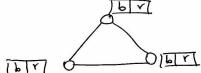
In the name of God, the Merciful, the Compassionate AI Midterm Solution

in the constraint graph. For instance, consider

graph Coloring problem and the following constraint

graph:

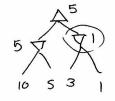


where b stands for blue and r stands for red. The graph is completely consistent but it is obviously a dead-end.

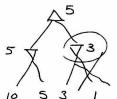
b) False. It depends on the step length  $\alpha$ . If it is small enough, then we could guarantee a decrease in the function value in the next iteration. For example consider the function  $f(x,y) = x^2 + y^2$  and the initial point (1,1). The gradient is  $\nabla f = (2x, 2y)$ . So in each iteration  $\int_{\alpha}^{\alpha} x^2 + \alpha = -3$ . Let  $\alpha = 2$ . We get  $\int_{\alpha}^{\alpha} x^2 - 3$ . The function value would increase to  $3^2 + 3^2 = 18$ .

c) False. The values of internal nodes (i.e. the ones other than the root) could be inaccurate if their successors are pruned. For example:

without pruning



with praning



- d) True. If the function value increases in a candidate next move, the probability of acceptance would be  $p=e^{-\left(\frac{1}{2}\text{new}-\frac{1}{2}\text{old}\right)}$ . Note that if  $\frac{1}{2}\text{new} > \frac{1}{2}\text{old} > \frac{1}{2}\text{min} = 0$  and the algorithm would not accept the candidate next move.
- 2) a) Let each state of the search tree be a set of selected nodes in the given graph. In each action, we add a node that is connected to all already selected nodes in the given graph. Search would terminate if we reach a state with K nodes in it. We could use DFS, for example, to search in this tree along with some heuristics to speed up the search.
  - b) Let each state be a subset of nodes in the given graph, with size K. Define the hearistic function as  $h(s) = {K \choose 2} |E_S|$ . where  $E_S$  is the set of edges between nodes in S in the given graph. We seek a state  $S^*$  with  $h(S^*) = 0$ . We could run a random search such as simulated annealing to optimize h(S) by removing a node and replacing it with another node at each iteration.

c) Let Xi be a binary variable indicating wether or not the i-th node belongs to the clique.

Obviously  $\sum_{i=1}^{n} X_i = K$  should be satisfied. Also,

 $\forall i,j: X_i=1 \land X_j=1 \Rightarrow W_{i,j}=1$ where  $W_{i,j}$  is the i,j entries of the graph adjacency matrix.

These are the heuristics that could guide the search to assign values to the variables;

- Minimum Remaining Value:

If a node doesn't have at least k neighbors in the graph, we could immediately assign of to it. Also, if a node is not connected to all already nodes in the cliques, we can safely ignore it, and assign of to it.

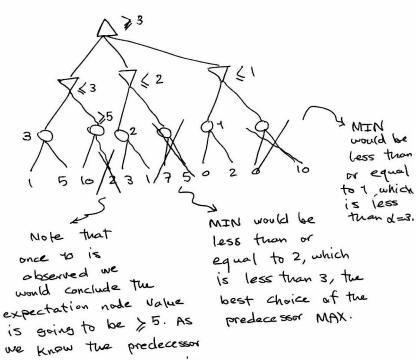
- Degree Constraint :

If a node has a higher degree in the given graph, assign its value first.

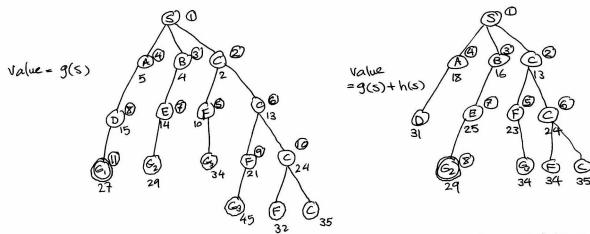
- Least Constraining Value:

If assigning 1 to a variable causes a lot of other variables to be 0, (that is the neighbors of that variable are not connected to all of already selected nodes), we should first try 0.





4) a) Uniform Cost Search Search



MIN node (3 there is no point in evaluating

the next child.

S,C,B,A,F,C,E,D,F,C,G, S,C,B,A,F,C,E,G2

not admissible, for node D, h(D)=16 b) Because h is while there is only cost 12 to reach the goal so h overestimates at the remaining cost-tostate. goal in state D and hence is not necessarily optimal.