

Charting the Potential of Description Logic for the Generation of Referring Expression

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Yuan Ren, Kees van Deemter and Jeff Z. Pan

Department of Computing Science
University of Aberdeen, UK



Background

- Generation of Referring Expressions (GRE) algorithms *identify a target* referent:
- Express info known to be
true of target
false of all else
- But how does GRE model knowledge?

Knowledge Representation in classic GRE

- Sets of <Attribute, Value> pairs, e.g.
 - <Type, Poodle>
 - <Colour, Grey>
- These are *atomic* facts. Can't say
 - **All** poodles are grey
 - Kees is Dutch **or** Belgian
 - **If** x is part of y
and y is part of z
then x is part of z

Modern KR can do all this (and more)

- KL-One and semantic nets
- Modern descendants: Conceptual Graphs and Description Logic
 - Represent complex knowledge
 - Perform efficient automatic deduction
- Why not use modern KR for GRE?
- This talk: Description Logic (DL)
 - DL/OWL is now the language of the semantic web

Advantages

- A richer model of reference
 - New targets become identifiable
- Re-use
 - existing algorithms
 - existing ontologies
 - deduction for proving uniqueness
- Info represented *succinctly*, e.g.
 - a part of b , b part of c , c part of d , ...
 - **For all x,y,z : if x part of y and y part of z then x part of z**

Remainder of this talk

- Introduction to DL
- Areces et al. on DL and GRE
- Extending the expressivity
- Extending the algorithm

Caveat: few empirical claims

This talk is about *what's possible* in GRE

Description Logics?

- A family of logic-based KR formalisms
- Describe domain in terms of **concepts** (classes), **roles** (relations) and **individuals**
- Smallest propositionally closed DL is **ALC** (equivalent to $\mathbf{K}_{(m)}$)
 - Concepts constructed using \cup , \cap , \forall , \exists

E.g., domain elements that have a child who is a Doctor:

$$(\exists hasChild.Doctor)^{\mathcal{I}} = \{x | \exists y. \langle x, y \rangle \in hasChild^{\mathcal{I}} \text{ and } y \in Doctor^{\mathcal{I}}\}$$

DL Knowledge Base

- A **TBox** is a set of “schema” axioms (sentences), e.g.:

{ Dog \subset Animal, Dog \subset \neg Woman
hasFather \circ hasBrother \subset hasUncle, feed \circ love \subset feed }

– i.e., a **background theory**

- An **ABox** is a set of “data” axioms (ground facts), e.g.:

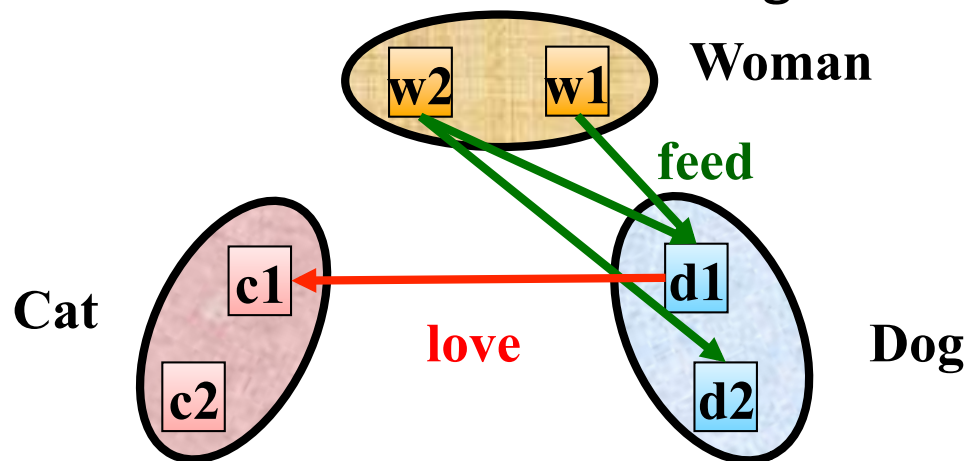
{d1:Dog, w1:Woman, c1:Animal,
(w1,d1):feed, (d1,c1):love
}

- An **NBox** (Negation as Failure Box) is a set of “**complete**” concepts and properties e.g.:

{Dog, Animal}

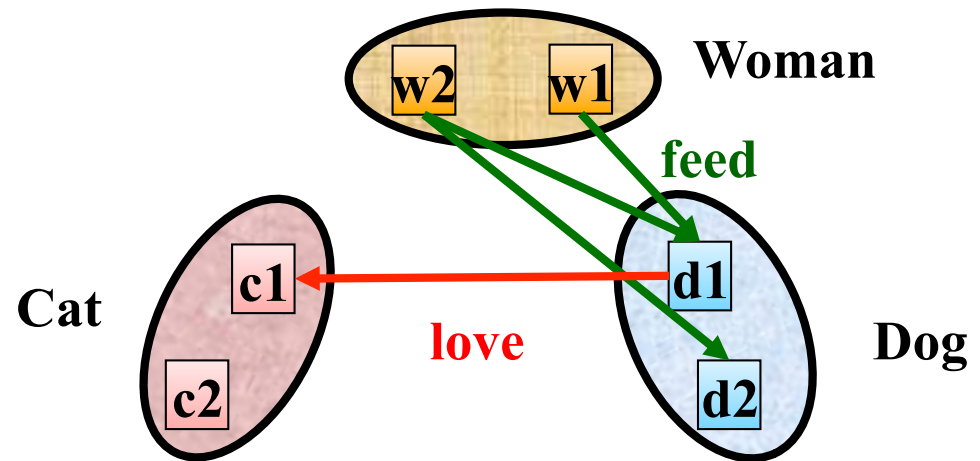
GRE example

- Uniquely identifying an object in context.
 - d1: the Dog that loves some Cat
 - w2: the Woman who feeds a Dog that doesn't love a Cat



From the DL point of view

- ABox assertion axioms
 - w1:Woman
 - w2:Woman
 - d1:Dog, d2:Dog
 - c1:Cat, c2:Cat
 - (w1,d1):feed
 - (w2,d2):feed
 - (w2,d1):feed
 - (d1,c1):love



DL for GRE: the story so far

- Areces et al. (2008) re-interpret GRE as a problem of computing **ALC** formulas.

\mathcal{L} -GRE PROBLEM

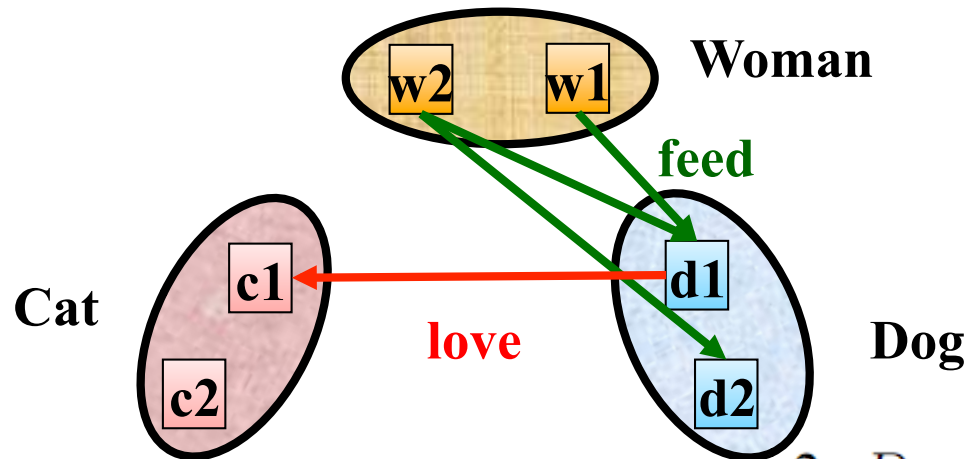
Input: A model \mathcal{M} and a target set $A \subseteq \Delta$.

Output: A formula $\varphi \in \mathcal{L}$ such that $\|\varphi\| = A$
(if such a formula exists).

- An algorithm to compute distinguishing REs (if one exists) for all objects:
 - Generate & Test strategy
 - Start from atomic concept names, then extend with (negative) **existential** quantifier.

DL for GRE example

- Referring Expression as a DL formula



$$\begin{aligned}
 1. \quad & Dog = \{d1, d2\}, \\
 & \neg Dog \sqcap Woman = \{w1, w2\}, \\
 & \neg Dog \sqcap \neg Woman = \{c1, c2\}.
 \end{aligned}$$

$$\begin{aligned}
 2. \quad & Dog \sqcap \exists love.(\neg Dog \sqcap \neg Woman) = \{d1\}, \\
 & Dog \sqcap \neg \exists love.(\neg Dog \sqcap \neg Woman) = \{d2\}.
 \end{aligned}$$

$$\begin{aligned}
 3. \quad & (\neg Dog \sqcap Woman) \sqcap \exists feed.(Dog \sqcap \\
 & \neg \exists love.(\neg Dog \sqcap \neg Woman)) = \{w2\}, \\
 & (\neg Dog \sqcap Woman) \sqcap \neg \exists feed.(Dog \sqcap \\
 & \neg \exists love.(\neg Dog \sqcap \neg Woman)) = \{w1\}.
 \end{aligned}$$

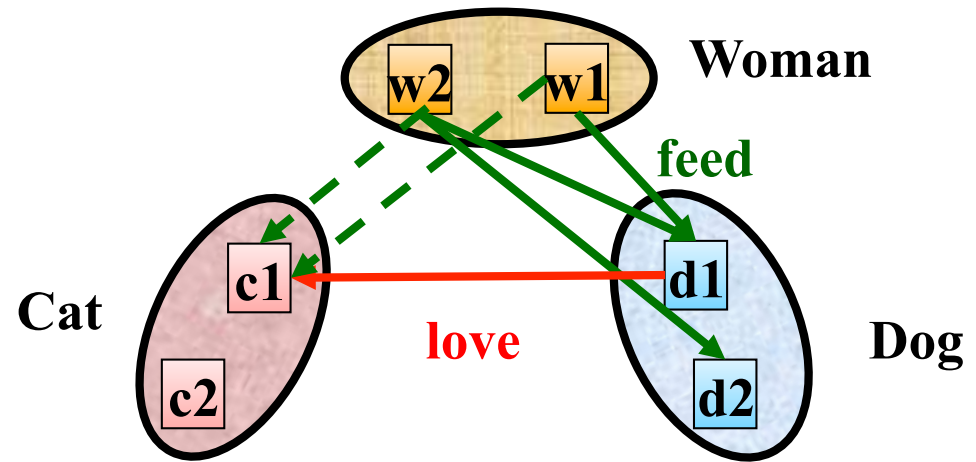
Logical foundations of GRE

DL forces us to make these explicit, allowing subtle distinctions

- GRE relies on a Unique Name Assumption (UNA)
 - If $\text{Dog}=\{d1,d2\}$ but $d1=d2$ then Dog refers uniquely
 - In DL, we can use UNA, or write $d1\neq d2$, etc.
- GRE relies on a closed world assumption (CWA)
 - In the earlier example, there are no more than two dogs, no more than two women, etc.
 - Without this, we would never be sure of uniqueness
 - DL: we can localise CWA, using an NBox

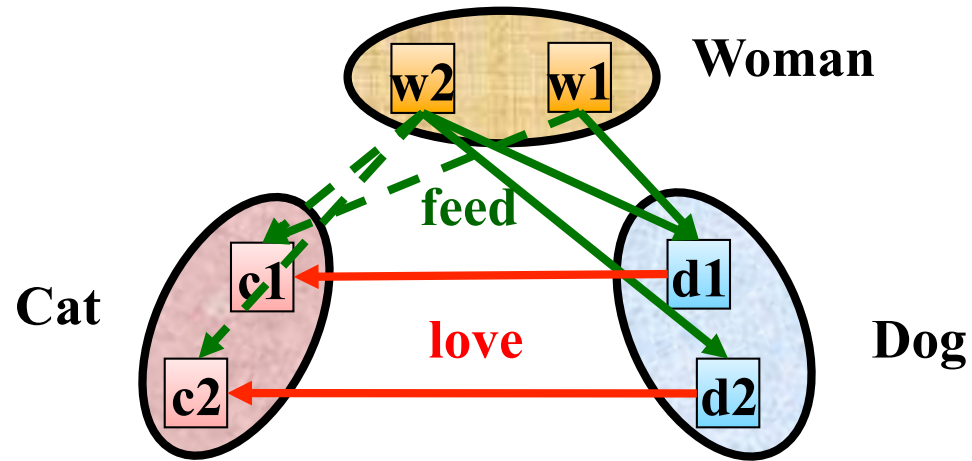
ABox Not Enough

- Tbox:
 - $\text{Woman} \sqsubset \neg \text{Dog}$
 - $\text{Dog} \sqsubset \neg \text{Cat}$
 - $\text{Cat} \sqsubset \neg \text{Woman}$
 - $\text{T} \sqsubset \text{Dog} \sqcup \text{Woman} \sqcup \text{Cat}$
 - $\text{feed} \circ \text{love} \sqsubset \text{feed}$



- Nbox:
 - $\{\text{Dog}, \text{Woman}, \text{Cat}, \text{love}, \text{feed}\}$

Additional Quantifiers



- Now Areces et al. cannot identify any of the 6 objects
- Additional quantifiers make them referable
 - c1: The cat which is fed by (at least) 2 women
 - w1: The woman feeding only those fed by at least 2 women
 - w2: The woman who feeds all dogs

Representing Quantifiers in Ontologies

We can use Generalised Quantifiers, e.g.,

“only”, “five” (5), “at least two” (≥ 2)

Example:

- English: *“The woman who loves at least two dogs”*
- Set theory + GQ: $\{y \in \text{Woman} : \geq 2 (\text{Dog}, \{z : \text{Love}(y, z)\})\}$

But which **quantifiers** exactly?

Proposal: use the numerical quantifiers

Table 1: Expressing GQ in DL

	QAB	DL
1	$\geq nN^2\{z : Ryz\}$	$y : \geq nR.N^2$
2	$\geq nN^2\neg\{z : Ryz\}$	$y : \geq n\neg R.N^2$
3	$\geq n\neg N^2\{z : Ryz\}$	$y : \geq nR.\neg N^2$
4	$\geq n\neg N^2\neg\{z : Ryz\}$	$y : \geq n\neg R.\neg N^2$
5	$\leq nN^2\{z : Ryz\}$	$y : \leq nR.N^2$
6	$\leq nN^2\neg\{z : Ryz\}$	$y : \leq n\neg R.N^2$
7	$\leq n\neg N^2\{z : Ryz\}$	$y : \leq nR.\neg N^2$
8	$\leq n\neg N^2\neg\{z : Ryz\}$	$y : \leq n\neg R.\neg N^2$

Example:

N^2 -- Dog

R -- Love

- Quantifiers of Type 1,
e.g., $n = 1$: the *existential* quantifier
- Quantifiers of Type 7,
e.g., $n = 0$: the *only* quantifier
- Quantifiers of Type 6,
e.g., $n = 0$: the *all* quantifier

Generating SROIQ-enabled REs

- GROWL: a GRE algorithm using OWL-2
 - Generate-and-test strategy
 - Using DL reasoning
 - Generating increasingly complex descriptions
 - Complexity measured by the structure of the expressions
 - 1 *Dog*
complexity of (negated) atomic concept is 1
 - 2 $\neg\textit{Dog} \sqcap \textit{Woman}$
complexity of conjunction (disjunction) is the maximal complexity of conjuncts (disjuncts) +1
 - 4 $\textit{Dog} \sqcap \neg\exists\textit{love} . (\neg\textit{Dog} \sqcap \neg\textit{Woman})$
complexity of existential (universal) restriction is the complexity of filler + 1

Generating SROIQ-enabled REs

- GROWL: a GRE algorithm using OWL-2
 - Starting from the basic terms, such as
 - Names (e.g. **Dog**, **feed**)
 - Inverse of named roles (such as **feed⁻**)
 - Negation of another basic term (e.g., **¬Dog**, **¬feed**, **¬feed⁻**)
 - For each existing term / description, create new description by
 - Extending with conjunction/disjunction
 - Extending with existential/universal restriction
 - Extending with cardinality restriction
 - Until no new non-empty description can be created

Generating SROIQ-enabled REs

- Deciding whether a new description should be accepted
 - Accept if no existing description has same extension
 - But still non-empty
- Using standard reasoning services
 - Concept **subsumption** checking
 - Concept **satisfiability** checking

Algorithm A-2: $Add(D, e)$

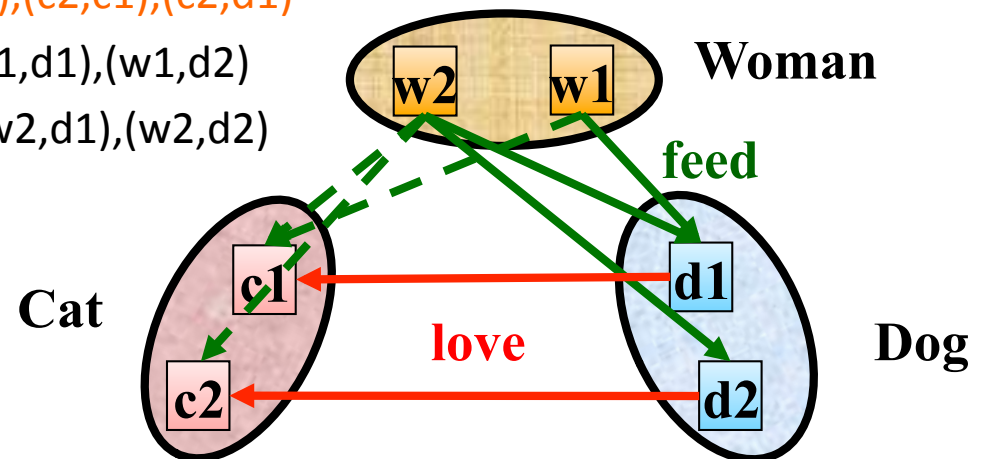
INPUT: D, e

OUTPUT: (Extended)Description Queue D

```
1: for  $d \in D$  do
2:   if  $Ins(d) = Ins(e)$  then
3:     return  $D$ 
4: if  $Ins(e) \neq \emptyset$  then
5:    $D := D \cup \{e\}$ 
6: return  $D$ 
```

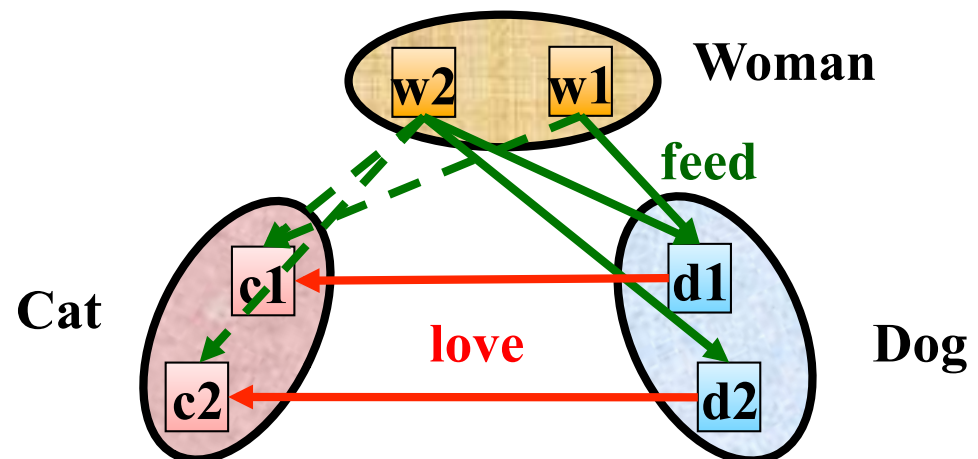
Example Revisited

- Starting from basic terms
 - $D = \{\text{Woman, Dog, Cat, } \neg \text{Woman, } \neg \text{Dog, } \neg \text{Cat, feed, feed}^{\neg}, \neg \text{feed}^{\neg} \text{ etc.}\}$
 - $\text{Ins}(\text{Woman}) = \{w1, w2\}$
 - $\text{Ins}(\text{Dog}) = \{d1, d2\}$
 - $\text{Ins}(\text{Cat}) = \{c1, c2\}$
 - $\text{Ins}(\neg \text{Woman}) = \{c1, c2, d1, d2\}$
 - ...
 - $\text{Ins}(\text{feed}) = \{(w1, d1), (w1, c1), (w2, d1), (w2, c1), (w2, d2), (w2, c2)\}$
 - $\text{Ins}(\text{feed}^{\neg}) = \{(d1, w1), (c1, w1), (d1, w2), (c1, w2), (d2, w2), (c2, w2)\}$
 - $\text{Ins}(\neg \text{feed}^{\neg}) = \{(d1, d2), (d1, c1), (d1, c2), (d1, d1), (d2, d1), (d2, d2), (d2, c1), (d2, c2), (d2, w1), (c1, c1), (c1, c2), (c1, d1), (c1, d2), (c2, c2), (c2, c1), (c2, d1), (c2, d2), (c2, w1), (w1, w1), (w1, w2), (w1, d1), (w1, d2), (w1, c1), (w1, c2), (w2, w1), (w2, w2), (w2, d1), (w2, d2), (w2, c1), (w2, c2)\}$



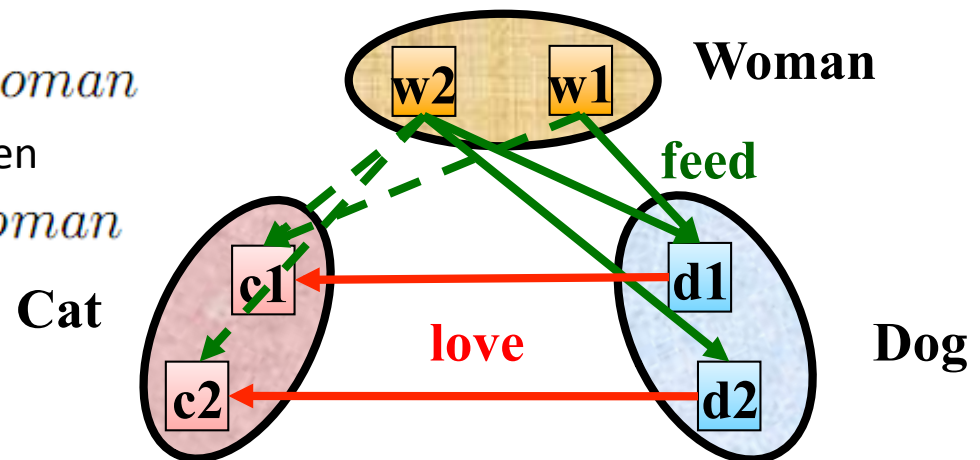
How to Refer to c1?

- Extending Woman with maximal Cardinality restriction
 $\leq 0(\neg \text{feed}^-). \text{Woman}$ (things that are fed by all women)
- $\text{Ins}(\leq 0(\neg \text{feed}^-). \text{Woman}) = \{\mathbf{c1}, \mathbf{d1}\}$. This is non-empty and not identical to any existing description
- Therefore, the new concept is added to D:
- $D = \{\text{Woman}, \text{Dog}, \text{Cat}, \neg \text{Woman}, \neg \text{Dog}, \neg \text{Cat}, \leq 0(\neg \text{feed}^-). \text{Woman}, \text{etc.}\}$
- Conjoining this with **Cat**:
 - $\text{Ins}(\text{Cat} \ \& \ \leq 0(\neg \text{feed}^-). \text{Woman}) = \{\mathbf{c1}\}$, c1 identified!



All objects become referable

- w1: the woman that does not feed all cats;
 $\{w1\} = Woman \sqcap \exists \neg feed.Cat$
- w2: the woman that feeds all cats;
 $\{w2\} = Woman \sqcap \leq 0 \neg feed.Cat$
- d1: the dog that is fed by all women;
 $\{d1\} = Dog \sqcap \leq 0 \neg feed^- .Woman$
- d2: the dog that is not fed by all women;
 $\{d2\} = Dog \sqcap \exists \neg feed^- .Woman$
- **c1: the cat that is fed by all women;**
 $\{c1\} = Cat \sqcap \leq 0 \neg feed^- .Woman$
- c2: the cat that is not fed by all women
 $\{c2\} = Cat \sqcap \exists \neg feed^- .Woman$



Conclusion (1)

We advocate using modern KR

- background knowledge should be considered
- expressing complex as well as atomic info
- reusing reasoning algorithms

Conclusion (2)

- Specific contributions:
 - Using DL reasoning to infer implicit knowledge (i.e., computing non-asserted information)
 - Generating REs taking into account incomplete knowledge
 - Extending the expressivity of GRE to OWL-2 (i.e., using new quantifiers, such as ≥ 2 feed.Dog)
 - Introducing Nbox for a partially closed world
- Open issues:
 - How useful are the newly generated REs?
 - How to choose the “best” RE?

Conclusion (2)

- Choosing the numerical quantifiers, have we used the “best” class?
- Note: GROWL generates descriptions of sets as well as individuals
- Description of sets will require new quantifiers
 - The dogs fed by between 3 and 7 women
- Some of these may even be non-first order, e.g.,
 - The dogs fed by most women
 - The dogs fed by an even number of women
- In finite domains, these were already expressible (though maybe less succinctly)

