## Repetition: Primal Dual for Set Cover

## Primal Relaxation:

| min |  | $\sum_{i=1}^{k} w_{i} x_{i}$ |  |
| :---: | ---: | ---: | :--- |
| s.t. | $\forall u \in U$ | $\sum_{i: u \in S_{i}} x_{i} \geq$ | 1 |
|  | $\forall i \in\{1, \ldots, k\}$ | $x_{i} \geq$ | 0 |
|  |  |  |  |

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| s.t. | $\forall u \in U$ | $\sum_{i: u \in S_{i}} x_{i} \geq 1$ |  |
|  | $\forall i \in\{1, \ldots, k\}$ | $x_{i} \geq 0$ |  |

Dual Formulation:

$$
\begin{array}{|ccr|}
\hline \max & & \sum_{u \in U} y_{u} \\
\text { s.t. } & \forall i \in\{1, \ldots, k\} & \\
& \sum_{u: u \in S_{i}} y_{u} & \leq w_{i} \\
y_{u} & \geq 0
\end{array}
$$

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- Increase dual variable $y_{e}$ until a dual constraint becomes tight (maybe increase by 0 !).


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- While $x$ not feasible
- Identify an element $e$ that is not covered in current primal integral solution.
- Increase dual variable $y_{e}$ until a dual constraint becomes tight (maybe increase by 0 !).
- If this is the constraint for set $S_{j}$ set $x_{j}=1$ (add this set to your solution).


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- Hence our cost is

$$
\begin{aligned}
\sum_{j} w_{j} x_{j}=\sum_{j} \sum_{e \in S_{j}} y_{e} & =\sum_{e}\left|\left\{j: e \in S_{j}\right\}\right| \cdot y_{e} \\
& \leq f \cdot \sum_{e} y_{e} \leq f \cdot \mathrm{OPT}
\end{aligned}
$$

Note that the constructed pair of primal and dual solution fulfills primal slackness conditions.

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$$

If we would also fulfill dual slackness conditions

$$
y_{e}>0 \Rightarrow \sum_{j: e \in S_{j}} x_{j}=1
$$

then the solution would be optimal!!!

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$$

This is sufficient to show that the solution is an $f$-approximation.

Suppose we have a primal/dual pair

\[

\]

$$
\begin{array}{|crrll|}
\hline \max & & \sum_{i} b_{i} y_{i} & \\
\text { s.t. } & \forall j & \sum_{i} a_{i j} y_{i} & \leq c_{j} \\
& \forall i & y_{i} & \geq 0 \\
\hline
\end{array}
$$

Suppose we have a primal/dual pair

| $\min$ |  | $\sum_{j} c_{j} x_{j}$ |  |  |
| ---: | ---: | ---: | ---: | :--- |
| s.t. | $\forall i$ | $\sum_{j:} a_{i j} x_{j}$ | $\geq$ | $b_{i}$ |
|  | $\forall j$ | $x_{j}$ | $\geq$ | 0 |
|  |  |  |  |  |


| $\max$ |  | $\sum_{i} b_{i} y_{i}$ |  |  |
| :---: | ---: | ---: | :--- | :--- |
| s.t. | $\forall j$ | $\sum_{i} a_{i j} y_{i}$ | $\leq$ | $c_{j}$ |
|  | $\forall i$ | $y_{i}$ | $\geq 0$ |  |

and solutions that fulfill approximate slackness conditions:

$$
\begin{aligned}
& x_{j}>0 \Rightarrow \sum_{i} a_{i j} y_{i} \geq \frac{1}{\alpha} c_{j} \\
& y_{i}>0 \Rightarrow \sum_{j} a_{i j} x_{j} \leq \beta b_{i}
\end{aligned}
$$

Then

$$
\sum_{j} c_{j} x_{j}
$$

Then


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$$
\begin{aligned}
& \sum_{j} c_{j} x_{j}
\end{aligned} \leq \alpha \sum_{j}\left(\sum_{i} a_{i j} y_{i}\right) x_{j}
$$

Then

$$
\begin{aligned}
& \sum_{j} c_{j} x_{j} \leq \alpha \sum_{j}\left(\sum_{i} a_{i j} y_{i}\right) x_{j} \\
& \uparrow \\
& \text { primal cost }=\alpha \sum_{i}\left(\sum_{j} a_{i j} x_{j}\right) y_{i}
\end{aligned}
$$

Then

$$
\begin{aligned}
\sum_{j} c_{j} x_{j} & \leq \alpha \sum_{j}\left(\sum_{i} a_{i j} y_{i}\right) x_{j} \\
{ } } & =\alpha \sum_{i}\left(\sum_{j} a_{i j} x_{j}\right) y_{i} \\
& \leq \alpha \beta \cdot \sum_{i} b_{i} y_{i}
\end{aligned}
$$

Then

$$
\begin{aligned}
& \sum_{j} c_{j} x_{j} \leq \alpha \sum_{j}\left(\sum_{i} a_{i j} y_{i}\right) x_{j} \\
& \frac{\text { primal cost }^{l}}{}= \alpha \sum_{i}\left(\sum_{j} a_{i j} x_{j}\right) y_{i} \\
& \leq \alpha \beta \cdot \sum_{i} b_{i} y_{i} \\
& \uparrow
\end{aligned}
$$

## Feedback Vertex Set for Undirected Graphs

- Given a graph $G=(V, E)$ and non-negative weights $w_{v} \geq 0$ for vertex $v \in V$.


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- Given a graph $G=(V, E)$ and non-negative weights $w_{v} \geq 0$ for vertex $v \in V$.
- Choose a minimum cost subset of vertices s.t. every cycle contains at least one vertex.

We can encode this as an instance of Set Cover

- Each vertex can be viewed as a set that contains some cycles.

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- However, this encoding gives a Set Cover instance of non-polynomial size.

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- Each vertex can be viewed as a set that contains some cycles.
- However, this encoding gives a Set Cover instance of non-polynomial size.
- The $O(\log n)$-approximation for Set Cover does not help us to get a good solution.

Let $\mathbb{C}$ denote the set of all cycles (where a cycle is identified by its set of vertices)

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## Primal Relaxation:

| $\min$ |  | $\sum_{v} w_{v} x_{v}$ |  |
| ---: | ---: | ---: | :--- | :--- |
| s.t. | $\forall C \in \mathbb{C}$ | $\sum_{v \in C} x_{v}$ | $\geq 1$ |
|  | $\forall v$ | $x_{v}$ | $\geq 0$ |

## Dual Formulation:

\[

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If we perform the previous dual technique for Set Cover we get the following:

- Start with $x=0$ and $y=0$

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If we perform the previous dual technique for Set Cover we get the following:

- Start with $x=0$ and $y=0$
- While there is a cycle $C$ that is not covered (does not contain a chosen vertex).
- Increase $y_{C}$ until dual constraint for some vertex $v$ becomes tight.
- set $x_{v}=1$.

Then

$$
\sum_{v} w_{v} x_{v}
$$

Then

$$
\sum_{v} w_{v} x_{v}=\sum_{v} \sum_{C: v \in C} y_{C} x_{v}
$$

Then

$$
\begin{aligned}
\sum_{v} w_{v} x_{v} & =\sum_{v} \sum_{C: v \in C} y_{C} x_{v} \\
& =\sum_{v \in S} \sum_{C: v \in C} y_{C}
\end{aligned}
$$

where $S$ is the set of vertices we choose.

Then

$$
\begin{aligned}
\sum_{v} w_{v} x_{v} & =\sum_{v} \sum_{C: v \in C} y_{C} x_{v} \\
& =\sum_{v \in S} \sum_{C: v \in C} y_{C} \\
& =\sum_{C}|S \cap C| \cdot y_{C}
\end{aligned}
$$

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& =\sum_{v \in S} \sum_{C: v \in C} y_{C} \\
& =\sum_{C}|S \cap C| \cdot y_{C}
\end{aligned}
$$

where $S$ is the set of vertices we choose.
If every cycle is short we get a good approximation ratio, but this is unrealistic.

```
Algorithm 1 FeedbackVertexSet
    1: \(y \leftarrow 0\)
    2: \(x \leftarrow 0\)
    3: while exists cycle \(C\) in \(G\) do
    4: \(\quad\) increase \(y_{C}\) until there is \(v \in C\) s.t. \(\sum_{C: v \in C} y_{C}=w_{v}\)
    5: \(\quad x_{v}=1\)
    6: \(\quad\) remove \(v\) from \(G\)
    7: repeatedly remove vertices of degree 1 from \(G\)
```


## Idea:

Always choose a short cycle that is not covered. If we always find a cycle of length at most $\alpha$ we get an $\alpha$-approximation.

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Always choose a short cycle that is not covered. If we always find a cycle of length at most $\alpha$ we get an $\alpha$-approximation.

Observation:
For any path $P$ of vertices of degree 2 in $G$ the algorithm chooses at most one vertex from $P$.

## Observation:

If we always choose a cycle for which the number of vertices of degree at least 3 is at most $\alpha$ we get a $2 \alpha$-approximation.

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If we always choose a cycle for which the number of vertices of degree at least 3 is at most $\alpha$ we get a $2 \alpha$-approximation.

## Theorem 92

In any graph with no vertices of degree 1, there always exists a cycle that has at most $\mathcal{O}(\log n)$ vertices of degree 3 or more. We can find such a cycle in linear time.

This means we have

$$
y_{C}>0 \Rightarrow|S \cap C| \leq \mathcal{O}(\log n)
$$

## Primal Dual for Shortest Path

Given a graph $G=(V, E)$ with two nodes $s, t \in V$ and edge-weights $c: E \rightarrow \mathbb{R}^{+}$find a shortest path between $s$ and $t$ w.r.t. edge-weights $c$.

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Here $\delta(S)$ denotes the set of edges with exactly one end-point in $S$, and $S=\{S \subseteq V: s \in S, t \notin S\}$.

## Primal Dual for Shortest Path

The Dual:

$\left.$| $\max$ | $\sum_{S} y_{S}$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| s.t. | $\forall e \in E$ | $\sum_{S: e \in \delta(S)} y_{S}$ |  |  |$\leq c(e) \right\rvert\,$

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| s.t. | $\forall e \in E$ | $\sum_{S: e \in \delta(S)} y_{S}$ | $\leq c(e)$ |
|  | $\forall S \in S$ | $y_{S}$ | $\geq 0$ |

Here $\delta(S)$ denotes the set of edges with exactly one end-point in
$S$, and $S=\{S \subseteq V: s \in S, t \notin S\}$.

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Each set can have its own moat but all moats must be disjoint.
An edge cannot be shorter than all the moats that it has to cross.

```
Algorithm 1 PrimalDualShortestPath
    1: \(y \leftarrow 0\)
    2: \(F \leftarrow \varnothing\)
    3: while there is no \(s-t\) path in \((V, F)\) do
    4: Let \(C\) be the connected component of \((V, F)\) con-
        taining \(s\)
    5: Increase \(y_{C}\) until there is an edge \(e^{\prime} \in \delta(C)\) such
        that \(\sum_{S: e^{\prime} \in \delta(S)} y_{S}=c\left(e^{\prime}\right)\).
    6: \(\quad F \leftarrow F \cup\left\{e^{\prime}\right\}\)
    7: Let \(P\) be an \(s\) - \(t\) path in \((V, F)\)
    8: return \(P\)
```


## Lemma 93

At each point in time the set $F$ forms a tree.

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## Proof:

- In each iteration we take the current connected component from $(V, F)$ that contains $s$ (call this component $C$ ) and add some edge from $\delta(C)$ to $F$.


## Lemma 93

At each point in time the set $F$ forms a tree.

## Proof:

- In each iteration we take the current connected component from $(V, F)$ that contains $s$ (call this component $C$ ) and add some edge from $\delta(C)$ to $F$.
- Since, at most one end-point of the new edge is in $C$ the edge cannot close a cycle.

$$
\sum_{e \in P} c(e)
$$

$$
\sum_{e \in P} c(e)=\sum_{e \in P} \sum_{S: e \in \delta(S)} y_{S}
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\sum_{e \in P} c(e) & =\sum_{e \in P} \sum_{S: e \in \delta(S)} y_{S} \\
& =\sum_{S: s \in S, t \notin S}|P \cap \delta(S)| \cdot y_{S} .
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If we can show that $y_{S}>0$ implies $|P \cap \delta(S)|=1$ gives

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\sum_{e \in P} c(e)=\sum_{S} y_{S} \leq \mathrm{OPT}
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by weak duality.

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$$

If we can show that $y_{S}>0$ implies $|P \cap \delta(S)|=1$ gives

$$
\sum_{e \in P} c(e)=\sum_{S} y_{S} \leq \mathrm{OPT}
$$

by weak duality.
Hence, we find a shortest path.

If $\delta(S)$ contains two edges from $P$ then there must exist a subpath $P^{\prime}$ of $P$ that starts and ends with a vertex from $S$ (and all interior vertices are not in $S$ ).

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When we increased $y_{S}, S$ was a connected component of the set of edges $F^{\prime}$ that we had chosen till this point.

If $\delta(S)$ contains two edges from $P$ then there must exist a subpath $P^{\prime}$ of $P$ that starts and ends with a vertex from $S$ (and all interior vertices are not in $S$ ).

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$F^{\prime} \cup P^{\prime}$ contains a cycle. Hence, also the final set of edges contains a cycle.

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$F^{\prime} \cup P^{\prime}$ contains a cycle. Hence, also the final set of edges contains a cycle.

This is a contradiction.

## Steiner Forest Problem:

Given a graph $G=(V, E)$, together with source-target pairs $s_{i}, t_{i}$, $i=1, \ldots, k$, and a cost function $c: E \rightarrow \mathbb{R}^{+}$on the edges. Find a subset $F \subseteq E$ of the edges such that for every $i \in\{1, \ldots, k\}$ there is a path between $s_{i}$ and $t_{i}$ only using edges in $F$.

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| min |  | $\sum_{e} c(e) x_{e}$ |  |
| :---: | ---: | :--- | :--- |
| s.t. | $\forall S \subseteq V: S \in S_{i}$ for some $i$ | $\sum_{e \in \delta(S)} x_{e}$ | $\geq 1$ |
|  | $\forall e \in E$ | $x_{e}$ | $\in\{0,1\}$ |

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|  | $\forall e \in E$ | $x_{e}$ | $\in\{0,1\}$ |

Here $S_{i}$ contains all sets $S$ such that $s_{i} \in S$ and $t_{i} \notin S$.

| $\max$ |  |  |
| ---: | ---: | ---: | ---: |
| s.t. | $\forall e \in E \quad$  <br> $S: \exists i$ s.t. $S \in S_{i}$ $y_{S}$ <br> $\sum_{S: e \in \delta(S)} y_{S}$ $\leq c(e)$ <br>  $y_{S} \geq 0$ |  |
|  |  |  |

The difference to the dual of the shortest path problem is that we have many more variables (sets for which we can generate a moat of non-zero width).

```
Algorithm 1 FirstTry
    1: \(y \leftarrow 0\)
    2: \(F \leftarrow \varnothing\)
    3: while not all \(s_{i}-t_{i}\) pairs connected in \(F\) do
    4: \(\quad\) Let \(C\) be some connected component of \((V, F)\) such
    that \(\left|C \cap\left\{s_{i}, t_{i}\right\}\right|=1\) for some \(i\).
    5: Increase \(y_{C}\) until there is an edge \(e^{\prime} \in \delta(C)\) s.t.
    \(\sum_{S \in S_{i}: e^{\prime} \in \delta(S)} y_{S}=c_{e^{\prime}}\)
    6: \(\quad F \leftarrow F \cup\left\{e^{\prime}\right\}\)
    7: return \(\bigcup_{i} P_{i}\)
```

$$
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\sum_{e \in F} c(e)=\sum_{e \in F} \sum_{S: e \in \delta(S)} y_{S}
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\sum_{e \in F} c(e)=\sum_{e \in F} \sum_{S: e \in \delta(S)} y_{S}=\sum_{S}|\delta(S) \cap F| \cdot y_{S} .
$$

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$$

If we show that $y_{S}>0$ implies that $|\delta(S) \cap F| \leq \alpha$ we are in good shape.

However, this is not true:

- Take a complete graph on $k+1$ vertices $v_{0}, v_{1}, \ldots, v_{k}$.

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- The first component $C$ could be $\left\{v_{0}\right\}$.

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If we show that $y_{S}>0$ implies that $|\delta(S) \cap F| \leq \alpha$ we are in good shape.

However, this is not true:

- Take a complete graph on $k+1$ vertices $v_{0}, v_{1}, \ldots, v_{k}$.
- The $i$-th pair is $v_{0}-v_{i}$.
- The first component $C$ could be $\left\{v_{0}\right\}$.
- We only set $y_{\left\{v_{0}\right\}}=1$. All other dual variables stay 0 .

$$
\sum_{e \in F} c(e)=\sum_{e \in F} \sum_{S: e \in \delta(S)} y_{S}=\sum_{S}|\delta(S) \cap F| \cdot y_{S} .
$$

If we show that $y_{S}>0$ implies that $|\delta(S) \cap F| \leq \alpha$ we are in good shape.

However, this is not true:

- Take a complete graph on $k+1$ vertices $v_{0}, v_{1}, \ldots, v_{k}$.
- The $i$-th pair is $v_{0}-v_{i}$.
- The first component $C$ could be $\left\{v_{0}\right\}$.
- We only set $y_{\left\{v_{0}\right\}}=1$. All other dual variables stay 0 .
- The final set $F$ contains all edges $\left\{v_{0}, v_{i}\right\}, i=1, \ldots, k$.

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- We only set $y_{\left\{v_{0}\right\}}=1$. All other dual variables stay 0 .
- The final set $F$ contains all edges $\left\{v_{0}, v_{i}\right\}, i=1, \ldots, k$.
- $y_{\left\{v_{0}\right\}}>0$ but $\left|\delta\left(\left\{v_{0}\right\}\right) \cap F\right|=k$.

```
Algorithm 1 SecondTry
    1: \(y \leftarrow 0 ; F \leftarrow \varnothing ; \ell \leftarrow 0\)
    2: while not all \(s_{i}-t_{i}\) pairs connected in \(F\) do
    3: \(\quad \ell \leftarrow \ell+1\)
    4: Let \(\mathbb{C}\) be set of all connected components \(C\) of \((V, F)\)
        such that \(\left|C \cap\left\{s_{i}, t_{i}\right\}\right|=1\) for some \(i\).
    5: \(\quad\) Increase \(y_{C}\) for all \(C \in \mathbb{C}\) uniformly until for some edge
        \(e_{\ell} \in \delta\left(C^{\prime}\right), C^{\prime} \in \mathbb{C}\) s.t. \(\sum_{s: e_{\ell} \in \delta(S)} y_{S}=c_{e_{\ell}}\)
    6: \(\quad F \leftarrow F \cup\left\{e_{\ell}\right\}\)
    7: \(F^{\prime} \leftarrow F\)
    8: for \(k \leftarrow \ell\) downto 1 do // reverse deletion
    9: \(\quad\) if \(F^{\prime}-e_{k}\) is feasible solution then
10: remove \(e_{k}\) from \(F^{\prime}\)
11: return \(F^{\prime}\)
```

The reverse deletion step is not strictly necessary this way. It would also be sufficient to simply delete all unnecessary edges in any order.

## Example

$$
\mathrm{o}_{S_{3}}
$$

$$
\circ_{S_{1}} \quad \circ_{S_{2}} \quad t_{2}^{\circ}
$$

${ }^{\circ} t_{1}$

- ${ }^{\circ} t_{3}$


## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Example



## Lemma 94

For any $\mathbb{C}$ in any iteration of the algorithm

$$
\sum_{C \in \mathbb{C}}\left|\delta(C) \cap F^{\prime}\right| \leq 2|\mathbb{C}|
$$

This means that the number of times a moat from $\mathbb{C}$ is crossed in the final solution is at most twice the number of moats.

Proof: later...

$$
\sum_{e \in F^{\prime}} c_{e}
$$

$$
\sum_{e \in F^{\prime}} c_{e}=\sum_{e \in F^{\prime}} \sum_{S: e \in \delta(S)} y_{S}
$$

$$
\sum_{e \in F^{\prime}} c_{e}=\sum_{e \in F^{\prime}} \sum_{S: e \in \delta(S)} y_{S}=\sum_{S}\left|F^{\prime} \cap \delta(S)\right| \cdot y_{S}
$$

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We want to show that

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\sum_{S}\left|F^{\prime} \cap \delta(S)\right| \cdot y_{S} \leq 2 \sum_{S} y_{S}
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- In the $i$-th iteration the increase of the left-hand side is

$$
\epsilon \sum_{C \in \mathbb{C}}\left|F^{\prime} \cap \delta(C)\right|
$$

and the increase of the right hand side is $2 \epsilon|\mathbb{C}|$.

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and the increase of the right hand side is $2 \epsilon|\mathbb{C}|$.

- Hence, by the previous lemma the inequality holds after the iteration if it holds in the beginning of the iteration.


## Lemma 95

For any set of connected components $\mathbb{C}$ in any iteration of the algorithm

$$
\sum_{C \in \mathscr{C}}\left|\delta(C) \cap F^{\prime}\right| \leq 2|\mathbb{C}|
$$

## Lemma 95

For any set of connected components $\mathbb{C}$ in any iteration of the algorithm

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## Proof:

- At any point during the algorithm the set of edges forms a forest (why?).


## Lemma 95

For any set of connected components $\mathbb{C}$ in any iteration of the algorithm

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## Proof:

- At any point during the algorithm the set of edges forms a forest (why?).
- Fix iteration $i$. Let $F_{i}$ be the set of edges in $F$ at the beginning of the iteration.


## Lemma 95

For any set of connected components $\mathbb{C}$ in any iteration of the algorithm

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## Proof:

- At any point during the algorithm the set of edges forms a forest (why?).
- Fix iteration $i$. Let $F_{i}$ be the set of edges in $F$ at the beginning of the iteration.
- Let $H=F^{\prime}-F_{i}$.


## Lemma 95

For any set of connected components $\mathbb{C}$ in any iteration of the algorithm

$$
\sum_{C \in \mathbb{C}}\left|\delta(C) \cap F^{\prime}\right| \leq 2|\mathbb{C}|
$$

## Proof:

- At any point during the algorithm the set of edges forms a forest (why?).
- Fix iteration $i$. Let $F_{i}$ be the set of edges in $F$ at the beginning of the iteration.
- Let $H=F^{\prime}-F_{i}$.
- All edges in $H$ are necessary for the solution.
- Contract all edges in $F_{i}$ into single vertices $V^{\prime}$.
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- We can consider the forest $H$ on the set of vertices $V^{\prime}$.
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- Color a vertex $v \in V^{\prime}$ red if it corresponds to a component from $\mathbb{C}$ (an active component). Otw. color it blue. (Let $B$ the set of blue vertices (with non-zero degree) and $R$ the set of red vertices)
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- We have

$$
\sum_{v \in R} \operatorname{deg}(v) \geq \sum_{C \in \mathbb{C}}\left|\delta(C) \cap F^{\prime}\right| \stackrel{?}{\leq} 2|\mathbb{C}|=2|R|
$$

- Suppose that no node in $B$ has degree one.
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- Then
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$$
\sum_{v \in R} \operatorname{deg}(v)
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- Then

$$
\sum_{v \in R} \operatorname{deg}(v)=\sum_{v \in R \cup B} \operatorname{deg}(v)-\sum_{v \in B} \operatorname{deg}(v)
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\begin{aligned}
\sum_{v \in R} \operatorname{deg}(v) & =\sum_{v \in R \cup B} \operatorname{deg}(v)-\sum_{v \in B} \operatorname{deg}(v) \\
& \leq 2(|R|+|B|)-2|B|
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- Every blue vertex with non-zero degree must have degree at least two.
- Suppose that no node in $B$ has degree one.
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- Every blue vertex with non-zero degree must have degree at least two.
- Suppose not. The single edge connecting $b \in B$ comes from $H$, and, hence, is necessary.
- Suppose that no node in $B$ has degree one.
- Then

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\begin{aligned}
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- Every blue vertex with non-zero degree must have degree at least two.
- Suppose not. The single edge connecting $b \in B$ comes from $H$, and, hence, is necessary.
- But this means that the cluster corresponding to $b$ must separate a source-target pair.
- Suppose that no node in $B$ has degree one.
- Then

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- Every blue vertex with non-zero degree must have degree at least two.
- Suppose not. The single edge connecting $b \in B$ comes from $H$, and, hence, is necessary.
- But this means that the cluster corresponding to $b$ must separate a source-target pair.
- But then it must be a red node.

