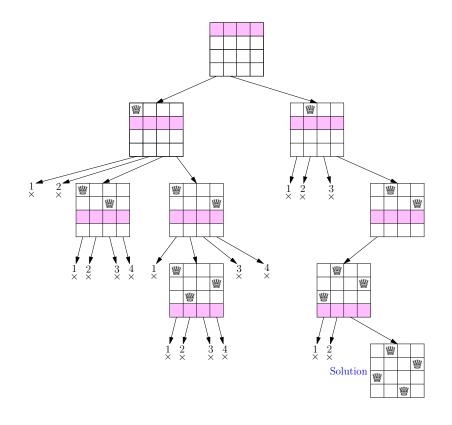
Algorithms and Analysis

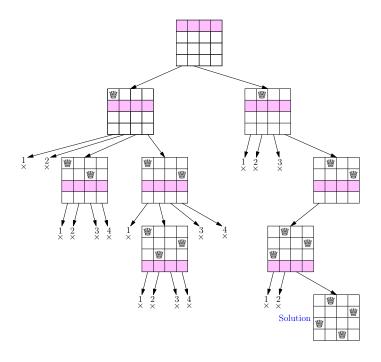
Lesson 22: Know how to Search



Backtracking, Branch and Bound

Outline

- 1. Search Trees
- 2. Backtracking
- 3. Branch and Bound
- 4. Search in Al



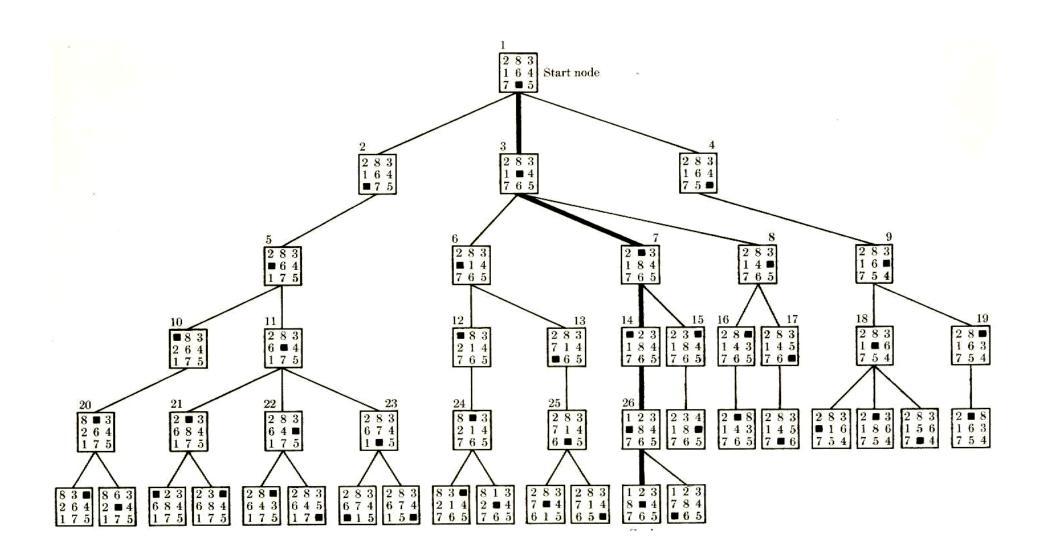
- Many real world problems involve taking a series of actions to manipulate the state of the system
- This is the area of planning and search which sits within the domain of artificial intelligence
- One of the key props to help us develop algorithms is to think of the states as nodes of a graph which are linked if there exists an action taking us from one state to another
- This provides a **state space representation** of the problem (we saw this before when we derived a low bound on sorting)

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8-Puzzle Example



Large State Spaces

- The search space typically increase exponentially with the problem size
- We can find the quickest solution to the 8-puzzle (and the 15 puzzle) using breadth first search, but larger puzzles soon become intractable
- Nevertheless, a lot of important problems involve very large state spaces and we have to find algorithms to explore them

Large State Spaces

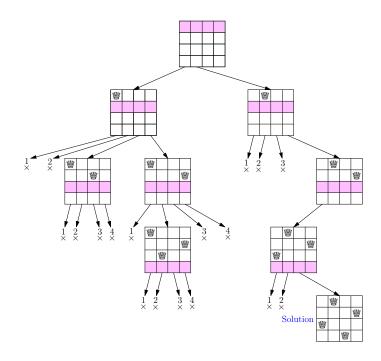
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- E.g. solving sudoku
- It works by growing partial solutions until either
 - * a feasible solution is found when we can finish
 - * no feasible solution is found when we backtrack
- We often search the state space using depth first search

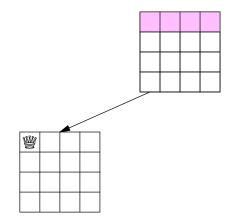
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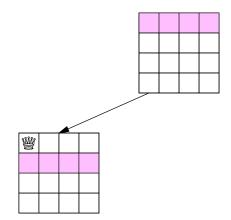
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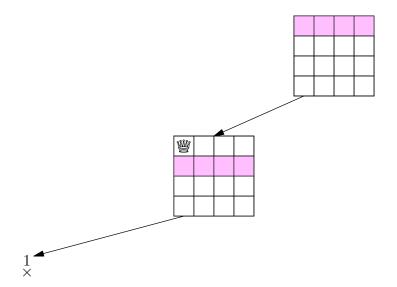
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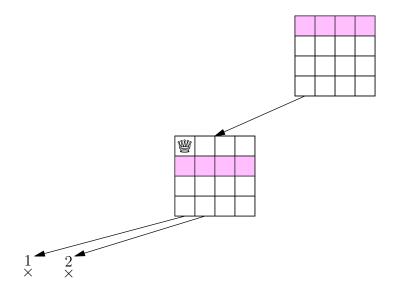


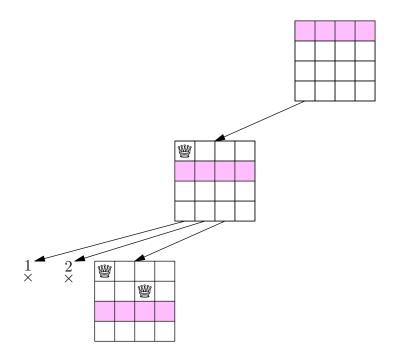


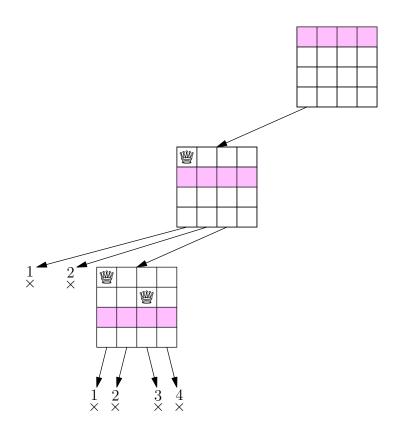


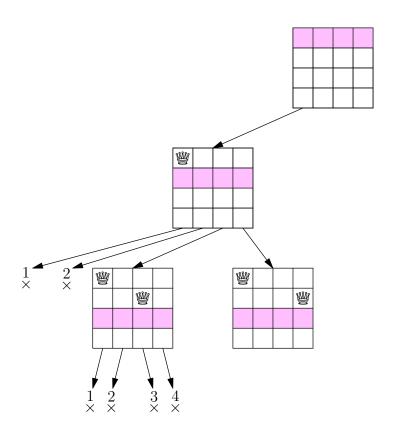


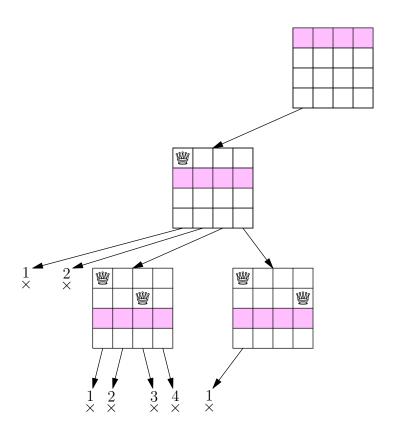


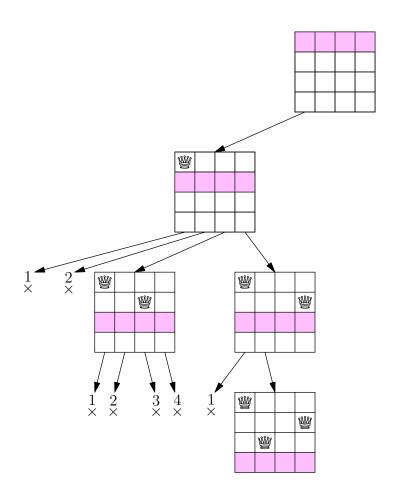


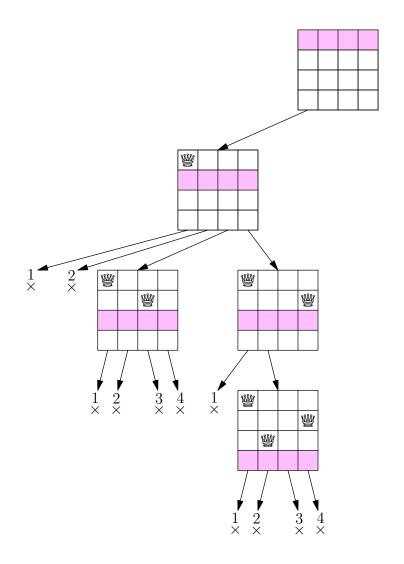


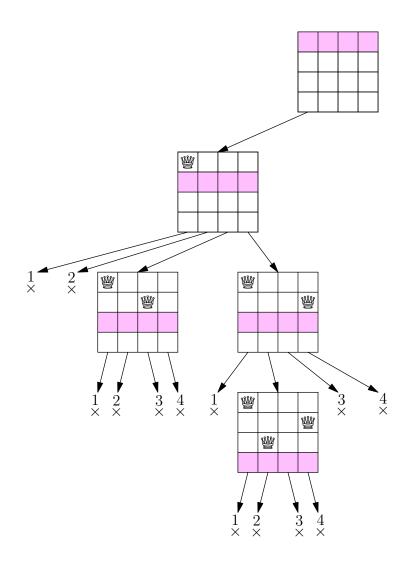


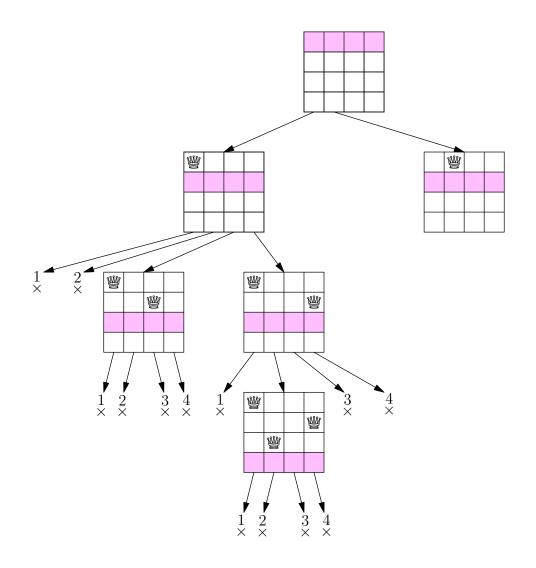


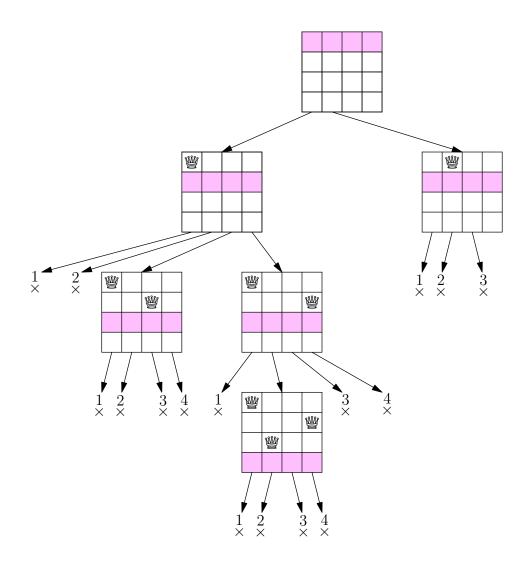


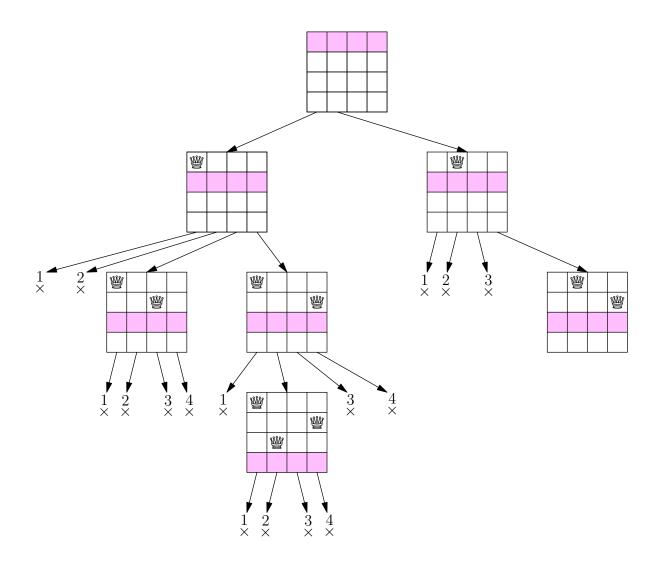


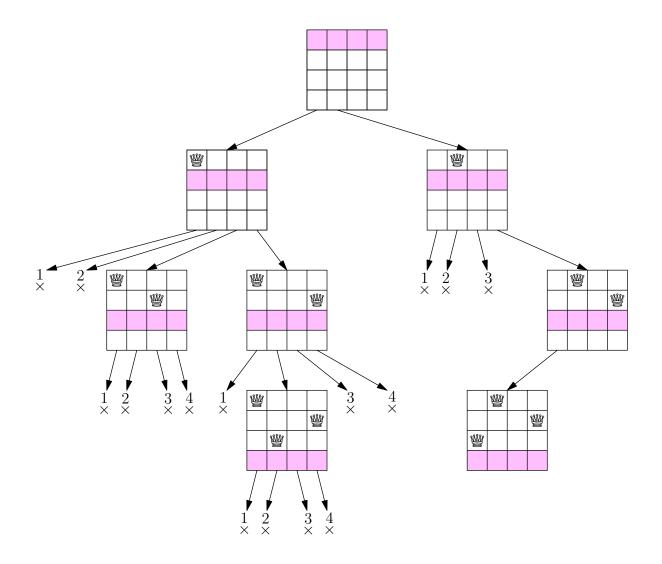


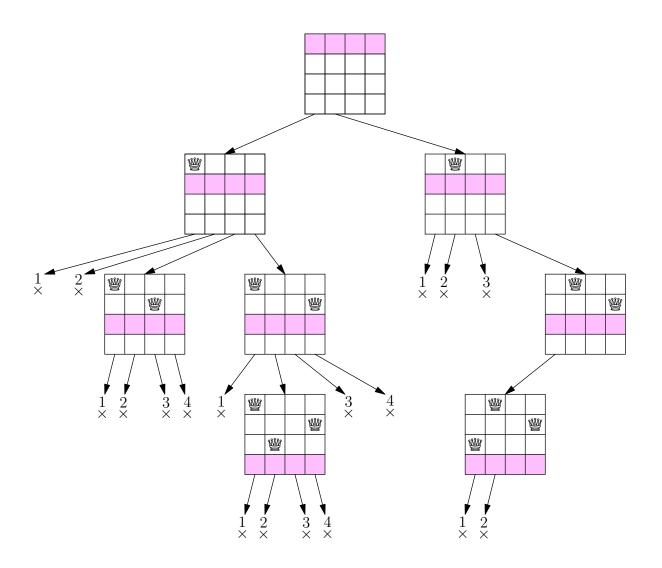


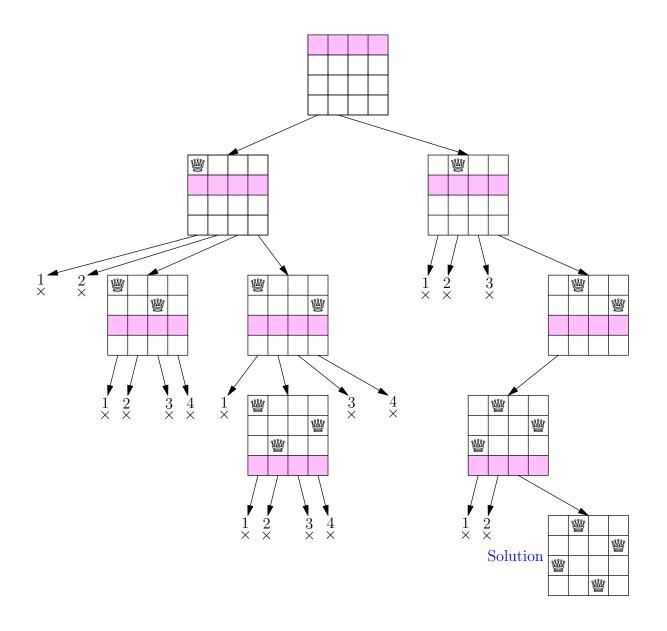


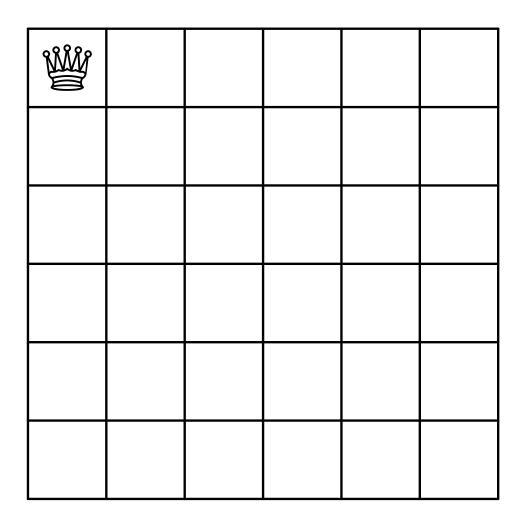












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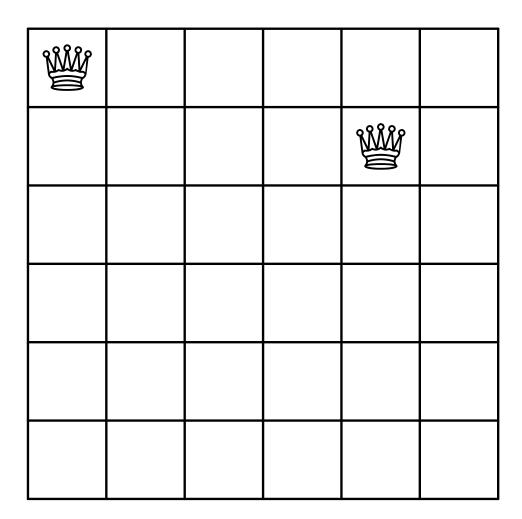
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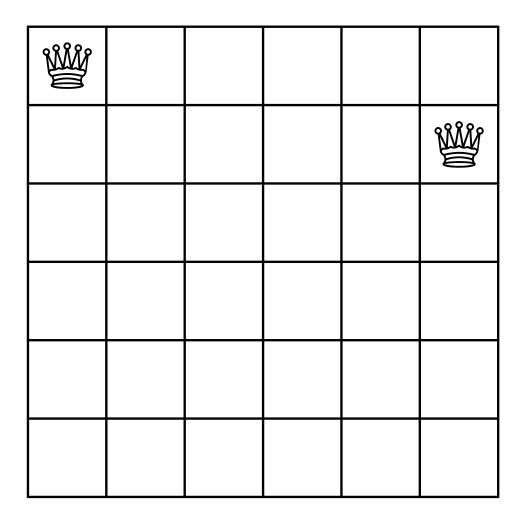
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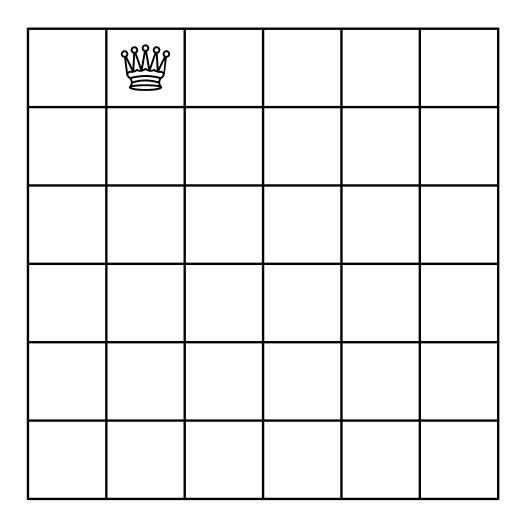


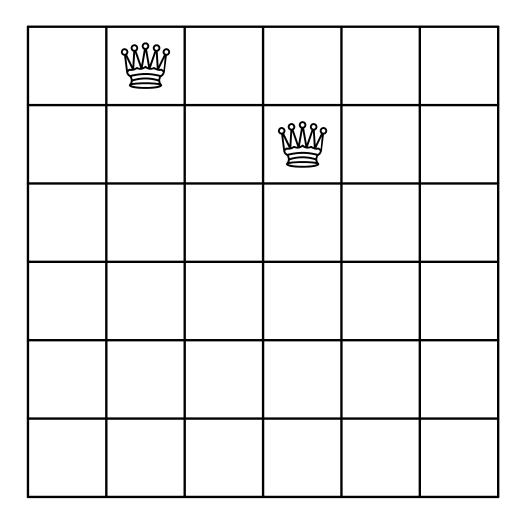
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- We just need a recursive function next(n, row, sol) which for a n-Queens problem searches new solutions in row given queens in previous rows given in sol
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Code

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List nextRow(int noRows, int row, List queenPositions) {
  if (row==noRows) {return queenPositions;}
  for (int col=0; col<noRows; ++col) {</pre>
    if (legalQueen(col, row, queenPositions)) {
      queenPositions.add(col);
      List solution = nextRow(noRows, row+1, queenPositions);
      if (solution!=null)
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bool LegalQueen(int col, int row, List sol) {
  for(int r=0; r<row: ++r) {
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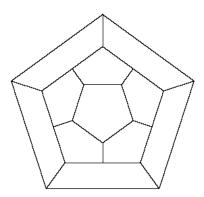
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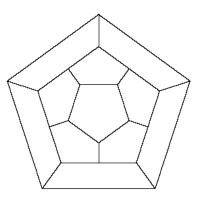
Hamiltonian Circuit

- A Hamiltonian cycle is a tour through a graph which visits every vertex once only and returns to the start
- It is a hard problem in that there are no known algorithms that are guaranteed to find a Hamiltonian cycle in polynomial time
- For many graphs it is not too hard



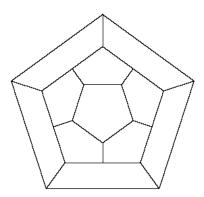
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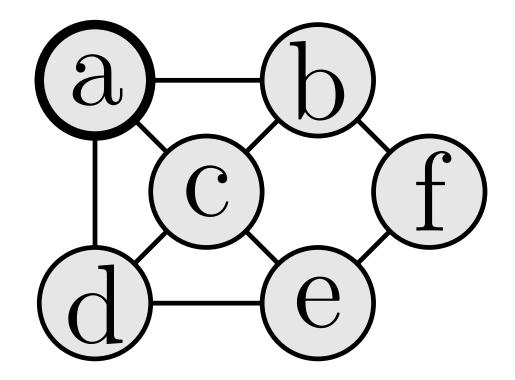
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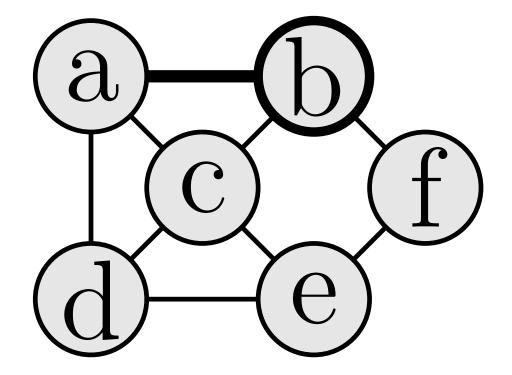


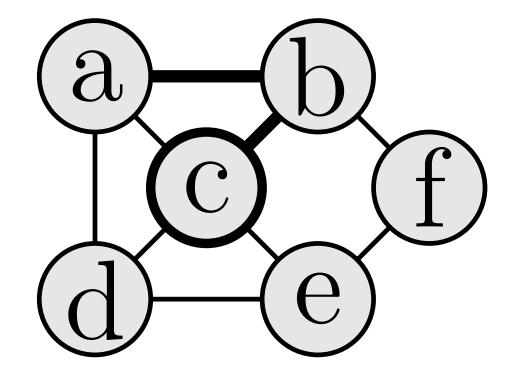
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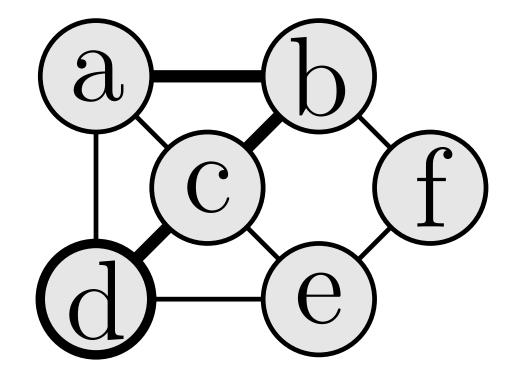
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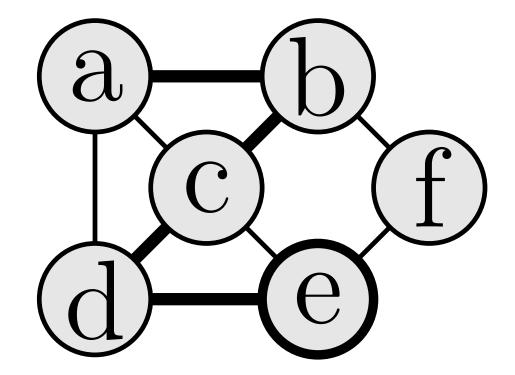


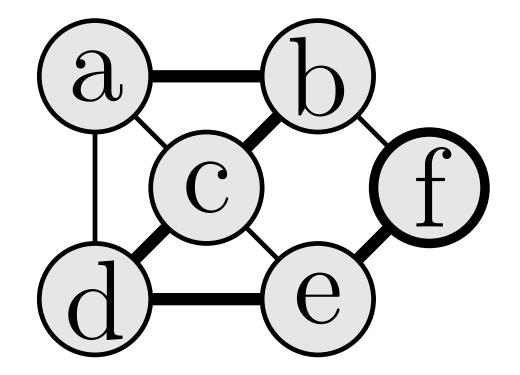


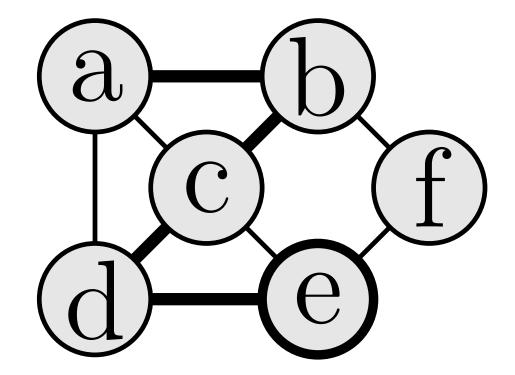


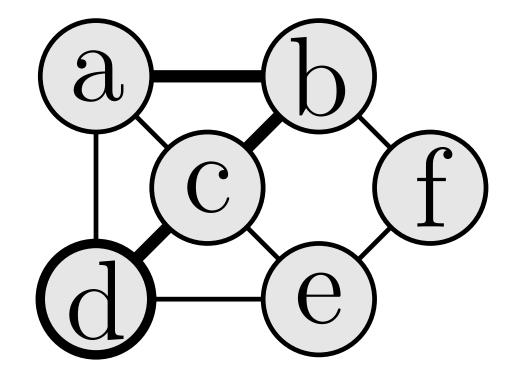


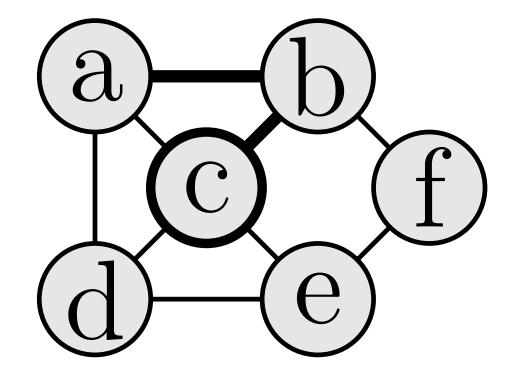


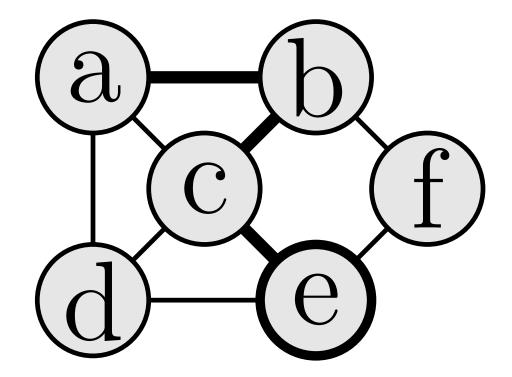


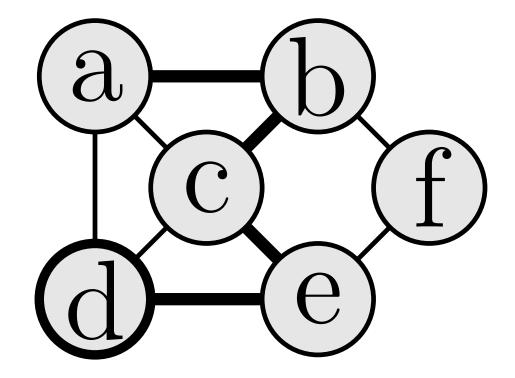


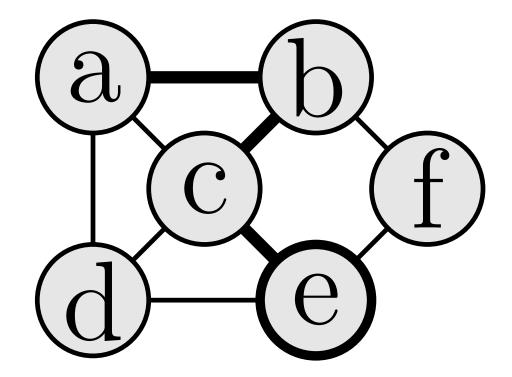


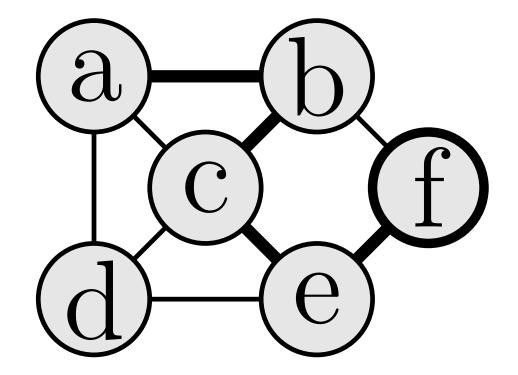


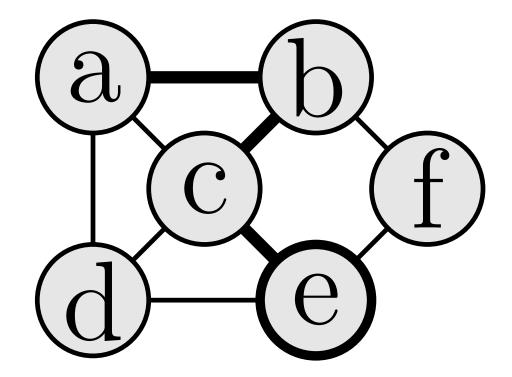


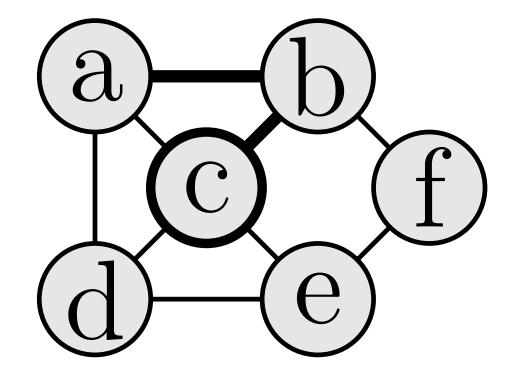


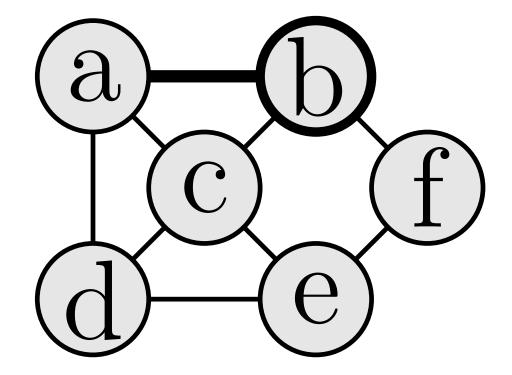


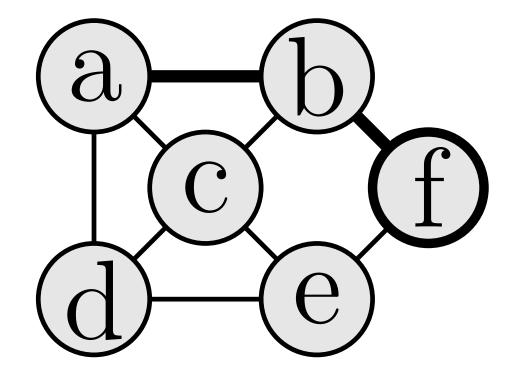


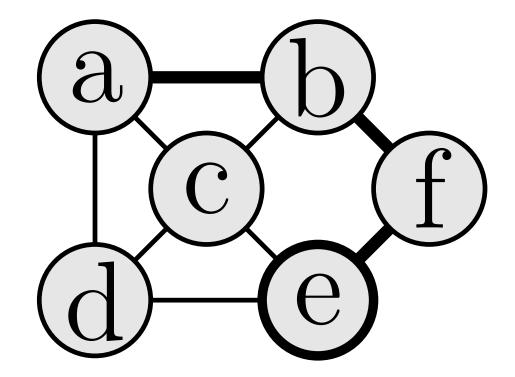


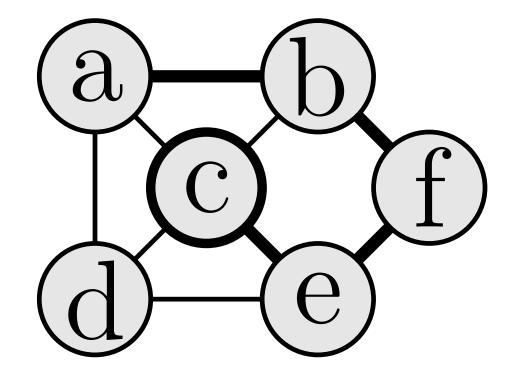


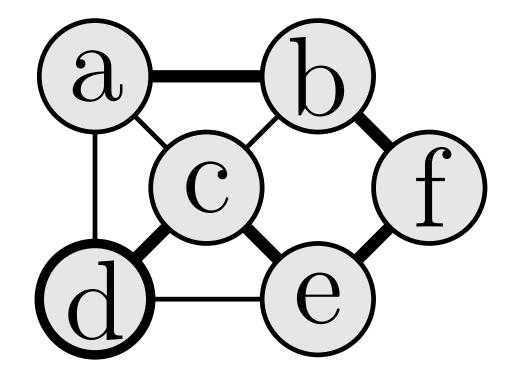


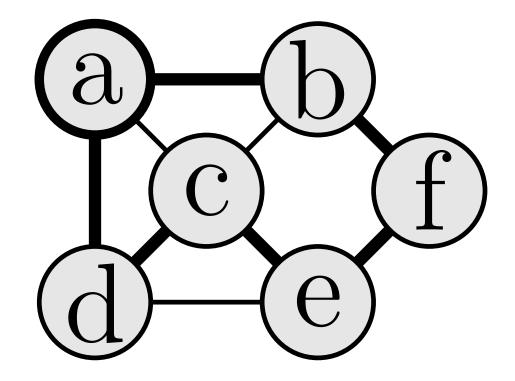












- Backtracking is a standard algorithm for solving constraint problems with large search spaces
- It can take exponential amount of time, however with many constraints it will often find solutions relatively quickly
- A backtracking algorithm does not solve, for example, sudoku in the same way as a human
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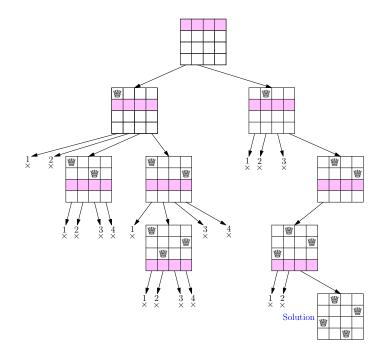
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- However, we don't have hard constraints
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- It beats exhaustive enumeration by eliminate many possible solutions without having to enumerate them all
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- By working harder we can sometimes strengthen the constraints thus eliminating much of the search space
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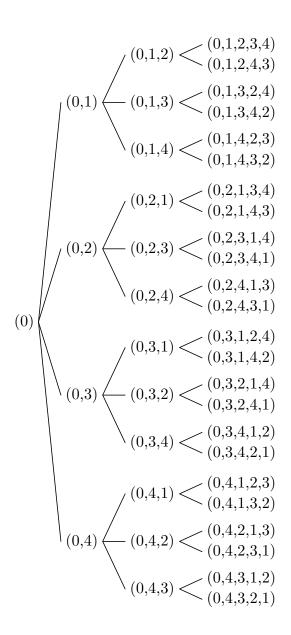
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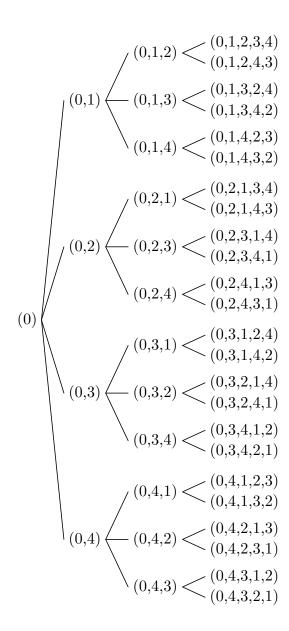
Cutting the Search Tree

- We can think of exact enumeration as exploring a giant search tree
- If we know a partial solution is worse than our bound we cut the search tree
- The earlier we cut the tree the more we can save



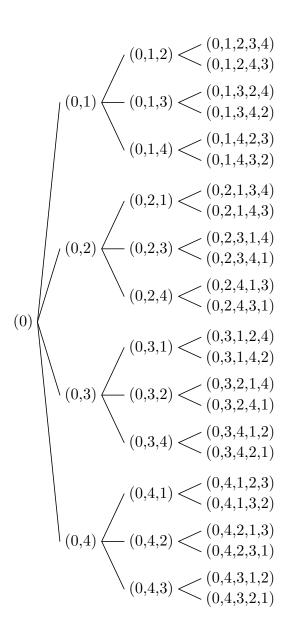
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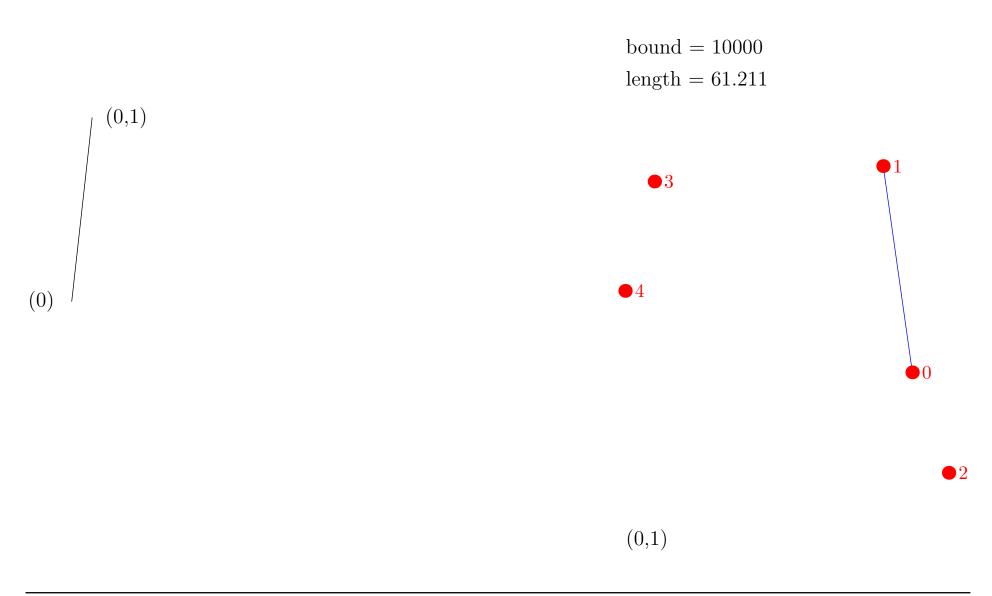
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- The earlier we cut the tree the more we can save

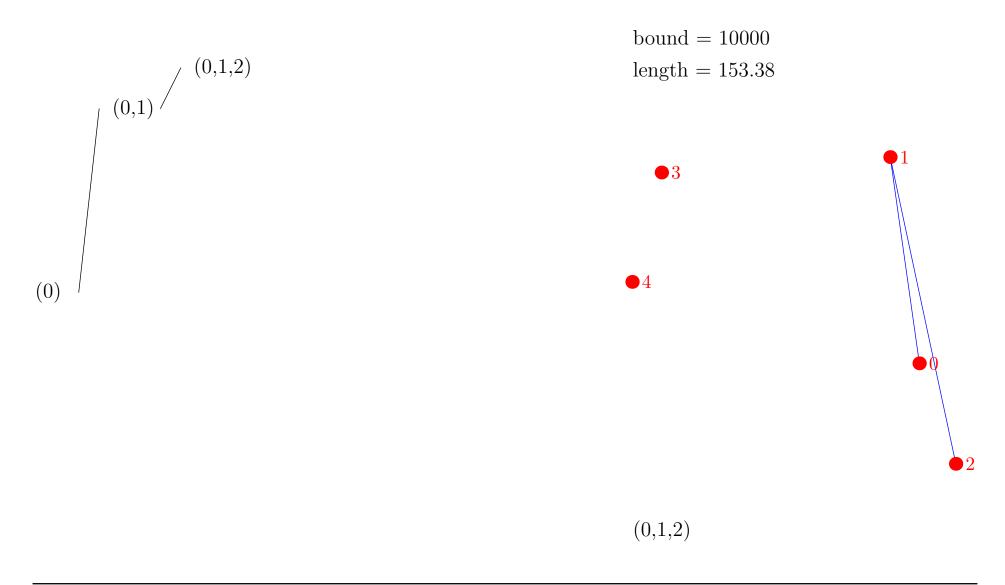


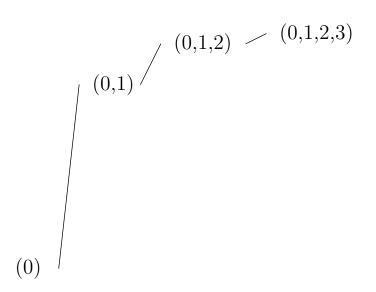
Cutting the Search Tree

- We can think of exact enumeration as exploring a giant search tree
- If we know a partial solution is worse than our bound we cut the search tree
- The earlier we cut the tree the more we can save

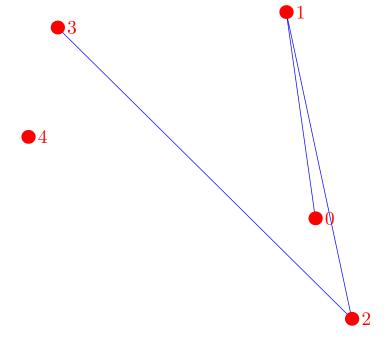




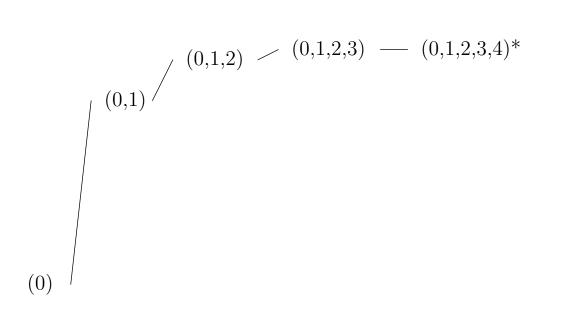




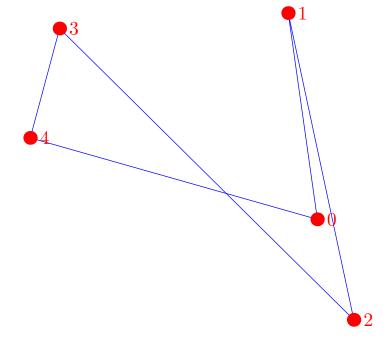
bound = 10000length = 275.05



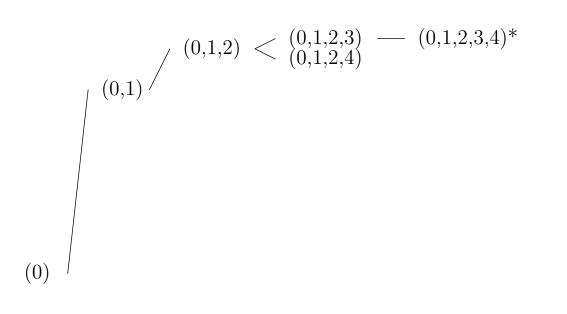
(0,1,2,3)

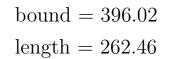


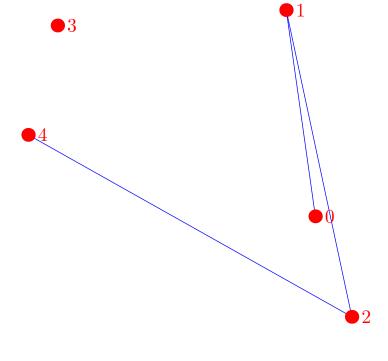
bound = 396.02length = 396.02



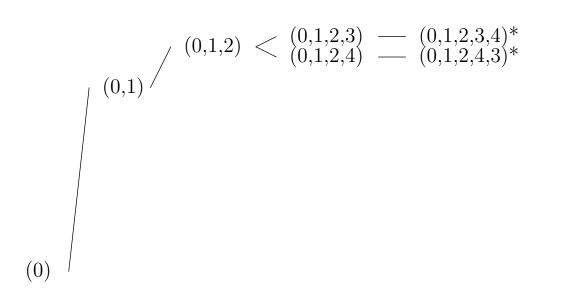
(0,1,2,3,4)



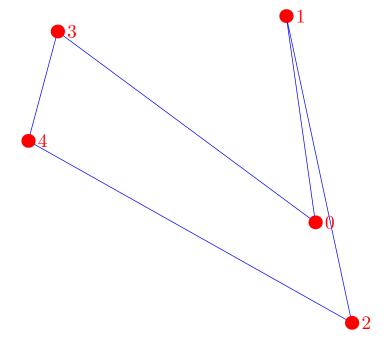




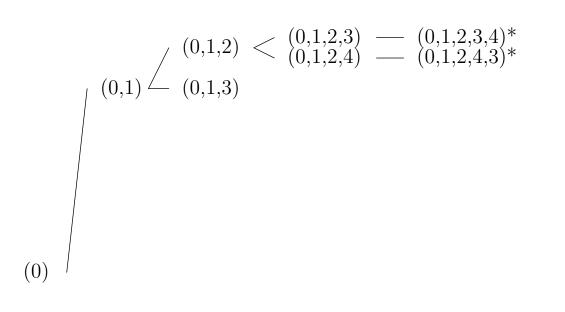
(0,1,2,4)

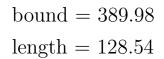


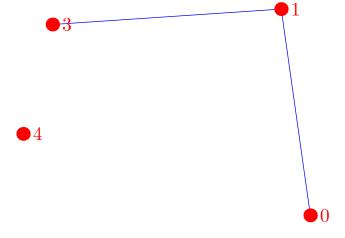
bound = 389.98length = 389.98



(0,1,2,4,3)

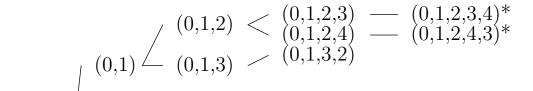




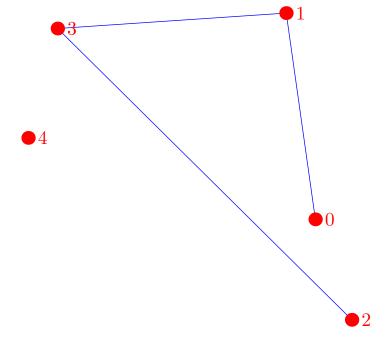


-2

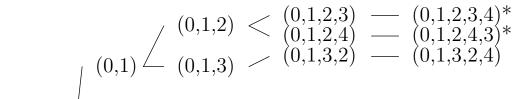
(0,1,3)



bound =
$$389.98$$
 length = 250.21

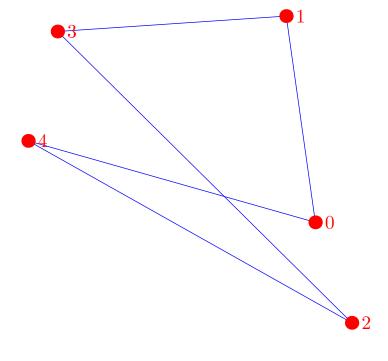


(0,1,3,2)

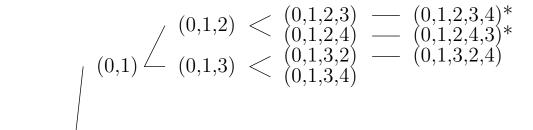


bound
$$= 389.98$$

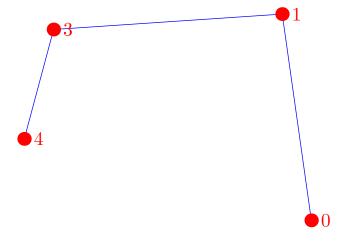
length $= 446.99$



(0,1,3,2,4)

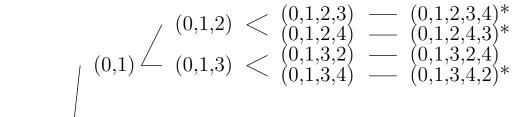


bound = 389.98length = 161.82



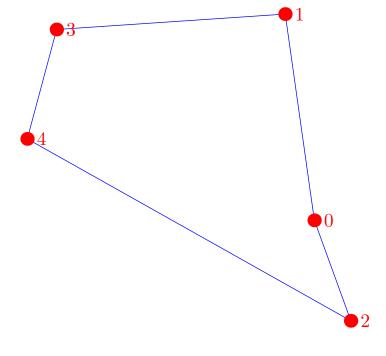
 $lue{2}$

(0,1,3,4)



bound =
$$302.31$$

length = 302.31

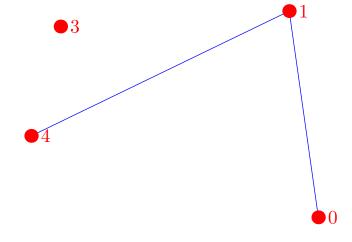


(0,1,3,4,2)

(0)

$$(0,1) \left\langle \begin{array}{c} (0,1,2) < \begin{pmatrix} (0,1,2,3) & --- & (0,1,2,3,4)^* \\ (0,1,2,4) & --- & (0,1,2,4,3)^* \\ (0,1,3) < \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\ (0,1,3,4) & --- & (0,1,3,4,2)^* \\ (0,1,4) & --- & (0,1,3,4,2)^* \end{array} \right.$$

$$bound = 302.31$$
$$length = 145.41$$

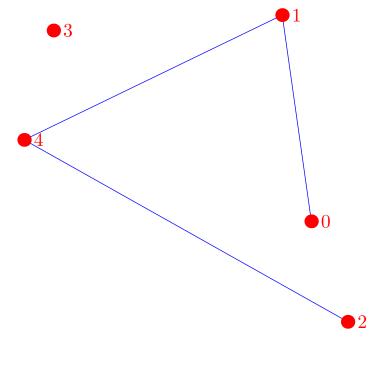


 $\bigcirc 2$

(0,1,4)

$$\begin{pmatrix}
(0,1,2) < (0,1,2,3) & --- & (0,1,2,3,4)^* \\
(0,1,2,4) & --- & (0,1,2,4,3)^* \\
(0,1,3) < (0,1,3,2) & --- & (0,1,3,2,4) \\
(0,1,3,4) & --- & (0,1,3,4,2)^* \\
(0,1,4) < (0,1,4,2)
\end{pmatrix}$$

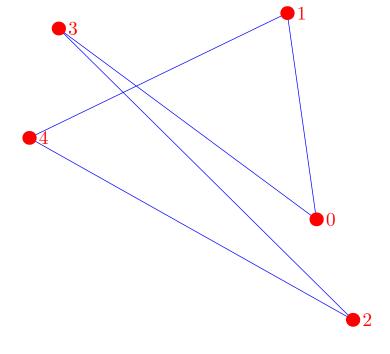
$$bound = 302.31$$
$$length = 254.49$$



(0,1,4,2)

$$(0,1) \left\langle \begin{array}{c} (0,1,2) < (0,1,2,3) & ---- (0,1,2,3,4)^* \\ (0,1,2,4) & ---- (0,1,2,4,3)^* \\ (0,1,3) < (0,1,3,2) & ---- (0,1,3,2,4) \\ (0,1,3) & ---- (0,1,3,4,2)^* \\ (0,1,4) & ---- (0,1,4,2,3) \end{array} \right\rangle$$

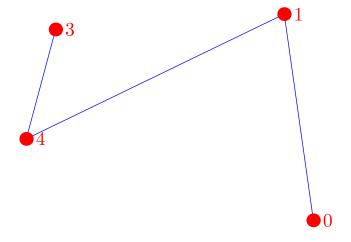
bound = 302.31length = 470.41



(0,1,4,2,3)

$$(0,1) \left\langle \begin{array}{c} (0,1,2) < \begin{pmatrix} (0,1,2,3) & --- & (0,1,2,3,4)^* \\ (0,1,2,4) & --- & (0,1,2,4,3)^* \\ (0,1,3) < \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\ (0,1,3,4) & --- & (0,1,3,4,2)^* \\ (0,1,4) < \begin{pmatrix} (0,1,4,2) & --- & (0,1,4,2,3) \\ (0,1,4,3) & --- & (0,1,4,2,3) \end{array} \right\rangle$$

$$bound = 302.31$$
$$length = 178.69$$

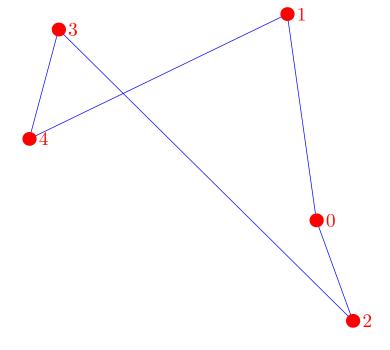


O 2

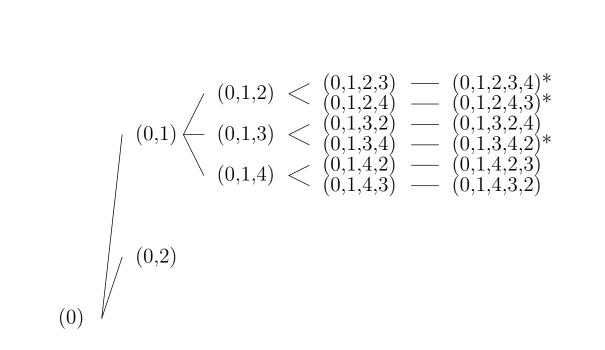
(0,1,4,3)

$$(0,1) \left\langle \begin{array}{c} (0,1,2) < \begin{pmatrix} (0,1,2,3) & \dots & (0,1,2,3,4)^* \\ (0,1,2,4) & \dots & (0,1,2,4,3)^* \\ (0,1,3) < \begin{pmatrix} (0,1,3,2) & \dots & (0,1,3,2,4) \\ (0,1,3,4) & \dots & (0,1,3,4,2)^* \\ (0,1,4) < \begin{pmatrix} (0,1,4,2) & \dots & (0,1,4,2,3) \\ (0,1,4,3) & \dots & (0,1,4,3,2) \\ \end{array} \right\rangle$$

bound = 302.31length = 331.77



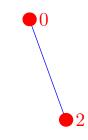
(0,1,4,3,2)



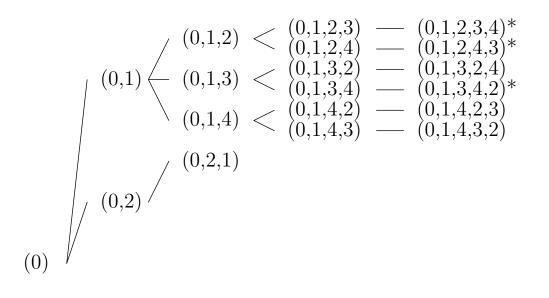
bound = 302.31length = 31.41

•3

left



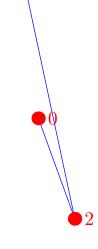
(0,2)



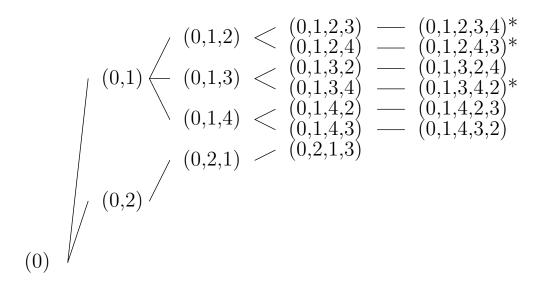
bound = 302.31length = 123.58

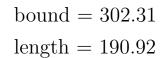
3

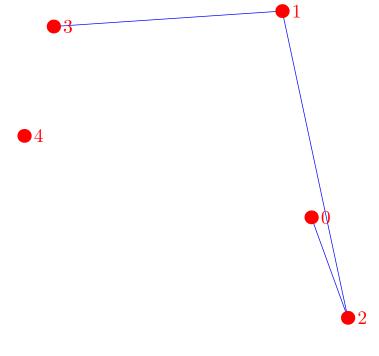
4



(0,2,1)





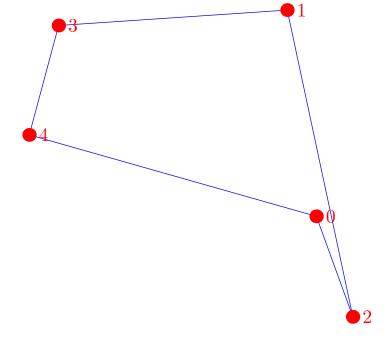


(0,2,1,3)

$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ---- (0,1,2,3,4)^* \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2)^* \\
(0,2,1) &> (0,2,1,3) & ---- (0,2,1,3,4)
\end{pmatrix}^*$$

$$(0,1) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,4,2,3) \\
(0,1,4,3) & ---- (0,1,4,3,2) \\
(0,2,1,3) & ---- (0,2,1,3,4)
\end{pmatrix}^*$$

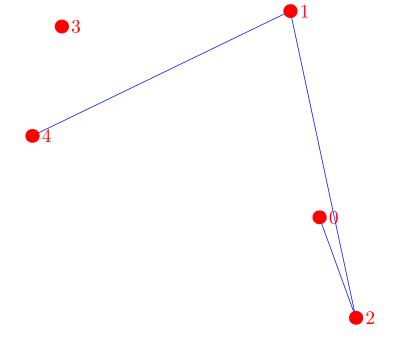
bound = 302.31length = 311.88



(0,2,1,3,4)

$$\begin{pmatrix}
(0,1,2) &< \begin{pmatrix} (0,1,2,3) & --- & (0,1,2,3,4)^* \\
(0,1,2,4) & --- & (0,1,2,4,3)^* \\
(0,1,3) &< \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\
(0,1,3,4) & --- & (0,1,3,4,2)^* \\
(0,1,4) &< \begin{pmatrix} (0,1,4,2) & --- & (0,1,4,2,3) \\
(0,1,4,3) & --- & (0,1,4,3,2) \\
(0,2,1) &< \begin{pmatrix} (0,2,1,3) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,4) \\
(0,2,1,4) & --- & (0,2,1,4) \\
(0,$$

bound = 302.31 length = 207.79

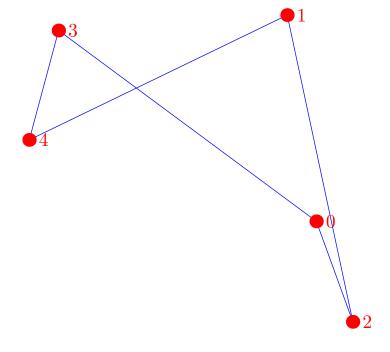


(0,2,1,4)

$$\begin{pmatrix}
(0,1,2) &< \begin{pmatrix} 0,1,2,3 \end{pmatrix} & --- & (0,1,2,3,4)^* \\
(0,1) &< \begin{pmatrix} 0,1,3 \end{pmatrix} &< \begin{pmatrix} 0,1,3,2 \end{pmatrix} & --- & (0,1,2,3,4)^* \\
(0,1,3) &< \begin{pmatrix} 0,1,3,2 \end{pmatrix} & --- & (0,1,3,2,4) \\
(0,1,4) &< \begin{pmatrix} 0,1,4,2 \end{pmatrix} & --- & (0,1,3,4,2)^* \\
(0,2,1) &< \begin{pmatrix} 0,2,1,3 \end{pmatrix} & --- & (0,2,1,3,4) \\
(0,2,1,4) &--- & (0,2,1,3,4) \\
(0,2,1,4) &--- & (0,2,1,4,3)
\end{pmatrix}$$

$$(0,2)$$

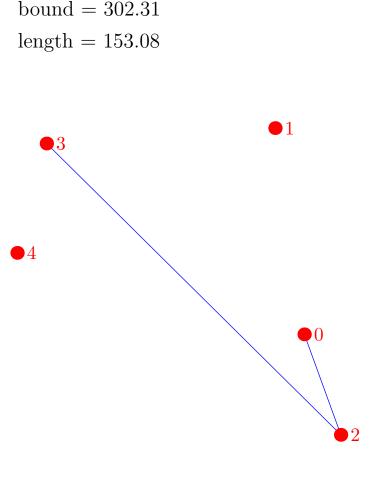
bound = 302.31length = 335.3



(0,2,1,4,3)

$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ---- (0,1,2,3,4)^* \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2)^* \\
(0,2,1) &< (0,2,1,3) & ---- (0,2,1,3,4) \\
(0,2,1,4) &= (0,2,1,4,3)
\end{pmatrix} (0,2)$$

$$(0) \qquad (0) \qquad$$



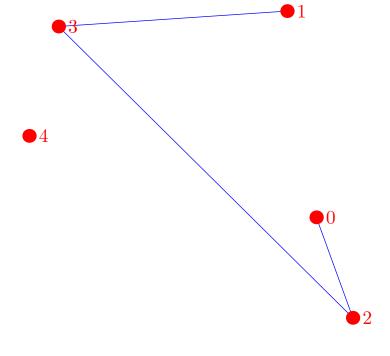
(0,2,3)

$$\begin{pmatrix}
(0,1,2) &< \begin{pmatrix} (0,1,2,3) & --- & (0,1,2,3,4) * \\ (0,1,2,4) & --- & (0,1,2,4,3) * \\ (0,1,3) &< \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\ (0,1,3,4) & --- & (0,1,3,4,2) * \\ (0,1,4) &< \begin{pmatrix} (0,1,4,2) & --- & (0,1,4,2,3) \\ (0,1,4,3) & --- & (0,1,4,3,2) \\ (0,2,1) &< \begin{pmatrix} (0,2,1,3) & --- & (0,2,1,3,4) \\ (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2) & --- & (0,2,3,1) \end{pmatrix}$$

$$(0,1) \begin{pmatrix}
(0,1,3) &< \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\ (0,1,4,2) & --- & (0,1,4,2,3) \\ (0,1,4,3) & --- & (0,1,4,3,2) \\ (0,2,1,4) & --- & (0,2,1,3,4) \\ (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2) & --- & (0,2,3,1) \end{pmatrix}$$

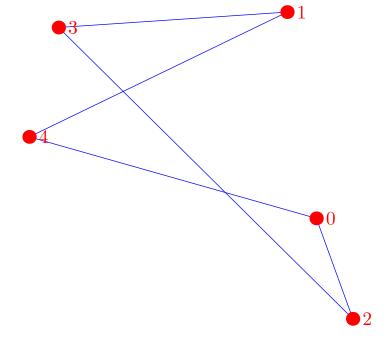
$$(0,2) \begin{pmatrix}
(0,2,3) &< \begin{pmatrix} (0,2,1,3) & --- & (0,2,1,3,4) \\ (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2) & --- & (0,2,2,3,1) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & --- & (0,2,2,2) \\ (0,2) & -$$

bound = 302.31length = 220.41



(0,2,3,1)

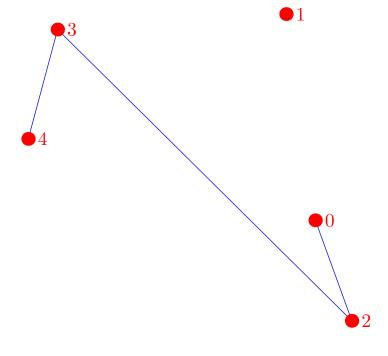
bound = 302.31length = 392.31



(0,2,3,1,4)

$$\begin{pmatrix}
(0,1,2) &< \begin{pmatrix} (0,1,2,3) & --- & (0,1,2,3,4)^* \\
(0,1,2,4) & --- & (0,1,2,4,3)^* \\
(0,1,3) &< \begin{pmatrix} (0,1,3,2) & --- & (0,1,3,2,4) \\
(0,1,3,4) & --- & (0,1,3,4,2)^* \\
(0,1,4) &< \begin{pmatrix} (0,1,4,2) & --- & (0,1,4,2,3) \\
(0,1,4,3) & --- & (0,1,4,3,2) \\
(0,2,1) &< \begin{pmatrix} (0,2,1,3) & --- & (0,2,1,3,4) \\
(0,2,1,4) & --- & (0,2,1,4,3) \\
(0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,4) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,3,1) & --- & (0,2,3,1,4) \\
(0,2) & --- & (0,2,2,3,1) & --- & (0,2,2,3,1) \\
(0,2) & --- & (0,2,2,3,1) & --- & (0,2,2,3,1) \\
(0,2) & --- & (0,2,2,3,1) & --- & (0,2,2,3,1) \\
(0,2) & --- & (0,2,2,3,1) & --- & (0,2,2,3,1) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2,2) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2,2) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2) \\
(0,2) & --- & (0,2,2,2,2) & --- & (0,2,2,2) \\
(0,2) & --- & (0,2,2,2) & --- & (0,2,2,2) \\
(0,2) & --- & (0,2,2,2) & ---$$

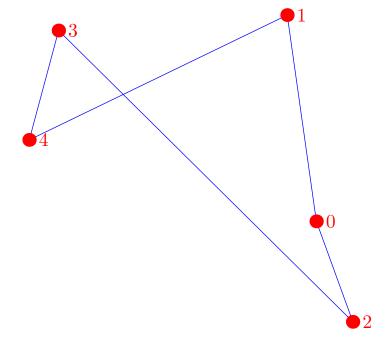
bound = 302.31length = 186.35



(0,2,3,4)

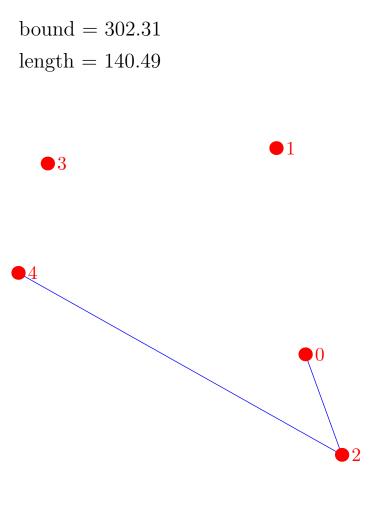
$$\begin{pmatrix}
(0,1,2) &< \begin{pmatrix} (0,1,2,3) & ---- & (0,1,2,3,4)^* \\
(0,1,2,4) & ---- & (0,1,2,4,3)^* \\
(0,1) &< (0,1,3) &< \begin{pmatrix} (0,1,3,2) & ---- & (0,1,3,2,4) \\
(0,1,3) &< & (0,1,3,4) & ---- & (0,1,3,4,2)^* \\
(0,1,4) &< \begin{pmatrix} (0,1,4,2) & ---- & (0,1,4,2,3) \\
(0,1,4,3) & ---- & (0,1,4,3,2) \\
(0,2,1) &< \begin{pmatrix} (0,2,1,3) & ---- & (0,2,1,3,4) \\
(0,2,1,4) & ---- & (0,2,1,4,3) \\
(0,2,3,4) & ---- & (0,2,3,1,4) \\
(0,2) &< \begin{pmatrix} (0,2,3,1) & ---- & (0,2,3,1,4) \\
(0,2,3,4) & ---- & (0,2,3,4,1) \end{pmatrix}$$

bound = 302.31length = 331.77



(0,2,3,4,1)

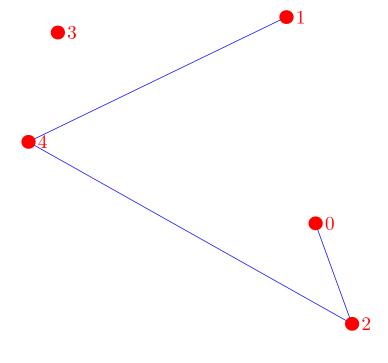
$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ---- (0,1,2,3,4)^* \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2)^* \\
(0,2,1) &< (0,2,1,3) & ---- (0,1,4,2,3) \\
(0,2,3) &< (0,2,3,1) & ---- (0,2,1,3,4) \\
(0,2,4) & (0,2,3,4) & ---- (0,2,3,1,4) \\
(0,2,4) & (0,2,3,4) & ---- (0,2,3,4,1)
\end{pmatrix}$$



(0,2,4)

$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ----- (0,1,2,3,4)^* \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2)^* \\
(0,2,1) &< (0,2,1,3) & ---- (0,1,4,2,3) \\
(0,2,1) &< (0,2,1,4) & ---- (0,2,1,3,4) \\
(0,2) &< (0,2,3,1) & ---- (0,2,3,1,4) \\
(0,2,4) &< (0,2,4,1)
\end{pmatrix}$$

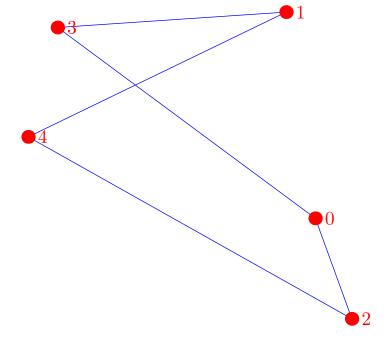
bound = 302.31length = 224.69



(0,2,4,1)

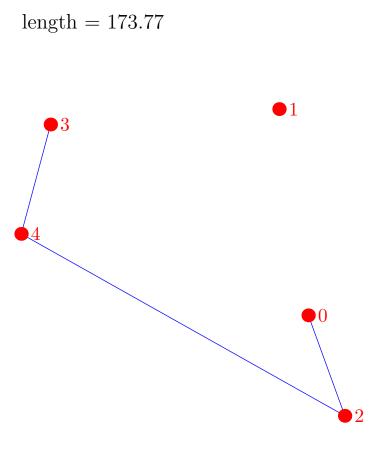
$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ---- (0,1,2,3,4)^* \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2)^* \\
(0,2,1) &< (0,2,1,3) & ---- (0,2,1,3,4) \\
(0,2,3) &< (0,2,3,4) & ---- (0,2,3,4,1) \\
(0,2,4) &> (0,2,4,1) & ---- (0,2,3,4,1)
\end{pmatrix}$$

bound = 302.31length = 386.27



(0,2,4,1,3)

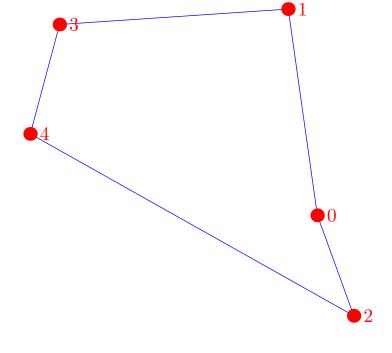
$$\begin{pmatrix}
(0,1,2) &< (0,1,2,3) & ---- (0,1,2,3,4) * \\
(0,1) &< (0,1,3) &< (0,1,3,2) & ---- (0,1,3,2,4) * \\
(0,1,4) &< (0,1,4,2) & ---- (0,1,3,4,2) * \\
(0,2,1) &< (0,2,1,3) & ---- (0,1,4,2,3) \\
(0,2,3) &< (0,2,1,4) & ---- (0,2,1,3,4) \\
(0,2,4) &< (0,2,4,1) & ---- (0,2,3,4,1) \\
(0,2,4) &< (0,2,4,3)
\end{pmatrix}$$



(0,2,4,3)

bound = 302.31

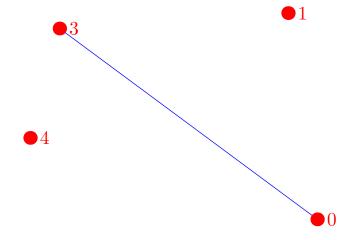
bound = 302.31length = 302.31



(0,2,4,3,1)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,3,4)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1,4) - (0,2,1,4,3) (0,2,3) < (0,2,3,1) - (0,2,3,1,4) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4,3) - (0,2,4,3,1)^* (0,3)$$

bound = 302.31length = 94.244

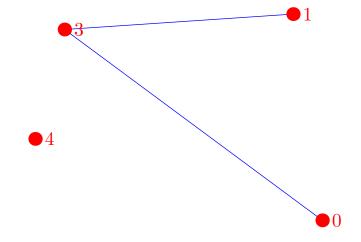


-2

(0,3)

$$(0,1,2) < \begin{pmatrix} 0,1,2,3 \end{pmatrix} & ---- & (0,1,2,3,4)^* \\ (0,1,2,4) & ---- & (0,1,2,4,3)^* \\ (0,1) & < & (0,1,3,2) & ---- & (0,1,3,2,4) \\ (0,1,3) & < & (0,1,3,4) & ---- & (0,1,3,4,2)^* \\ (0,1,4) & < & (0,1,4,2) & ---- & (0,1,4,2,3) \\ (0,2,1) & < & (0,2,1,3) & ---- & (0,2,1,3,4) \\ (0,2,1) & < & (0,2,1,4) & ---- & (0,2,1,4,3) \\ (0,2,1) & < & (0,2,1,4) & ---- & (0,2,3,1,4) \\ (0,2,3) & < & (0,2,3,4) & ---- & (0,2,3,4,1) \\ (0,2,4) & < & (0,2,4,1) & ---- & (0,2,4,1,3) \\ (0,3,1) & & & (0,3,1) \end{pmatrix}$$

bound = 302.31length = 161.58

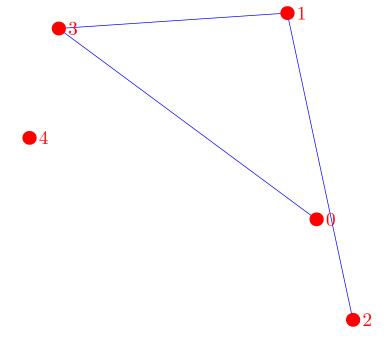


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(0,3,1)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1) < (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) (0,3)$$

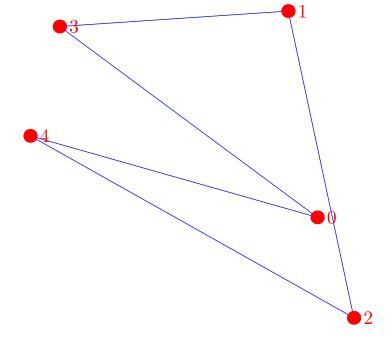
bound = 302.31length = 253.75



(0,3,1,2)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,3) < (0,2,1,4) - (0,2,1,4,3) (0,2,4) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3)$$

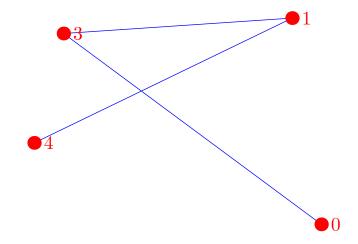
bound = 302.31length = 450.52



(0,3,1,2,4)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1) < (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3)$$

bound = 302.31 length = 245.78

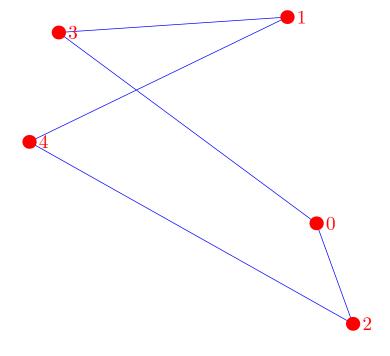


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(0,3,1,4)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,3) < (0,2,3,4) - (0,2,3,1,4) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,1,4,2) (0,3)$$

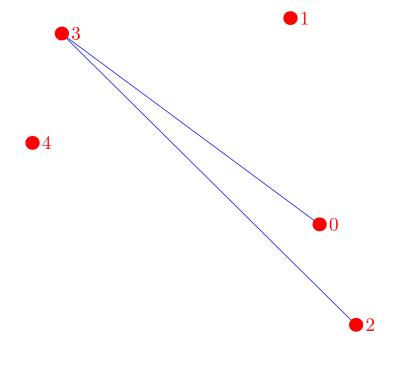
bound = 302.31length = 386.27



(0,3,1,4,2)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,4) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1,4) - (0,2,1,4,3) - (0,2,1,4,3) (0,2) < (0,2,3,1) - (0,2,3,1,4) (0,2,4) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,3) - (0,2,4,3,1) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,1,4) - (0,3,1,4,2) (0,3,2)$$

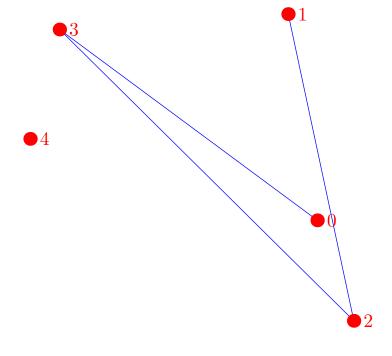
bound = 302.31length = 215.91



(0,3,2)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,3) < (0,1,3,4) - (0,1,3,4,2)^* (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,1) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,1) - (0,2,3,4,1) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3) - (0,3,2) < (0,3,2,1)$$

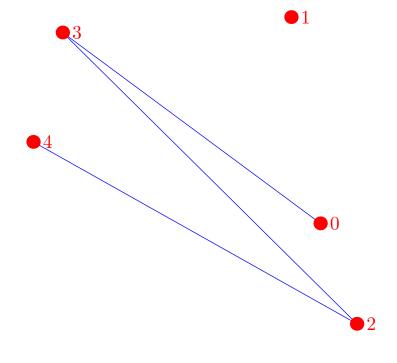
bound = 302.31length = 308.09



(0,3,2,1)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1) < (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,1,4) < (0,1,4,3) - (0,1,4,3,2) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1,4) - (0,2,1,4,3) (0,2,3) < (0,2,3,1) - (0,2,3,1,4) (0,2,4) < (0,2,4,1) - (0,2,3,4,1) (0,2,4) < (0,2,4,3) - (0,2,4,3,1)^* (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,1,4) - (0,3,1,4,2) (0,3,2,4) (0,3,2,4)$$

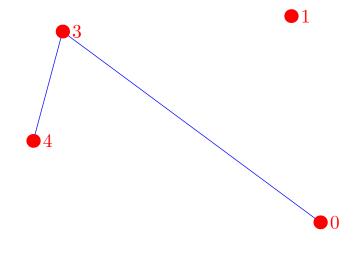
bound = 302.31length = 324.99



(0,3,2,4)

$$(0,1) \left\langle \begin{array}{c} (0,1,2) < (0,1,2,3) & \dots & (0,1,2,3,4)^* \\ (0,1) \left\langle \begin{array}{c} (0,1,2,4) & \dots & (0,1,2,3,4)^* \\ (0,1,3) < (0,1,3,2) & \dots & (0,1,3,2,4) \\ (0,1,4) < (0,1,4,2) & \dots & (0,1,4,2,3) \\ (0,1,4) < (0,1,4,3) & \dots & (0,1,4,2,3) \\ (0,2,1) < (0,2,1,3) & \dots & (0,2,1,3,4) \\ (0,2) \left\langle \begin{array}{c} (0,2,1,3) & \dots & (0,2,1,3,4) \\ (0,2,1,4) & \dots & (0,2,1,4,3) \\ (0,2,3,4) & \dots & (0,2,3,4,1) \\ (0,2,4) < (0,2,4,3) & \dots & (0,2,3,4,1) \\ (0,2,4) < (0,2,4,3) & \dots & (0,2,4,3,1)^* \\ (0,3,1) < (0,3,1,2) & \dots & (0,3,1,2,4) \\ (0,3,2) < (0,3,2,4) & \dots & (0,3,1,4,2) \\ \end{array} \right.$$

bound = 302.31length = 127.52

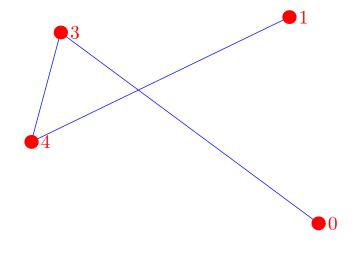


 $\bigcirc 2$

(0,3,4)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* - (0,1,2,3,4)^* - (0,1,2,4,3)^* - (0,1,2,4,3)^* - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,4,2)^* - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,3,4,1) - (0,2,4,1,3) - (0,2,4,1,3) - (0,2,4,3,1)^* - (0,3,1) < (0,3,1,2) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,4,2) - (0,3,1,4,2) - (0,3,2,4) - (0,3,2,4) - (0,3,2,4) - (0,3,2,4) - (0,3,4,1) - (0,3,4,1) - (0,3,4,1) - (0,3,4,1) - (0,3,4,4) - (0,4,4,4) - (0,4,4,4) - (0,4,4,4) - (0,4,4,4) - (0,4,4,4) - (0,4,4,4) - (0,4$$

bound = 302.31length = 211.72

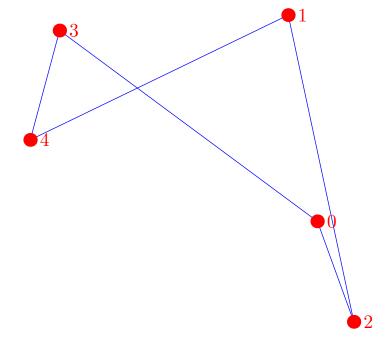


 $\bigcirc 2$

(0,3,4,1)

$$(0,1) \left\langle \begin{array}{c} (0,1,2,3) & ----- & (0,1,2,3,4)^* \\ (0,1) \left\langle \begin{array}{c} (0,1,2,4) & ---- & (0,1,2,3,4)^* \\ (0,1,3) & < (0,1,3,2) & ---- & (0,1,3,2,4) \\ (0,1,4) & < (0,1,3,4) & ---- & (0,1,3,4,2)^* \\ (0,2,1) & < (0,2,1,3) & ---- & (0,1,4,3,2) \\ (0,2,1) & < (0,2,1,3) & ---- & (0,2,1,3,4) \\ (0,2,1) & < (0,2,1,4) & ---- & (0,2,1,4,3) \\ (0,2,4) & < (0,2,3,4) & ---- & (0,2,3,4,1) \\ (0,2,4) & < (0,2,4,1) & ---- & (0,2,4,1,3) \\ (0,2,4) & < (0,2,4,3) & ---- & (0,2,4,3,1)^* \\ (0,3,1) & < (0,3,1,2) & ---- & (0,3,1,2,4) \\ (0,3) \left\langle \begin{array}{c} (0,3,2) & < (0,3,2,1) \\ (0,3,2,4) & ---- & (0,3,4,1,2) \\ \end{array} \right.$$

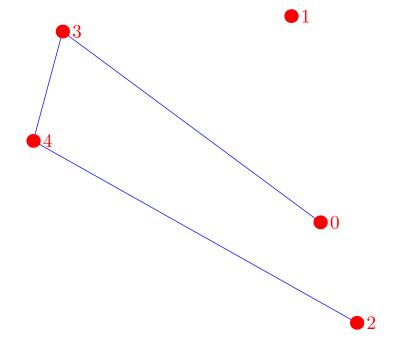
bound = 302.31length = 335.3



(0,3,4,1,2)

$$(0,1) < (0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1) < (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,1,4) < (0,1,4,3) - (0,1,4,3,2) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) - (0,2,3,1) - (0,2,1,3,4) (0,2,4) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,2) < (0,3,2,1) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,3,4,2) - (0,3,4,1,2)$$

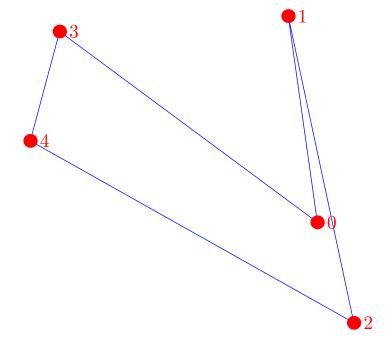
bound = 302.31length = 236.6



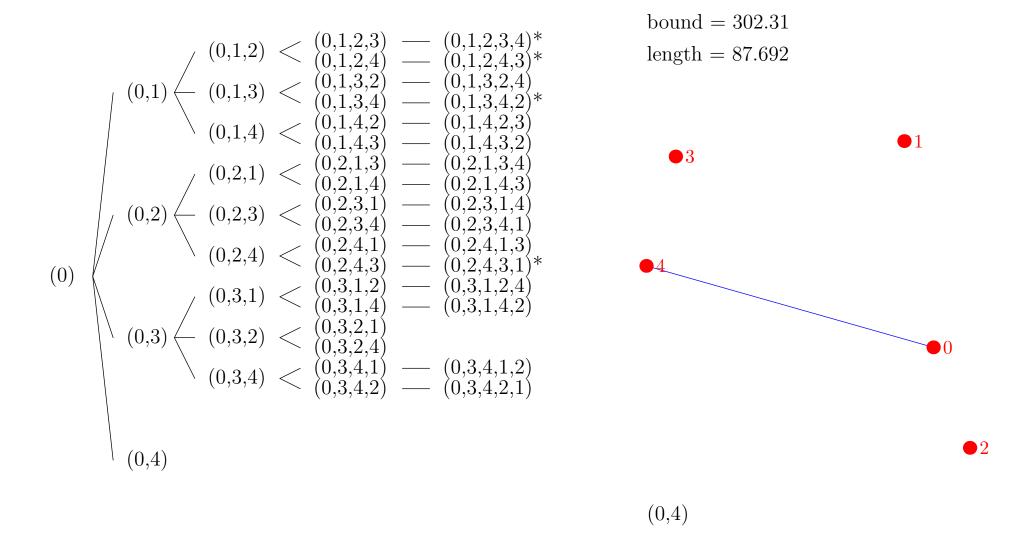
(0,3,4,2)

$$(0,1) < (0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,4) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,1,4) - (0,2,1,4,3) (0,2,4) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4) < (0,2,4,3) - (0,2,4,3,1)^* (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,2) < (0,3,2,4) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,3,4,2,1) (0,3,4,2,1)$$

bound = 302.31length = 389.98



(0,3,4,2,1)



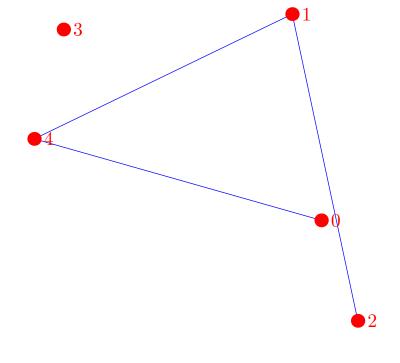
$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* - (0,1,2,3,4)^* - (0,1,2,4,3)^* - (0,1,2,4,3)^* - (0,1,2,4,3)^* - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,4,2)^* - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,3,2) - (0,2,1,3,4) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,3,4,1) - (0,2,3,4,1) - (0,2,4,3,1)^* - (0,2,4,3,1) - (0,2,4,3,1) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,4,2) - (0,3,2,4) - (0,3,4,1,2) - (0,3,4,1,2) - (0,3,4,1,2) - (0,3,4,2,1) - (0,4,1) - (0,4,1)$$

bound = 302.31length = 171.9**3 2**

(0,4,1)

$$(0,1) < (0,1,2,3) = (0,1,2,3,4)^* (0,1,2,4,3) = (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) = (0,1,3,2,4) (0,1,3,4) = (0,1,3,4,2)^* (0,1,4) < (0,1,4,2) = (0,1,4,2,3) (0,2,1) < (0,2,1,3) = (0,2,1,3,4) (0,2,1) < (0,2,1,4) = (0,2,1,4,3) (0,2,1) < (0,2,3,4) = (0,2,3,1,4) (0,2,4) < (0,2,3,4) = (0,2,3,4,1) (0,2,4) < (0,2,4,3) = (0,2,4,3,1)^* (0,3,1) < (0,3,1,2) = (0,3,1,2,4) (0,3,4) < (0,3,2,4) (0,3,4) < (0,3,4,1) = (0,3,4,1,2) (0,4,1) < (0,4,1,2) (0,4,1) < (0,4,1,2)$$

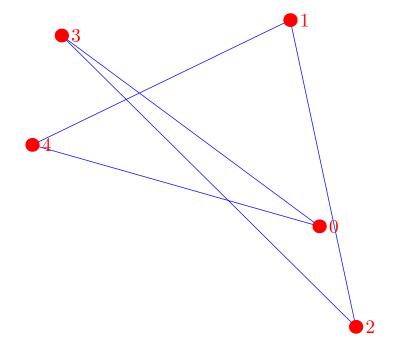
bound = 302.31length = 264.07



(0,4,1,2)

$$(0,1) < (0,1,2,3) = (0,1,2,3,4)^* (0,1,2,4) = (0,1,2,3,4)^* (0,1,2,4) = (0,1,2,3,4)^* (0,1,3) < (0,1,3,2) = (0,1,3,2,4) (0,1,3,4) = (0,1,3,4,2)^* (0,1,4) < (0,1,4,2) = (0,1,4,2,3) (0,2,1) < (0,2,1,3) = (0,2,1,3,4) (0,2,1) < (0,2,1,4) = (0,2,1,4,3) (0,2,1) < (0,2,3,4) = (0,2,3,4,1) (0,2,4) < (0,2,3,4) = (0,2,3,4,1) (0,2,4) < (0,2,4,3) = (0,2,4,3,1)^* (0,3,1) < (0,3,1,2) = (0,3,1,2,4) (0,3,2) < (0,3,2,4) (0,3,4) < (0,3,4,1) = (0,3,4,1,2) (0,3,4) < (0,3,4,2) = (0,3,4,2,1) (0,4,1) < (0,4,1,2) = (0,4,1,2,3)$$

bound = 302.31 length = 479.98



(0,4,1,2,3)

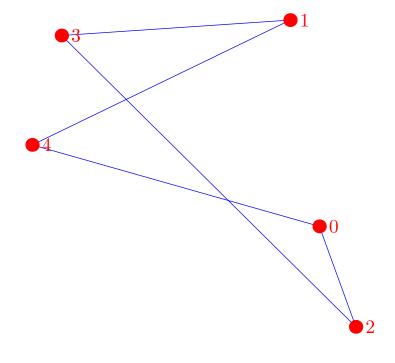
$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* - (0,1,2,3,4)^* - (0,1,2,4,3)^* - (0,1,2,4,3)^* - (0,1,2,4,3)^* - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,2,4) - (0,1,3,4,2)^* - (0,1,4,2) - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,1,4,2,3) - (0,2,1,3,4) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,1,4,3) - (0,2,3,4,1) - (0,2,3,4,1) - (0,2,3,4,1) - (0,2,4,3,1) - (0,2,4,3,1) - (0,2,4,3,1) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,2,4) - (0,3,1,4,2) - (0,3,1,4,2) - (0,3,2,4) - (0,3,2,4) - (0,3,4,2) - (0,3,4,2,1) - (0,3,4,2,1) - (0,4,1,2,3) - (0$$

bound = 302.31length = 239.23

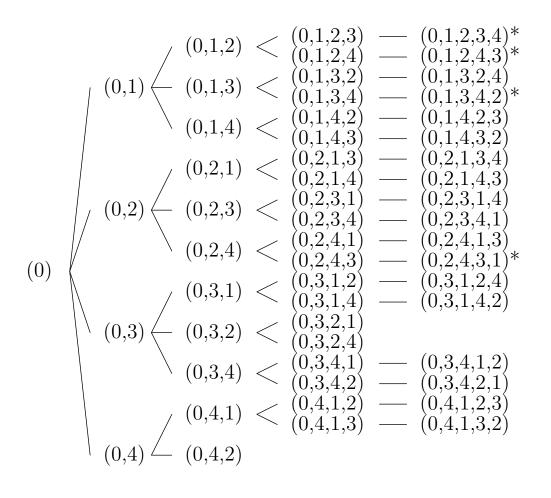
(0,4,1,3)

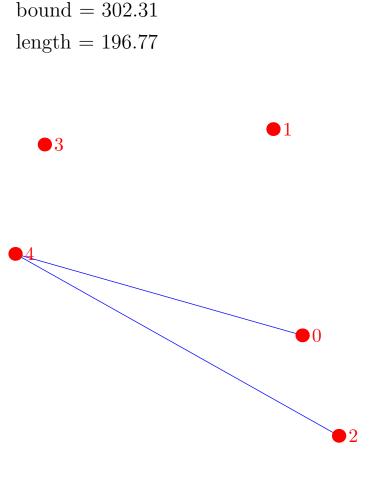
$$(0,1) < (0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,1,4) < (0,1,4,3) - (0,1,4,3,2) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) < (0,2,1,4) - (0,2,1,4,3) (0,2,1) < (0,2,3,4) - (0,2,3,1,4) (0,2,4) < (0,2,3,4) - (0,2,3,1,4) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3) < (0,3,1,4) - (0,3,1,2,4) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,3,4) < (0,3,4,2) - (0,3,4,2,1) (0,4,1) < (0,4,1,2) - (0,4,1,2,3) (0,4,1,3,2) (0,4,1,3,2)$$

bound = 302.31length = 392.31



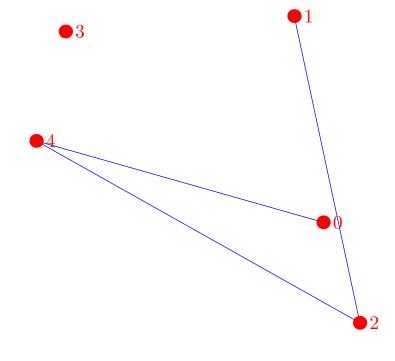
(0,4,1,3,2)





$$(0,1) \left\langle \begin{array}{c} (0,1,2) < (0,1,2,3) & \dots & (0,1,2,3,4)^* \\ (0,1,2,4) & \dots & (0,1,2,4,3)^* \\ (0,1,3) < (0,1,3,2) & \dots & (0,1,3,2,4) \\ (0,1,4) < (0,1,4,2) & \dots & (0,1,4,2,3) \\ (0,2,1) < (0,2,1,3) & \dots & (0,2,1,3,4) \\ (0,2) \left\langle \begin{array}{c} (0,2,1,3) & \dots & (0,2,1,3,4) \\ (0,2,3) < (0,2,3,4) & \dots & (0,2,1,4,3) \\ (0,2,4) < (0,2,3,4) & \dots & (0,2,3,4,1) \\ (0,2,4) < (0,2,4,3) & \dots & (0,2,4,1,3) \\ (0,3,1) < (0,3,1,2) & \dots & (0,3,1,2,4) \\ (0,3) \left\langle \begin{array}{c} (0,3,2) < (0,3,2,1) \\ (0,3,4) < (0,3,4,1) & \dots & (0,3,4,1,2) \\ (0,3,4,2) & \dots & (0,3,4,2,1) \\ (0,4,1) < (0,4,1,2) & \dots & (0,4,1,2,3) \\ (0,4,2) & (0,4,2,1) \end{array} \right.$$

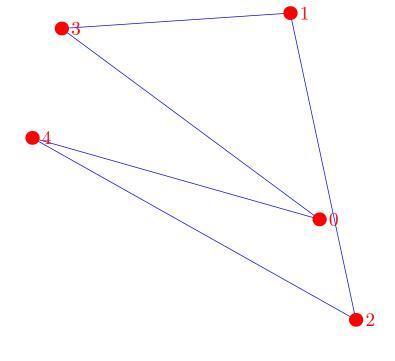
bound = 302.31length = 288.95



(0,4,2,1)

$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,3) + (0,1,3,4) - (0,1,3,4,2)^* (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,1,4,3) - (0,1,4,3,2) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,1) + (0,2,1,4) - (0,2,1,4,3) (0,2,4) < (0,2,3,1) - (0,2,3,1,4) (0,2,4) < (0,2,4,1) - (0,2,4,1,3) (0,2,4) < (0,2,4,3) - (0,2,4,3,1)^* (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,1,4) - (0,3,1,4,2) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,3,4,2) - (0,3,4,2,1) (0,4,1) < (0,4,1,2) - (0,4,1,2,3) (0,4,1) < (0,4,1,3) - (0,4,1,3,2) (0,4,2) < (0,4,2,1) - (0,4,2,1,3)$$

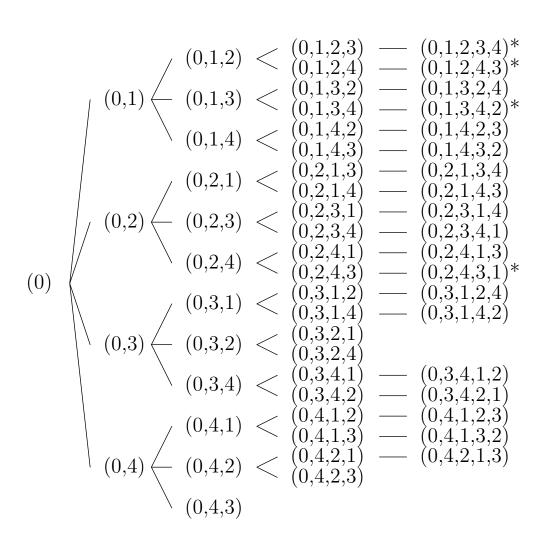
bound = 302.31length = 450.52

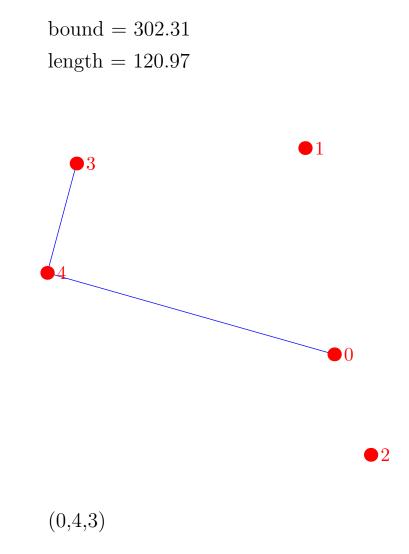


(0,4,2,1,3)

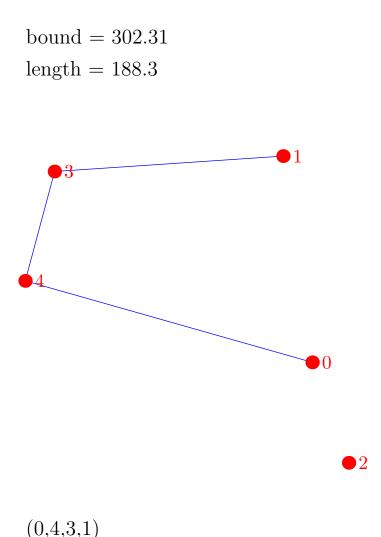
$$(0,1,2) < (0,1,2,3) - (0,1,2,3,4)^* (0,1,2,4) - (0,1,2,4,3)^* (0,1,3) < (0,1,3,2) - (0,1,3,2,4) (0,1,4) < (0,1,4,2) - (0,1,4,2,3) (0,2,1) < (0,2,1,3) - (0,2,1,3,4) (0,2,4) < (0,2,3,4) - (0,2,3,4,1) (0,2,4) < (0,2,4,3) - (0,2,4,1,3) (0,3,1) < (0,3,1,2) - (0,3,1,2,4) (0,3,4) < (0,3,2,4) - (0,3,4,1,2) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,3,4) < (0,3,4,1) - (0,3,4,1,2) (0,4,1) < (0,4,1,2) - (0,4,1,2,3) (0,4,2) < (0,4,2,1) - (0,4,2,1,3) (0,4,2,3) - (0,4,2,1,3) (0,4,2,1) - (0,4,2,1,3) (0,4,2,1,3) - (0,4,2,1,3)$$

bound = 302.31length = 318.44





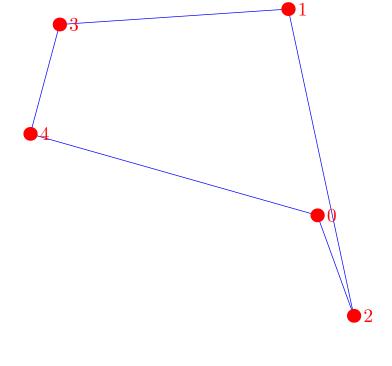
$$(0,1,2) < \begin{pmatrix} 0,1,2,3 & & & & & & & & & \\ 0,1,2,4 & & & & & & & & \\ 0,1,2,4,3 & & & & & & & \\ 0,1,2,4,3 & & & & & & & \\ & & & & & & & & \\ & & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & \\ & & & \\ & & \\ & & & \\ & & \\ & & & \\ & & \\ & & \\ & & & \\$$

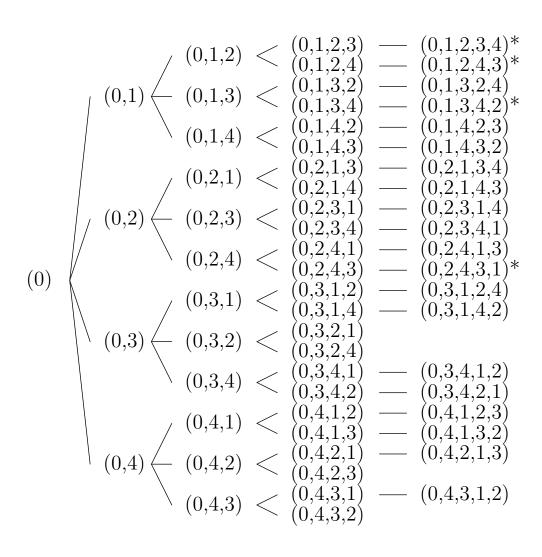


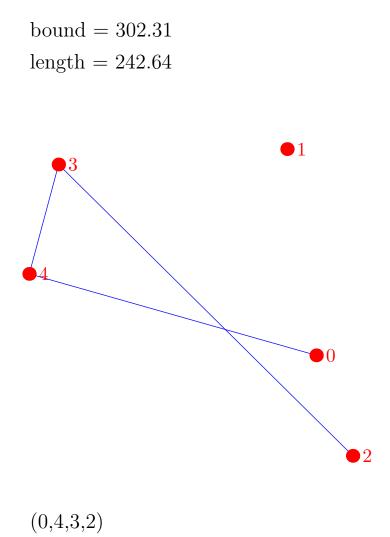
$$(0,1,2) < \begin{pmatrix} 0,1,2,3 \end{pmatrix} & ---- & (0,1,2,3,4)^* \\ (0,1,2,4) & --- & (0,1,2,4,3)^* \\ (0,1,3) < & (0,1,3,2) & --- & (0,1,3,2,4) \\ (0,1,4) < & (0,1,3,4) & --- & (0,1,4,2,3) \\ (0,1,4) < & (0,1,4,2) & --- & (0,1,4,2,3) \\ (0,2,1) < & (0,2,1,3) & --- & (0,2,1,3,4) \\ (0,2,1) < & (0,2,1,4) & --- & (0,2,1,4,3) \\ (0,2,1) < & (0,2,3,1) & --- & (0,2,3,1,4) \\ (0,2,4) < & (0,2,3,1) & --- & (0,2,3,4,1) \\ (0,2,4) < & (0,2,4,1) & --- & (0,2,4,1,3) \\ (0,2,4) < & (0,2,4,3) & --- & (0,2,4,3,1)^* \\ (0,3,1) < & (0,3,1,2) & --- & (0,3,1,2,4) \\ (0,3,1) < & (0,3,1,2) & --- & (0,3,1,2,4) \\ (0,3) < & (0,3,2,1) & --- & (0,3,4,1,2) \\ (0,3,4) < & (0,3,4,2) & --- & (0,3,4,2,1) \\ (0,4,1) < & (0,4,1,2) & --- & (0,4,1,2,3) \\ (0,4) < & (0,4,2,1) & --- & (0,4,2,1,3) \\ (0,4,3) < & (0,4,3,1) & --- & (0,4,3,1,2) \\ \end{pmatrix}$$

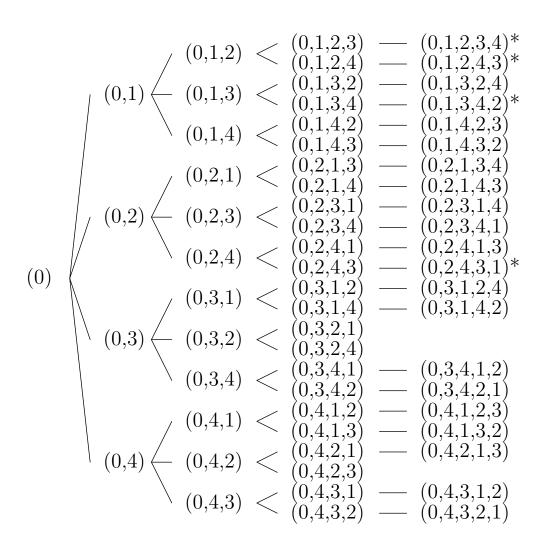
bound = 302.31 length = 311.88

(0,4,3,1,2)

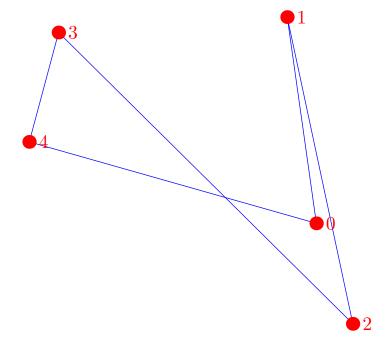






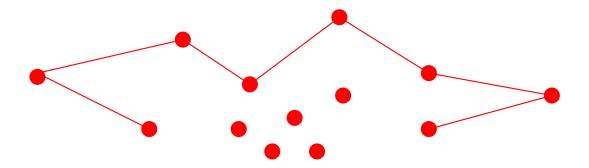


bound = 302.31length = 396.02



Bound on Partial Solution

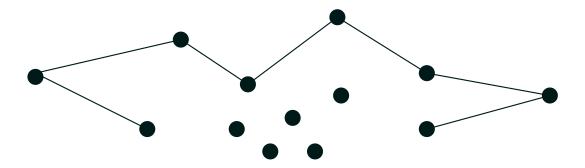
 We know that the partial solution has to include all the remaining cities



- We can use this to obtain a lower bound on the partial solution
- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this

Bound on Partial Solution

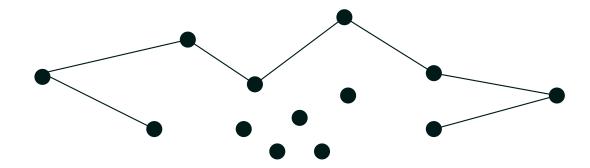
 We know that the partial solution has to include all the remaining cities



- We can use this to obtain a lower bound on the partial solution
- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this

Bound on Partial Solution

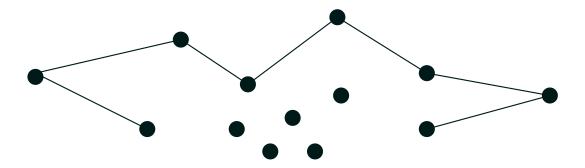
 We know that the partial solution has to include all the remaining cities



- We can use this to obtain a lower bound on the partial solution
- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this

Bound on Partial Solution

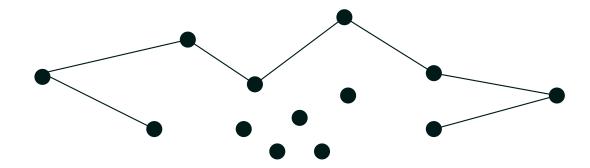
 We know that the partial solution has to include all the remaining cities



- We can use this to obtain a lower bound on the partial solution
- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this

Bound on Partial Solution

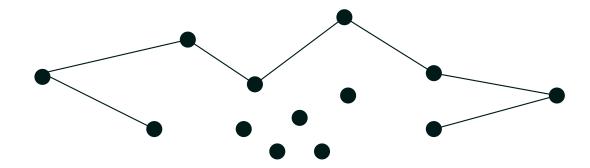
 We know that the partial solution has to include all the remaining cities



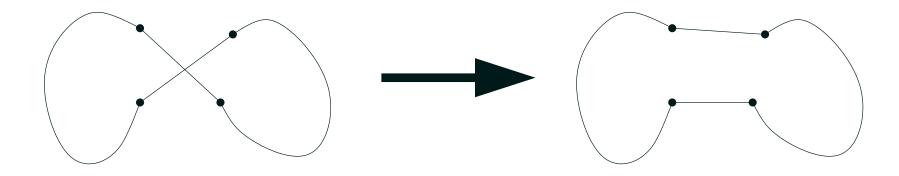
- We can use this to obtain a lower bound on the partial solution
- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this

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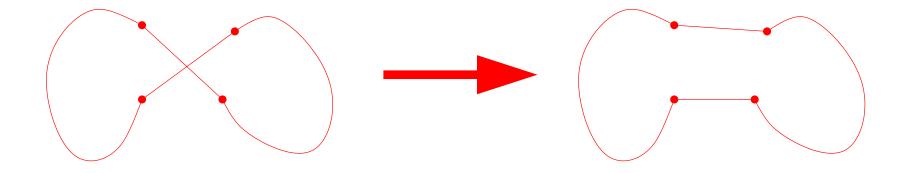
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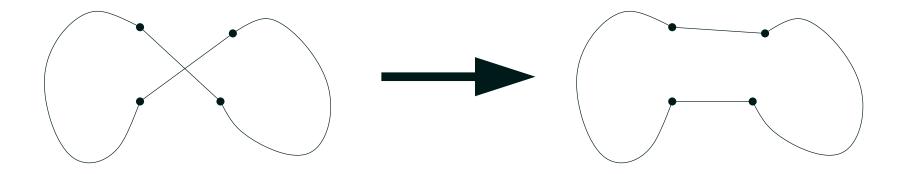
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- We know the remaining tour will go through each of the unvisited cities and the two edge cities
- In fact the remaining part of the tour is a spanning tree of these vertices (it connects all the vertices once and has no cycles)
- But we know a lower bound for this—the minimum spanning tree



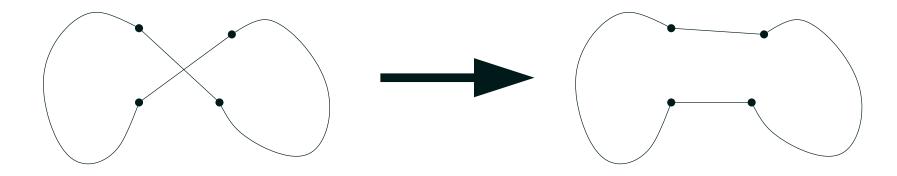
- In fact we can check that we cannot perform a 2-opt move
- We can also halve the search by considering only one direction—for example, by insisting we visit city 1 before city 2



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 It helps to start with a good bound

 We can use an incomplete heuristic algorithm to find a good solution which will act as a starting bound

 One very simple heuristic is a greedy algorithm **3**

4

•0

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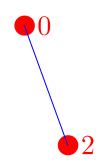
4

3

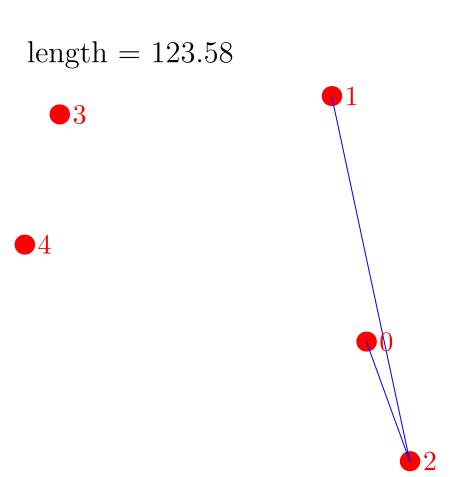
•0

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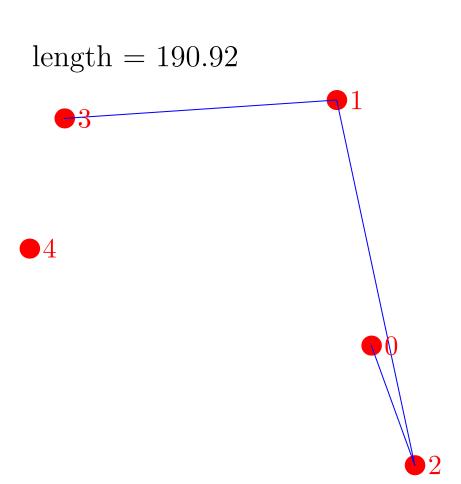




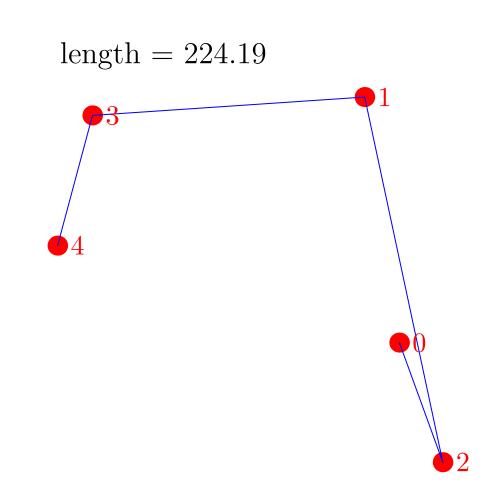
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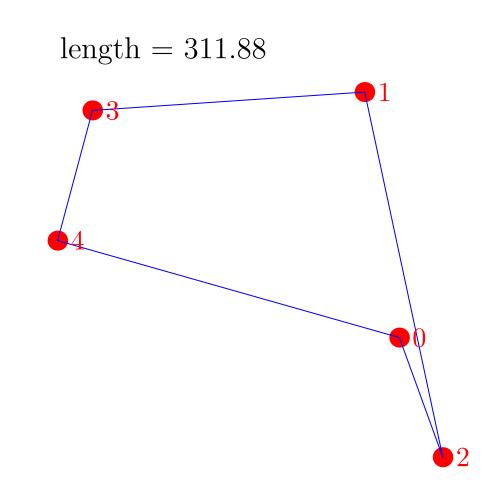
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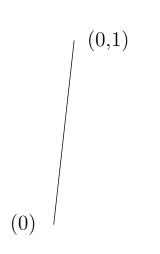


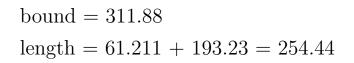
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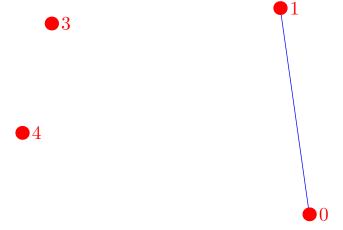


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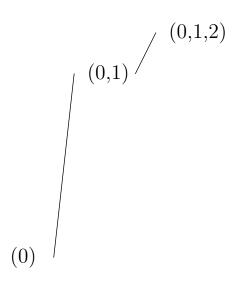


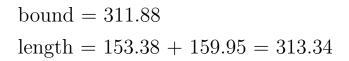


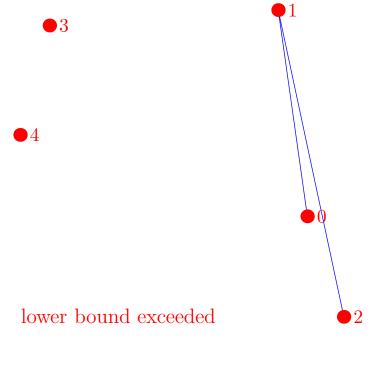




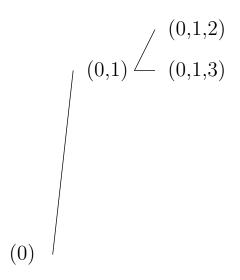
(0,1)

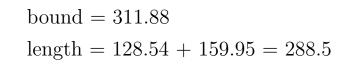


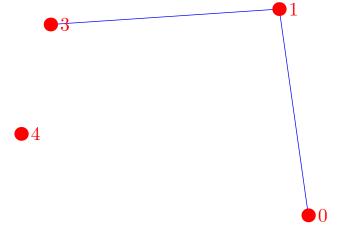




(0,1,2)

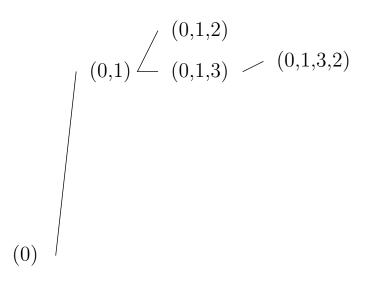






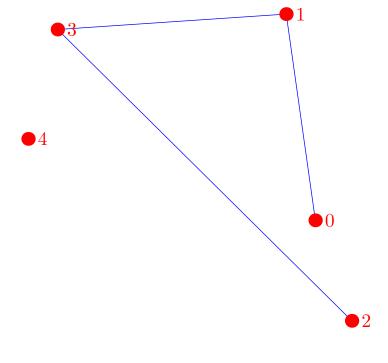
 $\bigcirc 2$

(0,1,3)

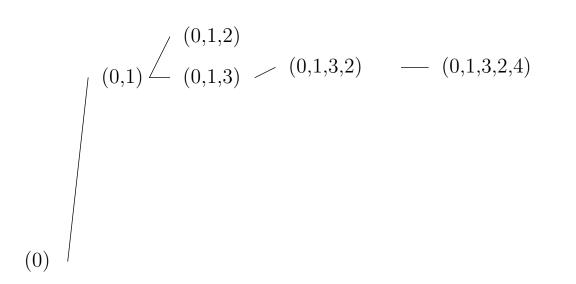


bound =
$$311.88$$

length = $250.21 + 31.41 = 281.63$

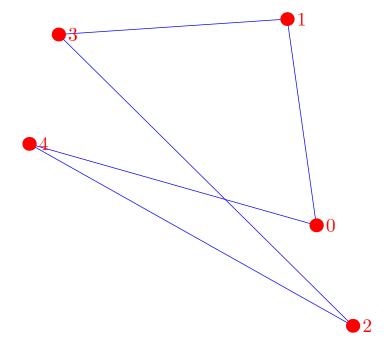


(0,1,3,2)

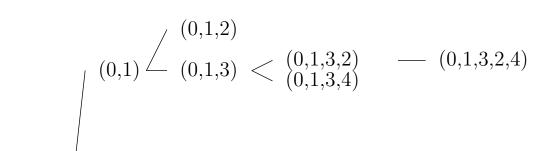


bound =
$$311.88$$

length = $446.99 + 0 = 446.99$

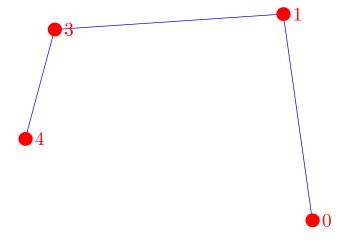


(0,1,3,2,4)



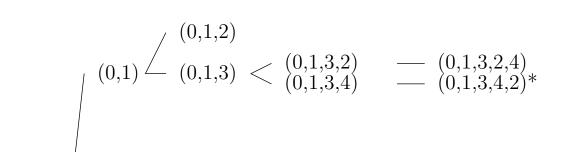
bound =
$$311.88$$

length = $161.82 + 31.41 = 193.23$



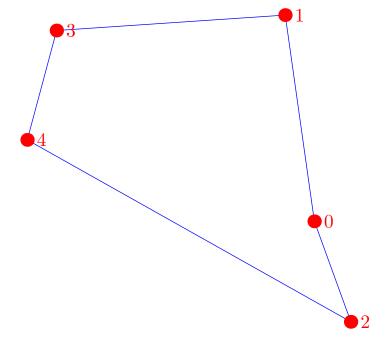
O2

(0,1,3,4)



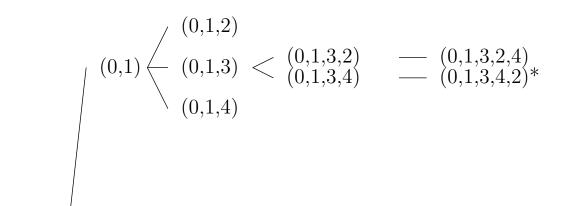
bound =
$$302.31$$

length = $302.31 + 0 = 302.31$



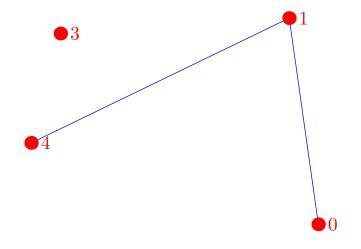
(0,1,3,4,2)

(0)



bound =
$$302.31$$

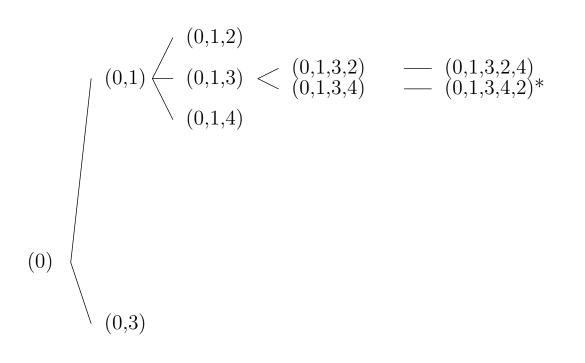
length = $145.41 + 159.95 = 305.37$



lower bound exceeded

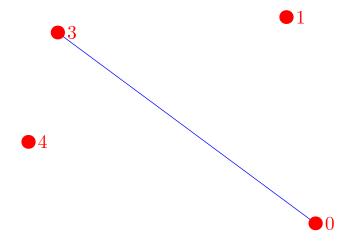
 $\bigcirc 2$

(0,1,4)



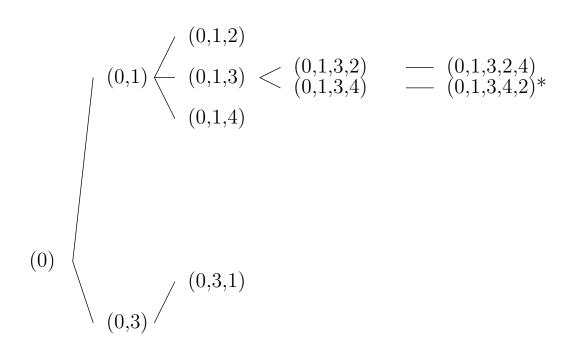
bound =
$$302.31$$

length = $94.244 + 193.23 = 287.47$



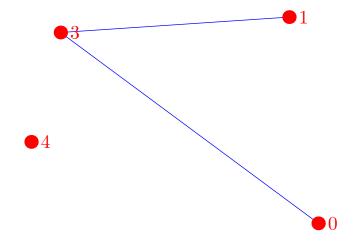
 $\bigcirc 2$

(0,3)



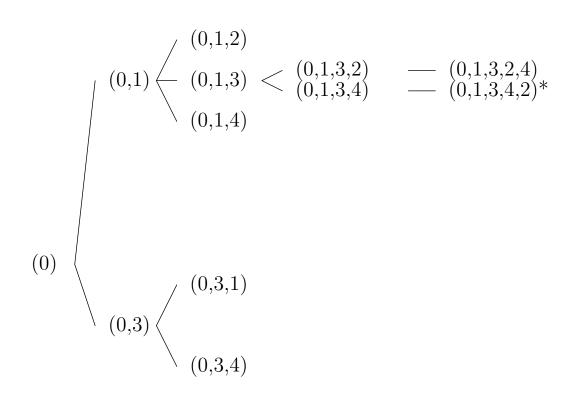
bound =
$$302.31$$

length = $161.58 + 159.95 = 321.53$



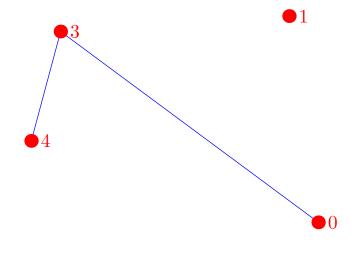
 $\bigcirc 2$

(0,3,1)



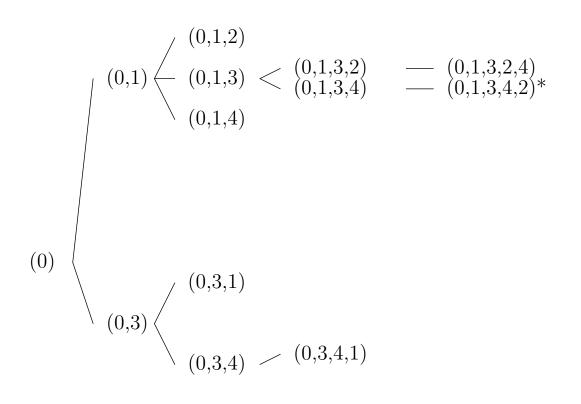
bound =
$$302.31$$

length = $127.52 + 159.95 = 287.47$



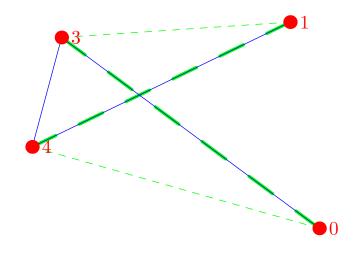
 $\bigcirc 2$

(0,3,4)



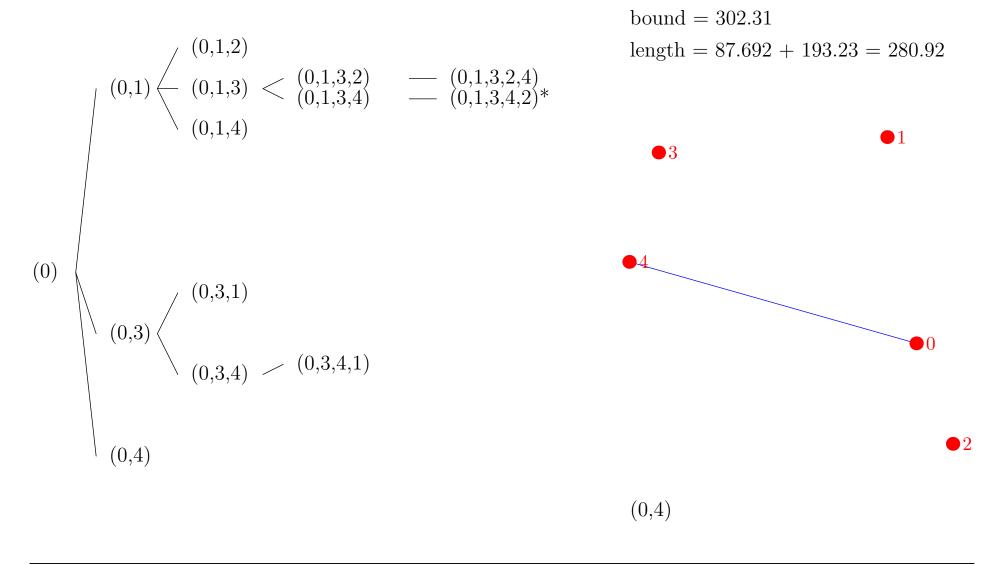
bound =
$$302.31$$

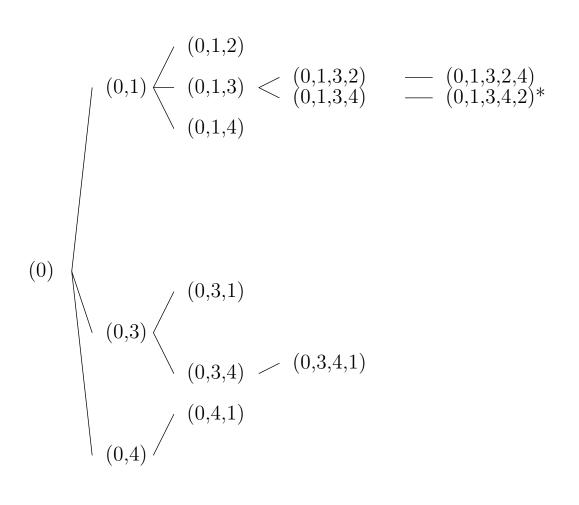
length = $211.72 + 61.211 = 272.93$



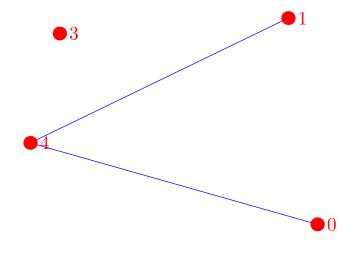
 $\bigcirc 2$

Not 2-opt (0,3,4,1)





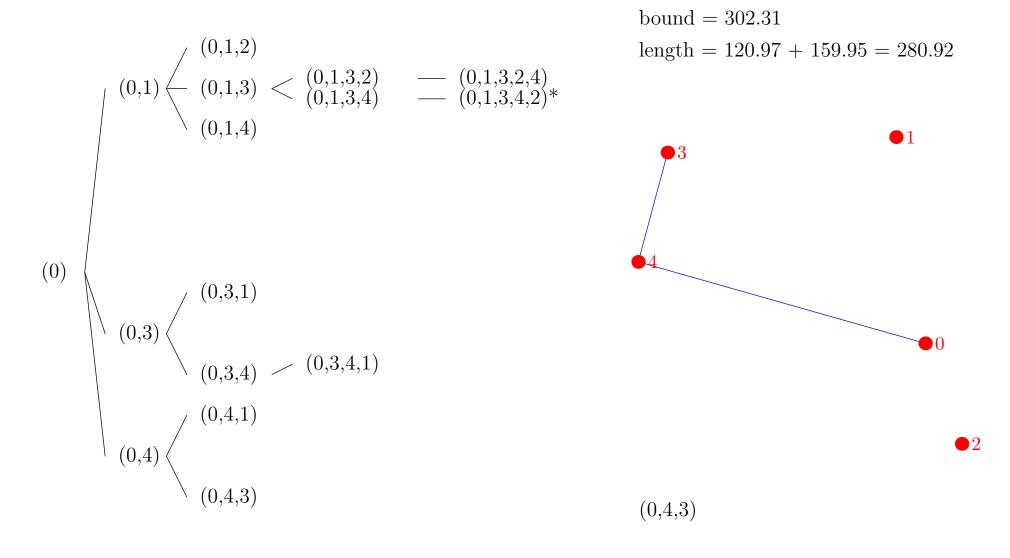
bound = 302.31length = 171.9 + 159.95 = 331.85

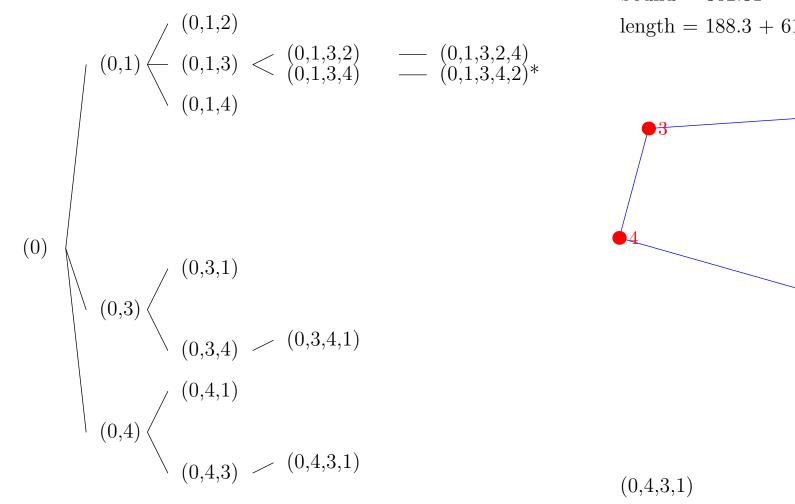


lower bound exceeded

1)

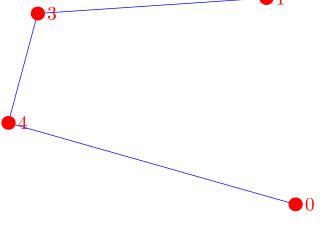
(0,4,1)

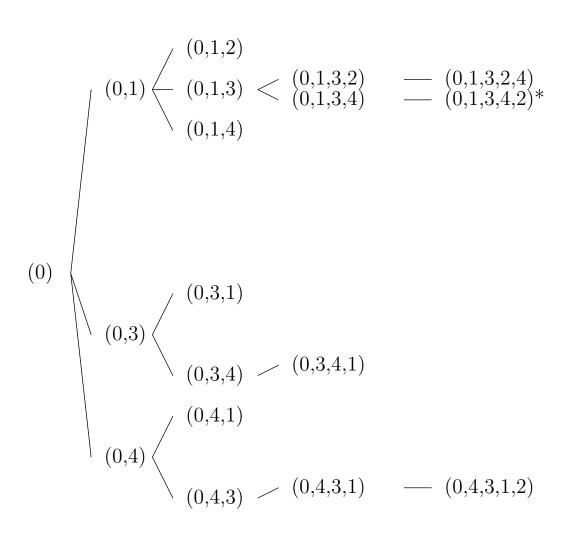




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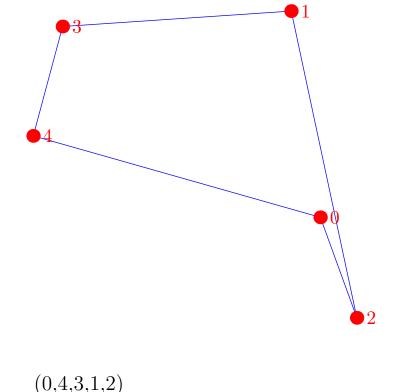
length = $188.3 + 61.211 = 249.51$





bound =
$$302.31$$

length = $311.88 + 0 = 311.88$



- Branch and bound works for many optimisation problems
- It's drawback is that you often end up still searching an exponentially large search space even though it might be massively faster than exhaustive enumeration
- To make it work well requires considerable work
- This is not an instantaneous algorithm, you may be waiting hours before you find a solution
- For really large problems branch and bound might be too slow

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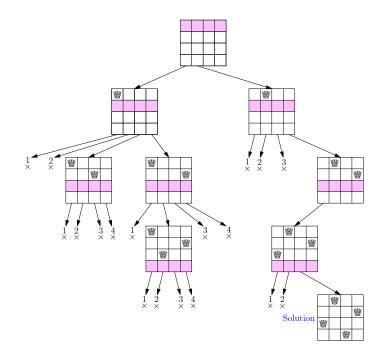
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Applications of Branch and Bound

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Outline

- 1. Search Trees
- 2. Backtracking
- 3. Branch and Bound
- 4. Search in Al



- Search is a big topic in Al
- The algorithms used depends on the information available
- A classic search scenario is when there is "heuristic" information which provides a hint as to where an optimal solution lies
- Algorithms such as A^* exist which will finds the best route given an (admissible) heuristic as efficiently as possible
- You should learn about this next year in Al

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- Search is also used to find the best action to take in planning problems and game playing (e.g. computer chess)
- Again it is useful to think in terms of a search tree
- Searching all paths on the search tree is usually infeasible
- Look for ways of pruning the search tree to focus on good moves
- Strategies include minimax and alpha-beta pruning

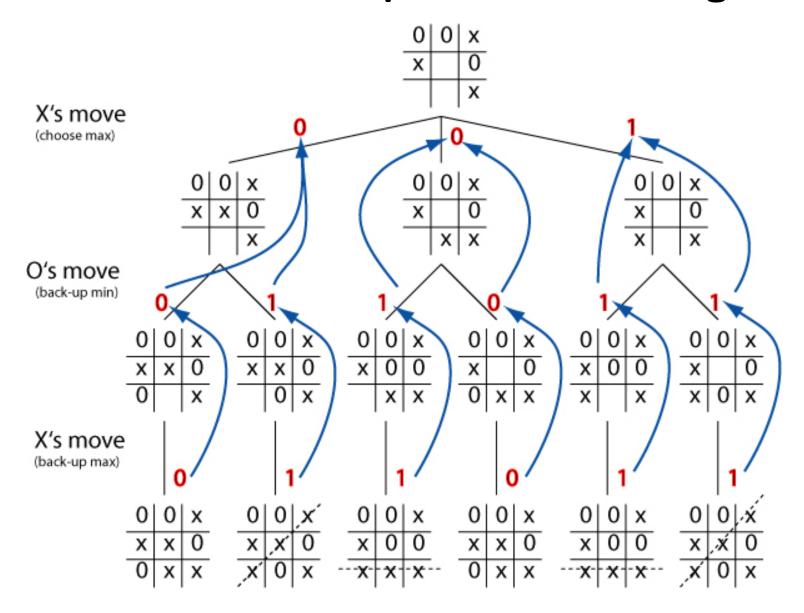
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Minimax with Alpha-Beta Prunning



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- It is helpful to consider the search space as a tree whose branch corresponds to possible actions
- Backtracking is useful in search trees with constraints
- For optimisation problems branch and bound uses backtracking and costs of partial solutions as constraints
- Widely applicable, but can take too long

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