$  can be put into canonical form.**

**Definition:** A set  $A$  of tasks is independent if there exists a schedule in which no task  $t \in A$  is late.

Let  $N_j(A) := |\{t_i \in A | d_i \leq j\}|$  for  $j = 0, \dots, n$ .

Lemma (3.5)

$A$  is independent  $\iff N_j(A) \leq j$  for  $j = 0, \dots, n$ .

**Theorem (3.6)**

Let  $S$  be a set of all tasks in  $T$ ,  $\mathcal{I}$  be the collection of all independent sets of tasks. Then  $(S, \mathcal{I})$  is a matroid.

**Proof:**

1. Hereditary: from the definition.
2. Exchange Property: consider two independent sets of tasks  $A$  and  $B$  with  $|A| < |B|$ .

Let  $k$  be largest  $j$  s.t.  $N_j(A) \geq N_j(B)$ , then  $k < n$  and for  $j = k + 1, \dots, n$ :  $N_j(A) < N_j(B)$ .

Choose  $x \in \{t_i \in B - A | d_i = k + 1\}$ , then  $N_j(A \cup \{x\}) = N_j(A)$  for  $j = 0, \dots, k$  and  $N_j(A \cup \{x\}) \leq N_j(A) + 1 \leq N_j(B) \leq j$  for  $j = k + 1, \dots, n$ . Thus,  $A \cup \{x\}$  is an independent set.  $\square$

1. Construct the weight function  $w$  by setting  $w(t_i) = p_i$  for all  $i = 1, \dots, n$ . Let  $M = (S, \mathcal{I})$  as we defined in Theorem 3.6.
2. Use GreedyBasis to find a maximum-weight basis  $A$  of tasks.
3. Create an optimal schedule by first scheduling tasks in  $A$  by order of non-decreasing deadlines.

**Time complexity:**  $O(n^2)$  (Sorting uses  $O(n \log n)$ . Checking independence costs  $O(n)$  for each iteration. In total  $O(n^2)$ .)