

Inductive Logic Programming (ILP) vs Deep Learning DL)

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Paper 1: ILP vs DL in One-Shot Learning

- Title: "Human-like rule learning from images using one-shot hypothesis derivation"
- Author: Muggleton
- Content: ILP vs Deep Learning in One Shot Learning
- Databases:
 - Omniglot dataset: character recognition
 - UK Biobank: neurodegenerative disease identification
- Link: https://www.researchgate.net/publication/355875180_Human-like_rule_learning_from_images_using_one-shot_hypothesis_derivation

Omniglot Dataset

- Character recognition
- 1623 different handwritten characters from 50 different alphabets
- Each character is a different class
- Each character has 20 examples
 - ✎ Written by 20 different people

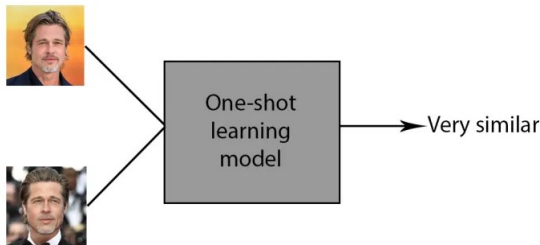
Bengali						Sanskrit					
ঐ	ঐ	ঐ	ঐ	ঐ	ঐ	प	झ	झ	ष	म	क
उ	क	म	ज	उ	व	ट	ठ	क	ज	फ	अ
द	थ	म	वा	ए	इ	ड	ए	न	ज	ग	थ
म	ह	उ	ड	म	ग	ल	आ	भ	ओ	य	उ
उ	त	ह	श	म	उ	र	छ	ण	इ	ल	थ
च	ग	ट	न	ड	अ	क	च	इ	ब	ह	श
ऊ	र	व									

...

- ✓ images_background
 - ✎ Alphabet_of_the_Magi
 - ✎ Anglo-Saxon_Futhorc
 - ✎ Arcadian
 - ✎ Armenian
 - ✎ Asomtavruli_(Georgian)
 - ✎ Balinese
 - ✓ Bengali
 - ✎ character01
 - 0132_01.png
 - 0132_02.png
 - 0132_03.png
 - 0132_04.png
 - 0132_05.png
 - 0132_06.png
 - 0132_07.png
 - 0132_08.png
 - 0132_09.png
 - 0132_10.png
 - ... 10 more
 - ✎ character02
 - ✎ character03

One Shot Learning

- Classification task in which one is provided with 1 (or few) example of each class
- Deep learning usually needs thousands of examples to learn a new class
- Human only need to see a picture of a giraffe once to be able to recognise another one
- Most well known use case: Facial recognition



Siamese Neural Networks

- Model trained to differentiate between images of different classes
- <https://www.cs.cmu.edu/~rsalakhu/papers/oneshot1.pdf>

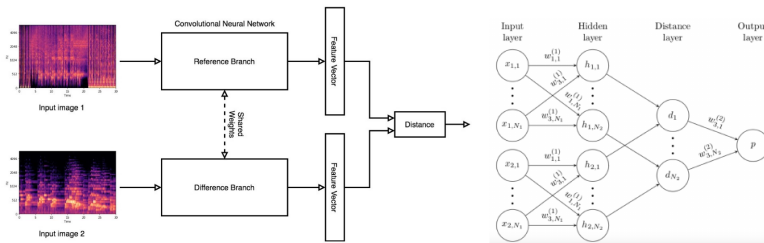


Figure: Architectures of a Siamese NN

ILP approach

- One-Shot Hypothesis Derivation (OSHD) based on
- Implemented in Toplog
- Background knowledge:

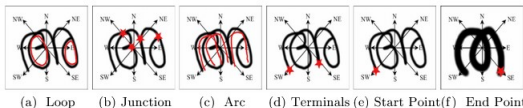


Fig.3: Human-like feature extraction criteria

Table 1: Geometrical and Directional Properties

Character ID	Geometrical Properties				Directional Properties	
	No. Loops	No. Junctions	No. Arcs	No. Terminals	Starting Point	Ending Point
1	2	4	3	2	sw	null
2	3	4	3	2	sw	null
3	3	4	3	2	sw	null
4	1	2	3	2	null	se
5	1	3	3	2	nw	se

- Modes:

```
:- modeh(1,alphabet(+character)).
:- modeb(*,has_gemproperties(+character,-properties)).
:- modeb(*,has_gemproperties_count(+properties,
                                     #geo_feature_name,#int)).
:- modeb(*,has_dirproperties(+character,-properties)).
:- modeb(*,has_dirproperties_feature(+properties,
                                     #dir_feature_name,#featurevalue)).
```

ILP vs DL

Background

- ILP:
 - Modes
 - Table of Geometrical and Directional Properties
- DL:
 - Many examples to train the model
 - 40 (of 50) alphabets used as background

Training

- ILP:
 - The 1 example generates a new rule
- DL:
 - The 1 example is kept in memory

Testing

- ILP:
 - The new example is verified against the new rule
- DL:
 - The new example, with the training example are compared to see whether they belong to the same class

- Results provided by Muggleton are not consistent with those provided by the original Siamese paper

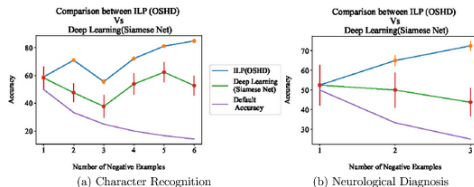


Fig. 6: Average Predictive accuracy of ILP (OSHD) vs Deep Learning (Siamese Net)

Table 1. Accuracy on Omniglot verification task (siamese convolutional neural net)

Method	Test
30k training	
no distortions	90.61
<i>affine distortions x8</i>	91.90
90k training	
no distortions	91.54
<i>affine distortions x8</i>	93.15
150k training	
no distortions	91.63
<i>affine distortions x8</i>	93.42

Figure: Comparison of the two models by Muggleton (left) and accuracy in the Siamese paper (right)

Approach 2: DeepProbLog

- Title: "From Statistical Relational AI to Neural Symbolic Computation: DeepProbLog"
- Author: Luc De Raedt
- Content: Adding a ProbLog brick at the end of Neural Networks
- Databases:
 - MNIST dataset: digit recognition
- Link: https://www.i-aida.org/wp-content/uploads/2022/03/Luc-De-Raedt_compressed.pdf

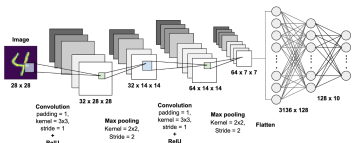
DeepMNIST addition

- Set of examples:

```
addition(3, 5, 8), addition(0, 4, 4), addition(9, 2, 11), ...
```

- Model:

- Neural network:



- ProbLog brick:

```
nn(mnist_net, [X], Y, [0 ... 9] ) :: digit(X,Y).
```

```
addition(X,Y,Z) :- digit(X,N1), digit(Y,N2), Z is N1+N2.
```

- Execution:

```
addition(3, 5, 8) :- digit(3, N1), digit(5, N2), 8 is N1 + N2.
```

