

RDN-Boost

A Guide

Task

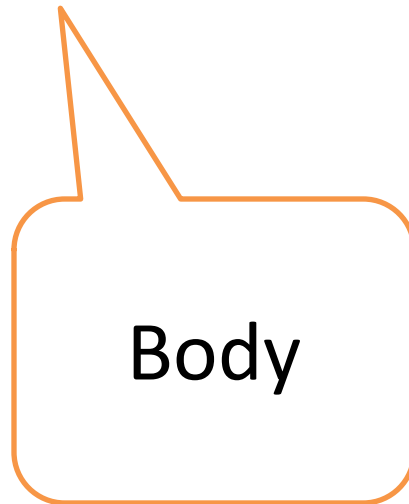
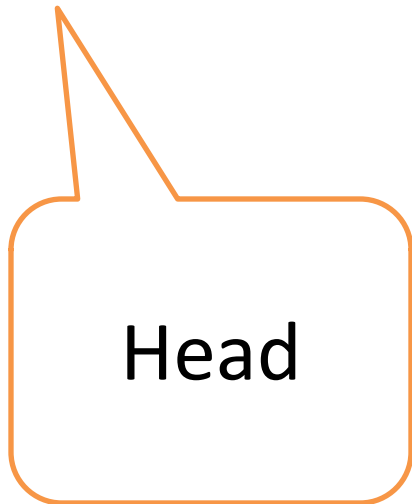
- Given:
 - train(t)
 - car(t, c)
- Todo:
 - Learn rules for target(t)

Facts

train(T1)
car(T1, C1)
train(T2)
train(T3)
car(T3, C2)
car(T3, C3)

Problem 1

- `target(X)` is true, if train has a car
 - `target(X) <- car(X,Y)`



Inductive Logic Programming

- Start with target(X)
 - target(X) <- car(X,Z)
 - target(X) <- car(Y,X)
 - Does not make sense since car has car id as the second argument and target has train id as the first argument
 - target(X) <- car(Y,Z)
 - Does not help since the rule says that a train is of target type if some train has a car

Provide type information

- To avoid `target(X) <- car(Y,X)` provide type information
- mode: `target(t)`
- mode: `car(t, c)`

- But what about `target(X) <- car(Y,Z)` ?

Modes to the rescue

- `car(t,c)` must use the current train variable
 - i.e. variable of type `t` should already be mentioned before
- `'+'` in a mode exactly does that
- But the variable of type `c` in `car` may not be seen before
- `'-'` in a mode exactly does that
mode: `car(+t, -c)`

Problem 2

- Additional facts
 - `big(c)`
 - `small(c)`
- `target(X)` is true if there is a big car and a small car in the train
 - `target(X) <- car(X,Y) , big(Y), car(X,Z), small(Z)`

Modes

- `car(+t, -c)`
- `big(+c)`
- `small(+c)`

- `big(-c)` would give us rules like
`target(X) <- big(Y)`

ILP search

- Target(X)
 - car(X,Y)
 - big(Y)
 - car(X,Z)
 - » small(Z)
 - small(Y)
 - small(Y)
 - car(X,Z)

But ILP is greedy search

- Target(X)

- car(X,Y)

- big(Y)

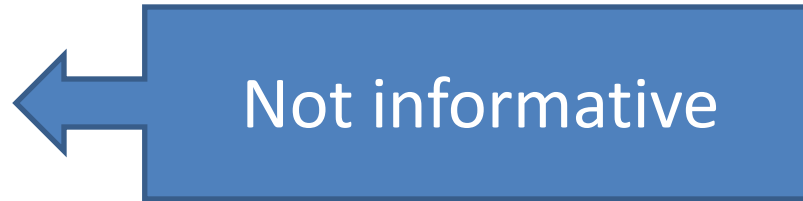
- car(X,Z)

- » small(Z)

- small(Y)

- small(Y)

- car(X,Z)



Don't be so greedy

- Increase lookahead to 2
- Target(X)
 - car(X,Y), big(Y)
 - car(X,Z), small(Z)
 - ...
 - ...
- setParam: nodeSize=2.

Problem 3

- Add facts
 - $\text{animal}(c, a)$
 - $a = \{\text{Dog, Cat, Mouse}\}$
- Target(X) if a car contains mouse

Possible rules

- Mode: animal(+c, -a)
- target(X) :- car(X, Y), animal(Y,A)
- We need to “ground” the variable A

#UseHash

- Mode: `animal(+c, #a)`.
- Generated clauses
 - `target(X) :- car(X, Y), animal(Y,"Dog")`
 - `target(X) :- car(X, Y), animal(Y,"Cat")`
 - `target(X) :- car(X, Y), animal(Y,"Mouse")`
- Still need `nodeSize=2`


Problem 4

- A big car contains a mouse
 - $\text{Target}(X) \text{ :- } \text{car}(X,Y), \text{big}(Y), \text{animal}(Y, \text{“Mouse”})$.
- Consider ILP search after $\text{target}(X) \text{ <- } \text{car}(X,Y), \text{big}(Y)$
 - $\text{small}(Y), \text{animal}(Y, \text{“Dog”})$
 - $\text{small}(Y), \text{animal}(Y, \text{“Cat”})$
 - $\text{animal}(Y, \text{“Dog”}), \text{animal}(Y, \text{“Cat”})$
 - ...
- small, big, animal : are informative
- car is not

Use bridgers

- Bridgers connect facts
 - E.g. age, parents, segment
- Bridgers should not be counted
 - Infinite bridgers : $\text{car}(X,Y)$, $\text{car}(X,Z)$, $\text{car}(X,A)$...
- First bridger is free
 - $\text{car}(X,Y)$ – size:0
 - $\text{car}(X,Y)$, $\text{big}(X)$ – size:1
 - $\text{car}(X,Y)$, $\text{animal}(X, \text{"Dog"})$ – size:1
 - $\text{car}(X,Y)$, $\text{car}(X,Z)$ – size:1
- bridger: $\text{car}/2$.
- Keep $\text{nodeSize}=1$.

Citeseer

- Citation segmentation
- Given: 
 - token(c, t)
 - punctuation(t)
 - wordString(t, w)
 - next(t, t)
- Todo:
 - field(t, f) f={author, title, venue}

Multi-valued classification

- Learn one model for each label
- Change n-valued classification into n binary classification models
 - infield_title(t)
 - infield_author(t)
 - infield_venue(t)

Joint model

- Model/Rules for `infield_title` might be useful for `infield_venue` and vice versa
 - `infield_title(T) <- next(T,P), punct(P), next(P,T1), infield_venue(T1)`
- Specify all three predicates as query predicates
 - query `infield_venue,infield_title,infield_author`
- During inference, pick the most likely label
 - Has to be a post-processing step. Not their in code

Cora

- Citation clustering
- Given:
 - Title(b, t)
 - Author(b,a)
 - Venue(b, v)
 - TitleWord(t, w)
 - AuthorWord(a,w)
 - VenueWord(v,w)
- Todo:
 - sameBib(b, b)

Transitivity

- We might want the model to learn rules like
 - `sameBib(X,Y) <- sameBib(X,Z), samebib(Z,Y)`
- If we use `sameBib(+b, -b)`
 - `sameBib(X,Y) <- sameBib(X,Y)`
- The rule is perfect but not really useful for inference
- Force one variable to not be in head of clause
 - `sameBib(`b, +b)`
 - `sameBib(+b, `b)`

Code issues

- Cannot handle same predicates in head and body
- RDN-Boost will create recursive_<predicate> automatically
 - recursive_sameBib for Cora
- Specify modes as
 - recursive_sameBib(`b, +b)
 - recursive_sameBib(+b, `b)

Greedy search issues

- Intuitively a good rule would be
 - $\text{sameBib}(X, Y) \leftarrow \text{Title}(X, T1), \text{Title}(Y, T2), \text{sameTitle}(T1, T2).$
- No subset of predicates is “informative”
- Needs a node size of 3 or $\text{Title}/2$ as bridger

Mode overview

- + : variable must have appeared before
- - : variable can be new but does not have to
- # : ground the variable/use constant
- ` : variable must not be in the head
- @<val>: variable must take value <val>



Tree representations

- `<train_folder>/models/`
 - `bRDNs/Trees/*tree` : Trees as list of clauses
 - `bRDNs/dotFiles/*dot`: .dot files that can be used by graphViz to visualize
 - `WILLtheories/*txt` : Prolog format for trees. Also has human readable text version of all trees in one file

Sample tree

%%%%% WILL-Produced Tree #1 @ 0:51:19 10/20/10. [Using 13,141,856 memory cells.] %%%%%%

```
% FOR advisedby(A, B):
%   if ( professor(B) )
%     then if ( professor(A) )
%       | then return -0.1418510649004878; // std dev = 0.000, 7.000 (wgt'ed) examples reached here. /* #neg=7 */
%       | else if ( publication(C, B) )
%         | | then return 0.739727882467934; // std dev = 0.323, 76.000 (wgt'ed) examples reached here. /* #neg=9 #pos=67 */
%         | | else return 0.3781489350995123; // std dev = 0.500, 25.000 (wgt'ed) examples reached here. /* #neg=12 #pos=13 */
%     else return -0.1418510649004879; // std dev = 0.000, 132.000 (wgt'ed) examples reached here. /* #neg=132 */
```

Additional flags

- modelSuffix
 - Run multiple experiments with different values for this flag to prevent overwriting
- negPosRatio(default=2)
 - Each boosting iteration samples negative examples so that negative:positive ratio is 2:1
 - Most datasets have too many negatives

Tree parameters

