Dynamic Programming II

Summer 2018 • Lecture 07/31

A Few Notes

Homework 4

Due Thursday at 5 p.m. on Gradescope.

Tutorial 6

RSVP at https://goo.gl/forms/PLJr5gknNWFgccww2

Outline for Today

Dynamic Programming

More DP algorithms

Knapsack (0/1 and Unbounded)

Maximal Independent Set

Dynamic Programming

Elements of dynamic programming

Large problems break up into small problems.

e.g. shortest path with at most k edges.

Optimal substructure the optimal solution of a problem can be expressed in terms of optimal solutions of smaller sub-problems.

```
e.g. d^{(k)}[b] = min\{d^{(k-1)}[b], min_a\{d^{(k-1)}[a] + w(a,b)\}\}
```

Overlapping sub-problems the sub-problems overlap a lot.

e.g. Lots of different entries of $d^{(k)}$ ask for $d^{(k-1)}[a]$.

This means we're save time by solving a sub-problem once and caching the answer.

Dynamic Programming

Steps of dynamic programming

- (1) Identify optimal substructure with overlapping subproblems.
- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
- (4) If necessary, track additional information so that the algorithm from (3) can solve a related problem.

Dynamic Programming

Two approaches for DP: bottom-up and top-down.

Bottom-up iterates through problems by size and solves the small problems first (Bellman-Ford solves $d^{(0)}$ then $d^{(1)}$ then $d^{(2)}$, etc.)

Top-down recurses to solve smaller problems, which recurse to solve even smaller problems.

How is this different than divide and conquer? **Memoization**, which keeps track of the small problems you've already solved to prevent resolving the same problem more than once.

Which items should I cram inside my backpack?

We have n items with weights and values.

item:					
weight:	6	2	4	3	11
value:	20	8	14	13	35

Which items should I cram inside my backpack?

We have n items with weights and values.

item:











This is Koko

the koala.

weight:

value:

6

20

2

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Which items should I cram inside my backpack?

We have n items with weights and values.

item: weight: 6 2 4 3 11 value: 20 8 14 13 35

And we have a knapsack that can only carry so much weight.



capacity: 10

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Which items should I cram inside my backpack?

We have n items with weights and values.

item:









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And we have a knapsack that can only carry so much weight.

This is called a "knapsack"; thank goodness since "Knapsack algorithm" sounds cooler than "Backpack algorithm."



capacity: 10











11

35

Unbounded Knapsack

Suppose I have infinite copies of all items.

weight 6 2 4 3 value 20 8 14 13

What's the most valuable way to fill the knapsack?









Total weight: 10

Total value: 42



capacity: 10

0/1 Knapsack

Suppose I only have one copy of each item.

What's the most valuable way to fill the knapsack?







Total weight: 9

Total value: 35

Some notation

item:

weight:

value:

 W_1

 V_{1}



 W_2

٧₂



 W_3

V₃

This is Koko the koala.



 \mathbf{W}_{n}

 V_{n}



capacity: W

Task Find the items to put in an unbounded knapsack.

Steps of dynamic programming

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Task Find the items to put in an unbounded knapsack.

(1) Identify optimal substructure with overlapping subproblems.

The problem statement restricts us from reducing the number of items.

By process of elimination, we reason that we must solve the problem for smaller knapsacks.



First solve the problem for small knapsacks

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Then larger knapsacks

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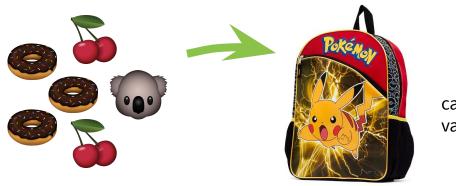
Then larger knapsacks



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Task Find the items to put in an unbounded knapsack.

(1) Identify optimal substructure with overlapping subproblems. If this is an optimal solution for capacity x



capacity x value V

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capacity x value V

Then this must be an optimal solution for capacity $x - w_i$ for item $i = w_i$



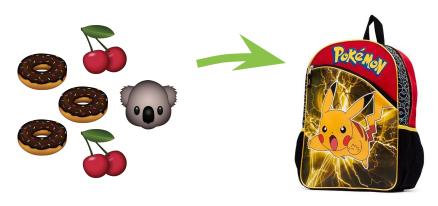




capacity x - w. value V - v.

Task Find the items to put in an unbounded knapsack.

(1) Identify optimal substructure with overlapping subproblems. If this is an optimal solution for capacity x



capacity x value V

Then this must be an optimal solution for capacity x - w, for item i =







capacity x - w, value V - v.

If there existed a more optimal solution, then adding a donut to that more optimal solution would improve the first solution.

Task Find the items to put in an unbounded knapsack.

Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.



- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
- (4) If necessary, track additional information so that the algorithm from (3) can solve a related problem.

Task Find the items to put in an unbounded knapsack.

(2) Define a recursive formulation.

Let **K[x]** be the optimal value for capacity x.

$$K[x] = max_i \{ + \}$$

The maximum over all i such that $w_i \le x$.

The optimal way to fill the smaller knapsack

The value of item i.

$$K[x] = \begin{cases} 0 & \text{if there are no i where } w_i \leq x \\ \max_i \{K[x-w_i] + v_i\} & \text{otherwise} \end{cases}$$

Task Find the items to put in an unbounded knapsack.

Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.

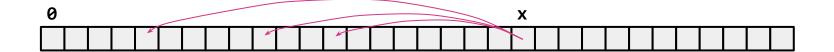


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Task Find the items to put in an unbounded knapsack.

(3) Use dynamic programming to solve the problem.

In what order do we need to fill our table according to the formulation from (2)?



$$K[x] = \begin{cases} 0 & \text{if there are no i where } w_i \le x \\ \max_i \{K[x-w_i] + v_i\} & \text{otherwise} \end{cases}$$

Task Find the items to put in an unbounded knapsack.

(3) Use dynamic programming to solve the problem.

In what order do we need to fill our table according to the formulation from (2)?



An element at position x in the table depends on elements at positions $x - w_i$ for all i. So we want to fill out the values at these positions before x.

$$K[x] = \begin{cases} 0 & \text{if there are no i where } w_i \le x \\ \max_i \{K[x-w_i] + v_i\} & \text{otherwise} \end{cases}$$

```
def unbounded_knapsack(capacity, weights, values):
    W = capacity
    n = weights.length
    K[0] = 0
    for x = 1 to W:
        K[x] = 0
    for i = 0 to n-1:
        W<sub>i</sub> = weights[i], v<sub>i</sub> = values[i]
        if w<sub>i</sub> ≤ x:
        K[x] = max{K[x], K[x-w<sub>i</sub>] + v<sub>i</sub>}
    return K[W]
```

Runtime: O(nW)

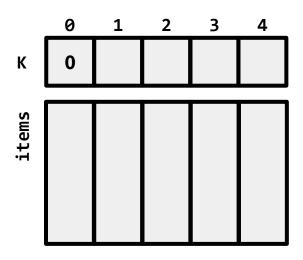
Task Find the items to put in an unbounded knapsack.

Steps of dynamic programming

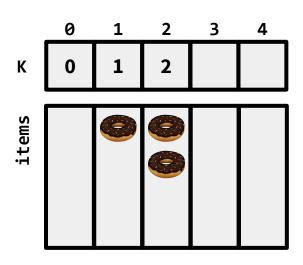
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```
def unbounded knapsack(capacity, weights, values):
  W = capacity
  n = weights.length
  K[0] = 0, items[0] = {}
  for x = 1 to W:
    K[x] = 0
    for i = 0 to n-1:
      w<sub>i</sub> = weights[i], v<sub>i</sub> = values[i]
      if W_i \leq x:
         K[x] = max\{K[x], K[x-w] + v\}
         if K[x] updated:
           items[x] = items[x-w_i] \cup \{i\}
  return items[W]
```

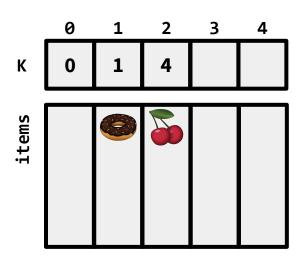
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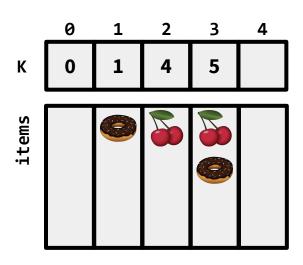




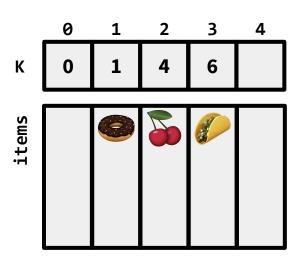




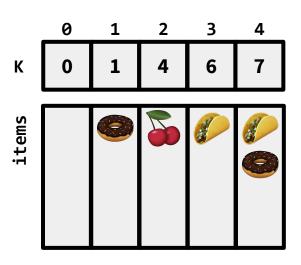




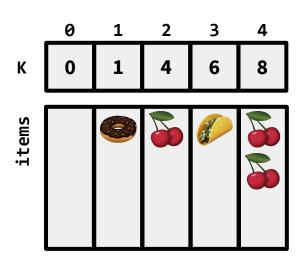














Unbounded Knapsack 👌













3

13



Suppose I have infinite copies of all items.

value

20

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11

35

What's the most valuable way to fill the knapsack?









Total weight: 10

Total value: 42



capacity: 10

0/1 Knapsack

Suppose I only have one copy of each item.

What's the most valuable way to fill the knapsack?







Total weight: 9

Total value: 35

Task Find the items to put in a 0/1 knapsack.

Steps of dynamic programming

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Task Find the items to put in a 0/1 knapsack.

(1) Identify optimal substructure with overlapping subproblems.

Can we use the same optimal substructure as unbounded knapsack?



First solve the problem for small knapsacks



Then larger knapsacks



Then larger knapsacks

Task Find the items to put in a 0/1 knapsack.

(1) Identify optimal substructure with overlapping subproblems.

Can we use the same optimal substructure as unbounded knapsack?

No, the subproblem needs information about which items have been used.





Task Find the items to put in a 0/1 knapsack.

(1) Identify optimal substructure with overlapping subproblems.

Can we use the same optimal substructure as unbounded knapsack?

No, the subproblem needs information about which items have been used.



Task Find the items to put in a 0/1 knapsack.

(1) Identify optimal substructure with overlapping subproblems.

We reason that we must solve the problem for a smaller number of items and

for smaller knapsacks.



First solve the problem for small knapsacks



Then larger knapsacks



Then larger knapsacks

First solve the problem for few items



Task Find the items to put in a 0/1 knapsack.

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First solve the problem for small knapsacks



Then larger knapsacks



Then larger knapsacks

First solve the problem for few items

Then more items



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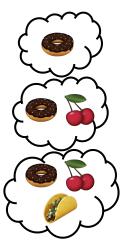


Then larger knapsacks

First solve the problem for few items

Then more items

Then more items



Task Find the items to put in a 0/1 knapsack.

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First solve the problem for small knapsacks



Then larger knapsacks



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We need a two-dimensional table!

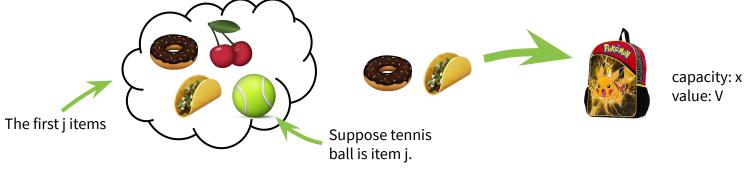
Then more items

Task Find the items to put in a 0/1 knapsack.

(1) Identify optimal substructure with overlapping subproblems.

Handle items in a similar way to how we handled vertices in Floyd-Warshall; restrict the set of items to be used to a specific set 0 to j-1.

Case 1 If the optimal solution for j items does not use item j.



Then this is an optimal solution for j - 1 items.

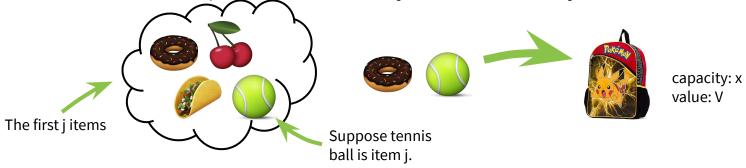


Task Find the items to put in a 0/1 knapsack.

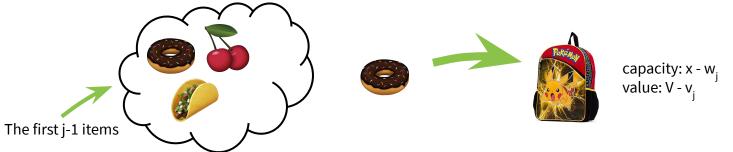
(1) Identify optimal substructure with overlapping subproblems.

Handle items in a similar way to how we handled vertices in Floyd-Warshall; restrict the set of items to be used to a specific set 0 to j-1.

Case 2 If the optimal solution for j items uses item j.



Then this is an optimal solution for j - 1 items.



Task Find the items to put in a 0/1 knapsack.

Steps of dynamic programming

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- (2) Define a recursive formulation.
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Task Find the items to put in an unbounded knapsack.

(2) Define a recursive formulation.

Let **K**[x,j] be the optimal value for capacity x with j items.

$$K[x,j] = \begin{cases} 0 & \text{if } x \text{ or } j \text{ are } 0 \\ \max\{K[x,j-1], K[x-w_j,j-1] + v_j\} & \text{otherwise} \end{cases}$$

Task Find the items to put in a 0/1 knapsack.

Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.



- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
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```
def zero_one_knapsack(capacity, weights, values):
  W = capacity
  n = weights.length
  K[x,0] = 0 for x = 0 to W
  K[0,i] = 0 for i = 0 to n
  for x = 1 to W:
    for j = 1 to n:
      K[x,j] = K[x,j-1]
      w<sub>i</sub> = weights[j], v<sub>i</sub> = values[j]
      if W_i \leq x:
         K[x,j] = max\{K[x,j], K[x-w_i] + v_i\}
  return K[W,n]
```

Runtime: O(nW)

Task Find the items to put in a 0/1 knapsack.

Steps of dynamic programming

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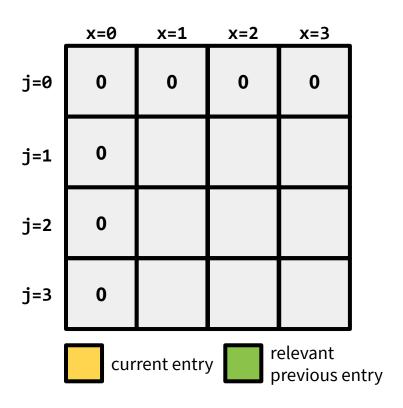
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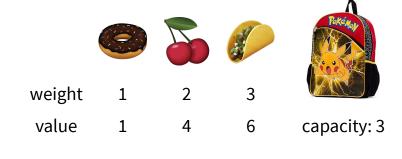


(3) Use dynamic programming to solve the problem.

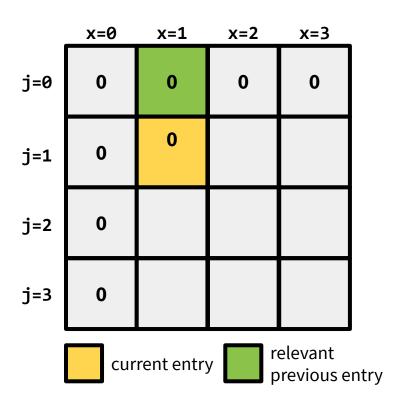


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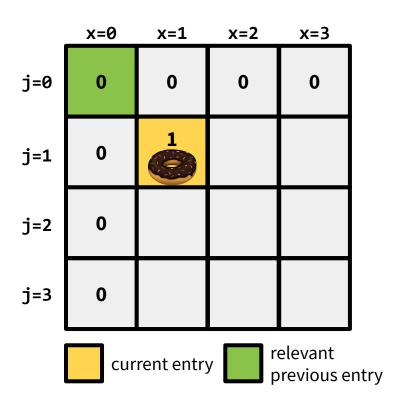


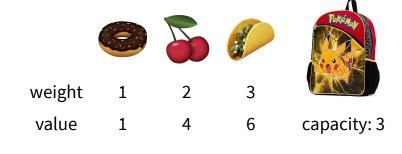
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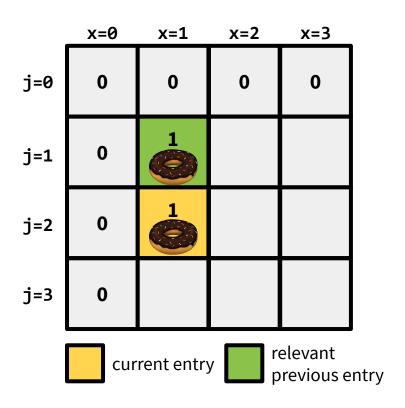


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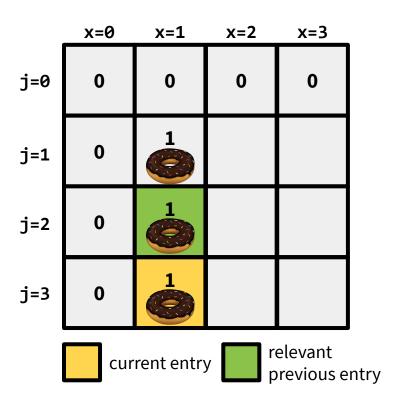


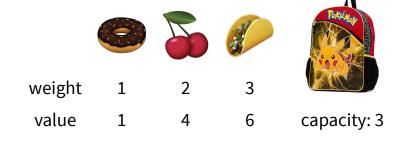
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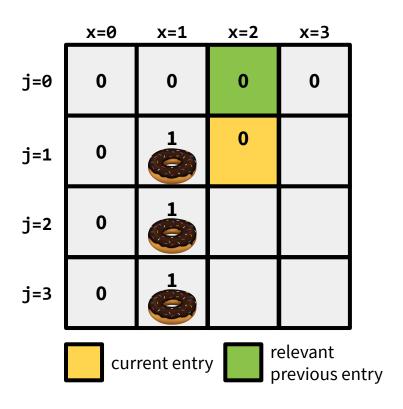


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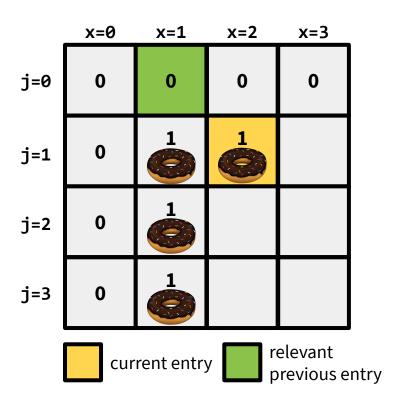


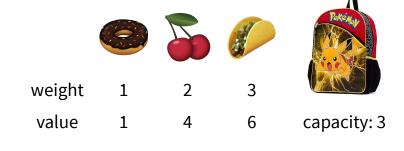
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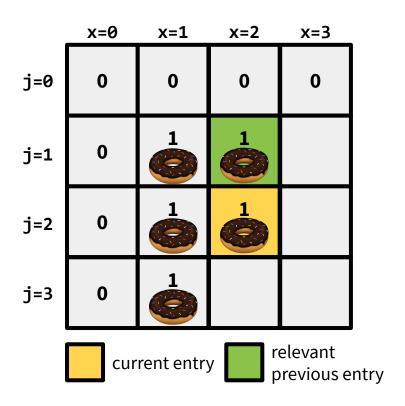


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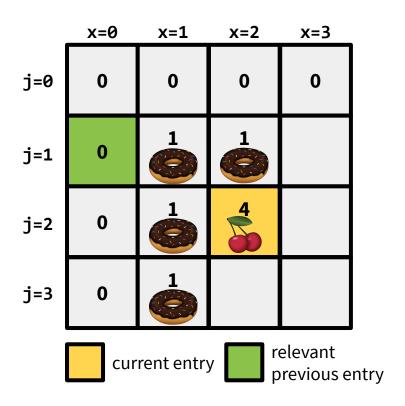


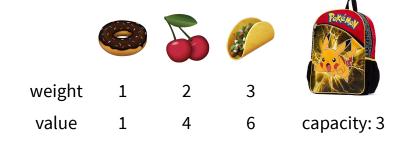
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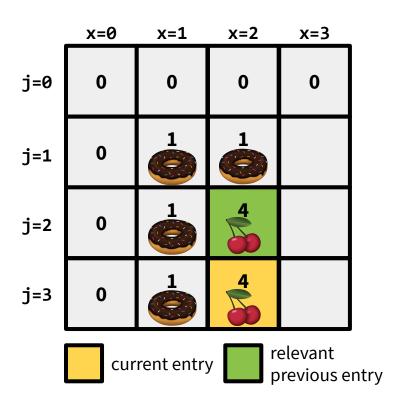


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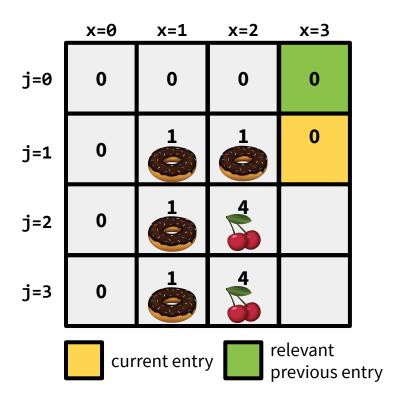


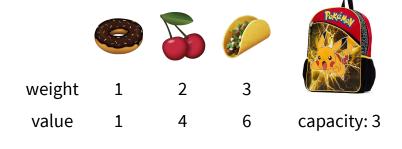
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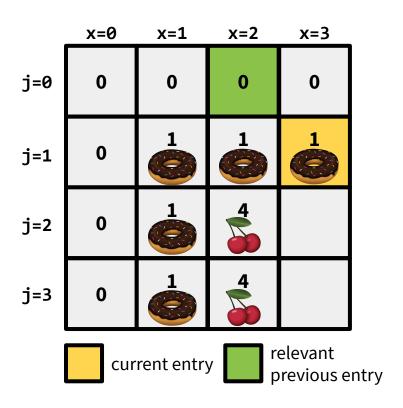


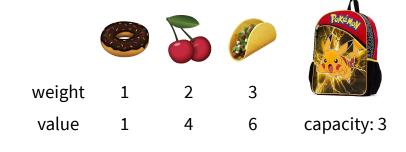
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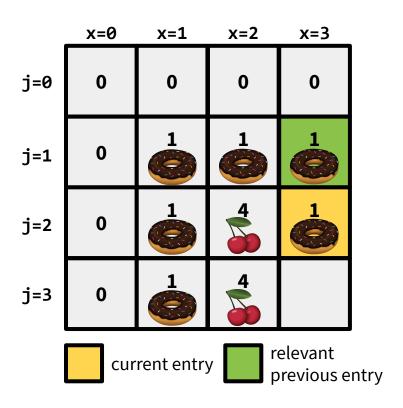


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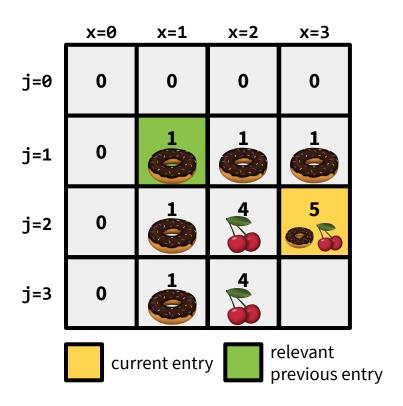


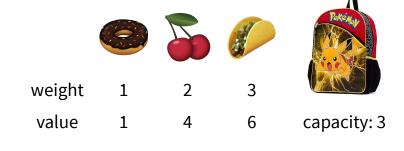
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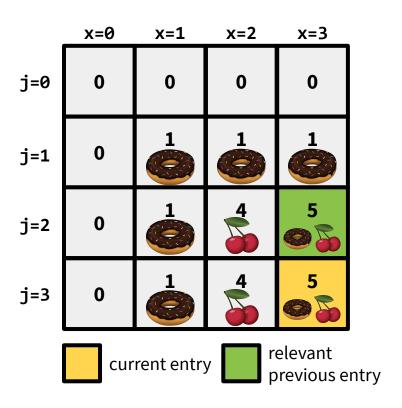


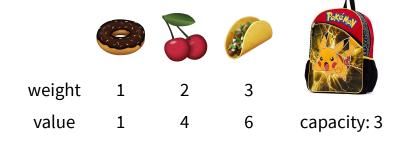
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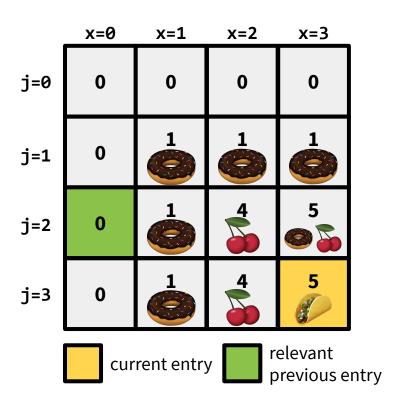


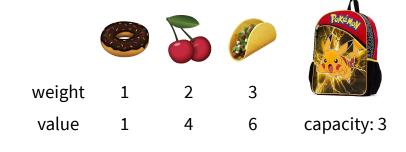
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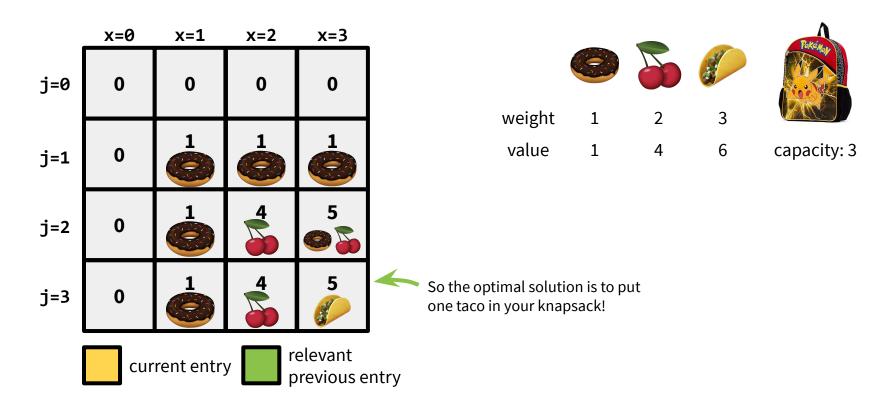


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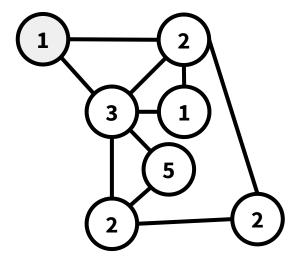
Independent Set

Maximum Independent Set

What is the maximum independent set in a graph?

An independent set describes a set of weighted vertices where no pair of vertices in the set shares an edge.

A maximum independent set has the largest weight.

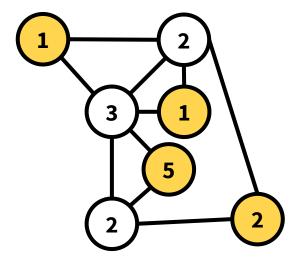


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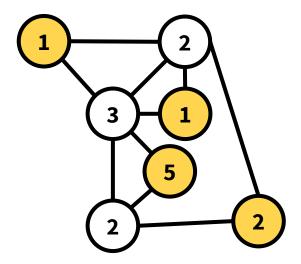


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This problem is NP-complete.

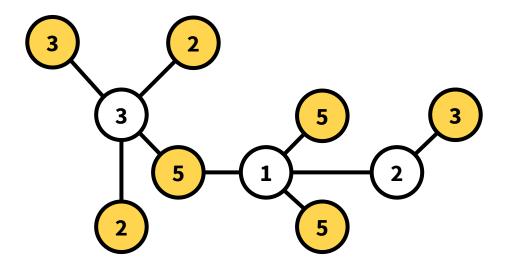


We'll learn what this means in a bit, but for now take it to mean we're unlikely to find an efficient algorithm.

What is the maximum independent set in a tree?

An independent set describes a set of weighted vertices where no pair of vertices in the set shares an edge.

A maximum independent set has the largest weight.



Task Find the maximum independent set in a tree.

Steps of dynamic programming

- (1) Identify optimal substructure with overlapping subproblems.
- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
- (4) If necessary, track additional information so that the algorithm from (3) can solve a related problem.

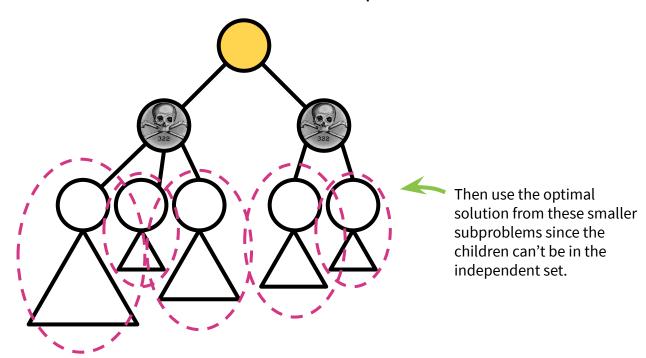
Task Find the maximum independent set in a tree.

(1) Identify optimal substructure with overlapping subproblems.

Subtree are a natural candidate.

Consider two cases:

Case 1 The root of this tree is in a maximum independent set.



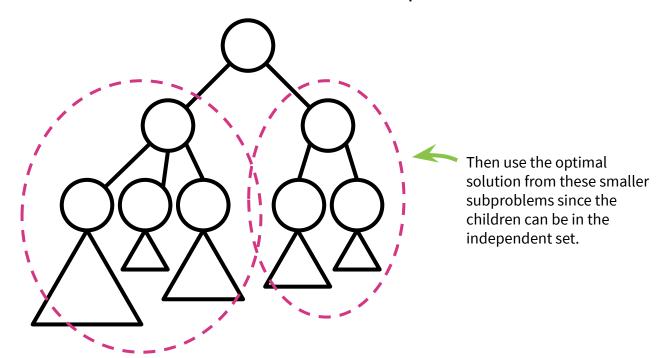
Task Find the maximum independent set in a tree.

(1) Identify optimal substructure with overlapping subproblems.

Subtree are a natural candidate.

Consider two cases:

Case 2 The root of this tree is not in a maximum independent set.



Task Find the maximum independent set in a tree.

Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.



- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
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Task Find the maximum independent set in a tree.

(2) Define a recursive formulation.

Let A[u] be the weight of a maximum independent set in the tree rooted at u.

$$A[u] = \max \left\{ \begin{array}{l} w(u) + \sum_{v \in u. \text{grandchildren}} & \text{case 1} \\ A[v] \\ \sum_{v \in u. \text{children}} A[v] & \text{case 2} \end{array} \right\}$$

Task Find the maximum independent set in a tree.

Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.

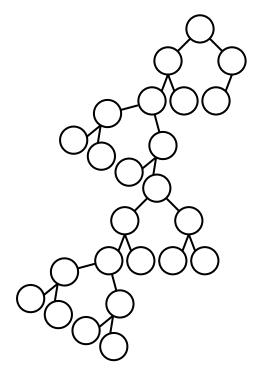


- (2) Define a recursive formulation.
- (3) Use dynamic programming to solve the problem.
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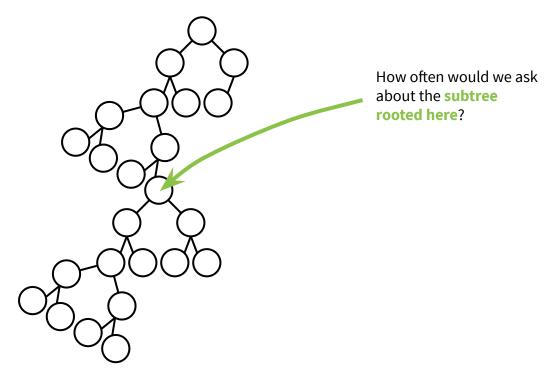
```
cache = {}
def max independent set helper(root):
 if is_leaf(root):
  return root.weight
cache[root] = w
  return w
```

Runtime: O(|V|)

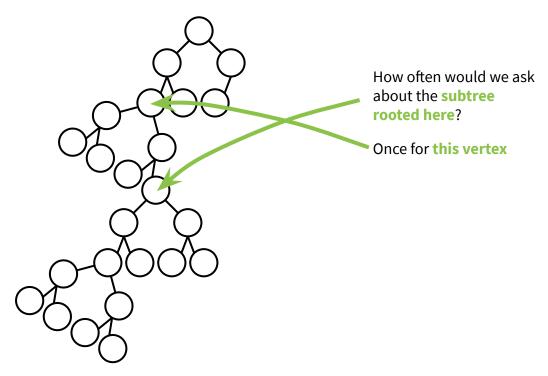
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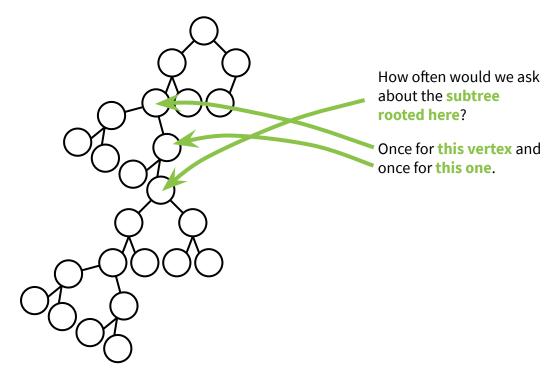
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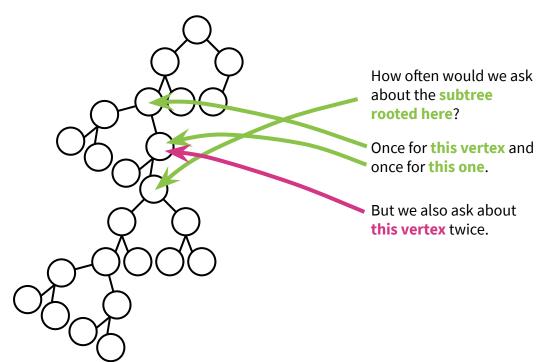
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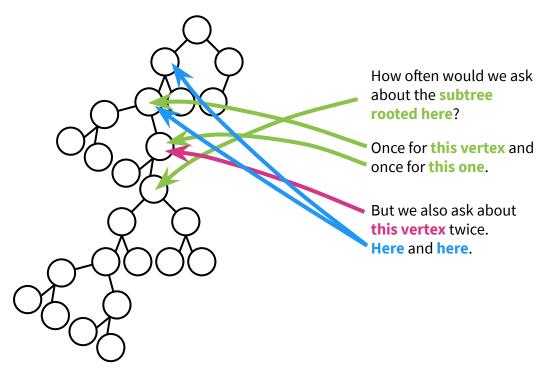
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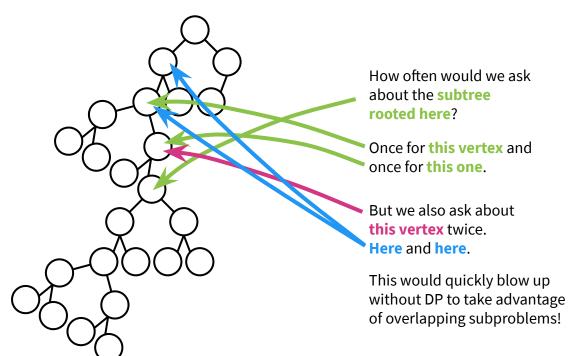
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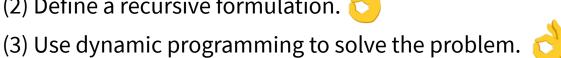
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Steps of dynamic programming

(1) Identify optimal substructure with overlapping subproblems.



(2) Define a recursive formulation.



(4) If necessary, track additional information so that the algorithm from (3) can solve a related problem.

Conclusion

	Runtime	Variables	Notes
LCS	O(X Y) worst-case	X and Y are the lengths of the strings being compared.	Finding the actual LCS is possible from first solving the length of the LCS.
Unbounded Knapsack	O(nW) worst-case	n is the number of items and W is the knapsack capacity.	Ditto
0/1 Knapsack	O(nW) worst-case	Ditto	Same as unbounded knapsack except can only use 1 of each item, so it requires a 2D table.
Maximum Independent Set	O(V) worst-case	V is the number of vertices in the tree.	NP-complete for graphs, this alg. works for trees; top-down easier.