

Dynamic programming

- memoization
- decorator memoized
- systematic subproblem computation

```

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

```

Binomial coefficient

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

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

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 Coefficient

Dynamic programming using a dictionary

`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
| | > 3
| | binomial(2, 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)
| | 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 lru_cache decorator

bionomial_lru_cache.py

```
from functools import lru_cache

@lru_cache(maxsize=None)
def binomial(n, k):
    if k==0 or k==n:
        return 1
    else:
        return binomial(n-1, k) + binomial(n-1, k-1)
```

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

Subset sum using dynamic programming

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

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

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

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

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

- 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 ($\text{sum of volume} \leq \text{capacity}$) 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**.

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

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

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)
```

Knapsack – maximum value and objects

knapsack.py

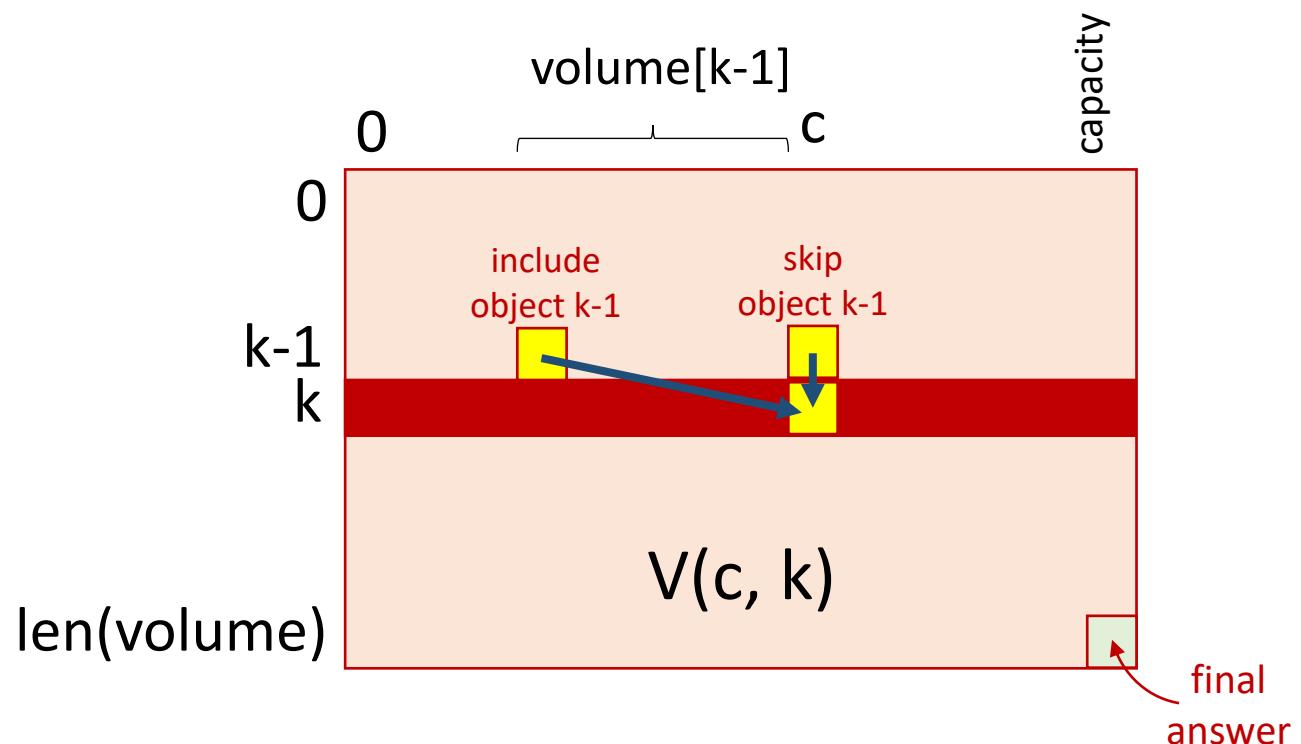
```
def knapsack(volume, value, capacity):
    @memoize
    def solve(c, k): # solve with capacity c and objects 0..k-1
        if k == 0: # no objects to put in knapsack
            return 0, []
        v, solution = solve(c, k-1) # try without object k-1
        if volume[k - 1] <= c: # try also with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

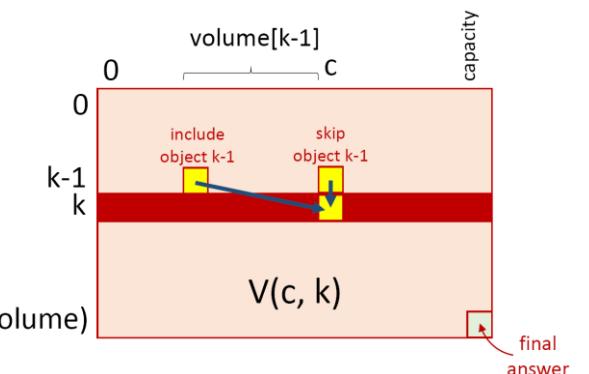
```
def knapsack(volume, value, capacity):
    ① solutions = [(0, [])] * (capacity + 1)
    ② for obj in range(len(volume)):
        for c in reversed(range(volume[obj], capacity + 1)):
            ④ prev_v, prev_solution = solutions[c - volume[obj]]
            v = value[obj] + prev_v
            if 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
- ④ $solutions[:c]$ previous row
- ⑤ compute next row right-to-left
- ⑥ $solutions[:volume[obj]]$
- ⑦ 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](#)
- [Google Code Jam](#)
- [CodeForces](#)
- [Topcoder](#)

Google Code Jam

codingcompetitions.withgoogle.com/codejam

code jam

print "hello, world!"

- Coding competition
- Qualification round 2020 (April 4, 01:00 – April 5, 04:00)
- In 2019 there was 30.000 participants for the qualification round

Scoreboard 2017

Rank	Contestant	Score	Penalty	A. Oversized Pancake Flipper		B. Tidy Numbers		C. Bathroom Stalls			D. Fashion Show	
				5pt	10pt	5pt	15pt	5pt	10pt	15pt	10pt	25pt
1	FatalEagle	100	55:12	4:02	4:29	9:02	9:25	22:30	22:50	23:13	54:48	55:12
2	ACMonster	100	57:32	3:41	4:02	11:37	11:58	25:03	25:27	26:01	57:08	57:32
3	y0105w49	100	58:00	8:28	8:58	16:20	16:47	31:34 1 wrong try	32:03	32:36	53:27	54:00
621	MathCrusader	75	11:52:45	1:52:49	1:53:20	2:11:37 1 wrong try	2:12:05	2:18:48	2:19:18	2:19:48	11:24:45 6 wrong tries	-- 1 wrong try
622	Siddhant22	75	12:07:37	2:15:58 1 wrong try	2:18:01	3:05:10	3:06:39	4:42:18	4:43:32	4:46:14	11:59:37 1 wrong try	Time expired
623	daimi89 Me	75	12:18:31	6:21:53	6:22:58	7:15:07	7:15:54	7:52:33	7:53:13	7:54:04	12:14:31 1 wrong try	--
624	plamenko	75	12:19:28	11:20	11:56	27:38	28:23	42:34	43:17	43:55	12:11:28 2 wrong tries	-- 1 wrong try

Google Code Jam - Qualification Round 2017

Problem A: Oversized Pancake Flipper (description)

- N pancakes each with exactly one happy chocolate side
- K-flipper that can flip K consecutive pancakes
- Problem: Find minimum number of flips to make all pancakes happy, if possible

