



The Logic of Hebbian Learning

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The Neuro-Symbolic Problem

Symbolic Systems

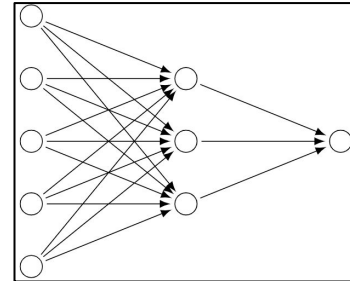
- ✓ Sophisticated rich reasoning
- ✓ Explainable decisions
- ✗ Notoriously rigid and static
- ✗ Manual knowledge-engineering

Neural Networks

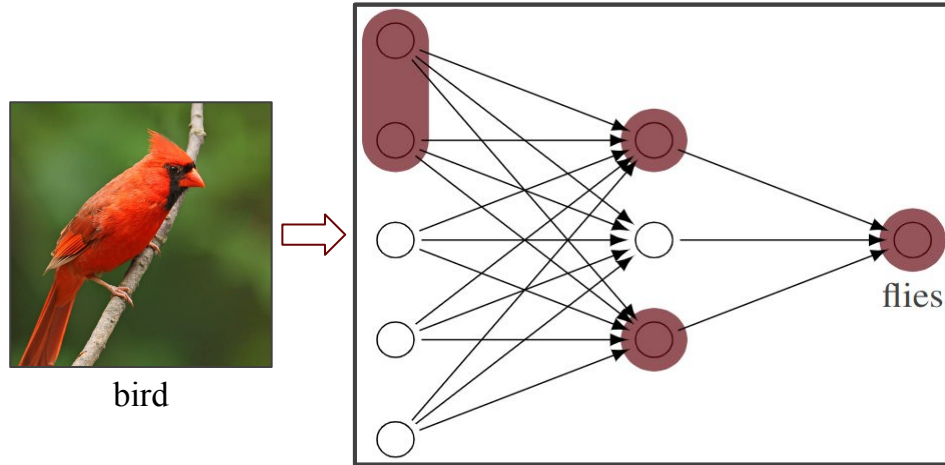
- ✗ Can't readily learn rich inference
- ✗ "Black Box" decisions
- ✓ Learns from experience
- ✓ Uses unstructured data

penguin \rightarrow bird
bird \Rightarrow flies

Problem:
How do we bridge
the two?



Forward Propagation & Conditionals

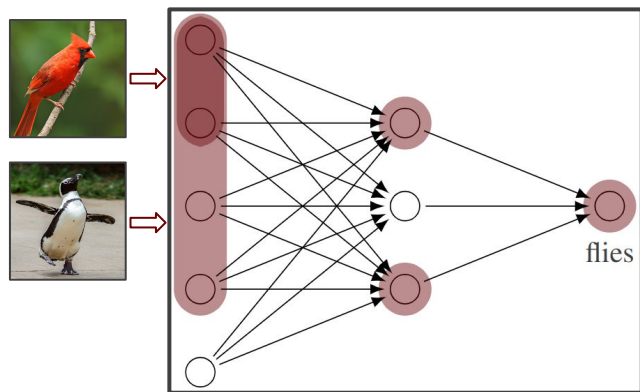


$$\boxed{\text{bird}} \quad \varphi \Rightarrow \psi \quad \text{iff} \quad \text{Prop}([\varphi]) \supseteq [\psi] \quad \boxed{\text{flies}}$$

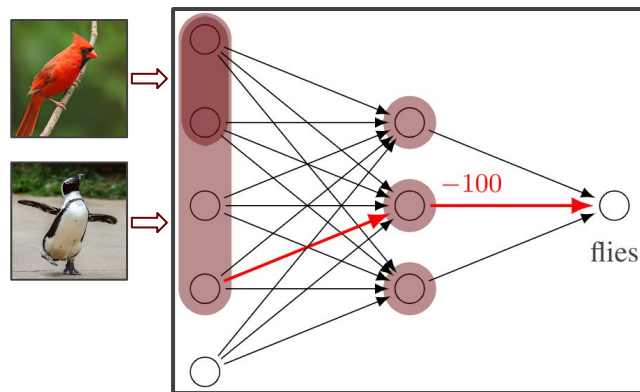
Learning Conditionals

penguin \rightarrow bird
bird \Rightarrow flies

Model Building



Learning



penguin \rightarrow bird
bird \Rightarrow flies
 \neg (penguin \Rightarrow flies)

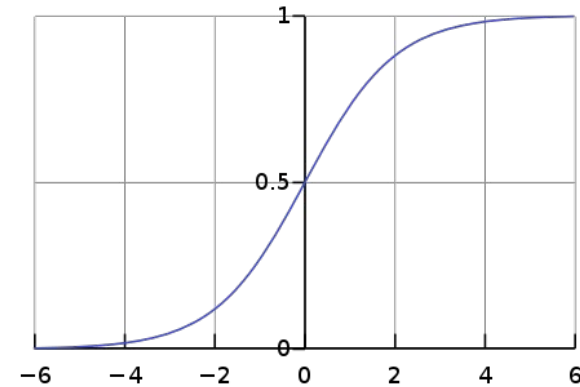
???



Simplifying Assumptions

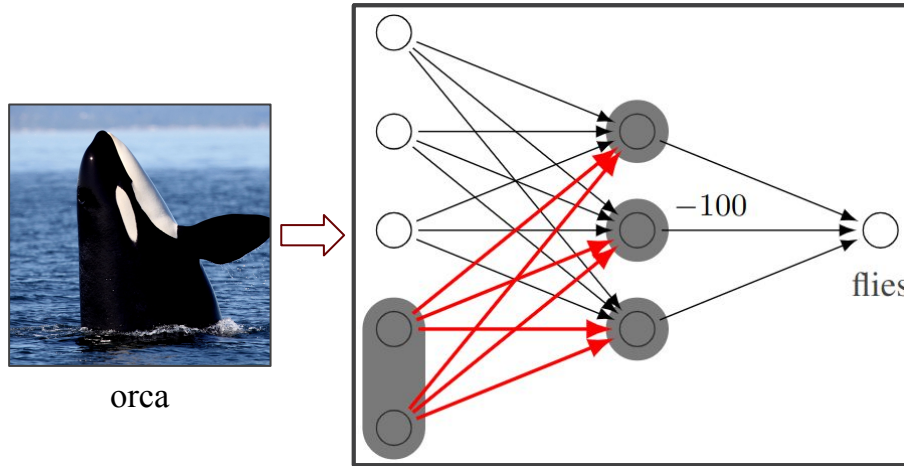
$$\mathcal{N} = \langle N, E, W, A, O, \eta \rangle$$

1. Net is **Feedforward**
2. Activations are **monotonically increasing**
3. Neuron outputs are **binary**



Hebbian Learning

Neurons that fire together wire together



$$\Delta W_{ij} = \eta x_i x_j$$

Prop and Inc

Prop : Set \rightarrow Set

Prop(S) means *forward-propagate* S in the net.

Inc : Net \times Set \rightarrow Net

Inc(\mathcal{N} , S) means *increase the weights* of edges within
Prop(S) by $\Delta W_{ij} = \eta x_i x_j$

The Logic

$p \mid \top \mid \neg\varphi \mid \varphi \wedge \varphi \mid \varphi \rightarrow \varphi \mid \varphi \Rightarrow \varphi \mid \mathbf{T}\varphi \mid [\varphi^+]\varphi$

$$[\mathbf{T}\varphi] = \text{Prop}([\varphi])$$

$$[[\varphi^+]\psi] = [\psi]_{\text{Inc}(\mathcal{N}, [\varphi])}$$

Some Axioms & Rules

Basic Axioms

(PC) All propositional tautologies

(DUAL) $\langle \mathbf{T} \rangle \varphi \leftrightarrow \neg \mathbf{T} \neg \varphi$

(N) $\mathbf{T}\mathbf{T}$

(T) $\mathbf{T}\varphi \rightarrow \varphi$

(4) $\mathbf{T}\varphi \rightarrow \mathbf{T}\mathbf{T}\varphi$

Inference Rules

(MP) $\frac{\varphi \quad \varphi \rightarrow \psi}{\psi}$

(TYP) $\frac{\frac{\varphi \Rightarrow \psi}{\mathbf{T}\varphi \rightarrow \psi} \quad \mathbf{T}\varphi \rightarrow \psi}{\varphi \Rightarrow \psi}$

(C \Rightarrow) $\frac{\varphi \rightarrow \psi \quad \psi \Rightarrow \varphi}{\varphi \leftrightarrow \psi}$

(LOOP \Rightarrow) $\frac{\varphi_0 \Rightarrow \varphi_1 \cdots \varphi_{k-1} \Rightarrow \varphi_k \quad \varphi_k \Rightarrow \varphi_0}{\varphi_0 \Rightarrow \varphi_k}$

(NEC $_+$) $\frac{\psi}{[\varphi^+] \psi}$

(C $_+$) $\frac{\psi \rightarrow \rho \quad [\varphi^+] \rho \rightarrow \psi}{[\varphi^+] \psi \leftrightarrow [\varphi^+] \rho}$

(LOOP $_+$) $\frac{[\varphi^+] \psi_0 \rightarrow \psi_1 \cdots [\varphi^+] \psi_{k-1} \rightarrow \psi_k \quad [\varphi^+] \psi_k \rightarrow \psi_0}{[\varphi^+] \psi_0 \rightarrow \psi_k}$

Reduction Axioms

(R $_p$) $[\varphi^+] p \leftrightarrow p$

(R $_{\neg}$) $[\varphi^+] \neg \psi \leftrightarrow \neg [\varphi^+] \psi$

(R $_{\wedge}$) $[\varphi^+] (\psi \wedge \rho) \leftrightarrow ([\varphi^+] \psi \wedge [\varphi^+] \rho)$

(NEST $_T$) $[\mathbf{T}\varphi^+] \psi \leftrightarrow [\varphi^+] \psi$

Key Axioms

(NS) $[\varphi^+] \mathbf{T}\psi \rightarrow \mathbf{T}[\varphi^+] \psi$

(TP) $\mathbf{T}[\varphi^+] \psi \wedge \mathbf{T}\varphi \rightarrow [\varphi^+] \mathbf{T}\psi$

The Key Axioms

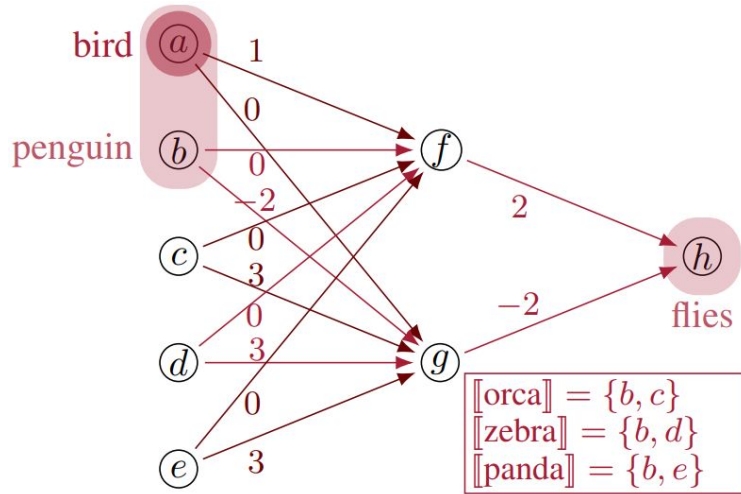
(No Surprises)

$$[\varphi^+] \mathbf{T}\psi \rightarrow \mathbf{T}[\varphi^+] \psi$$

(Typicality Preservation)

$$\mathbf{T}[\varphi^+] \psi \wedge \mathbf{T}\varphi \rightarrow [\varphi^+] \mathbf{T}\psi$$

Working Code!



$\mathcal{N} \models \mathbf{T}(\text{penguin}) \rightarrow \text{flies, yet}$

$\mathcal{N} \not\models [orca^+][zebra^+][panda^+](\mathbf{T}(\text{penguin}) \rightarrow \text{flies})$

```
print(model.is_model("(typ penguin) implies flies"))
```

True

```
print(model.is_model("orca+ zebra+ panda+ \  
((typ penguin) implies flies)"))
```

False

github.com/ais-climber/neural-semantic

The screenshot shows a GitHub repository page for 'ais-climber/neural-semantic'. The main content is the README file, which includes a title 'Neural Semantics', a description of the project as a neuro-symbolic interface, and a note about its development status. Below the text is a diagram illustrating the workflow: 'Model Building' (from knowledge to a neural network) and 'Extraction' (from a neural network back to knowledge). The diagram also shows 'Learning' and 'Knowledge Engineering' components.

github.com/ais-climber/neural-semantic

Apps Email & Social internships-sum... intern status CORE Computer... CORE Computer... cv.pdf blancocv.pdf CV.pdf 1988-Normal Mul...

☰ README.md

Neural Semantics

A neuro-symbolic interface, intended for both **model extraction** (extracting knowledge from a net) as well as **model building** (building a net from a knowledge base). The name comes from the core idea -- that the internal dynamics of neural networks can be used as formal semantics of knowledge bases.

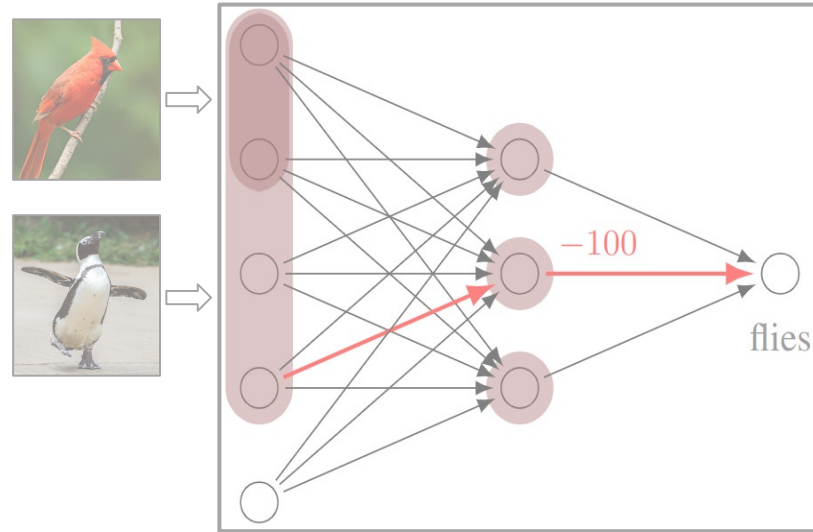
NOTE: This program is currently **very much in development**, and many of the planned features involve significant research efforts (this is my PhD). So what the program can do right now is somewhat limited.

The diagram illustrates the neuro-symbolic interface workflow. It shows a cycle between knowledge engineering and model building/extraction. Knowledge engineering involves defining rules like 'penguin → bird' and 'bird ⇒ flies'. Model building uses these rules to create a neural network. Learning involves training the network with images of a penguin and a bird. Extraction involves analyzing the network's internal dynamics to extract knowledge, such as 'penguin → bird', 'bird ⇒ flies', and '¬(penguin ⇒ flies)'. Knowledge engineering also includes a 'Knowledge Engineering' box that interacts with the extracted knowledge.

Future Work: The Dream

1. Model Building (i.e. Completeness)
2. First-order logic
3. Nonbinary (fuzzy-valued) output
4. More varied activation functions (e.g. ReLU)
5. Learning via backpropagation

Questions?



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github.com/ais-climber/neural-semantic



Appendix / Helper Slides



The Logic: Rules of Inference

(PC) All propositional tautologies

(DUAL) $\langle \mathbf{T} \rangle \varphi \leftrightarrow \neg \mathbf{T} \neg \varphi$

(N) $\mathbf{T} \top$

(T) $\mathbf{T} \varphi \rightarrow \varphi$

(4) $\mathbf{T} \varphi \rightarrow \mathbf{T} \mathbf{T} \varphi$

(MP)
$$\frac{\varphi \quad \varphi \rightarrow \psi}{\psi}$$

(TYP)
$$\frac{\varphi \Rightarrow \psi}{\mathbf{T} \varphi \rightarrow \psi} \quad \frac{\mathbf{T} \varphi \rightarrow \psi}{\varphi \Rightarrow \psi}$$

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$$\frac{\varphi_0 \Rightarrow \varphi_1 \cdots \varphi_{k-1} \Rightarrow \varphi_k \quad \varphi_k \Rightarrow \varphi_0}{\varphi_0 \Rightarrow \varphi_k}$$

The Logic: Rules of Inference

$$\begin{array}{l} (\text{NEC}_+) \quad \frac{\psi}{[\varphi^+] \psi} \\ (\text{C}_+) \quad \frac{\psi \rightarrow \rho \quad [\varphi^+] \rho \rightarrow \psi}{[\varphi^+] \psi \leftrightarrow [\varphi^+] \rho} \\ (\text{LOOP}_+) \quad \frac{[\varphi^+] \psi_0 \rightarrow \psi_1 \cdots [\varphi^+] \psi_{k-1} \rightarrow \psi_k \quad [\varphi^+] \psi_k \rightarrow \psi_0}{[\varphi^+] \psi_0 \rightarrow \psi_k} \end{array}$$

$$\begin{array}{l} (\text{R}_p) \quad [\varphi^+] p \leftrightarrow p \\ (\text{R}_{\neg}) \quad [\varphi^+] \neg \psi \leftrightarrow \neg [\varphi^+] \psi \\ (\text{R}_{\wedge}) \quad [\varphi^+] (\psi \wedge \rho) \leftrightarrow ([\varphi^+] \psi \wedge [\varphi^+] \rho) \\ (\text{NEST}_{\mathbf{T}}) \quad [\mathbf{T}\varphi^+] \psi \leftrightarrow [\varphi^+] \psi \end{array}$$

$$\begin{array}{l} (\text{NS}) \quad [\varphi^+] \mathbf{T}\psi \rightarrow \mathbf{T}[\varphi^+] \psi \\ (\text{TP}) \quad \mathbf{T}[\varphi^+] \psi \wedge \mathbf{T}\varphi \rightarrow [\varphi^+] \mathbf{T}\psi \end{array}$$

Future Work: Completeness

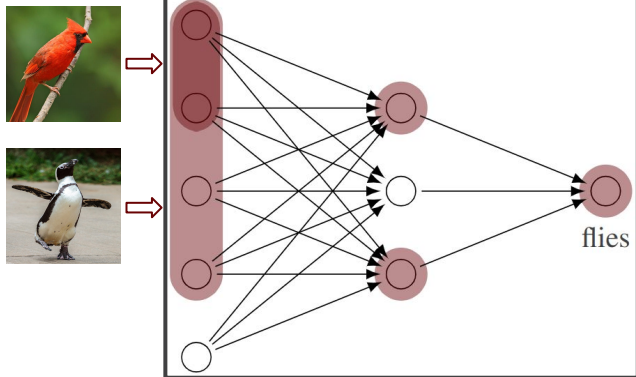
penguin \rightarrow bird
bird \Rightarrow flies

Knowledge Engineering

penguin \rightarrow bird
bird \Rightarrow flies
 \neg (penguin \Rightarrow flies)

Model Building

Extraction



Learning

