Manual for Join Bayes Net System

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Contents

1	Join Bayes Net Package	1
	1.1 Overview	1
	1.2 Join Bayes Net	2
2	Markov Logic Network Evaluation Package	5
	2.1 Overview	5
	2.2 Inference	5
	2.3 5-fold Cross-validation	6
3	Datasets	8
	3.1 Databases	8
4	Useful Scripts	9
	4.1 Generator	9
	4.2 Statistics	

1 Join Bayes Net Package

1.1 Overview

Join Bayes $Nets^1$ is a software package providing a series of learning algorithms for statistical relational learning and probabilistic inference. This package is written in Java and based on the open-source *The Tetrad Project*² and used $Weka^3$ for decision learning. Tetrad and Weka are open-source packages, of which the source codes are imported into our package.

Package jbn

Operating System Unix/Linux

Environments Java version 1.6 or greater

MySQL installation

 $^{^1\}mathrm{Learning~BNs~For}$ Relational Data http://www2.cs.sfu.ca/ \sim oschulte/jbn/

²The Tetrad Project http://www.phil.cmu.edu/projects/tetrad/

³Weka 3: Data Mining Software in Java http://www.cs.waikato.ac.nz/ml/weka/

1.2 Join Bayes Net

The source code for our join Bayes Net system can be found in the jbn folder. There are several parts in this package, structure learning, parameter learning of join Bayes Net, Markov Logic Network exporting and decision tree learning. To make the package work, first we need to configure some global settings. The global parameters of the system is defined in this file:

ca.sfu.jbn.common.global.java

You can change the default name of database, the URL, username and password of the database server and some other useful settings. Although it may be impossible to go through each package one by one in the source code, there are some important packages (classes) you should pay attention to. These packages are mentioned below:

$ca.sfu.jbn. Parameter Learning_main.java$

Main entry for structure and parameter learning of join Bayes Net; when you want to use the structure and parameter learning code, export a jar file that takes this class as main entry (argument formats are mentioned as comments in the source file)

ca.sfu.jbn.MLN_ParameterLearning_main.java

Main entry for Markov Logic Network exporting from learned Bayes Net, further used for MLN evaluation. Typically, we use this class together with the Markov Logic Network Evaluation Package. The *evaluate/jbn.jar* file is an exported runnable Jar file of the whole project launched from this class.

ca.sfu.jbn.MLN_DecisionTreeLearner_main.java

Main entry for Markov Logic Network exporting with Weka decision tree learner. Similar with MLN_ParameterLearning_main.java, it is also used for MLN evaluation. Typically, we use this class together with the Markov Logic Network Evaluation Package. The evaluate/jbndt.jar file is an exported runnable Jar file of the whole project launched from this class.

ca.sfu.jbn.structureLearning

Package which contains all the codes related to structure learning of join Bayes Net.

ca.sfu.jbn.parameterLearning

Package which contains *not all* the codes related to parameter learning of join Bayes Net.

- 1. ParamTet.java provides a main entry for parameter learning (As you see, it is kind of messy of these main entries for some historical reasons, do not bother too much).
- 2. Decisiontree.java is the most important code for decision tree learner, in

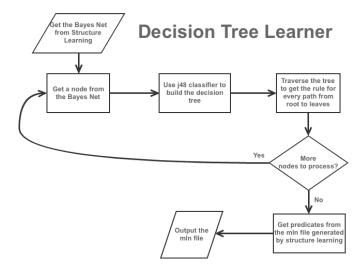


Figure 1: Flow chart to illustrate the process of decision tree learner

which we use Weka J48 decision learner to learn the tree structure and use the frequencies from the database to calculate the weights for each clause in MLN. This class is directly called from main entry of the learner ca.sfu.jbn.MLN_DecisionTreeLearner_main.java.

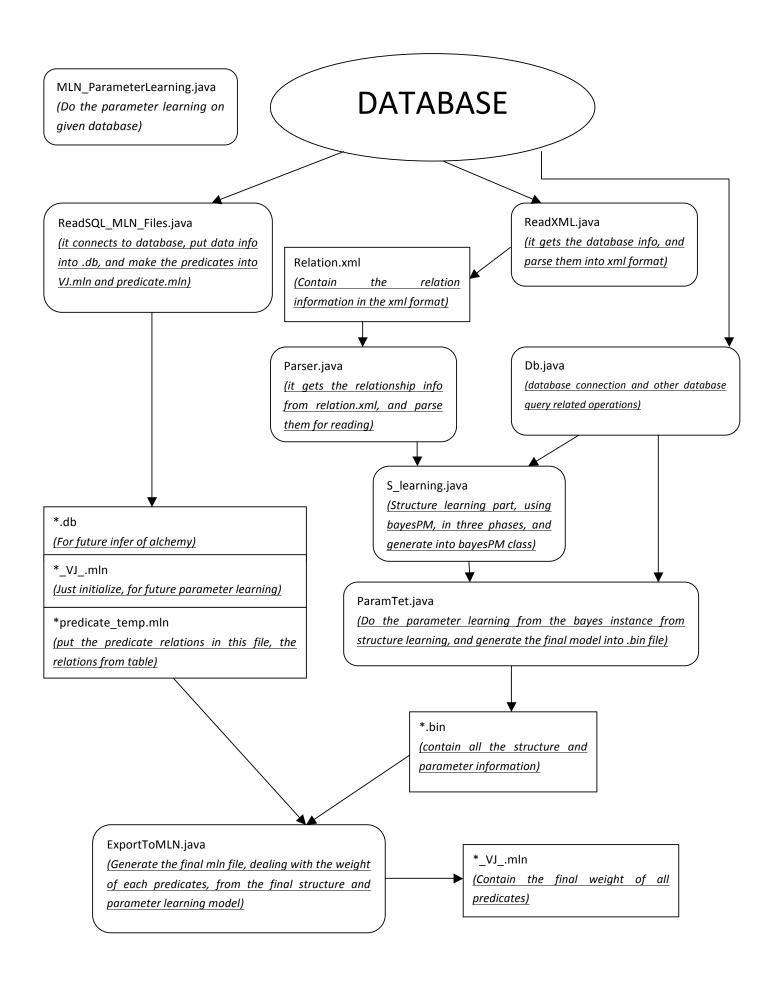
You may want to take a look at the package *ca.sfu.jbn.frequency* to know better about the process of parameter learning.

ca.sfu.jbn.frequency

Very important package for parameter learning. BayesStat.java is the key code to obtain frequency and probability statistics directly from the database by doing SQL queries. One thing to notice: there are some constraints for this code. It did work well for our research now, but you may have to modify it to satisfy new requirements (it may become buggy when there are multiple negative relations in your query).

edu.cmu.tetradapp.Tetrad.java

Tetrad has a GUI application for you to play with. We have done a lot of work to get our JBN system integrated with this user interface. Try this if you are interested. An exported Tetrad Jar file can be found in tetrad/SFUTetrad.jar.



2 Markov Logic Network Evaluation Package

2.1 Overview

This package is to help evaluate the structure and parameters of a learned Markov Logic Network file (.mln in the Alchemy⁴ format). The MLN evaluation package outputs statistics, such as the average accuracy, conditional log-likelihood, over all predicates in the dataset.

Package evaluate
Operating System Unix/Linux

Environments Java version 1.6 or greater

Korn shell (ksh)

Complied binaries of Alchemy (included in the package)

2.2 Inference

There are several parts of this package, of which each part is relatively independent with others. The MLN inference code is written in Java, you can find the code in *infer* folder. Here are some important packages (classes) in this code:

MLN.MLN.inference

Main entry for the MLN inference code, used for MLN evaluation. Typically, we use this class together with *evaluate* package. The *evaluate/evaluation/infer.jar* file is an exported runnable Jar file of the whole project launched from this class.

Usually, we do not use the *infer* project alone, it is just one part of the whole evaluation system.

Now, what is the most important to know about this package is how to get the codes work. First, we need to know the functionality of these files in the evaluate folder:

1. test.sh

Korn shell script for evaluating the performance of different methods by training and testing on a single database. This is one of the main entry of the whole system, of which the argument format will be introduced later.

2. testcross.sh

Korn shell script for evaluating the performance of different methods by training and testing on 5-fold cross-evaluation. This is the other one of the main entry of the whole system, of which the argument format will be introduced later.

3. learnwts

Compiled code for Alchemy MLN weight learner. We can get the code on

 $^{^4 {\}rm Alchemy}$ - Open Source AI http://alchemy.cs.washington.edu/

the website, but if you do not want to waste several weeks on changing and compiling it, just use this one.

4. jbn.jar

As we have seen, this is an exported Jar file of the Join Bayes Net package launched from *ca.sfu.jbn.MLN_ParameterLearning_main.java*.

5. jbndt.jar

As we have seen, this is an exported Jar file of the Join Bayes Net package launched from *ca.sfu.jbn.MLN_DecisionTreeLearner_main.java*.

6. evaluation/infer.jar

As we have seen, this is an exported Jar file of the Inference package launched from MLN.MLN.inference.

7. evaluation/lib

This is a folder that contains some ancient codes used for analyzing the evaluation results and generating measurement statistics such as accuracy and log-likelihood. Some of the codes come from the *ca.sfu.jbn.Analyzer* of the *Inference* package. However, I have never read through them, why not just put there and use them?

8. config.xml

Global configuration setting file that contains the database server, username and password.

Thus, we can use *test.sh* and *testcross.sh* to run the evaluation system. The argument format of these two scripts are the same; therefore, I will just explain the cross-validation script *testcross.sh*.

2.3 5-fold Cross-validation

Before we begin to use the script, make sure you have set up the cross-validation databases. Since 5-fold cross-validation is used, we create 5 training databases and corresponding 5 testing databases, which are a subset of the whole databases. The scripts in the cross folder are used to generate these databases, which will be explained later.

The names of these training and testing databases are derived from the whole database. For example, for a database named *MovieLens*, the generated training databases are named with suffix _*Training#* and _*Test#*:

 $MovieLens_Training1, MovieLens_Training2, \ldots, MovieLens_Training5,$ and the testing databases are named in a similar way:

 $MovieLens_Test1, MovieLens_Test2, \dots, MovieLens_Test5.$

After we set up these databases, we can finally start our experiments. We use the following command to run a cross-validation experiment:

 $$ testcross.sh \ dbName \ method \ geo-parameter$

Here is a detailed description of these three arguments:

1. dbName

The name of the whole database you want to do cross-validation on. For instance, the cross-validation databases are *MovieLens_Training#* and *MovieLens_Test#*, you should use the name *MovieLens*.

2. method

You have many options on *method* parameter. This parameter determines how to generate clauses and how to get the weights of clauses in MLN file. Here is a list:

(a) **mbn**

To export MLN structure file from Bayes Net and use Alchemy weight learner mbn to learn the weights of clauses in MLN file

(b) dtmbn

To export MLN structure file using decision tree learner from Bayes Net and use Alchemy weight learner *learnwts* to learn the weights of clauses in MLN file

(c) **log**

Deprecated To export weighted MLN file from Bayes Net, here we obtain the weight by taking log of the conditional probability

(d) dtlog

Deprecated To export weighted MLN file using decision tree learner from Bayes Net, here we obtain the weight by taking log of the conditional probability

(e) **lsn**

To export weighted MLN file from Bayes Net. Similar with *log*, here we obtain the weight by taking log of the conditional probability and then subtracting the log of unit prior

(f) dtlsn

To export weighted MLN file using decision tree learner from Bayes Net. Similar with log, here we obtain the weight by taking log of the conditional probability and then subtracting the log of unit prior

3. geo-parameter

This parameter determines whether to use *geometric mean* when we do MLN inference. There are two options:

(a) **geo**

To using *geometric mean* when we do inference; the performance test usually favors this option

(b) **nom**

To disable geometric mean option

The content of the output contains statistics for each run in the 5-fold cross-validation, including the accuracy, log-likelihood for each attribute and overall accuracy and log-likelihood in average with regard to attributes.

Here is an example to illustrate how to run a 5-fold cross-validation on *Movie-Lens* database (using *mbn* method and geometric mean):

\$ testcross.sh MovieLens mbn qeo

3 Datasets

3.1 Databases

Our datasets are some real-world and synthesized databases. There are five real-world databases⁵ that we most commonly use in our experiments. You can find more details about these datasets from the references in our published papers and on-line sources. *Schema* is the name of database in our database server.

1. MovieLens Database

Schema: MovieLens

This is a standard dataset from the UC Irvine machine learning repository.

2. Mutagenesis Database

Schema: Muta

This dataset is widely used in ILP research. It contains information on Atoms, Molecules, and Bonds between them.

3. Hepatitis

Schema: hepelwin

This data is a modified version of the PKDD02 Discovery Challenge database. The database contains information on the laboratory examinations of hepatitis B and C infected patients.

4. Mondial

Schema: mondelwin

This dataset contains data from multiple geographical web data sources. We used a subset of the tables and features for fast inference.

5. Financial

Schema: fin

A dataset from the PKDD 1999 cup.

 $^{^5\}mathrm{Some}$ of these databases are slightly modified from the original ones to meet specific requirements of our system

6. **UW**

Schema: uw

A dataset from University of Washington, containing information of faculty, students and courses.

The backup of these databases are stored in the *databases* folder. For convenience, each table is stored in one *.sql* file. Please always make sure you have the backups before you modify these databases.

4 Useful Scripts

4.1 Generator

In some of our experiments, we have to generate sub-databases from the original databases; or, make synthesized databases for some special purpose. The most frequently-used generator is to generate the 5-fold cross-validation sub-databases; other generators include the scripts to generate synthesized testbed used in our *parameter learning* evaluation.

Script Name cross Operating System Unix/Linux

Environments Java version 1.6 or greater

Korn shell (ksh)

Description Generate the 5-fold cross-validation databases. The name

of the cross-validation databases are named by appending a suffix to the name of the original database. For example, the original database is *MovieLens*, it will create ten databases (five training databases and five test databases) with names: *MovieLens_Training[1-5]* and

 $MovieLens_Test[1-5]$

Example See the file *example.txt* in the same folder to get the idea

Script Name TestbedGenerator.py

Environments Python version 2.4.3 or greater

Description deprecated This script is used to generate a synthesized uni-

versity database containing information about students and courses. We can control the number of students and courses

in the databases and the ratio of enrollment.

Example \$ python TestbedGenerator.py <Students> <Courses>

4.2 Statistics

We have some scripts to help us handle with the outputs of our systems. You can find them in the *scripts* folder. These scripts are very simple and most of them are written in *Python*. Be free to use them to simplify your work.

Script Name clause-count.py

Environments Python version 2.4.3 or greater

DescriptionCount the number of clauses in a MLN fileOutputThe number of clause in this MLN fileExample\$ python clause-count.py < MovieLens.mln</th>

Script Name clause-display.py

Environments Python version 2.4.3 or greater

Description Print out all the clauses in a MLN file
Example \$ python clause-display.py < MovieLens.mln

Script Name cross-stat.py

Environments Python version 2.4.3 or greater

Description Very useful. Print out statistics of the cross-validation re-

sult, including accuracy, log-likelihood for each attributes, also with average accuracy, average log-likelihood and stan-

dard deviation

Example First, we need to save all the output generated in one

cross-validation experiment into a temporary file, say

cross.output. Then run with this command:
\$ python cross-stat.py < cross.output</pre>

10