

Lecture 20

- Plan: 1) ~~Finish LCIS algo.~~
2) Matroid intersection polytope
3) Start min-cost arborescence
after today, ~4 more lectures! ~1 matroids ~3 ellipsoid

Matroid intersection polytope

• Let $M_1 = (E, I_1)$ & $M_2 = (E, I_2)$
be two matroids, rank functions r_1, r_2 .

• Analogously to the matroid polytope, let

$$X = \left\{ \sum_{e \in E} x_e e : x_e \in [0, 1], \sum_{e \in S} x_e \leq r_1(S) \wedge r_2(S) \right\}$$

ie. $X \subseteq \mathbb{R}^E$ is set of indicators of common independent sets.

- Define the matroid intersection polytope

$P_{M_1, M_2} := \text{conv}(X)$
(can use to optimize linear functions over X).

- Main result: P_{M_1, M_2} is the intersection $P_{M_1} \cap P_{M_2}$ of the matroid polytopes P_{M_1}, P_{M_2} of M_1, M_2 .

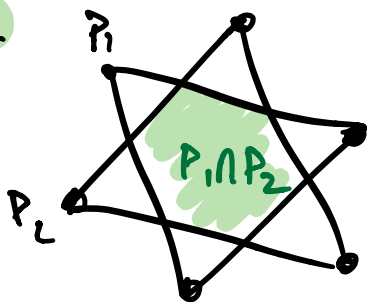
$$\Rightarrow \text{vertices}(P_{M_1}) \cap \text{vertices}(P_{M_2}) = \text{vertices}(P_{M_1} \cap P_{M_2})$$

- This is surprising! In general, for polytopes P_1, P_2

$$\text{verts}(P_1) \cap \text{verts}(P_2) \neq \text{verts}(P_1 \cap P_2)$$

$\rightarrow \text{vertices}(P_{M_i}) = \text{indicators of } I_i \Rightarrow \text{intersection is common indep sets.} = \text{vertices of } P_{M_1, M_2}$
if $P_{M_1, M_2} = P_{M_1} \cap P_{M_2}$, then same set of vertices.

e.g.



P_1, P_2 share no vertices but

$P_1 \cap P_2 \neq \emptyset$
(& hence has vertices).

• Intervals of inequalities?

• Recall matroid polytope:

for r rank function of M ,

$$P_M = \left\{ x \in \mathbb{R}^E : \begin{array}{l} x(S) \leq r(S) \quad \forall S \subseteq E \\ x_e \geq 0 \quad \forall e \in E \end{array} \right\}$$

$$:= \sum_{e \in S} x_e.$$

• $P_{M_1} \cap P_{M_2}$ has both sets of constraints, so

1981: can efficiently decide membership in P_{M_1, M_2} .

Theorem: Let $P = P_{M_1} \cap P_{M_2}$, i.e.

$$P = \left\{ x \in \mathbb{R}^E : \begin{array}{l} x(S) \leq r_1(S) \quad \forall S \subseteq E \\ x(S) \leq r_2(S) \quad \forall S \subseteq E \\ x_e \geq 0 \quad \forall e \in E \end{array} \right\}$$

Then

$$\boxed{P_{M_1, M_2} = P} \quad \text{i.e. } P_{M_1, M_2} = P_{M_1} \cap P_{M_2}.$$

Proof: Plan: similar to lecture 17, vertex proof for matroid polytope.

- Like second proof for matroid polytope, use vertex integrality
- Integrality^{of P} suffices by the usual logic:
- ▷ Clearly $\text{conv}(X) \subseteq P$, b/c $X \subseteq P_{M_1} \cap P_{M_2} = P$.

▷ on the other hand, if P integral then $P \subseteq \text{conv}(X)$ because integral points in P_{M_1}, P_{M_2} are indicators of indep sets in M_1, M_2 . \Rightarrow integral points in $P_{M_1} \cap P_{M_2} = P$ are common indep sets.

• Again, if $P = \{x: Ax \leq b, x \geq 0\}$, matrix A is not totally unimodular.

there are matrices

• But ~~submatrices~~ describing vertices will be T.U. (except).



$x^* \in \mathbb{R}^E$

$x^* = (1, 1, \sqrt{2})$?

Let $x^* \in \mathbb{R}^E$ be an extreme point of P .
 ↳ i.e. vertex

• We know x^* characterized by which inequalities are tight for it.

• For $i \in \{1, 2\}$, let $\vec{r}_i := \sum_{e \in S} x_e^* \mathbf{1}_e = x^* \cdot \mathbf{1}_S$.

$$T_i = \left\{ S \subseteq E : x^*(S) = r_i(S) \right\}$$

i.e. T_i sets of tight rank constraints in M_i .

• Let $J = \{e : x_e^* = 0\}$.

• Then x^* is unique solution to

$$F_1 \left[\begin{array}{l} x(S) = r_1(S) \quad \forall S \in T_1 \\ x(S) = r_2(S) \quad \forall S \in T_2 \\ x_e = 0 \quad \forall e \in J. \end{array} \right.$$

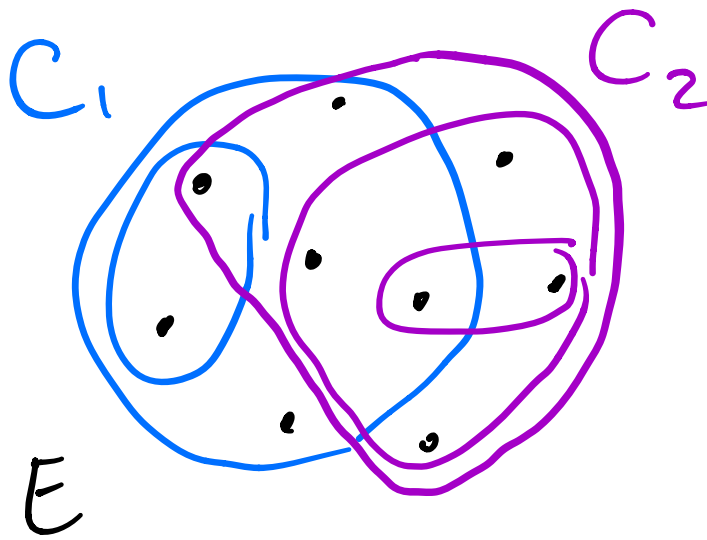
- That is, $\{x^*\}$ is the intersection of two faces F_1, F_2 in P_{M_1}, P_{M_2} .

$$F_i = \{x \in P_{M_i} : x(S) = r_i(S) \forall S \in T_i, x_e = 0 \forall e \in J\}.$$

- Recall from lec 17: T_i can be replaced by a chain C_i without changing F_i .
 $\exists C_1, C_2$ chains s.t.

$$F_i = \{x \in P_{M_i} : x(S) = r_i(S) \forall S \in C_i, x_e = 0 \forall e \in J\}.$$

e.g.



• Thus, assume x^* is solution to

$$x(s) = r_1(s) \quad \forall s \in C_1$$

$$x(s) = r_2(s) \quad \forall s \in C_2$$

$$x_e = 0 \quad \forall e \in J.$$

• This is $Ax=b$ for $b \in \mathbb{R}$

Claim: A T.U. $\Rightarrow x^*$ integral.

• Why? Rows of A are ± 1 s of S in chain C_1 or C_2 .

e.g.

can permute rows of A so it looks like this. doesn't change T.U.

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \left. \begin{array}{l} \} C_1 \\ \} C_2 \end{array} \right\}$$

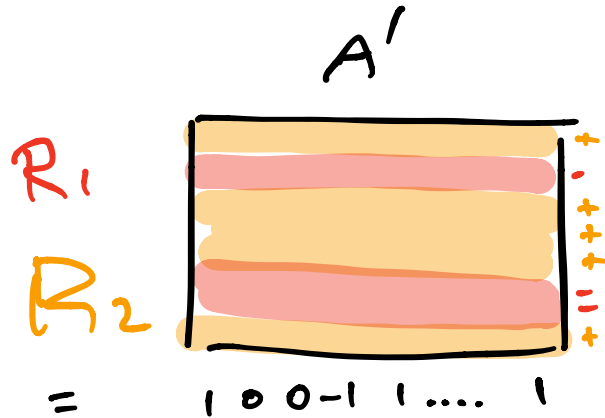
• We use discrepancy to prove. Theorem 3.14 in polyhedral notes

• Recall: A T.U. $\Leftrightarrow \forall$ submatrices

A' of A , \exists partition R_1, R_2 of rows of A'
 \uparrow
 a_i

$\sum_{i \in R_1} a_i - \sum_{i \in R_2} a_i$ has $\{-1, 0, +1\}$.

e.g.



- Consider submatrix A' of A corresponds to subchains C'_1, C'_2 (same form as A)

- Assign R_1, R_2 as follows:

▷ Assign target elmnt of C'_1 to R_1 , then alternately assign remaining elts of C'_1 to R_2, R_1

e.s.

$$C'_1 \begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{bmatrix} \begin{matrix} + \\ - \\ + \\ - \end{matrix} \begin{matrix} R_1 \\ R_2 \end{matrix}$$

sum has entries in $\{0, 1\}$.

▷ For C'_2 , assign oppositely.

e.g.

$$C'_2 \begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{bmatrix} \begin{matrix} - \\ + \\ - \\ + \end{matrix} \begin{matrix} R_2 \\ R_1 \end{matrix}$$

sum has entries in $\{0, -1\}$.

- Overall, sum has entries in $\{-1, 0, 1\}$.
- completes the proof. □.

Matroid intersection optimization

- Given a cost function $c: E \rightarrow \mathbb{R}$
can we efficiently compute

$$\max_{S \in \mathcal{I}_1 \cap \mathcal{I}_2} c(S) := \sum_{e \in S} c(e) = c \cdot \mathbf{1}_S.$$

equiv: optimize $c^T x$ over $x \in P_{M_1, M_2}$.

- For just one matroid: greedy alg works.
- For $c = \mathbf{1}$: just L.C.I.S.
(1, ..., 1) $|E|$ times
- For perfect matching: Hungarian algo.
e.g. min-cost P.M.
- can also compute min cost L.C.I.S.
Exercise: equiv. to max cost indep
for $c' = K - c$ for K large.

• In general, YES, can efficiently compute.

▷ ellipsoid

▷ complicated primal-dual algs.

→ strongly poly. time

≠ steps indep of C

if arithmetic is unit cost.

• ~~Today:~~

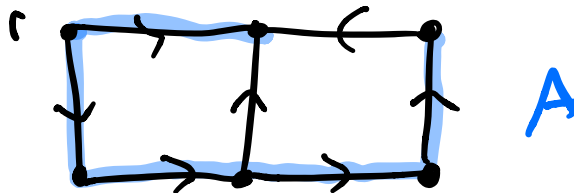
simplex primal dual alg. for,



Min-Cost arborescence

• Recall: given directed graph D & vertex r , arborescence A is a spanning tree in D directed away from r .

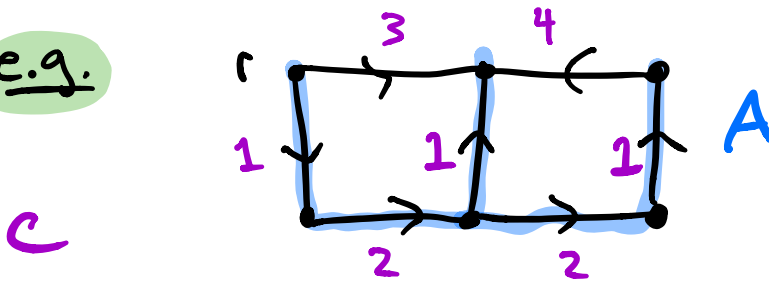
e.g.



- min-cost arborescence:

$$\min_{A \text{ arborescence}} \sum_{e \in A} c(e) = c(A)$$

e.g.



- e.g. edges = roads to be fixed
 r = distribution center
 Cost = expense of fixing road.

First, I.P. formulation:

assume c nonnegative.

$$\text{OPT} = \min_{x \in \mathbb{R}^E} \sum_{e \in E} c_e x_e$$

think $x = \mathbb{1}_A$
 for A arborescence
 $\sum c_e x_e = c(A)$

subject to

$x = \mathbb{1}_A$, where
 A has no cuts.



indegrees = 1
except s .

x is an
indicator.

$$\sum_{e \in \delta^-(s)} x_e \geq 1 \quad \forall s \in V-r$$

$$\sum_{e \in \delta^-(v)} x_e = 1 \quad \forall v \in V-s$$

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

• Check: only solutions are $\mathbb{1}_A$
where A is an arborescence.

in particular, all arbos satisfy constraints.

• Miraculously, we'll show even
w/out integrality constraint &
indegree constraint, there's still
an optimal solution that's an
arborescence.

- I.e. the following LP has optimizer $\mathbb{1}_A$ st. A is an arborescence.

$$\text{LP} = \min_{x \in \mathbb{R}^E} \sum_{e \in E} c(e) x_e$$

$$\text{subject to } \sum_{e \in \delta(S)} x_e \geq 1 \quad \forall S \subseteq V-r.$$

$$\text{(primal)} \quad x_e \geq 0 \quad \forall e \in E$$

(note $\text{LP} \leq \text{OPT}$ b/c LP has fewer constraints.)

(topo in pre-lecture!)

- Dual LP is "symmetric version"

$$LP = \max \sum_{S \in V-r} y_S$$

subject to

$$\sum_{S: e \in \delta(S)} y_S \leq c(e) \quad \forall e \in E$$

(dual)

$$y_S \geq 0 \quad \forall S \in V-r.$$

• Algorithm sketch: construct

▷ arb. A ,

▷ dual. feas. y

satisfying complementary slackness

Then $c(A) = LP$, but $LP \leq OPT$

$$c(A) \leq OPT \Rightarrow \boxed{c(A) = OPT.}$$

- Complementary slackness for $x \in \mathcal{I}_A$, y says:

a.) $y_s > 0 \Rightarrow |A \cap \delta^-(s)| = 1$

b.) $e \in A \Rightarrow \sum_{s: e \in \delta^-(s)} y_s = c(e).$

- Two phases of algorithm:

1) Construct

▷ dual feas y

▷ set F of edges s.t.

every vertex of V is

reachable from r in F

F might not be an arborescence.

& $y, A = F$ satisfies (b).

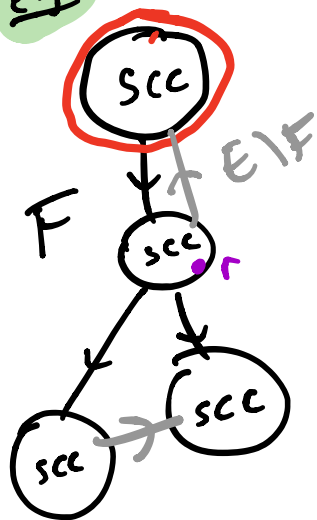
2) Remove unnecessary edges from F , get arborescence which satisfies both (a) & (b).

Phase 1 Initialize $F = \emptyset, y = 0$
counter $k = 1$

▷ While not everything reachable from r in F

▷ select $S \subseteq V - r$

eg.

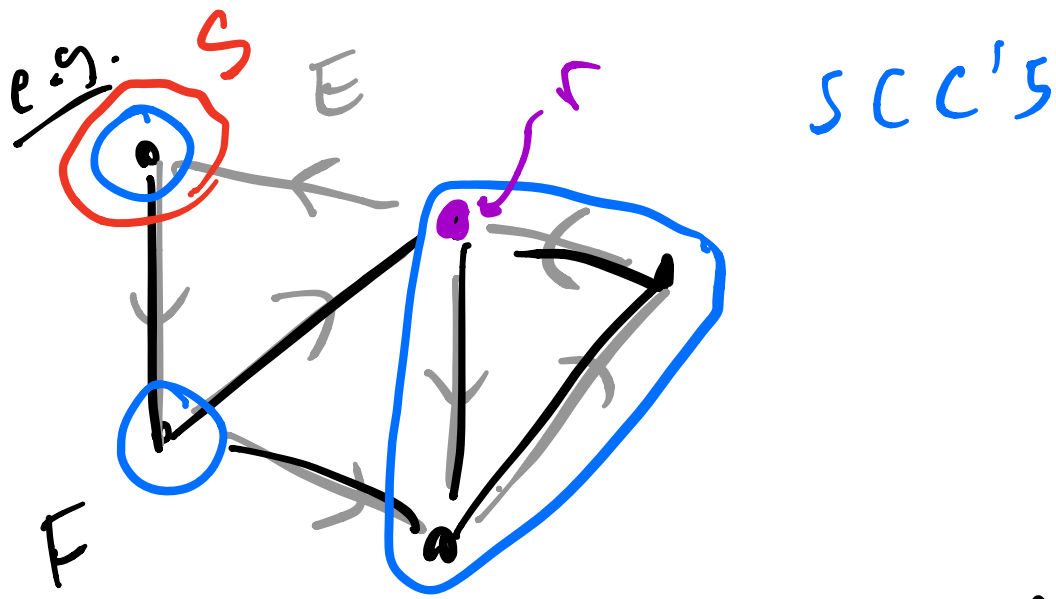


i) F strongly connected in S
(every vertex can reach every other using only edges contained entirely in S)

ii) $F \cap \delta^-(s) = \emptyset$

S is a "source" in decomp. of F into S.C.C.'s.

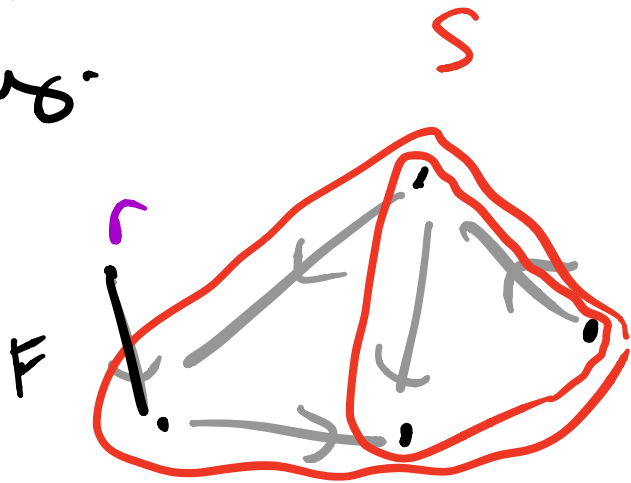
(digraph has decomp where if S.C.C.'s are contracted, left with DAG).



S is a subset of vertices,

F subset of edges.

does $S \subseteq$ vertices "touched by" F ?
 not necessarily.
 initially



▷ increase y_s until new inequality

$$\sum y_s \leq c(e_k)$$

$$S: e_k \in \delta^-(s)$$

becomes an equality.

note $e_k \notin F$
 y_c
 $F \cap \delta^-(s) = \emptyset$

(y remain dual feas,
b/c it was before)

$$▷ F \leftarrow F + e_k, k \leftarrow k+1$$

new F, y don't violate (b)
because e_k is tight.

▷ Return F, y satisfying (b),
& everything reachable from \uparrow in F .

Phase 2: eliminate as many edges as we can in reverse order they were added.

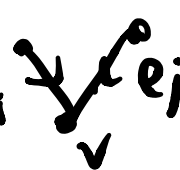
▷ For $i = k \dots 1$:

▷ If $F - e_i$ contains a directed path from r to every vertex,
 $F \leftarrow F - e_i$.

▷ Return $A = F$

Claim 1: A is an arborescence

Pf: • We'll show $|A| = |V| - 1$ & $d^-(v) = 1$
 $\forall v \in V - r$

- If indegree < 1 for $v \neq r$, contradicts reachability in A
- if $|A| > |V| - 1$ then collision $e_i \searrow \swarrow e_j$


- Suppose $i < j \Rightarrow$ in reverse delete, would have removed e_j . v/c our vertex reached through e_j is reachable through e_i . \square

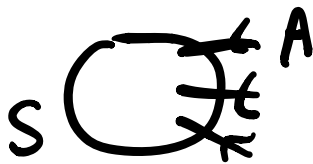
finally:

Claim 2: Condition (a) of complementary slackness holds.

$$a.) \gamma_s > 0 \Rightarrow |A \cap \delta^-(s)| = 1.$$

Pf: Assume not $\exists S$ s.t.

$$\gamma_s > 0 \text{ \& } |A \cap \delta^-(s)| > 1$$



- S was chosen at some step l of phase 1 when we added e_l to F .

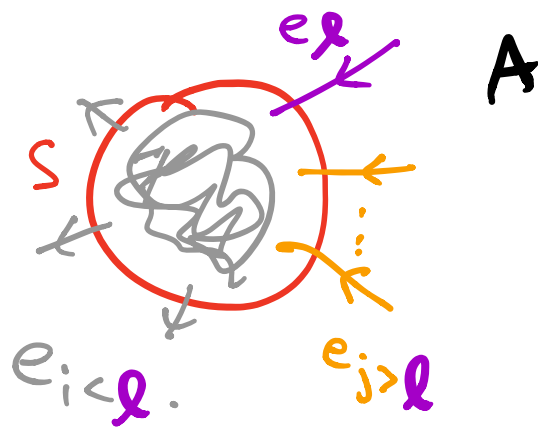
- F had no other edges in $\delta^-(S)$ when e_ℓ was added.

(by construction)

\Rightarrow all edges of $A \cap \delta^-(S)$ are e_j for $j > \ell$.

- When S chosen, F strongly connected within S

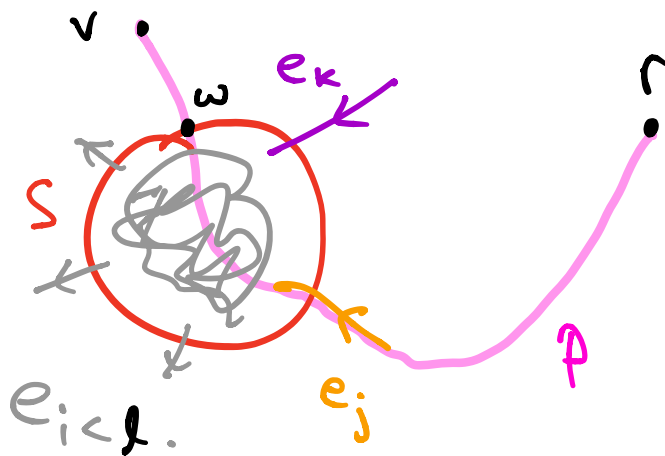
\Rightarrow S strongly connected using only e_i $i < \ell$.



- Subclaim: All $e_j, j > k$ should have been removed in Phase 2.

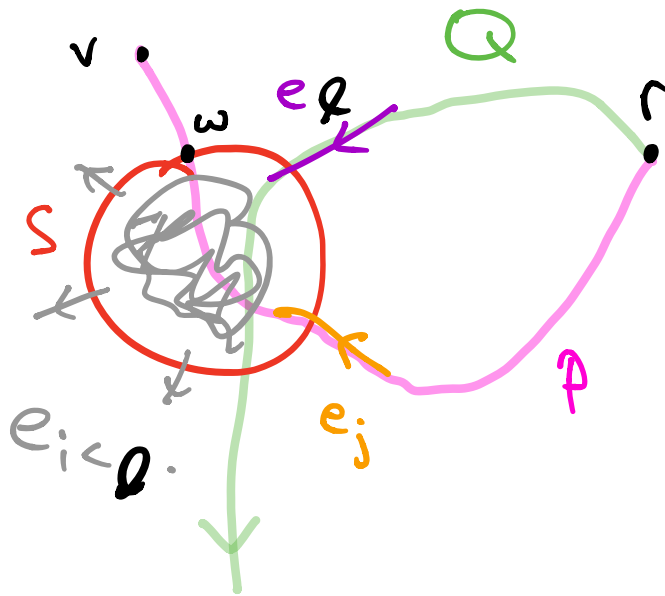
Why? suppose e_j necessary to visit some vertex v

- let P $r \rightarrow v$ path using e_j
- let w last vertex in S on P



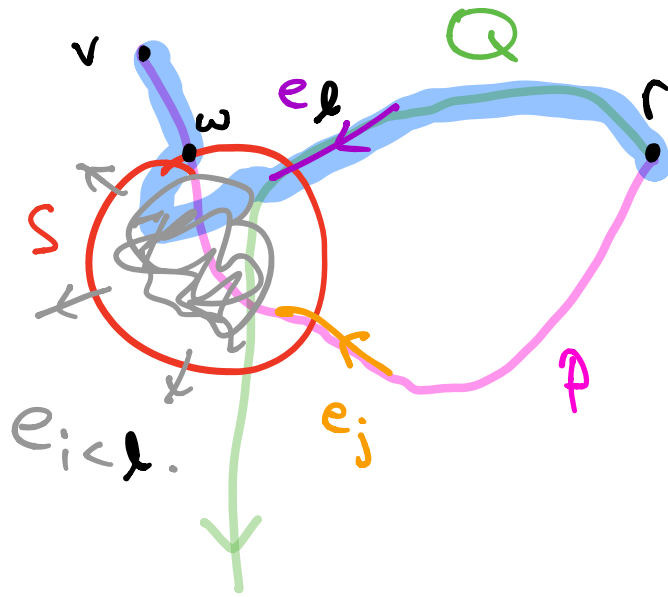
note: P first enters S thru e_j , else could shortcut e_j because S stay conn.

- Because e_l is necessary at Step 2 of Phase 2, there must be another path Q through e_l .



similarly: Q must enter S first than e_l .

- Can use Q to shortcut e_j ; thus e_j not necessary.



□.