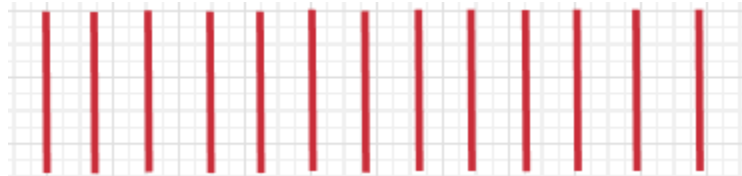


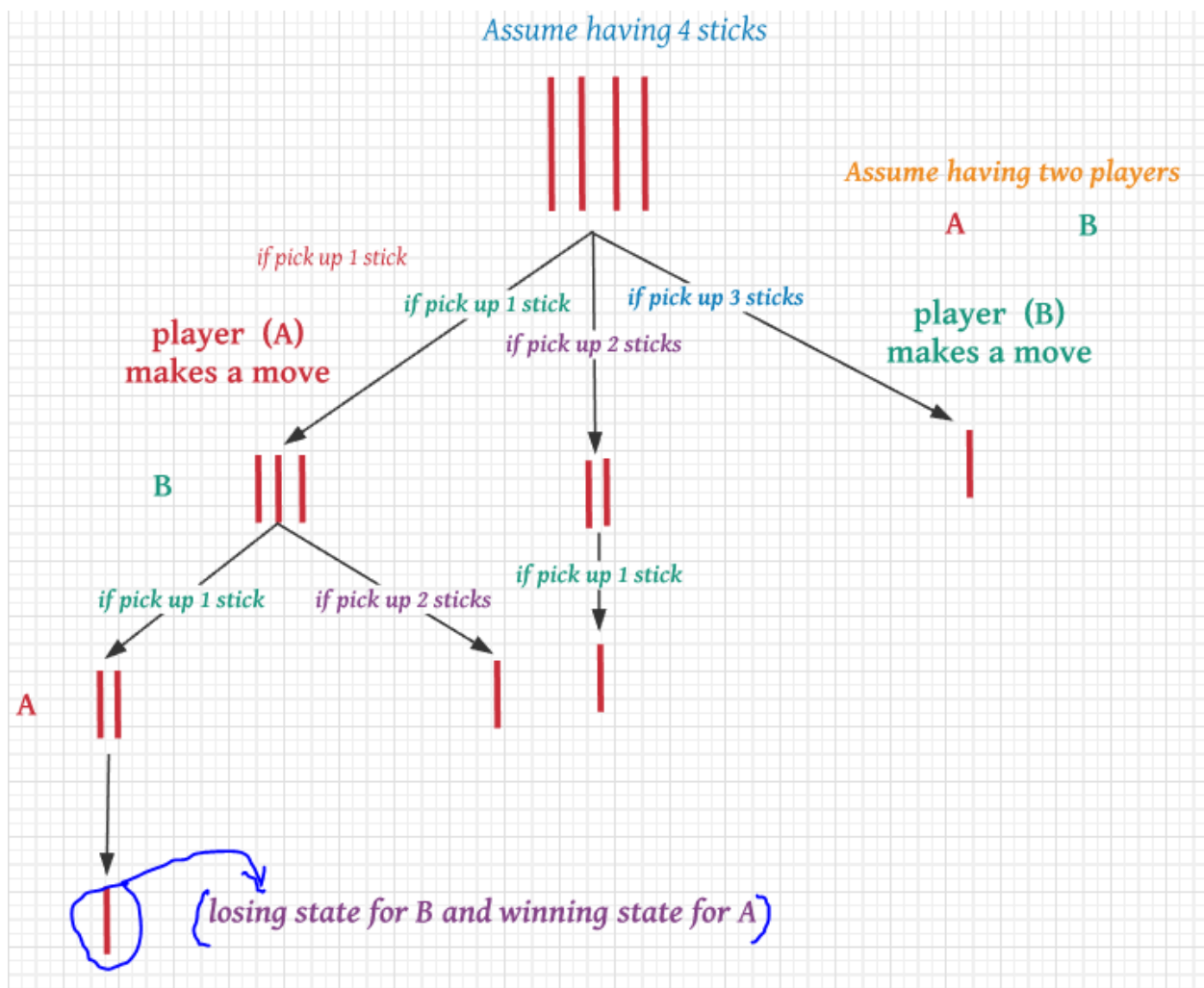
Searching Game trees

- Represent a situation where is two parities involve

Game: assume having 13 sticks as showing

Game rules

- Player can take 1, 2, or 3 sticks onetime
- The player who takes the last sticks loses the game



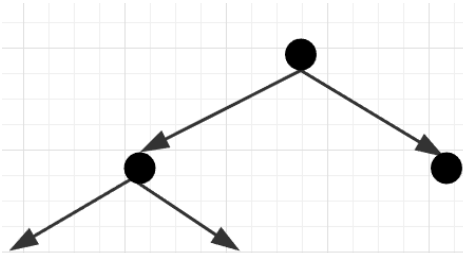
Search Game tree

Basic concept

1. Min /Max

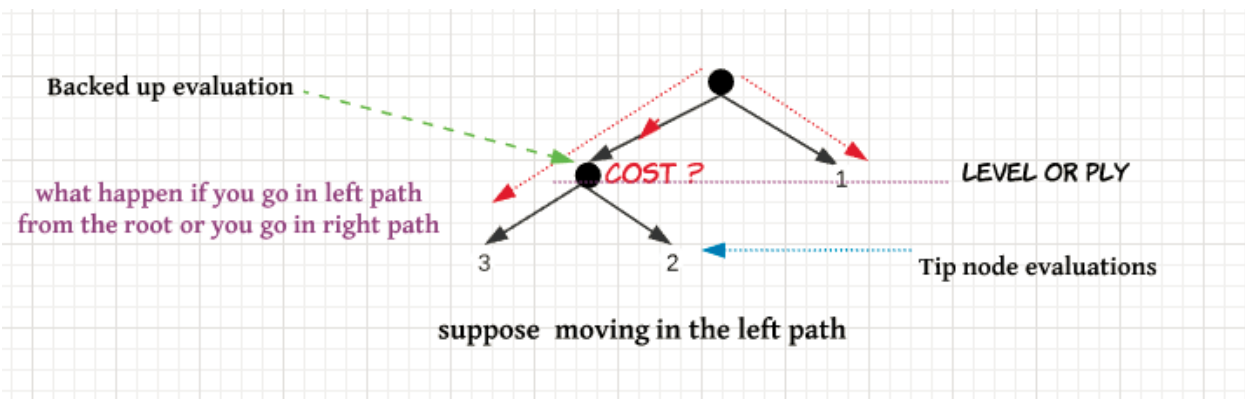
Maximizing player (Max)

Minimizing player (Min)



Player A tries to maximize her/his winner and minimize the winner for player B

Player B tries to maximize her/his winner and minimize the winner for Player A



There are two types of evaluation associated with node

1. Tip node evaluation (reward)
2. Backed up evaluation. (internal node of the tree)

To compute tip and backed up

- Use the min/max if knowing the tip node evaluation.

Alpha beta algorithm is about Reduce the number of evaluation

- For the max node back up value is the max of children values
- For the min nodes backup value is the min of children values

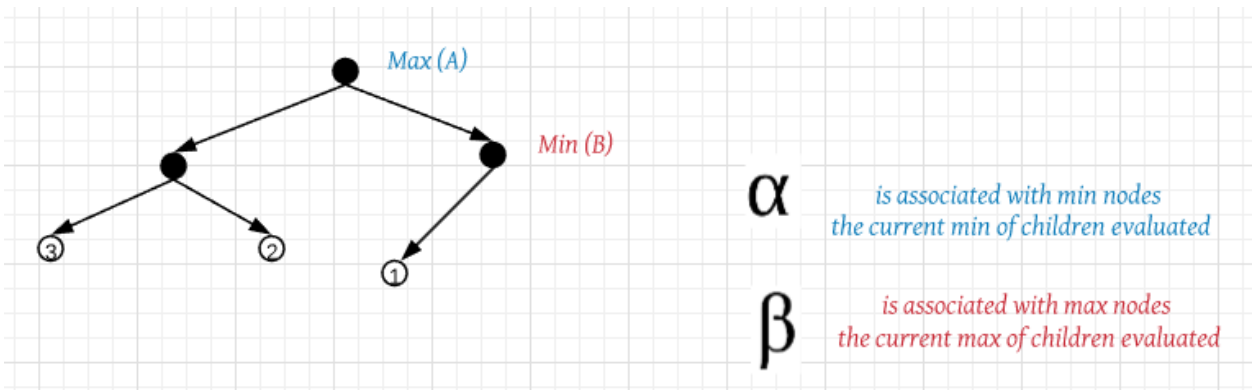
A! WINS the tip node reward is 1

B ! WINS the tip node reward is -1

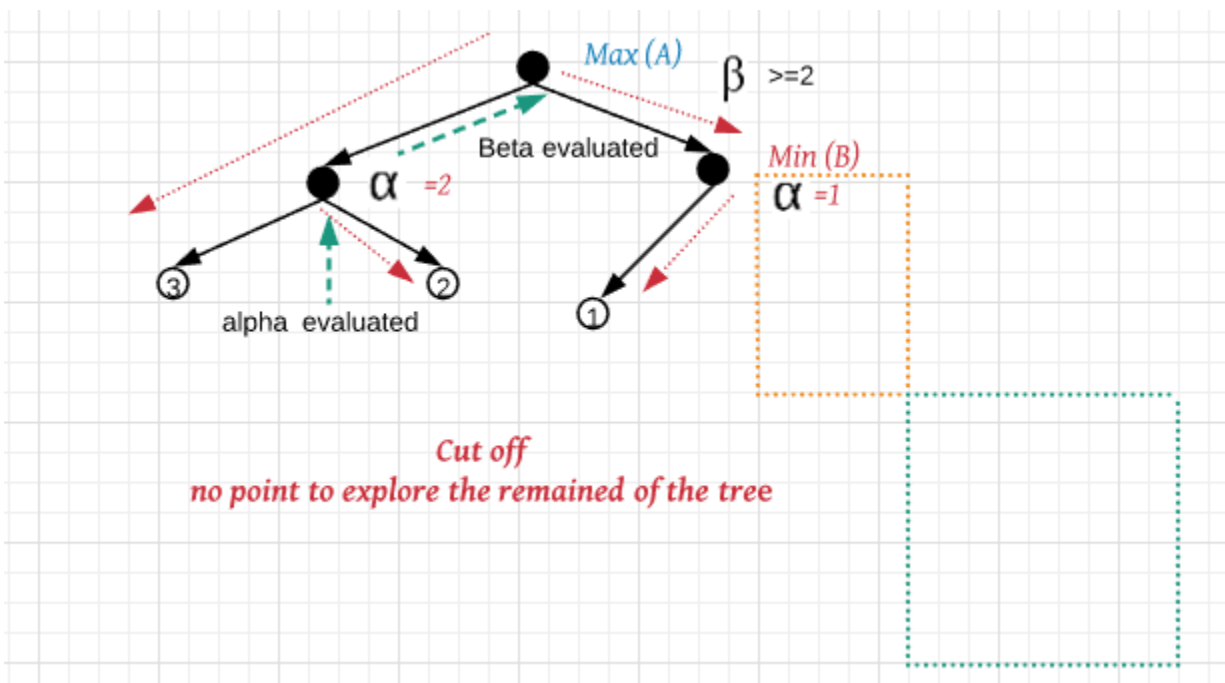
The diagram illustrates a game tree for a minimax problem. The root node is labeled $\text{MAX}(A)$ and has three branches. The left branch leads to a node labeled $\text{MIN}(B)$, which has three branches. The middle branch leads to a node labeled $\text{MAX}(B)$, which has two branches. The right branch leads to a node labeled $A! + 1$. The diagram includes various annotations such as Min , Max , $\text{MIN}(B)$, $\text{MAX}(B)$, $A! + 1$, $B! - 1$, and $A! + 1$.

What is good about it and what is bad about it?

It is good about the computation complex problem (game of checker and chess)

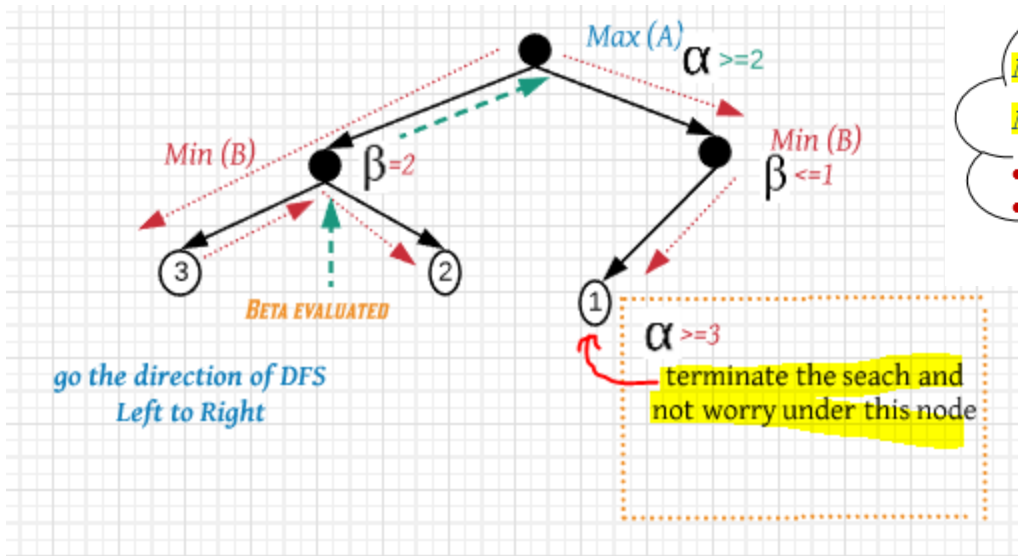


Continue exploring search tree DFS left first. Always do in DFS first and left to right direction



Continue with game tree

Alpha and beta pruning (search algorithm) $\alpha\beta$



Correction!!!

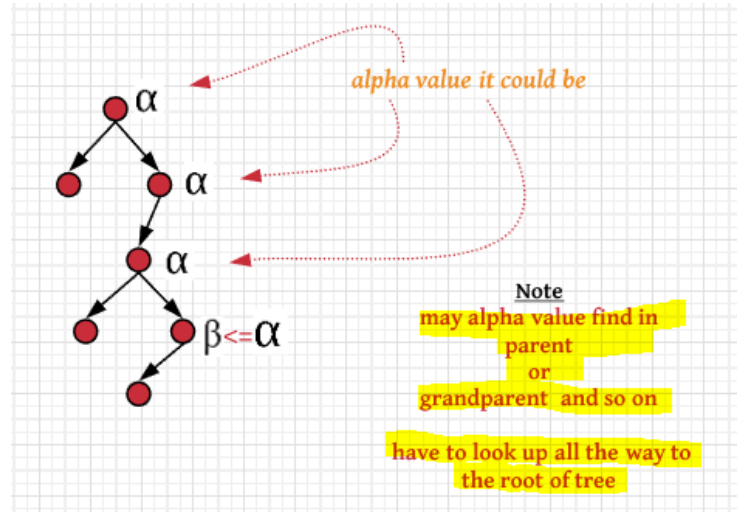
Max: α value (max of the children values)

Min: β value (min of children values)

- The alpha value never decrease
- The Beta value never increase

α/β pruning

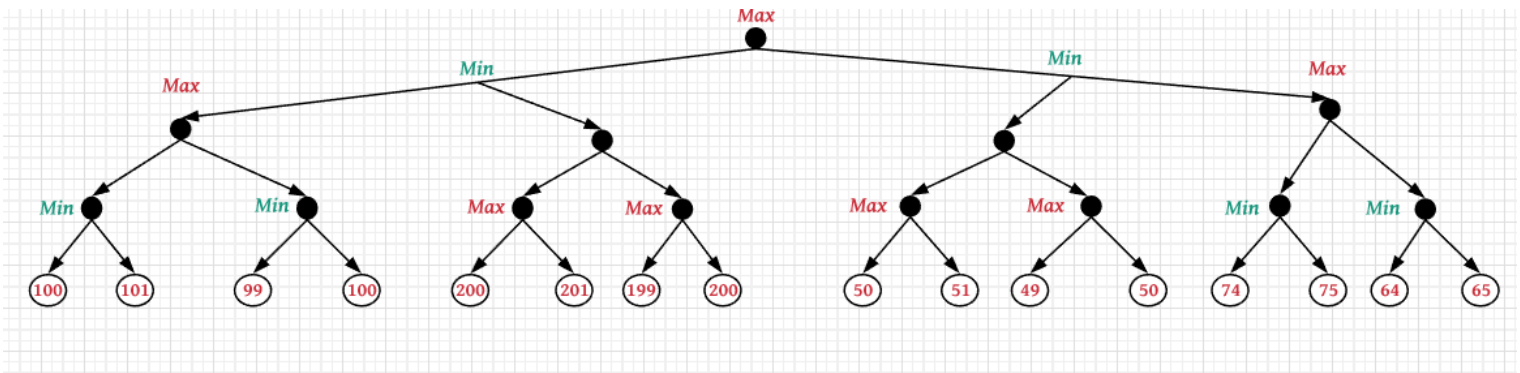
1. Searching can be discontinued below any **Min node** whose $\beta \leq \alpha$ of any of its **ancestors**. (alpha cut-off)



2. Search can be discontinued below an **Max node** whose $\alpha \geq \beta$ of any of its **ancestors** (beta cut-off)

not need to explore **L** because the value of **L (7)** is greater than **K(0)** and needs to get min value. it is **cut-off**

Using alpha beta pruning to find the winning path

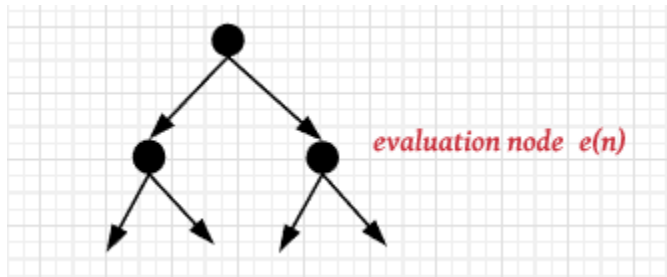


What is the limitation?

If have 16 parallel processors, it is easy to evaluate by using alpha beta.

Example

To demonstrate that partial state of the game can be assigned an evaluation to help alpha beta pruning.

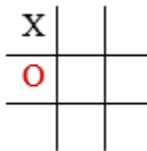


Example

Tic Tac Toe game

State

- X for Max player
- O for Min player
- $e(s)$ if S is a win for max
then $e(s) = \infty$
- $e(s)$ if s is a win for min
then $e(s) = -\infty$

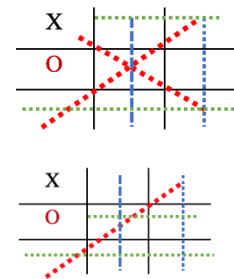


$e(s) = \{\text{number of rows, columns, and diagonal open for max}\}$

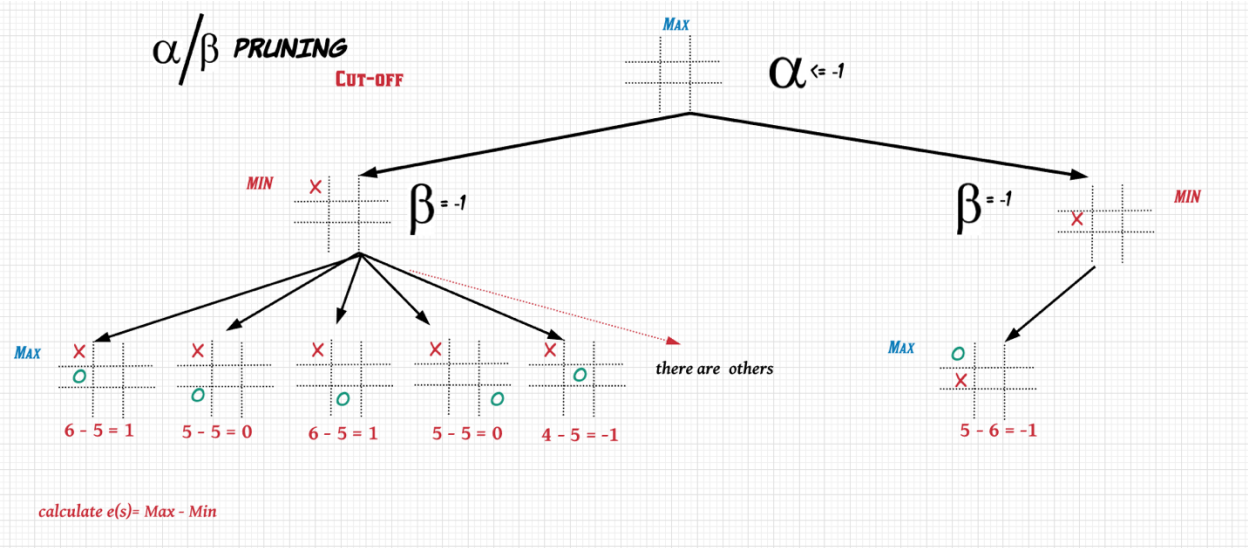
$e(s) = -\{\text{number of rows, columns, and diagonal open for min}\}$

- calculate the number of rows, column and diagonal for X
2 rows, 2 columns, 2 diagonals. $\rightarrow 2+2+2 = 6$
- calculate the number of rows, column and diagonal for O
2 rows, 2 columns, 1 diagonal. $\rightarrow 2+2+1 = 5$

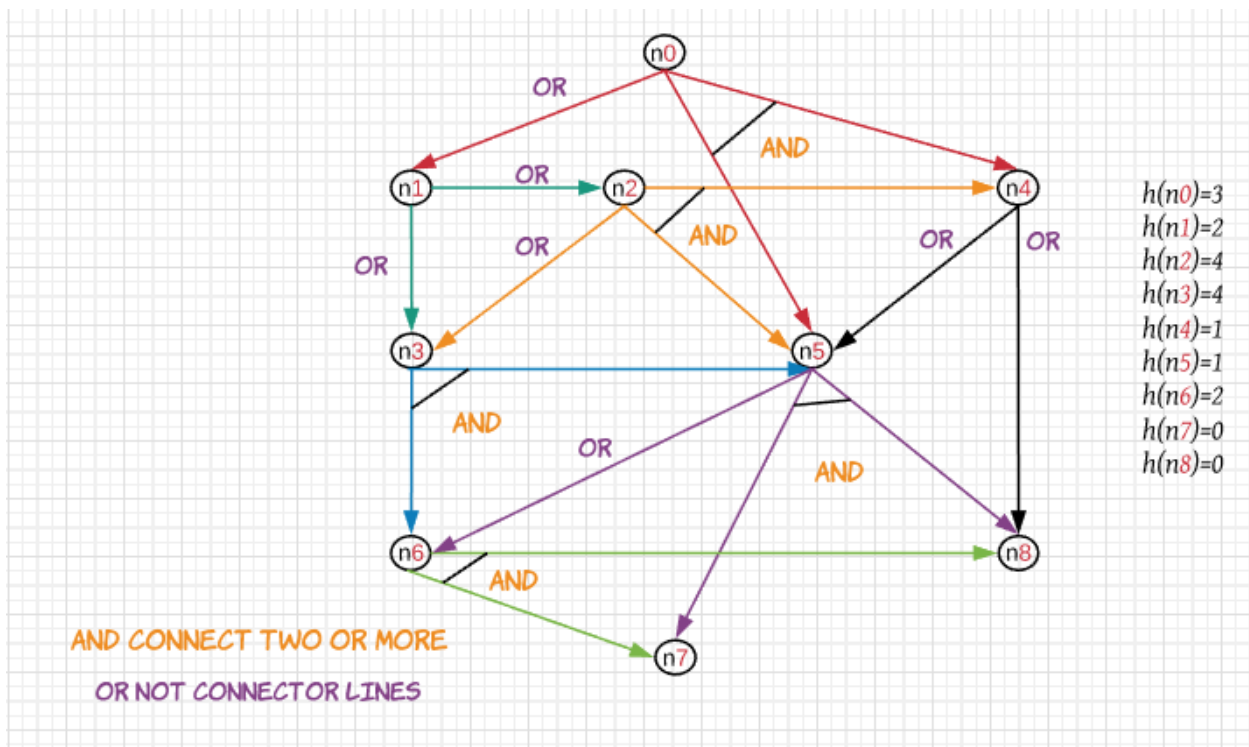
$e(s) = \text{Max} - \text{Min} = 6-5 = 1$



Example



AO*



- Algorithm has two loops
- Loops keep building partial solution while building, it is evaluated what it the best subgraph so far.
- Check to see if by any chances, it might be better solution that has better partial cost so far, if that the cases switch to pursue that particular solution.

- Circle node n7 and node n8 with green
- the green color indicates something solve

Example

