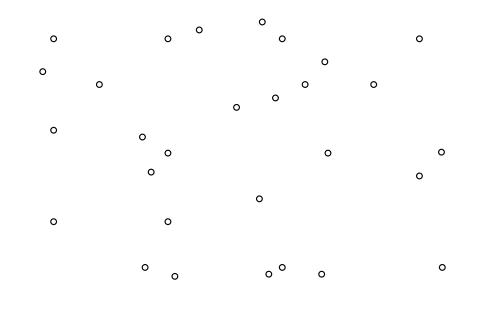
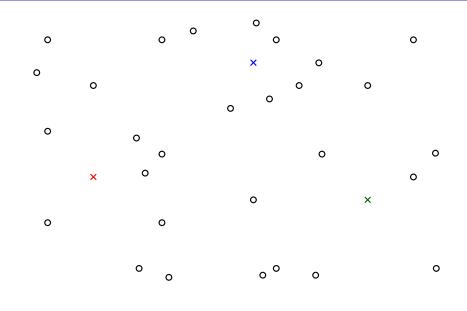
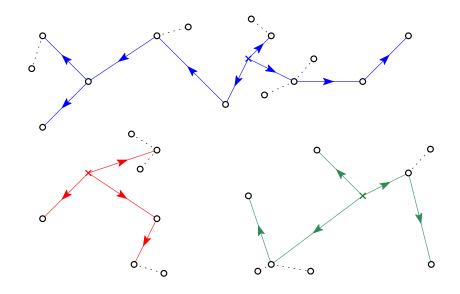
A branch-and-cut algorithm for the κ -connected Forest Star Problem

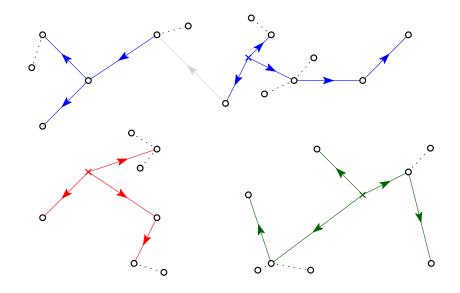
Alfie Plant University of Edinburgh

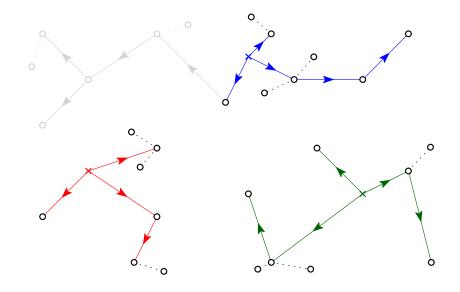
 25^{th} March 2025

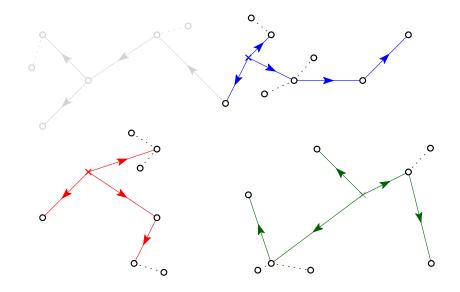


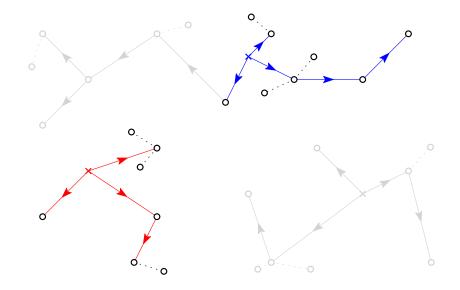


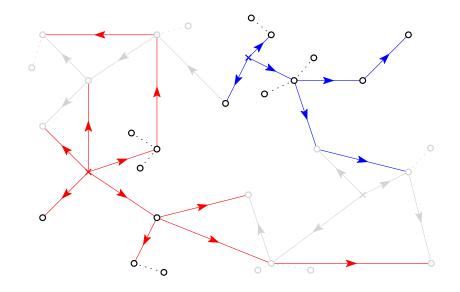


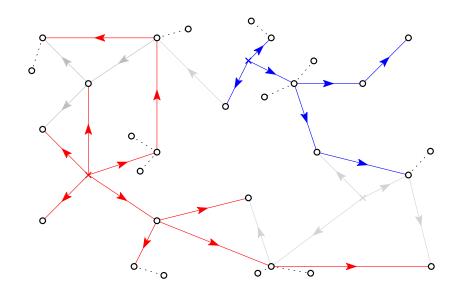










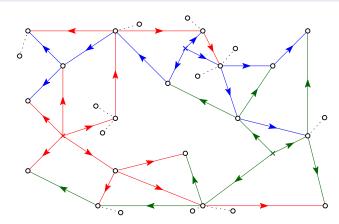


The κ -connected Forest Star Problem

Given a complete directed graph $\mathcal{G}=(\mathcal{V}\cup\mathcal{R},\mathcal{A})$, find a set of directed trees, each rooted at a different $r\in\mathcal{R}$, such that each vertex $i\in\mathcal{V}$ is either on κ trees, or is assigned a vertex that is, which minimises the total tree and assignment cost.

The κ -connected Forest Star Problem

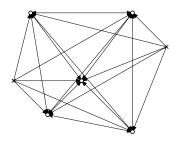
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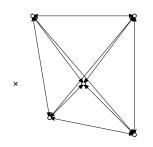
Agenda

- Naive IP formulation
- Reduced IP formulation
- Branch-and-bound algorithm
- Outline of proof for the equivalence of formulations
- Computational results
- 6 Conclusions

■ Consider a directed graph $\mathcal{G} = (\mathcal{V} \cup \mathcal{R}, \mathcal{A})$



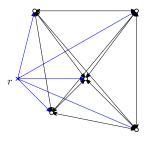
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The κ -connected Forest Star Problem

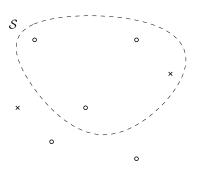
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- Consider a directed graph $\mathcal{G} = (\mathcal{V} \cup \mathcal{R}, \mathcal{A})$
- $\mathcal{A}_r = \mathcal{A}_{\mathcal{V}} \cup (\{r\} \times \mathcal{V})$ for some $r \in \mathcal{R}$

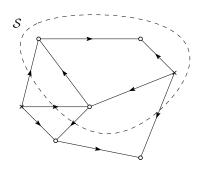


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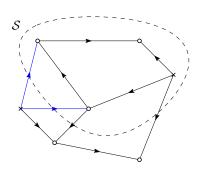
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- $A_r = A_{\mathcal{V}} \cup (\{r\} \times \mathcal{V}) \text{ for some } r \in \mathcal{R}$
- For some vertex set $S \subset V \cup R$ and some arc set $T \subseteq A$



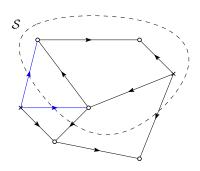
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- $\bullet \delta^+(\mathcal{T},\mathcal{S}) = \{(i,j) \in \mathcal{T} : i \notin \mathcal{S}, j \in \mathcal{S}\}\$
- If $\mathcal{T} = \mathcal{A}$, $\delta^+(\mathcal{S})$ is used for simplicity



Decision variables

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Note

If $y_{ii} = 1$, then vertex $i \in \mathcal{V}$ must be on a tree.

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Objective Function

$$\text{minimise} \sum_{(i,j,r) \in \mathcal{A} \times \mathcal{R}} c_{ij} x_{ij}^r \ + \sum_{(i,j) \in \mathcal{A}_{\mathcal{V}}} a_{ij} y_{ij}$$

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$$\sum_{j \in \mathcal{V}} y_{ij} = 1$$

$$\forall i \in \mathcal{V}$$

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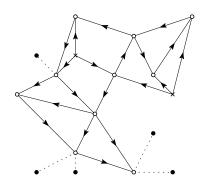
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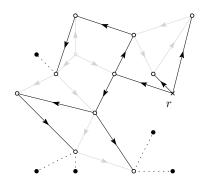
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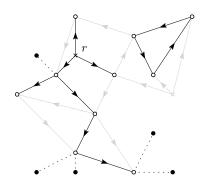
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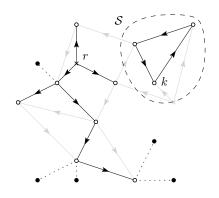


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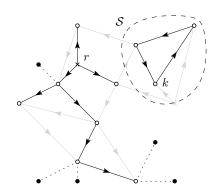


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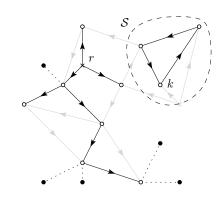
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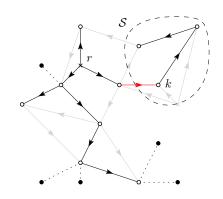
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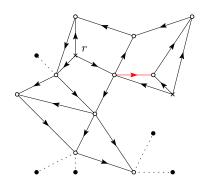
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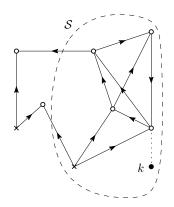
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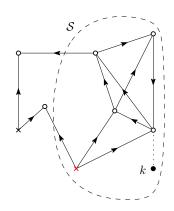
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$$\kappa = 2$$



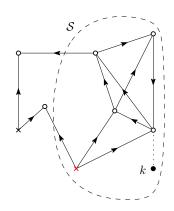
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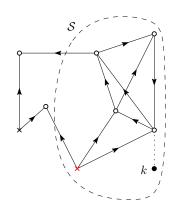
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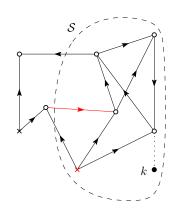
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Branch-and-cut Algorithm

■ Define the initial LP relaxation of the formulation as a sub-problem

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- **■** Select a sub-problem and determine the LP solution

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- 2 Select a sub-problem and determine the LP solution
- **3** Solve the separation problem

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- 2 Select a sub-problem and determine the LP solution
- Solve the separation problem
- If violated cycles are found, add them and return to step 1
- **5** Otherwise, create a branch and return to step 2

■ Assume $\kappa = |\mathcal{R}|$

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- lacksquare Decomposition : Reduced ightarrow Naive

Proposition 1

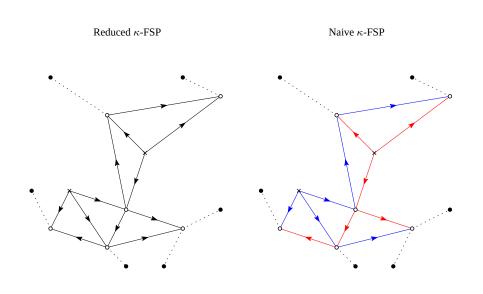
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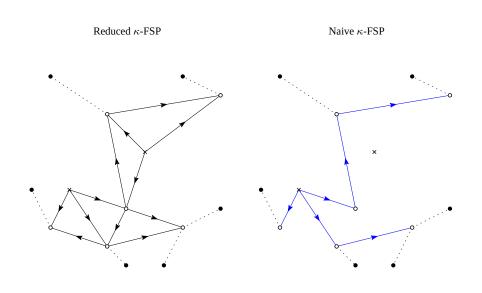
- Assume $\kappa = |\mathcal{R}|$
- \blacksquare Mapping : Naive \rightarrow Reduced
- Decomposition : Reduced \rightarrow Naive

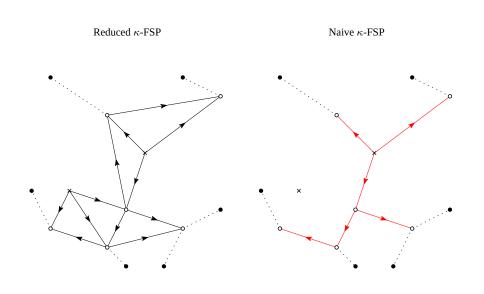
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$$\begin{array}{ll} \text{Let} & \bar{x}_{ij} = \sum_{r \in \mathcal{R}} x_{ij}^{r*} & & \forall (i,j) \in \mathcal{A} \\ \\ \text{and} & \bar{y}_{ij} = y_{ij}^* & & \forall (i,j) \in \mathcal{A}_{\mathcal{V}}. \end{array}$$







Theorem (Max-flow Min-cut)

Given $\mathcal{G} = (\mathcal{V} \cup \mathcal{R}, \mathcal{A})$ such that $\omega_{ij} \geq 0$ for $(i, j) \in \mathcal{A}$, together with some $s \in \mathcal{V} \cup \mathcal{R}$ and $t \in \mathcal{V} \cup \mathcal{R}$, the maximum flow in \mathcal{G} from s to t is equal to the minimum s-t cut of \mathcal{G} .

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- Maximum flow is zero, hence the minimum cut is zero.

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For any solution (\bar{x}, \bar{y}) to the reduced formulation there exists a path from each $r \in \mathcal{R}$ to k for all $k \in \mathcal{V}$ such that $\bar{y}_{kk} = 1$.

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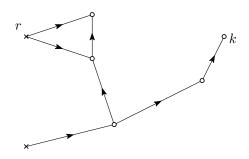
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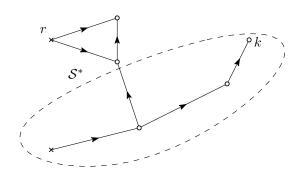
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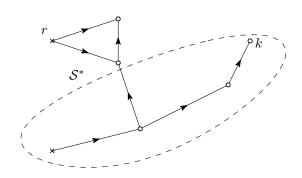
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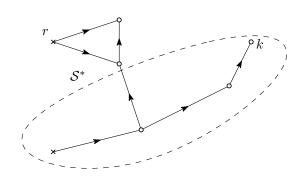
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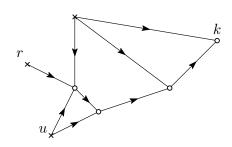


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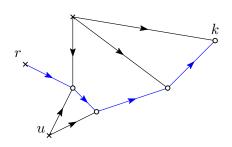
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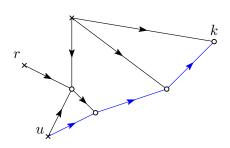
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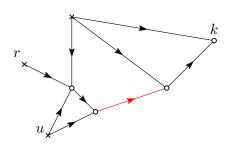
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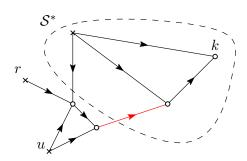
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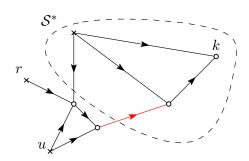
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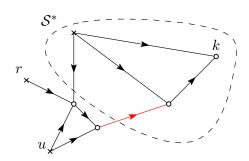
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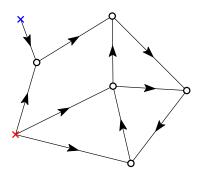
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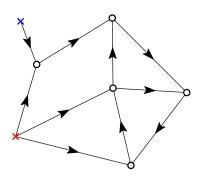
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$$\sum_{r \in \mathcal{R}} (x_{ij}^r + x_{ji}^r) \le 1 \quad \forall i \in \mathcal{V}, \ \forall j \in \mathcal{V}$$

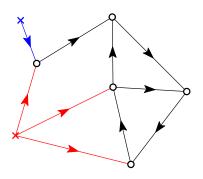
• A chokepoint is an arc that, when removed from the graph, there is no directed path from some $r \in \mathcal{R}$ to some $k \in \mathcal{V}$.



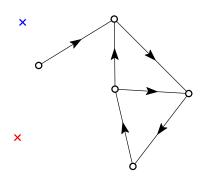
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- In other words, when a chokepoint arc is found, the path chosen from r to k must contain that arc.



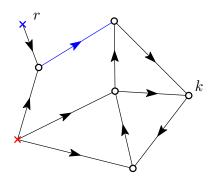
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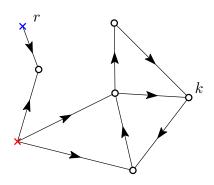
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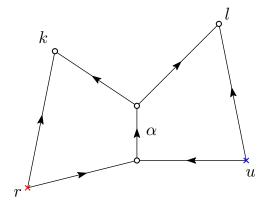
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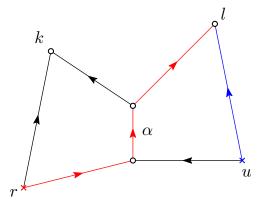
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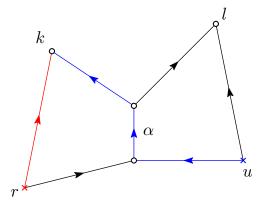
Lemma 3



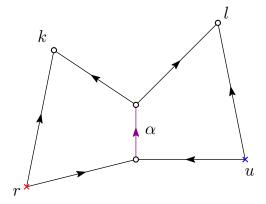
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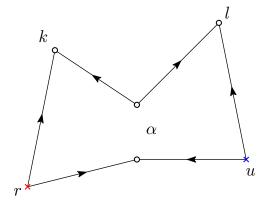
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Lemmas 1, 2 and 3 guarantee that an algorithm can be implemented to assign each arc in the solution to a tree.

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■ **Step 1**: Assign Root Arcs

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- Step 3a: Assign Chokepoint Paths
- Step 3b: Assign Remaining Paths

- Performance of naive and reduced formulations tested on five different seeds across 15 instances of data
- $|\mathcal{V}| = 20, 40, 60, 80, 100$
- $\kappa = 2, 4, 6$

instance			naive		reduced						
$ \mathcal{V} $	κ	build	cuts	cut time	opt.	solve	build	cuts	cut time	opt.	solve
	2	0.07	187	00.19	5	00.60	0.06	40	00.05	5	00.12
20	4	0.10	718	02.96	5	06.89	0.07	28	00.16	5	00.75
	6	0.16	656	03.19	5	08.28	0.06	11	00.11	5	00.76
	2	0.39	912	01.87	5	04.27	0.33	140	00.35	5	00.80
40	4	1.35	_	_	0	_	0.35	434	26.1 5	5	49.41
	6	1.23	_	-	0	-	0.37	16 5	06.72	5	19.43
	2	1.47	27 55	17.79	5	31.03	1.15	400	02.04	5	04.34
60	4	2.85	10848	01:27.39	2	12:11.83	1.26	3930	06:06.37	4	16:17.40
	6	4.22	_	-	0	-	1.48	2307	02:14.70	5	09:14.83
	2	6.83	8371	05:14.50	5	07:09.84	3.07	1289	11.32	5	18.97
80	4	13.67	14111	02:36.59	3	18:00.77	2.96	-	-	0	-
	6	16.69	-	-	0	-	3.90	3190	03:50.63	1	25:50.09
	2	11.60	9274	11:54.36	3	14:34.47	4.08	1306	11.80	5	19.52
100	4	25.09	_	-	0	-	4.24	-	-	0	-
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- Consequently, a tree decomposition is not given by a solution. A new algorithm has been presented that offers a structured approach to extracting this information
- Future research could explore the development of a primal heuristic for the κ -FSP and prove the result of equivalence for $\kappa < |\mathcal{R}|$