

"""

SAT ALGORITHM DISCOVERY SYSTEM - ENHANCED

Translated from Grimoire Codex to Python

NOW WITH:

- Real DPLL SAT solving algorithm
- Symbolic complexity analysis
- Formal solution verification
- Advanced meta-learning

"""

```
import random
import time
from dataclasses import dataclass, field
from typing import List, Dict, Tuple, Optional, Callable, Any, Set
from collections import defaultdict
from copy import deepcopy
import math

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# SAT PROBLEM STRUCTURES
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@dataclass
class SATInstance:
    """Represents a 3-SAT problem instance"""
    num_variables: int
    clauses: List[Tuple[int, ...]]

    @staticmethod
    def generate_random(num_vars: int, num_clauses: int, clause_size: int = 3) -> 'SATInstance':
        """Generate random SAT instance"""
        clauses = []
        for _ in range(num_clauses):
            vars_in_clause = random.sample(range(1, num_vars + 1), min(clause_size, num_vars))
            clause = tuple(random.choice([v, -v]) for v in vars_in_clause)
            clauses.append(clause)
        return SATInstance(num_vars, clauses)

    def verify_solution(self, assignment: Dict[int, bool]) -> bool:
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"""Verify if assignment satisfies all clauses"""
for clause in self.clauses:
    satisfied = False
    for literal in clause:
        var = abs(literal)
        if var in assignment:
            if (literal > 0 and assignment[var]) or (literal < 0 and not assignment[var]):
                satisfied = True
                break
    if not satisfied:
        return False
return True

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# REAL DPLL SAT SOLVER
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class DPLLSolver:
    """Real DPLL SAT solver with configurable heuristics"""

    def __init__(self, heuristics: List[str], strategies: List[str], parameters: Dict):
        self.heuristics = heuristics
        self.strategies = strategies
        self.parameters = parameters
        self.operation_count = 0
        self.decision_count = 0
        self.propagation_count = 0

    def solve(self, instance: SATInstance, max_operations: int = 10000) -> Tuple[bool,
Optional[Dict[int, bool]]], int]:
        """Solve SAT instance using DPLL algorithm"""
        self.operation_count = 0
        self.decision_count = 0
        self.propagation_count = 0

        clauses = [set(clause) for clause in instance.clauses]
        assignment = {}

        result = self._dpll(clauses, assignment, instance.num_variables, max_operations)

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if result:
    return True, assignment, self.operation_count
else:
    return False, None, self.operation_count

def _dpll(self, clauses: List[Set[int]], assignment: Dict[int, bool],
          num_vars: int, max_ops: int) -> bool:
    """Core DPLL recursive algorithm"""
    self.operation_count += 1

    if self.operation_count > max_ops:
        return False

    if not clauses:
        return True

    if any(len(clause) == 0 for clause in clauses):
        return False

    # Unit propagation
    if "unit_propagation" in self.strategies:
        while True:
            unit_clause = self._find_unit_clause(clauses)
            if unit_clause is None:
                break

            literal = next(iter(unit_clause))
            var = abs(literal)
            value = literal > 0

            assignment[var] = value
            clauses = self._propagate(clauses, literal)
            self.propagation_count += 1
            self.operation_count += 1

            if not clauses:
                return True
            if any(len(clause) == 0 for clause in clauses):
                return False

    # Pure literal elimination
    if "pure_literal_elimination" in self.strategies:
        pure_literal = self._find_pure_literal(clauses)
        if pure_literal is not None:

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        var = abs(pure_literal)
        value = pure_literal > 0
        assignment[var] = value
        clauses = self._propagate(clauses, pure_literal)
        self.operation_count += 1
        return self._dpll(clauses, assignment, num_vars, max_ops)

# Choose variable
var = self._choose_variable(clauses, assignment, num_vars)
if var is None:
    return True

self.decision_count += 1

# Try positive
assignment_copy = assignment.copy()
clauses_copy = deepcopy(clauses)
assignment_copy[var] = True
clauses_try = self._propagate(clauses_copy, var)

if self._dpll(clauses_try, assignment_copy, num_vars, max_ops):
    assignment.update(assignment_copy)
    return True

# Try negative
assignment[var] = False
clauses = self._propagate(clauses, -var)

return self._dpll(clauses, assignment, num_vars, max_ops)

def _find_unit_clause(self, clauses: List[Set[int]]) -> Optional[Set[int]]:
    for clause in clauses:
        if len(clause) == 1:
            return clause
    return None

def _find_pure_literal(self, clauses: List[Set[int]]) -> Optional[int]:
    literal_count = defaultdict(int)
    for clause in clauses:
        for literal in clause:
            literal_count[literal] += 1

    for literal in literal_count:
        if -literal not in literal_count:

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        return literal
    return None

def _propagate(self, clauses: List[Set[int]], literal: int) -> List[Set[int]]:
    new_clauses = []
    for clause in clauses:
        if literal in clause:
            continue
        elif -literal in clause:
            new_clause = clause - {-literal}
            new_clauses.append(new_clause)
        else:
            new_clauses.append(clause.copy())
    return new_clauses

def _choose_variable(self, clauses: List[Set[int]],
                     assignment: Dict[int, bool], num_vars: int) -> Optional[int]:
    unassigned = set(range(1, num_vars + 1)) - set(assignment.keys())

    if not unassigned:
        return None

    if not self.heuristics:
        return random.choice(list(unassigned))

    heuristic = self.heuristics[0]

    if heuristic == "random_variable_selection":
        return random.choice(list(unassigned))

    elif heuristic == "most_constrained_first":
        var_count = defaultdict(int)
        for clause in clauses:
            for literal in clause:
                var = abs(literal)
                if var in unassigned:
                    var_count[var] += 1
        if var_count:
            return max(var_count.items(), key=lambda x: x[1])[0]
        return random.choice(list(unassigned))

    elif heuristic == "activity_based":
        var_activity = defaultdict(float)
        for clause in clauses:

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clause_vars = [abs(lit) for lit in clause if abs(lit) in unassigned]
activity_boost = 1.0 / max(len(clause), 1)
for var in clause_vars:
    var_activity[var] += activity_boost
if var_activity:
    return max(var_activity.items(), key=lambda x: x[1])[0]
return random.choice(list(unassigned))

else:
    return random.choice(list(unassigned))

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# COMPLEXITY ANALYZER
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class ComplexityAnalyzer:
    """Mathara + Fractala - Analyze algorithmic complexity"""

    def __init__(self):
        self.measurements = []

    def analyze_run(self, num_variables: int, operations: int, solved: bool) -> Dict:
        if num_variables == 0:
            return {'empirical_complexity': 'N/A'}

        n = num_variables

        linear_ratio = operations / n if n > 0 else 0
        nlogn_ratio = operations / (n * math.log2(n)) if n > 1 else 0
        quadratic_ratio = operations / (n * n) if n > 0 else 0
        exponential_base = operations ** (1/n) if n > 0 and operations > 0 else 0

        analysis = {
            'operations': operations,
            'variables': num_variables,
            'linear_ratio': linear_ratio,
            'exponential_base': exponential_base,
            'solved': solved
        }

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        self.measurements.append(analysis)
        return analysis

def estimate_complexity_class(self, recent_runs: int = 10) -> str:
    if len(self.measurements) < 3:
        return "INSUFFICIENT_DATA"

    recent = self.measurements[-recent_runs:]

    exp_bases = [m['exponential_base'] for m in recent if m['exponential_base'] > 0]
    if exp_bases:
        avg_base = sum(exp_bases) / len(exp_bases)
        if avg_base > 1.1:
            return f"O(~{avg_base:.2f}^n) EXPONENTIAL"

    linear_ratios = [m['linear_ratio'] for m in recent]
    if linear_ratios:
        avg_linear = sum(linear_ratios) / len(linear_ratios)
        return f"O(n) ratio: {avg_linear:.1f}"

    return "UNKNOWN"

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# MEMORY & STATE
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class MemorySubstrate:
    def __init__(self):
        self.state_history = []
        self.journey_log = []
        self.complexity_history = []

    def track_journey(self, event: str):
        self.journey_log.append({'event': event, 'time': time.time()})

    def record_complexity(self, analysis: Dict):
        self.complexity_history.append(analysis)

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# ALGORITHM TEMPLATE
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@dataclass
class AlgorithmTemplate:
    name: str
    heuristics: List[str]
    strategies: List[str]
    parameters: Dict[str, Any]
    fitness: float = 0.0
    success_count: int = 0
    total_runs: int = 0
    total_operations: int = 0

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# GENERATION ENGINE
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class GenerationEngine:
    def __init__(self, memory: MemorySubstrate):
        self.memory = memory
        self.templates_created = 0

        self.heuristic_pool = [
            "random_variable_selection",
            "most_constrained_first",
            "activity_based"
        ]

        self.strategy_pool = [
            "unit_propagation",
            "pure_literal_elimination"
        ]

    def generate_algorithm(self) -> AlgorithmTemplate:

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algorithm = AlgorithmTemplate(
    name=f"Algo_{self.templates_created}",
    heuristics=random.sample(self.heuristic_pool, random.randint(1, 2)),
    strategies=random.sample(self.strategy_pool, random.randint(1, 2)),
    parameters={'max_operations': random.randint(2000, 5000)}
)
self.templates_created += 1
return algorithm

def mutate_algorithm(self, base: AlgorithmTemplate) -> AlgorithmTemplate:
    mutated = AlgorithmTemplate(
        name=f"{base.name}_m{random.randint(0,99)}",
        heuristics=base.heuristics.copy(),
        strategies=base.strategies.copy(),
        parameters=base.parameters.copy()
    )
    mutation_type = random.choice(['heuristic', 'strategy', 'parameter'])

    if mutation_type == 'heuristic' and mutated.heuristics:
        mutated.heuristics[0] = random.choice(self.heuristic_pool)
    elif mutation_type == 'strategy' and mutated.strategies:
        mutated.strategies[0] = random.choice(self.strategy_pool)
    else:
        mutated.parameters['max_operations'] += random.randint(-500, 500)
        mutated.parameters['max_operations'] = max(1000,
            mutated.parameters['max_operations'])
    return mutated

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# EXECUTION ENGINE
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class ExecutionEngine:
    def __init__(self, memory: MemorySubstrate):
        self.memory = memory
        self.complexity_analyzer = ComplexityAnalyzer()

    def execute_algorithm(self, algorithm: AlgorithmTemplate, instance: SATInstance) -> Dict:

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solver = DPLLSolver(algorithm.heuristics, algorithm.strategies, algorithm.parameters)

start_time = time.time()
solved, assignment, operations = solver.solve(instance,
algorithm.parameters['max_operations'])
execution_time = time.time() - start_time

verified = False
if solved and assignment:
    verified = instance.verify_solution(assignment)

complexity_analysis = self.complexity_analyzer.analyze_run(instance.num_variables,
operations, solved)
self.memory.record_complexity(complexity_analysis)

symbol = '✓' if verified else '?' if solved else '✗'
self.memory.track_journey(f'{algorithm.name}: {symbol} {operations} ops')

return {
    'algorithm': algorithm.name,
    'solved': solved,
    'verified': verified,
    'operations': operations,
    'time': execution_time,
    'instance_size': instance.num_variables
}

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# META-LEARNING
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class MetaLearningEngine:
    def __init__(self, memory: MemorySubstrate):
        self.memory = memory
        self.generation = 0
        self.fitness_history = []

    def evaluate_population(self, algorithms: List[AlgorithmTemplate], results: List[Dict]) ->
List[AlgorithmTemplate]:
        algo_results = defaultdict(list)

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for result in results:
    algo_results[result['algorithm']].append(result)

for algo in algorithms:
    algo_results_list = algo_results[algo.name]
    if algo_results_list:
        verified_count = sum(1 for r in algo_results_list if r['verified'])
        total_ops = sum(r['operations'] for r in algo_results_list)
        avg_ops = total_ops / len(algo_results_list)

        verification_rate = verified_count / len(algo_results_list)
        ops_penalty = avg_ops / 10000

        algo.fitness = verification_rate - ops_penalty
        algo.success_count = verified_count
        algo.total_runs = len(algo_results_list)
        algo.total_operations = total_ops
    else:
        algo.fitness = 0.0

return sorted(algorithms, key=lambda a: a.fitness, reverse=True)

def evolve_generation(self, algorithms: List[AlgorithmTemplate], generator: GenerationEngine) -> List[AlgorithmTemplate]:
    self.generation += 1

    next_gen = algorithms[:3].copy()

    while len(next_gen) < len(algorithms):
        parent = random.choice(algorithms[:3])
        child = generator.mutate_algorithm(parent)
        next_gen.append(child)

    avg_fitness = sum(a.fitness for a in algorithms) / len(algorithms)
    self.fitness_history.append(avg_fitness)

    return next_gen

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# ORCHESTRATOR

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class SystemOrchestrator:
    def __init__(self):
        self.memory = MemorySubstrate()
        self.generator = GenerationEngine(self.memory)
        self.executor = ExecutionEngine(self.memory)
        self.meta_learner = MetaLearningEngine(self.memory)

        self.population_size = 8
        self.instances_per_gen = 5
        self.current_population = []

    def initialize(self):
        print("🌟 ENHANCED SAT ALGORITHM DISCOVERY SYSTEM")
        print(" Real DPLL + Formal Verification + Complexity Analysis")
        print("==" * 70)

        self.current_population = [self.generator.generate_algorithm() for _ in
range(self.population_size)]
        print(f"✓ Generated {self.population_size} algorithms\n")

    def run_generation(self, gen_num: int):
        print(f"{'='*70}")
        print(f"GENERATION {gen_num}")
        print(f"{'='*70}")

        test_instances = [SATInstance.generate_random(random.randint(10, 15),
random.randint(25, 40))
                         for _ in range(self.instances_per_gen)]

        all_results = []
        for algo in self.current_population:
            for instance in test_instances:
                result = self.executor.execute_algorithm(algo, instance)
                all_results.append(result)

        total = len(all_results)
        verified = sum(1 for r in all_results if r['verified'])
        avg_ops = sum(r['operations'] for r in all_results) / total

        print(f"\n📊 RESULTS:")

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        print(f" Verified: {verified}/{total} ({verified/total:.1%})")
        print(f" Avg Operations: {avg_ops:.0f}")

        self.current_population = self.meta_learner.evaluate_population(self.current_population,
all_results)

        best = self.current_population[0]
        print(f"\n🏆 BEST: {best.name}")
        print(f" Fitness: {best.fitness:.4f}")
        print(f" Verified: {best.success_count}/{best.total_runs}")
        print(f" Heuristics: {best.heuristics}")
        print(f" Strategies: {best.strategies}")

complexity = self.executor.complexity_analyzer.estimate_complexity_class()
print(f" Complexity: {complexity}")

        self.current_population = self.meta_learner.evolve_generation(self.current_population,
self.generator)

def run_experiment(self, num_generations: int = 8):
    self.initialize()

    for gen in range(1, num_generations + 1):
        self.run_generation(gen)

    print(f"\n{'='*70}")
    print("EXPERIMENT COMPLETE")
    print(f"{'='*70}")

    print(f"\n📈 Fitness Evolution:")
    for i, fitness in enumerate(self.meta_learner.fitness_history):
        print(f" Gen {i+1}: {fitness:.4f}")

    best = self.current_population[0]
    print(f"\n🏆 FINAL BEST ALGORITHM:")
    print(f" {best.name}")
    print(f" Fitness: {best.fitness:.4f}")
    print(f" Success: {best.success_count}/{best.total_runs} verified")
    print(f" Heuristics: {best.heuristics}")
    print(f" Strategies: {best.strategies}")

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# MAIN
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if __name__ == "__main__":
    system = SystemOrchestrator()
    system.run_experiment(num_generations=8)

    print("\n" + "="*70)
    print("✨ ENHANCED FEATURES:")
    print(" ✓ Real DPLL SAT solver")
    print(" ✓ Formal solution verification")
    print(" ✓ Symbolic operation counting")
    print(" ✓ Complexity estimation")
    print(" ✓ Evolutionary meta-learning")
    print("="*70)
```

🌟 ENHANCED SAT ALGORITHM DISCOVERY SYSTEM

Real DPLL + Formal Verification + Complexity Analysis

✓ Generated 8 algorithms

GENERATION 1

📊 RESULTS:

Verified: 40/40 (100.0%)

Avg Operations: 23

🏆 BEST: Algo_1

Fitness: 0.9984

Verified: 5/5

Heuristics: ['activity_based', 'most_constrained_first']

Strategies: ['unit_propagation']

Complexity: O(~1.28^n) EXPONENTIAL

GENERATION 2

📊 RESULTS:

Verified: 40/40 (100.0%)

Avg Operations: 24

🏆 BEST: Algo_1

Fitness: 0.9984

Verified: 5/5

Heuristics: ['activity_based', 'most_constrained_first']

Strategies: ['unit_propagation']

Complexity: $O(\sim 1.28^n)$ EXPONENTIAL

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GENERATION 3

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📊 RESULTS:

Verified: 40/40 (100.0%)

Avg Operations: 24

🏆 BEST: Algo_1_m25_m85

Fitness: 0.9980

Verified: 5/5

Heuristics: ['random_variable_selection', 'most_constrained_first']

Strategies: ['unit_propagation']

Complexity: $O(\sim 1.29^n)$ EXPONENTIAL

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GENERATION 4

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📊 RESULTS:

Verified: 32/40 (80.0%)

Avg Operations: 40

🏆 BEST: Algo_1

Fitness: 0.7984

Verified: 4/5

Heuristics: ['activity_based', 'most_constrained_first']

Strategies: ['unit_propagation']

Complexity: $O(\sim 1.28^n)$ EXPONENTIAL

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GENERATION 5

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📊 RESULTS:

Verified: 40/40 (100.0%)

Avg Operations: 17

🏆 BEST: Algo_1

Fitness: 0.9984
Verified: 5/5
Heuristics: ['activity_based', 'most_constrained_first']
Strategies: ['unit_propagation']
Complexity: $O(\sim 1.30^n)$ EXPONENTIAL

GENERATION 6

 RESULTS:
Verified: 40/40 (100.0%)
Avg Operations: 16

 BEST: Algo_1
Fitness: 0.9986
Verified: 5/5
Heuristics: ['activity_based', 'most_constrained_first']
Strategies: ['unit_propagation']
Complexity: $O(\sim 1.24^n)$ EXPONENTIAL

GENERATION 7

 RESULTS:
Verified: 40/40 (100.0%)
Avg Operations: 17

 BEST: Algo_1
Fitness: 0.9985
Verified: 5/5
Heuristics: ['activity_based', 'most_constrained_first']
Strategies: ['unit_propagation']
Complexity: $O(\sim 1.23^n)$ EXPONENTIAL

GENERATION 8

 RESULTS:
Verified: 40/40 (100.0%)
Avg Operations: 13

 BEST: Algo_1
Fitness: 0.9987
Verified: 5/5

Heuristics: ['activity_based', 'most_constrained_first']

Strategies: ['unit_propagation']

Complexity: O($\sim 1.23^n$) EXPONENTIAL

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EXPERIMENT COMPLETE

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 Fitness Evolution:

Gen 1: 0.9977

Gen 2: 0.9976

Gen 3: 0.9976

Gen 4: 0.7960

Gen 5: 0.9983

Gen 6: 0.9984

Gen 7: 0.9983

Gen 8: 0.9987

 FINAL BEST ALGORITHM:

Algo_1

Fitness: 0.9987

Success: 5/5 verified

Heuristics: ['activity_based', 'most_constrained_first']

Strategies: ['unit_propagation']

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 ENHANCED FEATURES:

- ✓ Real DPLL SAT solver
- ✓ Formal solution verification
- ✓ Symbolic operation counting
- ✓ Complexity estimation
- ✓ Evolutionary meta-learning

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