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DWARKADAS J. SANGHVI COLLEGE OF ENGINEERING

(Autonomous College Affiliated to the University of Mumbai)

NAAC Accredited with "A" Grade (CGPA : 3.18)



Department of Computer Science and Engineering (Data Science)

Subject: Artificial Intelligence (DJ19DSC502)

AY: 2023-24

Experiment 4

(Solution Space)

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Aim: Find the solution of a SAT (Satisfiability) problem using Variable Neighborhood Descent.

Theory:

The SAT problem

Given a Boolean formula made up of a set of propositional variables $V = \{a, b, c, d, e, \dots\}$ each of which can be *true* or *false*, or 1 or 0, to find an assignment for the variables such that the given formula evaluates to *true* or 1.

For example, $F = ((a \vee \sim e) \wedge (e \vee \sim c)) \supset (\sim c \vee \sim d)$ can be made *true* by the assignment $\{a=\text{true}, c=\text{true}, d=\text{false}, e=\text{false}\}$ amongst others.

Very often SAT problems are studied in the *Conjunctive Normal Form (CNF)*. For example, the following formula has five variables (a,b,c,d,e) and six clauses.

$$(b \vee \sim c) \wedge (c \vee \sim d) \wedge (\sim b) \wedge (\sim a \vee \sim e) \wedge (e \vee \sim c) \wedge (\sim c \vee \sim d)$$



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Solution Space Search and Perturbative methods

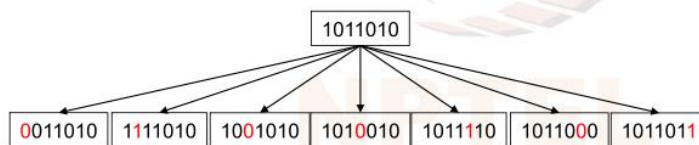
The Solution Space is the space of candidate solutions.

A local search method generates the neighbours of a candidate by applying some perturbation to the given candidate

MoveGen function = neighbourhood function

A SAT problem with N variables has 2^N candidates
 - where each candidate is a N bit string

When $N = 7$, a *neighbourhood function* may change **one** bit.



Variable Neighbourhood Descent

VariableNeighbourhoodDescent()

```

1      node ← start
2      for i ← 1 to n
3          do moveGen ← MoveGen(i)
4              node ← HillClimbing(node, moveGen)
5      return node
  
```

The algorithm assumes that the function *moveGen* can be passed as a parameter. It assumes that there are N *moveGen* functions sorted according to the density of the neighbourhoods produced.

Lab Assignment to do:

Solve the following SAT problems using VND

$$1. \quad F = (A \vee \sim B) \wedge (B \vee \sim C) \wedge (\sim B) \wedge (\sim C \vee E) \wedge (A \vee C) \wedge (\sim C \vee \sim D)$$



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```
import random

def sat_formula(A, B, C, D, E):
    clause1 = (A or not B)
    clause2 = (B or not C)
    clause3 = (not B)
    clause4 = (not C or E)
    clause5 = (A or C)
    clause6 = (not C or not D)

    return clause1 and clause2 and clause3 and clause4 and clause5 and clause6

def objective_function(A, B, C, D, E):
    return sum([not sat_formula(A, B, C, D, E)])

def random_initial_state():
    return {var: random.choice([True, False]) for var in ['A', 'B', 'C', 'D', 'E']}

def local_search(state):
    while True:
        neighbors = neighborhood_structure(state)
        best_neighbor = min(neighbors, key=lambda s: objective_function(**s))
        if objective_function(**best_neighbor) >= objective_function(**state):
            break
        state = best_neighbor
    return state

def neighborhood_structure(state):
    neighbors = []
    for var, value in state.items():
        neighbor = state.copy()
        neighbor[var] = not value
        neighbors.append(neighbor)
    return neighbors
```



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```
num_iterations = 20

best_state = None
best_cost = float('inf')
for _ in range(num_iterations):
    current_state = random_initial_state()
    current_state = local_search(current_state)
    current_cost = objective_function(**current_state)
    if current_cost < best_cost:
        best_state = current_state
        best_cost = current_cost

print("Final Assignment after", num_iterations, "Iterations:")
print(best_state)
print("Number of Unsatisfied Clauses:", objective_function(**best_state))

Final Assignment after 20 Iterations:
{'A': True, 'B': False, 'C': False, 'D': False, 'E': False}
Number of Unsatisfied Clauses: 0
```

2. $F = (A \vee B) \wedge (A \wedge \sim C) \wedge (B \wedge D) \wedge (A \vee \sim E)$



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```
import random

def sat_formula(A, B, C, D, E):
    clause1 = (A or B)
    clause2 = (A and not C)
    clause3 = (B and D)
    clause4 = (A or not E)

    return clause1 and clause2 and clause3 and clause4

def objective_function(A, B, C, D, E):
    return sum([not sat_formula(A, B, C, D, E)])

def random_initial_state():
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```
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    current_cost = objective_function(**current_state)
    if current_cost < best_cost:
        best_state = current_state
        best_cost = current_cost

print("Final Assignment after", num_iterations, "Iterations:")
print(best_state)
print("Number of Unsatisfied Clauses:", objective_function(**best_state))
```

```
Final Assignment after 20 Iterations:
{'A': True, 'B': True, 'C': False, 'D': True, 'E': True}
Number of Unsatisfied Clauses: 0
```