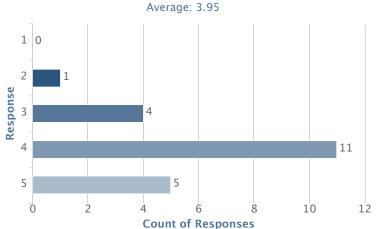
G53KRR SEM Feedback and Revision 2017

SEM feedback

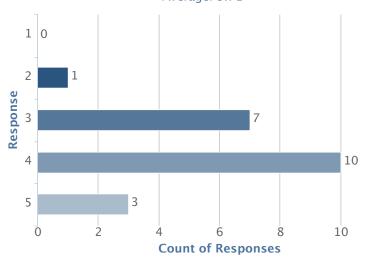
- No big issues raised
- I will go through SEM questions one by one

1) The module has provided me with opportunities to explore ideas or concepts in depth.



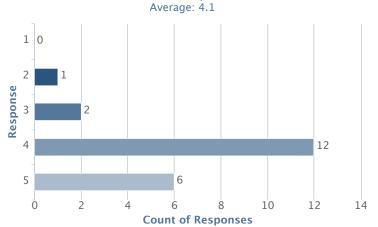
- I hope that you had an opportunity to study formal languages for knowledge representation in depth
- the fundamental concept of logical entailment
- the idea of logical reasoning/formal derivation

2) The module has challenged me to deliver my best work. Average: 3.71



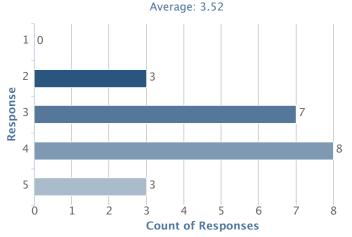
■ this is not easy to answer until the exam...

3) The module has been well organised and has been running smoothly.



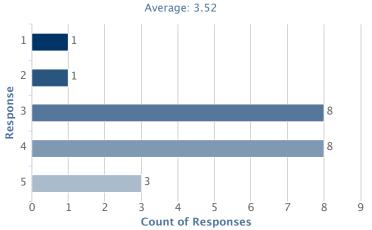
- did not get a text comment what was bad about organisation of the module
- would be grateful for feedback, anonymous or not

4) The resources in Moodle for this module have helped me to complete my work.



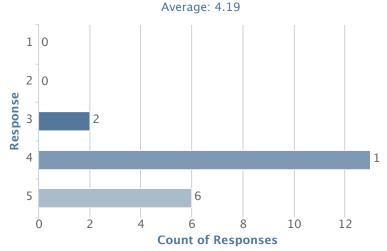
- there were clearly no resources on moodle apart from Logic in CS book
- there are however resources on the module web page, www.cs.nott.ac.uk/~psznza/G53KRR which is linked from moodle
- hope everyone found it it is essential for revision

5) The criteria used in marking my work have been made clear to me.



- exam papers are on the web
- sample model answers and marking schemes are on the web
- any correct answer will get full marks
- small mistakes will get very small mark reductions or none at all, if there is evidence of clear understanding
- if you want to do a sample exam/informal coursework and get my feedback, please email it to me

6) The workload on this module was reasonable for the number of credits



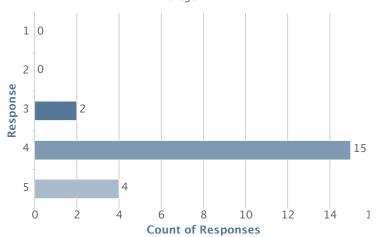
Q₆b

6b) Do you feel there was too much work involved or there was too little work involved?



Count of Responses





Exam format

- same as in 2014 2016: 4 questions out of 4
- previous papers and answers on the G53KRR module web page

Exam marking

- Answers to the exams published on the web are sample answers
- Often there are many different correct answers possible (for example, resoulution derivations)
- Any correct answer will get full marks
- Answers which have minor mistakes will have minor marks reductions (or sometimes no reduction, if the answer shows clear understanding)
- Big mistakes which suggest poor understanding will lose more marks

Exam topics = overview of the module

- First order logic
- Resolution
- Horn clauses, backward chaining, forward chaining
- Knowledge elicitation, lifecycle of a knolwedge based system
- Description logic
- Non-monotonic reasoning
- Bayesian networks

How to revise

- For each lecture, check recommended reading (on G53KRR web page)
- Do informal exercises (on G53KRR web page)
- Do exam questions from previous years (on G53KRR web page)
- Revision lecture in January?
- Send me topics for revision in advance
- Send me your exam attempts for feedback

Some common mistakes

■ Next I will go through some common mistakes, topic by topic

Common mistakes 1

First order logic question: show that S1, S2 do not logically entail S3

- Correct answer: describe an interpretation which makes S1 and S2 true and S3 false
- Don't:
 - use truth tables for first order sentences
 - attempt a resolution derivation of [] from S1, S2 and ¬S3, and then stop and say 'see, it does not work, so S3 is not entailed'

Common mistakes 2

Resolution

don't apply resolution to two literals at the same time:

MISTAKE:
$$\frac{[A,B], [\neg A, \neg B]}{[]}$$

it is not sound! $A \vee B$ and $\neg A \vee \neg B$ should not derive false.

• only substitute for variables (not constants or functional terms) Don't do f(x)/a or a/f(x).

MISTAKE :
$$\frac{[(P(f(x))], [\neg P(a)]}{[]}$$



Bayesian networks

- Directed acyclic graph
- Nodes: propositional variables; a directed edge from p_i to p_j if the truth of p_i affects the truth of p_i . p_i parent of p_i .

$$J(\langle P_1,\ldots,P_n\rangle)=Pr(P_1\wedge\ldots\wedge P_n)$$

Chain rule

$$Pr(P_1 \wedge \ldots \wedge P_n) = Pr(P_1) \cdot Pr(P_2|P_1) \cdot \cdots \cdot Pr(P_n|P_1 \wedge \ldots \wedge P_{n-1})$$

■ Independence assumption Each propositional variable in the belief network is conditionally independent from non-parent variables given its parent variables:

$$Pr(P_i \mid P_1 \land ... \land P_{i-1}) = Pr(P_i \mid parents(P_i))$$

where $parents(P_i)$ is the conjunction of literals which correspond to parents of p_i in the network.

Mistake 3

- Mistake: suppose a network consists of two variables, p_1 and p_2 , such that there is an edge from p_1 to p_2 . The mistake is to say that $Pr(p_1 \mid p_2) = Pr(p_1)$ because p_2 is not a parent of p_1 (so apply the independence assumption 'in reverse order of indices')
- This is a much more subtle (and understandable given the way the independenc assumption is stated) mistake.
- The independence assumption statement assumes that in the state description, the variables are listed in topological sort order (if there is an edge from p_i to p_j , then p_i appears before p_j in the order). This is always possible since the graph is acyclic. So we never check probability of parent conditioned on a child or a set of descendants.

Some revision slides

 Next I will go through some revision topics and previous exam answers

Circumscription

- The main idea: formalise common sense rules which admit exceptions.
- Rules like 'Birds fly' formalised as

$$\forall x (Bird(x) \land \neg Ab(x) \supset Flies(x))$$

- To check whether something is entailed by a knowledge base which contains such rules, we only check if it is entailed under the assumption that the set of exceptions *I*(*Ab*) is as small as possible
- This is called circumscription or minimal entailment

Example

$$KB = \{Bird(tweety), \forall x(Bird(x) \land \neg Ab(x) \supset Flies(x))\}$$

- Classically, KB \(\notin \) Flies(tweety) because there are interpretations of KB where Tweety is an exceptional bird (it is in I(Ab)) and it does not fly
- But such interpretations do not minimise the set of exceptions: nothing which is said in KB forces us to think that Tweety is exceptional, so it does not have to be in I(Ab)
- If we only consider interpretation which satisfy KB and where the set of exceptions is as small as possible, Tweety is not in this set, so Bird(tweety) ∧ ¬Ab(tweety) holds and hence Flies(tweety) holds
- KB entails Flies(tweety) on 'minimal' interpretations where I(Ab) is circumscribed (made as small as possible)

Definition of minimal entailment

- Let $M_1 = (D, I_1)$ and $M_2 = (D, I_2)$ be two interpretations over the same domain such that every constant and function are interpreted the same way.
- $\blacksquare M_1 \leq M_2 \Leftrightarrow I_1(Ab) \subseteq I_2(Ab)$
- $M_1 < M_2$ if $M_1 \le M_2$ but not $M_2 \le M_1$. (There are strictly fewer abnormal things in M_1).
- Minimal entailment: $KB \models_{\leq} \alpha$ iff for all interpretations M which make KB true, either $M \models \alpha$ or M is not minimal (exists M' such that M' < M and $M' \models KB$).

Back to the example

- $KB = \{Bird(tweety), \forall x(Bird(x) \land \neg Ab(x) \supset Flies(x))\}$
- $KB \models_{\leq} Flies(tweety)$ because for every interpretation M which makes KB true and Flies(tweety) false, it has to be that $I(tweety) \in I(Ab)$.
- So for for every such interpretation there is an interpretation M' which is just like M, but $l'(tweety) \notin l'(Ab)$ and $l'(tweety) \in l'(Flies)$, and M' still makes KB true and M' < M.

Defaults

- A default rule consists of a prerequisite α , justification β , conclusion γ and says 'if α holds and it is consistent to believe β , then believe γ ': $\frac{\alpha:\beta}{\gamma}$
- For example:

$$\frac{Bird(x):Flies(x)}{Flies(x)}$$

■ Default rules where justification and conclusion are the same are called *normal default rules* and are writted $Bird(x) \Rightarrow Flies(x)$.

Default theories and extensions

- A default theory *KB* consists of a normal first-order knowledge base *F* and a set of default rules *D*
- A set of reasonable beliefs given a default theory $KB = \{F, D\}$ is called an *extension* of KB
- E is an *extension* of (F, D) iff for every sentence π ,

$$\pi \in E \Leftrightarrow F \cup \{\gamma \mid \frac{\alpha : \beta}{\gamma} \in D, \alpha \in E, \neg \beta \notin E\} \models \pi$$

How one could construct an extension

$$\pi \in \mathcal{E} \iff \mathcal{F} \cup \{\gamma \mid \frac{\alpha : \beta}{\gamma} \in \mathcal{D}, \alpha \in \mathcal{E}, \neg \beta \not\in \mathcal{E}\} \models \pi$$

- 1 E := F;
- 2 close *E* under classical entailment: $E := \{\pi : E \models \pi\}$
- 3 choose some (substitution instance of) $\frac{\alpha:\beta}{\gamma}\in D$
- 4 if $\alpha \in E$, and $\neg \beta \notin E$ (meaning, β is consistent with E), $E := E \cup \{\gamma\}$
- 5 go back to 2



Example

$$F = \{Bird(tweety)\}, D = \{\frac{Bird(x) : Flies(x)}{Flies(x)}\}\$$

- *E* := {*Bird*(*tweety*)}
- close *E* under classical entailment: $E := \{\pi : Bird(tweety) \models \pi\}$
- $\blacksquare \frac{\textit{Bird(tweety):Flies(tweety)}}{\textit{Flies(tweety)}} \in D$
- $Bird(tweety) \in E$, and $\neg Flies(tweety) \notin E$ $E := E \cup \{Flies(tweety)\}$
- \blacksquare $E := \{\pi : Bird(tweety), Flies(tweety) \models \pi\}$
- there are no more rules to apply



Example from 2008 exam, Q6e

$$F = \{ Dutchman(peter), Dutchman(hans), Dutchman(johan), \\ peter \neq hans, hans \neq johan, peter \neq johan, \\
\neg Tall(peter) \lor \neg Tall(hans) \} \\ D = \{ \frac{Dutchman(x) : Tall(x)}{Tall(x)} \}$$

Three instances of the default rule:

Exam 2008 example continued

- Suppose we start constructing E_1 with the first rule, for Peter. Since $\neg Tall(peter) \notin E_1$, we can add Tall(peter) to E_1 .
- After we close E_1 under consequence, from Tall(peter) and $\neg Tall(peter) \lor \neg Tall(hans)$ we get $\neg Tall(hans) \in E_1$.
- So now the second rule for Hans is not applicable.
- The third rule is applicable, since $\neg Tall(johan) \notin E_1$, we can add Tall(johan) to E_1
- Another possible extension is E_2 : we use the second rule first, and add Tall(hans) to E_2 .
- Now the first rule is not applicable, because E_2 contains $\neg Tall(peter)$
- The third rule is applicable, since $\neg Tall(johan) \notin E_2$, we can add Tall(johan) to E_2



Another example with two extensions

- Facts: $F = \{Republican(dick), Quaker(dick)\}$
- Default rules: $Republican(x) \Rightarrow \neg Pacifist(x)$, $Quaker(x) \Rightarrow Pacifist(x)$.
- Extension E_1 (pick the rule $Republican(x) \Rightarrow \neg Pacifist(x)$ first) is all consequences of $\{Republican(dick), Quaker(dick), \neg Pacifist(dick)\}$. Because we start with $E_1 = \{Republican(dick), Quaker(dick)\}$, $\neg \neg Pacifist(dick) \notin E_1$, so we can add $\neg Pacifist(dick)$ to E_1 .
- Extension E₂ (pick the rule Quaker(x) ⇒ Pacifist(x) first) is all consequences of {Republican(dick), Quaker(dick), Pacifist(dick)}. Because we start with E₂ = {Republican(dick), Quaker(dick)}, ¬Pacifist(dick) ∉ E₂, so we can add Pacifist(dick) to E₂.

Example with one extension

- Facts: $F = \{Republican(dick), Quaker(dick), \forall x (Republican(x)) \supset MemberOfPoliticalParty(x))\}$
- Default rules: $Republican(x) \Rightarrow \neg Pacifist(x)$,

$$\frac{\textit{Quaker}(x): \textit{Pacifist}(x) \land \neg \textit{MemberOfPoliticalParty}(x)}{\textit{Pacifist}(x)}$$

- Closure of F under consequence includes: {Republican(dick), Quaker(dick), ∀x(Republican(x) ⊃ MemberOfPoliticalParty(x)), MemberOfPoliticalParty(dick)}
- The second default rule is not applicable, because $\neg\neg MemberOfPoliticalParty(dick) \in E$
- only the first rule is applicable, since $\neg\neg Pacifist(dick) \notin E$, so $\neg Pacifist(dick)$ is added.

Any questions?