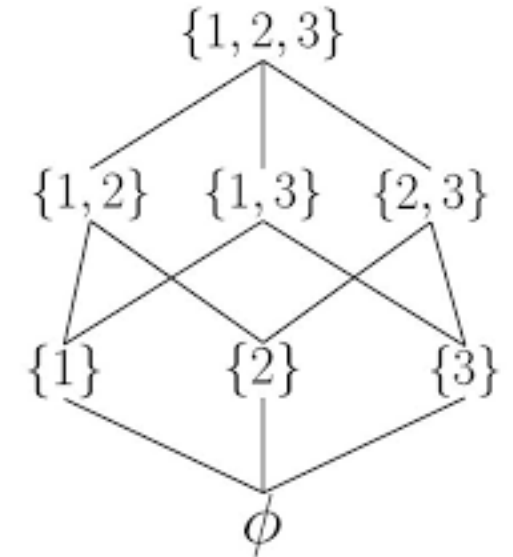


DP for Optimal Subset Selection Problem (Abstract View)

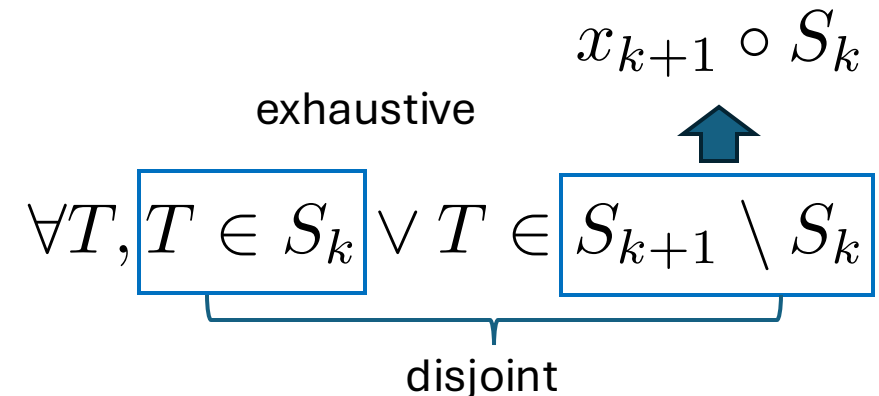
Optimal Subset Selection Problem

- Let S be a set of n item: x_1, x_2, \dots, x_n
- Consider any subset T from S , define
 - Eligibility: $\text{valid?}(T) = T/F$
 - Reward:
 - $R(T) \rightarrow \mathbb{R}_{\geq 0}$
 - Monotone: $R(T \cup \{x_j\}) = f(R(T), R(x_j))$, f is monotone
- Problem: find valid T^* with the minimum reward

Solution space



- Since we only need the optimal solution, can we partition the powerset of S into exhaustive and disjoint partitions, and only keep the best solution in each partition?
 - A natural way is the prefix partitioning
 - $S_k :=$ set of first k items of S



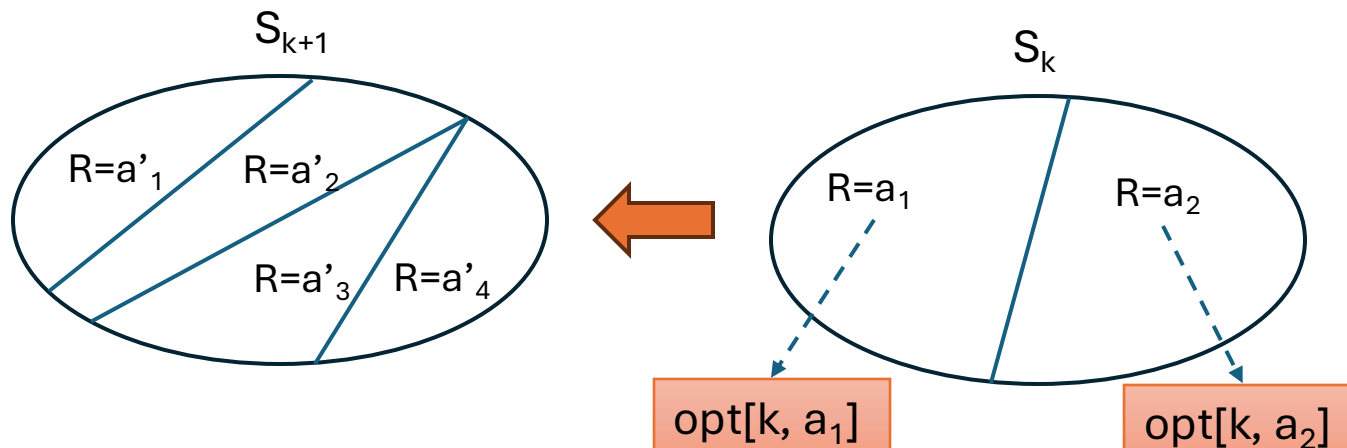
Recurrence

- Let T_k be the optimal subset in S_k , T_k^c be the optimal subset in $S_{k+1} \setminus S_k$
- Can we **grow** T_k to T_{k+1} ? \rightarrow Greedy
- Can we **grow** T_k or T_k^c to T_{k+1} \rightarrow Dynamic Programming

Exercise: Draw and Trace these ‘compressed solution’ spaces for a simple (knapsack) problem.

Knapsack Problem

- Try 1:
 - Grow T_k with x_{k+1}
 - Only choice: take x_{k+1} if the result is valid (i.e., not violating the weight constraint)
 - Observe that this may not be T_{k+1}
 - T_k cannot take x_{k+1} (weight limit)
 - $T_k^{(2)}$ can take x_{k+1} , and result in better solution
 - Pushing it further, all $\{\tau \mid \tau \in S_k, R(\tau) > R(T_k) - R(x_{k+1})\}$ need to be considered!



$T_k^{(2)}$: 2nd-best solution in S_k

If we do not have a bound on $R(x_{k+1})$, then $R(\tau)$ can be as small as 0

→ **Any** subset in S_k may be grown into T_{k+1} !!!

Nevertheless, among subsets with the same reward value, only need to “keep” one subset
→ (solution space) Compression achieved !

Exercise: Balanced Partition Problem

- (1) Do the similar reasoning on BP
- (2) Find a (smallest) example where T_{k+1} is not grown from T_k

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$$S = \{2, 10, 8, 5\}$$

$$T_1 = \{2\}$$

$$T_2 = \{2\}$$

$$T_3 = \{2, 8\}$$

$$T_3^{(2)} = \{2, 10\}$$



$$T_4 = \{2, 8\}$$

$$T_4 = \{2, 10\}$$

Knapsack Problem

- Try 2:

- What about $opt[W] = \min_{x \in S} \{opt[W - w(x)] + R(x)\}$

Knapsack Problem

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- What about $opt[W] = \min_{x \in S} \{opt[W - w(x)] + R(x)\}$

- Bug

- $opt[W]$ and $opt[W-w(x)]$ may both require using the same x
- But each x has only one supply!
- → Must track available items

- Fix: $opt[U, W] = \min_{x \in U} \{opt[U \setminus \{x\}, W - w(x)] + R(x)\}$

- Correct but useless

- Subproblems in U alone require enumerating ALL subsets
- → Use the compression idea

Always valid as long as opt solution contains at least one x

We consider $U \setminus \{x\}$ because we need to consider set of solutions that contains x

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 - Partition all subsets into S_k , k in $[n]$, based on the largest element it contains
 - The recurrence becomes: $opt[U, W] = \min_{i \in [n]} \{opt[S_i, W]\}$
- Overlapping substructure emerges!

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 - Partition all subsets into S_k , k in $[n]$, based on the largest element it contains
 - The recurrence becomes: $opt[U, W] = \min_{i \in [n]} \{opt[S_i, W]\}$
- Overlapping substructure emerges!

$$opt[S_{i+1}, W] = \min (opt[S_i, W], opt[S_i, W - w(x_{i+1})] + R(x_{i+1}))$$

Opt solution do not
uses x_{i+1}

Opt solution uses x_{i+1}