

All About Fuzzy Description Logics

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About Vagueness

On the Existence of Vague Concepts

What are vague concepts and do they exist?

- What are the pictures about?



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- A **concept is vague** whenever its extension is deemed lacking in clarity
 - **Aboutness** of a picture or piece of text
 - **Tall** person
 - **High** temperature
 - **Nice** weather
 - **Adventurous** trip
 - **Similar** proof
- Vague concepts:
 - Are abundant in everyday speech and almost inevitable
 - Their meaning is often subjective and context dependent

On the Existence of Vague Objects

What are vague objects and do they exist?

- Are there vague objects in the pictures?



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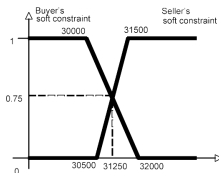


- An **object is vague** whenever its identity is lacking in clarity
 - **Dust**
 - **Cloud**
 - **Dunes**
 - **Sun**
- Vague objects:
 - Are not identical to anything, except to themselves (reflexivity)
 - Are characterised by a **vague identity** relation (e.g. a **similarity** relation)
- BTW: example of *uncertain object*: “habitable Earth-like planet in universe”

Vague Statements

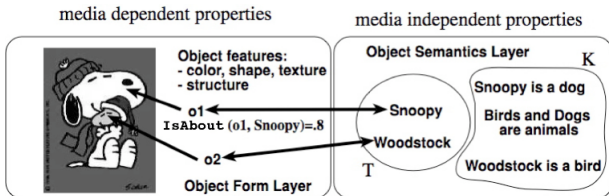
- A **statement is vague** whenever it involves vague concepts or vague objects
 - **Heavy** rain
 - **Tall** person
 - **Hot** temperature
- The **truth** of a vague statement is a matter of **degree**, as it is intrinsically difficult to establish whether the statement is entirely true or false
 - There are 33 °C. Is it **hot**?

Sources of Vagueness: Matchmaking



- A car seller sells an Audi TT for 31500 €, as from the catalog price.
- A buyer is looking for a sports-car, but wants to pay not more than around 30000 €
- Classical DLs: the problem relies on the crisp conditions on price.
- More fine grained approach: to consider prices as vague constraints (fuzzy sets) (as usual in negotiation)
 - Seller would sell above 31500 €, but can go down to 30500 €
 - The buyer prefers to spend less than 30000 €, but can go up to 32000 €
 - Highest degree of matching is 0.75 . The car may be sold at 31250 €.

Sources of Vagueness: Multimedia information retrieval

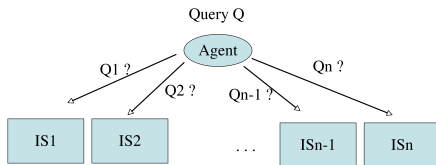


<i>IsAbout</i>		
<i>ImageRegion</i>	<i>Object ID</i>	<i>degree</i>
$o1$	<i>snoopy</i>	0.8
$o2$	<i>woodstock</i>	0.7
\vdots	\vdots	
\vdots	\vdots	

“Find top- k image regions about animals”

$Query(x) \leftarrow ImageRegion(x) \wedge isAbout(x, y) \wedge Animal(y)$

Sources of Vagueness: Distributed Information Retrieval



Then the agent has to perform **automatically** the following steps:

- 1 The agent has to select a subset of relevant resources $\mathcal{S}' \subseteq \mathcal{S}$, as it is not reasonable to assume to access to and query all resources (**resource selection/resource discovery**);
- 2 For every selected source $S_i \in \mathcal{S}'$ the agent has to reformulate its information need Q_A into the query language \mathcal{L}_i provided by the resource (**schema mapping/ontology alignment**);
- 3 The results from the selected resources have to be merged together

Sources of Vagueness: Vague database query

<i>HotelID</i>	<i>hasLoc</i>	<i>ConferenceID</i>	<i>hasLoc</i>
<i>h1</i>	<i>hl1</i>	<i>c1</i>	<i>cl1</i>
<i>h2</i>	<i>hl2</i>	<i>c2</i>	<i>cl2</i>
⋮	⋮	⋮	⋮

<i>hasLoc</i>	<i>hasLoc</i>	<i>distance</i>	<i>hasLoc</i>	<i>hasLoc</i>	<i>close</i>	<i>cheap</i>
<i>hl1</i>	<i>cl1</i>	300	<i>hl1</i>	<i>cl1</i>	0.7	0.3
<i>hl1</i>	<i>cl2</i>	500	<i>hl1</i>	<i>cl2</i>	0.5	0.5
<i>hl2</i>	<i>cl1</i>	750	<i>hl2</i>	<i>cl1</i>	0.25	0.8
<i>hl2</i>	<i>cl2</i>	800	<i>hl2</i>	<i>cl2</i>	0.2	0.9
⋮	⋮		⋮	⋮	⋮	

“Find top-*k* cheapest hotels close to the train station”

$$q(h) \leftarrow \text{hasLocation}(h, hl) \wedge \text{hasLocation}(\text{train}, cl) \wedge \text{close}(hl, cl) \wedge \text{cheap}(h)$$

Sources of Vagueness: Health-care: diagnosis of pneumonia



Health Care Guideline:

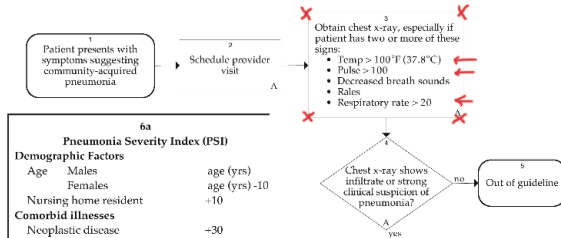
Community-Acquired Pneumonia in Adults

INSTITUTE FOR CLINICAL
SYSTEMS IMPROVEMENT

Seventh Edition
May 2006

Work Group Leader
John Degelau, MD
*Internal Medicine,
HealthPartners Medical Group*

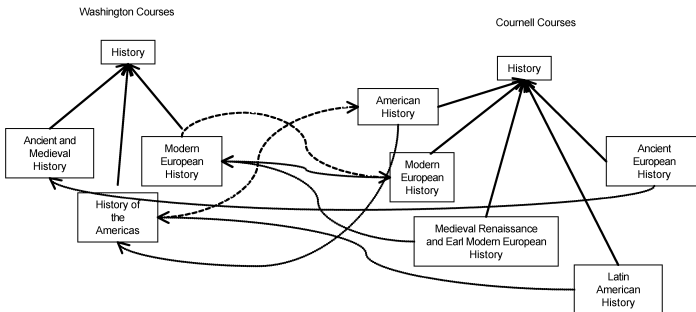
Work Group Members
Family Medicine
Garrett Trobec, MD



- E.g., $Temp = 37.5$, $Pulse = 98$, $RespiratoryRate = 18$ are in “danger zone” already
- Temperature, Pulse and Respiratory rate: these constraints are rather vague

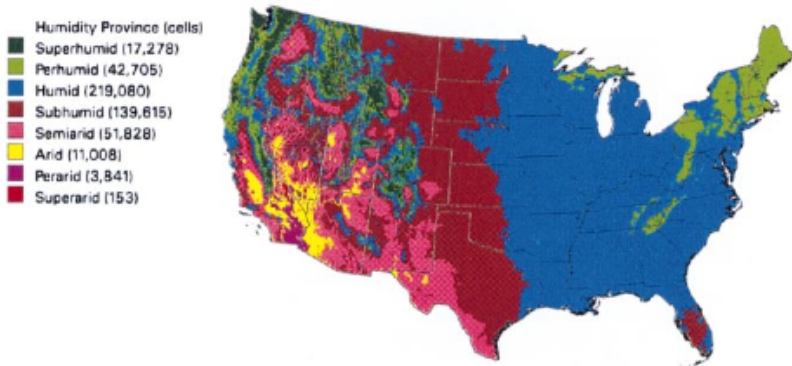
Sources of Vagueness: Ontology alignment (schema matching)

- To which **degree** are two concepts of two ontologies similar?



Sources of Vagueness: Lifezone mapping

- To which **degree** do certain areas have a specific bioclimate



Holdridge life zones of USA

Sources of Vagueness: ARPAT, Air quality in the province of Lucca

I dati di domenica 21/02/2010

Sintesi dei dati rilevati dalle ore 0 alle ore 24 del giorno domenica 21/02/2010

Stazione		Tipo stazione	SO ₂ µg/m ³ (media su 24h)	NO ₂ µg/m ³ (max oraria)	CO mg/m ³ (max oraria)	O ₃ µg/m ³ (max oraria)	PM ₁₀ µg/m ³ (media su 24h)	Giudizio di qualità dell'aria
Lucca	P.za San Michele (RETE REGIONALE **)	urbana - traffico	1	75	---	---	37	Accettabile
Lucca	V.le Carducci	urbana - traffico	1	---	2,3	---	49	Accettabile
Lucca	Carignano (RETE REGIONALE **)	rurale - fondo	---	---	---	86 (h.16*)	---	Buona
Viareggio	Largo Risorgimento	urbana - traffico	---	---	1,8	---	n.d.	Buona
Viareggio	Via Maroncelli (RETE REGIONALE **)	urbana - fondo	4	97	---	61 (h.15*)	33	Accettabile
Capannori	V. di Piaggia (RETE REGIONALE **)	urbana - fondo	---	62	1,3	---	25	Accettabile
Porcari	V. Carrara (RETE REGIONALE **)	periferica - fondo	1	51	---	84 (h.15*)	24	Accettabile

Giudizio di qualità	SO ₂ µg/m ³ (media su 24h)	NO ₂ µg/m ³ (max oraria)	CO mg/m ³ (max oraria)	O ₃ µg/m ³ (max oraria)	PM ₁₀ µg/m ³ (media su 24h)
Buona	0-50	0-50	0-2,5	0-120	0-25
Accettabile	51-125	51-200	2,6-15	121-180	26-50
Scadente	126-250	201-400	15,1-30	181-240	51-74
Pessima	>250	>400	>30	>240	>74

Sintesi dei dati rilevati dalle ore 0 alle ore 24 del giorno domenica 14/02/2010

Stazione		Tipo stazione	SO ₂ µg/m ³ (media su 24h)	NO ₂ µg/m ³ (max oraria)	CO mg/m ³ (max oraria)	O ₃ µg/m ³ (max oraria)	PM ₁₀ µg/m ³ (media su 24h)	Giudizio di qualità dell'aria
Lucca	P.za San Micheletto (RETE REGIONALE **)	urbana - traffico	1	75	---	---	56	Scadente
Lucca	V.le Carducci	urbana - traffico	2	---	2	---	75	Pessima
Lucca	Carignano (RETE REGIONALE **)	rurale - fondo	---	---	---	87 (h.18*)	---	Buona
Viareggio	Largo Risorgimento	urbana - traffico	---	---	1,7	---	n.d.	Buona
Viareggio	Via Maroncelli (RETE REGIONALE **)	urbana - fondo	1	121	---	60 (h.17*)	45	Accettabile
Capannori	V. di Piaggia (RETE REGIONALE **)	urbana - fondo	---	79	2	---	59	Scadente
Porcari	V. Carrara (RETE REGIONALE **)	periferica - fondo	2	72	---	82 (h.16*)	63	Scadente

Giudizio di qualità	SO ₂ µg/m ³ (media su 24h)	NO ₂ µg/m ³ (max oraria)	CO mg/m ³ (max oraria)	O ₃ µg/m ³ (max oraria)	PM ₁₀ µg/m ³ (media su 24h)
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TripAdvisor: Hotel User Judgments

2,889 Reviews from our TripAdvisor Community

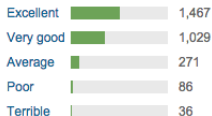


Your overall rating of this property



[Click to rate](#)

Traveler rating



See reviews for



Rating summary



Uncertainty vs Vagueness: a clarification

- Initial difficulty:
 - Understand the conceptual differences between **uncertainty** and **vagueness**
- Main problem:
 - Interpreting a **degree** as a measure of **uncertainty** rather than as a measure of **vagueness**

Uncertain Statements

- A statement is **true** or **false** in any world/interpretation
 - We are “**uncertain**” about which world to consider
 - We may have e.g. a probability distribution over possible worlds
- E.g., “it will rain tomorrow”
 - We cannot exactly establish whether it will rain tomorrow or not, due to our **incomplete** knowledge about our world
 - We can estimate to which **degree** this is **probable**

- Consider a propositional statement (formula) ϕ
- Interpretation (world) $\mathcal{I} \in \mathcal{W}$,

$$\mathcal{I} : \mathcal{W} \rightarrow \{0, 1\}$$

- $\mathcal{I}(\phi) = 1$ means ϕ is true in \mathcal{I} , denoted $\mathcal{I} \models \phi$
- Each interpretation \mathcal{I} depicts some concrete world
- Given n propositional letters, $|\mathcal{W}| = 2^n$
- In uncertainty theory, we do not know which interpretation \mathcal{I} is the actual one

- One may construct a **probability distribution** over the worlds

$$Pr : \mathcal{W} \rightarrow [0, 1]$$
$$\sum_{\mathcal{I}} Pr(\mathcal{I}) = 1$$

- $Pr(\mathcal{I})$ indicates the probability that \mathcal{I} is the actual world
- **Probability** $Pr(\phi)$ of a statement ϕ in Pr

$$Pr(\phi) = \sum_{\mathcal{I} \models \phi} Pr(\mathcal{I})$$

- $Pr(\phi)$ is the probability of the event: " ϕ is true"

Vague Statements

- A statement is **true** to some **degree**, which is taken from a truth space (usually $[0, 1]$)
- The convention prescribing that a proposition is either true or false is changed towards **graded propositions**
- E.g., “heavy rain”
 - The compatibility of “heavy” in the phrase “heavy rain” is graded and the degree depends on the amount of rain is falling
 - The intensity of precipitation is expressed in terms of a precipitation rate R : volume flux of precipitation through a horizontal surface, i.e. $m^3/m^2s = ms^{-1}$
 - It is usually expressed in mm/h

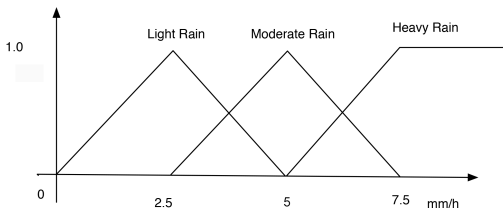
“Heavy rain” continued...E.g., in weather forecasts one may find:

- Rain intensity measured as precipitation rate R : volume flux of precipitation through a horizontal surface, i.e. $m^3/m^2h = mh^{-1}$

Rain.	Falling drops of water larger than 0.5 mm in diameter. “Rain” usually implies that the rain will fall steadily over a period of time;
Light rain.	Rain falls at the rate of 2.6 mm or less an hour;
Moderate rain.	Rain falls at the rate of 2.7 mm to 7.6 mm an hour;
Heavy rain.	Rain falls at the rate of 7.7 mm an hour or more.

- Quite harsh distinction: $R = 7.7mm/h \rightarrow$ heavy rain
 $R = 7.6mm/h \rightarrow$ moderate rain
- This is clearly unsatisfactory, as quite naturally
 - The more rain is falling, the more the sentence “heavy rain” is true
 - Vice-versa, the less rain is falling the less the sentence is true

- In other words, that the sentence “heavy rain” is no longer either true or false, but is **intrinsically graded**
 - Even if we have complete knowledge about the current world, i.e. exact specification of the precipitation rate
- More fine grained approach:
 - Define the various types of rains as



- Light rain, moderate rain and heavy rain are **vague concepts**

- Consider a propositional statement ϕ
- A propositional interpretation \mathcal{I} maps ϕ to a truth degree in $[0, 1]$

$$\mathcal{I}(\phi) \in [0, 1]$$

- I.e., we are unable to establish whether a statement is entirely true or false due the occurrence of vague concept
- Vague statements are truth-functional
 - Degree of truth of a statement can be calculated from the degrees of truth of its constituents
 - Note that this is not possible for uncertain statements
- Example of truth functional interpretation of vague statements:

$$\begin{aligned}\mathcal{I}(\phi \wedge \psi) &= \min(\mathcal{I}(\phi), \mathcal{I}(\psi)) \\ \mathcal{I}(\phi \vee \psi) &= \max(\mathcal{I}(\phi), \mathcal{I}(\psi)) \\ \mathcal{I}(\neg\phi) &= 1 - \mathcal{I}(\phi)\end{aligned}$$

Uncertain Vague Statements

- Recap:

- In a **probabilistic** setting each statement is either true or false, but there is e.g. a probability distribution telling us how probable each interpretation/sentence is

$$\mathcal{I}(\phi) \in \{0, 1\}, Pr(\mathcal{I}) \in [0, 1] \text{ and } Pr(\phi) = \sum_{\mathcal{I} \models \phi} Pr(\mathcal{I}) \in [0, 1]$$

- In **vagueness** theory instead, sentences are **graded**

$$\mathcal{I}(\phi) \in [0, 1]$$

- Are there sentences combining the two orthogonal concepts of uncertainty and vagueness?
- Yes, and we use them daily !
 - E.g. “there will be heavy rain tomorrow”
- This type of sentences are called **uncertain vague sentences**
- Essentially, there is
 - **uncertainty** about the world we will have tomorrow
 - **vagueness** about the various types of rain

- Consider a propositional statement ϕ
- A model for uncertain vague sentences:
 - Define probability distribution over worlds $\mathcal{I} \in \mathcal{W}$, i.e.

$$Pr(\mathcal{I}) \in [0, 1], \sum_{\mathcal{I}} Pr(\mathcal{I}) = 1$$

- Sentences are graded: each interpretation $\mathcal{I} \in \mathcal{W}$ is truth functional and maps sentences into $[0, 1]$

$$\mathcal{I}(\phi) \in [0, 1]$$

- For a sentence ϕ , consider the **expected truth** of ϕ

$$ET(\phi) = \sum_{\mathcal{I}} Pr(\mathcal{I}) \cdot \mathcal{I}(\phi) .$$

- Note: if \mathcal{I} is bivalent (that is, $\mathcal{I}(\phi) \in \{0, 1\}$) then $ET(\phi) = Pr(\phi)$

From Fuzzy Sets to Mathematical Fuzzy Logic

Fuzzy Sets Basics

From Crisp Sets to Fuzzy Sets.

- Let X be a **universal set** of objects
- The **power set**, denoted 2^A , of a set $A \subset X$, is the set of subsets of A , i.e.,

$$2^A = \{B \mid B \subseteq A\}$$

- Often sets are defined as

$$A = \{x \mid P(x)\}$$

- $P(x)$ is a statement “ x has property P ”
- $P(x)$ is either **true** or **false** for any $x \in X$

- Examples of universe X and subsets $A, B \in 2^X$ may be

$$X = \{x \mid x \text{ is a day}\}$$

$$A = \{x \mid x \text{ is a rainy day}\}$$

$$B = \{x \mid x \text{ is a day with precipitation rate } R \geq 7.5\text{mm/h}\}$$

- In the above case: $B \subseteq A \subseteq X$
- The **membership function** of a set $A \subseteq X$:

$$\chi_A: X \rightarrow \{0, 1\}$$

where $\chi_A(x) = 1$ iff $x \in A$

- Note that for sets $A, B \in 2^X$

$$A \subseteq B \text{ iff } \forall x \in X. \chi_A(x) \leq \chi_B(x)$$

- **Complement** of a set A , i.e. $\bar{A} = X \setminus A$: $\forall x \in X$:

$$\chi_{\bar{A}}(x) = 1 - \chi_A(x)$$

- **Intersection** and **union**: $\forall x \in X$

$$\chi_{A \cap B}(x) = \min(\chi_A(x), \chi_B(x))$$

$$\chi_{A \cup B}(x) = \max(\chi_A(x), \chi_B(x))$$

- **Cartesian product** of two sets $A, B \in 2^X$

$$A \times B = \{\langle a, b \rangle \mid a \in A, b \in B\}$$

- A relation $R \subseteq X \times X$

- is **reflexive** if for all $x \in X$

$$\chi_R(x, x) = 1$$

- is **symmetric** if for all $x, y \in X$

$$\chi_R(x, y) = \chi_R(y, x)$$

- **Inverse** of R , $\chi_{R^{-1}}: X \times X \rightarrow \{0, 1\}$: $\forall x, y \in X$:

$$\chi_{R^{-1}}(y, x) = \chi_R(x, y)$$

- **Fuzzy set** A : $\chi_A: X \rightarrow [0, 1]$, or simply

$$A: X \rightarrow [0, 1]$$

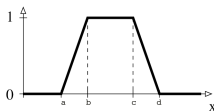
- **Fuzzy power set** over X , is denoted $\tilde{2}^X$, i.e. the set of all fuzzy sets over X
- Example: the fuzzy set

$$C = \{x \mid x \text{ is a day with heavy precipitation rate } R\}$$

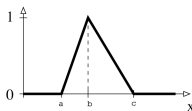
is defined via the membership function

$$\chi_C(x) = \begin{cases} 1 & \text{if } R \geq 7.5 \\ (x - 5)/2.5 & \text{if } R \in [5, 7.5) \\ 0 & \text{otherwise} \end{cases}$$

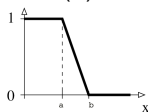
- Fuzzy membership functions may depend on the context and may be subjective
- **Shape** may be quite different
- Usually, it is sufficient to consider functions



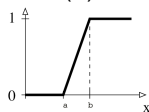
(a)



(b)



(c)



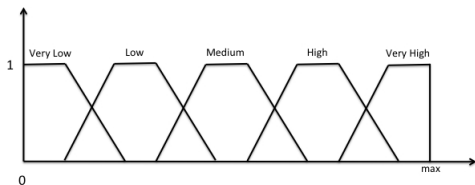
(d)

(a) Trapezoidal $trz(a, b, c, d)$; (b) Triangular $tri(a, b, c)$; (c) left-shoulder $ls(a, b)$; (d) right-shoulder $rs(a, b)$

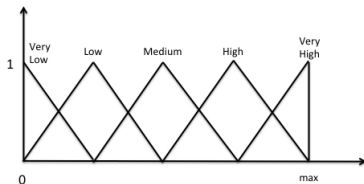
Fuzzy Sets Construction

- The usefulness of fuzzy sets depends critically on appropriate membership functions
- Methods for fuzzy membership functions construction is largely addressed in literature

- Easy and typically satisfactory method (numerical domain)
 - uniform partitioning into 5 fuzzy sets

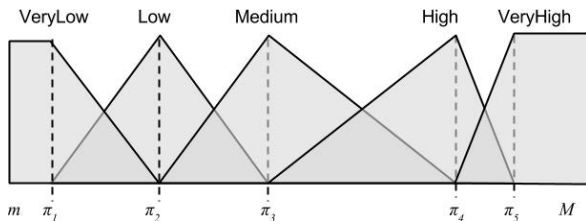


Fuzzy sets construction using trapezoidal functions



Fuzzy sets construction using triangular functions

- Another popular method is based on **clustering**
- Use **Fuzzy C-Means** to cluster data into 5 clusters
 - Fuzzy C-Means extends K-Means to accommodate graded membership
- From the clusters c_1, \dots, c_5 take the centroids π_1, \dots, π_5
- Build the fuzzy sets from the centroids



Fuzzy sets construction using clustering

Norm-Based Fuzzy Set Operations

- Standard fuzzy set operations are not the only ones
- Most notable ones are **triangular norms**
 - **t-norm** \otimes for set intersection
 - **t-conorm** \oplus (also called **s-norm**) for set union
 - **negation** \ominus for set complementation
 - **implication** \Rightarrow
 - set inclusion $A \sqsubseteq B$ is defined as

$$\inf_{x \in X} A(x) \Rightarrow B(x)$$

- \Rightarrow is often defined from \otimes as *r-implication*

$$a \Rightarrow b = \sup \{c \mid a \otimes c \leq b\} .$$

- These functions satisfy some properties that one expects to hold

Properties for t-norms and s-norms

Axiom Name	T-norm	S-norm
Taututology/Contradiction	$a \otimes 0 = 0$	$a \oplus 1 = 1$
Identity	$a \otimes 1 = a$	$a \oplus 0 = a$
Commutativity	$a \otimes b = b \otimes a$	$a \oplus b = b \oplus a$
Associativity	$(a \otimes b) \otimes c = a \otimes (b \otimes c)$	$(a \oplus b) \oplus c = a \oplus (b \oplus c)$
Monotonicity	if $b \leq c$, then $a \otimes b \leq a \otimes c$	if $b \leq c$, then $a \oplus b \leq a \oplus c$

Properties for implication and negation functions

Axiom Name	Implication Function	Negation Function
Tautology / Contradiction	$0 \Rightarrow b = 1, a \Rightarrow 1 = 1, 1 \Rightarrow 0 = 0$	$\ominus 0 = 1, \ominus 1 = 0$
Antitonicity	if $a \leq b$, then $a \Rightarrow c \geq b \Rightarrow c$	if $a \leq b$, then $\ominus a \geq \ominus b$
Monotonicity	if $b \leq c$, then $a \Rightarrow b \leq a \Rightarrow c$	

- By commutativity, \otimes and \oplus are monotone also in the first argument
- \otimes is **idempotent** if $a \otimes a = a$, for all $a \in [0, 1]$
- Negation function \ominus is **involution** iff $\ominus \ominus a = a$, for all $a \in [0, 1]$.
- Salient negation functions are:
 - Standard or Łukasiewicz negation: $\ominus_l a = 1 - a$;
 - Gödel negation: $\ominus_g a$ is 1 if $a = 0$, else is 0.
- Łukasiewicz negation is involutive, Gödel negation is not.

- Salient t-norm functions are:

Gödel t-norm: $a \otimes_g b = \min(a, b)$;

Bounded difference or Łukasiewicz t-norm:

$$a \otimes_l b = \max(0, a + b - 1);$$

Algebraic product or product t-norm: $a \otimes_p b = a \cdot b$;

Drastic product: $a \otimes_d b =$

$$\begin{cases} 0 & \text{when } (a, b) \in [0, 1[\times [0, 1[\\ \min(a, b) & \text{otherwise} \end{cases}$$

- Salient s-norm functions are:

Gödel s-norm: $a \oplus_g b = \max(a, b)$;

Bounded sum or Łukasiewicz s-norm:

$$a \oplus_l b = \min(1, a + b);$$

Algebraic sum or product s-norm: $a \oplus_p b = a + b - ab$;

Drastic sum: $a \oplus_d b =$

$$\begin{cases} 1 & \text{when } (a, b) \in]0, 1] \times]0, 1] \\ \max(a, b) & \text{otherwise} \end{cases}$$

Salient properties of norms:

- Ordering among t-norms (\otimes is any t-norm):

$$\otimes_d \leq \otimes \leq \otimes_g$$

$$\otimes_d \leq \otimes_l \leq \otimes_p \leq \otimes_g .$$

- The only idempotent t-norm is \otimes_g .
- The only t-norm satisfying $a \otimes a = 0$ for all $a \in [0, 1[$ is \otimes_d .
- Ordering among s-norms (\oplus is any s-norm):

$$\oplus_g \leq \oplus \leq \oplus_d$$

$$\oplus_g \leq \oplus_p \leq \oplus_l \leq \oplus_d .$$

- The only idempotent s-norm is \oplus_g .
- The only s-norm satisfying $a \oplus a = 1$ for all $a \in]0, 1]$ is \oplus_d .
- The **dual s-norm** of \otimes is defined as

$$a \oplus b = 1 - (1 - a) \otimes (1 - b) .$$

- **Kleene-Dienes implication**: $x \Rightarrow y = \max(1 - x, y)$ is called
- **Fuzzy modus ponens**: let $a \geq n$ and $a \Rightarrow b \geq m$
 - Under Kleene-Dienes implication, we infer that if $n > 1 - m$ then $b \geq m$
 - Under r-implication relative to a t-norm \otimes , we infer that $b \geq n \otimes m$
- **composition** of two fuzzy relations $R_1 : X \times X \rightarrow [0, 1]$ and $R_2 : X \times X \rightarrow [0, 1]$: for all $x, z \in X$
 - $(R_1 \circ R_2)(x, z) = \sup_{y \in X} R_1(x, y) \otimes R_2(y, z)$
- A fuzzy relation R is **transitive** iff for all $x, z \in X$
 $R(x, z) \geq (R \circ R)(x, z)$

Łukasiewicz, Gödel, Product logic and Standard Fuzzy logic

- One distinguishes three different sets of fuzzy set operations (called **fuzzy logics**)
 - Łukasiewicz, Gödel, and Product logic
 - Standard Fuzzy Logic (SFL) is a sublogic of Łukasiewicz
 - $\min(a, b) = a \otimes_I (a \Rightarrow_I b)$, $\max(a, b) = 1 - \min(1 - a, 1 - b)$

	Łukasiewicz Logic	Gödel Logic	Product Logic	SFL
$a \otimes b$	$\max(a + b - 1, 0)$	$\min(a, b)$	$a \cdot b$	$\min(a, b)$
$a \oplus b$	$\min(a + b, 1)$	$\max(a, b)$	$a + b - a \cdot b$	$\max(a, b)$
$a \Rightarrow b$	$\min(1 - a + b, 1)$	$\begin{cases} 1 & \text{if } a \leq b \\ b & \text{otherwise} \end{cases}$	$\min(1, b/a)$	$\max(1 - a, b)$
$\ominus a$	$1 - a$	$\begin{cases} 1 & \text{if } a = 0 \\ 0 & \text{otherwise} \end{cases}$	$\begin{cases} 1 & \text{if } a = 0 \\ 0 & \text{otherwise} \end{cases}$	$1 - a$

- Mostert–Shields theorem: any continuous t-norm can be obtained as an ordinal sum of these three

Some additional properties

Property	Łukasiewicz Logic	Gödel Logic	Product Logic	SFL
$x \otimes \ominus x = 0$	•			
$x \oplus \ominus x = 1$	•			
$x \otimes x = x$		•		•
$x \oplus x = x$		•		•
$\ominus \ominus x = x$	•			•
$x \Rightarrow y = \ominus x \oplus y$	•			•
$\ominus(x \Rightarrow y) = x \otimes \ominus y$	•			•
$\ominus(x \otimes y) = \ominus x \oplus \ominus y$	•	•	•	•
$\ominus(x \oplus y) = \ominus x \otimes \ominus y$	•	•	•	•

- **Note:** If all conditions in the upper part of a column have to be satisfied then we collapse to classical two-valued logic

Fuzzy Modifiers

- Fuzzy modifiers: interesting feature of fuzzy set theory
- A fuzzy modifier apply to fuzzy sets to change their membership function
 - Examples: **very**, **more_or_less**, and **slightly**
- A **fuzzy modifier** m represents a function

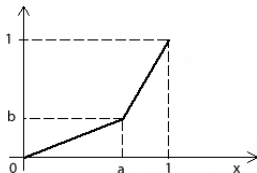
$$f_m: [0, 1] \rightarrow [0, 1]$$

Example: $f_{\text{very}}(x) = x^2$, $f_{\text{more_or_less}}(x) = \text{tri}(0, x, 1)$, $f_{\text{slightly}}(x) = \sqrt{x}$

- Modelling the fuzzy set of **very heavy rain**:

$$\begin{aligned}\chi_{\text{very heavy rain}}(x) &= f_{\text{very}}(\chi_{\text{heavy rain}}(x)) \\ &= (\chi_{\text{heavy rain}}(x))^2 \\ &= (rs(5, 7.5)(x))^2\end{aligned}$$

- A typical shape of modifiers: **linear modifiers** $lm(a, b)$



- Note: linear modifiers require one parameter c only

$$lm(a, b) = lm(c)$$

where $a = c/(c + 1)$, $b = 1/(c + 1)$

Mathematical Fuzzy Logics Basics

- OWL 2 is grounded on Mathematical Logic
- Fuzzy OWL 2 is grounded on **Mathematical Fuzzy Logic**
- A statement is no longer either true or false, but is graded
- **Truth space**: set of truth values L with some structure
- Given a statement ϕ
 - **Fuzzy Interpretation**: a function \mathcal{I} mapping ϕ into L , i.e.

$$\mathcal{I}(\phi) \in L$$

- Usually

$$L = [0, 1]$$
$$L_n = \left\{0, \frac{1}{n}, \dots, \frac{n-2}{n-1}, \dots, 1\right\} \quad (n \geq 1)$$

- **Fuzzy statement:** for $r \in [0, 1]$

$$\langle \phi, r \rangle$$

The degree of truth of ϕ is equal or greater than r

- **Examples:**
 - Fuzzy FOL: $\langle \text{RainyDay}(d), 0.75 \rangle$
 - Fuzzy LPs: $\langle \text{RainyDay}(d) \leftarrow, 0.75 \rangle$
 - Fuzzy RDFS: $\langle \langle d, \text{type}, \text{RainyDay} \rangle, 0.75 \rangle$
 - Fuzzy DLs: $\langle d:\text{RainyDay}, 0.75 \rangle$

- **Fuzzy interpretation \mathcal{I} :**

- Maps each basic statement p_i into $[0, 1]$
- Extended inductively to all statements

$$\mathcal{I}(\phi \wedge \psi) = \mathcal{I}(\phi) \otimes \mathcal{I}(\psi)$$

$$\mathcal{I}(\phi \vee \psi) = \mathcal{I}(\phi) \oplus \mathcal{I}(\psi)$$

$$\mathcal{I}(\phi \rightarrow \psi) = \mathcal{I}(\phi) \Rightarrow \mathcal{I}(\psi)$$

$$\mathcal{I}(\phi \leftrightarrow \psi) = \mathcal{I}(\phi \rightarrow \psi) \otimes \mathcal{I}(\psi \rightarrow \phi)$$

$$\mathcal{I}(\neg\phi) = \ominus \mathcal{I}(\phi)$$

$$\mathcal{I}(\exists x.\phi) = \sup_{a \in \Delta^{\mathcal{I}}} \mathcal{I}_x^a(\phi)$$

$$\mathcal{I}(\forall x.\phi) = \inf_{a \in \Delta^{\mathcal{I}}} \mathcal{I}_x^a(\phi),$$

where

- $\Delta^{\mathcal{I}}$ is the domain of \mathcal{I}
- \otimes , \oplus , \Rightarrow , and \ominus are the t-norms, t-conorms, implication functions, a negation functions
- The function \mathcal{I}_x^a is as \mathcal{I} except that x is interpreted as a

Example

- In Lukasiewicz logic:

$$\varphi = \text{Cold} \wedge \text{Cloudy}$$

\mathcal{I}	Cold	Cloudy	$\mathcal{I}(\varphi)$
\mathcal{I}_1	0	0.1	$\max(0, 0 + 0.1 - 1) = 0.0$
\mathcal{I}_2	0.3	0.4	$\max(0, 0.3 + 0.4 - 1) = 0.0$
\mathcal{I}_3	0.7	0.8	$\max(0, 0.7 + 0.9 - 1) = 0.6$
\mathcal{I}_4	1	1	$\max(0, 1 + 1 - 1) = 1.0$
\vdots	\vdots	\vdots	\vdots

- Note:** given m propositional letters
 - Fuzzy interpretations over $L = [0, 1]$ are **not recursively enumerable**
 - There are n^m fuzzy interpretations over L_n

- One may also consider the following abbreviations:

$$\phi \wedge_g \psi \stackrel{\text{def}}{=} \phi \wedge (\phi \rightarrow \psi)$$

$$\phi \vee_g \psi \stackrel{\text{def}}{=} (\phi \rightarrow \psi) \rightarrow \phi \wedge_g (\psi \rightarrow \phi) \rightarrow \psi$$

$$\neg_{\otimes} \phi \stackrel{\text{def}}{=} \phi \rightarrow 0$$

$$\langle \phi \leq r \rangle \stackrel{\text{def}}{=} \langle \neg_I \phi, 1 - r \rangle$$

- In case \Rightarrow is the r-implication based on \otimes , then
 - \wedge_g is Gödel t-norm
 - \vee_g is Gödel s-norm
 - \neg_{\otimes} is interpreted as the negation function related to \otimes

- \mathcal{I} **satisfies** $\langle \phi, r \rangle$, or \mathcal{I} is a **model** of $\langle \phi, r \rangle$

$$\mathcal{I} \models \langle \phi, r \rangle \text{ iff } \mathcal{I}(\phi) \geq r$$

- \mathcal{I} is a **model** of ϕ if $\mathcal{I}(\phi) = 1$
- **Fuzzy knowledge base** \mathcal{K} : finite set of fuzzy statements
- \mathcal{I} **satisfies** (is a **model** of) \mathcal{K} : $\mathcal{I} \models \mathcal{K}$ iff it satisfies each element in it
- **Best entailment degree** of ϕ w.r.t. \mathcal{K} :

$$\text{bed}(\mathcal{K}, \phi) = \sup \{r \mid \mathcal{K} \models \langle \phi, r \rangle\}$$

- **Best satisfiability degree** of ϕ w.r.t. \mathcal{K} :

$$\text{bsd}(\mathcal{K}, \phi) = \sup_{\mathcal{I}} \{\mathcal{I}(\phi) \mid \mathcal{I} \models \mathcal{K}\}$$

Proposition (Fuzzy Modus Ponens)

For *r*-implication \rightarrow , for $r, s \in [0, 1]$:

$$\langle \phi, r \rangle, \langle \phi \rightarrow \psi, s \rangle \models \langle \psi, r \otimes s \rangle$$

Proposition

Salient equivalences:

$$\begin{aligned}\neg\neg\phi &\equiv \phi \quad (\perp, SFL) \\ \phi \wedge \phi &\equiv \phi \quad (G, SFL) \\ \neg(\phi \wedge \neg\phi) &\equiv 1 \quad (\perp, G, \Pi) \\ \phi \vee \neg\phi &\equiv 1 \quad (\perp) \\ \forall x.\phi &\equiv \neg\exists x.\neg\phi \quad (\perp, SFL)\end{aligned}$$

Proposition

Salient equivalences:

$$\mathcal{L} + G \equiv \textit{Boolean Logic}$$

$$\mathcal{L} + \Pi \equiv \textit{Boolean Logic}$$

$$G + \Pi \equiv \textit{Boolean Logic}$$

Proposition (BED)

$bed(\mathcal{K}, \phi) = \min x$. *such that $\mathcal{K} \cup \{\langle \phi \leq x \rangle\}$ satisfiable.*

Proposition (BSD)

$bsd(\mathcal{K}, \phi) = \max x$. *such that $\mathcal{K} \cup \{\langle \phi, x \rangle\}$ satisfiable.*

On Witnessed Models

- Witnessed interpretation \mathcal{I} :

$$\mathcal{I}(\exists x.\phi) = \mathcal{I}_x^a(\phi), \text{ for some } a \in \Delta^{\mathcal{I}} \quad (1)$$

$$\mathcal{I}(\forall x.\phi) = \mathcal{I}_x^a(\phi), \text{ for some } a \in \Delta^{\mathcal{I}} \quad (2)$$

- The supremum (resp. infimum) are attained at some point
- Classical interpretations are witnessed
- Fuzzy interpretations **may not be witnessed**
- E.g., \mathcal{I} is not witnessed as Eq. (1) not satisfied:

$$\begin{aligned} \Delta^{\mathcal{I}} &= \mathbb{N} \\ \mathcal{I}_x^n(A(x)) &= 1 - 1/n < 1, \text{ for all } n \\ \mathcal{I}(\exists x.A(x)) &= \sup_n \mathcal{I}_x^n(A(x)) \\ &= \sup_n 1 - 1/n = 1 \end{aligned}$$

Proposition (Witnessed model property)

In Łukasiewicz logic and SFL over $L = [0, 1]$, or for all cases in which the truth space L is finite, a fuzzy KB has a witnessed fuzzy model iff it has a fuzzy model.

- Not true for Gödel and product logic over $L = [0, 1]$
 - $\neg\forall x p(x) \wedge \neg\exists x \neg p(x)$ has no classical model
 - In Gödel logic it has no finite model, but has an **infinite** model: for integer $n \geq 1$, let \mathcal{I} such that $\mathcal{I}(p(n)) = 1/n$

$$\begin{aligned}\mathcal{I}(\forall x p(x)) &= \inf_n 1/n = 0 \\ \mathcal{I}(\exists x \neg p(x)) &= \sup_n \neg 1/n = \sup_n 0 = 0\end{aligned}$$

- IMHO: non-witnessed models make little sense in KR
- We will always assume that interpretations are witnessed

Fuzzy Propositional Logic: Reasoning

- We need to distinguish if truth space is $L = [0, 1]$ or $L_n = \{0, \frac{1}{n}, \dots, \frac{n-2}{n-1}, \dots, 1\}$
- Case L_n easier: given m propositional letters, there are m^n possible interpretations
- We may use
 - Operational Research
 - Analytic Tableaux, Non-Deterministic Analytic Tableaux
 - Reduction into Classical Propositional Logic

Operational Research: Case Łukasiewicz Logic & SFL

- Basic idea: translate formulae into equational constraints about truth degrees
- For a formula ϕ consider a variable x_ϕ
 - Intuition: x_ϕ will hold the degree of truth of statement ϕ
 - Example: constraints under Łukasiewicz for $\langle \neg\phi, 0.6 \rangle$

$$x_{\neg\phi} \in [0, 1]$$

$$x_\phi \in [0, 1]$$

$$x_{\neg\phi} = 1 - x_\phi$$

- We may use **Mixed Integer Linear Programming** for the encodings of constraints

For Łukasiewicz:

- $x_1 \otimes_l x_2 = z$
 $\mapsto \{x_1 + x_2 - 1 \leq z, x_1 + x_2 - 1 \geq z - y, z \leq 1 - y, y \in \{0, 1\}\}$,
where y is a new variable.
- $x_1 \oplus_l x_2 = z \mapsto \{x_1 + x_2 \leq z + y, y \leq z, x_1 + x_2 \geq z, y \in \{0, 1\}\}$,
where y is a new variable.
- $x_1 \Rightarrow_l x_2 = z \mapsto \{(1 - x_1) \oplus_l x_2 = z\}$.

For SFL:

- $x_1 \otimes_g x_2 = z$
 $\mapsto \{z \leq x_1, z \leq x_2, x_1 \leq z + y, x_2 \leq z + (1 - y), y \in \{0, 1\}\}$,
where y is a new variable.
- $x_1 \oplus_g x_2 = z$
 $\mapsto \{z \geq x_1, z \geq x_2, x_1 + y \geq z, x_2 + (1 - y) \geq z, y \in \{0, 1\}\}$,
where y is a new variable.
- $x_1 \Rightarrow_{kd} x_2 = z \mapsto (1 - x_1) \oplus_g x_2 = z$.

- **Negation Normal Form, $nnf(\phi)$**

$$\neg \perp = \top$$

$$\neg \top = \perp$$

$$\neg \neg \phi \mapsto \phi$$

$$\neg(\phi \wedge \psi) \mapsto \neg\phi \vee \neg\psi$$

$$\neg(\phi \vee \psi) \mapsto \neg\phi \wedge \neg\psi$$

$$\neg(\phi \rightarrow \psi) \mapsto \phi \wedge \neg\psi .$$

- 1 Transform \mathcal{K} into NNF
- 2 Initialize the fuzzy theory $\mathcal{T}_{\mathcal{K}}$ and the initial set of constraints $\mathcal{C}_{\mathcal{K}}$ by

$$\begin{aligned}\mathcal{T}_{\mathcal{K}} &= \{\phi \mid \langle \phi, n \rangle \in \mathcal{K}\} \\ \mathcal{C}_{\mathcal{K}} &= \{x_{\psi} \geq n \mid \langle \phi, n \rangle \in \mathcal{K}\}\end{aligned}$$

- 3 Apply the following inference rules until no more rules can be applied

- (*var*). For variable x_{ϕ} occurring in $\mathcal{C}_{\mathcal{K}}$ add $x_{\phi} \in [0, 1]$ to $\mathcal{C}_{\mathcal{K}}$
- (*v̄ar*). For variable $x_{\neg\phi}$ occurring in $\mathcal{C}_{\mathcal{K}}$ add $x_{\phi} = 1 - x_{\neg\phi}$ to $\mathcal{C}_{\mathcal{K}}$
- (\perp). If $\perp \in \mathcal{T}_{\mathcal{K}}$ then $\mathcal{C}_{\mathcal{K}} := \mathcal{C}_{\mathcal{K}} \cup \{x_{\perp} = 0\}$
- (\top). If $\top \in \mathcal{T}_{\mathcal{K}}$ then $\mathcal{C}_{\mathcal{K}} := \mathcal{C}_{\mathcal{K}} \cup \{x_{\top} = 1\}$
- (\wedge). If $\phi \wedge \psi \in \mathcal{T}_{\mathcal{K}}$, then
 - 1 add ϕ and ψ to $\mathcal{T}_{\mathcal{K}}$
 - 2 $\mathcal{C}_{\mathcal{K}} := \mathcal{C}_{\mathcal{K}} \cup \{x_{\phi} \otimes x_{\psi} = x_{\phi \wedge \psi}\}$
- (\vee). If $\phi \vee \psi \in \mathcal{T}_{\mathcal{K}}$, then
 - 1 add ϕ and ψ to $\mathcal{T}_{\mathcal{K}}$
 - 2 $\mathcal{C}_{\mathcal{K}} := \mathcal{C}_{\mathcal{K}} \cup \{x_{\phi} \oplus x_{\psi} = x_{\phi \vee \psi}\}$
- (\rightarrow). If $\phi \rightarrow \psi \in \mathcal{T}_{\mathcal{K}}$, then
 - 1 add $\text{nnf}(\neg\phi)$ and ψ to $\mathcal{T}_{\mathcal{K}}$
 - 2 $\mathcal{C}_{\mathcal{K}} := \mathcal{C}_{\mathcal{K}} \cup \{(1 - x_{\text{nnf}(\neg\phi)}) \Rightarrow x_{\psi} = x_{\phi \rightarrow \psi}\}$

***sat*(\mathcal{K}):** \mathcal{K} is satisfiable iff the final set of constraints $\mathcal{C}_{\mathcal{K}}$ has a solution

- bed*(\mathcal{K}, ϕ):**
- Add $\neg\phi$ to $\mathcal{T}_{\mathcal{K}}$
 - Add $x_{\neg\phi} \geq 1 - x, x \in [0, 1]$ to $\mathcal{C}_{\mathcal{K}}$, x new
 - Compute final set of constraints $\mathcal{C}_{\mathcal{K}}$
 - Then, solve the optimisation problem

$bed(\mathcal{K}, \phi) = \min x.$ such that $\mathcal{C}_{\mathcal{K}}$ has a solution

- bsd*(\mathcal{K}, ϕ):**
- Add ϕ to $\mathcal{T}_{\mathcal{K}}$
 - Add $x_{\phi} \geq x, x \in [0, 1]$ to $\mathcal{C}_{\mathcal{K}}$, x new
 - Compute final set of constraints $\mathcal{C}_{\mathcal{K}}$
 - Then, solve the optimisation problem

$bsd(\mathcal{K}, \phi) = \max x.$ such that $\mathcal{C}_{\mathcal{K}}$ has a solution

Analytical Fuzzy Tableau: Case SFL

- Main property the method is based on:
 - if \mathcal{I} is model of $\langle \phi \wedge \psi, n \rangle$ then \mathcal{I} is a model of both $\langle \phi, n \rangle$ and $\langle \psi, n \rangle$;
 - if \mathcal{I} is model of $\langle \phi \vee \psi, n \rangle$ then \mathcal{I} is a model of either $\langle \phi, n \rangle$ or $\langle \psi, n \rangle$.
 - \mathcal{I} cannot be a model of both $\langle p, n \rangle$ and $\langle \neg p, m \rangle$ if $n > 1 - m$.
- A **clash** is either
 - a fuzzy statement $\langle \perp, n \rangle$ with $n > 0$; or
 - a pair of fuzzy statements $\langle p, n \rangle$ and $\langle \neg p, m \rangle$ with $n > 1 - m$
- **Clash-free**: does not contain a clash

- 1 Transform \mathcal{K} into NNF
- 2 Initialize the completion $\mathcal{S}_{\mathcal{K}} = \mathcal{K}$
- 3 Apply the following inference rules to $\mathcal{S}_{\mathcal{K}}$ until no more rules can be applied
- 4 We call a set of fuzzy statements $\mathcal{S}_{\mathcal{K}}$ **complete** iff none of the rules below can be applied to $\mathcal{S}_{\mathcal{K}}$
- 5 Note that rule (\vee) is non-deterministic
 - (\wedge). If $\langle \phi \wedge \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$ and $\{\langle \phi, n \rangle, \langle \psi, n \rangle\} \not\subseteq \mathcal{S}_{\mathcal{K}}$, then add both $\langle \phi, n \rangle$ and $\langle \psi, n \rangle$ to $\mathcal{S}_{\mathcal{K}}$
 - (\vee). If $\langle \phi \vee \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$ and $\{\langle \phi, n \rangle, \langle \psi, n \rangle\} \cap \mathcal{S}_{\mathcal{K}} = \emptyset$, then add either $\langle \phi, n \rangle$ or $\langle \psi, n \rangle$ to $\mathcal{S}_{\mathcal{K}}$
 - (\rightarrow). If $\langle \phi \rightarrow \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$ and $\langle nnf(\neg\phi) \vee \psi, n \rangle \notin \mathcal{S}_{\mathcal{K}}$, then add $\langle nnf(\neg\phi) \vee \psi, n \rangle$ to $\mathcal{S}_{\mathcal{K}}$

***sat*(\mathcal{K})**: \mathcal{K} is satisfiable iff we find a complete and clash-free completion $\mathcal{S}_{\mathcal{K}}$ of \mathcal{K}

- For BED and BSD we need some more work
- Given \mathcal{K} , define

$$\begin{aligned} N^{\mathcal{K}} &= \{0, 0.5, 1\} \cup \{n \mid \langle \phi, n \rangle \in \mathcal{K}\} \\ \bar{N}^{\mathcal{K}} &= N^{\mathcal{K}} \cup \{1 - n \mid n \in N^{\mathcal{K}}\} \\ \epsilon &= \min\{d/2 \mid n, m \in \bar{N}^{\mathcal{K}}, n \neq m, d = |n - m|\} \end{aligned}$$

Proposition

Under SFL, given \mathcal{K} , then for $n > 0$

$\mathcal{K} \models \langle \phi, n \rangle$ iff $\mathcal{K} \cup \{\langle \neg\phi, 1 - n + \epsilon \rangle\}$ is not satisfiable .

Moreover, \mathcal{K} is satisfiable iff it has a model over $\bar{N}^{\mathcal{K}}$.

$bed(\mathcal{K}, \phi)$: Find greatest $n \in \bar{N}^{\mathcal{K}}$ such that $\mathcal{K} \models \langle \phi, n \rangle$
 $bsd(\mathcal{K}, \phi)$: Find greatest $n \in \bar{N}^{\mathcal{K}}$ such that $\mathcal{K} \cup \{\langle \phi, n \rangle\}$
satisfiable

Non Deterministic Analytic Fuzzy Tableau

- Works for finitely-valued fuzzy propositional logic over L_n
- Works also for SFL (as in place of $[0, 1]$, we may use \bar{N}^K)
- Basic idea is as for fuzzy tableau, but now we **guess** the truth degrees
 - (\wedge). If $\langle \phi \wedge \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$, $n_1, n_2 \in L_n$ such that $n_1 \otimes n_2 = n$ and $\{\langle \phi, n_1 \rangle, \langle \psi, n_2 \rangle\} \not\subseteq \mathcal{S}_{\mathcal{K}}$, then add both $\langle \phi, n_1 \rangle$ and $\langle \psi, n_2 \rangle$ to $\mathcal{S}_{\mathcal{K}}$
 - (\vee). If $\langle \phi \vee \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$, $n_1, n_2 \in L_n$ such that $n_1 \oplus n_2 = n$ and $\{\langle \phi, n_1 \rangle, \langle \psi, n_2 \rangle\} \not\subseteq \mathcal{S}_{\mathcal{K}}$, then add both $\langle \phi, n_1 \rangle$ and $\langle \psi, n_2 \rangle$ to $\mathcal{S}_{\mathcal{K}}$
 - (\rightarrow). If $\langle \phi \rightarrow \psi, n \rangle \in \mathcal{S}_{\mathcal{K}}$, $n_1, n_2 \in L_n$ such that $n_1 \Rightarrow n_2 = n$ and $\{\langle \phi, n_1 \rangle, \langle \psi, n_2 \rangle\} \not\subseteq \mathcal{S}_{\mathcal{K}}$, then add both $\langle \phi, n_1 \rangle$ and $\langle \psi, n_2 \rangle$ to $\mathcal{S}_{\mathcal{K}}$
- A **clash** is either
 - a fuzzy statement $\langle \perp, n \rangle$ with $n > 0$; or
 - a pair of fuzzy statements $\langle p, n \rangle$ and $\langle \neg p, m \rangle$ such that

$$x_p \geq n, \quad \ominus x_p \geq m, \quad x_p \in L_n$$

has no solution

Reduction to Classical Propositional Logic: Case SFL over $[0, 1]$

- Given \mathcal{K} , we know that we can use

$$L_n = \bar{N}^{\mathcal{K}} = \{\gamma_1, \dots, \gamma_n\}$$

with $\gamma_i < \gamma_{i+1}$, $1 \leq i \leq n-1$

- Basic idea: use atom $A_{\geq r}$ to represent

The truth degree of A has to be equal or greater than r

- Similarly for $A_{>r}$, $A_{\leq r}$ and $A_{<r}$

- To start with, build $Crisp_{L_n}$
 - For all atoms A , for all $1 \leq i \leq n-1, 2 \leq j \leq n-1$

$$A_{\geq \gamma_{i+1}} \rightarrow A_{> \gamma_i}$$

$$A_{> \gamma_j} \rightarrow A_{\geq \gamma_j}$$

- Build $Crisp_{\mathcal{K}}$:

$$Crisp_{\mathcal{K}} = \{\rho(\phi, n) \mid \langle \phi, n \rangle \in \mathcal{K}\} \cup Crisp_{L_n},$$

x	y	$\rho(x, y)$
\top	c	\top
\perp	0	\top
\perp	c	\perp if $c > 0$
A	c	$A_{> c}$
$\neg A$	c	$\neg A_{> 1-c}$
$\phi \wedge \psi$	c	$\rho(\phi, c) \wedge \rho(\psi, c)$
$\phi \vee \psi$	c	$\rho(\phi, c) \vee \rho(\psi, c)$

Proposition

Given \mathcal{K} under SFL over L_n , then $\mathcal{K} \models \langle \phi, c \rangle$ iff $\mathcal{K} \cup \{ \langle \neg \phi, 1 - c^- \rangle \}$ is not satisfiable, where c^- is the next smaller value than c in L_n

sat(\mathcal{K}): \mathcal{K} is satisfiable iff $\text{Crisp}_{\mathcal{K}}$ satisfiable

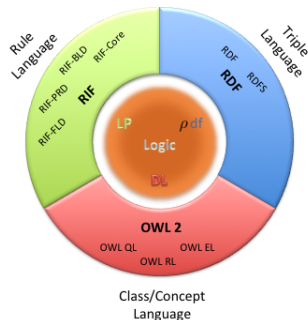
bed(\mathcal{K}, ϕ): Find greatest $c \in L_n$ such that $\mathcal{K} \models \langle \phi, c \rangle$

bsd(\mathcal{K}, ϕ): Find greatest $c \in L_n$ such that $\mathcal{K} \cup \{ \langle \phi, c \rangle \}$ satisfiable

Fuzzy Description Logics and OWL 2

The Semantic Web Family of Languages

- Wide variety of languages
 - **RDFS**: *Triple language*, -Resource Description Framework
 - The logical counterpart is ρdf
 - **RIF**: *Rule language*, -Rule Interchange Format,
 - Relate to the *Logic Programming* (LP) paradigm
 - **OWL 2**: *Conceptual language*, -Ontology Web Language
 - Relate to **Description Logics** (DLs)



OWL 2 Profiles

- OWL 2 EL
 - Useful for large size of properties and/or classes
 - Basic reasoning problems solved in polynomial time
 - The EL acronym refers to the \mathcal{EL} family of DLs
- OWL 2 QL
 - Useful for very large volumes of instance data
 - Conjunctive query answering via query rewriting and SQL
 - OWL 2 QL relates to the DL family *DL-Lite*
- OWL 2 RL
 - Useful for scalable reasoning without sacrificing too much expressive power
 - OWL 2 RL maps to Datalog
 - Computational complexity: same as for Datalog, polynomial in size of the data, EXPTIME w.r.t. size of knowledge base

Description Logics (DLs)

- **Concept/Class**: names are equivalent to unary predicates
 - In general, concepts equiv to formulae with one free variable
- **Role or attribute**: names are equivalent to binary predicates
 - In general, roles equiv to formulae with two free variables
- **Taxonomy**: Concept and role hierarchies can be expressed
- **Individual**: names are equivalent to constants
- **Operators**: restricted so that
 - Language is decidable and, if possible, of low complexity
 - No need for explicit use of variables
 - Restricted form of \exists and \forall
 - Features such as counting can be succinctly expressed

- **Basic ingredients:** descriptions of classes, properties, and their instances, such as
 - $a:C$, meaning that individual a is an instance of concept/class C

$a:\text{Person} \sqcap \forall \text{hasChild.Femal}$

- $(a, b):R$, meaning that the pair of individuals $\langle a, b \rangle$ is an instance of the property/role R

$(\text{tom}, \text{mary}):\text{hasChild}$

- $C \sqsubseteq D$, meaning that the class C is a subclass of class D

$\text{Person} \sqsubseteq \forall \text{hasChild.Person}$

The DL Family

- A given DL is defined by set of concept and role forming operators
- Basic language: \mathcal{ALC} (*A*ttributive *L*anguage with *C*omplement)

Syntax	Semantics	Example
$C, D \rightarrow$	$\top(x)$	
	$\perp(x)$	
	A	<i>Human</i>
	$C \sqcap D$	<i>Human</i> \sqcap <i>Male</i>
	$C \sqcup D$	<i>Nice</i> \sqcup <i>Rich</i>
	$\neg C$	\neg <i>Meat</i>
	$\exists R.C$	\exists <i>has_child.Blond</i>
	$\forall R.C$	\forall <i>has_child.Human</i>
$C \sqsubseteq D$	$\forall x.C(x) \Rightarrow D(x)$	<i>Happy_Father</i> \sqsubseteq <i>Man</i> \sqcap \exists <i>has_child.Female</i>
$a:C$	$C(a)$	<i>John:Happy_Father</i>

DL Semantics

- Semantics is given in terms of an **interpretation** $\mathcal{I} = (\Delta^{\mathcal{I}}, \cdot^{\mathcal{I}})$, where
 - $\Delta^{\mathcal{I}}$ is the **domain** (a non-empty set)
 - $\cdot^{\mathcal{I}}$ is an **interpretation function** that maps:
 - **Concept** (class) name A into a subset $A^{\mathcal{I}}$ of $\Delta^{\mathcal{I}}$
 - **Role** (property) name R into a subset $R^{\mathcal{I}}$ of $\Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$
 - **Individual** name a into an element of $\Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$ s.t. $a^{\mathcal{I}} \neq b^{\mathcal{I}}$ if $a \neq b$ (UNA)
 - Interpretation function $\cdot^{\mathcal{I}}$ is extended to concept expressions:

$$\begin{aligned}
 \top^{\mathcal{I}} &= \Delta^{\mathcal{I}} \\
 \perp^{\mathcal{I}} &= \emptyset \\
 (C_1 \sqcap C_2)^{\mathcal{I}} &= C_1^{\mathcal{I}} \cap C_2^{\mathcal{I}} \\
 (C_1 \sqcup C_2)^{\mathcal{I}} &= C_1^{\mathcal{I}} \cup C_2^{\mathcal{I}} \\
 (\neg C)^{\mathcal{I}} &= \Delta^{\mathcal{I}} \setminus C^{\mathcal{I}} \\
 (\exists R.C)^{\mathcal{I}} &= \{x \in \Delta^{\mathcal{I}} \mid \exists y. \langle x, y \rangle \in R^{\mathcal{I}} \wedge y \in C^{\mathcal{I}}\} \\
 (\forall R.C)^{\mathcal{I}} &= \{x \in \Delta^{\mathcal{I}} \mid \forall y. \langle x, y \rangle \in R^{\mathcal{I}} \Rightarrow y \in C^{\mathcal{I}}\}
 \end{aligned}$$

- Finally, we say that
 - \mathcal{I} is a model of $C \sqsubseteq D$, written $\mathcal{I} \models C \sqsubseteq D$, iff $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$
 - \mathcal{I} is a model of $a:C$, written $\mathcal{I} \models a:C$, iff $a^{\mathcal{I}} \in C^{\mathcal{I}}$
 - \mathcal{I} is a model of $(a, b):R$, written $\mathcal{I} \models (a, b):R$, iff $\langle a^{\mathcal{I}}, b^{\mathcal{I}} \rangle \in R^{\mathcal{I}}$

Note on DL Naming

- \mathcal{AL} : $C, D \rightarrow \top \mid \perp \mid A \mid C \sqcap D \mid \neg A \mid \exists R.C \mid \forall R.C$
- \mathcal{C} : Concept negation, $\neg C$. Thus, $\mathcal{ALC} = \mathcal{AL} + \mathcal{C}$
 - \mathcal{S} : Used for \mathcal{ALC} with transitive roles \mathcal{R}_+
 - \mathcal{U} : Concept disjunction, $C_1 \sqcup C_2$
 - \mathcal{E} : Existential quantification, $\exists R.C$
 - \mathcal{H} : Role inclusion axioms, $R_1 \sqsubseteq R_2$, e.g. *is_component_of* \sqsubseteq *is_part_of*
 - \mathcal{N} : Number restrictions, $(\geq n R)$ and $(\leq n R)$, e.g. $(\geq 3 \text{ has_Child})$ (has at least 3 children)
 - \mathcal{Q} : Qualified number restrictions, $(\geq n R.C)$ and $(\leq n R.C)$, e.g. $(\leq 2 \text{ has_Child.Adult})$ (has at most 2 adult children)
 - \mathcal{O} : Nominals (singleton class), $\{a\}$, e.g. $\exists \text{has_child.}\{mary\}$.
Note: $a:C$ equiv to $\{a\} \sqsubseteq C$ and $(a, b):R$ equiv to $\{a\} \sqsubseteq \exists R.\{b\}$
 - \mathcal{I} : Inverse role, R^- , e.g. *isPartOf* = *hasPart*⁻
 - \mathcal{F} : Functional role, f , e.g. *functional(hasAge)*
 - \mathcal{R}_+ : transitive role, e.g. *transitive(isPartOf)*

For instance,

$$\begin{aligned}
 SHIF &= S + H + I + F = \mathcal{ALCR}_+HIF && \text{OWL-Lite} \\
 SHOIN &= S + H + O + I + N = \mathcal{ALCR}_+HOIN && \text{OWL-DL} \\
 SROIQ &= S + R + O + I + Q = \mathcal{ALCR}_+ROIN && \text{OWL 2}
 \end{aligned}$$

Semantics of Additional Constructs

- \mathcal{H} : Role inclusion axioms, $\mathcal{I} \models R_1 \sqsubseteq R_2$ iff $R_1^{\mathcal{I}} \subseteq R_2^{\mathcal{I}}$
- \mathcal{N} : Number restrictions,
 $(\geq n R)^{\mathcal{I}} = \{x \in \Delta^{\mathcal{I}} : |\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\}| \geq n\}$,
 $(\leq n R)^{\mathcal{I}} = \{x \in \Delta^{\mathcal{I}} : |\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\}| \leq n\}$
- \mathcal{Q} : Qualified number restrictions,
 $(\geq n R.C)^{\mathcal{I}} = \{x \in \Delta^{\mathcal{I}} : |\{y \mid \langle x, y \rangle \in R^{\mathcal{I}} \wedge y \in C^{\mathcal{I}}\}| \geq n\}$,
 $(\leq n R.C)^{\mathcal{I}} = \{x \in \Delta^{\mathcal{I}} : |\{y \mid \langle x, y \rangle \in R^{\mathcal{I}} \wedge y \in C^{\mathcal{I}}\}| \leq n\}$
- \mathcal{O} : Nominals (singleton class), $\{a\}^{\mathcal{I}} = \{a^{\mathcal{I}}\}$
- \mathcal{I} : Inverse role, $(R^-)^{\mathcal{I}} = \{\langle x, y \rangle \mid \langle y, x \rangle \in R^{\mathcal{I}}\}$
- \mathcal{F} : Functional role, $\mathcal{I} \models \text{fun}(f)$ iff $\forall x \forall y \forall z$ if $\langle x, y \rangle \in f^{\mathcal{I}}$ and $\langle x, z \rangle \in f^{\mathcal{I}}$ then $y = z$
- \mathcal{R}_+ : transitive role,
 $(R_+)^{\mathcal{I}} = \{\langle x, y \rangle \mid \exists z \text{ such that } \langle x, z \rangle \in R^{\mathcal{I}} \wedge \langle z, y \rangle \in R^{\mathcal{I}}\}$

Basics on Concrete Domains

- **Concrete domains:** reals, integers, strings, ...

(tim, 14):hasAge

(sf, "SoftComputing"):hasAcronym

(source1, "ComputerScience"):isAbout

(service2, "InformationRetrievalTool"):Matches

Minor = Person \sqcap \exists hasAge. ≤ 18

- Semantics: a clean separation between "object" classes and concrete domains
 - $D = \langle \Delta_D, \Phi_D \rangle$
 - Δ_D is an interpretation domain
 - Φ_D is the set of concrete domain predicates d with a predefined arity n and **fixed** interpretation $d^D \subseteq \Delta_D^n$
 - Concrete properties: $R^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}} \times \Delta_D$
- Notation: (D) . E.g., $\mathcal{ALC}(D)$ is \mathcal{ALC} + concrete domains

DL Knowledge Base

- A DL **Knowledge Base** is a pair $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$, where
 - \mathcal{T} is a **TBox**
 - containing general inclusion axioms of the form $C \sqsubseteq D$,
 - concept definitions of the form $A = C$
 - primitive concept definitions of the form $A \sqsubseteq C$
 - role inclusions of the form $R \sqsubseteq P$
 - role equivalence of the form $R = P$
 - \mathcal{A} is a **ABox**
 - containing assertions of the form $a:C$
 - containing assertions of the form $(a, b):R$
- An interpretation \mathcal{I} is a model of \mathcal{K} , written $\mathcal{I} \models \mathcal{K}$ iff $\mathcal{I} \models \mathcal{T}$ and $\mathcal{I} \models \mathcal{A}$, where
 - $\mathcal{I} \models \mathcal{T}$ (\mathcal{I} is a model of \mathcal{T}) iff \mathcal{I} is a model of each element in \mathcal{T}
 - $\mathcal{I} \models \mathcal{A}$ (\mathcal{I} is a model of \mathcal{A}) iff \mathcal{I} is a model of each element in \mathcal{A}

Basic Inference Problems (Formally)

Consistency: Check if knowledge is meaningful

- Is \mathcal{K} satisfiability? \mapsto Is there some model \mathcal{I} of \mathcal{K} ?
- Is C satisfiability? $\mapsto C^{\mathcal{I}} \neq \emptyset$ for some some model \mathcal{I} of \mathcal{K} ?

Subsumption: structure knowledge, compute taxonomy

- $\mathcal{K} \models C \sqsubseteq D$? \mapsto Is it true that $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$ for all models \mathcal{I} of \mathcal{K} ?

Equivalence: check if two classes denote same set of instances

- $\mathcal{K} \models C = D$? \mapsto Is it true that $C^{\mathcal{I}} = D^{\mathcal{I}}$ for all models \mathcal{I} of \mathcal{K} ?

Instantiation: check if individual a instance of class C

- $\mathcal{K} \models a:C$? \mapsto Is it true that $a^{\mathcal{I}} \in C^{\mathcal{I}}$ for all models \mathcal{I} of \mathcal{K} ?

Retrieval: retrieve set of individuals that instantiate C

- Compute the set $\{a \mid \mathcal{K} \models a:C\}$

Reduction to Satisfiability

Problems are all **reducible** to KB satisfiability

Subsumption: $\mathcal{K} \models C \sqsubseteq D$ iff $\langle \mathcal{T}, \mathcal{A} \cup \{a:C \sqcap \neg D\} \rangle$ not satisfiable, where a is a new individual

Equivalence: $\mathcal{K} \models C = D$ iff $\mathcal{K} \models C \sqsubseteq D$ and $\mathcal{K} \models D \sqsubseteq C$

Instantiation: $\mathcal{K} \models a:C$ iff $\langle \mathcal{T}, \mathcal{A} \cup \{a:\neg C\} \rangle$ not satisfiable

Retrieval: The computation of the set $\{a \mid \mathcal{K} \models a:C\}$ is reducible to the instance checking problem

Reasoning in DLs: Basics

- OWL 2: **tableaux** based algorithms
- OWL 2 EL: **structural** based algorithms
- OWL 2 QL: **query rewriting** based algorithms
- OWL 2 RL: **logic programming** based algorithms

Tableaux: Basics

- Tableaux algorithm deciding satisfiability
- Try to build a **tree-like model** \mathcal{I} of the KB
- Decompose concepts C syntactically
 - Apply tableau **expansion rules**
 - Infer constraints on elements of model
- Tableau rules correspond to constructors in logic (\sqcap, \sqcup, \dots)
 - Some rules are **nondeterministic** (e.g., \sqcup, \leq)
 - In practice, this means **search**
- Stop when no more rules applicable or **clash** occurs
 - Clash is an obvious contradiction, e.g., $A(x), \neg A(x)$
- Cycle check (**blocking**) may be needed for termination

Negation Normal Form (NNF)

- We have to transform concepts into **Negation Normal Form**: push negation inside using de Morgan' laws

$$\begin{aligned}\neg \top &\mapsto \perp \\ \neg \perp &\mapsto \top \\ \neg \neg C &\mapsto C \\ \neg(C_1 \sqcap C_2) &\mapsto \neg C_1 \sqcup \neg C_2 \\ \neg(C_1 \sqcup C_2) &\mapsto \neg C_1 \sqcap \neg C_2\end{aligned}$$

and

$$\begin{aligned}\neg(\exists R.C) &\mapsto \forall R.\neg C \\ \neg(\forall R.C) &\mapsto \exists R.\neg C\end{aligned}$$

Completion-Forest

- This is a forest of trees, where
 - each node x is labelled with a set $\mathcal{L}(x)$ of concepts
 - each edge $\langle x, y \rangle$ is labelled with $\mathcal{L}(\langle x, y \rangle) = \{R\}$ for some role R (edges correspond to relationships between pairs of individuals)
- The forest is initialized with
 - a root node a , labelled $\mathcal{L}(a) = \emptyset$ for each individual a occurring in the KB
 - an edge $\langle a, b \rangle$ labelled $\mathcal{L}(\langle a, b \rangle) = \{R\}$ for each $(a, b):R$ occurring in the KB
- Then, for each $a:C$ occurring in the KB, set $\mathcal{L}(a) \rightarrow \mathcal{L}(a) \cup \{C\}$
- The algorithm expands the tree either by extending $\mathcal{L}(x)$ for some node x or by adding new leaf nodes.
- Edges are added when expanding $\exists R.C$
- A completion-forest contains a **clash** if, for a node x , $\{C, \neg C\} \subseteq \mathcal{L}(x)$
- If nodes x and y are connected by an edge $\langle x, y \rangle$, then y is called a successor of x and x is called a predecessor of y . Ancestor is the transitive closure of predecessor.
- A node y is called an R -successor of a node x if y is a successor of x and $\mathcal{L}(\langle x, y \rangle) = \{R\}$.
- The algorithm returns "satisfiable" if rules can be applied s.t. they yield a clash-free, complete (no more rules can be applied) completion forest

\mathcal{ALC} Tableau rules without GCI's

Rule	Description
(\sqcap)	if 1. $C_1 \sqcap C_2 \in \mathcal{L}(x)$ and 2. $\{C_1, C_2\} \not\subseteq \mathcal{L}(x)$ then $\mathcal{L}(x) \rightarrow \mathcal{L}(x) \cup \{C_1, C_2\}$
(\sqcup)	if 1. $C_1 \sqcup C_2 \in \mathcal{L}(x)$ and 2. $\{C_1, C_2\} \cap \mathcal{L}(x) = \emptyset$ then $\mathcal{L}(x) \rightarrow \mathcal{L}(x) \cup \{C\}$ for some $C \in \{C_1, C_2\}$
(\exists)	if 1. $\exists R.C \in \mathcal{L}(x)$ and 2. x has no R -successor y with $C \in \mathcal{L}(y)$ then create a new node y with $\mathcal{L}(\langle x, y \rangle) = \{R\}$ and $\mathcal{L}(y) = \{C\}$
(\forall)	if 1. $\forall R.C \in \mathcal{L}(x)$ and 2. x has an R -successor y with $C \notin \mathcal{L}(y)$ then $\mathcal{L}(y) \rightarrow \mathcal{L}(y) \cup \{C\}$

Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

$$\mathcal{L}(x) = \{\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D\}$$

x

Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

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x

Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

x

Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

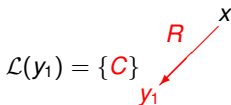
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

x

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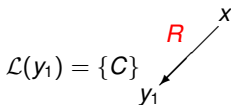
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Example

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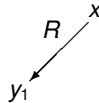
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$



Example

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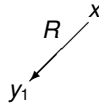
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

$$\mathcal{L}(y_1) = \{C, \neg C \sqcup \neg D\}$$


Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

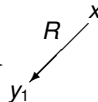
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

$$\mathcal{L}(y_1) = \{C, \neg C \sqcup \neg D\}$$


Example

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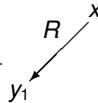
Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

$$\mathcal{L}(y_1) = \{C, \neg C \sqcup \neg D, \neg C\}$$

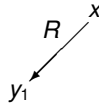
Clash



Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

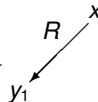
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

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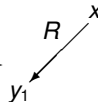
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Example

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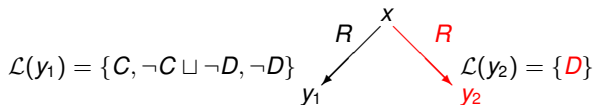
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

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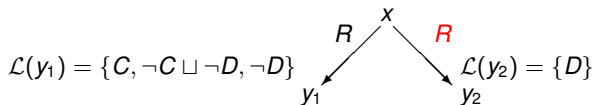
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$



Example

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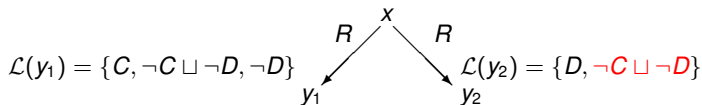
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$



Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

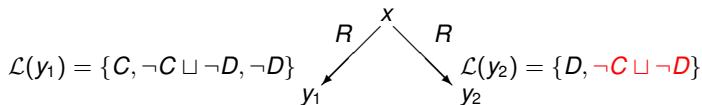
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$



Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes. w

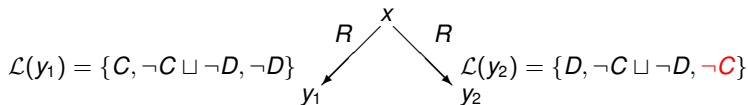
$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$



Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

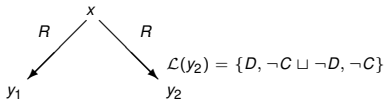


Example

Is $\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D$ satisfiable? Yes.

$$\mathcal{L}(x) = \{\exists R.C, \forall R.(\neg C \sqcup \neg D), \exists R.D\}$$

$$\mathcal{L}(y_1) = \{C, \neg C \sqcup \neg D, \neg D\}$$



- Finished. No more rules applicable and the tableau is complete and clash-free
- Hence, the concept is **satisfiable**
- The tree corresponds to a **model** $\mathcal{I} = (\Delta^{\mathcal{I}}, \cdot^{\mathcal{I}})$
 - The nodes are the elements of the domain: $\Delta^{\mathcal{I}} = \{x, y_1, y_2\}$
 - For each atomic concept A , set $A^{\mathcal{I}} = \{z \mid A \in \mathcal{L}(z)\}$
 - $C^{\mathcal{I}} = \{y_1\}, D^{\mathcal{I}} = \{y_2\}$
 - For each role R , set $R^{\mathcal{I}} = \{\langle x, y \rangle \mid \text{there is an edge labeled } R \text{ from } x \text{ to } y\}$
 - $R^{\mathcal{I}} = \{\langle x, y_1 \rangle, \langle x, y_2 \rangle\}$
 - It can be shown that $x \in (\exists R.C \sqcap \forall R.(\neg C \sqcup \neg D) \sqcap \exists R.D)^{\mathcal{I}} \neq \emptyset$

Soundness and Completeness

Theorem

Let \mathcal{A} be an \mathcal{ALC} ABox and F a completion-forest obtained by applying the tableau rules to \mathcal{A} . Then

- 1 The rule application terminates;*
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KBs with GCIs

- We have seen how to test the satisfiability of an ABox \mathcal{A}
- But, how can we check if a KB $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$ is satisfiable with $\mathcal{T} \neq \emptyset$?
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KB Satisfiability

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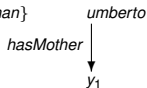
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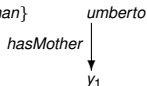


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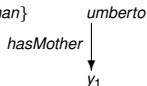


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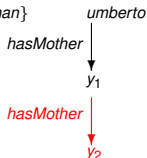
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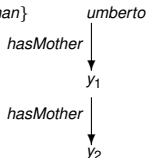
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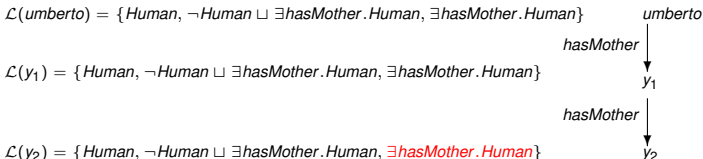
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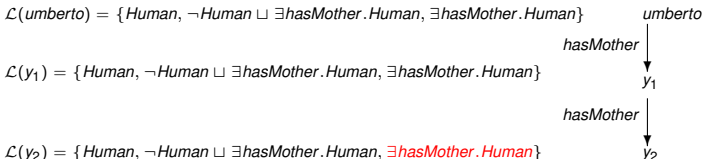
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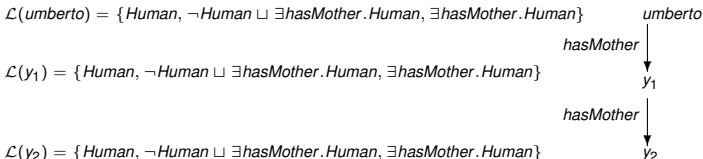
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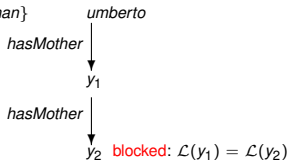
Node Blocking in \mathcal{ALC}

- When creating new node, check ancestors for equal label set
- If such a node is found, new node is **blocked**
- No rule is applied to blocked nodes

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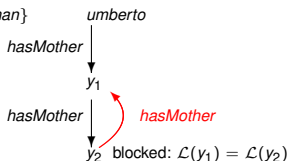
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- Block represents **cyclical** model
 - $\Delta^{\mathcal{I}} = \{umberto, y_1, y_2\}$
 - $Human^{\mathcal{I}} = \{umberto, y_1, y_2\}$
 - $hasMother^{\mathcal{I}} = \{\langle umberto, y_1 \rangle, \langle y_1, y_2 \rangle, \langle y_2, y_1 \rangle\}$

Blocking in \mathcal{ALC}

- A non-root node x is blocked if for some ancestor y , y is blocked or $\mathcal{L}(x) = \mathcal{L}(y)$, where y is not a root node
- A blocked node x is indirectly blocked if its predecessor is blocked, otherwise it is directly blocked
- If x is directly blocked, it has a unique ancestor y such that $\mathcal{L}(x) = \mathcal{L}(y)$
- if there existed another ancestor z such that $\mathcal{L}(x) = \mathcal{L}(z)$ then either y or z must be blocked
- If x is directly blocked and y is the unique ancestor such that $\mathcal{L}(x) = \mathcal{L}(y)$, we will say that y blocks x

\mathcal{ALC} Tableau rules with GCI's

Rule	Description
(\sqcap)	if 1. $C_1 \sqcap C_2 \in \mathcal{L}(x)$, x is not indirectly blocked and 2. $\{C_1, C_2\} \not\subseteq \mathcal{L}(x)$ then $\mathcal{L}(x) \rightarrow \mathcal{L}(x) \cup \{C_1, C_2\}$
(\sqcup)	if 1. $C_1 \sqcup C_2 \in \mathcal{L}(x)$, x is not indirectly blocked and 2. $\{C_1, C_2\} \cap \mathcal{L}(x) = \emptyset$ then $\mathcal{L}(x) \rightarrow \mathcal{L}(x) \cup \{C\}$ for some $C \in \{C_1, C_2\}$
(\exists)	if 1. $\exists R.C \in \mathcal{L}(x)$, x is not blocked and 2. x has no R -successor y with $C \in \mathcal{L}(y)$ then create a new node y with $\mathcal{L}(\langle x, y \rangle) = \{R\}$ and $\mathcal{L}(y) = \{C\}$
(\forall)	if 1. $\forall R.C \in \mathcal{L}(x)$, x is not indirectly blocked and 2. x has an R -successor y with $C \notin \mathcal{L}(y)$ then $\mathcal{L}(y) \rightarrow \mathcal{L}(y) \cup \{C\}$
(\sqsubseteq)	if 1. $C \sqsubseteq D \in \mathcal{T}$, x is not indirectly blocked and 2. $\{nfn(\neg C), D\} \cap \mathcal{L}(x) = \emptyset$ then $\mathcal{L}(x) \rightarrow \mathcal{L}(x) \cup \{E\}$ for some $E \in \{nfn(\neg C), D\}$ ($nfn(\neg C)$ is NNF of $\neg C$)

Soundness and Completeness

Theorem

Let \mathcal{K} be an \mathcal{ALC} KB and F a completion-forest obtained by applying the tableau rules to \mathcal{K} . Then

- 1 The rule application terminates;*
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Fuzzy DLs Basics

- We have seen how to “fuzzify” classical sets and FOL
 - Fuzzy statements are of the form $\langle \phi, n \rangle$, where ϕ is a statement and $n \in [0, 1]$
- The natural extension to fuzzy DLs consists then in replacing ϕ with a DL expression
- Several **fuzzy** variants of DLs have been proposed: they can be classified according to
 - The DL resp. ontology language that they generalize
 - The allowed fuzzy constructs
 - The underlying fuzzy logic
 - Their reasoning algorithms and computational complexity results

- In classical DLs, a concept C is interpreted by an interpretation \mathcal{I} as a set of individuals
- In fuzzy DLs, a concept C is interpreted by \mathcal{I} as a fuzzy set of individuals
- Each individual is instance of a concept to a degree in $[0, 1]$
- Each pair of individuals is instance of a role to a degree in $[0, 1]$

- $\langle a:C, n \rangle$ states that a is an instance of concept/class C with degree at least n
- $\langle (a, b):R, n \rangle$ states that $\langle a, b \rangle$ is an instance of relation R with degree at least n
- $\langle C_1 \sqsubseteq C_2, n \rangle$ states a vague subsumption relationship
 - The FOL statement $\forall x. C_1(x) \rightarrow C_2(x)$ is true to degree at least n
- **Note:** one may find also fuzzy DL expressions $\langle \alpha \geq n \rangle$, $\langle \alpha \leq n \rangle$, $\langle \alpha > n \rangle$, $\langle \alpha < n \rangle$, and $\langle \alpha = n \rangle$
- We use the form $\langle \alpha, n \rangle$, i.e. $\langle \alpha \geq n \rangle$ only
 - Remind that graded axioms are intended to be produced semi- or automatically
 - Hardly they may have the form $\langle \alpha \leq n \rangle$, $\langle \alpha > n \rangle$ or $\langle \alpha < n \rangle$

The semantics is an immediate consequence of the First-Order-Logic translation of DLs expressions

Interpretation:

\mathcal{I}	=	$\Delta^{\mathcal{I}}$	\otimes	=	t-norm
$C^{\mathcal{I}}$:	$\Delta^{\mathcal{I}} \rightarrow [0, 1]$	\oplus	=	s-norm
$R^{\mathcal{I}}$:	$\Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}} \rightarrow [0, 1]$	\neg	=	negation
			\Rightarrow	=	implication

	Syntax	Semantics
Concepts:	$C, D \rightarrow$	$\top^{\mathcal{I}}(x) = 1$
		$\perp^{\mathcal{I}}(x) = 0$
		$A^{\mathcal{I}}(x) \in [0, 1]$
	$C \sqcap D$	$(C_1 \sqcap C_2)^{\mathcal{I}}(x) = C_1^{\mathcal{I}}(x) \otimes C_2^{\mathcal{I}}(x)$
	$C \sqcup D$	$(C_1 \sqcup C_2)^{\mathcal{I}}(x) = C_1^{\mathcal{I}}(x) \oplus C_2^{\mathcal{I}}(x)$
	$C \rightarrow D$	$(C \rightarrow D)^{\mathcal{I}}(x) = C^{\mathcal{I}}(x) \Rightarrow D^{\mathcal{I}}(x)$
	$\neg C$	$(\neg C)^{\mathcal{I}}(x) = \neg C^{\mathcal{I}}(x)$
	$\exists R.C$	$(\exists R.C)^{\mathcal{I}}(x) = \sup_{y \in \Delta^{\mathcal{I}}} R^{\mathcal{I}}(x, y) \otimes C^{\mathcal{I}}(y)$
	$\forall R.C$	$(\forall R.C)^{\mathcal{I}}(x) = \inf_{y \in \Delta^{\mathcal{I}}} R^{\mathcal{I}}(x, y) \Rightarrow C^{\mathcal{I}}(y)$
	$\{a\}$	$\{a\}^{\mathcal{I}}(x) = 1 \text{ if } a^{\mathcal{I}} = x, \text{ else } 0$

Assertions: $\langle a:C, r \rangle, \mathcal{I} \models \langle a:C, r \rangle$ iff $C^{\mathcal{I}}(a^{\mathcal{I}}) \geq r$ (similarly for roles)

General Inclusion Axioms: $\langle C \sqsubseteq D, r \rangle,$

● $\mathcal{I} \models \langle C \sqsubseteq D, r \rangle$ iff $\inf_{x \in \Delta^{\mathcal{I}}} C^{\mathcal{I}}(x) \Rightarrow D^{\mathcal{I}}(x) \geq r$

Some Remarks

- Like for fuzzy FOL, \forall and \exists are not complementary in general: i.e. $\forall R.C \not\equiv \neg \exists R.\neg C$
- $\forall R.C \equiv \neg \exists R.\neg C$ under Łukasiewicz logic and SFL
- $\langle C \sqsubseteq D, n \rangle$ may be rewritten as $\langle \top \sqsubseteq C \rightarrow D, n \rangle$
- In early works, a fuzzy GCI is of the form $C \sqsubseteq D$ with semantics:
 - \mathcal{I} is a model of $C \sqsubseteq D$ iff for every $x \in \Delta^{\mathcal{I}}$ we have that $C^{\mathcal{I}}(x) \leq D^{\mathcal{I}}(x)$
 - This is the same of fuzzy axiom $\langle \top \sqsubseteq C \rightarrow_x D, 1 \rangle$, where \rightarrow_x is an r -implication
- **Disjointness**: use $\langle C \sqcap D \sqsubseteq \perp, 1 \rangle$ rather than $\langle C \sqsubseteq \neg D, 1 \rangle$
 - they are not the same, e.g. $\langle A \sqsubseteq \neg A, 1 \rangle$ says that $A^{\mathcal{I}}(x) \leq 0.5$, for all \mathcal{I} and for all $x \in \Delta^{\mathcal{I}}$

Witnessed Interpretation

- **Witnessed Interpretation:**
 - Infima and suprema are attained at some point

$$\begin{aligned}(\exists R.C)^{\mathcal{I}}(x) &= R^{\mathcal{I}}(x, y) \otimes C^{\mathcal{I}}(y) \text{ for some } y \in \Delta^{\mathcal{I}} \\ (\forall R.C)^{\mathcal{I}}(x) &= R^{\mathcal{I}}(x, y) \Rightarrow C^{\mathcal{I}}(y) \text{ for some } y \in \Delta^{\mathcal{I}} \\ (C \sqsubseteq D)^{\mathcal{I}} &= C^{\mathcal{I}}(x) \Rightarrow D^{\mathcal{I}}(x) \text{ for some } x \in \Delta^{\mathcal{I}}\end{aligned}$$

- It is customary to stick to witnessed interpretations only

- **Fuzzy knowledge base:** $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$
 - \mathcal{T} is a **fuzzy TBox**, that is a finite set of fuzzy GCI
 - \mathcal{A} is a **fuzzy ABox**, that is a finite set of fuzzy assertions
- **Acyclic fuzzy ontologies:** TBox with axioms of the form

$A \sqsubseteq_n C$ (**primitive** GCI)

$A \sqsubseteq_{\tilde{}} C$ (**primitive** GCI)

$A \simeq C$ (**definitional** GCI)

- A concept name
- $A \sqsubseteq_n C$ shorthand for $\langle \top \sqsubseteq A \rightarrow C, n \rangle$
- No nominal $\{a\}$ occurs in the TBox

- We say that
 - concept name A **directly uses** concept name B w.r.t. \mathcal{T} , denoted $A \rightarrow_{\mathcal{T}} B$, if A is the head of some axiom $\tau \in \mathcal{T}$ such that B occurs in the body of τ
 - concept name A **uses** concept name B w.r.t. \mathcal{T} , denoted $A \rightsquigarrow_{\mathcal{T}} B$, if there exist concept names A_1, \dots, A_n , such that $A_1 = A$, $A_n = B$ and, for every $1 \leq i < n$, it holds that $A_i \rightarrow_{\mathcal{T}} A_{i+1}$
- TBox \mathcal{T} is **cyclic (acyclic)** if there is (no) A such that $A \rightsquigarrow_{\mathcal{T}} A$
- TBox \mathcal{T} is **unfoldable** if
 - \mathcal{T} is acyclic
 - If $A \stackrel{\mathcal{T}}{\equiv} C \in \mathcal{T}$ then A does not occur in the head of any other axiom

- \mathcal{I} **satisfies (is a model of)** $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$ iff it satisfies each element in \mathcal{A} and \mathcal{T}
- A fuzzy KB $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$ **entails** an axiom E , denoted $\mathcal{K} \models E$, iff every model of \mathcal{K} satisfies E
- We say that two concepts C and D are **equivalent**, denoted $C \equiv_{\mathcal{K}} D$ iff in every model \mathcal{I} of \mathcal{K} and for all $x \in \Delta^{\mathcal{I}}$, $C^{\mathcal{I}}(x) = D^{\mathcal{I}}(x)$
- **Best entailment degree**: for assertion of GCI ϕ

$$bed(\mathcal{K}, \phi) = \sup \{r \mid \mathcal{K} \models \langle \phi, r \rangle\}$$

- **Best satisfiability degree**: for concept C

$$bsd(\mathcal{K}, C) = \sup_{\mathcal{I} \models \mathcal{K}} \sup_{x \in \Delta^{\mathcal{I}}} C^{\mathcal{I}}(x).$$

Some Salient Fuzzy Concept Equivalences

Property	Łukasiewicz	Gödel	Product	SFL
$C \sqcap \neg C \equiv \perp$	•	•	•	
$C \sqcup \neg C \equiv \top$	•			
$C \sqcap C \equiv C$		•		•
$C \sqcup C \equiv C$		•		•
$\neg \neg C \equiv C$	•			•
$C \rightarrow D \equiv \neg C \sqcup D$	•			•
$C \rightarrow D \equiv \neg D \rightarrow \neg C$	•			•
$\neg(C \rightarrow D) \equiv C \sqcap \neg D$	•			•
$\neg(C \sqcap D) \equiv \neg C \sqcup \neg D$	•	•	•	•
$\neg(C \sqcup D) \equiv \neg C \sqcap \neg D$	•	•	•	•
$C \sqcap (D \sqcup E) \equiv (C \sqcap D) \sqcup (C \sqcap E)$		•		•
$C \sqcup (D \sqcap E) \equiv (C \sqcup D) \sqcap (C \sqcup E)$		•		•
$\exists R.C \equiv \neg \forall R.\neg C$	•			•

Towards Fuzzy OWL 2 and its Profiles

- Recall that OWL 2 relates to $SR_{OIQ}(D)$
- We need to extend the semantics to fuzzy $SR_{OIQ}(D)$
- Additionally, we add
 - **modifiers** (e.g., *very*)
 - **concrete fuzzy concepts** (e.g., *Young*)
 - both additions have **explicit** membership functions
 - other extensions:
 - aggregation functions: weighted sum, OWA, fuzzy integrals
 - fuzzy rough sets, fuzzy spatial, fuzzy numbers

Number Restrictions, Inverse, Transitive roles, ...

- The semantics of the concept $(\geq n R.C)$ is: \wedge interpreted as Gödel t-norm

$$\exists y_1, \dots, y_n. \bigwedge_{i=1}^n R(x, y_i) \wedge C(y_i) \wedge \bigwedge_{1 \leq i < j \leq n} y_i \neq y_j$$

- The semantics of the concept $(\leq n R.C)$ is: \wedge interpreted as Gödel t-norm

$$(\leq n R)^{\mathcal{I}}(x) = \forall y_1, \dots, y_{n+1}. \bigwedge_{i=1}^{n+1} (R(x, y_i) \wedge C(y_i)) \Rightarrow \bigvee_{1 \leq i < j \leq n+1} y_i = y_j$$

- Note: $(\geq 1 R) \equiv \exists R.T$
- For transitive roles R we impose: for all $x, y \in \Delta^{\mathcal{I}}$

$$R^{\mathcal{I}}(x, y) \geq \sup_{z \in \Delta^{\mathcal{I}}} R^{\mathcal{I}}(x, z) \otimes R^{\mathcal{I}}(z, y)$$

- For inverse roles we have for all $x, y \in \Delta^{\mathcal{I}}$

$$R^{\mathcal{I}}(x, y) = R^{\mathcal{I}}(y, x)$$

- The semantics of functional roles $fun(R)$ is

$$\forall x \forall y \forall z. R(x, y) \wedge R(x, z) \Rightarrow y = z$$

- Similar for other \mathcal{SROIQ} constructs

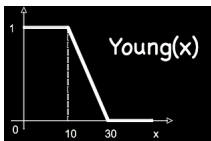
Fuzzy Concrete Domains

- E.g., *Small*, *Young*, *High*, etc. with **explicit** membership function
- Use the idea of concrete domains:
 - $D = \langle \Delta_D, \Phi_D \rangle$
 - Δ_D is an interpretation domain
 - Φ_D is the set of concrete unary fuzzy domain predicates d and **fixed** interpretation $d^D: \Delta_D \rightarrow [0, 1]$
- Specifically,

$$\mathbf{d} \quad \rightarrow \quad |s(a, b) \mid rs(a, b) \mid tri(a, b, c) \mid trz(a, b, c, d) \\ \mid \geq_v \mid \leq_v \mid =_v$$

$$C, D \quad \rightarrow \quad \forall T.\mathbf{d} \mid \exists T.\mathbf{d}$$

- Representation of **Young Person**:



$$\begin{aligned} \text{Minor} &= \text{Person} \sqcap \exists \text{hasAge} . \leq 18 \\ \text{YoungPerson} &= \text{Person} \sqcap \exists \text{hasAge} . \text{Is}(10, 30) \end{aligned}$$

- Representation of **Heavy Rain**:

$$\text{HeavyRain} = \text{Rain} \sqcap \exists \text{hasPrecipitationRate} . \text{rs}(5, 7.5)$$

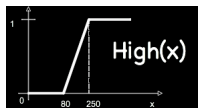
Modifiers

- *Very, moreOrLess, slightly, etc.*
- **Fuzzy modifier** m with function $f_m: [0, 1] \rightarrow [0, 1]$

$$C \rightarrow m(C) \mid \forall T.m(\mathbf{d}) \mid \exists T.m(\mathbf{d})$$

where m is a linear modifier

- Representation of **Sport Car**



$$\text{SportsCar} = \text{Car} \sqcap \exists \text{speed}.\text{very}(\text{rs}(80, 250))$$

- Representation of **Very Heavy Rain**

$$\text{VeryHeavyRain} = \text{Rain} \sqcap \exists \text{hasPrecipitationRate}.\text{very}(\text{rs}(5, 7.5)) .$$

Aggregation Operators

- **Aggregation operators**: aggregate concepts, using functions such as the mean, median, weighted sum operators
- Given an n -ary aggregation operator $@ : [0, 1]^n \rightarrow [0, 1]$
 - We fuzzy concepts by allowing to apply $@$ to n concepts C_1, \dots, C_n , i.e.

$$C \rightarrow @(C_1, \dots, C_n)$$

- Semantics:

$$@(C_1, \dots, C_n)^{\mathcal{I}}(x) = @(C_1^{\mathcal{I}}(x), \dots, C_n^{\mathcal{I}}(x)).$$

- Allows to express the concept

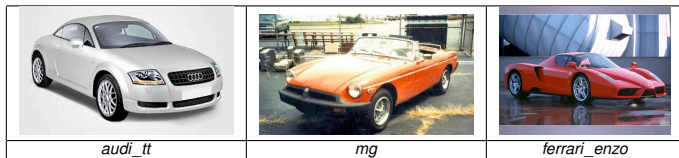
$$GoodHotel = 0.3 \cdot ExpensiveHotel + 0.7 \cdot LuxuriousHotel$$

- The membership function of good hotels is the weighted sum of being an expensive and luxurious hotel

Some Applications

- Information retrieval
- Recommendation systems
- Image interpretation
- Ambient intelligence
- Ontology merging
- Matchmaking
- decision making
- Summarization
- Robotics perception
- Software design
- Machine learning

Example (Graded Entailment)

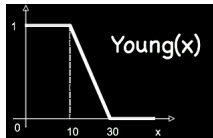


Car	speed
audi_tt	243
mg	≤ 170
ferrari_enzo	≥ 350

SportsCar = *Car* \sqcap \exists hasSpeed.very(High)

$\mathcal{K} \models \langle \text{ferrari_enzo}:\text{SportsCar}, 1 \rangle$
 $\mathcal{K} \models \langle \text{audi_tt}:\text{SportsCar}, 0.92 \rangle$
 $\mathcal{K} \models \langle \text{mg}:\neg\text{SportsCar}, 0.72 \rangle$

Example (Graded Subsumption)

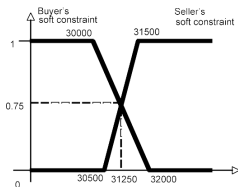


$$\begin{aligned} \text{Minor} &= \text{Person} \sqcap \exists \text{hasAge} . \leq_{18} \\ \text{YoungPerson} &= \text{Person} \sqcap \exists \text{hasAge} . \text{Young} \\ &\quad \text{fun}(\text{hasAge}) \end{aligned}$$

$$\mathcal{K} \models \langle \text{Minor} \sqsubseteq \text{YoungPerson}, 0.6 \rangle$$

Note: without an explicit membership function of *Young*, **this inference cannot be drawn**

Example (Simplified Matchmaking)



- A car seller sells an Audi TT for 31500€, as from the catalog price.
- A buyer is looking for a sports-car, but wants to pay not more than around 30000€
- Classical sets: the problem relies on the crisp conditions on price
- More fine grained approach: to consider prices as fuzzy sets (as usual in negotiation)
 - Seller may consider optimal to sell above 31500€, but can go down to 30500€
 - The buyer prefers to spend less than 30000€, but can go up to 32000€
 $AudiTT = SportsCar \sqcap \exists hasPrice.rs(30500, 31500)$
 $Query = SportsCar \sqcap \exists hasPrice.ls(30000, 32000)$
 - Highest degree to which the concept
 $C = AudiTT \sqcap Query$
is satisfiable is 0.75 (the degree to which the Audi TT and the query **matches** is 0.75)
 - The car may be sold at 31250€

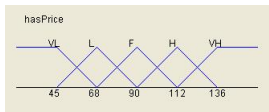
Example: Learning fuzzy GCIs from data

- Learning of fuzzy GCIs from crisp data
- Use Case: What are **Good hotels**, using TripAdvisor data?
 - Given
 - OWL 2 Ontology about meaningful city entities and their descriptions
 - TripAdvisor data about hotels and user judgments
 - We may learn that in e.g., Pisa, Italy

$\langle \exists \text{hasAmenity.Babysitting} \sqcap \exists \text{hasPrice.fair} \sqsubseteq \text{Good_Hotel}, 0.282 \rangle$

“A hotel having babysitting as amenity and a fair price is a good hotel (to degree 0.282)”

- Real valued price attribute *hasPrice* has been automatically fuzzyfied



Example: Multi-Criteria Decision Making

- We have to select among two sites, A_1, A_2
- There are two criteria (C_1 -Transportation Issues, and C_2 -Public Nuisance) for judgement
- There are two experts (E_1, E_2) that make judgments
- The decision matrix of the experts is shown below:

E_1		Criteria	
		0.48	0.52
Alter.		C_1	C_2
x_1	A_1	$tri(0.6, 0.7, 0.8)$	$tri(0.9, 0.95, 1.0)$
x_2	A_2	$tri(0.6, 0.7, 0.8)$	$tri(0.4, 0.5, 0.6)$

E_2		Criteria	
		0.52	0.48
Alter.		C_1	C_2
x_1	A_1	$tri(0.55, 0.6, 0.7)$	$tri(0.4, 0.45, 0.5)$
x_2	A_2	$tri(0.35, 0.4, 0.45)$	$tri(0.5, 0.55, 0.6)$

- For each expert $k = 1, 2$, for each alternative $i = 1, 2$ and for each criteria $j = 1, 2$, we define the concept

$$P_{ij}^k = \exists \text{hasScore}.a_{ij}^k$$

- Now, for each expert k and alternative i , we define the weighted concept

$$A_i^k = w_1^k \cdot P_{i1}^k + w_2^k \cdot P_{i2}^k$$

- Finally, we combine the two experts outcome, by defining the weighted concept

$$A_i = 0.5 \cdot A_i^1 + 0.5 \cdot A_i^2$$

- It can be verified that $rv(\mathcal{K}, A_1) = bsd(\mathcal{K}, A_1) = 0.26$ and $rv(\mathcal{K}, A_2) = bsd(\mathcal{K}, A_2) = 0.37$

Representing Fuzzy OWL Ontologies in OWL

- OWL 2 is W3C standard, with classical logic semantics
 - Hence, cannot support natively Fuzzy Logic
- However, **Fuzzy OWL 2**, has been defined using OWL 2
 - Uses the axiom annotation feature of OWL 2
- Any Fuzzy OWL 2 ontology is a legal OWL 2 ontology

- A java parser for Fuzzy OWL 2 exists
- Protégé plug-in exists to encode Fuzzy OWL ontologies

Reasoning Problems and Algorithms

Consistency problem:

- Is \mathcal{K} satisfiable?
- Is \mathcal{C} coherent, i.e. is $C^{\mathcal{I}}(x) > 0$ for some $\mathcal{I} \models \mathcal{K}$ and $x \in \Delta^{\mathcal{I}}$?

Instance checking problem:

- Does $\mathcal{K} \models \langle a:C, n \rangle$ hold?

Subsumption problem:

- Does $\mathcal{K} \models \langle C \sqsubseteq D, n \rangle$ hold?

Best entailment degree problem:

- What is $bed(\mathcal{K}, \phi)$?

Best satisfiability degree problem:

- What is $bsd(\mathcal{K}, \phi)$?

Instance retrieval problem:

- Compute the set $\{\langle a, n \rangle \mid n = bed(\mathcal{K}, a:C)\}$

Top-k retrieval problem:

- Compute the top-k ranked elements of $\{\langle a, n \rangle \mid n = bed(\mathcal{K}, a:C)\}$

Some Reductions

- \mathcal{K} is satisfiable iff $bsd(\mathcal{K}, a:\perp) > 0$, where a is a new individual.
- C is coherent w.r.t. \mathcal{K} if one of the following holds:
 - $\mathcal{K} \cup \{\langle a:C > 0 \rangle\}$ is satisfiable, where a is a new individual
 - $\mathcal{K} \not\models \langle C \sqsubseteq \perp, 1 \rangle$
 - $bsd(\mathcal{K}, C) > 0$
- $\mathcal{K} \models \langle a:C, n \rangle$ if one of the following holds:
 - $\mathcal{K} \cup \{\langle a:C < n \rangle\}$ is not satisfiable
 - $bed(\mathcal{K}, a:C) \geq n$
- $\mathcal{K} \models \langle C \sqsubseteq D, n \rangle$ if one of the following holds:
 - $\mathcal{K} \cup \{\langle a:C \rightarrow D < n \rangle\}$ is not satisfiable, where a is a new individual
 - $bed(\mathcal{K}, C \sqsubseteq D) \geq n$
- We have that

$$bed(\mathcal{K}, \phi) = \min x. \text{ such that } \mathcal{K} \cup \{\langle \phi \leq x \rangle\} \text{ satisfiable}$$

$$bsd(\mathcal{K}, \phi) = \max x. \text{ such that } \mathcal{K} \cup \{\langle \phi \geq x \rangle\} \text{ satisfiable}$$

Reasoning in Fuzzy DLs: Basics

- Algorithms for fuzzy DLs: are a mixture of classical DLs reasoning algorithms and algorithms for Mathematical Fuzzy Logic
- Fuzzy OWL 2:
 - **Fuzzy tableaux** based algorithms
 - Tableaux and non deterministic tableaux
 - Operational Research
 - **Reduction** into classical DLs
- Fuzzy OWL 2 EL: **fuzzy structural** based algorithms
- Fuzzy OWL 2 QL: **fuzzy query rewriting** based algorithms
- fuzzy OWL 2 RL: **fuzzy logic programming** based algorithms

OR Fuzzy Tableaux: \mathcal{ALC} under SFL over $[0, 1]$

- Works as for classical \mathcal{ALC} on completion forests
 - Blocking is as for classical \mathcal{ALC}
 - The completion forest is expanded by repeatedly applying inference rules
 - The completion-forest is complete when none of the rules are applicable
- Additionally, at each inference step we add equational constraints that have to hold
- Eventually, the initial KB is satisfiable if the final set of equational constraints has a solution
 - For the latter case, we may use a MILP solver

Rule	Description
(var)	For variable $x_{v:C}$ add $x_{v:C} \in [0, 1]$ to $\mathcal{C}_{\mathcal{F}}$. For variable $x_{(v,w):R}$, add $x_{(v,w):R} \in [0, 1]$ to $\mathcal{C}_{\mathcal{F}}$
(\bar{A})	if $\neg A \in \mathcal{L}(v)$ then add $x_{v:A} = 1 - x_{v:\neg A}$ to $\mathcal{C}_{\mathcal{F}}$
(\perp)	If $\perp \in \mathcal{L}(v)$ then add $x_{v:\perp} = 0$ to $\mathcal{C}_{\mathcal{F}}$
(\top)	If $\top \in \mathcal{L}(v)$ then add $x_{v:\top} = 1$ to $\mathcal{C}_{\mathcal{F}}$
(\sqcap)	if $C_1 \sqcap C_2 \in \mathcal{L}(v)$, v is not indirectly blocked then $\mathcal{L}(v) \rightarrow \mathcal{L}(v) \cup \{C_1, C_2\}$, and add $x_{v:C_1} \otimes x_{v:C_2} \geq x_{v:C_1 \sqcap C_2}$ to $\mathcal{C}_{\mathcal{F}}$
(\sqcup)	if $C_1 \sqcup C_2 \in \mathcal{L}(v)$, v is not indirectly blocked then $\mathcal{L}(v) \rightarrow \mathcal{L}(v) \cup \{C_1, C_2\}$, and add $x_{v:C_1} \oplus x_{v:C_2} \geq x_{v:C_1 \sqcup C_2}$ to $\mathcal{C}_{\mathcal{F}}$
(\forall)	if $\forall R.C \in \mathcal{L}(v)$, v is not indirectly blocked then $\mathcal{L}(w) \rightarrow \mathcal{L}(w) \cup \{C\}$, and add $x_{w:C} \geq x_{v:\forall R.C} \otimes x_{(v,w):R}$ to $\mathcal{C}_{\mathcal{F}}$
(\exists)	if $\exists R.C \in \mathcal{L}(v)$, v is not blocked then create new node w with $\mathcal{L}(\langle v, w \rangle) = \{R\}$ and $\mathcal{L}(w) = \{C\}$, and add $x_{w:C} \otimes x_{(v,w):R} \geq x_{v:\exists R.C}$ to $\mathcal{C}_{\mathcal{F}}$
(\sqsubseteq)	if $\langle C \sqsubseteq D, n \rangle \in \mathcal{T}$, v is not indirectly blocked then $\mathcal{L}(v) \rightarrow \mathcal{L}(v) \cup \{C, D\}$, and add $x_{v:D} \geq x_{v:C} \otimes n$ to $\mathcal{C}_{\mathcal{F}}$

Analytical Fuzzy Tableaux: \mathcal{ALC} under SFL over $[0, 1]$

- Works as for classical \mathcal{ALC} on completion forests
 - Node labels $\mathcal{L}(v)$ contain, rather than DL concept expressions, expressions of the form $\langle C, n \rangle$

“The truth degree of being v instance of C is $\geq n$ ”
 - Blocking is as for classical \mathcal{ALC}
 - The completion forest is expanded by repeatedly applying inference rules
 - The completion-forest is complete when none of the rules are applicable
- Additionally, we adapt the notion of **clash**: a clash is either
 - $\langle \perp, n \rangle$ with $n > 0$; or
 - a pair $\langle C, n \rangle$ and $\langle \neg C, m \rangle$ with $n > 1 - m$
- Eventually, the initial KB is satisfiable if there is a clash-free complete completion forest

- (\sqcap). If (i) $\langle C_1 \sqcap C_2, n \rangle \in \mathcal{L}(v)$, (ii) $\{\langle C_1, n \rangle, \langle C_2, n \rangle\} \not\subseteq \mathcal{L}(v)$, and (iii) node v is not indirectly blocked, then add $\langle C_1, n \rangle$ and $\langle C_2, n \rangle$ to $\mathcal{L}(v)$.
- (\sqcup). If (i) $\langle C_1 \sqcup C_2, n \rangle \in \mathcal{L}(v)$, (ii) $\{\langle C_1, n \rangle, \langle C_2, n \rangle\} \cap \mathcal{L}(v) = \emptyset$, and (iii) node v is not indirectly blocked, then add some $\langle C, n \rangle \in \{\langle C_1, n \rangle, \langle C_2, n \rangle\}$ to $\mathcal{L}(v)$.
- (\forall). If (i) $\langle \forall R.C, n \rangle \in \mathcal{L}(v)$, (ii) $\langle R, m \rangle \in \mathcal{L}(\langle v, w \rangle)$ with $m > 1 - n$, (iii) $\langle C, n \rangle \notin \mathcal{L}(w)$, and (iv) node v is not indirectly blocked, then add $\langle C, n \rangle$ to $\mathcal{L}(w)$.
- (\exists). If (i) $\langle \exists R.C, n \rangle \in \mathcal{L}(v)$, (ii) there is no $\langle R, n_1 \rangle \in \mathcal{L}(\langle v, w \rangle)$ with $\langle C, n_2 \rangle \in \mathcal{L}(w)$ such that $\min(n_1, n_2) \geq n$, and (iii) node v is not blocked, then create a new node w , add $\langle R, n \rangle$ to $\mathcal{L}(\langle v, w \rangle)$ and add $\langle C, n \rangle$ to $\mathcal{L}(w)$.
- (\sqsubseteq). If (i) $\langle \top \sqsubseteq D, n \rangle \in \mathcal{T}$, (ii) $\langle D, n \rangle \notin \mathcal{L}(v)$, and (iii) node v is not indirectly blocked, then add $\langle D, n \rangle$ to $\mathcal{L}(v)$.

Non-Deterministic Analytic Fuzzy Tableaux

- It's a combination of the analogous method for fuzzy propositional logic and analytical fuzzy tableau
- Rule example:
 - (\sqcap). If (i) $\langle C_1 \sqcap C_2, m \rangle \in \mathcal{L}(v)$, (ii) there are $m_1, m_2 \in L_n$ such that $m_1 \otimes m_2 = m$ with $\{\langle C_1, m_1 \rangle, \langle C_2, m_2 \rangle\} \not\subseteq \mathcal{L}(v)$, and (iii) node v is not indirectly blocked, then add $\langle C_1, m_1 \rangle$ and $\langle C_2, m_2 \rangle$ to $\mathcal{L}(v)$

Reduction to Classical DLs

- Same principle as for the reduction for propositional fuzzy logic
- Needs adaption to the DL constructs: e.g. \exists , \forall and \sqsubseteq
- Examples of reduction rules for SFL:

$$\begin{aligned}\rho(A, \geq \gamma) &= A_{\geq \gamma} \\ \rho(C \sqcap D, \geq \gamma) &= \rho(C, \geq \gamma) \sqcap \rho(D, \geq \gamma) \\ \rho(C \sqcap D, \leq \gamma) &= \rho(C, \leq \gamma) \sqcup \rho(D, \leq \gamma) \\ \rho(\forall R.C, \geq \gamma) &= \forall \rho(R, > 1 - \gamma). \rho(C, \geq \gamma) \\ \rho(\forall R.C, \leq \gamma) &= \exists \rho(R, \geq 1 - \gamma). \rho(C, \leq \gamma) \\ \rho(\exists R.C, \geq \gamma) &= \exists \rho(R, \geq \gamma). \rho(C, \geq \gamma) \\ \rho(\exists R.C, \leq \gamma) &= \forall \rho(R, > \gamma). \rho(C, \leq \gamma) \\ \rho(R, \geq \gamma) &= R_{\geq \gamma} \\ \rho(\langle a:C, \gamma \rangle) &= \{a:\rho(C, \geq \gamma)\} \\ \rho(\langle C \sqsubseteq D, n \rangle) &= \bigcup_{\alpha \in \bar{N}_+^K, \alpha \leq n} \{ \rho(C, \geq \alpha) \sqsubseteq \rho(D, \geq \alpha) \}\end{aligned}$$

Computational Complexity

The bad news...**undecidability!**

Proposition

Assume that fuzzy GCIs are restricted to be classical, i.e. of the form $\langle \alpha, 1 \rangle$ only. Then for the following fuzzy DLs, the KB satisfiability problem is undecidable over $[0, 1]$:

- 1 \mathcal{ELC} with classical axioms only under Łukasiewicz logic and product logic;
- 2 \mathcal{ELC} under any non Gödel-t-norm \otimes ;
- 3 \mathcal{ELC} with concept assertions of the form $\langle \alpha = n \rangle$ only under any non Gödel-t-norm \otimes ;
- 4 \mathcal{AL} with concept implication operator \rightarrow and concept assertions of the form $\langle \alpha = n \rangle$ only under any non Gödel-t-norm \otimes .
- 5 \mathcal{ELC} under SFL with weighted sum constructor.

Some decidability results..

Proposition

The KB satisfiability problem is decidable for

- *$SROIQ$ under SFL over $[0, 1]$ and Gödel logic over L_n*
- *$SROI\mathcal{N}$ under Łukasiewicz logic over L_n*
- *$S\mathcal{H}I$ under any continuous t -norm over L_n without $TBox$*
- *\mathcal{ALC} with concept implication operator \rightarrow , for any continuous t -norm over $[0, 1]$ with acyclic $TBox$*
- *$S\mathcal{H}IF$ with concept implication operator \rightarrow , for Łukasiewicz logic over $[0, 1]$ with acyclic $TBox$*
- *$S\mathcal{I}$ under any continuous t -norm over $[0, 1]$ without $TBox$*

Reasoners

Languages supported by fuzzy ontology reasoners:

Reasoner	Fuzzy DL	Logic	Degrees	Other constructors	GUI
fuzzyDL	$SHIF(\mathbf{D})$	Z, \perp	General	Modifiers, rough, aggregation	•
Fire	$SHIN$	Z	Numbers		•
FPLGERDS	ACC	\perp	Numbers	Role negatio/top/bottom	
YADLR	$ALCOQ$	Z, \perp	General	Local reflexivity	
DeLorean	$SROIQ(\mathbf{D})$	Z, G	General	Modifiers, rough DL	•
GURDL	ACC	General	Numbers		No
FRESG	$ACC(\mathbf{D})$	Z	Numbers	Fuzzy datatype expressions	•
LiFR	DLP fragment	Z	Numbers	Weighted concepts	
SMT-based solver	$AL\mathcal{E}$	Π	No	No	
DLMedia	$DLR-Lite$	Z, G	Numbers	Image similarity	•
SoftFacts	$DLR-Lite$	Z, G	Numbers	Fuzzy datatypes	•
ONTOSEARCH2	$DL - Lite_R$	General	Numbers		•

Reasoning services offered by fuzzy ontology reasoners

Reasoner	CON	ENT	CSAT	SUB	IR	BDB	Other tasks	OPT
fuzzyDL	•	•	•	•	•	•	Defuzzification	•
Fire	•	•	•	•		•	Classification	•
FPLGERDS		•						
YADLR		Partial			•	Partial	Realisation	
DeLorean	•	•	•	•		•		•
GURDL	•	•		•				•
FRESG	•	•	•		•		Realisation	
LiFR		Partial	•	•		•		
SMT-based solver			•					
DLMedia							Top-k Image Retrieval	•
SoftFacts							Top-k CQA	•
ONTOSEARCH2							Retrieval	

“CON”, “ENT”, “CSAT”, “SUB”, “IR”, “BED”, and “OPT” represent consistency, entailment, concept satisfiability, subsumption, instance retrieval, BED, and optimisations, respectively

That's it !