Dynamic Programming

Find the solution of a large instance by finding and efficiently memorizing the solutions of small instances.

Ex.

Finding the n-th Fibonacci number.

Fibonacci numbers: 1 1 2 3 5 8 13 21 ...

$f(0) = f(1) = 1$, $f(n) = f(n-1) + f(n-2)$, for all $n >= 2$.

Python function

```python
def fib_DP(n):
    a, b = 1, 1
    for i in range(2, n+1):
        a, b = b, a+b
    return b
```

Notice the difference of a recursive call

```python
def fib_RC(n):
    if n<=1:
        return 1
    else:
        return fib_RC(n-1) + fib_RC(n-2)
```


Demo: fib.py

Methods for IP : branch - and - bound, dynamic programming (DP) etc.

(DP)

Ex. 0-1 Knapsack problem

Given n items with size $a_i > 0$

value $c_i > 0$

1 container capacity $b > 0$

Object: pack the items so that the total value is maximized.

formulation

$$\max \quad z = \sum c_i x_i$$

$$s.t. \quad \sum a_i x_i \leq b$$

$$x_i \in \{0,1\}$$

Here, we assume $a_i, b, c_i \in \mathbb{Z}$ (integer)

$$\max \quad 3x_1 + 4x_2 + x_3 + 2x_4$$

$$s.t. \quad 2x_1 + 3x_2 + x_3 + 3x_4 \leq 4$$

$$x_i \in \{0,1\}$$

A "simple" yet difficult problem

enumeration method $\Rightarrow O(n2^n)$ time.

DP

Let $f(i,k) = \max\limits_{x_i \in \{0,1\}} \sum\limits_{j=1}^{i} c_j x_j$ $\quad i=1,2,\cdots,n$

$\quad k=0,1,\cdots,b$

$\sum\limits_{j=1}^{i} a_j x_j \leq k$

Then $f(1,k) = \begin{cases} 0 & k < a_1 \\ \\ c_1 & k \geq a_1 \end{cases}$

and

$$f(i,k) = \max\left\{ f(i-1,k),\ f(i-1, k-a_i) + c_i \right\},\quad i \geq 2$$

$\quad$ (assume $f(i-1, k-a_i) = -\infty$ if $k < a_i$)

Ex.

i \ k	1	2	3	k
1	0	3	3	3
2	0	3	4	4
3	1	3	4	5
4	1	3	4	(5)

optimal

running time $O(nb)$

better if $b < 2^n$.

结构时间分析

Shortest Path Problem

Input: Graph G=(V,E), edge length l(u,v), s, t
Output: a shortest s-t path (or its nonexistence)

Note: it may not exist if there exists a negative cycle.

In the following, we assume there is no such a cycle.

Method 1: find the shortest one from ALL paths

=> Too many paths! (see Movie 1)

Method 2: Pulling method (=> Dijkstra's method)

=> Movie 2
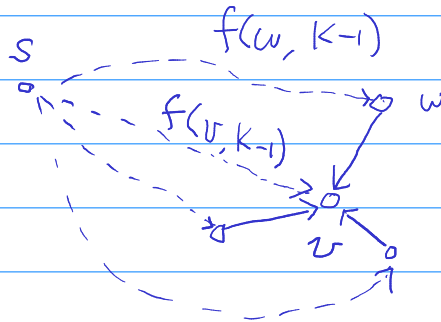
Bellman-Ford algo for the shortest path problem

Define

f(v, k) = length of a shortest s-v path that uses at most k edges

$\Rightarrow$ We want f(v, n-1) for all v.

$$f(v, 0) = \begin{cases} 0 & v = s \\ \infty & v \neq s \end{cases}$$

$$f(v, k) = \min\left\{ f(v, k-1), \min_{w:(w,v)\in E}\{f(w, k-1) + \ell(w, v)\} \right\}$$

Observation

We can safely drop the second parameter in f(v, k), i.e., consider

$$f(v) = \min\left\{ f(v), \min_{w:(w,v)\in E}\{f(w) + \ell(w, v)\} \right\}.$$

$\Longleftrightarrow$ Bellman-Ford algo

** Advanced topic (optional)

* Dijkstra's algorithm: 1-1 or 1-many/all
* Bellman-Ford algorithm: 1-all
* Floyd-Warshall: all-all

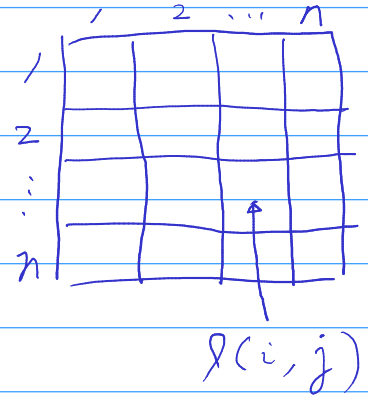More efficent than n times of the first two.

Let V = {1, 2, ..., n}, and we use the incidence matrix.

main

        for i = 1, 2, ..., n
            for j = 1, 2, ..., n

$$dist[i, j] = \begin{cases} 0 & i = j \\ \ell(i,j) & (i,j) \in E \\ \infty & otherwise \end{cases}$$

$\ell(i, j)$

        for k = 1, 2, ..., n
            for i = 1, 2, ..., n
                for j = 1, 2, ..., n
                    if dist[i, j] > dist[i, k] + dist[k, j] {
                        dist[i, j] = dist[i, k] + dist[k, j]
                    }

time: $O(n^3)$
space: $O(n^2)$

Correctness

f(i, j, k) = length of a shortest i-j path that uses only nodes 1, ..., k

$$\Rightarrow \quad f(i, j, 0) = \begin{cases} 0 & i = j \\ \ell(i,j) & (i,j) \in E \\ \infty & otherwise \end{cases}$$

$$f(i, j, k) = \min \{ f(i, j, k-1), f(i, k, k-1) + f(k, j, k-1) \}$$

nodes 1, 2, ..., k-1

(i) (k) (j)