Weka tutorial and Inductive Logic Programming

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Weka – What you should know

- arff file format
 - loading files
 - looking at basic statistics of data
- classifying
 - picking classifier
 - setting parameters
 - methods of evaluation
- clustering
 - k-means
 - comparison to classification
- There is a lot more to weka than this

Announcements

- Projects Due Thursday
- Homework 5 due next Thursday
- Final
 - December 19th
 - Room 1325 (classroom)
 - 10:05am
 - Emphasis on information since midterm

Weka

- watch in class tutorial
- additional information at weka site: http://www.cs.waikato.ac.nz/ml/weka/

Inductive Logic Programming – Motivation

- So far all models
 - have used fixed length feature vectors
 - have a fixed, previously decided, class feature (except Bayesian networks)
- We have been learning propositional logic models
- But a lot of real world data doesn't fit this paradigm, you can make the data fit into a fixed length feature vector

Staring Example

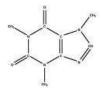
example of East-West trains (Michalski)

What makes a train go Eastward?

Real World Example

Mutagenicity of chemical molecules (King, Srinivasan, Muggleton, Sternberg, 1994)





What makes a molecule to be mutagenic?

ILP can use Multiple Tables

Many feature vectors

has_car					
Train	Car				
t1	c11				
t1	c12				
t1	c13				
t1	c14				
t2	c21				

car_properties							
Car	Length	Shape	Axes	Roof			
c11	short	rectangle	2	none			
c12	long	rectangle	3	none			
c13	short	rectangle	2	peaked			
c14	long	rectangle	2	none			
c21	short	rectangle	2	flat			

ILP Setup

- First Order Definite Clauses
- Learns a Theory of clauses (rules)
- Using:
 - Background Knowledge and
 - Examples (positive and negative)

ILP Algorithm

• ILP is a covering algorithm:

```
theory=empty
While (positives left uncovered):
   construct a rule
   theory=theory+rule
   remove positives covered by rule
return theory
```

Example

- Learn grandparent(X,Y) relation
- Background Knowledge:

father(philip, charles)
 mother(mum, margaret)
 married(diana, charles)
 father(philip, anne)
 mother(mum, elizabeth)
 married(elizabeth, philip)

male(philip)female(beatrice)female(margaret)

parent(X,Y):-father(X,Y).parent(X,Y):-mother(X,Y).

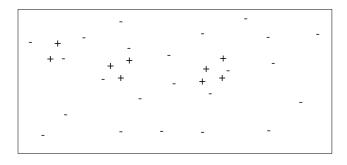
• Examples:

- grandparent(mum, charles) grandparent(elizabeth, beatrice)

- ¬grandparent(mum, harry), ¬grandparent(spencer, peter)

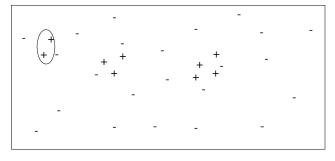
ILP Algorithm

• Covering Algorithms while (positives left uncovered)



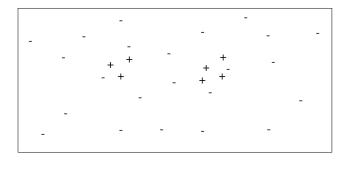
ILP Algorithm

• Covering Algorithms construct a rule theory=theory+rule



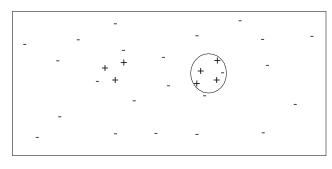
ILP Algorithm

• Covering Algorithms remove positives covered by rule



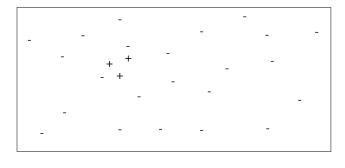
ILP Algorithm

• Covering Algorithms learn a rule theory=theory+rule



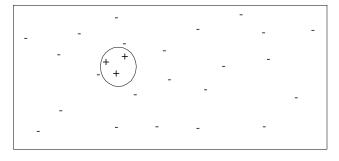
ILP Algorithm

• Covering Algorithms remove positives covered by rule



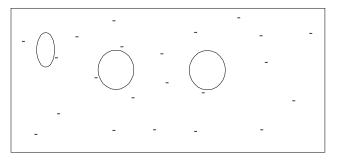
ILP Algorithm

• Covering Algorithms learn a rule theory = theory+rule



ILP Algorithm

• Covering Algorithms return theory



Learning a Rule

- Several Methods for learning Rules:
- Top Down Approaches
 - start with most general rule (cover everything)
 - specialize the rule until it covers only (mostly) positives
- Bottom Up Approaches
 - start with most specific rule (covers a single example)
 - generalize the rule as long as it doesn't cover any (to many) negatives

Learning a Rule

- Top Down Approaches
 - FOIL (1993)
 - PROGOL (1995)
- Bottom Up Approaches
 - GOLEM (1992)

FOIL

- Start with most general rule: grandparent(X,Y):-.
- Consider adding all possible predicates to the body of the rule:
 - grandparent(X,Y):-parent(X,Y). grandparent(X,Y):-parent(X,Z). grandparent(X,Y):-female(X).
- score each possible rule using foil_gain (similar to information gain)
- Keep the rule that has the highest score
- Repeat until you have a rule that covers only (mostly) positives

GOLEM

- Pick several pairs of positive examples
- Construct the Least General Generalization for each pair
 - List ALL the things the two examples have in common
- Select the one that covers the most positives
- Now select a new set of positive examples
- Construct the Least General Generalization between each new positive and the current rule
- Keep the one that covers the most positives
- Repeat as long as not covering any (to many) negatives

PROGOL

- Pick a positive seed example from the remaining uncovered positive examples:
 - grandparent(mary,john)
- Create a list of all true information about that seed example
 - female(mary), male(john), parent(mary,fred),male(fred), parent(george,mary),parent(fred,john),...
- Variabilize the information
 - female(M),male(J),parent(M,F),male(F),parent(G,M),
 parent(F,J)...
- Only consider adding variabilized predicates from this list (bottom clause)
- This guarantees that the seed example (at least) will be covered by the rule

Conclusion

- Weka
- Benefits of Inductive Logic Programming (ILP)
- Covering Algorithms
- top down / bottom up approaches to rule generation
- FOIL, PROGOL, and GOLEM