Problem Solving and Search in Artificial Intelligence

Tabu Search

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Introduction

- Local search techniques
 - Tabu search
 - Simulated annealing
 - Stochastic Hill-Climber
 - ...
- Tabu Search uses memory during the search
- In memory are stored relevant information about the history of search
- The memory should help to avoid the cycles during the search
- Tabu search is a deterministic heuristic technique

Procedure Tabu-Suche begin Initialize tabu list



Procedure Tabu-Suche begin Initialize tabu list Generate randomly Initial Solution s_c



Procedure Tabu-Suche begin Initialize tabu list Generate randomly Initial Solution s_c Evaluate s_c





```
Procedure Tabu-Suche
begin
Initialize tabu list
Generate randomly Initial Solution s_c
Evaluate s_c
repeat
Generate all neighborhood solutions of the solution s_c
```



```
Procedure Tabu-Suche
begin
Initialize tabu list
Generate randomly Initial Solution s_c
Evaluate s_c
repeat
Generate all neighborhood solutions of the solution s_c
Find best solution s_x in the neighborhood
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Procedure Tabu-Suche

begin

Initialize tabu list

Generate randomly Initial Solution s_c

Evaluate s_c

repeat

Generate all neighborhood solutions of the solution s_c

Find best solution s_x in the neighborhood

if s_x is not tabu solution then s_c = s_x
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else if 'aspiration criteria' is fulfilled then
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find best not tabu solution in the neighborhood s_{nt}
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Procedure Tabu-Suche
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 Evaluate S
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   Generate all neighborhood solutions of the solution s_{c}
   Find best solution s_{r} in the neighborhood
   if s_x is not tabu solution then s_c = s_x
   else if 'aspiration criteria' is fulfilled then
         S_{c} = S_{x}
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      S_c = S_{nt}
   Update tabu list
 until (terminate-condition)
end
```



Example



- SAT problem:
 - Make a a compound statement of Boolean variables to evaluate to true
 - Suppose we have to solve the SAT problem with 8 variables:

$$F(x) = (x_1 \lor \overline{x}_3 \lor x_7) \land (\overline{x}_1 \lor \overline{x_2}) \land \dots \land (\overline{x}_2 \lor x_4 \lor x_7)$$

 Find the truth assignment for each variable x_i such that F(x)=TRUE

General questions



- Representation of solution
 - Candidate solution is represented with a binary string of length n (number of variables)
 - Example: X=(0,0,0,1,1,1,0,1) represents this solution: x1=0, x2=0, x3=0, x4=1, x5=1, x6=1, x7=0, x8=1
- Evaluation of function:
 - Weighted sum of a number of satisfied clauses (weights depends on the number of variables in clause)
- Initial Solution
 - Can be generated for example by random assignment of variables with 0 or 1: X=(0,1,1,1,0,0,0,0)

Neighborhood generation

- Moves
 - A simple move is defined, which flips the value of one variable from 1 to 0 or from 0 to 1
 - More moves can be defined...
- If we apply only the first move the whole neighborhood of solution can be generated by flipping of value of each variable
- In tabu search usually the whole neighborhood is generated during each iteration

Tabu seach specific questions

- Memory
 - Which information should we store during the search to possibly avoid the cycles?

Memory

- Recency-based memory
 - Some parameters of few past iterations are stored
 - For example for SAT problem we could store the information for the flipped variables in past 5 iterations
 - Based on that we could forbid (make tabu) the flipping of variables which were flipped in last 5 iterations





Memory



- Frequency-based memory
 - Stores information for larger number of iterations
 - For example for SAT problem we could store the information about number of flipps for each variable during the last 100 iterations
 - Based on that we could prefer some of flipps of variables more than others during the search



Selection of solutions



- The acceptance of solution for next iteration depends not only from its quality
- The memory has also the impact in the selection process
- Solution are classified in tabu and not tabu solutions
- Usually the best non tabu solutions is accepted for the next iteration

Selection of solutions

- Aspiration criteria
 - Tabu solution may be accepted if it fulfills some conditions
 - Example: The tabu solution is the best solution so far
- Based on frequency based memory
 - Search can be intensified
 - More frequent moves are preferred
 - Search can be diversified
 - Less used moves during the search are preferred



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 Evaluate s<sub>a</sub>
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   Find best solution s_r in the neighborhood
   if s_x is not tabu solution then s_c = s_x
     else if 'aspiration criteria' is fulfilled then
         S_{c} = S_{x}
     else
      find best not tabu solution in the neighborhood s_{nt}
      S_c = S_{nt}
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```

SAT Problem

- Initialize memory
 - Recency based memory
 - M: 0 0 0 0 0 0 0 0
 - Frequency based memory





```
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```

- Initial Solution
 - Random generated solution

$$s_c$$
: 1 0 1 0 0 0 1 1





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- Evaluate solution
 - Suppose that the fitness of solution is 30



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```

Neighborhood of current solution

S_c

 S_1





Neighborhood of current solution





S_c



Neighborhood of current solution







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Evaluation of solutions



Best solution







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      find best not tabu solution in the neighborhood s_{nt}
      S_c = S_{nt}
   Update tabu list
 until (terminate-condition)
end
```

• Recency based memory



Recency based memory





()

• M:

Recency based memory

5



0

0

0

• F: 0 1 0 0 0 0 0 0

0

0

0

Flipping of bit in position 2 is Tabu in next 5 iterations



- Recency based memory
 - M: 0 5 0 0 0 0 0 0
 - Frequency based memory

()

0

0

0

0

0

0

• F:

Flipping of bit in position 2 is Tabu in next 5 iterations







 Suppose that in next iteration the best solution is obtained by flipping the bit in position 4. The content of memory after the second iteration will be:

All non zeros entries are decreased by one at every iteration

SAT problem

• Suppose that after 8 iterations the short term memory has following content:

- Suppose that the following solutions are obtained from the neighborhood of the current solution:
 - $eval(s_1)=35$, $eval(s_2)=38$, $eval(s_3)=36$, $eval(s_4)=34$, $eval(s_5)=32$, $eval(s_6)=30$, $eval(s_7)=34$, $eval(s_8)=33$



SAT problem

• Suppose that after 8 iterations the short term memory has following content:

M: 0 4 5 3 1 0 2 0

- The following solutions are obtained from the neighborhood of the current solution:
 - eval(s₁)=35, eval(s₂)=38, eval(s₃)=36, eval(s₄)=34, eval(s₅)=32, eval(s₆)=30, eval(s₇)=34, eval(s₈)=33
- Best solution in neighborhood has the fitness 38, but it is obtained by flipping bit 2



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      S_c = S_{nt}
   Update tabu list
 until (terminate-condition)
end
```

Aspiration criteria



- The tabu solution may be accepted, if it fulfills some conditions
 - For example if the solution is the best solution found so far
- Suppose that in SAT Example the best solution found is far has fitness 39



```
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 Generate randomly Initial Solution s_{a}
 Evaluate s<sub>a</sub>
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   if s_x is not tabu solution then s_c = s_x
      else if 'aspiration criteria' is fulfilled then
         S_{c} = S_{x}
      else
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      \boldsymbol{S}_{c} = \boldsymbol{S}_{nt}
   Update tabu list
 until (terminate-condition)
end
```



 eval(s₁)=35, eval(s₂)=38, eval(s₃)=36, eval(s₄)=34, eval(s₅)=32, eval(s₆)=30, eval(s₇)=34, eval(s₈)=33

- If the aspiration criteria is not fulfiled, only non tabu solutions will be taken in consideration
 - Solutions: S₁, S₆, S₈



- eval(s₁)=35, eval(s₂)=38, eval(s₃)=36, eval(s₄)=34, eval(s₅)=32, eval(s₆)=30, eval(s₇)=34, eval(s₈)=33
- If the aspiration criteria is not fulfiled, only non tabu solutions will be taken in consideration
 - Solutions: S₁, S₆, S₈

Solution s_1 is accepted for the next iteration



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      S_c = S_{nt}
   Update tabu list
 until (terminate-condition)
end
```

Termination condition

- Optimal solution found
- Number of iterations
- Time
- Empty Neighborhood
- No improves of solution for a determined time/number of iterations
- User interaction
- ...



 Suppose that the content of frequencybased memory for SAT problem after 100 iterations has the following content

- The following solutions are obtained from the neighborhood of the current solution:
 - eval(s₁)=46, eval(s₂)=43, eval(s₃)=46, eval(s₄)=45, eval(s₅)=44, eval(s₆)=43, eval(s₇)=46, eval(s₈)=46



 Suppose that the content of frequency-based memory for SAT problem after 100 iterations has the following content

- The following solutions are obtained from the neighborhood of the current solution:
 - eval(s₁)=46, eval(s₂)=43, eval(s₃)=46, eval(s₄)=45, eval(s₅)=44, eval(s₆)=43, eval(s₇)=46, eval(s₈)=46
- Suppose that only solutions s₃, s₇, s₈ are non-tabu solutions



 $eval(s_1)=46$, $eval(s_2)=43$, $eval(s_3)=46$, $eval(s_4)=45$, $eval(s_5)=44$, $eval(s_6)=43$, $eval(s_7)=46$, $eval(s_8)=46$

- Solutions s₃, s₇, s₈ are non-tabu solutions
- Possible use of memory
 - Make less frequently used moves more attractive
 - Diversification of search



 $eval(s_1)=46$, $eval(s_2)=43$, $eval(s_3)=46$, $eval(s_4)=45$, $eval(s_5)=44$, $eval(s_6)=43$, $eval(s_7)=46$, $eval(s_8)=46$

- Solutions s₃, s₇, s₈ are non-tabu solutions
- Possible use of memory
 - Make less frequently used moves more attractive
 - Diversification of search

Solution s₈ will be accepted for next iteration



- Other possibilities of use of frequency-based memory
 - Aspiration by default
 - Select a move that is the "oldest" of all considered
 - Aspiration by search direction
 - Memorize also whether or not the moves generated improvements
 - Aspiration by influence
 - Particular move can have larger influence if a "larger" step is made from the old solution to the new

Tabu list



- Length of tabu list (for how many iteration should the solution be made tabu)
 - Usually depends from size of problem
 - The length of tabu list could also change during the search
 - Reactive tabu search

Tabu List

- Hashing
- FIFO list
- Storage of last usage time of moves

Adaptive length of tabu list

- Length is 1 in the beginning
- Length increases when the repetitions of solutions happens
- Length decreases when the repetition of solutions disappears

Literature



- Z. Michalewicz and D. B. Fogel. *How to Solve It: Modern Heuristics*. -- Chapter 5
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