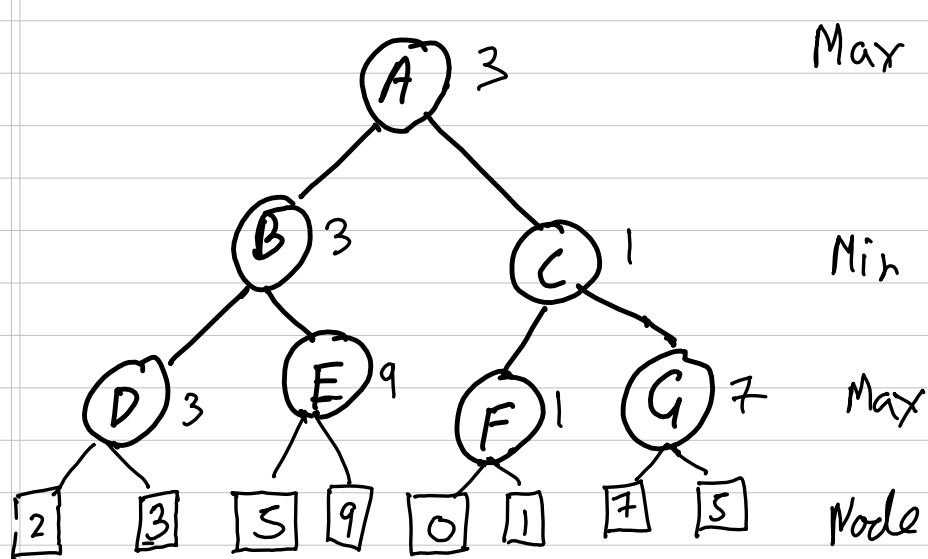
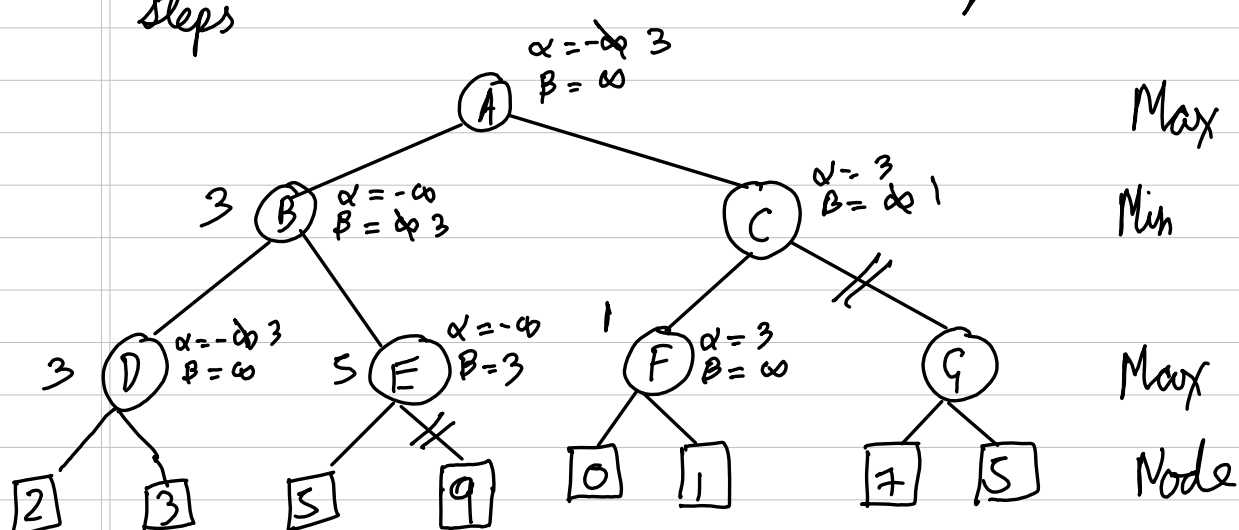


① Min Max \rightarrow Adversarial search



α -B - Pruning
Used to reduce the number of steps



At D $\alpha = -\infty$
 $\beta = \infty$

Comparing 2, $\alpha < 2 \therefore$ update $\alpha = 2$

$\alpha = 2$ $\alpha \geq \beta \rightarrow \text{False}$
 $\beta = \infty$

Comparing 3 $\alpha < 3 \therefore$ update $\alpha = 3$

$\alpha = 3$ $\alpha \geq \beta \rightarrow \text{False}$
 $\beta = \infty$

Path to D $\rightarrow 3$

At B $\alpha = -\infty$
 $\beta = \infty$

Compare D $\beta > 3 \therefore$ update β

$\alpha = -\infty$ $\alpha \geq \beta \rightarrow \text{false}$
 $\beta = 3$

At E $\alpha = -\infty$ } Passed from parent
 $\beta = 3$ $\alpha \geq \beta \rightarrow \text{false}$

Compare 5

$\alpha < 5 \therefore$ update 5

$\alpha = 5$ $\alpha \geq \beta \rightarrow \text{True}$
 $\beta = 3$

\therefore Prune branch to 9
Cost to E = 5

At B Backtracking \rightarrow

$\beta < 5 \therefore$ No update
Cost to B = 3

At A Backtracking

$\alpha < 3 \therefore$ update $\alpha = 3$

$\alpha = 3$ $\alpha \geq \beta \rightarrow \text{False}$
 $\beta = \infty$

At C, $\alpha = 3$ } Pass to child
 $\beta = \infty$

At F, $\alpha = 3$ } Pass to child
 $\beta = \infty$

Compare with 0, $\alpha > 0 \therefore$ No update
Compare with 1, $\alpha > 1 \therefore$ No update

$\alpha \geq \beta \rightarrow \text{False}$

Maximum value at F

0, 1 \rightarrow 1

Backtrack to C

$\beta > 1 \therefore$ update $\beta = 1$

$\alpha = 3$ $\alpha \geq \beta$
 $\beta = 1$ \therefore Prune

② Adversarial Search

Enemy, opponent changing state of problem in every step

Alternating Min & Max levels (Self & opponent)

Used to play games like checkers, chess

MinMax algorithm \rightarrow complexity $O(b^d)$

α - β pruning \rightarrow Reduce complexity $O(\sqrt{b^d})$
Prune certain nodes

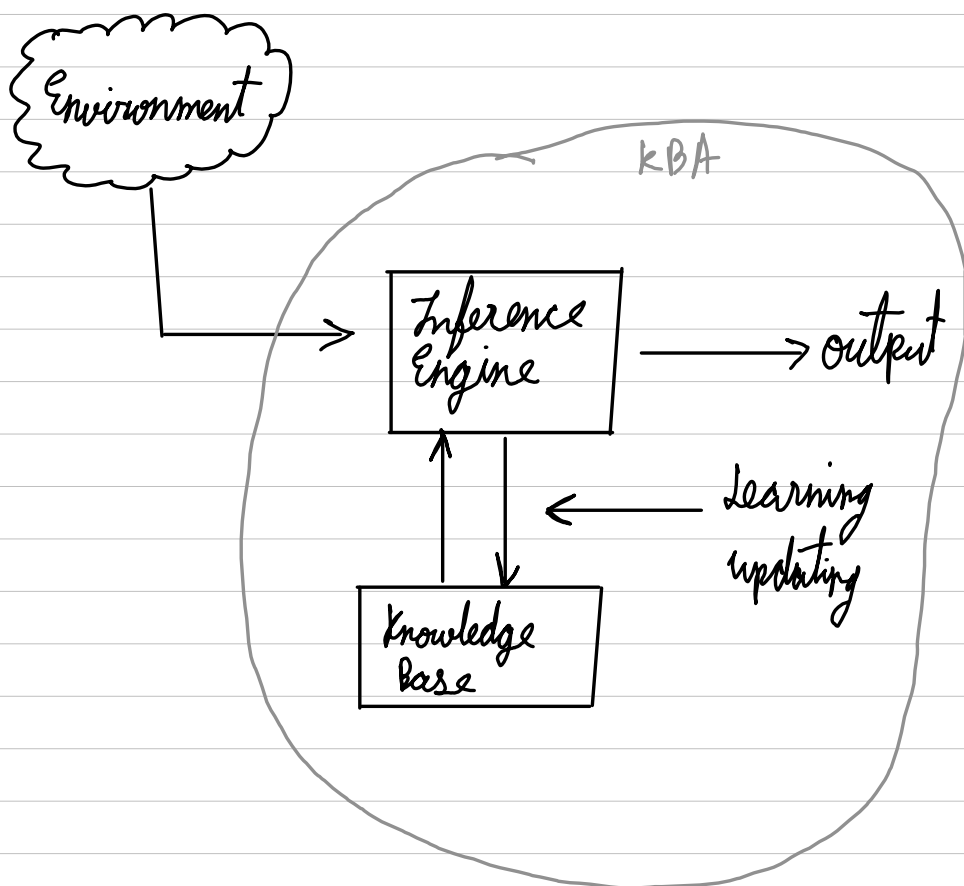
Condition $\alpha \geq \beta$

$\alpha \rightarrow$ Max } Path from current node
 $\beta \rightarrow$ Min } to root Node

③ Knowledge Based Agents

→ Agents that rely on knowledge & reasoning to take decisions over simple reflexes

They make deductions
Decisions
Conclusions } Based on knowledge base



Knowledge based agents have the ability to

- T — Take actions
- R — Reason over knowledge
- U — Update knowledge
- M — Maintain internal knowledge base

Represent the world with some formal representation and act intelligently

Three operations performed by KBA in order to show intelligent behaviour

- T — Tell the knowledge base the situation
- A — Ask what to Do
- P — Perform the selected Action

④ Wumpus World

P - Navigate the world kill Wumpus, get gold & return safely

+1000 getting gold
+100 killing a Wumpus

-10 using up arrow
-1000 getting eaten by Wumpus
-1000 falling in pit
-1 every Move taken

Game ends when agent climbs out with gold or when agent dies

E - Deterministic
Single Agent
Static
Discrete
Partially Observable
Known
Sequential



4x4 Grid rooms

Random locations for Wumpus & gold & pit

A - Move left
Move right
Move up
Move Down
shoot arrow Left
Right
up
Down

S - feel Breeze
feel stretch
feel glitter
feel Bump (On wall)
feel scream (when Wumpus is killed)

Prove that Wumpus is in room (1,3) using FOL

SSS stretch		$\neg B$	$\neg P$
 Wumpus	$\approx B$ $\approx S$ $\approx P$	Pit	$\neg P$
SSS stretch		$\neg B$	
	$\neg B$	P	$\neg B$

Rules

$\neg S_{11} \rightarrow \neg W_{12} \wedge \neg W_{21}$ ①
 $S_{12} \rightarrow W_{11} \vee W_{22} \vee W_{13}$ ②
 $\neg S_{21} \rightarrow \neg W_{11} \wedge \neg W_{22} \wedge \neg W_{31}$ ③

from ① using AND elimination

$\neg S_{11} \rightarrow \neg W_{12}$
 $\neg S_{11} \rightarrow \neg W_{21}$ } Apply Modus Ponens
 $\neg S_{11}$
 $\neg W_{12} \quad \neg W_{21}$

from ③ similarly

$\neg W_{11} \quad \neg W_{12} \quad \neg W_{31}$

from ② Modus Ponens

$W_{11} \vee W_{12} \vee W_{13}$ ④

Unit resolution ④ & $\neg W_{11}, \neg W_{12}$

$\therefore W_{13}$ obtained

⑤ Limitations of PL logic

cannot represent states like ALL, some or none

cannot model real world situations

limited expressive power

cannot handle uncertainty

⑥ Hill Climbing Problem

Continuously Move in direction of increasing value

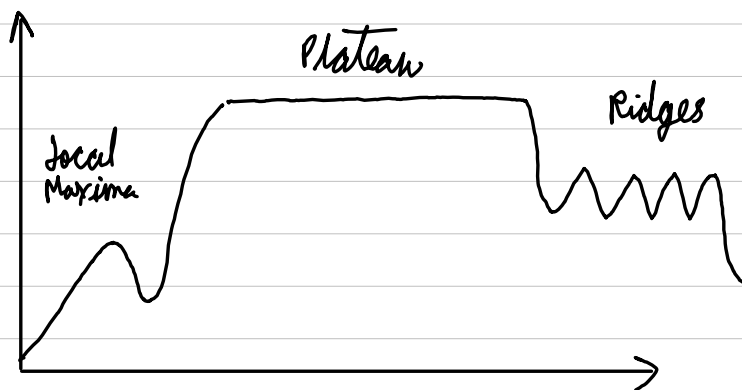
Terminate when peak value is reached

Does not maintain search tree. Only current state & value of objective function

Dont look beyond immediate neighbours

Heuristic local search algorithm

Disadvantages → Local Maxima reached
∴ Cant Navigate ridges, plateaus



start with a initial solution and take small steps to choose best neighbours

⑦ Convert to FOL

- | | | |
|---|------------------------|--------------------|
| ① | Fred is a colie | Colie (Fred) |
| ② | Sam is Freds Master | Master (Sam, Fred) |
| ③ | Day is saturday | Day (saturday) |
| ④ | It is cold on saturday | Cold (saturday) |
| ⑤ | Fred is trained | Trained (Fred) |

⑥ Trained colies are good dogs $\forall x \text{ Colie}(x) \wedge \text{Trained}(x) \rightarrow \text{good dog}(x)$

⑦ If a dog is a good dog and has a master at some place then he will be at that place

$\forall x \forall y \forall z \quad \text{good dog}(x) \wedge \text{Master}(y, x) \wedge \text{At}(y, z) \rightarrow \text{At}(x, z)$

⑧ If day is saturday & day is cold then sam is at Museum

$\text{Day}(\text{saturday}) \wedge \text{Day}(\text{cold}) \rightarrow \text{At}(\text{Sam}, \text{Museum})$

⑧ Prove by Resolution

Fred is at Museum

Writing in CNF

$$\begin{aligned} & \sim (\text{Trained}(x) \wedge \text{Collie}(x)) \vee \text{GoodDog}(x) \\ & \text{DeMorgan's Law} \\ & \equiv \sim \text{Trained}(x) \vee \sim \text{Collie}(x) \vee \text{GoodDog}(x) \quad (9) \end{aligned}$$

Similarly

$$\begin{aligned} & \sim \text{Day}(\text{Saturday}) \vee \sim \text{Cold}(\text{Saturday}) \\ & \vee \text{At}(\text{Sam}, \text{Museum}) \quad (10) \end{aligned}$$

$$\sim \text{GoodDog}(i) \vee \sim \text{Master}(j, i)$$

$$\vee \sim \text{At}(j, z) \vee \text{At}(i, z) \quad (11)$$

Assume Fred is Not at Museum

$$(11) \quad \sim \text{At}(\text{Fred}, \text{Museum})$$

i | Fred
2 | Museum

Resolution

$$a \vee \sim b, \sim b \therefore a$$

$$\sim \text{GoodDog}(\text{Fred}) \vee \sim \text{Master}(j, \text{Fred})$$

$$\vee \sim \text{At}(j, \text{Museum})$$

(10)

j | Sam

Resolution

$$\sim \text{Day}(\text{Saturday}) \vee \sim \text{Cold}(\text{Saturday})$$

$$\vee \sim \text{GoodDog}(\text{Fred}) \vee \sim \text{Master}(\text{Sam}, \text{Fred})$$

③, ④, ②

Resolution

(9)

$$\sim \text{GoodDog}(\text{Fred})$$

$$x(\text{Fred})$$

$$\sim \text{Trained}(\text{Fred}) \vee \sim \text{Collie}(\text{Fred})$$

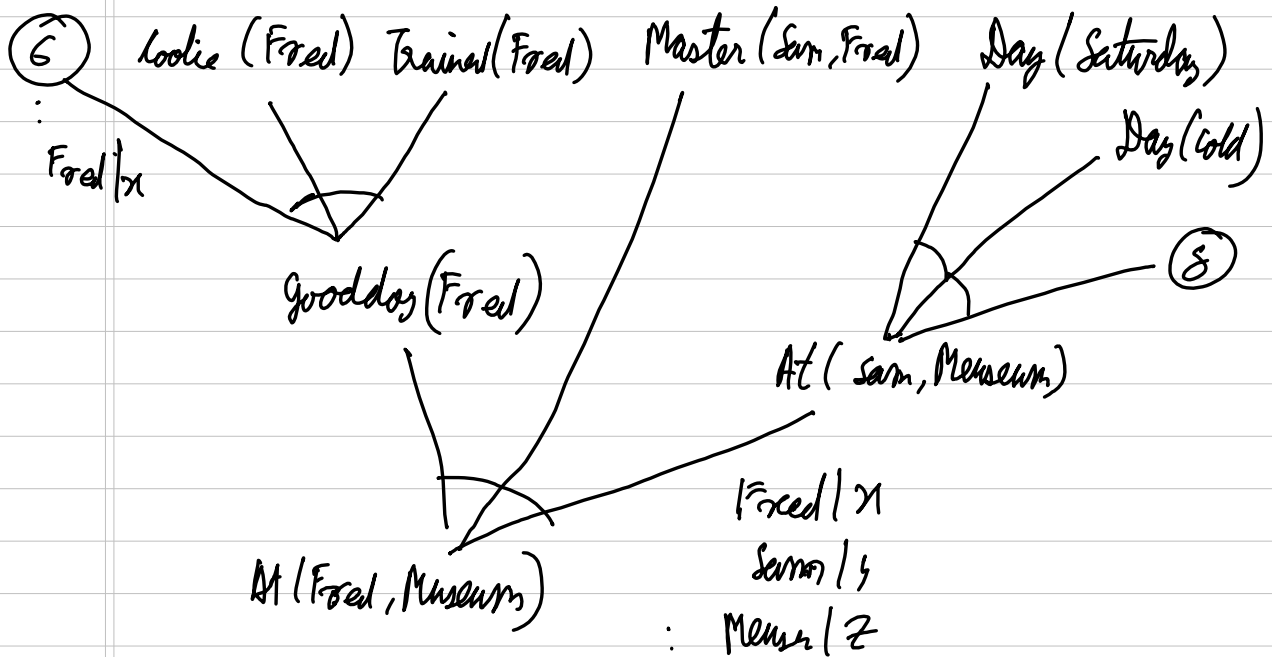
(1) ⑤

Resolution

F

Hence proved

⑨ Prove by forward / Backward chaining



(10)

Forward chaining vs Backward Chaining

Forward chaining	Backward Chaining
Start with known facts and proceed towards the goal	start from the goal & proceed towards known facts
Bottom up	Top down
Breadth first	Depth first
Any conclusion	Only required data
Planning & control	Diagnosis & debugging
∞ no of conclusions	Only finite conclusion

⑪ What is Uncertainty in AI & how to solve it?

→ In real world, there are a lot of scenarios where certainty of something is not confirmed.

In order to represent uncertain knowledge where we are not sure about the predicates, we need uncertain reasoning or probabilistic reasoning

When there are unpredictable outcomes
too large probabilities

Unknown errors / events

In order to tackle uncertainty, probabilistic reasoning is used.

Probabilistic models represent the relationships between events in a probabilistic way

Then inference uses the probability to take decisions.

⑫

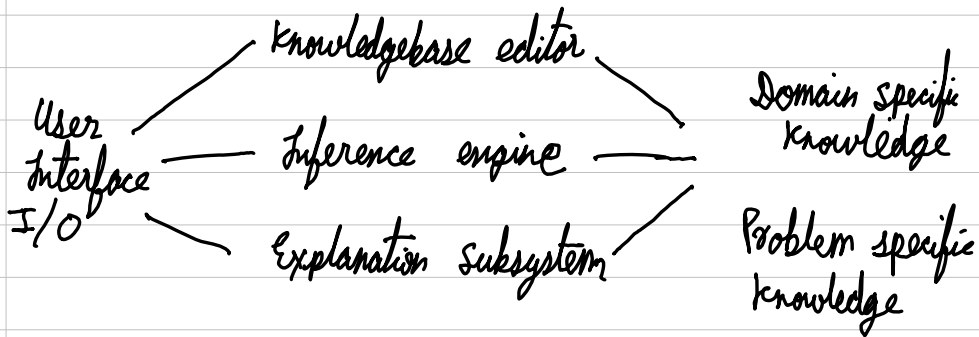
ADL vs STRIPS

ADL	STRIPS
Action description language	Stanford Research Institute Problem Solver
Allows Negative literals	Negative literals not allowed
Open world Assumption Unknown literals are unknown	Closed world assumption Unknown literals are false
$P \wedge \sim Q$ means add P , $\sim Q$ delete $\sim P$ & Q	$P \wedge \sim Q$ means add P delete Q
Goals are conjunctions & disjunctions $A \wedge (B \vee C)$	Only conjunctions is goals $(A \wedge B)$
Type support $x = y$	No type support
equality support P: plane	No equality support

⑬ Explain expert systems with diagram

→ Program that implements knowledge & reasoning process of a human expert.

"An intelligent computer program that employs knowledge and inference procedures to solve problems that are considered difficult enough to require significant human expertise for their solutions"



They are efficient, accurate & solve problems like human experts

Limitations of expert systems —

Can't handle unforeseen situation
High Cost
Difficult to maintain
No creativity

① STRIPS for Air Cargo Problem



A₁



A₂

Move C₁ from A₁ to A₂
Move C₂ from A₂ to A₁

Init (At (C₁, A₁), At (C₂, A₂),
At (P₁, A₁), At (P₂, A₂),
Cargo (C₁), Cargo (C₂), Airport (A₁),
Plane (P₁), Plane (P₂), Airport (A₂))

Goal (At (C₁, A₂), At (C₂, A₁))

Actions (load (C, P, a)

Precondition : Cargo (C) ∧ Plane (P) ∧ Airport (a)
∧ ~ In (C, P)

Postcondition : In (C, P) ∧ ~ At (C, a)

Fly (P, from, to)

Precond : Plane (P) ∧ Airport (from) ∧ Airport (to)
∧ at (P, from)

Postcond : ~ at (P, from) ∧ at (P, to)

Unload (P, C, a)

Precond : In (C, P) ∧ Cargo (C) ∧ Plane (P) ∧
airport (P)

Postcond : at (C, a) ∧ ~ In (C, P)

⑮ POP for air cargo

→ Goal: $at(C_1, a_2) \wedge at(C_2, a_1)$

Consider action $Unload(C_1, P_1, a_2)$

Post cond: $at(C_1, a_2) \wedge \neg h(C_1, P_1)$

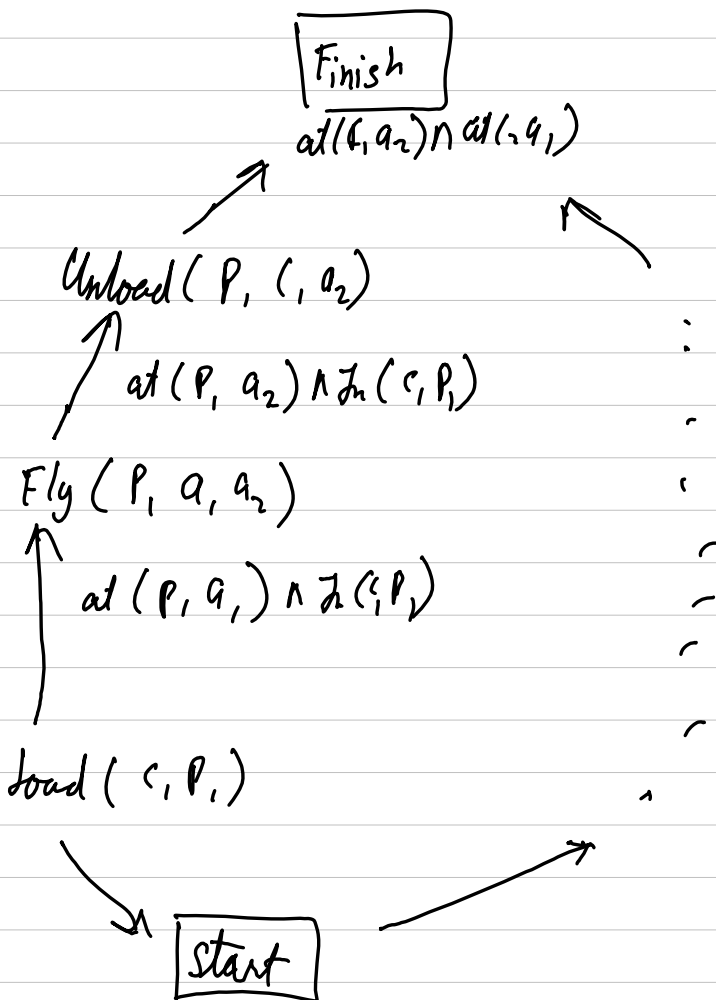
Precond: $h(C_1, P_1) \wedge at(P_1, a_1), n \dots$ Type

Consider action $Fly(P_1, a_1, a_2)$

Post cond...

Precond ..

⋮



⑩ POP vs TOP

POP	TOP
Parallel execution of action sequences	Sequential execution of action
No specific order of actions	Exact ordering of actions
Single graph obtained	Multiple orderings obtained

⑪ Inference rules in FOL

Universal instantiation

$$\frac{\forall x P(x)}{P(c)}$$

if all people are donkeys then c is a donkey

Universal Generalization

$$\frac{P(c)}{\forall x P(x)}$$

$P(c)$ is true for any c then $\forall x P(x)$

Any cat is a animal \rightarrow all cats are animals
A byte has 8 bits \rightarrow all bytes have 8 bits

Existential instantiation

$$\frac{\exists x P(x)}{P(c)}$$

There exists an intelligent monkey $\exists x P(x)$

let monkey be named c $P(c)$

Existential Generalization

$$\frac{P(c)}{\exists x P(x)}$$

Tinky is an intelligent monkey $P(c)$

\therefore There exists an intelligent monkey $\exists x P(x)$

⑮ Inference rules for knowledge reasoning

If you study you will get good marks

Modus Ponens

$$\frac{P \rightarrow Q, P}{\therefore Q}$$

You studied

\therefore You have got good marks

Modus Tollens

(Law of contrapositive)

$$\frac{P \rightarrow Q, \sim Q}{\therefore \sim P}$$

You didn't get good marks

\therefore you haven't studied

And elimination

$$\frac{a \wedge b}{a}$$

You are smart & cute

\therefore you are smart can be inferred

Bidirectional elimination

$$\frac{(a \rightarrow b) \wedge (b \rightarrow a)}{a \leftrightarrow b}$$

$$\frac{a \leftrightarrow b}{(a \rightarrow b) \wedge b \rightarrow a}$$

Resolution

$$\frac{P \vee Q, \sim Q \vee R}{\therefore P \vee R}$$

Unit resolution

$$\frac{P \vee Q, \sim Q}{\therefore P}$$

①⑨ Process of building expert system

i) Problem identification

ii) Conceptualization

iii) Formalization

iv) Implementation

v) Testing

② steps of NLP

→

- 1) Segmentation
- 2) Tokenization
- 3) stemming
- 4) Lemmatization
- 5) Identifying stop words
- 6) Dependency parsing
- 7) POS tagging
- 8) Named entity Recognition
- 9) chunking

(21)

Unification & Lifting

Unification is the process of making two different logical expressions identical by finding a substitution

$\text{king}(x)$, $\text{king}(\text{John})$

$x \mid \text{John}$

Unify algorithm takes in two atomic sentences and returns a unifier if it exists