



# Evidence-Based Reasoning in Classifier Hierarchies

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# Overview

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- Motivation
- Hierarchical Neural Network Classifiers
- Dempster-Shafer Theory of Evidence / Transferable Belief Model
- Evidence-Based Reasoning
- Results
- Conclusions

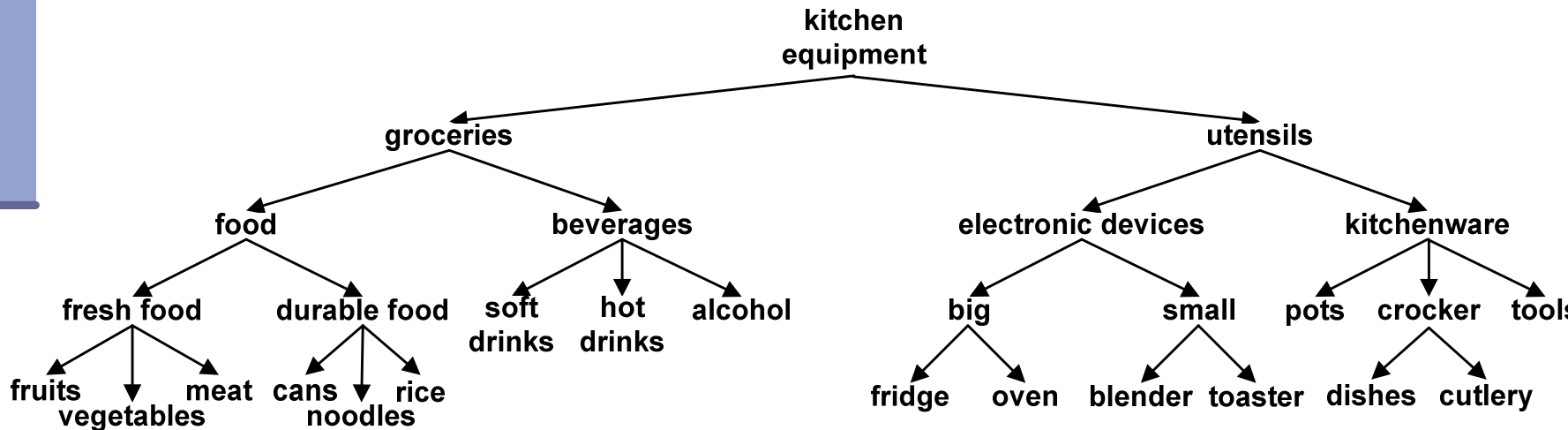
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# Motivation

- Object recognition problems are **multi-class problems**
  - Number of classes greater than 2
- Object recognition problems often show a **hierarchical structure**



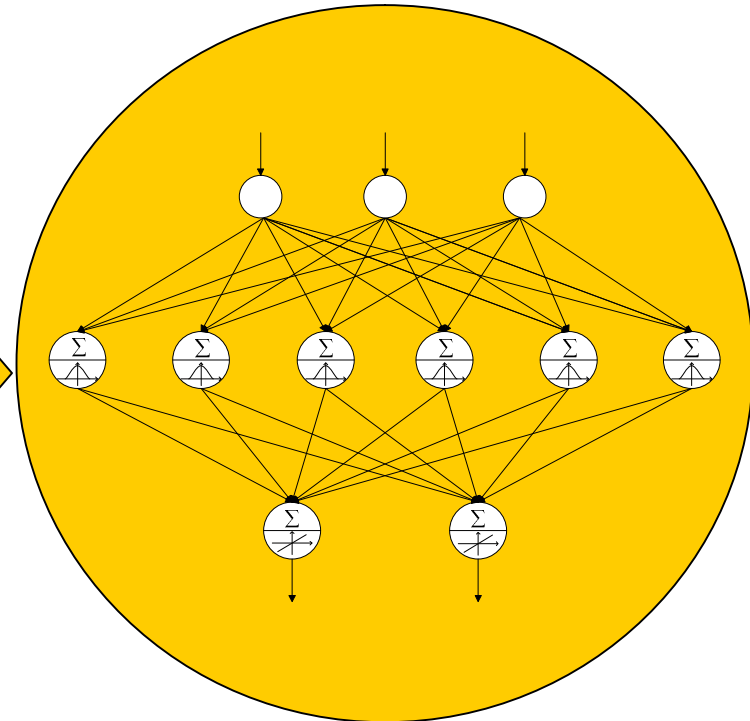
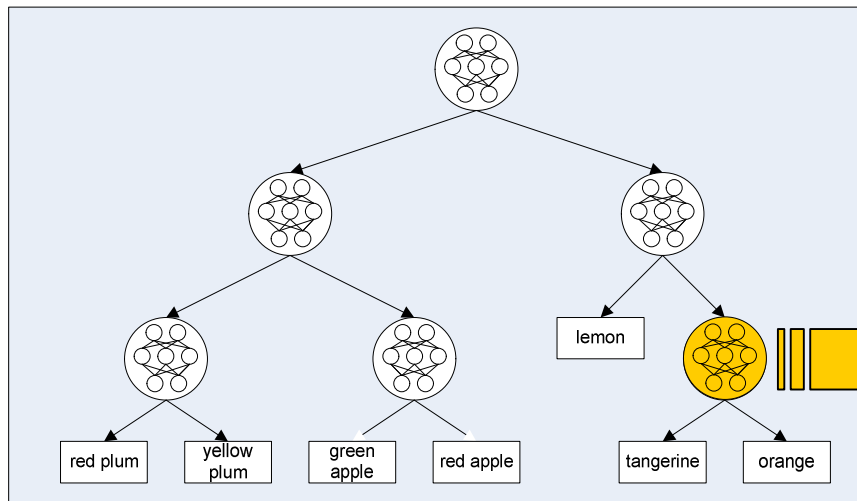


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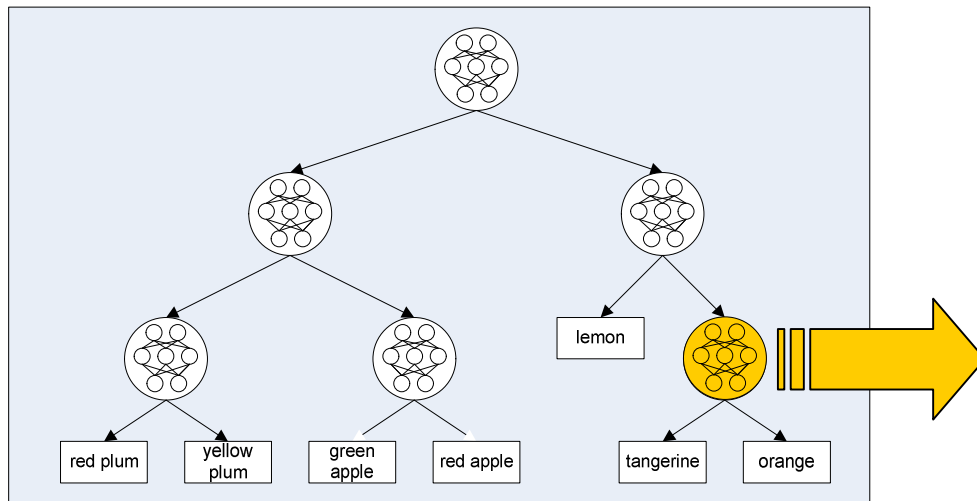
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# Hierarchical Neural Network Classifiers



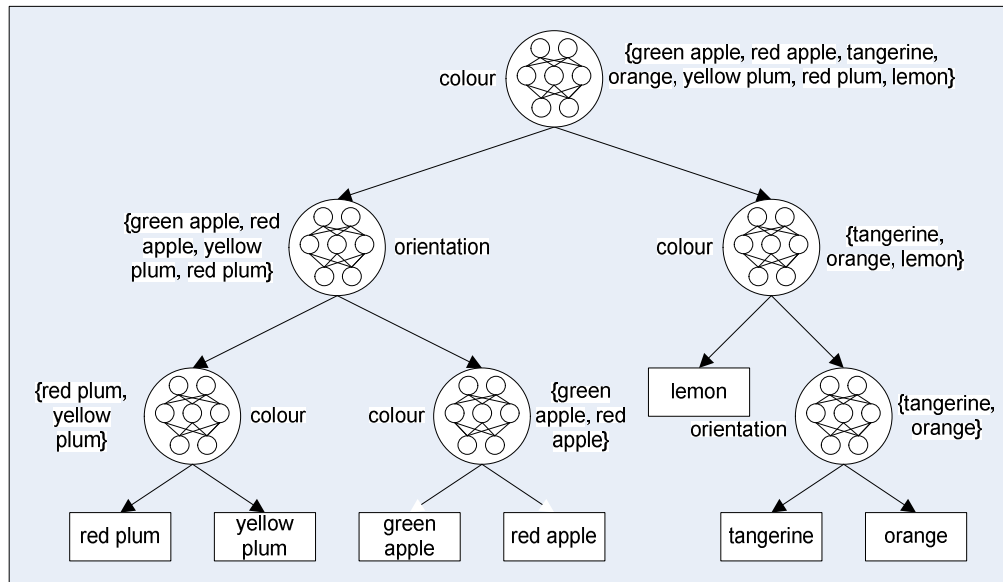
# Hierarchical Neural Network Classifiers



## Classifiers

- RBF
- k-NN
- Fuzzy k-NN
- LVQ

# Hierarchical Neural Network Classifiers



Fay et al. *Learning Object Recognition in a NeuroBotic System*. SOAVE 2004

- Hierarchically organised neural networks
- Coarse to fine output space division
- Hierarchical class grouping
- Different strategies to combine evidences of the individual classifiers to a collective output

# Hierarchical Neural Network Classifiers

- Classifiers provide evidence at different levels of abstraction
- Classifiers provide evidence for single classes or sets of classes
- The classifier outputs indicate how likely a given sample belongs to a class or group of classes
- Hierarchical neural network classifiers can be interpreted as a group of experts making hierarchically structured statements

# Benefits of Hierarchical Neural Network Classifiers

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- Division of complex classification problems into several less complex classification problems
- Links symbolic and sub-symbolic information
- Results are available at different levels of abstraction

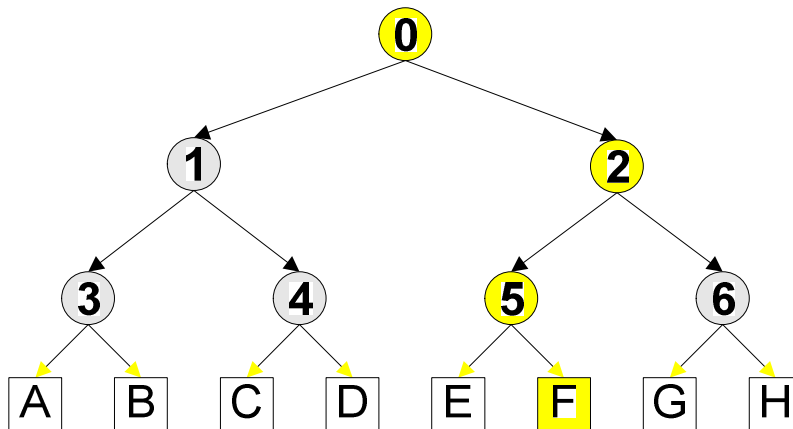
# Hierarchy Retrieval

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- Retrieval of combined classification result
  - Similar to decision tree approach
  - Utilising Dempster-Shafer theory of evidence

# Hierarchy Retrieval

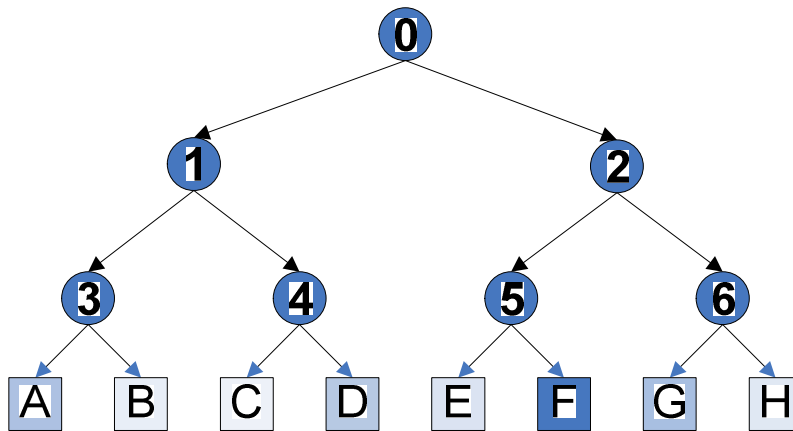
## Similar to Decision Trees



- Only part of the experts are consulted
- The decision is successively delegated to an expert at the next lower level resulting in a path from the root node to an end node
- Result is a class



# Hierarchy Retrieval Utilising Dempster-Shafer Theory of Evidence



- All experts are consulted
- Each expert gives his opinion
- Result is a class as well as evidences for the individual classes

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# Dempster-Shafer Theory of Evidence

- Generalisation of the Bayesian probability theory
- Provides means for dealing with incomplete knowledge, uncertainty and ignorance
- Allows for assigning evidence not only to atomic hypotheses but also to sets of hypotheses
- Provides rule for combining evidence (orthogonal sum)
- Suitable for combining hierarchically structured knowledge incorporating uncertainty

# Dempster-Shafer Theory of Evidence

## ■ Basic concepts

- Frame of discernment  $\Omega = \{q_1, q_2, \dots, q_q\}$

Set of mutually exclusive atomic hypotheses

- Basic probability assignment (or mass)  $m$

$$m : 2^\Omega \mapsto [0,1] \quad m(\emptyset) = 0 \quad \sum_{A \subseteq \Omega} m(A) = 1$$

- Dempster's combination rule (orthogonal sum)

$$m_{1,2} = m_1 \oplus m_2$$

Combines evidence from different sources

$$m_{1,2}(C) = K^{-1} \sum_{A,B:A \cap B = C} m_1(A) \cdot m_2(B), \quad \forall C \neq \emptyset$$

$$K = 1 - \sum_{A,B:A \cap B = \emptyset} m_1(A) \cdot m_2(B)$$

# Transferable Belief Model by Smets

- Interpretation of the Dempster-Shafer theory of evidence
- Positive masses can be assigned to the empty set  $m(\emptyset) \geq 0$ 
  - Unnormalised orthogonal sum

$$m_{1,2}(C) = \cancel{K^{-1}} \sum_{A, B: A \cap B = C} m_1(A) \cdot m_2(B), \forall C \subseteq \Omega$$

- Unnormalised basic probability assignments

$$m : 2^\Omega \mapsto [0, 1] \quad m(\emptyset) = 0 \quad \sum_{A \subseteq \Omega} m(A) = 1$$

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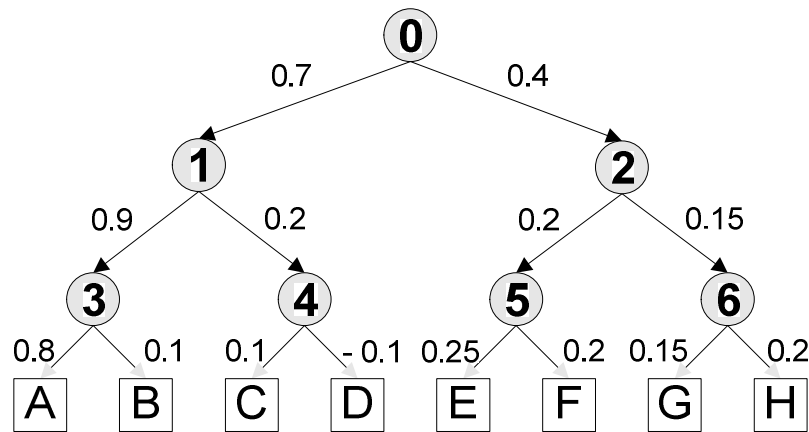
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# Evidence-Based Reasoning in Hierarchical Neural Network Classifiers

- Mass functions obtained from classifier outputs
  - Enforce positive classifier outputs
  - Normalise classifier outputs with sum greater than one
  - Discount classifier outputs
  - Assign belief to frame of discernment  $O$  (Representing ignorance)
- Combine mass functions of all classifiers
- Apply pignistic transformation to obtain final classification result

# Example

## Evidence-Based Reasoning

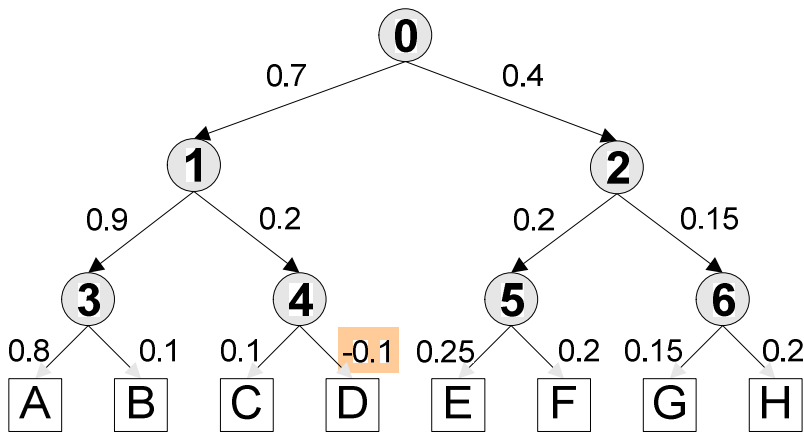


- Enforce positive outputs
- Normalise outputs  $> 1$
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- Combine evidence of all classifiers
- Calculate pignistic transformation
- Obtain combined classification result



# Example

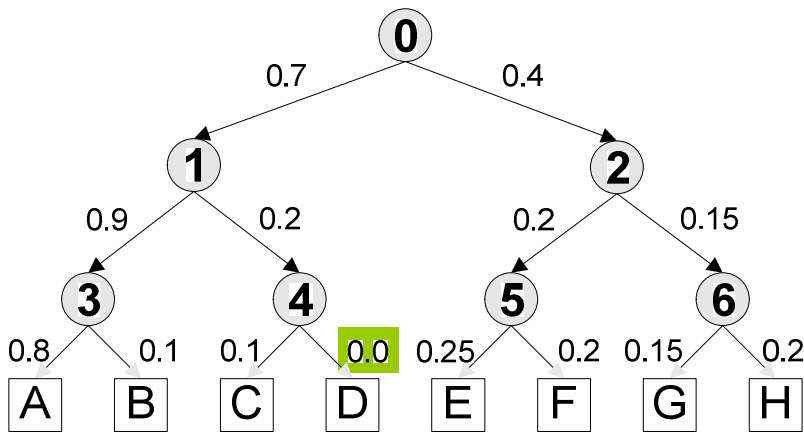
## Evidence-Based Reasoning



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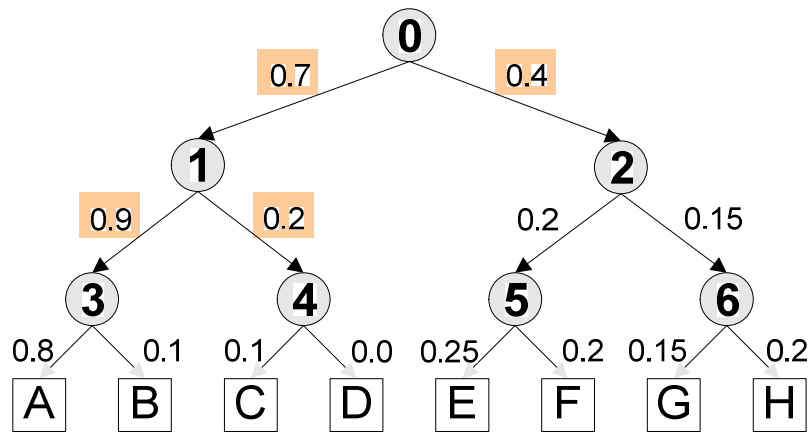
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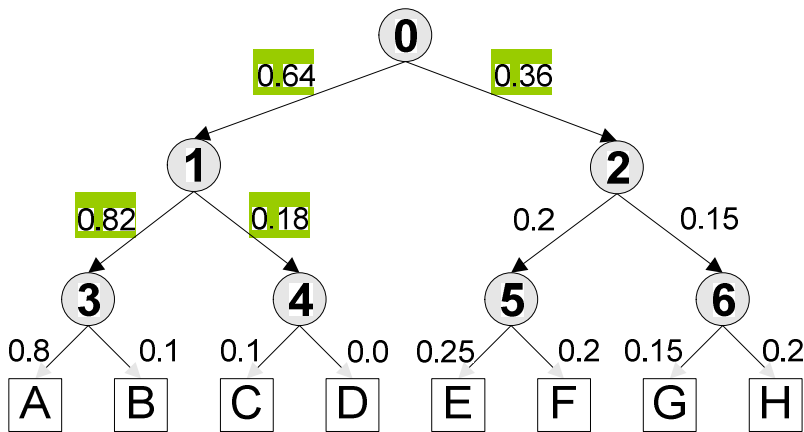
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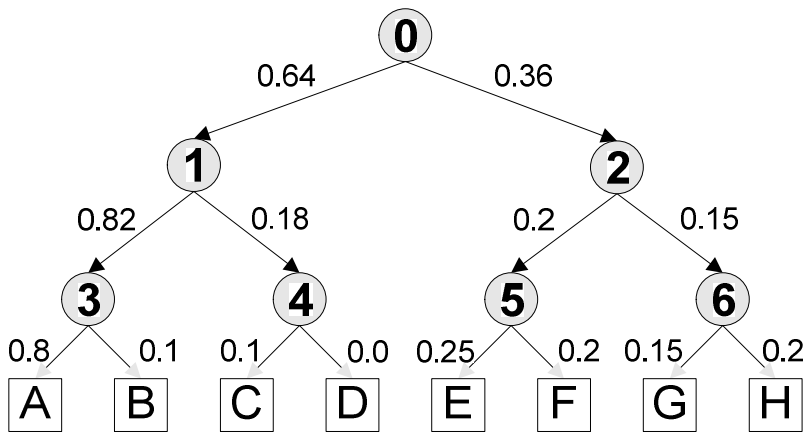
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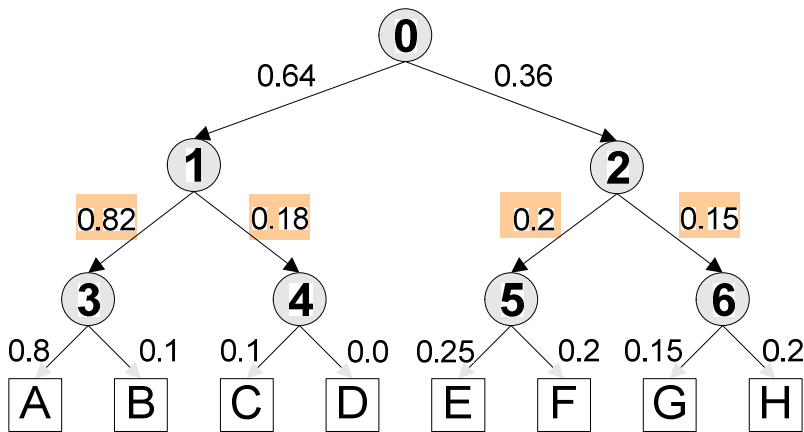
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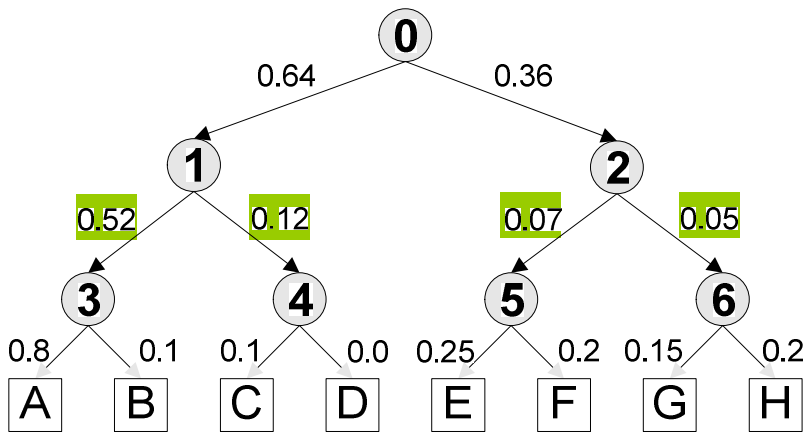
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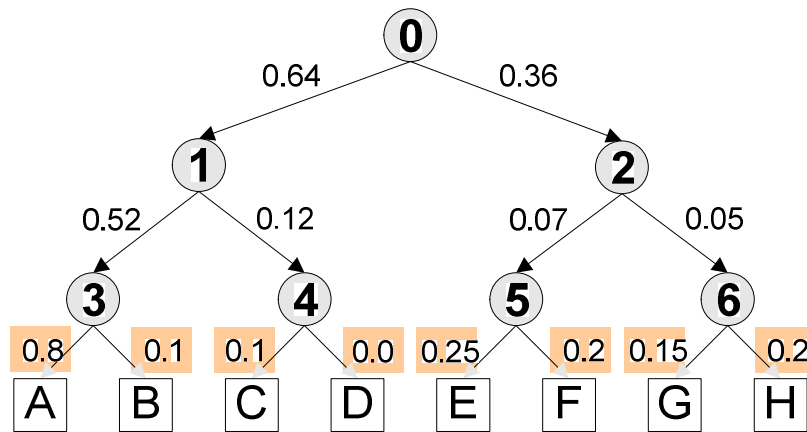
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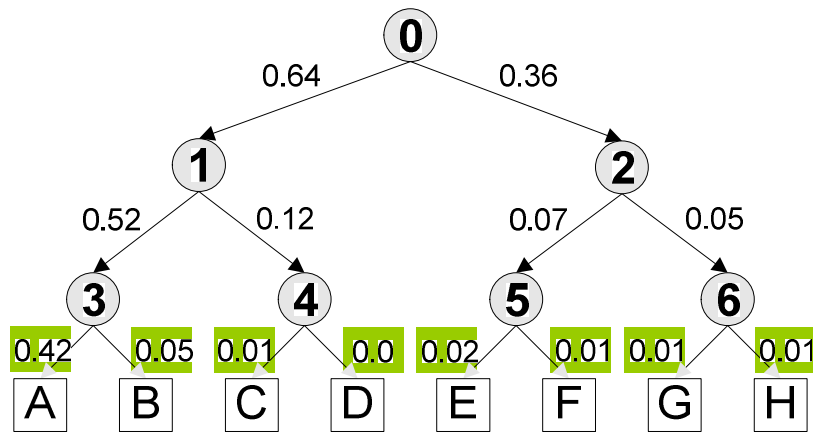


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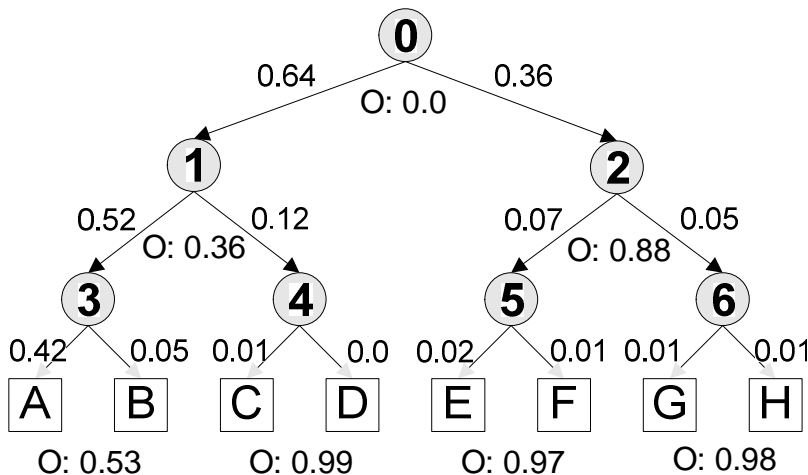
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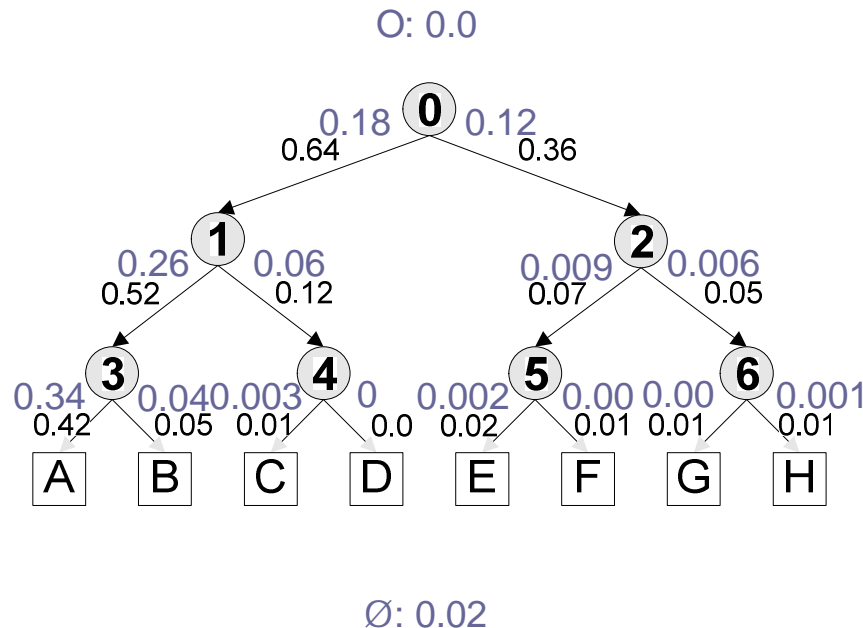
## Evidence-Based Reasoning



- Enforce positive outputs
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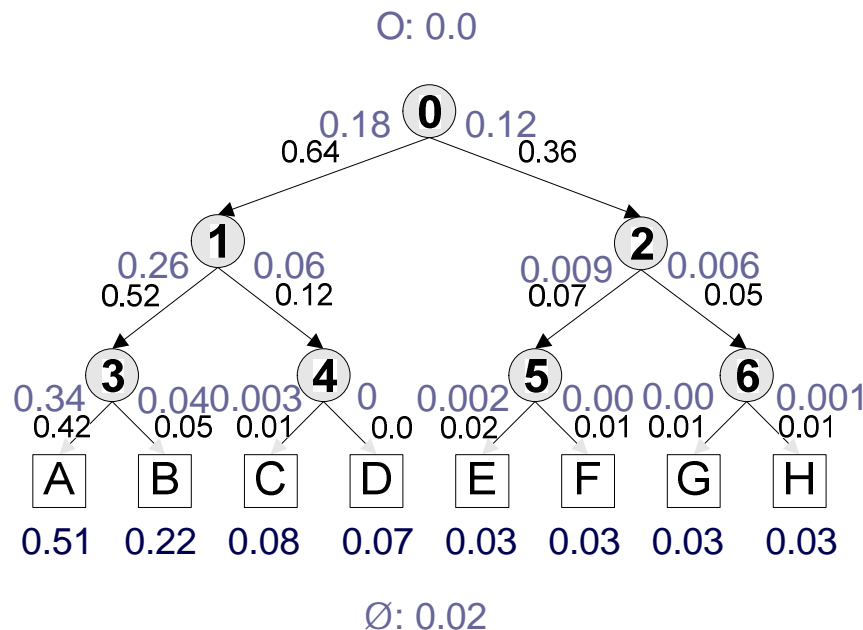
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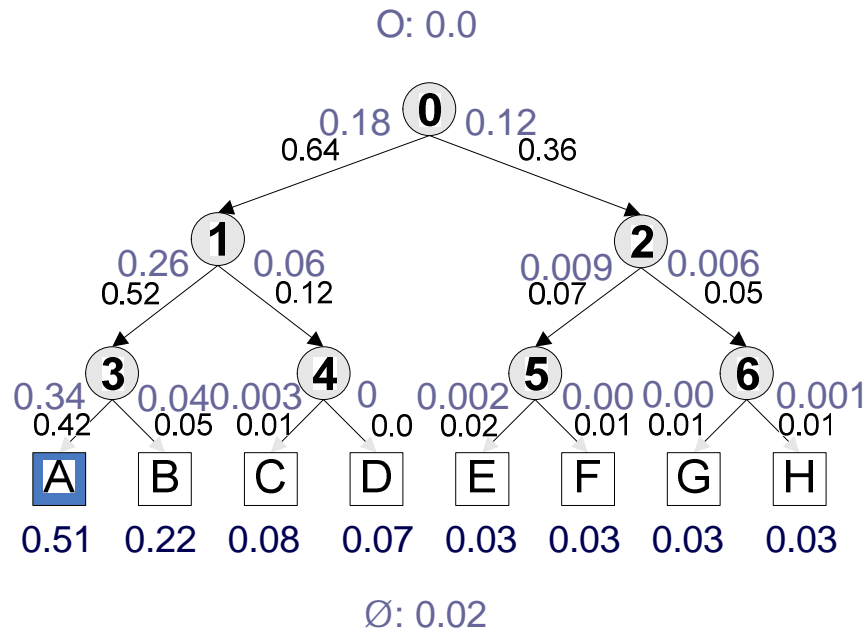
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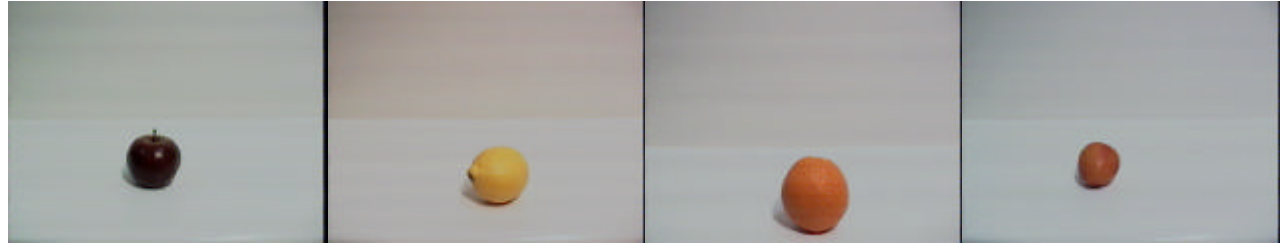
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# Data Sets

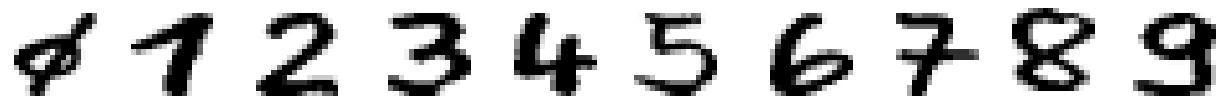
- Fruits



- Objects (COIL)



- Digits



- Letters



# Evaluation

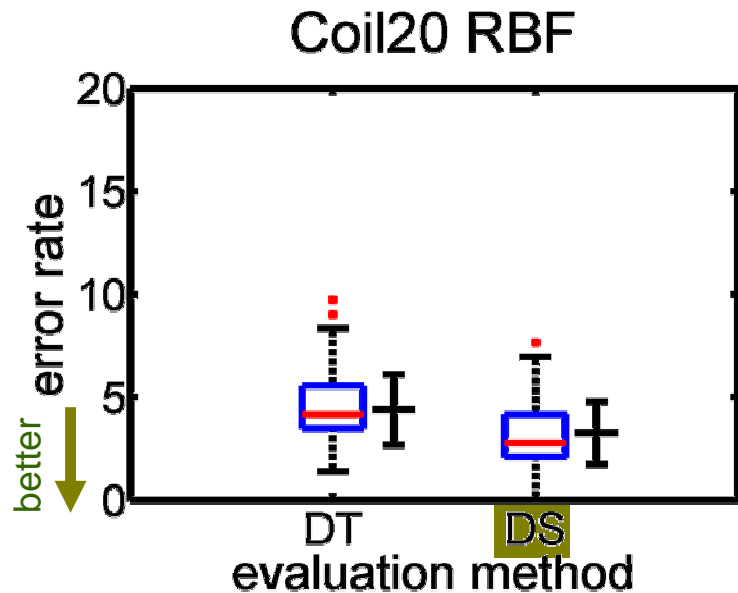
## Hierarchy Retrieval

- Compare **evidence theoretic** retrieval strategy and **decision tree** retrieval strategy
- Experiments on different data sets:
  - COIL-20
  - Digits
  - Letters
- Different neural classifiers:
  - RBF
  - Fuzzy k-NN classifier
- 10-times 10-fold cross-validation
- Both strategies are evaluated on the same hierarchy



# Results

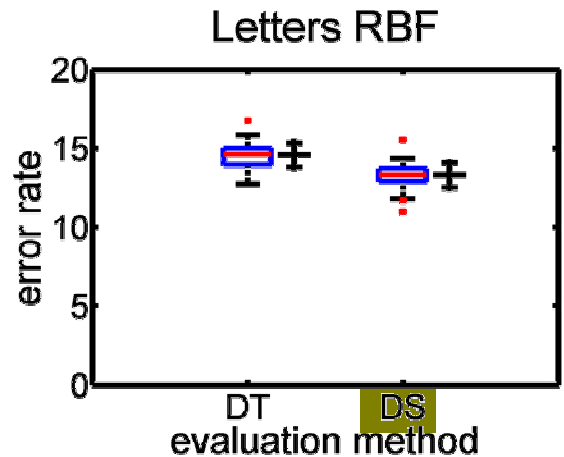
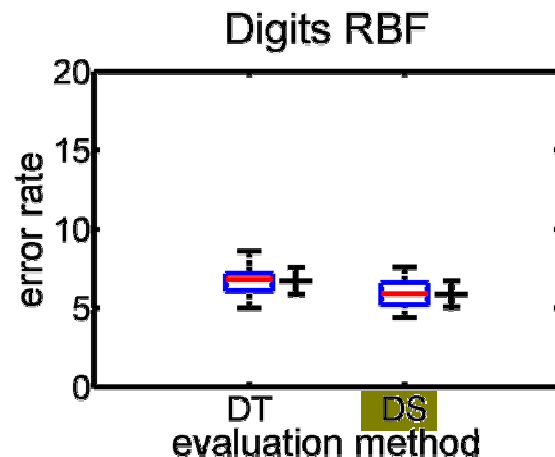
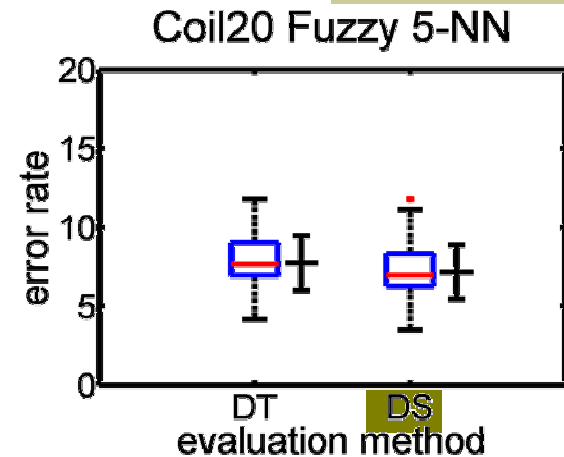
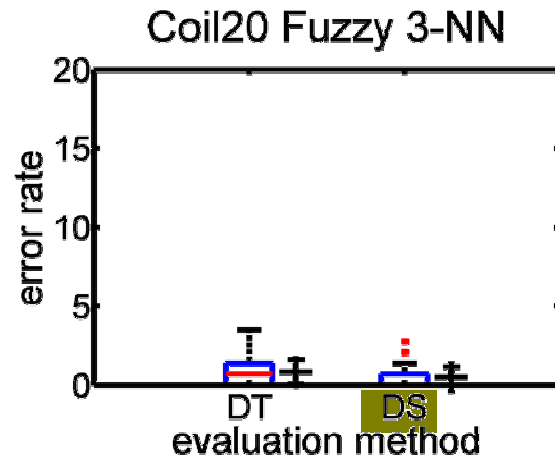
## Hierarchy Retrieval



- Data set:
  - COIL-20
- Classifier:
  - RBF
- Dempster-Shafer (DS):
  - Mean error: **3.24±1.52%**
- Decision tree (DT):
  - Mean error: **4.38±1.71%**
- p-value: 0.025
- t-value: 2.26
- Significant at 5%

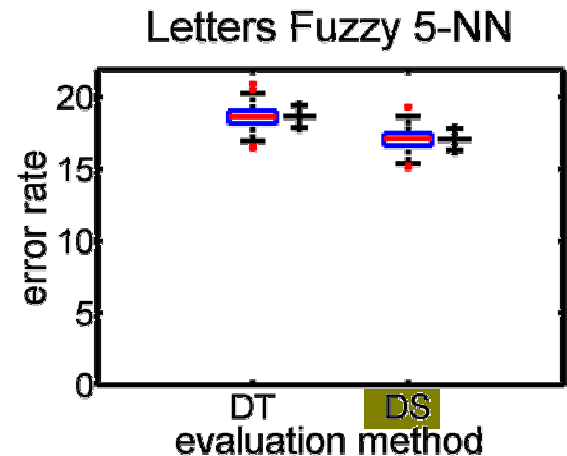
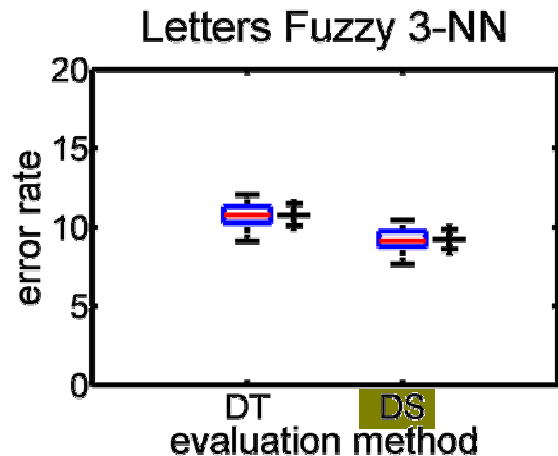
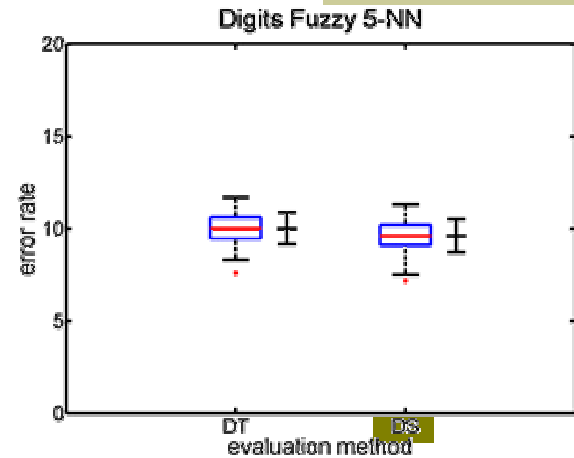
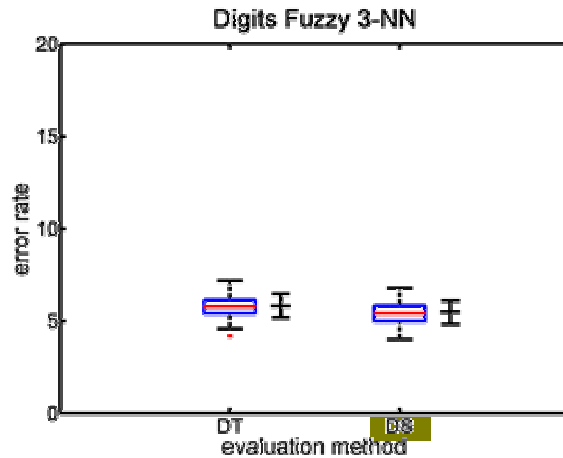
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# Conclusions

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- Dempster-Shafer theory of evidence is suitable for the application to hierarchical neural network classifiers
- Evidence theoretic hierarchy retrieval outperformed the decision tree retrieval in all tested cases



# Thank you for your attention!

This work was partially supported by the European Union  
award #IST-2001-35282 of the Mirrorbot project.