using Maximum Entropy



Explaining default intuitions using maximum entropy

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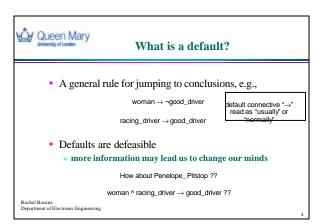




Outline

- · Defaults & default reasoning
- Maximum entropy & variable strength defaults
- The ME-solution and ranked model(s)
- How ME handles default intuitions
- Conclusion
 - » ME as normative default reasoning?
 - » Eliciting default information

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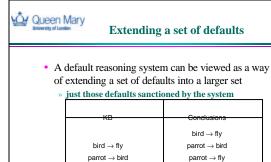
How to reason using defaults?

- Need a means to systematically manipulate a set of defaults to arrive at defeasible conclusions
 - » e.g. from the defaults

 $\begin{array}{c} \mathsf{bird} \to \mathsf{fly} \\ \mathsf{parrot} \to \mathsf{bird} \end{array}$

- » if we know an object is a parrot, we might conclude that, by virtue of it being a bird, it can fly
- » i.e. we infer another default parrot \rightarrow fly

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...etc...

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Systems of default reasoning

- Originally, many different approaches were taken
 - » McCarthy's circumscription
 - » Reiter's default logic
 - » Nonmonotonic logic, inheritance networks, conditional logics...
- · Later, attempts were made to standardise
 - » Gabbay & Makinson's inference rules for nonmonotonic systems
 - » Shoham's preferential logics
 - » Kraus, Lehman and Magidor's preferential reasoning
 - » also, preferential and rational consequence relations, rank-based systems, epsilon semantics, etc.

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Preferential reasoning

- Using this framework and applying a set of fairly uncontentious inference rules, a consensus was reached about minimal/basic default reasoning
 - inference rules of System P
- Many different semantics for System P all leading to the same conclusions
 - » preferential consequence relations, system P, probabilistic or e-semantics, possibilistic semantics
- · BUT these conclusions are extremely conservative
 - $\ \ \, \text{$>$ can't even infer} \quad \mathsf{parrot} \to \mathsf{fly}$

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Probabilistic or e-semantics

 Interprets a default as a statement of extremely high conditional probability, so bird →fly means

 $P(fly \mid bird) \geq 1 - e \qquad \text{ (or } \quad P(\neg fly \mid bird) \leq \ e \text{)}$

e > 0

- Sanctions conclusions that also have extremely high probability as $e \rightarrow 0$
 - » exactly the same conclusions as System

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How to capture stronger conclusions justifiably?

- · Use principle of maximum entropy (ME).
- Intuitively, the principle is simple
 - » model all that is known and assume nothing about that which is unknown
 - » given a set of defaults, choose a distribution that is consistent with all the defaults, but otherwise as uniform as possible
- ME-solution is both
 - » adventurous (so leads to more conclusions) and
 - » justifiable (since assumes no more than is known)

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Applying maximum entropy to default reasoning

- ε-semantics based on probability so can apply ME
- First done by Goldszmidt, Morris & Pearl (1993)
 - » derived unique ME solution + algorithm to find it in small class of problems
- Bourne & Parsons (1999) adjusted underlying semantics
 - » found more general solution that allows greater expressiveness of defaults + algorithm to find it
 - » multiple solutions in cases of redundancy

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Both cases, result in a ranked model (or models)

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New semantics: Adding strengths to defaults

Old semantics	New semantics	Differences
$a_i \rightarrow b_i$	$a_i \mathop{\to_{si}} b_i$	s _i is a (specified) relative strength
$P(b_i \mid a_i) \ge 1 - \epsilon$	$P(b_i \mid a_i) = 1 - O_i(\epsilon^{si})$	Constraints are equalities O _i is an (unspecified) convergence function
$P(\neg b_i \mid a_i) \le \varepsilon$	$P(\neg b_i \mid a_i) = O_i(\varepsilon^{si})$	

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Applying maximum entropy to default reasoning

• Find that distribution with maximum entropy subject to the constraints imposed by the defaults

 $H[P] = -\sum_{m} P(m) log P(m)$

- * As $e \rightarrow 0$, find the order of magnitude abstraction of this distribution
 - » the ME-solution or ME-ranked model

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Implication of assumptions

- By being vague about what the input defaults represent, we can nevertheless obtain a concrete and unique result in many cases
 - » i.e. can obtain firm results without considering the convergence functions themselves, just their order of magnitude
 - $\hspace{-1em} \hspace{-1em} \hspace{-1$
 - » AND since we assume the constraints are satisfied as equalities, it's possible they could be inconsistent and lead to no solution (not all strength assignments are valid)

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Problems with redundancy

- Troublesome cases are caused by defaults being redundant (i.e. ME-entailed by other defaults)
 - » in such a case, constraints may not be satisfied and model breaks down
- By identifying and ignoring redundant defaults the ME-solution can again be computed
 - » [though this can be subjective since under the revised semantics there may be several candidates for redundancy]

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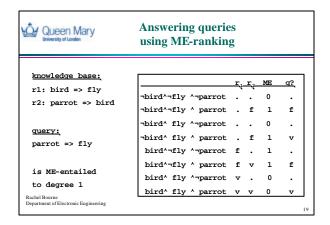
Unique ME-solutions

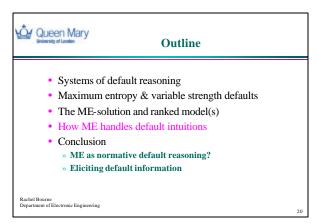
- When all goes well, obtain a unique ME-solution/ranked model, based on the relative strengths assigned to defaults.
 - » each default has an ME-rank (different from its strength)
 - » ranked model assigns a rank to each possible world being the sum of ranks of those defaults it falsifies (i.e. contradicts)
- ME-ranking determines defaults belonging to the extension
 a default is ME-entailed if its minimal verifying model is lower than minimal falsifying model
 - » also can determine degree of entailment (or strength)

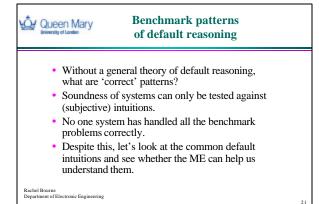
» some defaults ME-entailed under any strength assignment (we term this uncontroversial entailment)

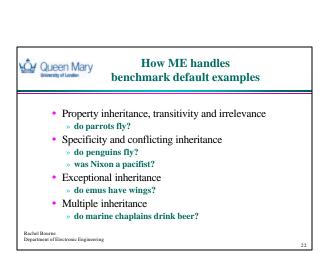
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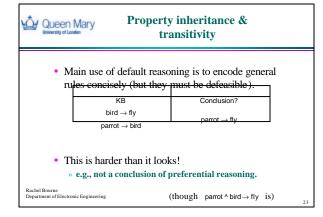
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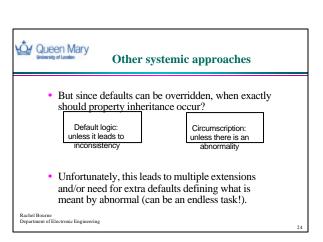




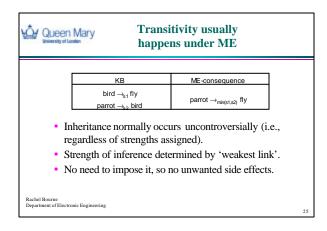


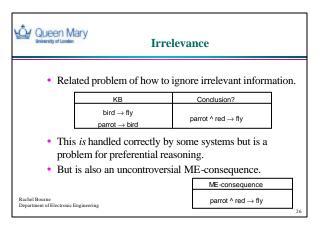


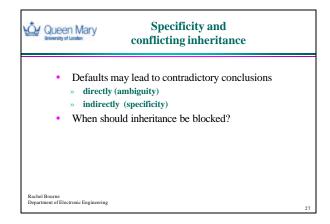


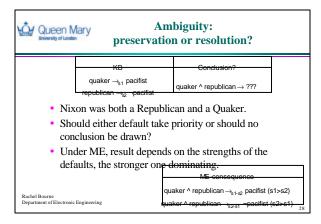


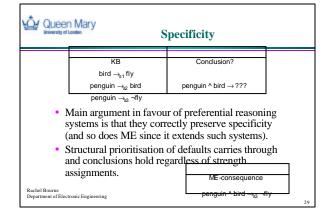
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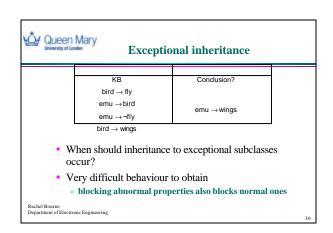












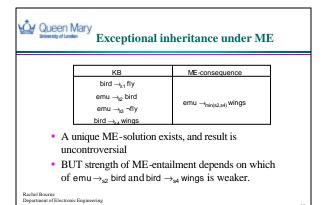
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Exceptional inheritance

- · Also, not easy to define behaviour required
 - » e.g. how exceptional can a subclass be before we doubt its classification rather than its inherited properties?
- Intuition here is that objects belonging to the same class should be similar in all features that define the class
 - » as many typical features as possible should be inherited

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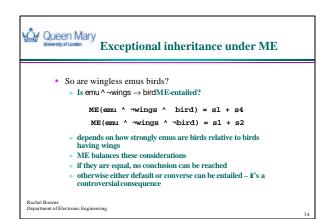
Exceptional inheritance under ME

 While ME gets it 'right' for this simple example, it throws up an interesting issue...

bird \rightarrow_{s1} fly emu \rightarrow_{s2} bird bird \rightarrow_{s4} wings

- » minimal falsifying model falsifies
 - either emu → bird
 - or bird → fly and bird → wings
- » which of these is relevant depends on strengths
- » emus without wings, while abnormal may be more normal than emus that are not birds

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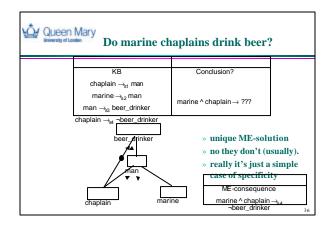




Multiple inheritance

- As seen in the 'Nixon Diamond', some cases involve inheriting from different sources leading to conflict and confusion.
- As problems get bigger, it is practically impossible to determine what the 'correct' inferences are.
- In the past this has lead to much debate and controversy...[cf. OO programming]
- The following example has been much discussed...

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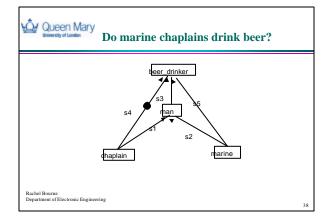


Do marine chaplains drink beer?

- But aren't marines usually heavy drinkers?
- · Need to model this by adding a new default...

KB	Conclusion?
chaplain → _{s1} man	
marine → _{s2} man	
$man \rightarrow_{s3} beer_drinker$	marine ^ chaplain → ???
$chaplain \rightarrow_{s4} \neg beer_drinker$	
marine → _{s5} beer_drinker	

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Do marine chaplains drink beer?

- Now the problem potentially contains redundancy
 - » if strength of new default is lower than it was previously ME-entailed, it can't be a constraint and therefore doesn't affect solution
 - » if same strength, multiple solutions possible
 - » if stronger, unique solution, but the query is now a controversial ME-consequence

```
ME(chaplain ^ marine ^ ¬beer_drinker) = s3 + s5 - min(s2,s3)
ME(chaplain ^ marine ^ beer_drinker) = s4
```

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ME as normative default reasoning?

- Main problem with default reasoning is that people can't always agree on what correct inferences are.
- ME matches intuitions in simple cases and demonstrates an intuitively acceptable way of examining complex ones.
- Can be used as a means of disambiguating and resolving arguments, since it only uses the information that has been supplied and not hidden semantics of propositions.

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Eliciting default information

- Remember that ME assumes just what is known and uniformly distributes rest of the uncertainty.
- So can be used to help clarify our own biases
 - » e.g. in multiple inheritance case, adding the extra default helped us to better understand the problem
- If we can't accept a particular inference sanctioned by ME, it may be that we are loading a proposition with extra semantic information
 - » can use ME to help design default knowledge bases

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