## Termination of the Ford-Fulkerson Algorithm

## Theorem 7.6.

a If all capacities are rational, then the algorithm terminates with a maximum $s-t$-flow.
b If all capacities are integral, it finds an integral maximum $s$ - $t$-flow.

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When an arbitrary $x$-augmenting path is chosen in every iteration, the Ford-Fulkerson Algorithm can behave badly:


## Running Time of the Ford-Fulkerson Algorithm

## Theorem 7.7.

If all capacities are integral and the maximum flow value is $K<\infty$, then the Ford-Fulkerson Algorithm terminates after at most $K$ iterations. Its running time is $O(m \cdot K)$ in this case.

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## Theorem 7.8.

The Edmonds-Karp Algorithm terminates after at most $n \cdot m$ iterations; its running time is $O\left(n \cdot m^{2}\right)$.

Remark: The Edmonds-Karp Algorithm can be implemented with running time $O\left(n^{2} \cdot m\right)$.

Arc-Based LP Formulation
Straightforward LP formulation of the maximum $s$-t-flow problem:

$$
\begin{aligned}
& \max \sum_{a \in \delta^{+}(s)} x_{a}-\sum_{a \in \delta^{-}(s)} x_{a} \text { net flow out of } S \\
& \text { set. } \quad \sum_{a \in \delta^{-}(v)} x_{a}-\sum_{a \in \delta^{+}(v)} x_{a}=0 \\
& x_{a} \leq u(a) \\
& x_{a} \geq 0 \\
& \text { flow conservation } \\
& \text { for all } v \in V \backslash\{s, t\} \quad y_{v} \\
& \begin{array}{c}
\text { capacity constronints } \\
\text { for all } a \in A
\end{array} \\
& z_{a} \\
& \text { for all } a \in A
\end{aligned}
$$

Column of primal


$$
\begin{aligned}
\min \sum_{a \in A} u(a) \cdot z_{a} & & y_{+}=0 \\
\text { s.t. }-y_{v}+y_{w}+z_{v w} & \geqslant 0 & \forall(v, w) \in A=v, w \\
y_{w}+z_{s w} & \geqslant 1 & \forall(s, w) \in A \\
-y_{v}+z_{v s} & \geqslant-1 & \forall(v, s) \in A \\
-y_{v}+z_{v t} & \geqslant 0 & \forall(v, t) \in A \\
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z_{a} & \geqslant 0 & \forall a \in A
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& x_{a} \leq u(a) & \\
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Dual LP:

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\begin{array}{lll}
\min & \sum_{a \in A} u(a) \cdot z_{a} & \\
\text { s.t. } & y_{w}-y_{v}+z_{(v, w)} \geq 0 & \text { for all }(v, w) \in A \\
& y_{s}=1, \quad y_{t}=0 & \\
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## Dual Solutions and $s$ - $t$-Cuts

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Observation: An s-t-cut $\delta^{+}(U)$ (with $U \subseteq V \backslash\{t\}, s \in U$ ) yields feasible dual solution $(y, z)$ of value $u\left(\delta^{+}(U)\right)$ :

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- let $y$ be the characteristic vector $\chi^{U}$ of $U$
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## Theorem 7.9.

There exists an $s$ - $t$-cut $\delta^{+}(U)$ (with $U \subseteq V \backslash\{t\}, s \in U$ ) such that the corresponding dual solution $(y, z)$ is an optimal dual solution.

Proof The 4.9:
opt value of dual $\min$ cap of a cut 11 strong duality max - flow 11 min -cat opt value of primal $=$ max flow value them

