

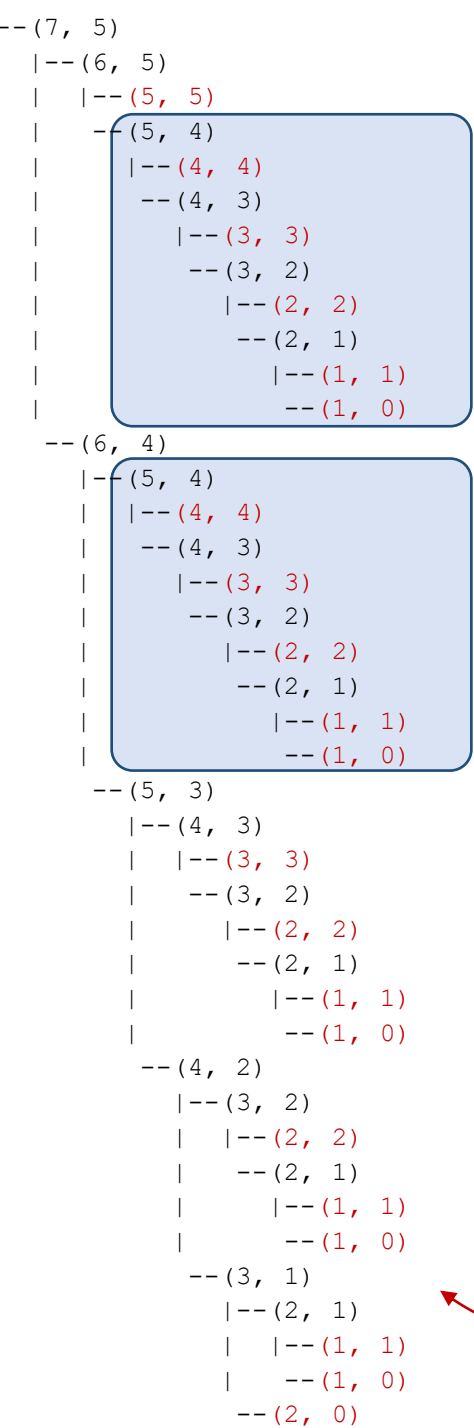
# Dynamic programming

- memoization
- decorator memoized / `functools.cache`
- systematic subproblem computation

# Binomial coefficient

$$\binom{n}{k} = \begin{cases} 1 & \text{if } k = 0 \text{ or } k = n \\ \binom{n-1}{k} + \binom{n-1}{k-1} & \text{otherwise} \end{cases}$$

identical computations



**binomial\_recursive.py**

```
def binomial(n, k):
    if k == 0 or k == n:
        return 1
    return binomial(n - 1, k) + binomial(n - 1, k - 1)
```

recursion tree for binomial(7, 5)

# Dynamic Programming



Remember solutions already found  
(memoization)

- Technique sometimes applicable when running time otherwise becomes exponential
- Only applicable if stuff to be remembered is manageable

# Binomial Coefficient

## Dynamic programming using a dictionary

recursion tree for  
`binomial(7, 5)`

```
-- (7, 5)
| -- (6, 5)
| | -- (5, 5)
| | -- (5, 4)
| | | -- (4, 4)
| | | -- (4, 3)
| | | | -- (3, 3)
| | | | -- (3, 2)
| | | | | -- (2, 2)
| | | | | -- (2, 1)
| | | | | -- (1, 1)
| | | | | -- (1, 0)
| | -- (6, 4)
| | | -- (5, 4)
| | | -- (5, 3)
| | | | -- (4, 3)
| | | | -- (4, 2)
| | | | | -- (3, 2)
| | | | | -- (3, 1)
| | | | | -- (2, 1)
| | | | | -- (2, 0)
```

`binomial_dictionary.py`

```
answers = {} # answers[(n, k)] = binomial(n, k)

def binomial(n, k):
    if (n, k) not in answers:
        if k == 0 or k == n:
            answer = 1
        else:
            answer = binomial(n - 1, k) + binomial(n - 1, k - 1)
        answers[(n, k)] = answer
    return answers[(n, k)]
```

Python shell

```
> binomial(6, 3)
| 20
> answers
| {(3, 3): 1, (2, 2): 1, (1, 1): 1, (1, 0): 1, (2, 1): 2, (3, 2): 3, (4, 3): 4, (2, 0): 1, (3, 1): 3, (4, 2): 6, (5, 3): 10, (3, 0): 1, (4, 1): 4, (5, 2): 10, (6, 3): 20}
```

- Use a dictionary `answers` to store already computed values

reuse value stored in dictionary `answers`


Question – What is the order of the size of the dictionary answers after calling `binomial(n, k)` ?

```
binomial_dictionary.py
```

```
answers = {} # answers[(n, k)] = binomial(n, k)
def binomial(n, k):
    if (n, k) not in answers:
        if k == 0 or k == n:
            answer = 1
        else:
            answer = binomial(n - 1, k) + binomial(n - 1, k - 1)
        answers[(n, k)] = answer
    return answers[(n, k)]
```

a)  $\max(n, k)$

b)  $n + k$

 c)  $n * k$

d)  $n^k$

e)  $k^n$

f) Don't know

# Binomial Coefficient

## Dynamic programming using decorator

- Use a decorator (@memoize) that implements the functionality of remembering the results of previous function calls

binomial\_decorator.py

```
def memoize(f):  
    # answers[args] = f(*args)  
    answers = {}  
  
    def wrapper(*args):  
        if args not in answers:  
            answers[args] = f(*args)  
        return answers[args]  
  
    return wrapper
```

```
@memoize  
def binomial(n, k):  
    if k == 0 or k == n:  
        return 1  
    else:  
        return binomial(n - 1, k) + binomial(n - 1, k - 1)
```

## binomial\_decorator\_trace.py

```

def trace(f): # decorator to trace recursive calls
    indent = 0

    def wrapper(*args):
        nonlocal indent

        spaces = '| ' * indent
        arg_str = ', '.join(map(repr, args))
        print(spaces + f'{f.__name__}({arg_str})')
        indent += 1
        result = f(*args)
        indent -= 1
        print(spaces + f'> {result}')
        return result

    return wrapper

def memoize(f):
    answers = {}

    def wrapper(*args):
        if args not in answers:
            answers[args] = f(*args)
        return answers[args]

    wrapper.__name__ = f.__name__ + '_memoize'

    return wrapper

@trace
@memoize
def binomial(n, k):
    if k == 0 or k == n:
        return 1
    return binomial(n - 1, k) + binomial(n-1, k-1)

print(binomial(5, 2))

```

## Python shell (without @memoize)

```

binomial(5, 2)
| binomial(4, 2)
| | binomial(3, 2)
| | | binomial(2, 2)
| | | > 1
| | | binomial(2, 1)
| | | | binomial(1, 1)
| | | | > 1
| | | | binomial(1, 0)
| | | | > 1
| | | > 2
| | > 3
| | binomial(3, 1)
| | | binomial(2, 1)
| | | | binomial(1, 1)
| | | | > 1
| | | | binomial(1, 0)
| | | | > 1
| | | > 2
| | > 3
| | binomial(2, 0)
| | > 1
| > 3
> 6
binomial(4, 1)
| binomial(3, 1)
| | binomial(2, 1)
| | | binomial(1, 1)
| | | > 1
| | | binomial(1, 0)
| | | > 1
| | > 2
| | binomial(2, 0)
| | > 1
| > 3
| binomial(3, 0)
| > 1
| > 4
> 10
10

```

## Python shell (with @memoize)

```

binomial_memoize(5, 2)
| binomial_memoize(4, 2)
| | binomial_memoize(3, 2)
| | | binomial_memoize(2, 2)
| | | > 1
| | | binomial_memoize(2, 1)
| | | | binomial_memoize(1, 1)
| | | | > 1
| | | | binomial_memoize(1, 0)
| | | | > 1
| | | > 2
| | > 3
| | binomial_memoize(3, 1)
| | | binomial_memoize(2, 1)
| | | > 2
| | | binomial_memoize(2, 0)
| | | > 1
| | > 3
| > 6
| binomial_memoize(4, 1)
| | binomial_memoize(3, 1)
| | > 3
| | binomial_memoize(3, 0)
| | > 1
| > 4
> 10
10

```

without assigning `wrapper.__name__`  
the name shown would be `wrapper`

saved recursive calls  
when using memoization

# Dynamic programming using cache decorator

```
bionomial_cache.py
```

```
from functools import cache

@cache
def binomial(n, k):
    if k == 0 or k == n:
        return 1
    else:
        return binomial(n - 1, k) + binomial(n - 1, k - 1)
```

- The decorators `@cache` (since Python 3.9) and `@lru_cache(maxsize=None)` in the standard library `functools` supports the same as the decorator `@memoize`
- By default `@lru_cache` at most remembers (caches) 128 previous function calls, always evicting **Least Recently Used** entries from its dictionary



# Subset sum using dynamic programming

- In the subset sum problem (Exercise 13.4) we are given a number  $x$  and a list of numbers  $\mathbb{L}$ , and want to determine if a subset of  $\mathbb{L}$  has sum  $x$

$$\mathbb{L} = [3, 7, 2, 11, 13, 4, 8] \quad x = 22 = 7 + 11 + 4$$

- Let  $S(v, k)$  denote if it is possible to **achieve value  $v$  with a subset of  $\mathbb{L}[:k]$** , i.e.  $S(v, k) = \text{True}$  if and only if a subset of the first  $k$  values in  $\mathbb{L}$  has sum  $v$
- $S(v, k)$  can be computed from the recurrence

$$S(v, k) = \begin{cases} \text{True} & \text{if } k = 0 \text{ and } v = 0 \\ \text{False} & \text{if } k = 0 \text{ and } v \neq 0 \\ S(v, k-1) \text{ or } S(v - \mathbb{L}[k-1], k-1) & \text{otherwise} \end{cases}$$

# Subset sum using dynamic programming

```
subset_sum_dp.py
```

```
def subset_sum(x, L):  
    @memoize  
    def solve(value, k):  
        if k == 0:  
            return value == 0  
        return solve(value, k - 1) or solve(value - L[k - 1], k - 1)  
    return solve(x, len(L))
```

```
Python shell
```

```
> subset_sum(11, [2, 3, 8, 11, -1])  
| True  
> subset_sum(6, [2, 3, 8, 11, -1])  
| False
```

Question – What is a bound on the size order of the memoization table if all values are positive integers?

subset\_sum\_dp.py

```
def subset_sum(x, L):  
    @memoize  
    def solve(value, k):  
        if k == 0:  
            return value == 0  
        return solve(value, k-1) or solve(value - L[k-1], k-1)  
    return solve(x, len(L))
```


Python shell

```
> subset_sum(11, [2, 3, 8, 11, -1])  
| True  
> subset_sum(6, [2, 3, 8, 11, -1])  
| False
```


a)  $\text{len}(L)$

b)  $\text{sum}(L)$

c)  $x$

 d)  $2^{\text{len}(L)}$

e)  $\text{len}(L)$

 f)  $\text{len}(L) * \text{sum}(L)$

g) Don't know

# Subset sum using dynamic programming

```
subset_sum_dp.py
```

```
def subset_sum_solution(x, L):  
    @memoize  
    def solve(value, k):  
        if k == 0:  
            if value == 0:  
                return []  
            else:  
                return None  
        solution = solve(value, k - 1)  
        if solution != None:  
            return solution  
        solution = solve(value - L[k - 1], k - 1)  
        if solution != None:  
            return solution + [L[k - 1]]  
        return None  
    return solve(x, len(L))
```

```
Python shell
```

```
> subset_sum_solution(11, [2, 3, 8, 11, -1])  
| [3, 8]  
> subset_sum_solution(6, [2, 3, 8, 11, -1])  
| None
```

# Knapsack problem

	Volume	Value
0	3	6
1	3	7
2	2	8
3	5	9

- Given a **knapsack** with volume **capacity C**, and set of **objects** with different **volumes and value**
- **Objective:** Find a subset of the objects that fits in the knapsack (sum of volume  $\leq$  capacity) and has maximal value
- **Example:** If  $C = 5$  and the volume and weights are given by the table, then the maximal value 15 can be achieved by the 2nd and 3rd object
- Let  $V(c, k)$  denote the **maximum value achievable by a subset of the first k objects within capacity c**

$$V(c, k) = \begin{cases} 0 & \text{if } k = 0 \\ V(c, k - 1) & \text{volume}[k-1] > c \\ \max\{V(c, k - 1), \text{value}[k - 1] + V(c - \text{volume}[k - 1], k - 1)\} & \text{otherwise} \end{cases}$$

# Knapsack – maximum value

`knapsack.py`

```
def knapsack_value(volume, value, capacity):  
    @memoize  
    def solve(c, k): # solve with capacity c and objects 0..k-1  
        if k == 0: # no objects to put in knapsack  
            return 0  
        v = solve(c, k - 1) # try without object k-1  
        if volume[k - 1] <= c: # try also with object k-1 if space  
            v = max(v, value[k - 1] + solve(c - volume[k - 1], k - 1))  
        return v  
    return solve(capacity, len(volume))
```

Python shell

```
> volumes = [3, 3, 2, 5]  
> values = [6, 7, 8, 9]  
> knapsack_value(volumes, values, 5)  
| 15
```

# Knapsack – maximum value and objects

knapsack.py

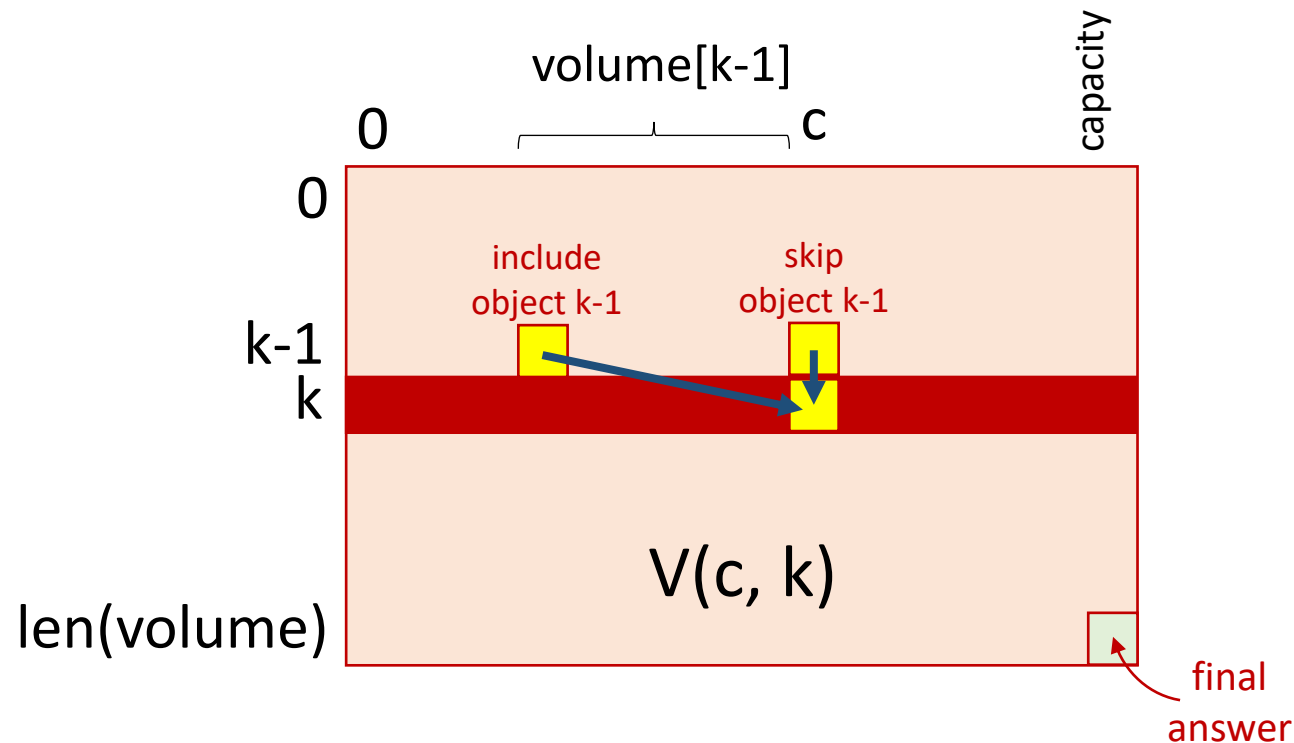
```
def knapsack(volume, value, capacity):  
    @memoize  
    def solve(c, k): # solve with capacity c and objects 0..k-1  
        if k == 0: # no objects to put in knapsack  
            return 0, []  
        v, solution = solve(c, k - 1) # try without object k-1  
        if volume[k - 1] <= c: # try also with object k-1 if space  
            v2, sol2 = solve(c - volume[k - 1], k - 1)  
            v2 = v2 + value[k - 1]  
            if v2 > v:  
                v = v2  
                solution = sol2 + [k - 1]  
        return v, solution  
    return solve(capacity, len(volume))
```

Python shell

```
> volumes = [3, 3, 2, 5]  
> values = [6, 7, 8, 9]  
> knapsack(volumes, values, 5)  
| (15, [1, 2])
```

# Knapsack - Table

$$V(c, k) = \begin{cases} 0 & \text{if } k = 0 \\ V(c, k - 1) & \text{value}[k-1] > c \\ \max\{V(c, k - 1), \text{value}[k - 1] + V(c - \text{volume}[k - 1], k - 1)\} & \text{otherwise} \end{cases}$$



- systematic fill out table
- only need to remember two rows



# Knapsack – Systematic table fill out

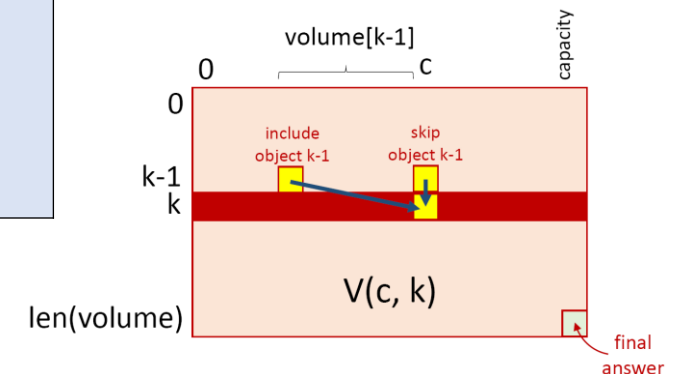
## knapsack\_systematic.py

```
def knapsack(volume, value, capacity):  
    ① solutions = [(0, [])] * (capacity + 1)  
    ② for obj in range(len(volume)): ⑤  
        for c in reversed(range(volume[obj], capacity + 1)): ④  
            prev_v, prev_solution = solutions[c - volume[obj]]  
            v = value[obj] + prev_v  
            if solutions[c][0] < v:  
                ③ solutions[c] = v, prev_solution + [obj]  
  
    return solutions[capacity]
```

## Python shell

```
> volumes = [3, 3, 2, 5]  
> values = [6, 7, 8, 9]  
> knapsack(volumes, values, 5)  
| (15, [1, 2])
```

- ① base case  $k = 0$
- ② consider each object
- ③ solutions[c:] current row  
solutions[:c] previous row
- ④ compute next row right-to-left
- ⑤ solutions[:volume[obj]]  
unchanged from previous row



# Summary

- Dynamic programming is a general approach for recursive problems where one tries to avoid recomputing the same expressions repeatedly
- **Solution 1: Memoization**
  - add dictionary to function to remember previous results
  - decorate with a `@memoize` decorator
- **Solution 2: Systematic table fill out**
  - can need to compute more values than when using memoization
  - can discard results not needed any longer (reduced memory usage)

# Coding competitions and online judges

If you like to practice your coding skills, there are many online “judges” with numerous exercises and where you can upload and test your solutions.

- [Project Euler](#)
- [Kattis](#)
- [CodeForces](#)
- [Topcoder](#)