

REOP - Session 3

Flows

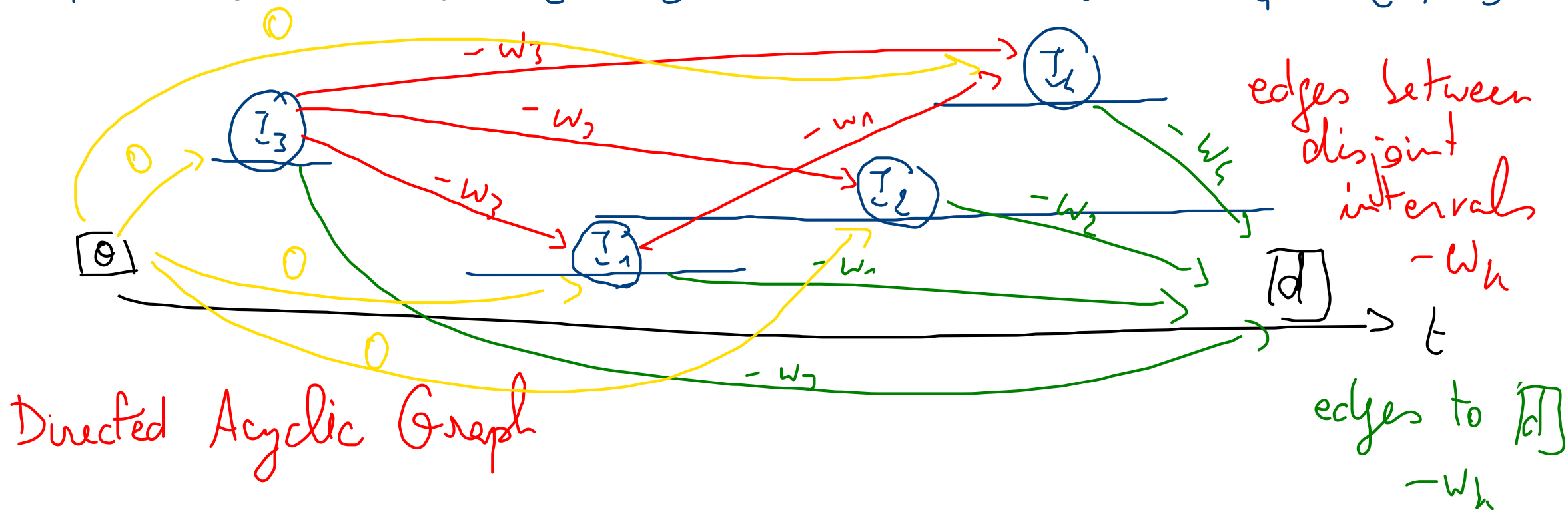
I / Homework

A) Ex 5.12: Flat rental

We have a collection \mathcal{I} of
each of them having weight w_k

bounded closed intervals $I_k = [a_k, b_k]$

1) $I_1 = [2, 4]$ $I_2 = [3, 7]$ $I_3 = [0, 1]$ $I_4 = [5, 6]$



We look for a shortest path in a graph with negative weights
~~Dijkstra~~ \rightarrow topological ordering works anyway

A shortest path is a set of disjoint intervals with max weight
(min negative weight)

2) An Airbnb host wanting to plan bookings | Disjoint bookings!
Interval = dates of the guest's stay | Maximize your
Weight = price they will pay | revenue: ∞

Q: What if we shift the weights to \mathbb{R}_+ ?
Then we get biased towards paths with
smaller number of edges

B) Ex 5.19: Longest common subword

$w_1 = \underline{a} \underline{b} \underline{c} \underline{d} \underline{a} \underline{c}$ common subword $bcd a$ of length 4
 $w_2 = \underline{b} \underline{c} \underline{a} \underline{d} \underline{e} \underline{a}$

Naive method: enumerate all subwords of w_1 & w_2 ⊕ Compare
 Complexity: $2^{m_1} \times 2^{m_2} \times \min(m_1, m_2)$ exponential

2 nested loops

↳ nb of subwords in w_1
 (= subsets of letters)

↳ worst-case comparison

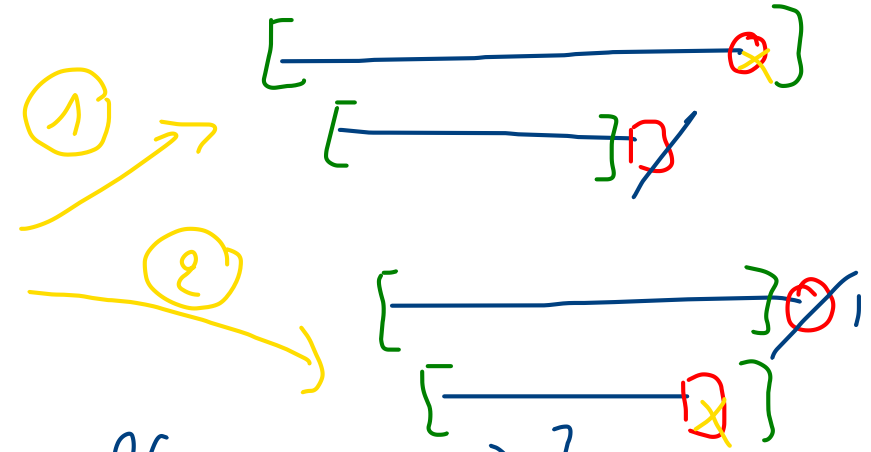
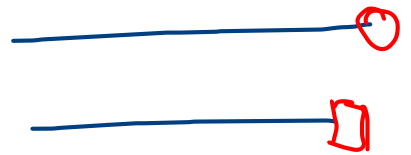
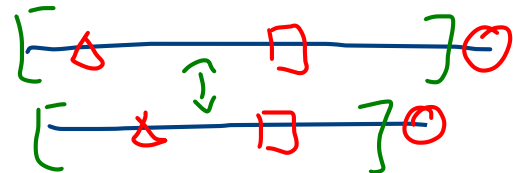
Find a dynamic programming approach: Bellman equation

Generalize the problem:

Compute $l(i, j)$ = length of the longest common subword between $w_1[1:i]$ & $w_2[1:j]$

2 cases:

- Last letters are identical: $l(m_1, m_2) = l(m_1 - 1, m_2 - 1) + 1$
- Last letters are different:



(B) $l(m_1, m_2) = \max \{ l(m_1, m_2 - 1), l(m_1 - 1, m_2) \}$

Initialization: $l(0, m_2) = l(m_1, 0) = 0$

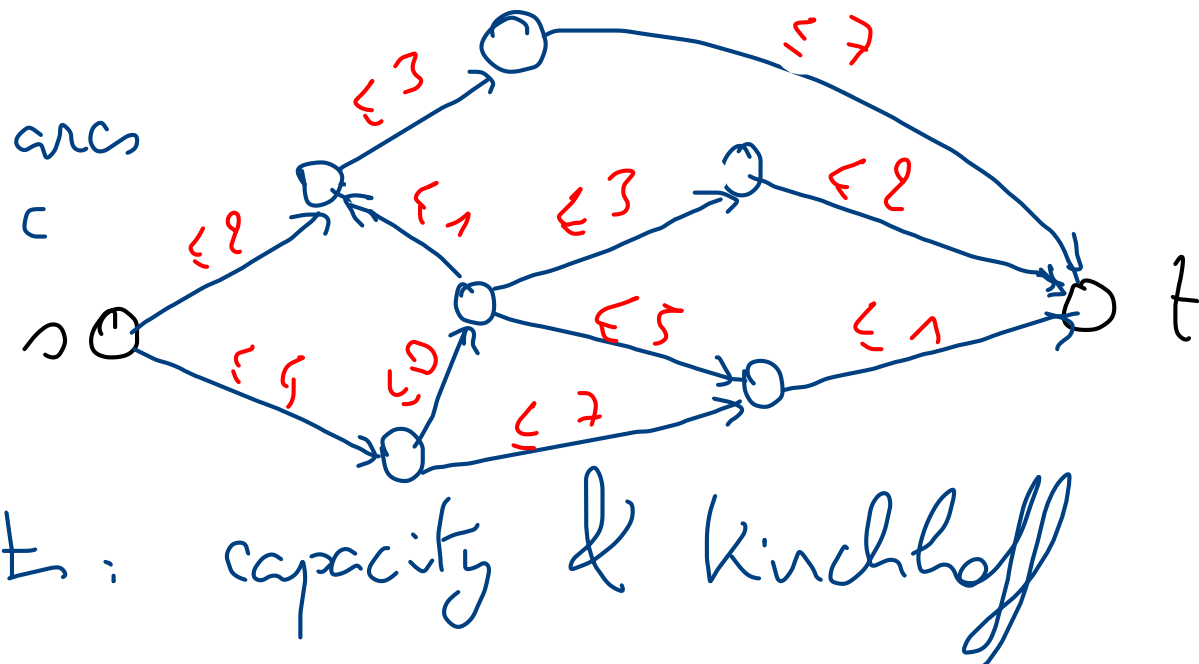
DP alg: compute $l(i,j)$ for $(i,j) \in [1, n_1] \times [1, n_2]$
 each application of the Bellman eq has $O(1)$ cost
 total complexity $O(n_1 \times n_2)$

This gives us the length, not the word
 To recover the subword we store the argmax in (B)

II/ Flow vocabulary

A) s-t flow (source - target)

Input: - a digraph $D = (V, A)$
 - 2 special nodes s & t
 - capacities $u(a) \geq 0$ on every arc

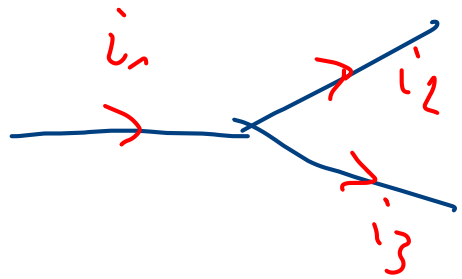


An s-t flow is a vector
 $f \in \mathbb{R}_+^A$ (s.t. $f(a) \geq 0$)
 satisfying 2 constraints: capacity & Kirchhoff

Capacity constraint: $\forall a \in A, f(a) \leq u(a)$

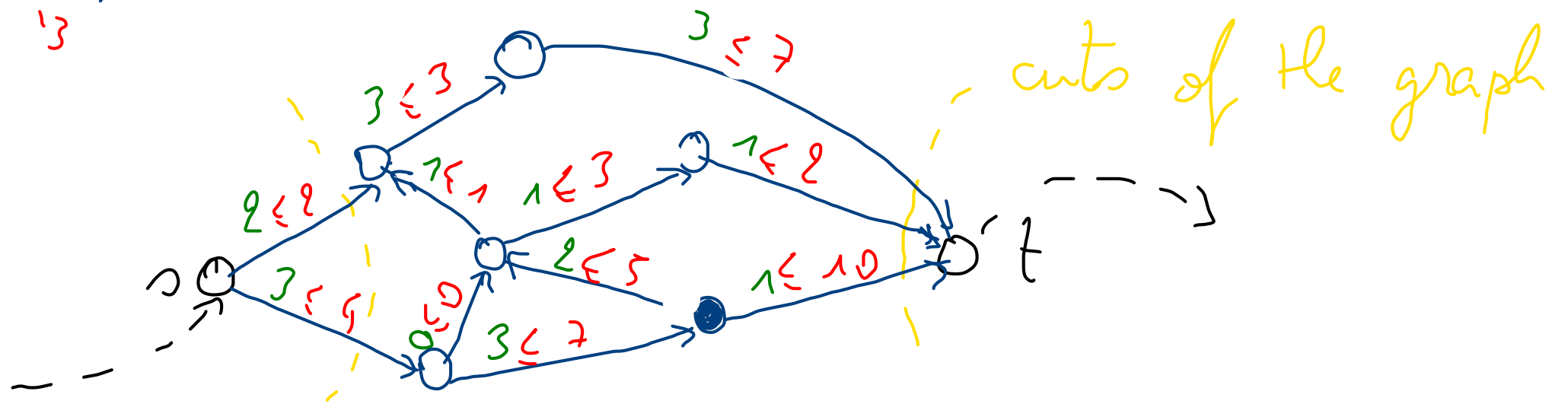
Kirchhoff constraint: $\forall v \in V,$
 conservation of the flow $\sum_{a \in \delta^-(v)} f(a) = \sum_{a \in \delta^+(v)} f(a)$

\hookrightarrow in-neighbors of v
 except for s & t



$$i_1 = i_2 + i_3$$

f



The value of an $s-t$ flow is the total quantity flowing out of the source

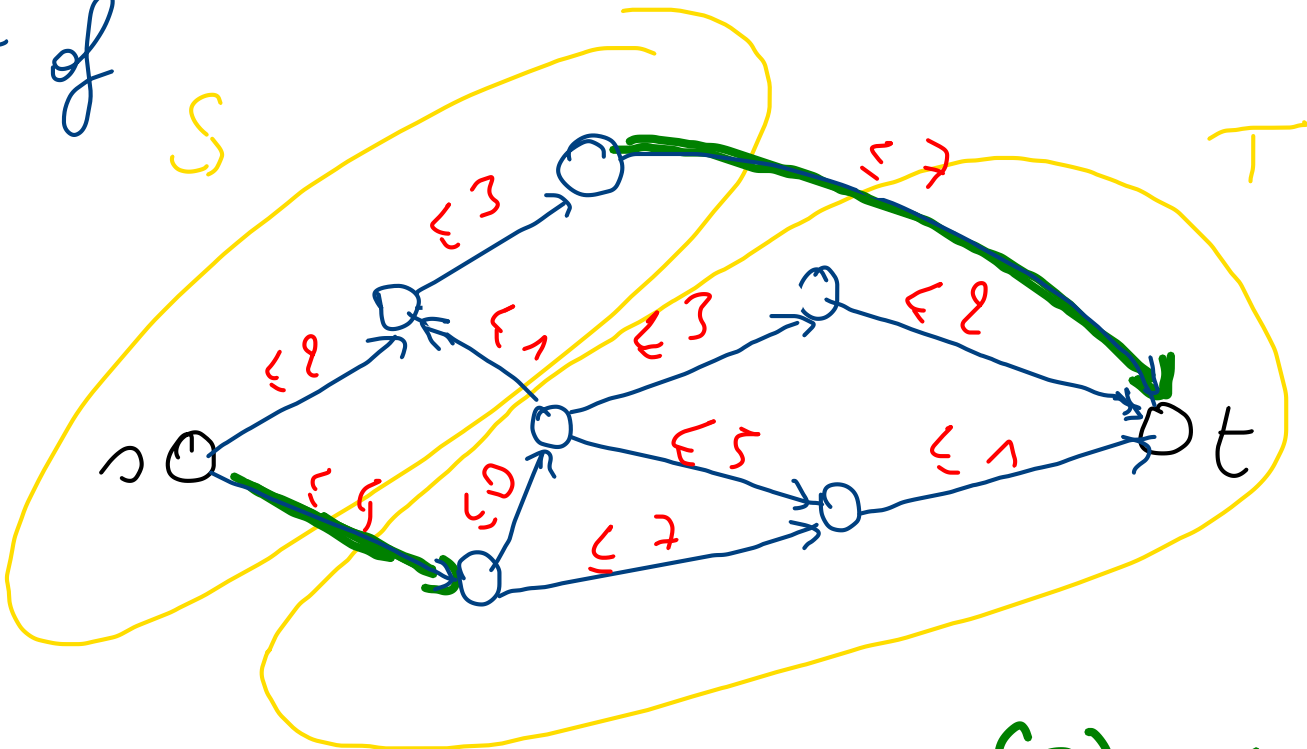
$$\text{val}(f) = \sum_{a \in \delta^+(s)} f(a) - \sum_{a \in \delta^-(s)} f(a) = (3+8) - 0$$

B) s-t cuts

An s-t cut is a partition (S, T) of the vertices such that $s \in S$ & $t \in T$

$$\begin{cases} S \cup T = V \\ S \cap T = \emptyset \end{cases}$$

It can also be viewed as a set of edges $B = \delta^+(S)$ splitting the graph between s & t



The capacity of the cut is

$$u(S, T) = \sum_{\substack{i \in S, j \in T \\ (i, j) \in A}} u(i, j)$$

$$u(B) = \sum_{a \in B} u(a)$$

$$u(B) = 4 + 7$$

2 problems:
 • Maximum flow: find an s-t flow of max value
 • Minimum cut: find an s-t cut of min capacity
 equiv. \leftrightarrow

Th 6.5: The value of a max flow is equal to the capacity of a min cut

C) Minimum cost flows (other framework, more general)

- Input:
- a digraph $D = (V, A)$ (with no "special vertices" s.t.)
 - lower & upper capacities $0 \leq l(a) \leq u(a)$ on each arc
 - cost values $c(a) \geq 0$ on each arc
 - inputs / inflows $b(v)$ for each vertex

A b -flow is a vector $f \in \mathbb{R}_+^A$ satisfying

- Capacity constraint: $\forall a \in A, l(a) \leq f(a) \leq u(a)$

- Kirchhoff constraint: $\forall v \in V, b(v) + \sum_{a \in \delta^-(v)} f(a) = \sum_{a \in \delta^+(v)} f(a)$

We must have $\sum_{v \in V} b(v) = 0$

algebraic
 ≥ 0 : flows in
 ≤ 0 : flows out

Problem: find a b -flow
with minimum cost

$$c(f) = \sum_{a \in A} c(a) f(a)$$

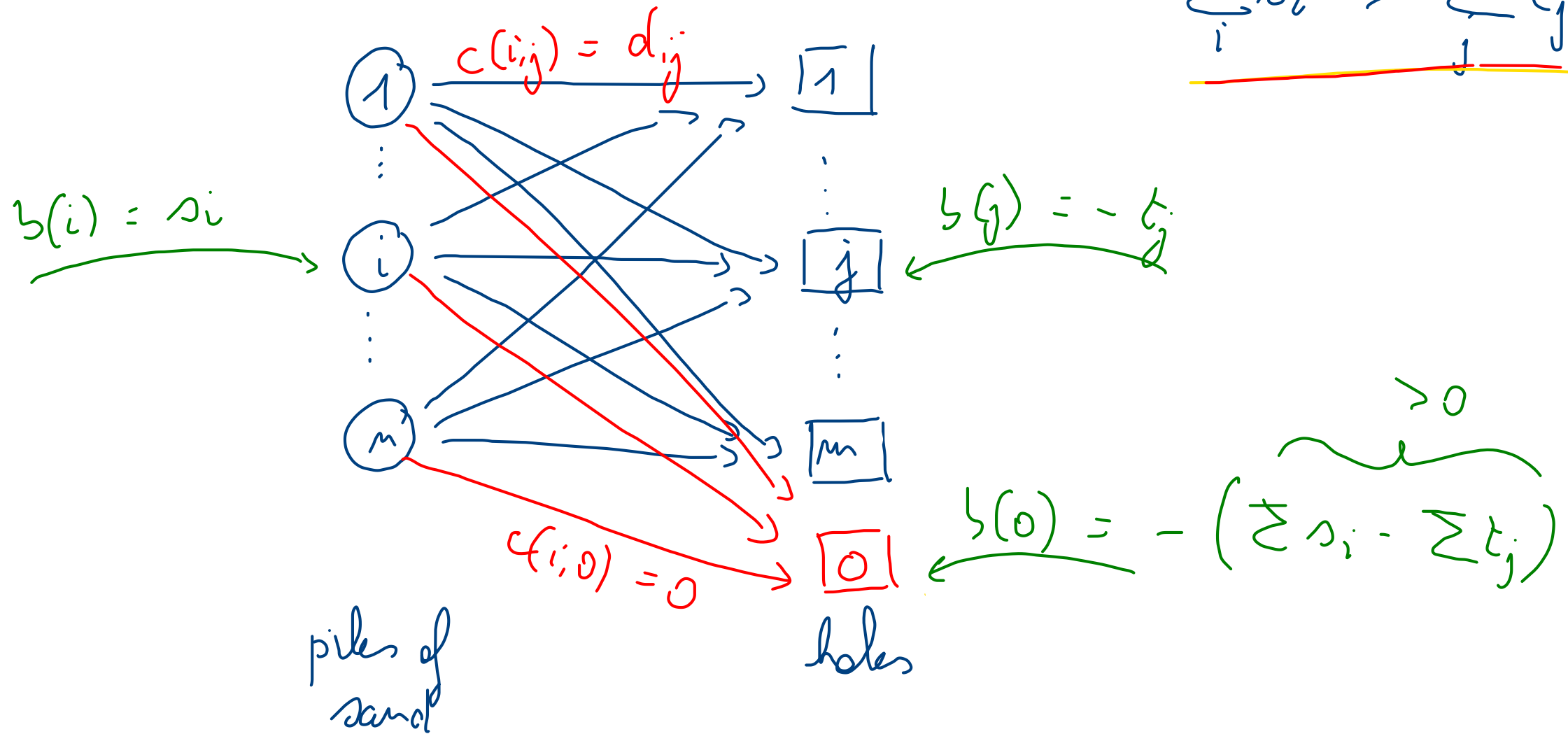
↙ with cost

D) Exercises

6.9 & 6.13

Ex 6.9 (Monge's transportation pl)

What if
 $\sum_i s_i > \sum_j t_j$



Flow variable x_{ij} = quantity of sand transported from i to j
 Solve a minimum cost flow problem (no capacity constraints)

IV/ Linear Programming for flows

General technique for flow problems

Useful if there additional "nonstandard constraints"

Formulation for max s-t flow:

$$\begin{array}{l} \text{(LP)} \quad \max_f \quad \sum_{a \in \delta^+(s)} f_a - \sum_{a \in \delta^-(s)} f_a = \text{val}(f) \quad \text{linear} \\ \text{s.t.} \quad \left| \begin{array}{l} 0 \leq f_a \leq u(a) \quad \forall a \in A \quad \text{linear} \\ \sum_{a \in \delta^-(v)} f_a = \sum_{a \in \delta^+(v)} f_a \quad \forall v \in V \setminus \{s, t\} \quad \text{linear} \end{array} \right. \end{array}$$

Ex: do the same for min cost s-flow

Prop 6.11: Integer capacities \implies integer flow values at the optimum

Prop 6.12: The dual of (LP) is the minimum cut problem

III/ Algorithms for the max s-t flow problem

Rg: Of course there are similar algorithms for the min cost s-flow
(see notes)

A) Optimality condition

Prop 6.3: If f is an s-t flow & (S, T) is an s-t cut, then

$$\text{val}(f) \leq u(S, T) \quad = 0 \text{ by conservation}$$

Proof: $\text{val}(f) = \sum_{a \in \delta^+(s)} f(a) - \sum_{a \in \delta^-(s)} f(a) + \sum_{\substack{v \in S \\ v \neq s}} \left(\sum_{a \in \delta^+(v)} f(a) - \sum_{a \in \delta^-(v)} f(a) \right)$

$$= \sum_{a \in \delta^+(S)} f(a) - \sum_{a \in \delta^-(S)} f(a)$$

$$\leq \sum_{a \in \delta^+(S)} u(a) - \sum_{a \in \delta^-(S)} 0$$

$$= \underbrace{\sum_{a \in \delta^+(S)} u(a)}_{u(S, T)}$$

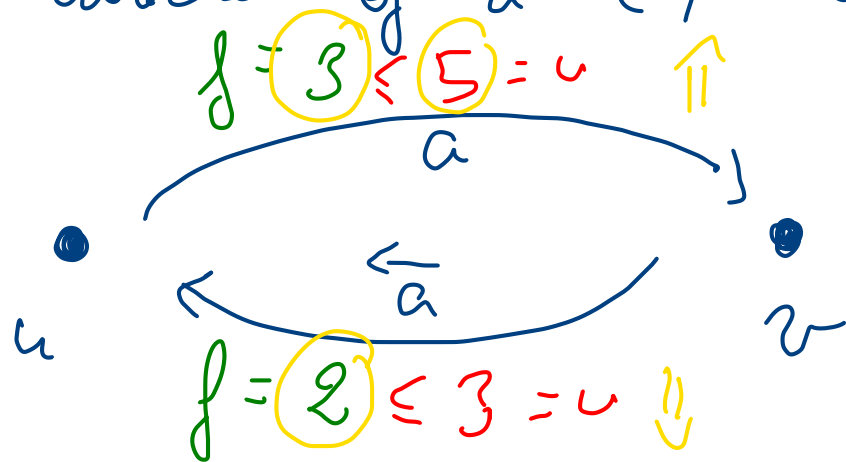
capacity

edges within S cancel out

For every arc $a = (u, v)$ we define $\bar{a} = (v, u)$
and residual capacities

$$u_r(a) = (u(a) - f(a)) + (f(\bar{a}))$$

They tell us how much we can increase the flow in the direction of a (\neq along the arc a)



2 ways to \uparrow the flow

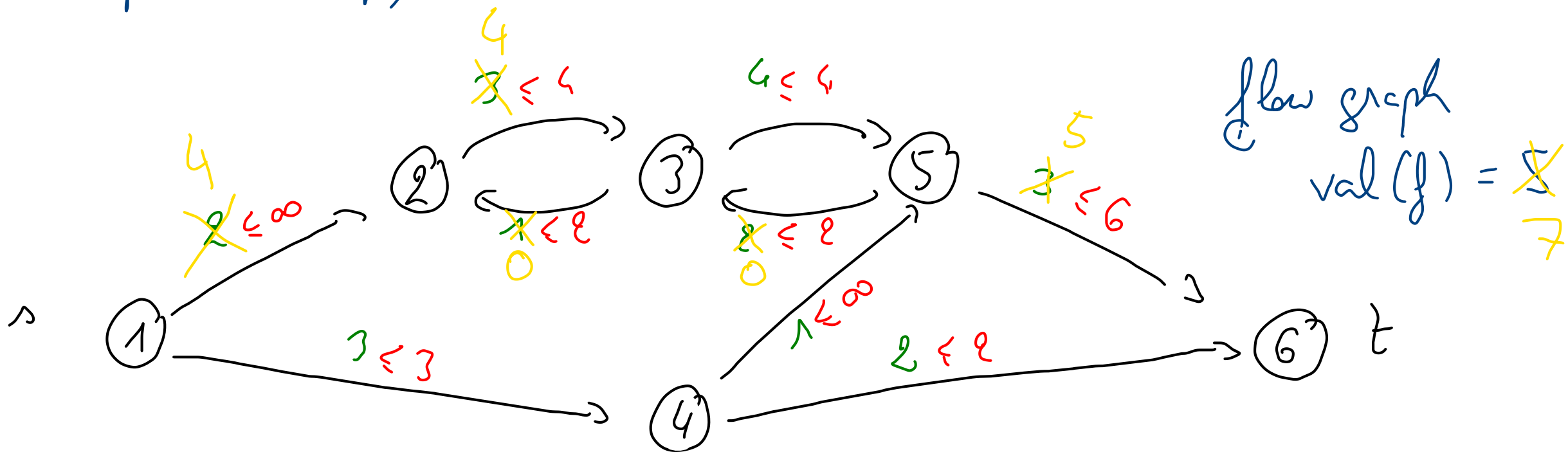
- more flow on $a + (5 - 3)$
- less flow on $\bar{a} - 2$

The residual graph with $D_r = (V, A_r)$ is the capacitated

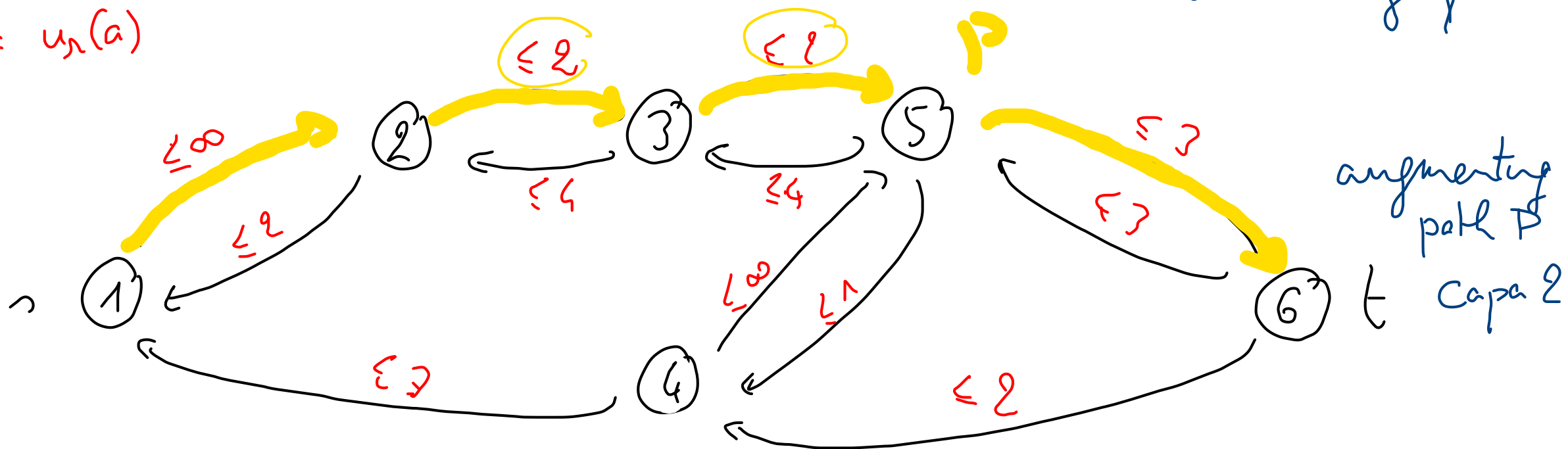
$$A_r = \{ a \in A \cup \bar{A} : u_r(a) > 0 \}$$

An augmenting path P is an s - t path in the residual graph

Example (Woodlap)



cap u_r(a)



If we rebuild the residual graph, s & t are now in separate connected components, so there is no augmenting path in D_n

Thm 6.4: An s - t flow is maximal iff there is no augmenting path in the residual graph

B) Ford-Fulkerson algorithm

1. Start with $f(a) = 0 \quad \forall a \in A$
 2. While there is an augmenting path:
 - 1. Select an augmenting path P with min number of edges
 - 2. Increase the flow along P by $\min_{a \in P} c_n(a)$
 3. Return f
- Breadth-First Search

Questions:

- How do we find an augmenting path?
 \Leftrightarrow Path from s to t in D_n
- Which path do we select? Important for complexity
Edmonds-Karp algorithm \rightarrow polynomial runtime
 $O(|A|^2 \times |V|)$ (can be better)

VI/ Homework

Ex 6.13, 6.10

if you want: Ex 6.16, 6.18