Artificial Intelligence

Lecturer 4 – Constraint Satisfaction Problems & Logical Agents

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Constraints Satisfaction Problems (CSPs)

- CSPs example
- Backtracking search
- Problem structure
- Local search for CSPs

Constraint Satisfaction Problems

General Idea

- Factored representation for each state: a set of variables, each of which has a value.
- Problem is solved when each variable has a value that satisfies all the constraints on the variable.
- A problem described this way: a constraint satisfaction problem, or CSP.
- CSP search algorithms use *general-purpose* rather than *problem-specific* heuristics
- Key idea: eliminate large portions of the search space all at once by identifying variable/value combinations that violate the constraints.

Definition: Constraint Satisfaction Problem

- A constraint satisfaction problem consists of three components, X, D, and C
 - X is a set of variables, $\{X_1, \dots, X_n\}$
 - D is a set of domains, {D₁, ..., D_n}, one for each variable.
 - C is a set of constraints that specify allowable combinations of values.

Example: Map Coloring

- Variables
 - WA, NT, Q, NSW, V , SA
- Domain
 - D_i = {red, green, blue}
- Constraint
 - Neighbor regions must have different colors
 - Color(WA) \neq color(NT)
 - Color(WA) \neq color(SA)
 - Color(NT) \neq color(SA)
 - ...



Example: Map Coloring

- Solution is an assignment of variables satisfying all constraints
 - WA=red, and
 - NT=green, and
 - Q=red, and
 - NSW=green, and
 - V=red, and
 - SA=blue



Constraint Graph

Binary CSPs

- Each constraint relates at most two variables
- Constraint graph
 - Node is variable
 - Edge is constraint



Varieties of CSPs

Discrete variables

- Finite domain, e.g, SAT Solving
- Infinite domain, e.g., work scheduling
 - Variables is start/end of working day
 - **Precedence constraints:** Constraint language, e.g., StartJob₁ + 5 ≤ StartJob₃
 - Linear constraints are decidable, non-linear constraints are undecidable

Continuous variables

- e.g., start/end time of observing the universe using Hubble telescope
- Linear constraints are solvable using Linear Programming

Varieties of Constraints

Single-variable constraints

□ e.g., SA \neq green

Binary constraints

• e.g., $SA \neq WA$

Multi-variable constraints

Relate at least 3 variables

Soft constraints

- □ Priority, e.g., red better than green
- Cost function over variables

Disjunctive constraints

- Four workers install wheels, but need to share a tool that helps to put the axle in place
- Axle_F and Axle_B must not overlap: either one comes first, or the other takes place

Example: Cryptarithmetic Puzzle

- Variables
 - $\square F,T,O,U,R,W, X_1,X_2,X_3$
- Domain
 - {0,1,2,3,4,5,6,7, 8,9}
- Constraints
 - Alldiff(F,T,O,U,R,W)
 - $O+O = R+10*C_1$
 - $C_1 + W + W = U + 10^* C_2$
 - $C_2 + T + T = O + 10 C_3$
 - □ C₃=F





Real World CSP

Assignment

- E.g., who teach which class
- Scheduling
 - E.g., when and where the class takes place
- Hardware design
- Spreadsheets
- Transport scheduling
- Manufacture scheduling

CSPs by Standard Search

State

Defined by the values assigned so far

Initial state

- The empty assignment
- Successor function
 - Assign a value to a unassigned variable that does not conflict with current assignment
 - Fail if no legal assignment
- Goal test
 - All variables are assigned and no conflict

CSP by Standard Search

- Every solution appears at depth *d* with *n* variables
 - Use depth-first search
- Path is irrelevant
- Number of leaves
 - □ *n!d*ⁿ
 - Two many

Backtracking Search

Variable assignments are commutative, e.g.,

- {WA=red, NT =green}
- {NT =green, WA=red}
- Single-variable assignment
 - Only consider one variable at each node
 - □ *dⁿ* leaves
- Backtracking search
 - Depth-first search+ Single-variable assignment
- Backtracking search is the basic uninformed algorithm for CSPs

Can solve n-Queen with n = 25

Backtracking Search Algorithm

```
function BACKTRACKING-SEARCH(csp) returns solution/failure
return RECURSIVE-BACKTRACKING({ }, csp)
```

function RECURSIVE-BACKTRACKING (assignment, csp) returns soln/failure if assignment is complete then return assignment $var \leftarrow SELECT-UNASSIGNED-VARIABLE(VARIABLES[csp], assignment, csp)$ for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do if value is consistent with assignment given CONSTRAINTS[csp] then add {var = value} to assignment result \leftarrow RECURSIVE-BACKTRACKING(assignment, csp) if result \neq failure then return result remove {var = value} from assignment return failure

Backtracking Search Algorithm



Improving Backtracking Search

- Which variable should be assigned next?
- In what order should its values be tried?
- Can we detect inevitable failure early?
- Can we take advantage of problem structure?

Choosing Variables

- Minimum remaining values (MRV)
 - Choose the variable with the fewest legal values
 - Degree heuristic
 - Choose the variable with the most constraints on remaining variables

Choosing Values

- Least constraining value (LCV)
 - Choose the least constraining value
 - the one that rules out the fewest values in the remaining variables
- Keep track of remaining legal values for unassigned variables
 - Terminate search when any variable has no legal values



- NT and SA cannot both be blue
- Simplest form of propagation makes each arc consistent
 - X -> Y is consistent iff for each value x of X there is some allowed value y for Y

Iterative Algorithms for CSPs

Hill-climbing, Simulated Annealing can be used for CSPs

- Complete state, e.g., all variables are assigned at each node
- Allow states with unsatisfiable constraints
- Operators reassign variables
- Variable selection
 - Random
- Value selection by min-conflicts heuristic
 - Choose value that violates the fewest constraints
 - i.e., hill climbing with h(n) = total number of violated constraints

Example: 4-Queens

- State: 4 queens in four columns (4*4 = 256 states)
- Operators: move queen in column
- Goal test: no attacks
- Evaluation: h(n) = number of attacks



Summary

- CSPs are a special kind of problem:
 - states defined by values of a fixed set of variables
 - goal test defined by constraints on variable values
- Backtracking = depth-first search with one variable assigned per node
- Variable ordering and value selection heuristics help significantly
- Forward checking prevents assignments that guarantee later failure
- Constraint propagation (e.g., arc consistency) does additional work to constrain values and detect inconsistencies
- The CSPs representation allows analysis of problem structure
- Tree-structured CSPs can be solved in linear time
- Iterative min-conflicts is usually effective in practice

Exercice

- Solve the following cryptarithmetic problem by combining the heuristics
 - Constraint Propagation
 - Minimum Remaining Values
 - Least Constraining Values





Exercice

- $O+O = R+10*C_1$
- \Box C₁+W+W= U+10*C₂
- $C_2 + T + T = O + 10 C_3$
- □ C₃= F

- 1. Choose C_3 : domain {0,1}
- 2. Choose $C_3=1$: use constraint propagation $F \neq 0$
- 3. F = 1
- 4. Choose C_2 : C_1 and C_2 have the same remaining values
- 5. Choose $C_2^{-}=0$
- 6. Choose $C_1: C_1$ has Minimum remaining values (MRV)
- 7. Choose $C_1=0$
- 8. Choose O: O must be even, less than 5 and therefore has MRV (T+T=O +10; O+O=R+10*0)
- 9. Choose O = 4
- 10. R=8
- 11. T=7
- 12. Choose U: U must be even, less than 9
- 13. U=6: constraint propagation
- 14. W=3

Reading & Suggested Exercises

Chapter 6: 6.3, 6.7, 6.8

LOGICAL AGENTS

Outline

- What is Logic
- Propositional Logic
 - Syntax
 - Semantic
- Inference in Propositional Logic
 - Forward Chaining
 - Backward Chaining

Knowledge-based Agents

Know about the world

 They maintain a collection of facts (sentences) about the world, their Knowledge Base, expressed in some formal language.

Reason about the world

 They are able to derive new facts from those in the KB using some inference mechanism.

Act upon the world

 They map percepts to actions by querying and updating the KB.

What is Logic ?

A logic is a triplet <L,S,R>

- L, the language of the logic, is a class of sentences described by a precise syntax, usually a formal grammar
- S, the logic's semantic, describes the meaning of elements in L
- R, the logic's inference system, consisting of derivation rules over L

Examples of logics:

 Propositional, First Order, Higher Order, Temporal, Fuzzy, Modal, Linear, …

Propositional Logic

- Propositional Logic is about facts in the world that are either true or false, nothing else
- Propositional variables stand for basic facts
- Sentences are made of
 - propositional variables (A,B,...),
 - Iogical constants (TRUE, FALSE), and
 - logical connectives (not,and,or,..)
- The meaning of sentences ranges over the Boolean values {True, False}
 - Examples: It's sunny, John is married

Language of Propositional Logic

Symbols

- Propositional variables: A,B,...,P,Q,...
- Logical constants: TRUE, FALSE
- Logical connectives:

$$\neg, \wedge, \lor, \Rightarrow, \Leftrightarrow$$

Sentences

- Each propositional variable is a sentence
- Each logical constant is a sentence
- If α and β are sentences then the following are sentences

$$(\alpha), \neg \alpha, \alpha \land \beta, \alpha \lor \beta, \alpha \Rightarrow \beta, \alpha \Leftrightarrow \beta$$

Formal Language of Propositional

- Logic
- Symbols
 - Propositional variables: A,B,...,P,Q,...
 - Logical constants: T, F
 - □ Logical connectives: $\neg, \land, \lor, \Rightarrow, \Leftrightarrow$
- Formal Grammar
 - Sentence -> Asentence | Csentence
 - Asentence -> TRUE | FALSE | A | B |...
 - Csentence -> (Sentence) | ¬ Sentence | Sentence
 Connective Sentence
 - Connective -> \neg | \land | \checkmark | \Longrightarrow | \iff

- The meaning of TRUE is always True, the meaning of FALSE is always False
- The meaning of a propositional variable is either True or False
 - depends on the interpretation
 - assignment of Boolean values to propositional variables
- The meaning of a sentence is either True or False
 - depends on the interpretation

True table

Ρ	Q	Not P	P and Q	P or Q	P implies Q	P equiv Q
False	False	True	False	False	True	True
False	True	True	False	True	True	False
True	False	False	False	True	False	False
True	True	False	True	True	True	True

 $a \Rightarrow b \Leftrightarrow \neg a \lor b \Leftrightarrow \neg b \Rightarrow \neg a$

Satisfiability

- A sentence is satisfiable if it is true under some interpretation
- Ex: P or H is satisfiable

P and ¬P is unsatisfiable (not satisfiable)

- Validity
 - A sentence is valid if it is true in every interpretation

Entailment

- Given
 - A set of sentences $~\Gamma~$
 - A sentence ψ
- Logical Entailment

$$\Gamma \models \Psi$$

if and only if every interpretation that makes all sentences in Γ true also makes ψ true

 $extsf{ }$ We said that $\Gamma extsf{entails} \ \psi$

Satisfiability vs. Validity vs. Entailment

- Ψ is valid iff True $\models \Psi$ (also written $\models \Psi$)
- Ψ is unsatisfiable iff $\Psi \models$ False
- $\Gamma \models \Psi$ iff $\Gamma \cup \{\neg \Psi\}$ is unsatisfiable

Inference in Propositional Logic

- Forward Chaining
- Backward Chaining

Given a set of rules, i.e., formulae of the form

 $p_1 \wedge p_2 \wedge \ldots \wedge p_n \Longrightarrow q$

and a set of known facts, i.e., formulae of the form

q, *r*,...

- A new fact *p* is added
- Find all rules that have p as a premise
- If the other premises are already known to hold then
 - add the consequent to the set of know facts, and
 - trigger further inferences

Example

$$P \Rightarrow Q$$
$$L \land M \Rightarrow P$$
$$B \land L \Rightarrow M$$
$$A \land P \Rightarrow L$$
$$A \land B \Rightarrow L$$
$$A$$
$$B$$















- Soundness
 - Yes
- Completeness
 - Yes

- Given a set of rules, and a set of known facts
- We ask whether a fact P is a consequence of the set of rules and the set of known facts
- The procedure check whether P is in the set of known facts
- Otherwise find all rules that have P as a consequent
 - If the premise is a conjunction, then process the conjunction conjunct by conjunct

Example

$$P \Rightarrow Q$$
$$L \land M \Rightarrow P$$
$$B \land L \Rightarrow M$$
$$A \land P \Rightarrow L$$
$$A \land B \Rightarrow L$$
$$A$$
$$B$$











- Soundness
 - Yes
- Completeness
 - Yes

Transformation rules

$$\begin{array}{c} (\alpha \land \beta) \equiv (\beta \land \alpha) \\ (\alpha \lor \beta) \equiv (\beta \lor \alpha) \end{array} \end{array} \right\} \text{ Commutativity rules} \\ ((\alpha \land \beta) \land \gamma) \equiv (\alpha \land (\beta \land \gamma)) \\ ((\alpha \lor \beta) \lor \gamma) \equiv (\alpha \lor (\beta \lor \gamma)) \end{array} \right\} \text{ Associativity rules} \\ \neg (\neg \alpha) \equiv \alpha \qquad \text{Double-negation elimination} \\ (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) \qquad \text{Contraposition} \\ (\alpha \Rightarrow \beta) \equiv (\neg \alpha \lor \beta) \\ (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)) \\ \neg (\alpha \land \beta) \equiv (\neg \alpha \lor \neg \beta) \qquad \right\} \text{ de Morgan} \\ \neg (\alpha \lor \beta) \equiv ((\alpha \land \beta) \lor (\alpha \land \gamma)) \\ (\alpha \lor (\beta \land \gamma)) \equiv ((\alpha \lor \beta) \land (\alpha \lor \gamma)) \end{array} \right\} \text{ Distributivity}$$

Transformation rules (con't)

• $(A \lor (A \land B) \equiv A$ • $(A \land (A \lor B)) \equiv A$

- $A \land 0 \Leftrightarrow 0$
- $A \lor 1 \Leftrightarrow 1$
- ¬1 ⇔ 0

- $A \lor 0 \Leftrightarrow A$
- $A \land 1 \Leftrightarrow A$
- ¬0 ⇔ 1

• $\neg A \lor A \Leftrightarrow 1$

• $\neg A \land A \Leftrightarrow 0$

Transform into CNF

$\mathsf{B}_{1,1} \iff (\mathsf{P}_{1,2} \lor \mathsf{P}_{2,1})$

- 1. Remove \Leftrightarrow , replace $\alpha \Leftrightarrow \beta$ by $(\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)$. $(B_{1,1} \Rightarrow (P_{1,2} \lor P_{2,1})) \land ((P_{1,2} \lor P_{2,1}) \Rightarrow B_{1,1})$
- 2. Remove \Rightarrow , replace $\alpha \Rightarrow \beta$ by $\neg \alpha \lor \beta$. $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg (P_{1,2} \lor P_{2,1}) \lor B_{1,1})$
- 3. Move negation inward using the de Morgan's rule : $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land ((\neg P_{1,2} \land \neg P_{2,1}) \lor B_{1,1})$
- 4. Applying the "and" distribution rule : $(\neg B_{1,1} \lor P_{1,2} \lor P_{2,1}) \land (\neg P_{1,2} \lor B_{1,1}) \land (\neg P_{2,1} \lor B_{1,1})$

Example

 $(A \lor B) \rightarrow (C \rightarrow D)$

- 1. Remove \Rightarrow $\neg(A \lor B) \lor (\neg C \lor D)$
- 2. Move negation inward $(\neg A \land \neg B) \lor (\neg C \lor D)$
- 3. Distribution $(\neg A \lor \neg C \lor D) \land (\neg B \lor \neg C \lor D)$

Exercises

Transform the following expression into CNF.

- 1. $P \vee (\neg P \land Q \land R)$
- 2. $(\neg P \land Q) \lor (P \land \neg Q)$
- 3. $\neg(P \Rightarrow Q) \lor (P \lor Q)$
- 4. (P \Rightarrow Q) \Rightarrow R
- 5. $(P \Rightarrow (Q \Rightarrow R)) \Rightarrow ((P \land S) \Rightarrow R)$
- 6. $(\mathsf{P} \land (\mathsf{Q} \Rightarrow \mathsf{R})) \Rightarrow \mathsf{S}$
- 7. $\mathsf{P} \land \mathsf{Q} \Longrightarrow \mathsf{R} \land \mathsf{S}$
- 8. **((a∨b)∧c)→(c∧d)**

Priority: $\neg \land \lor \rightarrow \leftrightarrow$

$$\begin{array}{c} (\alpha \land \beta) \equiv (\beta \land \alpha) \\ (\alpha \lor \beta) \equiv (\beta \lor \alpha) \end{array} \\ ((\alpha \land \beta) \land \gamma) \equiv (\alpha \land (\beta \land \gamma)) \\ ((\alpha \lor \beta) \lor \gamma) \equiv (\alpha \land (\beta \land \gamma)) \\ \neg (\neg \alpha) \equiv \alpha \\ (\alpha \Rightarrow \beta) \equiv (\neg \beta \Rightarrow \neg \alpha) \\ (\alpha \Rightarrow \beta) \equiv (\neg \alpha \lor \beta) \\ (\alpha \Leftrightarrow \beta) \equiv ((\alpha \Rightarrow \beta) \land (\beta \Rightarrow \alpha)) \\ \neg (\alpha \land \beta) \equiv (\neg \alpha \lor \gamma) \\ \neg (\alpha \land \beta) \equiv (\neg \alpha \land \neg \beta) \end{array}$$

- $\mathsf{P} \lor (\neg \mathsf{P} \land \mathsf{Q} \land \mathsf{R})$
- $(\neg P \land Q) \lor (P \land \neg Q)$
- 3. $\neg(\mathsf{P} \Rightarrow \mathsf{Q}) \lor (\mathsf{P} \lor \mathsf{Q})$
- 4. $(P \Rightarrow Q) \Rightarrow R$
- 5. $(P \Rightarrow (Q \Rightarrow R)) \Rightarrow ((P \land S) \Rightarrow R)$
- $\text{6.} \quad (\mathsf{P} \land (\mathsf{Q} \Rightarrow \mathsf{R})) \Rightarrow \mathsf{S}$
- 7. $P \land Q \Longrightarrow R \land S$

Definitive and Horn Clauses

Definitive Clause

- Disjunction of literals of which *exactly one* is *positive*.
- Example: clause (¬L1,1 ∨ ¬Breeze ∨ B1,1) is a definite clause, whereas (¬B1,1 ∨ P1,2 ∨ P2,1) is not.

Horn Clause

- Disjunction of literals of which at most one is positive.
- All definite clauses are Horn clauses, as are clauses with no positive literals; these are called goal clauses.

Reading & Suggested Exercises

- Chapter 7
- Exercises 7.4, 7.5, 7.10, 7.19, 7.20