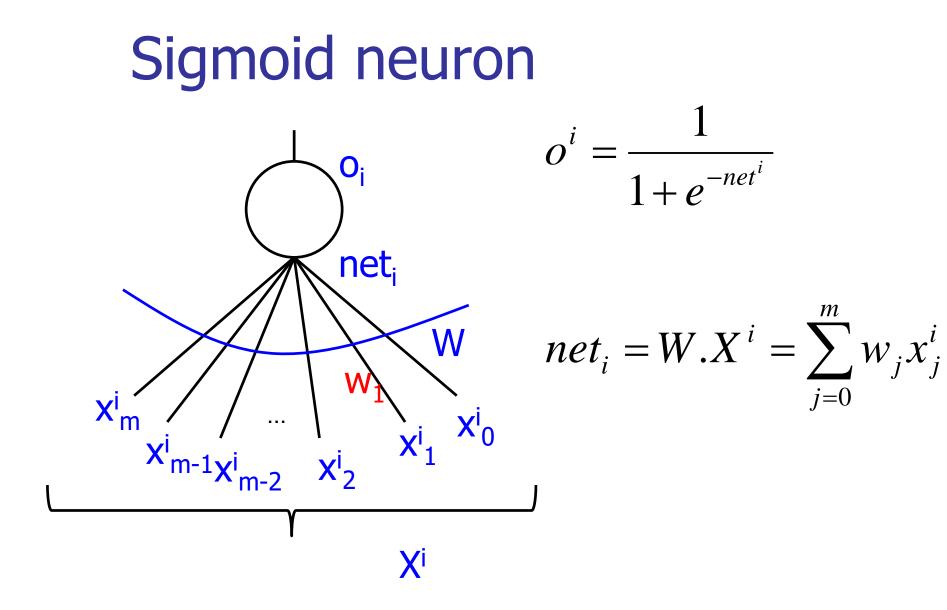
CS217: Artificial Intelligence and Machine Learning (associated lab: CS240)

> Pushpak Bhattacharyya CSE Dept., IIT Bombay Week6 of 10feb25, BP finishing, Logic

Main points covered: week5 of 3feb25



Softmax

$$\sigma(\overline{Z})_i = \frac{e^{Z_i}}{\sum_{j=1}^{K} e^{Z_j}}$$

- σ is the **softmax** function
- Z is the input vector of size K
- The RHS gives the *ith* component of the output vector
- Input to softmax and output of softmax are of the same dimension

Cross Entropy Function

$H(P,Q) = -\sum_{x=1,N} \sum_{k=1,C} P(x,k) \log_2 Q(x,k)$

x varies over *N* data instances, *c* varies over *C* classes *P* is target distribution; *Q* is observed distribution

Sigmoid and Softmax: weight change rule

With Cross Entropy Loss, the change in any weight is

*learning rate * diff between target and observed outputs * input at the connection*

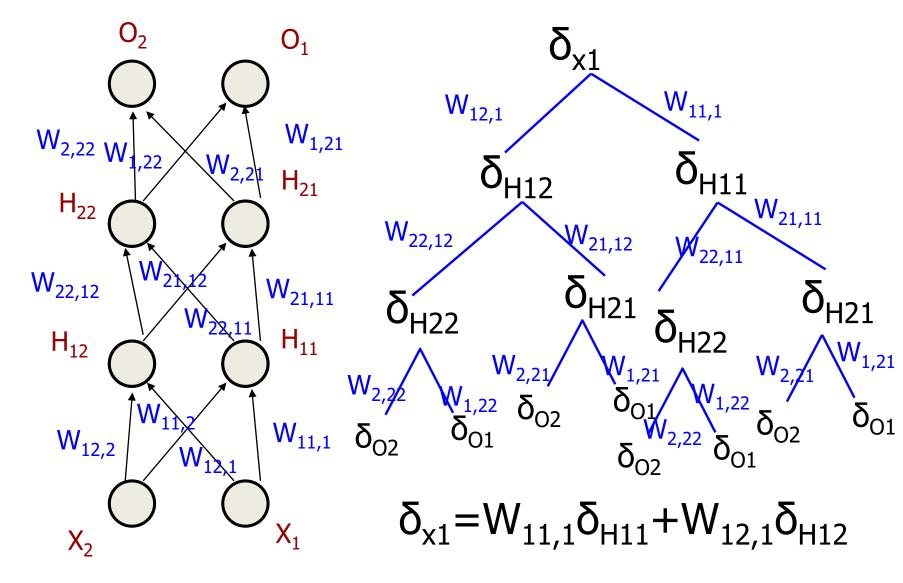
General Backpropagation Rule

- General weight updating rule: $\Delta w_{ji} = \eta \delta j o_i$
- Where

$$\delta_j = (t_j - o_j)o_j(1 - o_j)$$
 for outermost layer

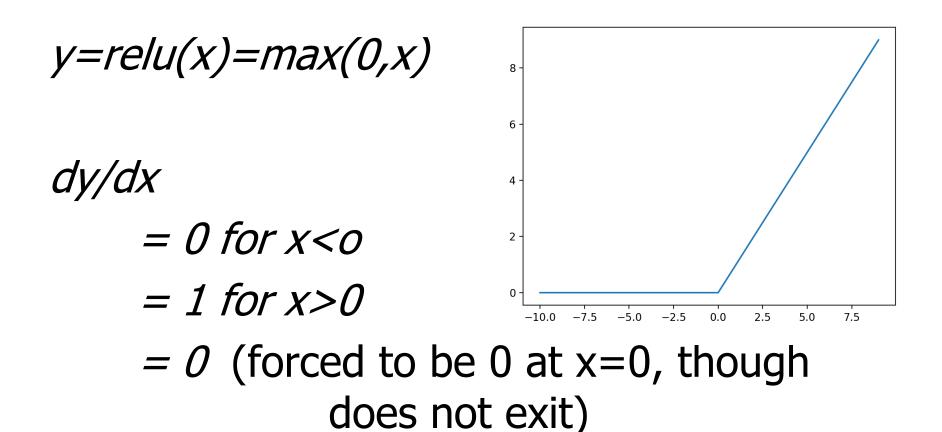
$$= \sum_{k \in \text{next layer}} (w_{kj} \delta_k) o_j (1 - o_j) \text{ for hidden layers}$$

Vanishing/Exploding Gradient



8

RELU and Vanishing Gradient

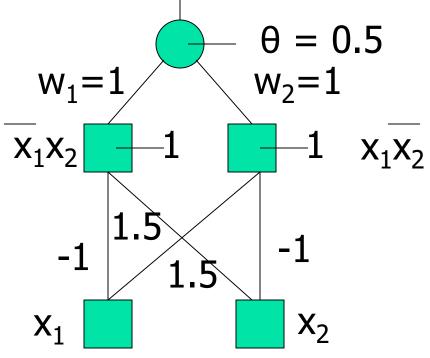


End of main points

Important concepts associated with FFNN-BP

How does BP work?

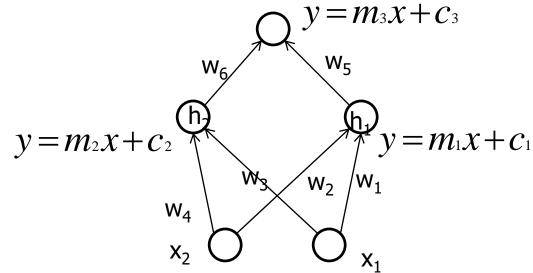
 Input propagation forward and error propagation backward (e.g. XOR)



Work it out !

- In the XOR network, if the activation function of the hidden layer neurons is changed from sigmoid to the ReLU function how will the weight update rule change for minimizing the 'total sum-squared error' of the network?
- 2) Suppose we have two neurons each in both the hidden and the output layer. Softmax is used at the output. Find out the weight update expressions for the following two cases:
 - a) The hidden layer uses ReLU activation.
 - b) The hidden layer uses sigmoid activation.

Can Linear Neurons Work?



$$h_{1} = m_{1}(w_{1}x_{1} + w_{2}x_{2}) + c_{1}$$

$$h_{2} = m_{2}(w_{3}x_{1} + w_{4}x_{2}) + c_{2}$$

$$Out = (w_{5}h_{1} + w_{6}h_{2}) + c_{3}$$

$$= k_{1}x_{1} + k_{2}x_{2} + k_{3}$$

Note: The whole structure shown in earlier slide is reducible to a single neuron with given behavior

 $Out = k_1 x_1 + k_2 x_2 + k_3$

Claim: A neuron with linear I-O behavior can't compute X-OR. **Proof:** Considering all possible cases:

[assuming 0.1 and 0.9 as the lower and upper thresholds]

For (0,0), Zero class: $m(w_1.0+w_2.0-\theta)+c<0.1$ $\Rightarrow c-m.\theta<0.1$

For (0,1), One class:

 $m(w_1.1+w_2.0-\theta)+c>0.9$ $\Rightarrow m.w_1-m.\theta+c>0.9$

For (1,0), One class: $m.w_2 - m.\theta + c > 0.9$

For (1,1), Zero class: $m.(w_1 + w_2) - m \cdot \theta + c < 0.1$

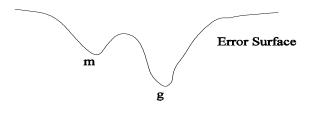
These equations are inconsistent. Because when we add these inequalities after adjusting for sign, we get 0>1.6!

Hence X-OR can't be computed.

Observations:

- 1. A linear neuron can't compute X-OR.
- A multilayer FFN with linear neurons is collapsible to a single linear neuron, hence no a additional power due to hidden layer.
- 3. Non-linearity is essential for power.

Local Minima Due to the Greedy nature of BP, it can get stuck in local minimum *m* and will never be able to reach the global minimum *q* as the error can only decrease by weight change.



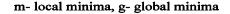
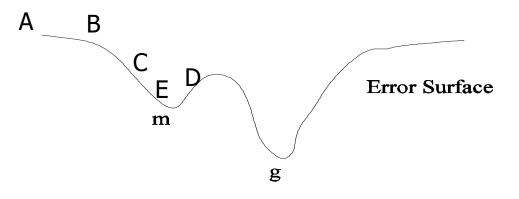


Figure- Getting Stuck in local minimum

Momentum factor

- 1. Introduce momentum factor.
- Accelerates the movement out of the trough.
- > Dampens oscillation inside the trough.
- > Choosing β : If β is large, we may jump over the minimum.

$$(\Delta w_{ji})$$
nth – iteration = $\eta \delta_j O_i + eta (\Delta w_{ji})$ (n – 1)th – iteration

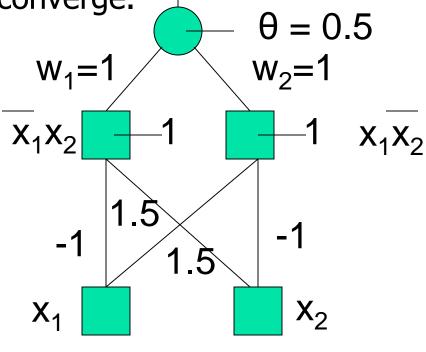


m- local minima, g- global minima

Figure- Getting Stuck in local minimum

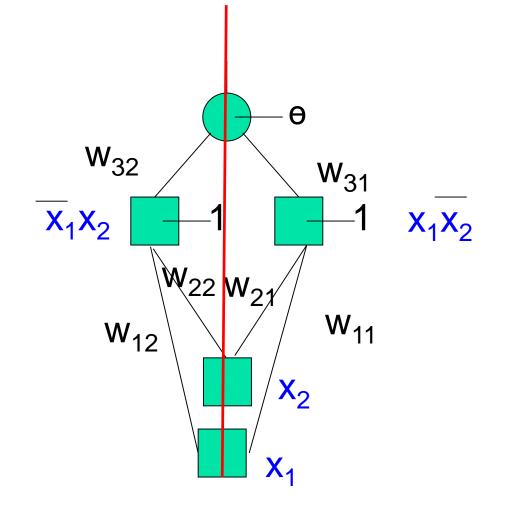
Symmetry breaking

 If mapping demands different weights, but we start with the same weights everywhere, then BP will never converge.



XOR n/w: if we s started with identical weight everywhere, BP will not converge

Symmetry breaking: understanding with proper diagram



Symmetry About The red Line should Be broken

Formal Logic

Theory of CS

Theory A

Logic

Theory B

Algorithm an Complexity

Concepts, Axioms, Rule

- Some foundational questions for Mechanization or Automation of Knowledge Representation and Reasoning:
 - What are symbols and concepts (well formed formulae)
 - What are the self evident and ground truths in the system (axiomatization)
 - What is the validity of the inference (soundness and consistency)
 - Is the inference system powerful enough to capture reality (completeness)
 - Can it be implemented in Turing machine (derivability and complexity)

Case study: Propositional calculus

Propositions

- Stand for facts/assertions
- Declarative statements
 - As opposed to interrogative statements (questions) or imperative statements (request, order)

Operators

AND (/\), OR (\/), NOT (\neg), IMPLICATION (=>)

 \Rightarrow and \neg form a minimal set (can express other operations)

- Prove it.

<u>Tautologies</u> are formulae whose truth value is always T, whatever the assignment is

Model

In propositional calculus any formula with n propositions has 2^n models (assignments)

- Tautologies evaluate to *T* in all models.

Examples:

¹⁾ $P \lor \neg P$

²⁾
$$\neg (P \land Q) \Leftrightarrow (\neg P \lor \neg Q)$$

-De Morgan with AND

Example

Prove $\sim (P \land Q) \rightarrow (\sim P \lor \sim Q)$ is a Tautology.

Q	Р	$L = \sim (P \land Q)$	$R = \sim P \lor \sim Q$	$L \rightarrow R$
Т	Т	F	F	Т
Т	F	Т	Т	Т
F	Т	Т	Т	Т
F	F	Т	Т	Т

Formal Systems

- Rule governed
- Strict description of structure and rule application
- Constituents
 - Symbols
 - Well formed formulae
 - Inference rules
 - Assignment of semantics
 - Notion of proof
 - Notion of soundness, completeness, consistency, decidability etc.

Hilbert's formalization of propositional calculus

- 1. Elements are *propositions* : Capital letters
- 2. Operator is only one : \rightarrow (called implies)
- 3. Special symbol *F* (called 'false')
- 4. Two other symbols : '(' and ')'
- 5. Well formed formula is constructed according to the grammar $WFF \rightarrow P/F/WFF \rightarrow WFF$
- 6. Inference rule : only one

Given $A \rightarrow B$ and

A

write *B*

known as MODUS PONENS

7. Axioms : Starting structures A1: $(A \rightarrow (B \rightarrow A))$

A2:
$$((A \to (B \to C)) \to ((A \to B) \to (A \to C)))$$

A3
$$(((A \to F) \to F) \to A)$$

This formal system defines the propositional calculus

Notion of proof

- 1. Sequence of well formed formulae
- 2. Start with a set of hypotheses
- 3. The expression to be proved should be the last line in the sequence
- 4. Each intermediate expression is either one of the hypotheses or one of the axioms or the result of modus ponens
- 5. An expression which is proved only from the axioms and inference rules is called a THEOREM within the system

Example of proof

From *P* and $P \rightarrow Q$ and $Q \rightarrow R$ prove R H1: *P*

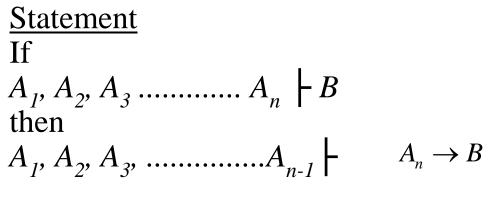
- H2: $P \rightarrow Q$
- H3: $Q \rightarrow R$
- i) *P* H1
- ii) $P \rightarrow Q$ H2
- iii) Q MP, (i), (ii)
- iv) $Q \rightarrow R$ H3
- v) *R* MP, (iii), (iv)

Prove that $(P \rightarrow P)$ is a THEOREM i) $P \rightarrow (P \rightarrow P)$ A1 : P for A and B ii) $P \rightarrow ((P \rightarrow P) \rightarrow P)$ A1: P for A and $(P \rightarrow P)$ for B iii) $[(P \rightarrow ((P \rightarrow P) \rightarrow P)) \rightarrow ((P \rightarrow (P \rightarrow P))) \rightarrow (P \rightarrow P))]$ A2: with P for A, $(P \rightarrow P)$ for B and P for C iv) $(P \rightarrow (P \rightarrow P)) \rightarrow (P \rightarrow P)$ MP, (ii), (iii) v) $(P \rightarrow P)$ MP, (i), (iv)

Shorthand

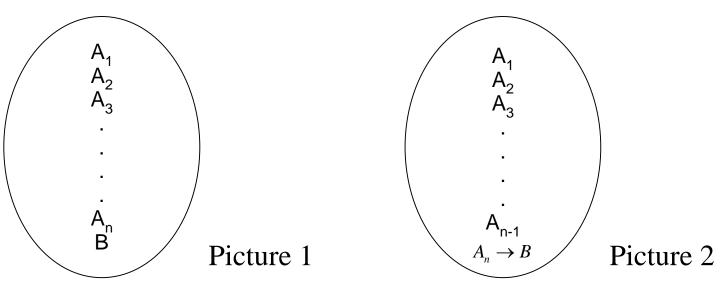
- 1. $\neg P$ is written as $P \rightarrow F$ and called '*NOT P*'
- 2. $((P \rightarrow F) \rightarrow Q)$ is written as $(P \lor Q)$ and called 'P OR Q'
- 3. $((P \rightarrow (Q \rightarrow F)) \rightarrow F)$ is written as $(P \land Q)$ and called 'P AND Q'
- Exercise: (Challenge)
- Prove that $A \rightarrow \neg(\neg(A))$

A very useful theorem (Actually a meta theorem, called deduction theorem)



- is read as 'derives'

Given



Use of Deduction Theorem Prove $A \to \neg(\neg(A))$ $A \to ((A \to F) \to F)$ *i.e.*, $A, A \rightarrow F$ - F (M.P) $A \vdash (A \to F) \to F$ (D.T) $- A \to ((A \to F) \to F)$ (D.T)

Very difficult to prove from first principles, *i.e.*, using axioms and inference rules only

Prove $P \rightarrow (P \lor Q)$ i.e. $P \rightarrow ((P \rightarrow F) \rightarrow Q)$ $P, P \rightarrow F, Q \rightarrow F \vdash F$ $P, P \to F \vdash (Q \to F) \to F$ (D.T) $\downarrow Q$ (M.P with A3) $\mathsf{P} \vdash (P \to F) \to Q$ $\vdash P \to ((P \to F) \to Q)$

More proofs

1. $(P \land Q) \rightarrow (P \lor Q)$ 2. $(P \rightarrow Q) \rightarrow (\neg Q \rightarrow \neg P)$ 3. $(P \rightarrow Q) \rightarrow ((\neg Q \rightarrow P) \rightarrow Q)$

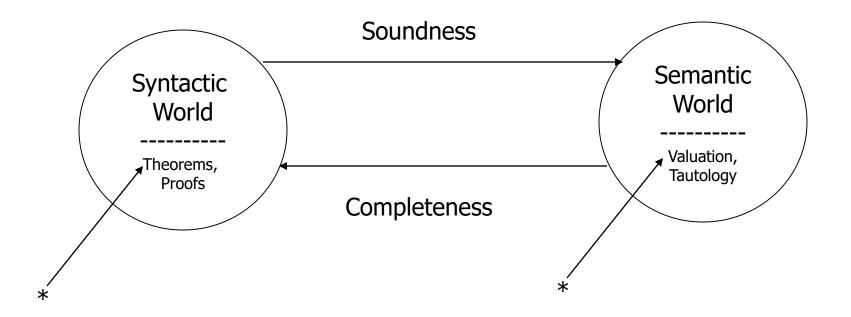
Important to note

- Deduction Theorem is a meta-theorem (statement **about** the system)
- P→P is a theorem (statement
 belonging to the system)
- The distinction is crucial in AI
- Self reference, diagonalization
- Foundation of Halting Theorem, Godel Theorem etc.

Example of `*of-about'* confusion

- "This statement is false"
- Truth of falsity cannot be decided
- Another example: "A city has a barber that shaves ALL AND ONLY those who do NOT shave themselves; Questiondoes the barber shave himself?"
 - Cannot be answered

Soundness, Completeness & Consistency





Provability — Truth

Completeness

■ Truth — Provability

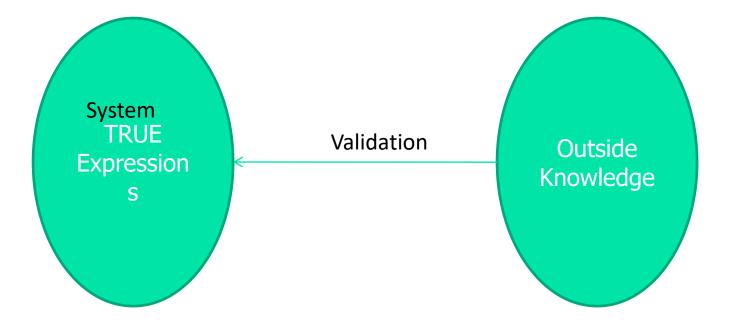


Correctness of the System

Proved entities are indeed true/valid

Completeness: Power of the System

True things are indeed provable



Consistency

The System should not be able to

prove both P and ~P, *i.e.*, should not be

able to derive



Examine the relation between

Soundness & Consistency

Soundness \equiv Consistency

If a System is inconsistent, *i.e.*, can derive

 \mathcal{T} , it can prove any expression to be a

theorem. Because

 $\mathcal{F} \rightarrow \mathsf{P}$ is a theorem

Inconsistency→Unsoundness

To show that

 \mathcal{F} P is a theorem

Observe that

Thus, inconsistency implies unsoundness

Unsoundness → Inconsistency

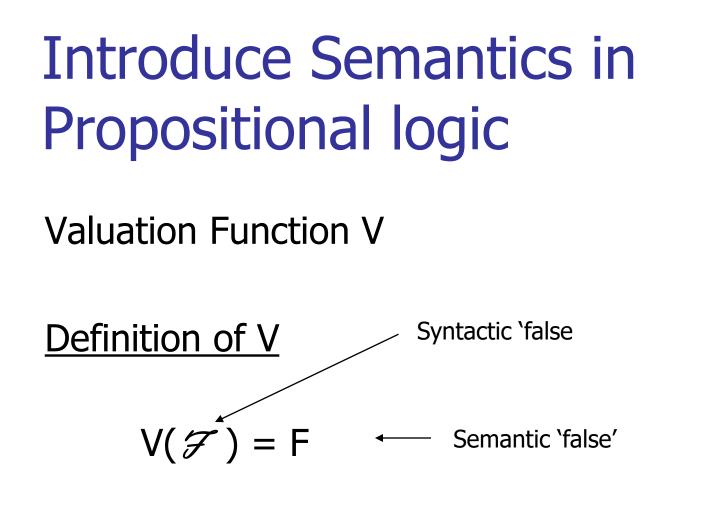
- Suppose we make the Hilbert System of propositional calculus unsound by introducing (A / | B) as an axiom
- Now AND can be written as
 - (A→(B→𝒫)) →𝒫
- If we assign \mathcal{F} to A, we have
 - (𝒫→(B→𝒫)) →𝒫
 - But $(\mathcal{F} \rightarrow (B \rightarrow \mathcal{F}))$ is an axiom (A1)
 - Hence *F* is derived

Inconsistency is a <u>Serious</u> issue.

Informal Statement of Godel Theorem:

If a sufficiently powerful system is complete it is inconsistent.

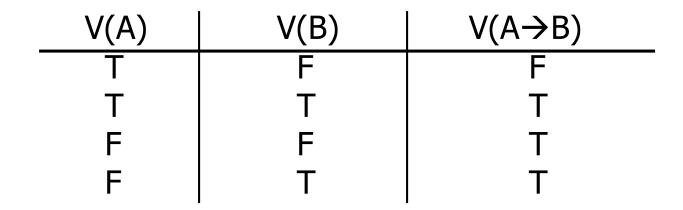
Sufficiently powerful: Can capture at least Peano Arithmetic



Where F is called 'false' and is one of the two symbols (T, F)

$V(\mathcal{F}) = F$

$V(A \rightarrow B)$ is defined through what is called the truth table



Tautology

An expression 'E' is a tautology if

V(E) = Tfor all valuations of constituent propositions

Each 'valuation' is called a 'model'.

To see that

 $(\mathcal{F} P)$ is a tautology

two models V(P) = TV(P) = F

 $V(\mathcal{F}) = T$ for both

⇒ P is a theorem Soundness Completeness ⇒ P is a tautology

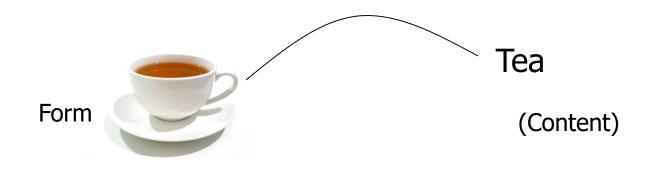
If a system is Sound & Complete, it does not <u>matter</u> how you "Prove" or "show the validity"

Take the Syntactic Path or the Semantic Path

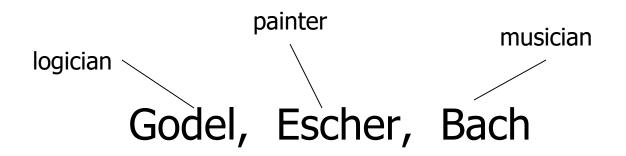
Syntax vs. Semantics issue

Refers to

FORM VS. CONTENT







By D. Hofstadter

Problem

 $(\mathsf{P} \land \mathsf{Q}) \not\rightarrow (\mathsf{P} \lor \mathsf{Q})$

Semantic Proof В Α $P \land Q$ Ρ A→B $P \bigvee$ Q \mathbf{O} Т F F Т F F F F F F Т Т Т

To show syntactically $(P \land Q) \rightarrow (P \lor Q)$

i.e. $[(P \to (Q \to \mathscr{F})) \to \mathscr{F}]$ $\to [(P \to \mathscr{F}) \to Q]$

If we can establish

$$(\mathsf{P} \longrightarrow (\mathsf{Q} \longrightarrow \mathscr{F})) \longrightarrow \mathscr{F},$$
$$(\mathsf{P} \longrightarrow \mathscr{F}), \, \mathsf{Q} \longrightarrow \mathscr{F} \vdash \mathscr{F}$$

This is shown as

 $Q \rightarrow \mathcal{F}$ hypothesis $(Q \rightarrow \mathcal{F}) \rightarrow (P \rightarrow (Q \rightarrow \mathcal{F}) \land \mathcal{A})$

Q → F, hypothesis (Q → F) → (P → (Q → F)); A1 P → (Q → F); MP F, MP Thus we have a proof of the line we

started with

Predicate calculus

Introduce through the "Himalayan Club Example"

Himalayan Club example

- Introduction through an example (Zohar Manna, 1974):
 - Problem: A, B and C belong to the Himalayan club. Every member in the club is either a mountain climber or a skier or both. A likes whatever B dislikes and dislikes whatever B likes. A likes rain and snow. No mountain climber likes rain. Every skier likes snow. Is there a member who is a mountain climber and not a skier?
- Given knowledge has:
 - Facts
 - Rules

Example contd.

- Let *mc* denote mountain climber and *sk* denotes skier.
 Knowledge representation in the given problem is as follows:
 - 1. member(A)
 - 2. member(B)
 - 3. member(C)
 - 4. $\forall x [member(x) \rightarrow (mc(x) \lor sk(x))]$
 - 5. $\forall x[mc(x) \rightarrow \sim like(x, rain)]$
 - 6. $\forall x[sk(x) \rightarrow like(x, snow)]$
 - $z \quad \forall x[like(B, x) \rightarrow \sim like(A, x)]$
 - 8. $\forall x [\sim like(B, x) \rightarrow like(A, x)]$
 - <u>9.</u> like(A, rain)
 - *10. like(A, snow)*
 - 11. Question: $\exists x [member(x) \land mc(x) \land \sim sk(x)]$
- We have to infer the 11th expression from the given 10.
- Done through Resolution Refutation.

Club example: Inferencing

- 1. *member(A)*
- 2. member(B)
- 3. *member(C)*
- 4. $\forall x[member(x) \rightarrow (mc(x) \lor sk(x))]$
 - Can be written as - $\sim member(x) \bigvee mc(x) \lor sk(x)$ $(mc(x) \lor sk(x))]$
- 5. $\forall x[sk(x) \rightarrow lk(x, snow)]$ - $\sim sk(x) \lor lk(x, snow)$
- 6. $\forall x[mc(x) \rightarrow \sim lk(x, rain)]$

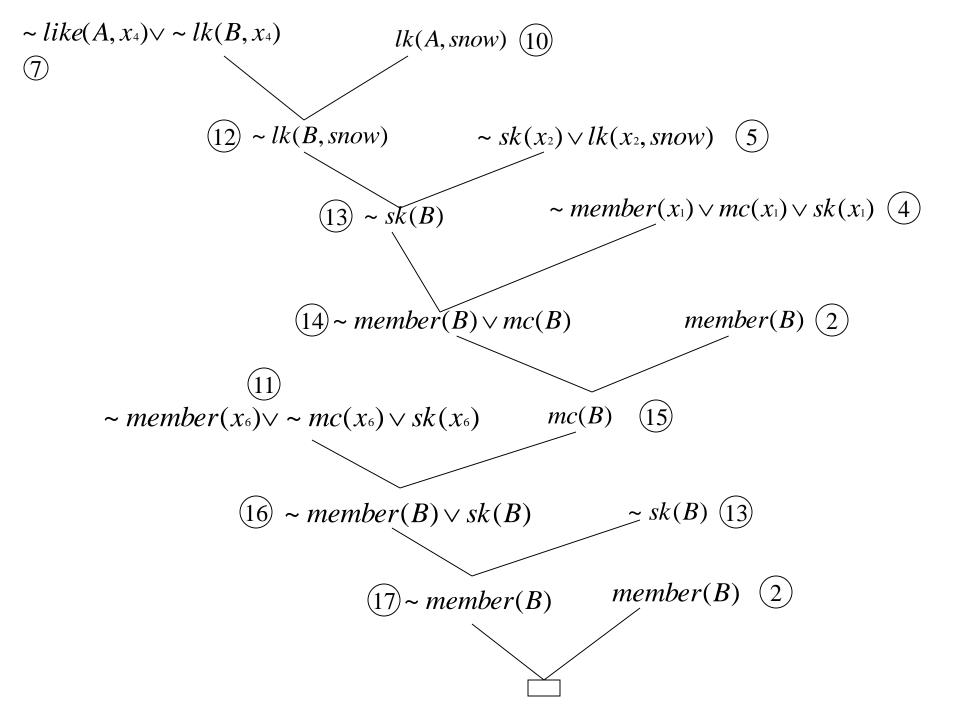
 $\sim mc(x) \lor \sim lk(x, rain)$

7. $\forall x[like(A, x) \rightarrow ~lk(B, x)]$

$$\sim like(A, x) \lor \sim lk(B, x)$$

- 8. $\forall x [\sim lk(A, x) \rightarrow lk(B, x)]$ _ $lk(A, x) \lor lk(B, x)$
- 9. lk(A, rain)
- 10. lk(A, snow)
- 11. $\exists x [member(x) \land mc(x) \land \sim sk(x)]$
 - Negate- $\forall x [\sim member(x) \lor \sim mc(x) \lor sk(x)]$

- Now standardize the variables apart which results in the following
- 1. *member(A)*
- 2. *member(B)*
- 3. *member(C)*
- 4. ~ member(x_1) \lor mc(x_1) \lor sk(x_1)
- 5. ~ $sk(x_2) \lor lk(x_2, snow)$
- 6. ~ $mc(x_3) \lor \sim lk(x_3, rain)$
- 7. ~ $like(A, x_4) \lor \sim lk(B, x_4)$
- 8. $lk(A, x_5) \vee lk(B, x_5)$
- 9. *lk*(*A*, *rain*)
- 10. lk(A, snow)
- 11. ~ member(x_6) \lor ~ $mc(x_6) \lor$ $sk(x_6)$



Well known examples in Predicate Calculus

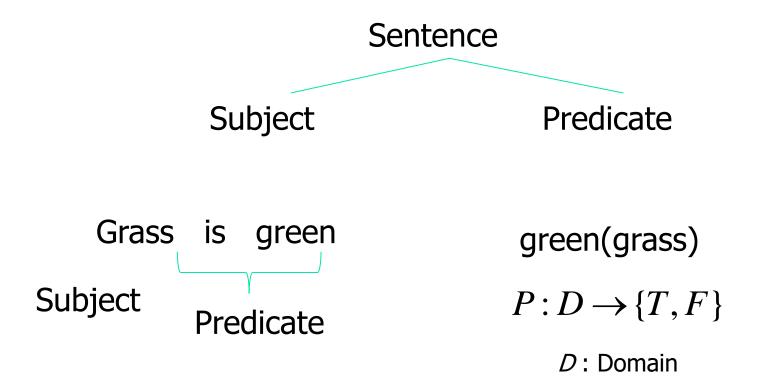
Man is mortal : rule

 $\forall x[man(x) \rightarrow mortal(x)]$

- shakespeare is a man man(shakespeare)
- To infer shakespeare is mortal mortal(shakespeare)

Predicate Calculus: origin

Predicate calculus originated in language



Predicate Calculus: only for declarative sentences

Is grass green? (Interrogative)Oh, grass is green! (Exclamatory)

Declarative Sentence

Subject

Predicate

Grass which is supple is green

 $\forall x(\operatorname{grass}(x)) \land \operatorname{supple}(x) \rightarrow \operatorname{green}(x))$

Predicate Calculus: more expressive power than propositional calculus

- 2 is even and is divisible by 2: P1
- 4 is even and is divisible by 2: P2
 6 is even and is divisible by 2: P3
 Generalizing,

 $\forall x ((Integer(x) \land even(x) \Rightarrow divides(2, x)))$

Predicate Calculus: finer than propositional calculus

- Finer Granularity (Grass is green, ball is green, leaf is green (green(x)))
- 2. Succinct description for infinite number of statements which would need \propto number of properties
- 3 place predicate Example: x gives y to z

give(x,y,z)

4 place predicate Example: x gives y to z through w

give(x,y,z,w)

Double causative in Hindi giving rise to higher place predicates

- जॉन ने खाना खाया John ne khana khaya John <CM> food ate John ate food *eat(John, food)*
- जॉन ने जैक को खाना खिलाया John ne Jack ko khana khilaya John <CM> Jack <CM> food fed John fed Jack *eat(John, Jack, food)*
- जॉन ने जैक को जिल के द्वारा खाना खिलाया John ne Jack ko Jill ke dvara khana khilaya John <CM> Jack <CM> Jill <CM> food made-to-eat John fed Jack through Jill *eat(John, Jack, Jill, food)*