

Laurea MAGISTRALE in COMPUTER SCIENCE

ARTIFICIAL INTELLIGENCE

Logics / 2

Modal Logics – Non-Monotonic Reasoning, Reasoning under Uncertainty

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Non-classic Logics

- *Classic* logic
 - Propositional logic
 - First-Order Predicate logic
- *Non-classic* logics adopt different or more additional rules
 - To solve different problems than deductive calculus
 - To represent other forms of reasoning
 - Weaker
 - Linked to contextual factors

Extension/Modification Directions

- Using classical logic in a different way
- Abandoning truth-functionality principle
 - Truth value of a proposition not required anymore to be only a function of its components
 - Dismission of truth tables
- Abandoning bivalence principle
 - A proposition is not assumed anymore to be only true or false

Assumptions in Predicate Logic Reasoning

- 1. descriptions in the form of predicates must be “sufficient” to represent the application domain
 - Each fact is known to be true or false
 - What if information is incomplete or uncertain?
 - Logic programming (CWA): if a fact cannot be proved true, it is assumed to be false
 - Human reasoning: if a fact cannot be proven false, it is assumed to be true
- 2. knowledge base must be consistent
 - In human reasoning, alternate hypotheses are considered, even conflicting ones, and are removed as soon as new evidence is available

Assumptions in Predicate Logic Reasoning

- 3. knowledge monotonically grows through the use of inference rules
 - Need for mechanisms to
 - Add knowledge based on assumptions
 - Non-monotonic reasoning
 - Delete inferences based on such assumptions, in case subsequent evidence shows the hypothesized assumptions are wrong
 - Truth maintenance

Minsky's Criticisms to Logic

- “logical” reasoning not flexible for thinking
- Cannot handle inconsistent data
- Feasibility of representing knowledge by “true” propositions is doubtful
- Separation of knowledge and rules too radical
- Logic is monotonic
- Procedural descriptions over declarative descriptions

Classical Reasoning

- Systems based on predicate logic are conceptually elegant and rigorous
- Albeit logic is often unsuitable to represent common knowledge, it is used for its ease in automatizing the deduction process
- The *truth* of an assumed or derived assertion is such *indefinitely* and assumes the existence of a static world of objects
- Propositions in classical logic are declarative and atemporal
 - E.g.: Property P holds for object x
- In real problems less strict claims may happen

- Moreover, one does not want / cannot express
 - that some property is necessarily true, or
 - the difference between believing, supposing and knowing something
 - E.g., when formalizing juridic norms one must express
 - That taking an action is *prohibited* or *permitted*
 - And that if someone takes a prohibited action he *must* be punished
 - To represent common-sense reasoning, *modal formalism* were defined that allow to express in what way a proposition is true
 - Propositions may be characterized by a *mode* (necessary, possible, compulsory, ...)

Modal Logics

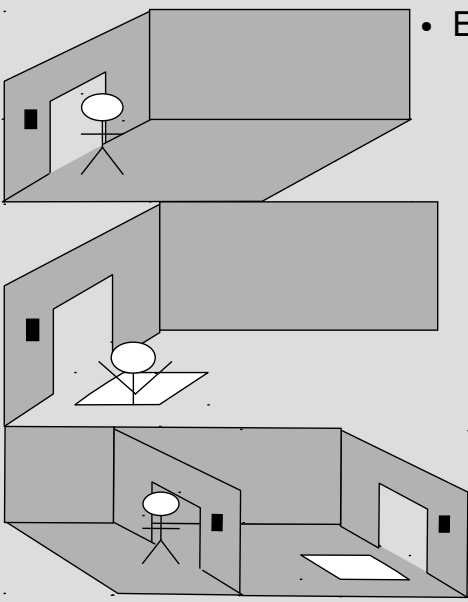
- A family of logics

- *Temporal*, handles propositions such as “It is always true that p ”, “Sometimes it is true that p ”, ...
- *Deontic*, handles propositions such as “ p is mandatory”, “ p is permitted”, ...
- *Epistemic*, handles propositions such as “I know that p ”, “I believe that p ”, ...
- *Alethic*, handles propositions such as “ p is necessary”, “ p is possible”, ...
- *Ethic*, handles propositions such as “ p is good”, “ p is bad”, ...

Modal Logics

- Allows to model/formalize intelligence

- To build intelligent agents that can operate in dynamic environments, like humans do

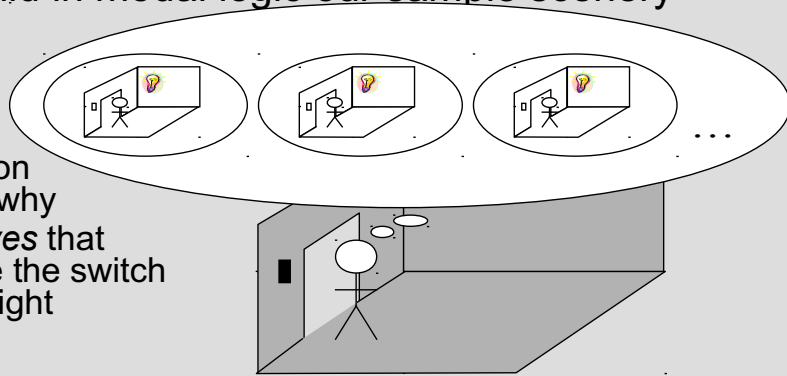


- Example:
 - Consider an agent moving in a dark room and, gropingly, discovering there is a switch on the wall
 - Understandably, the agent thinks the switch turns on the light...
- Instead, after pressing it, a trapdoor opens beneath him and he falls down...
- The agent escapes the trapdoor and, gropingly (he did not turn on the light), he reaches another room, again dark...
- He finds another switch... but this time he thinks twice before turning it on!

Modal Logics

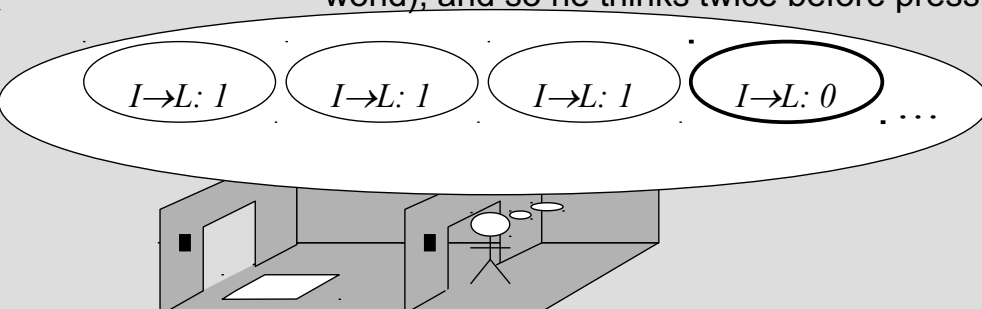
- Agent's action was wrong (he shouldn't turn on the switch), not stupid
 - Objective: building *rational*, not *omniscient* software agents
 - At any time, makes the choice it believes is the best
- Modal logics allow to build *rational agents*
 - Let us try and build in modal logic our sample scenery

- In all rooms in which the agent was in the past, the switch on the wall turned on the light; this is why the agent *believes* that also in this case the switch will turn on the light



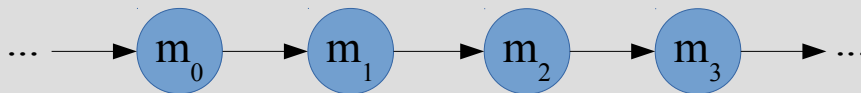
Modal Logics

- The rooms in which the agent remembers having been are just worlds accessible to him. In them, formula $I \rightarrow L$ holds, where I is fact “pushed button” and L “light is on”
- As a consequence, in the current room, i.e. the world in which the agent is moving, $(I \rightarrow L)$ is *necessarily/probably* true, even if, as the agent will discover, in this world $I \rightarrow L$ is false
- **Warning!** The agent has no access to the world in which he moves: he falls in the trapdoor because he cannot see that, in it, $I \rightarrow L$ is false
- When the agent is in the new room, he recalls that, in the previous one, $I \rightarrow L$ was false (now, he has access to an additional possible world), and so he thinks twice before pressing the button



Temporal Modal Logic

- Allows representing knowledge with temporal references
- Extend classical logic with operators such as
 - $\Box A$ “A is always true”
 - $\Diamond A$ “A will be sooner or later true”
 - OA “A will be true in the next temporal moment”
- Interpreted in a structure consisting of a discrete set of temporal intervals (*worlds*)



Temporal Operators

- Meaning defined by determining when a formula is true in a world w
 - $\Box A$ is true in w if it is true at timepoint w itself and in all subsequent ones
 - $\Diamond A$ is true in w if it is true at timepoint w itself or there exist a timepoint w' following w in which A is true
 - OA is true in w if A is true at the timepoint immediately following w

- Example: proposition $\text{leave}(\text{train}, \text{rome})$
 - “the train to Rome is leaving”
 - (a) $[\neg \text{leave}(\text{train}, \text{rome})] \Rightarrow [\Box \neg \text{leave}(\text{train}, \text{rome})]$
 - If the train to Rome is not leaving now, then it will never leave
 - (b) $[\neg \text{leave}(\text{train}, \text{rome})] \Rightarrow [\bigcirc \neg \text{leave}(\text{train}, \text{rome})]$
 - If the train to Rome is not leaving now, then it will leave in a moment
 - It might be appropriate to explicitly add time
 - $\text{leave}(t, \text{train}, \text{rome})$
 - In which case the logic must include an arithmetic
 - (a') $\neg \text{leave}(0, \text{train}, \text{rome}) \Rightarrow \forall t (0 \leq t \Rightarrow \neg \text{leave}(t, \text{train}, \text{rome}))$
 - Equivalent to (a)
 - (b') $\neg \text{leave}(0, \text{train}, \text{rome}) \Rightarrow \text{leave}(1, \text{train}, \text{rome})$
 - Equivalent to (b)

- Temporal aspects may also be represented by reifying propositions
 - treating them as objects to talk about
- using *higher level* predicates such as
 - $\text{holds}(p, t)$
 - “ p holds at time t ”
- Example: Formulas (a) and (b) above become
 - (a'') $\neg \text{holds}(\text{leave}(\text{train}, \text{rome}), 0) \Rightarrow \forall t (0 \leq t \Rightarrow \neg \text{holds}(\text{leave}(\text{train}, \text{rome}), t))$
 - (b'') $\neg \text{holds}(\text{leave}(\text{train}, \text{rome}), 0) \Rightarrow \text{holds}(\text{leave}(\text{train}, \text{rome}), 1)$

- One may even talk about the duration of an event, thanks to a predicate such as
 - $\text{happens}(p, t1, t2)$
 - “the duration of event p covers the temporal interval between $t1$ and $t2$ ”
 - Example: This allows to express the fact that if the train travels from Rome to Turin in interval $(t1, t2)$, then it arrives in Turin
 - $\text{happens}(\text{travel}(\text{train}, \text{rome}, \text{turin}), t1, t2) \Rightarrow \text{holds}(\text{arrive}(\text{train}, \text{turin}), t2)$

Deontic Logic

- Allows to express obligation or permission
 - E.g., in juridic norms, one must be able to express
 - Making an action is *prohibited* or *permitted*
 - If someone makes a *prohibited* action he *must* be punished
- Operators: given a proposition A
 - OA A is mandatory
 - PA A is permitted
 - $O \neg A$ A is prohibited
- Axioms describing operators features
 - $PA \Leftrightarrow \neg O \neg A$
 - All and only non-prohibited things are permitted
 - $OA \Rightarrow PA$
 - What is mandatory must be permitted

Reasoning about Knowledge

- Sometimes one needs to reason about knowledge and about reasoning itself
 - “If I know that the train to Turin leaves at time X I go to the station ten minutes before time X; if I have no idea about the time then I check the Internet”
 - “If it is not possible to prove that the defendant is guilty then one must conclude he is innocent”
- Need for modal logics that can work on a *meta-level*
 - Meta-knowledge
 - Meta-reasoning

Epistemic Modal Logics

- Knowledge \neq Belief
- Operators
 - K denotes what an agent knows
 - KA the agent knows that proposition A is true
 - B denotes what an agent believes
 - BA the agent believes that proposition A is true
- Examples
 - $K \exists x \text{ thief}(x,C)$
 - The agent knows that *someone* stole the picture
 - $K \forall x (\text{thief}(x,C) \Rightarrow \text{previous_offender}(x))$
 - $\exists x K \text{ thief}(x,C)$
 - The agent knows *who* stole the picture

Epistemic Modal Logics

- What is known is true ($KA \Rightarrow A$), but an agent may believe something false
- Possibility of representing situations with many agents and to use modal operators to represent what agents know or believe about each other
 - Example: agents α and β
 - $K(\alpha, A) \Leftrightarrow K_\alpha(A)$ agent α knows that A is true
 - $B_\alpha(A)$ agent α believes that A is true
 - $K_\alpha K_\beta A$ agent α knows that agent β knows A

Well-Formed Formulas

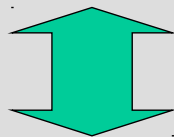
- Syntax in first-order modal languages
 - All wff in first-order predicate calculus are also wff in the modal language
 - If Φ is a closed formula in the modal language
 - Without free variablesand α is a ground term, then $K(\alpha, \Phi)$ is a wff in the modal language
 - If Φ and Ψ are wff, then all expressions that may be built from Φ and Ψ using the usual propositional connectives are wff, as well

Monotonic Reasoning

- Scheme based on certainty of deduction
 - Principle 1: In a monotonic reasoning system, the number of propositions proved true continuously increases
 - Knowledge is
 - Complete
 - All facts needed to solve a problem are present or may be derived)
 - Consistent
 - Updatable only by adding new facts that are consistent with those already asserted (monotonicity)

Non-monotonic Reasoning

- Often information may be incomplete, albeit temporarily, or conditions may change in time



- Need to formulate (even incorrect) hypotheses, to restart the reasoning process when it reaches a deadend
- *Non-monotonic* reasoning system
 - Keeps track of a set of hypothetical assumptions and continuously revises them based on new observed or deduced knowledge

Non-monotonic Reasoning

- Classical logic (FOPL)
 - Once a fact is asserted, its' forever true
 - Theorems never decrease with the increase of axioms
- Non-monotonic Reasoning
 - The set of currently true (or believed) facts does NOT increase monotonically
 - Can shrink or grow with reasoning
 - Adding a new fact might lead to an inconsistency
 - Need to remove one of the contradictory facts
 - Hypotheses based on default assumptions evolve as long as new information is acquired
 - Adding new information may change the correctness of the conclusion
 - From a formal viewpoint, adding new axioms to a theory T, not necessarily all theorems in T are preserved

Non-monotonic Reasoning

- Example
 - You are a student, it's 8am, you are in bed.
 - You slip out of your dreams and think: Today is Sunday. No classes today. I don't have to get up. You go back to sleep.
 - You wake up again. It's 9:30am now and it is slowly coming to your mind: Today is Tuesday. What an unpleasant surprise.

P1 = today-is-Tuesday

P3 = have-class-at-10am

P5 = have-to-get-up

P2 = today-is-Sunday

P4 = no-classes

P6 = can-stay-in-bed

Non-monotonic Reasoning

- 3 problems to be solved
 - How to extend the knowledge base so as to allow inferences based both on the absence and on the presence of knowledge
 - It is known that $\neg P$
 - It is not known if P
 - How to update correctly the knowledge base when a new fact is added or an old one is removed
 - How to use knowledge to try and solve conflicts that are generated when different non-monotonic, mutually inconsistent, inferences can be carried out

Non-monotonic Reasoning

- A formalism should:
 - 1. Be able to define the set of worlds that may exist, given the facts that are certain
 - I.e., allow to define the set of models of any set of wffs
 - Interpretation of a set of wffs:
a domain (set of objects) D with a function that maps
 - Each predicate to a relationship
 - Each n -ary function to an operator from D^n to D
 - Each constant to a member of D
 - Model of a set of wffs: an interpretation that satisfies them
 - 2. provide a way to choose a model
 - 3. provide the basis for implementing reasoning
 - 4. provide a model whose conclusions correspond to our intuitions

Non-monotonic Reasoning

- Many formalisms proposed
 - Default logic (Reiter, 1987)
 - Non-monotonic logic (McDermott, Doyle, 1980)
 - Autoepistemic logic (Moore, 1985)
 - Circumscription (McCarthy, 1980)
 - Negation as Failure (Clark, 1978)
- All have to do with what we “know” at a given moment
 - Central question in the logic formalism has to do with the *extension* of operator \exists

Non-monotonic Reasoning

- Example: “All A’s are B’s”
 - A statement that many logics try to “weaken”, to reach more generality
 - Classical logic
 - $\forall x : x \in A \Rightarrow x \in B$
 - Default logic
 - An A is a B, unless exceptions
 - Non-monotonic modal logic
 - If B is conceivable and if A, then B
 - Autoepistemic logic
 - If a given A were not B, we should know
 - Circumscription
 - Any A which is not abnormal is B
 - Negation as Failure
 - If, given A, B cannot be proved, then B is assumed to be false

Non-monotonic Logic

- Extends the FOPL language with a modal operator M, meaning “is consistent”
 - Example: formula
 - $\forall x,y: \text{brothers}(x,y) \wedge M \text{ go_along_well}(x,y) \Rightarrow \text{defends}(x,y)$
 - “For all x’s and y’s, if x and y are brothers and it is conceivable that x goes along well with y, then one concludes that x will defend y”
- Need to define the meaning “is consistent” for the theory to be at least semi-decidable
 - Normally the Prolog notion of Negation as Failure or some variant of it is used
 - To show that P is consistent:
 - Try and prove $\neg P$
 - If failure: assume $\neg P$ to be false, declare P as consistent

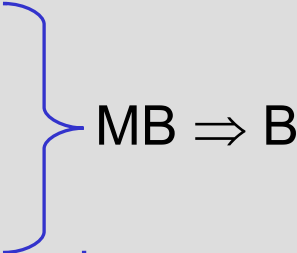
Non-monotonic Reasoning

- Problem: possibility that many non-monotonic propositions, taken singularly, suggest extensions that are overall inconsistent
 - Example
 - $\forall x : \text{ignorant}(x) \wedge M \text{ studies}(x) \Rightarrow \text{studies}(x)$
 - $\forall x : \text{studies}(x) \wedge M \text{ passes_test}(x) \Rightarrow \text{passes_test}(x)$
 - $\forall x : \text{ignorant}(x) \Rightarrow \neg \text{passes_test}(x)$
 - $\text{ignorant}(\text{pippo}) \quad \Rightarrow \quad \text{studies}(\text{pippo})$

\Downarrow

\Downarrow
 - $\neg \text{passes_test}(\text{pippo}) \quad \quad \quad \text{passes_test}(\text{pippo})$

Non-monotonic Logics

- Defines the set of theorems that can be derived from a set A of wffs as the intersection of the sets of theorems resulting from all possible ways in which the wffs in A can be combined
 - Assertions, albeit looking like rules, are wffs that can be manipulated using the traditional laws to combine logistic expressions
 -
 - $A \wedge MB \Rightarrow B$
 -
 - $\neg A \wedge MB \Rightarrow B$
- 
- $MB \Rightarrow B$

Default Logics (DL)

- Introduces a new class of rules

$$\frac{A : B}{C}$$

- “If A is true (provable) and it is consistent to assume B , then conclude C ”
 - A *requirement*
 - B *justification*
 - C *conclusion*

Default Logics (DL)

- Example

- Sentence “Birds typically fly” can be expressed as

$$\frac{\text{bird}(x) : \text{fly}(x)}{\text{fly}(x)}$$

- or

$$\frac{: \text{bird}(x) \Rightarrow \text{fly}(x)}{\text{bird}(x) \Rightarrow \text{fly}(x)}$$

- “If it is conceivable that birds fly then I conclude that birds fly”

- A belief becomes a theorem!

Default Logic (DL)

- Extension: a set of beliefs

- The new inference rules are used as a basis to compute a set of plausible extensions of the knowledge base
- Each extension ‘increments’ the knowledge base
 - Maximal consistent augmentation
- Generated by applying inference rules without violating consistency
- Logics admits as theorem an expression that is satisfied in some extension
- If it’s needed to decide among extension, a decision is not made

Default Logic (DL)

- Expressions are inference rules rather than expressions in the language
 - Cannot be manipulated by other inference rules
 - Example: given two rules and no assertion about A
 - From

$$\frac{A : B}{B}$$

$$\frac{\neg A : B}{B}$$

- no conclusion can be drawn about B

Non-monotonic Reasoning

- Inheritance
 - Non-monotonic reasoning typically used in object-centered representation in order for instances of a class to inherit some attributes from a typical description of the class itself
 - Algorithm based on the fact that an object inherits an attribute value from all classes to which it belongs, *unless this leads to a contradiction*,
 - in which case a value coming from a more specific class has priority over that of a more general one

Non-monotonic Reasoning

- Use of *delineation* to deduce properties
 - Always requires to proceed from specific to general
 - With such an ordering defined on matching and retrieval processes, information present in lower levels protects the system from contradictory deductions based on higher levels
 - Example
 - Rule that considers inheritance of a default value for the height of a basketball player
$$\frac{\text{basketball_player}(x) : \text{height}(x, 1.85)}{\text{height}(x, 1.85)}$$
 - In any case, a declared value blocks inheritance of a default value

Non-monotonic Reasoning

- When conflicts arise due to multiple inheritance, *delineation by default*, using this kind of rules, can be used

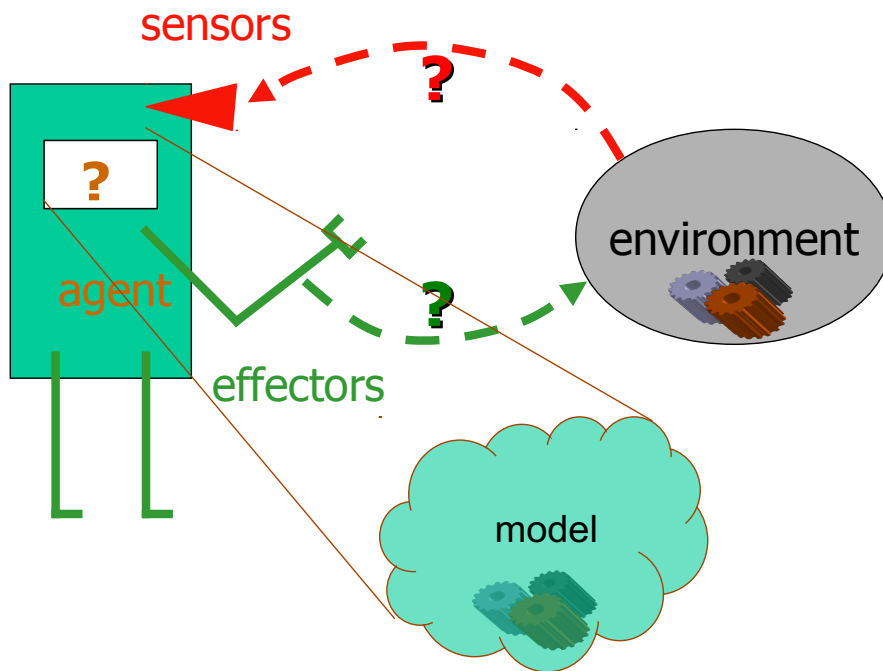
$$\frac{\text{elephant}(x) : \text{color}(x, \text{gray})}{\text{color}(x, \text{gray})}$$

$$\frac{\text{albino}(x) : \text{color}(x, \text{white})}{\text{color}(x, \text{white})}$$

$$\frac{\text{elephant}(x) : \neg \text{albino}(x) \wedge \text{color}(x, \text{gray})}{\text{color}(x, \text{gray})}$$

- This rule can prevent application of default knowledge about elephants when more specific knowledge coming from class albino is available

Agent under Uncertain Conditions



Kinds of Uncertainty

- Uncertainty in prior knowledge
 - Example: in an expert system for medical diagnosis, some causes of the disease are not known and not represented in the knowledge base
- Uncertainty in action
 - Example: it is expected that the agent must turn on the light when entering a room, but it is necessary/expected that there is power, that the switch works, that the bulb is not blown, etc.
- Uncertainty in perception
 - Example: sensors do not return the exact position, the environment is not enough well-lit, etc.

Fuzzy Theory

- A mathematical theory that encodes *qualitative* evaluations
 - Particularly suited for what people express in everyday language
 - Allows to deal with knowledge expressed qualitatively
- An extension of Boolean logics to values in the *continuous* range $[0,1]$
 - A *fuzzy* expression must not be true or false, but several *degrees of truth* are admitted

Fuzzy Logics

- Used to describe and operate with vague definitions
 - Example (controlling a cement plant)
 - If temperature is high, add few cement and much increase water
- Based on the idea that the elements of a set are defined through a *degree of membership*
 - Increased expressive power: many quantities can be represented in a fuzzy way
 - Examples:
 - The engine is very hot
 - John is very tall

Fuzzy Logics

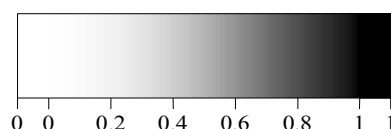
- Typically used
 - Strictly speaking
 - It is a branch of fuzzy set theory, that deals with knowledge representation and inference
 - It deals with imprecise knowledge
 - Broadly speaking
 - Considered as a synonym of fuzzy set theory

Fuzzy Logics

- A set of mathematical principles for knowledge representation based on *degrees of membership* to a set
- Uses a continuum of logic values between 0 (completely false) and 1 (completely true)



(a) Boolean Logic.



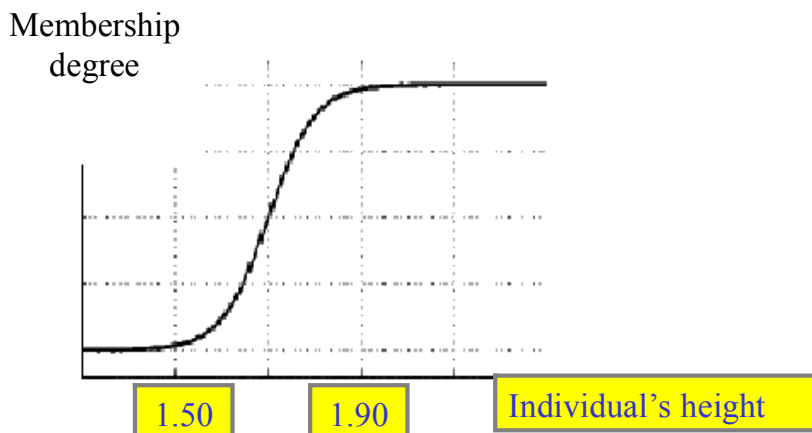
(b) Multi-valued Logic.

Fuzzy Logics

- The degree of membership can be interpreted as a probability
 - Logic mechanisms of probabilistic type can be defined
 - Example
 - $\text{very_bad} = [0, 0.2]$; $\text{bad} = [0.2, 0.5]$;
 $\text{good} = [0.5, 0.8]$; $\text{very_good} = [0.8, 1]$
 - With different degrees of membership, e.g.:
 - 0 is very_bad with degree 1
 - 0.2 is bad with degree 0.5 and bad with degree 0.5
 - etc.
 - The *membership function* of a fuzzy set associates the features of each element (or *instance*, or *individual*) the value of the degree of membership to the set itself

Fuzzy Logics

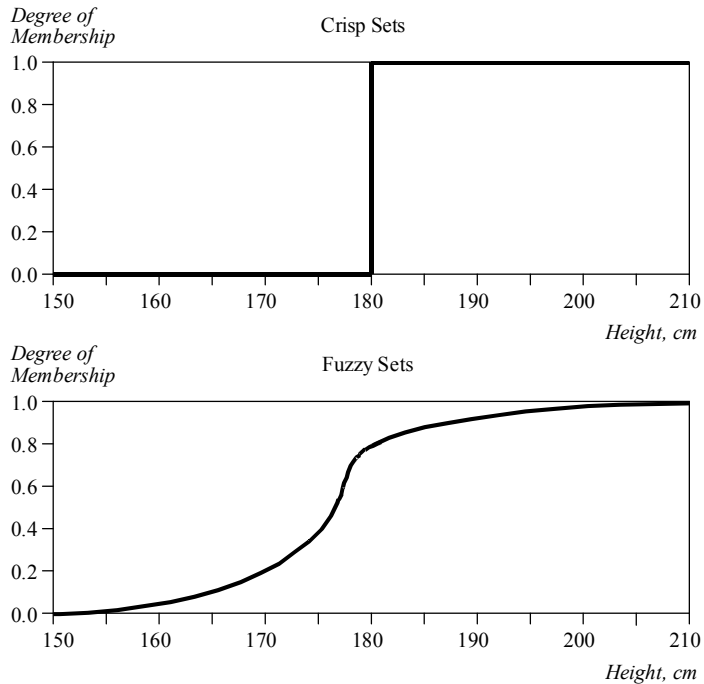
- Many kinds of membership functions are available
 - Can be chosen for pure mathematical convenience or determined through experimental observations
 - Example: membership function for fuzzy category “quite tall”



Fuzzy Logics

- Traditional (*Crisp*) sets vs Fuzzy sets

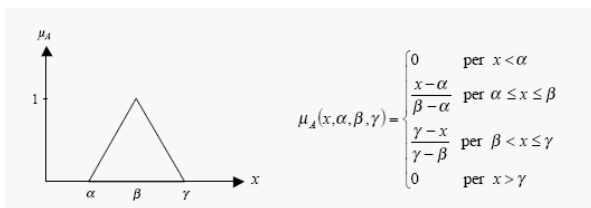
- x-axis = universe of discourse
 - All possible values applicable to a given variable
- y-axis = value of membership to the fuzzy set



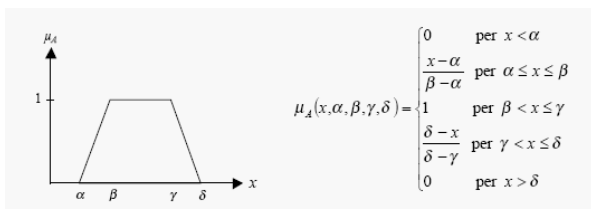
Fuzzy Theory

- Membership functions

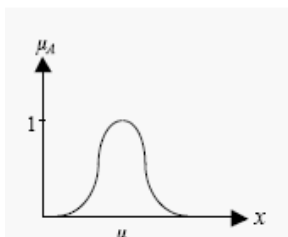
- Triangular



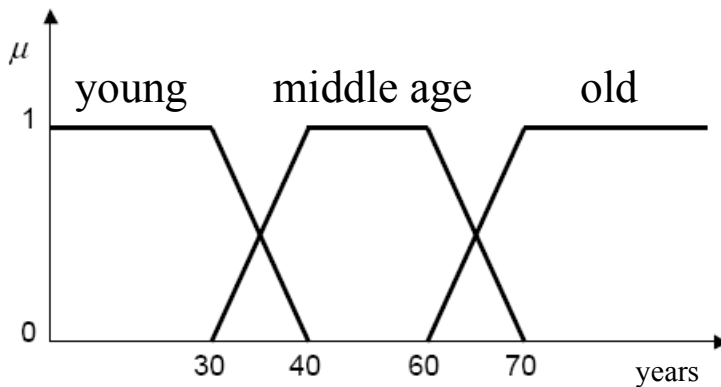
- Trapezoidal



- Gaussian

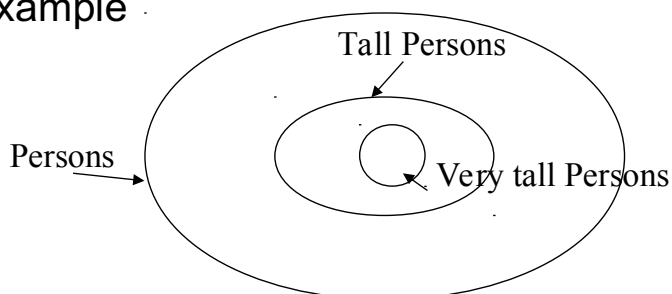


- Example: age of a person
 - A 30-years old person is young
 - How is a 31-years old person defined?
- The fuzzy approach:



Fuzzy Logics

- Membership
 - In traditional sets, all elements in a set entirely belong to the superset
 - In fuzzy sets, each element may or may not belong both to the subset and to the superset
 - An element of a fuzzy set may have less degree of membership to the subset than to the superset
 - Example

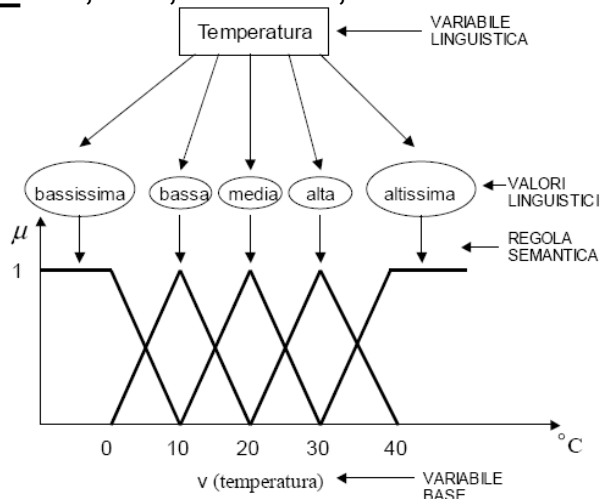


Fuzzy Theory

- Based on *linguistic variables*
- Linguistic variable
 - A fuzzy variable whose values are linguistic terms
 - E.g., the claim “John is tall” implies that linguistic variable John has linguistic value “tall”
 - In rule-based fuzzy systems, linguistic variables are used in *fuzzy rules*
 - E.g.:
 - IF wind is strong
 - THEN sailing is fine
 - or
 - IF speed is slow
 - THEN stopping_distance is short

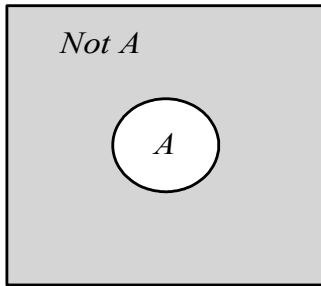
Fuzzy Theory

- Linguistic variables
 - Variable whose values are words or sentences in a natural or artificial language
 - Example: Variable *temperature* whose values are *very_low*, *low*, *medium*, etc.

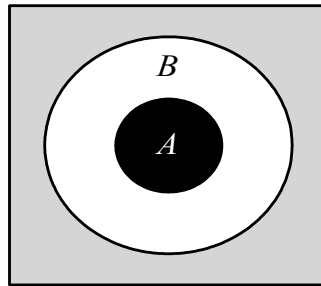


Fuzzy Sets

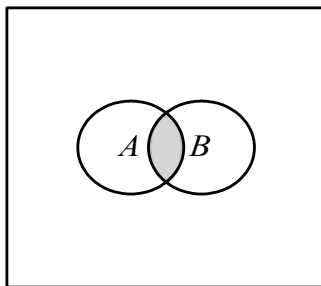
- Operations



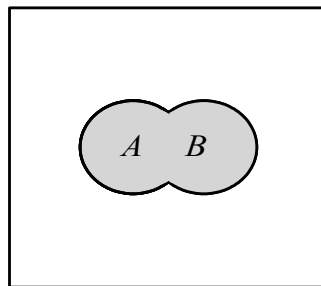
Complement



Containment



Intersection



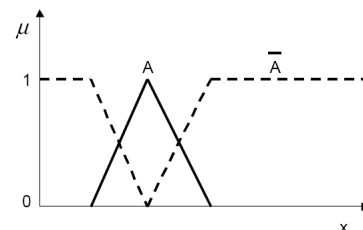
Union

Fuzzy Sets

- Complement

- Traditional sets: which elements do not belong to the set?
 - Opposite of the set
 - E.g.: the complement of the set of tall persons is the set of NOT tall persons
 - Removing tall persons from the universe of discourse, the complement is obtained
- Fuzzy sets: by what degree an element does not belong to the set?
 - Complement $\sim A$ of a fuzzy set A obtained as follows:

$$\mu_{\sim A}(x) = 1 - \mu_A(x)$$

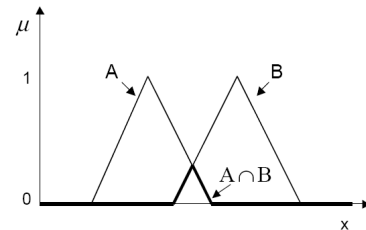


Fuzzy Sets

- Intersection

- Traditional sets: Which elements belong to both sets?
 - Shared elements
- Fuzzy sets: How much an element belongs to both sets?
 - An element may partially belong to the two sets with different degrees of membership
 - Defined as the lowest degree of belonging
 - Intersection between two fuzzy sets A and B on the universe of discourse X:

- $$\mu_{A \cap B}(x) = \min[\mu_A(x), \mu_B(x)]$$
$$= \mu_A(x) \cap \mu_B(x)$$
where $x \in X$

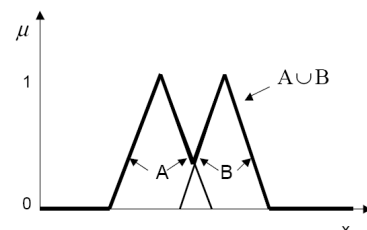


Fuzzy Sets

- Union

- Traditional sets: which elements belong to either or both sets?
 - Elements belonging to at least one set
- Fuzzy sets: how much an element belongs to either or both sets?
 - Reverse of intersection
 - Defined as the highest degree of belonging
 - Intersection between two fuzzy sets A and B on the universe of discourse X:

- $$\mu_{A \cup B}(x) = \max[\mu_A(x), \mu_B(x)]$$
$$= \mu_A(x) \cup \mu_B(x)$$
where $x \in X$



Fuzzy Sets

• Operators

• AND (Intersection)

$$\mu_{A \cap B}(x) = \min(\mu_A(x), \mu_B(x)) \quad \forall x \in X$$

•

•

• OR (Union)

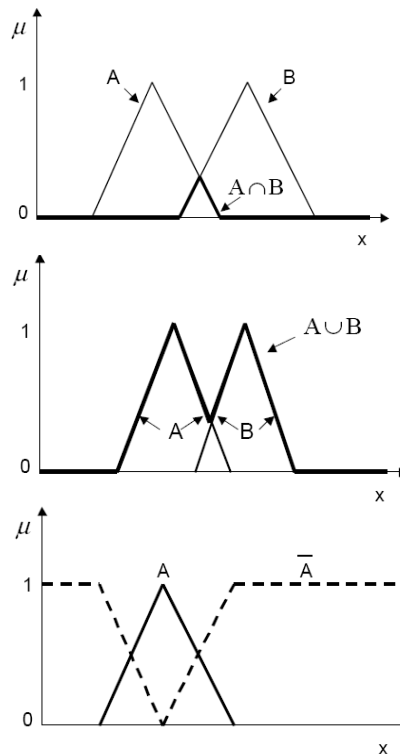
$$\mu_{A \cup B}(x) = \max(\mu_A(x), \mu_B(x)) \quad \forall x \in X$$

•

•

• NOT (Complement)

$$\mu_{\bar{A}}(x) = 1 - \mu_A(x) \quad \forall x \in X$$



Fuzzy Sets

• Properties

• Equality

– A fuzzy set is equal to another if

$$\mu_A(x) = \mu_B(x), \quad \forall x \in X$$

• Example

$$A = 0.3/1 + 0.5/2 + 1/3$$

$$B = 0.3/1 + 0.5/2 + 1/3$$

• Thus, $A = B$

• Inclusion

– A fuzzy set A, $A \subseteq X$, is included in a fuzzy set B, $B \subseteq X$, if

$$\mu_A(x) \leq \mu_B(x), \quad \forall x \in X$$

– A is a subset of B

• Example

• Consider $X = \{1, 2, 3\}$ and sets A and B

$$A = 0.3/1 + 0.5/2 + 1/3$$

$$B = 0.5/1 + 0.55/2 + 1/3$$

• Then, $A \subseteq B$

Further Readings

- P. Smets, E.H. Mamdani, D. Dubois, H. Prade (Eds.) “Non–Standard Logics for Automated Reasoning” Academic Press, 1988.
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- G. Brewka, I. Niemela, M. Truszcynski “Non Monotonic Reasoning” 2007 Elsevier
- L. Zadeh “Fuzzy Logic and approximate reasoning” Synthese, 30: 407-428, 1975