#### **Dynamic Programming vs. Divide-&-conquer**

- Divide-&-conquer works best when all subproblems are independent. So, pick the partition that makes the algorithm most efficient. Then simply combine their solutions to solve the entire problem.
- Dynamic programming is needed when subproblems are <u>dependent</u> and we don't know where to partition the problem. For example, let  $S_1$ = {ALPHABET}, and  $S_2$  = {HABITAT}. Consider the subproblem with  $S_1'$  = {ALPH},  $S_2'$  = {HABI}.

Then, <u>LCS  $(S_{1'}, S_{2'}) + LCS (S_{1}-S_{1'}, S_{2}-S_{2'}) \neq LCS(S_{1}, S_{2})$ </u>

- Divide-&-conquer is best suited for the case when no "overlapping subproblems" are encountered.
- In dynamic programming algorithms, we typically solve each subproblem only once and store their solutions. But this is at the cost of space.

# **Dynamic programming vs Greedy**

 Dynamic Programming solves the sub-problems bottom up. The problem can't be solved until we find all solutions of sub-problems. The solution comes up when the whole problem appears.

Greedy solves the sub-problems from top down. We first need to find the greedy choice for a problem, then reduce the problem to a smaller one. The solution is obtained when the whole problem disappears.

2. Dynamic Programming has to try every possibility before solving the problem. It is much more expensive than greedy. However, there are some problems that greedy can not solve while dynamic programming can. Therefore, we first try greedy algorithm. If it fails then try dynamic programming.

### Fractional Knapsack Problem

- Burglar's choices: Items: x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>n</sub> Value: v<sub>1</sub>, v<sub>2</sub>, ..., v<sub>n</sub> Max Quantity: q<sub>1</sub>, q<sub>2</sub>, ..., q<sub>n</sub> Weight per unit quantity: w<sub>1</sub>, w<sub>2</sub>, ..., w<sub>n</sub> Getaway Truck has a weight limit of B. Burglar can take "fractional" amount of any item. How can burglar maximize value of the loot?
  Greedy Algorithm works!
- Greedy Algorithm works!
   Pick the maximum possible quantity of highest value per weight item. Continue until weight limit of truck is reached.

## 0-1 Knapsack Problem

- Burglar's choices: Items: x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>n</sub> Value: v<sub>1</sub>, v<sub>2</sub>, ..., v<sub>n</sub> Weight: w<sub>1</sub>, w<sub>2</sub>, ..., w<sub>n</sub> Getaway Truck has a weight limit of B. Burglar cannot take "fractional" amount of item. How can burglar maximize value of the loot?
- Greedy Algorithm does not work! Why?
- Need dynamic programming!

## 0-1 Knapsack Problem

- Subproblems?
  - V[j, L] = <u>Optimal</u> solution for knapsack problem assuming a truck of weight limit L and choice of items from set {1,2,..., j}.
  - V[n, B] = <u>Optimal</u> solution for original problem
  - V[1, L] = easy to compute for all values of L.
- Table of solutions?
  - V[1..n, 1..B]
- Ordering of subproblems?
  - Row-wise
- Recurrence Relation? [Either ×<sub>i</sub> included or not]
  - V[j, L] = max { V[j-1, L],

 $v_{j} + V[j-1, L-w_{j}]$