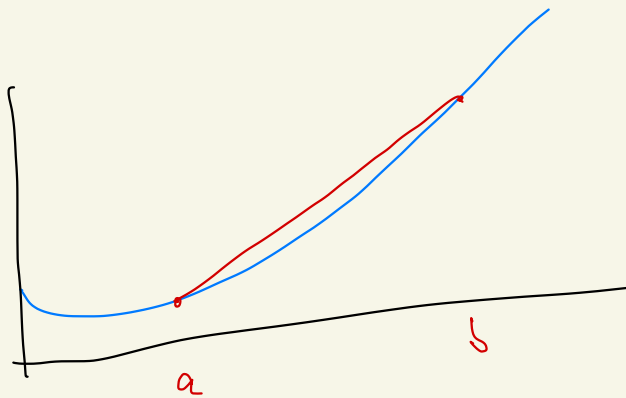
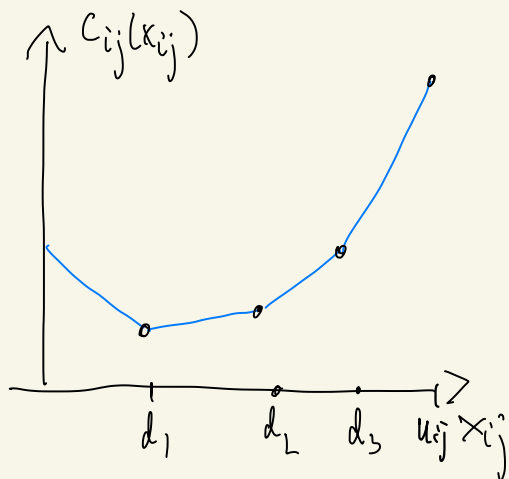


- Standard cost function in min cost flow problem c_{ij} is a constant so cost of sending x_{ij} units along c_j is $c_{ij}x_{ij}$

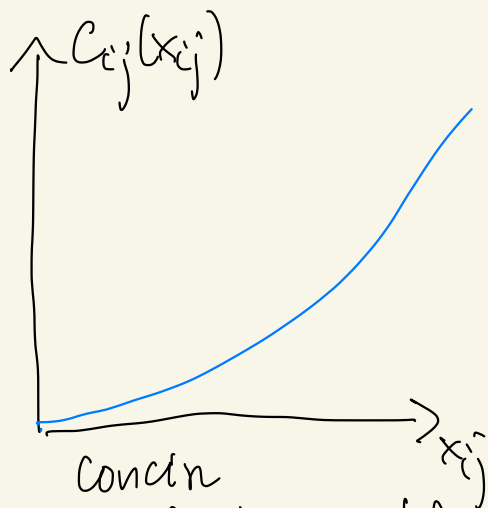
- A function f is **convex** if $\forall a < b$ in the domain of f the values of f in the interval $[a, b]$ lie below the straight line from $(a, f(a))$ to $(b, f(b))$



Ahuja chapter 14 convex cost flows



discrete piecewise linear
 $O(\# \text{break points})$ space



concn
cost function specified
in functional form
e.g. x_{ij}^4 $O(1)$ space

Min convex cost flow:

$$\min \sum c_{ij}(x_{ij})$$

$$\text{s.t. } b_x(i) = d(i) \quad \forall i \in V \quad (*)$$

$$0 \leq x_{ij} \leq u_{ij} \quad \forall ij \in A$$

$$x_{ij} \in \mathbb{Z} \quad \forall ij \in A$$

consider only integer flows!

assumptions on $N = (V, A, c, u, b, CC)$

(1) $D = (V, A)$ is a digraph

(2) $b \in \mathbb{Z}^n$ (integer balances)

(3) $c_{ij} = 0 \forall ij$

(4) $\exists (ij)$ -paths of ∞ capacity $\forall ij \in V$

(5) $c_{ij}(0) = 0 \forall ij \in A$

Section 14.2 on applications \Rightarrow self study

14.3 Transformation to a normal
min cost flow problem when c_{ij} is
convex and piecewise linear for all arcs

Drawback of solution: The new network
may be much larger than N as # arcs will
depend on # break points of the cost
functions.

Assumption: each c_{ij} has exactly p break points (consists of p linear segments)

So c_{ij} is constant in each of the segments

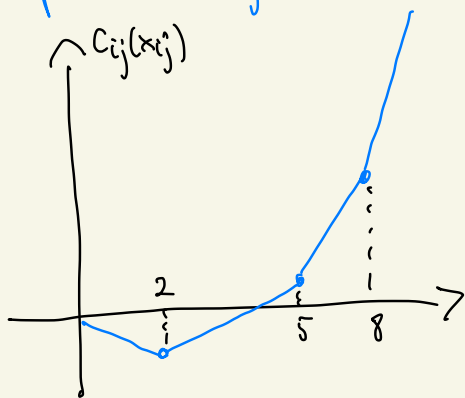
$$[d_{ij}^0, d_{ij}^1], [d_{ij}^1, d_{ij}^2] \dots [d_{ij}^{p-1}, d_{ij}^p]$$

c_{ij}^0 c_{ij}^2 c_{ij}^p

When $d_{ij}^0 = 0$ and $d_{ij}^p = u_{ij}$

Note that p is finite as u_{ij} is finite

look at x_{ij} and split in p parts which represent the part of x_{ij} belongs to each of the p segments:



fill up from below:
 in example, if $x_{ij} = 6$
 then segments 1 and 2
 are filled and the flow in
 segment 3 is 1

Given x_{ij} we define segment flows y_{ij}^k for $k \in [p]$:

$$\textcircled{\square} \quad y_{ij}^k = \begin{cases} 0 & \text{if } x_{ij} \leq d_{ij}^{k-1} \\ x_{ij} - d_{ij}^{k-1} & \text{if } x_{ij} \in [d_{ij}^{k-1}, d_{ij}^k] \\ d_{ij}^k - d_{ij}^{k-1} & \text{if } x_{ij} \geq d_{ij}^k \end{cases}$$

Then

$$x_{ij} = \sum_{k=1}^p y_{ij}^k \quad \text{and} \quad c_{ij}(x_{ij}) = \sum_{k=1}^p c_{ij}^k y_{ij}^k$$

This transforms the model $(*)$ into $(**)$:

$$\min \sum_{ij \in A} \sum_{k=1}^p c_{ij}^k y_{ij}^k$$

$$\sum_{ij} \sum_{k=1}^p y_{ij}^k - \sum_{ji} \sum_{k=1}^p y_{ji}^k = b(i) \quad (**)$$

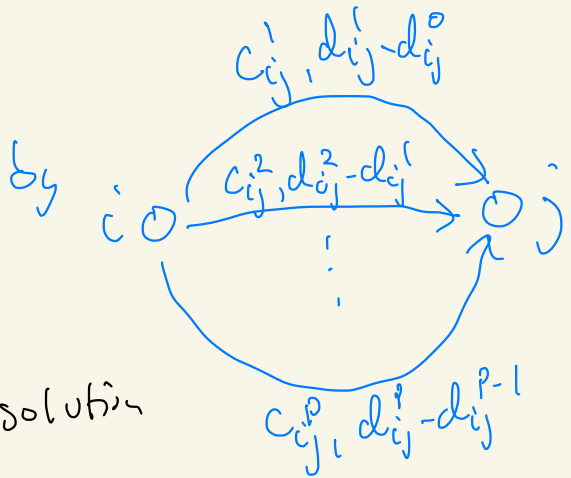
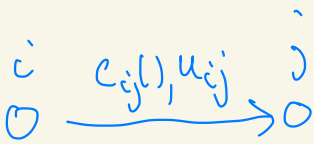
$$0 \leq y_{ij}^k \leq d_{ij}^k - d_{ij}^{k-1} \quad \forall ij$$

$$\min \sum_{ij \in A} \sum_{k=1}^p c_{ij}^k y_{ij}^k$$

$$\sum_{ij} \sum_{k=1}^p y_{ij}^k - \sum_{ji} \sum_{k=1}^p y_{ji}^k = b(i) \quad (**)$$

$$0 \leq y_{ij}^k \leq d_{ij}^k - d_{ij}^{k-1} \quad \forall ij$$

Claim: (**) is a min cost flow problem in the network N' when we replace



Def y is a **contiguous** solution

to (**) if $y_{ij}^q > 0$

$$\forall y_{ij}^r = d_{ij}^r - d_{ij}^{r-1} \quad \text{for all } r < q$$

Show equivalence between $(*)$ and $(**)$

Suppon x solves $(*)$

Def $y_{ij}^1, \dots, y_{ij}^p$ from x_{ij} as in $(*)$

that is, fill up the intervals $[d_{ij}^0, d_{ij}^1] \dots [d_{ij}^{p-1}, d_{ij}^p]$

In that order.

Then y is a contiguous solution and

$$c_{ij}(x_{ij}) = \sum_{k=1}^p c_{ij}^k y_{ij}^k$$

so the cost of x and y are the same

Suppon y is a contiguous solution to $(**)$

Define x_{ij} as $x_{ij} = \sum_{k=1}^p y_{ij}^k$ then

x is a solution to $(*)$ and its cost

is the same as the cost of y .

Note that there may be non contiguous solutions to $(**)$ but they are not optimal:

Suppose $y_{ij}^l = M > 0$ but $y_{ij}^k < d_{ij}^k - d_{ij}^{k-1}$
 some $k < l$.

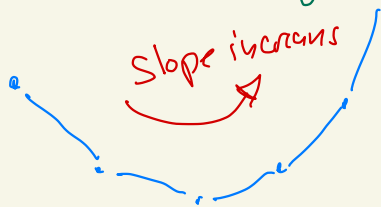
Then we get a dual solution to $(**)$

by setting $y_{ij}^k \leftarrow y_{ij}^k + Q$ and

$$y_{ij}^l \leftarrow y_{ij}^l - Q$$

when $Q = \min \{ M, (d_{ij}^k - d_{ij}^{k-1}) - y_{ij}^k \}$

why? because $C_{ij}^k < C_{ij}^l$ by convexity of $C_{ij}(\cdot)$



Conclusion All optimal solutions to $(**)$ are contiguous.

• Algorithmic challenge with reduction:

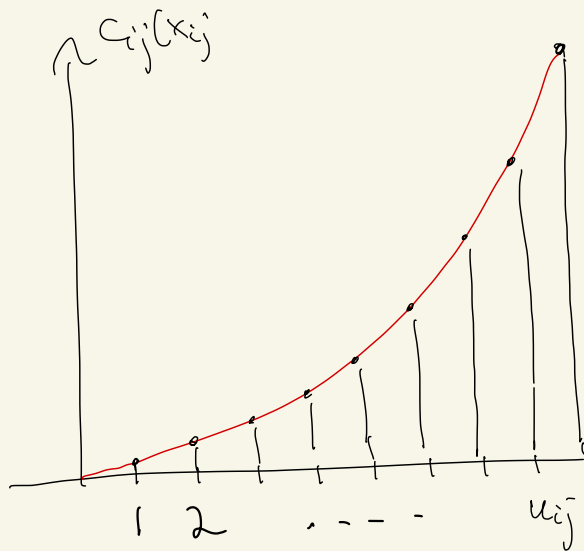
Need p copies of each original arc

• Not a problem if $c_{ij}(x_{ij})$ is specified as a piecewise linear function, since then we need (at to) p breakpoints for each arc ij .

• If instead $c_{ij}(x_{ij})$ is in functional form (e.g. $c_{ij}(x_{ij}) = x_{ij}^2$)

and we represent $c_{ij}(x_{ij})$ as a piecewise linear approximation with u_{ij} lines

then we may need much more space in model ($\otimes x$)



14.4 pseudopolynomial algorithms

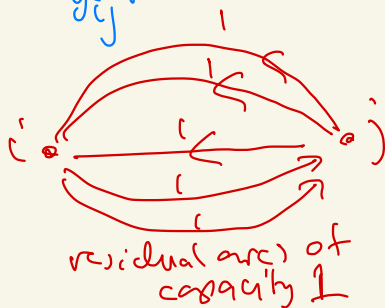
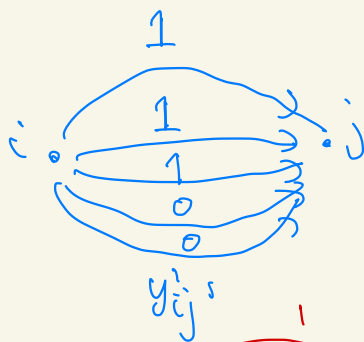
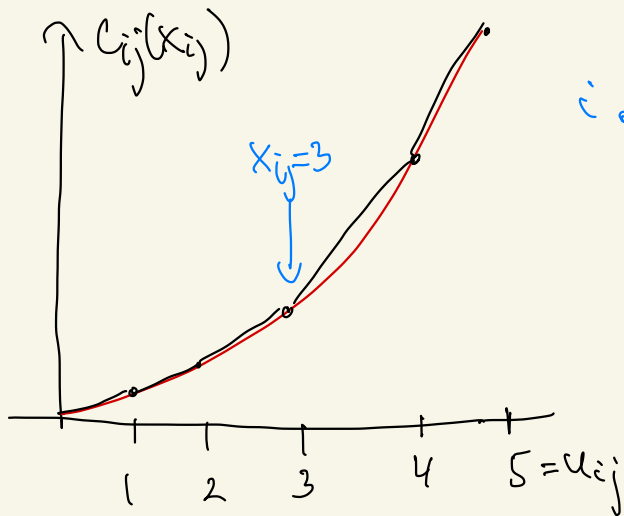
Modifications of Cycle Cancelling and Buildup method for standard min cost flow.

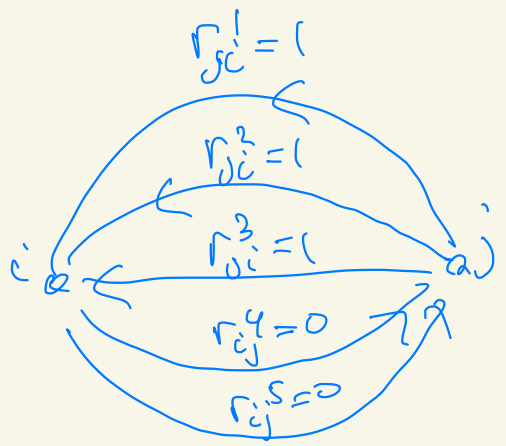
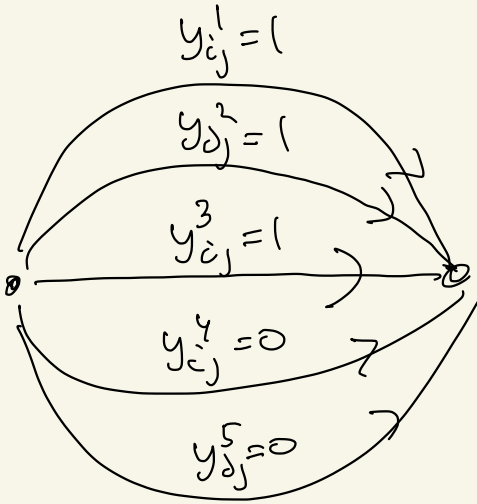
New approach:

Do not make all p copies of an arc ij but introduce only the relevant copies.

The residual network has (potentially) many arcs in both directions between i and j

Example based on figure 14.2:





Effect of increasing $x_{ij} \rightarrow x_{ij} + 1$:

$$y_{ij}^4 < 1 \text{ as } c_{ij}^4 < c_{ij}^5$$

Effect of decreasing $x_{ij} \rightarrow x_{ij} - 1$:

$$y_{ij}^3 < 0 \text{ as } c_{ij}^3 > c_{ij}^2 > c_{ij}^1$$

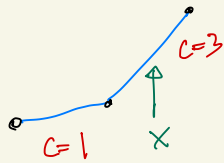
$$\text{so } -c_{ij}^3 < -c_{ij}^2 < -c_{ij}^1$$

Conclusion: Enough to keep ij^3 and ij^4 in the residual network

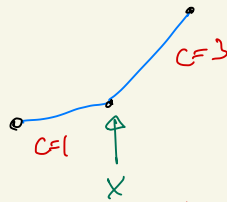
This holds in general: we just need to store two arcs in $N(x)$ per arc ij

Construction of $N(x) \quad \forall ij \in A:$

- if $x_{ij} < u_{ij}$ then ij is an arc in $N(x)$
with cost $C_{ij}(x_{ij}+1) - C_{ij}(x_{ij})$
- if $0 < x_{ij}$ then $j\bar{i}$ is in $N(x)$
with cost $C_{ij}(x_{ij}-1) - C_{ij}(x_{ij})$



Cost of ij in $N(x)$
is 3
Cost of $j\bar{i}$ in $N(x)$
is -3



Cost of ij in $N(x)$
is 3
Cost of $j\bar{i}$ in $N(x)$
is -1

$r_{ij} =$ distance from x_{ij} to next breakpoint
 $r_{j\bar{i}} =$ distance from previous breakpoint to x_{ij}

Cycle cancelling for convex cost flows

Same as normal, except for construction and maintenance of $N(x)$:

Given feasible flow x

- Construct $N(x)$
- While $\exists W \in N(x) : c(W) < 0$
 - let $\delta(W)$ be capacity of W (defined below)
 - $x \leftarrow x \oplus \delta(W)W$
 - Construct $N(x)$

$\delta(W)$ is now the minimum residual capacity of an arc ij on W

So we only increase/decrease flow in ij to the next breakpoint

Note that, by the equivalence of models (x) and (x') and the definition of $N(x)$ x is optimal if and only if $N(x)$ has no negative cycle.

Possible improvement

- After augmenting by $\delta(w)$ units along w the flow on one or more arcs i_j of w has reached breakpoint so it is possible that $c(w)$ remains negative in the new residual network.

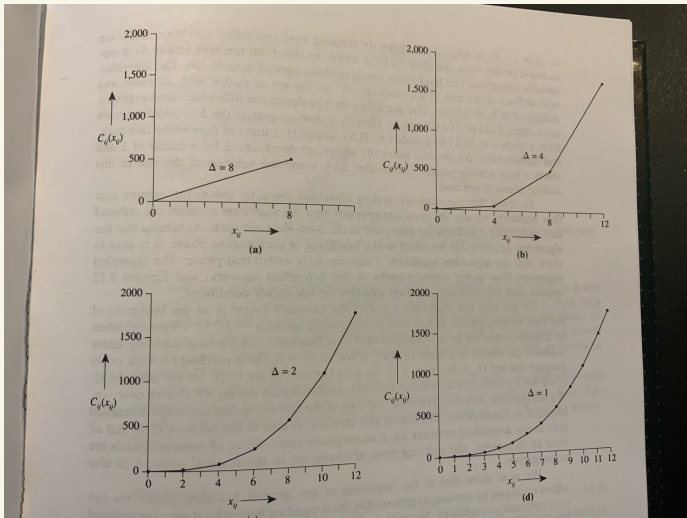
In that case we can augment along w with $\delta(w)$ units for the new value $\delta(w)$.

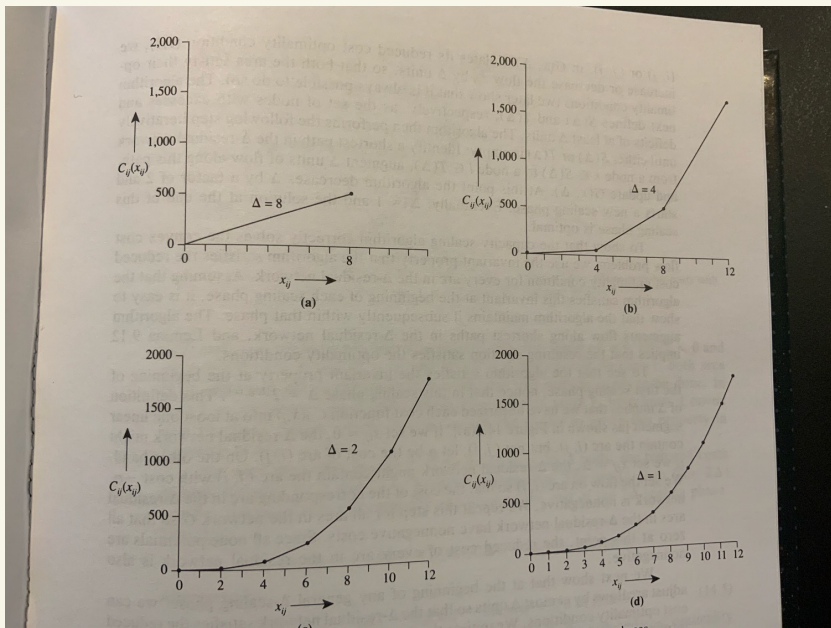
Buildup algorithm:

Again very similar except for
det. of $N(x)$.
(self study!)

14.5 Obtaining a polynomial algorithm.

- Build on the capacity scaling algorithm in Section 10.2.
- The algorithm does not linearize $C_{ij}(x_{ij})$ in one step but in a logarithmic # of iterations in which the approximation becomes more and more accurate





$$C_{ij}(x_{ij}) = x_{ij}^4 \quad u_{ij} = 12 \in [2^3, 2^4 - 1) \Rightarrow \lceil \log_2 u_{ij} \rceil = 3 \quad \Delta = 2^3 = 8$$

- step 1 one segment of length 8 and slope $\frac{8^4}{8} = 8^3$
- step 2 segment above divided into 2 segments of length 4 with slopes $\frac{4^4}{4} = 4^3$ and $\frac{8^4 - 4^4}{4}$ respectively
- in step 4 all segments have length 1

In the Δ -scaling phase we maintain
the Δ -residual network $N(x, \Delta)$

• If $x_{ij} + \Delta \leq u_{ij}$ then $i'j \in N(x, \Delta)$
and has cost $\frac{c_{ij}(x_{ij} + \Delta) - c_{ij}(x_{ij})}{\Delta}$

• if $x_{ij} \geq \Delta$ then $j'c \in N(x, \Delta)$
and has cost $\frac{c_{ij}(x_{ij} - \Delta) - c_{ij}(x_{ij})}{\Delta}$

let $u = \max_{ij} u_{ij}$

initially $\Delta = 2^{\lfloor \log_2 u \rfloor}$ $x \equiv 0$ and $\pi \equiv 0$

x, π show that x is optimal

as $c_{ij}(0) = 0 \quad \forall ij$ by assumption

Phase Δ :

1. Construct $N(x, \Delta)$
2. $\forall ij \in A$: if ij or ji is in $N(x, \Delta)$ and has negative reduced cost we increase or decrease the flow by Δ units
3. $S(\Delta) \leftarrow \{i \mid b(i) - b_x(i) \geq \Delta\}$
 $T(\Delta) \leftarrow \{i \mid b_x(i) - b(i) \geq \Delta\}$
4. Find shortest path distances $d(i)$ from some $k \in S(\Delta)$ and let P be a shortest (k, ℓ) -path for some $\ell \in T(\Delta)$
5. $\pi \leftarrow \pi - d$
6. Augment x by sending Δ units along P
7. Update $S(\Delta), T(\Delta)$ and if both are nonempty go to 4
8. If $\Delta \geq 1$ set $\Delta \leftarrow \frac{\Delta}{2}$ and goto 1

Note that on the 1-scalars phase the discretization is complete, we now prove that the final x is optimal.

Invariant $C_{ij}^{\pi} \geq 0$ for every arc $ij \in N(x, \Delta)$

Invariant $C_{ij}^T \geq 0$ for every arc $ij \in N(x, \Delta)$

Note that it is enough to show if the invariant holds initially and when the Δ -plan finishes

Then step 2 (changing x_{ij} by $\pm \Delta$ if $C_{ij}^T < 0$ or $C_{ij}^T < 0$ for arcs in residual network) will establish invariant and steps 4, 5 maintains the invariant

Invariant at the beginning of $\Delta = \frac{\lfloor \log_2 u \rfloor}{2}$ plan:

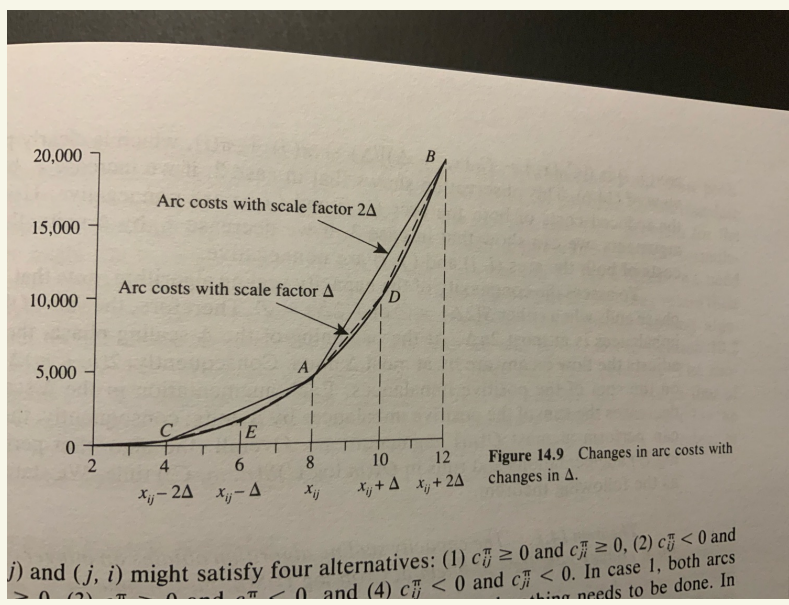
- $C_{ij}(x_{ij})$ is linearized into at most one segment (0 segments if $u_{ij} < \Delta$)
- let $\alpha = C_{ij}^1$ (slope of first section as discretization)
 - if we set $x_{ij} \leftarrow 0$ then $ij \in N(x, \Delta)$ with cost α and jc is not in $N(x, \Delta)$
 - if we set $x_{ij} \leftarrow \Delta$ then $jc \in N(x, \Delta)$ with cost $-\alpha$ and ij is not in $N(x, \Delta)$
- The algorithm assigns x_{ij} value 0 or Δ s.t. the resulting arc in $N(x)$ has cost ≥ 0

• We have $\pi \equiv 0$ initially so all arcs of $N(x, \Delta)$ have reduced cost ≥ 0 when the first phase starts.

So the invariant holds.

General Δ -phen: assuming invariant holds when the 2Δ -phen finishes.

Possible problem: in the 2Δ -phen $c_{ij}(x_{ij})$ is linearized into segments of length 2Δ and in the Δ -phen c_{ij} into segments of length $\Delta \Rightarrow$ arc costs change so we risk having arcs with negative cost in $N(x, \Delta)$:



In $N(x, 2\Delta)$:
 c_{ij} cost = slope of AB
 j_i cost = -slope of AC

In $N(x, \Delta)$:
 c_{ij} cost = slope of AD
 j_i cost = -slope of AE
 so cost $c_{ij} \nearrow$ and
 cost $j_i \searrow$

i) and (j, i) might satisfy four alternatives: (1) $c_{ij}^\pi \geq 0$ and $c_{ji}^\pi \geq 0$, (2) $c_{ij}^\pi < 0$ and $c_{ji}^\pi \geq 0$, (3) $c_{ij}^\pi \geq 0$ and $c_{ji}^\pi < 0$, and (4) $c_{ij}^\pi < 0$ and $c_{ji}^\pi < 0$. In case 1, both arcs are in the tree. In case 2, i needs to be done. In case 3, j needs to be done. In case 4, i and j need to be done.

Recall: in Δ -phen the reduced cost of c_{ij} is
 $c_{ij}^\pi = (c_{ij}(x_{ij+\Delta}) - c_{ij}(x_{ij})) / \Delta - \pi(i) + \pi(j)$

Then are 4 possible combinations of the signs
of c_{ij}^π and c_{ji}^π in $N(x, \Delta)$ when $i, j \in N(x)$

$$\begin{array}{cccc} \text{(1)} & \text{(2)} & \text{(3)} & \text{(4)} \\ c_{ij}^\pi \geq 0, c_{ji}^\pi \geq 0, & c_{ij}^\pi < 0, c_{ji}^\pi \geq 0, & c_{ij}^\pi \geq 0, c_{ji}^\pi < 0, & c_{ij}^\pi < 0, c_{ji}^\pi < 0 \end{array}$$

Case (1): Nothing needed for the arc c_{ij}

Case (2): $c_{ij}^\pi < 0$ and $c_{ji}^\pi \geq 0$

We want to increase x_{ij} by Δ

$$\frac{[c_{ij}(x_{ij} + 2\Delta) - c_{ij}(x_{ij})]}{2\Delta} - \pi(i) + \pi(j) \geq 0 \quad (2\Delta\text{-phen})$$

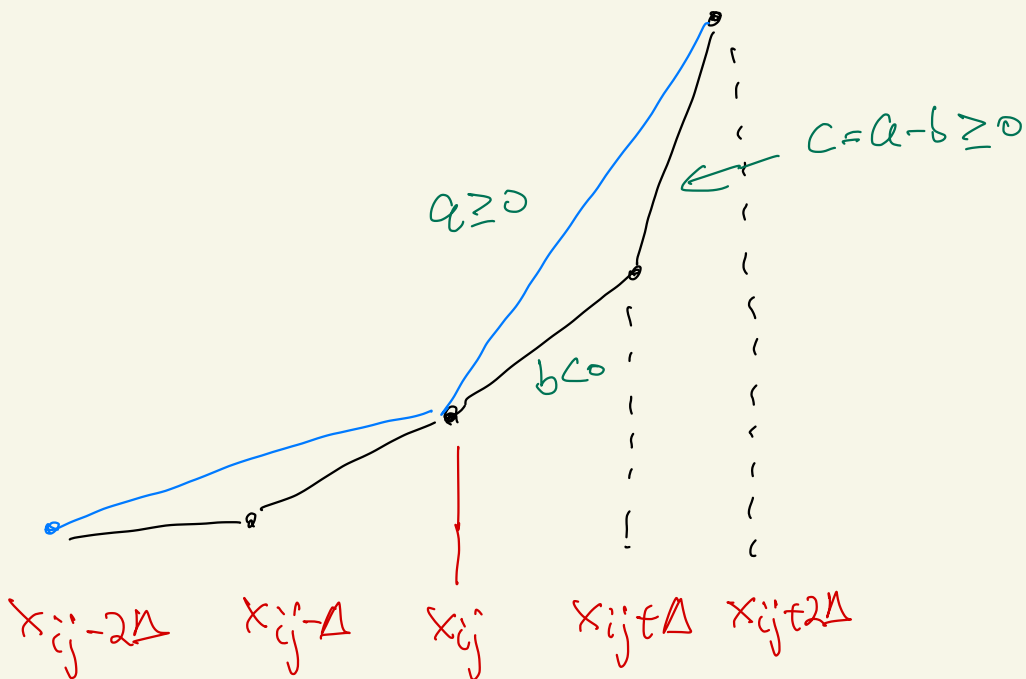
$$\Leftrightarrow c_{ij}(x_{ij} + 2\Delta) - c_{ij}(x_{ij}) - 2\Delta\pi(i) + 2\Delta\pi(j) \geq 0 \quad (\square)$$

$$c_{ij}(x_{ij} + 2\Delta) - c_{ij}(x_{ij} + \Delta) - \Delta\pi(i) + \Delta\pi(j)$$

$$= [c_{ij}(x_{ij} + 2\Delta) - c_{ij}(x_{ij}) - 2\Delta\pi(i) + 2\Delta\pi(j)]$$

$$- [c_{ij}(x_{ij} + \Delta) - c_{ij}(x_{ij}) - \Delta\pi(i) + \Delta\pi(j)]$$

$$\geq 0 \Leftrightarrow \underbrace{[\]}_{\geq 0} \geq 0 \text{ and } \underbrace{[\]}_{< 0} < 0 \text{ since } c_{ij}^\pi < 0$$



after increasing x_{ij} to $x_{ij} + \Delta$ the arc $j \in N(x)$ which is ok since

$$C_{j \in N(x)}^{\pi} = [C_{ij}(x_{ij}) - C_{ij}(x_{ij} + \Delta)] / \Delta - \pi(0) + \pi(\Delta)$$

$$= - [(C_{ij}(x_{ij} + \Delta) - C_{ij}(x_{ij})) / \Delta - \pi(\Delta) + \pi(0)]$$

$$= -C_{ij}^{\pi} \geq 0 \quad \text{so} \quad C_{ij}^{\pi} < 0$$

Case 3 $c_{ij}^{\pi} \geq 0, c_{ji}^{\pi} < 0$

very similar calculations:

we decrease x_{ij} by Δ to obtain

that both ij and ji have
non negative reduced cost

(ji may disappear if we set $x_{ij} = 0$.)

Case 4 $c_{ij}^{\pi} < 0, c_{ji}^{\pi} < 0$

impossible when $c_{ij}(x_{ij})$ is convex

or exercise 14.17

Complexity (similar analysis as for sec 10.2)

- When 2Δ -phase terminates we have

$$E = \sum_{b(i) > b_x(i)} (b(i) - b_x(i)) \leq 2n\Delta$$

a) $S(\Delta) = \emptyset$
o) $T(\Delta) = \emptyset$

- modifying x_{ij} at the beginning of the phase changes E by at most $2m\Delta$ so $E \leq 2(n+ms)\Delta$ when we start augmenting by exactly Δ -units along shortest paths

↓ $O(m)$ augmentations in Δ -phase

• $O(\log U)$ phases

- When the $2^{\lfloor \log_2 U \rfloor}$ phase begins, we

$$\text{have } \sum b_{ij} \leq mU \leq 2m2^{\lfloor \log_2 U \rfloor} = 2m\Delta$$

⇒ Complexity is $O(m \log_2 U \cdot \text{Dijkstra})$