

Course Project:
A Symbiotic Relation between Big Data and Big Compute:
The Fraud Detection Case Study – Stage 2 Update

Description:

As some of you may have heard during one of this semester's Distinguished Lectures, DOE's Lucy Nowell mentioned that there exists a symbiotic relation between the emerging field of **Big Data** and the established field **Big Computing** (i.e. Supercomputing). In this project, you will learn such a close relationship through a specific case study: ***On-Line Fraud Detection***.

On-Line Fraud Detection involves identifying a fraud as soon as it has been perpetrated into the system. This technique is only successful by having a training algorithm that can produce a model suitable to be used by a real-time detector. In this project, we will focus on fraud detection for credit card transactions (see Figure 1), using ***Markov chains*** to train the model off-line and a ***parallel implementation*** of a concurrent queue for on-line detection.

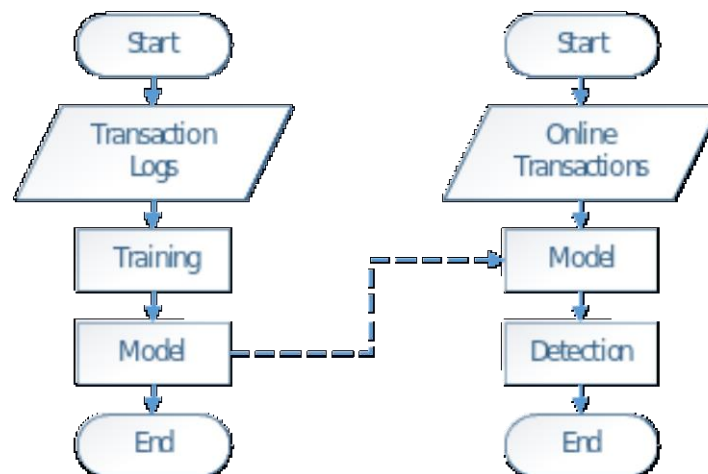


Fig 1. Offline training and online detection

The project will be divided in three stages to give the opportunity to students to make partial progresses towards a final presentation and report due at the end of the semester. This document presents the general project's overview and Stage 1's description. Other stage's descriptions will be posted in upcoming weeks.

Students will be divided in the following ***three groups***, with each group having ***two mentors*** and a ***leader*** (which will be nominated by the groups themselves). It is expected students have weekly meetings with their mentors to address any questions and doubts about the project.

The groups and their mentors are as follows:

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The stages for the project are the following:

Stage 1:

- **Title: Literature Review**
- **Objective:** Students will perform a background review on two relevant topics: Markov models and Hadoop framework. To be more efficient, each group may divide itself in subteams of 2 students for this review. Although each subteam will take charge of one topic, the results of their reviews should be communicated internally to all group's members.
- **Result:** Students will submit the answers to the following questions before **October 31st at 23.59.**

Markov Models:

A Markov chain (MC) is a stochastic process with the Markov property. The term "Markov chain" refers to the sequence of random variables such a process moves through, with the Markov property defining serial dependence only between adjacent periods (as in a "chain"). It can thus be used for describing systems that follow a chain of linked events, where what happens next depends only on the current state of the system [1].

This project trains a MC (unsupervised learning) to create a model that a fraud detection system can use to identify potential transaction outliers. The resulting model will be used to classify sequence of transactions as normal or anomalous.

In this part of the project your job is to get familiar with the basic concepts of Markov Chain and its application for this particular implementation of fraud detection. This process includes the reading of scientific papers, the proposal of a serial pseudo code for both the training and classification, the answering to some questions, and the creation of a report.

1. Background
 - a. Read the following documents
 - i. Reference [2] is the main reference for the implementation of both the training and the classification. In reading this reference, you may wish to focus on section 2 for the outline of the methodology, section 3.1 for the training of the model, and section 3.2 for the classifier.
 - ii. Additional references for estimating the Markov transition matrix are provided in [3], [4]. In reference [3], focus on section 2.A (Transition Matrices When Individual Transitions Known) and appendix A (Estimating Transition Probabilities with Observed Transitions) for the training of the model. In reference [4], focus on section 1.2 (Markov chains) and 1.3 (Training algorithm).
 - b. 1.2. For the report
 - i. Write down a small summary about Markov chain. Within your summary take into account:
 1. Definition
 2. Properties
 3. Relationship between past, present, and future (states)
 4. Transition matrix
 5. Representation as a graph (vertex and edges)
 - ii. With your own words explain the methodology presented in [2] for fraud detection.
 - iii. With your own words explain the process to train the Markov chain.
 - iv. With your own words explain the local classifier and the miss-rate metric.
 - v. When possible add some working example to make clear your explanations.
2. From the readings, propose a serial pseudo code for the following:
 - a. Training the model (estimating the Markov chain transition matrix).
 - b. Creating a local classifier when the miss-rate metric is used.

Explain clearly the steps of the algorithms and refer to the background concepts.
3. Answer the following question about Markov Chains:
 - a. Decide whether each of the following vectors could be a probability vector.
 - i. $[0.4, 0.2, 0]$
 - ii. $[\frac{1}{4}, \frac{1}{8}, \frac{5}{8}]$
 - iii. $[0.07, 0.04, 0.37, 0.52]$

iv. $[0.3, -0.1, 0.8]$

- b. Decide whether each of the following matrices could be a transition matrix, by definition. Sketch a transition diagram for any transition matrices.

$$\begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$$

$$\begin{bmatrix} \frac{2}{3} & \frac{1}{3} \\ 1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} \frac{1}{4} & \frac{3}{4} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$\begin{bmatrix} \frac{1}{4} & \frac{3}{4} & 0 \\ 2 & 0 & 1 \\ 1 & \frac{2}{3} & 3 \end{bmatrix}$$

- c. Think about the following sentence: “for a Markov chain, the conditional distribution of any future state X_{n+1} , given the past states X_0, X_1, \dots, X_{n-1} and the present state X_n , is independent of the past states and depends only on the present state.” [5]. What are the implications of this sentence when designing the training algorithm? HINT: consider a toy training sequence of states $Q = q_1, q_2, \dots, q_L$, and how the training algorithm should “analyze” the sequence.

References

- [1] Markov chain. (2014, October 16). In Wikipedia, The Free Encyclopedia. http://en.wikipedia.org/w/index.php?title=Markov_chain&oldid=629904997
- [2] S. Jha, K. Tan, and R. A. Maxion, “Markov chains, classifiers, and intrusion detection,” in Proceedings. 14th IEEE Computer Security Foundations Workshop, 2001., 2001, pp. 206–219. http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=930147&tag=1
- [3] Jones, Matthew T. Estimating Markov transition matrices using proportions data: an application to credit risk. International Monetary Fund, 2005. <https://www.imf.org/external/pubs/cat/longres.aspx?sk=18387.0>
- [4] <http://www.eecis.udel.edu/~lliao/cis841s06/hmmtutorialpart1.pdf>
- [5] Ross, Sheldon M. Introduction to probability models. Academic press, 2006.

Hadoop:

Apache Hadoop is an open-source software framework for storage and large-scale processing of data-sets on clusters of commodity hardware. MapReduce is a software framework for easily writing applications which process vast amounts of data in-parallel on large clusters (thousands of

nodes) in a reliable, fault-tolerant manner. Please follow installation tutorial[1] and set up Hadoop(version 2.5.0) environment on a single-node in a pseudo-distributed mode. Read the MapReduce paper [2] and tutorial [3] for basic understanding, and [4] for more details if necessary. Compile and run the WordCount v1.0 example in [3].

(1) Explain using your own words MapReduce programming model, using WordCount as example.

(2) Submit a WordCount v1.0 source code with your comments;

(3) - Considering the API definition for a Mapper and a Reducer, What are the data type requirements for a key and value Class?

- Should the Key and Value class implement a special interface?

- What data types can be used directly as a key or value, is it possible to use a list as a key or value?

(4) - What are combiners?

- When should I use a combiner in a MapReduce Job?

- If you remove the use of the combiner (`job.setCombinerClass(IntSumReducer.class)`) in the example source code, Is there any performance difference?

- Why does it work using the Reducer class as combiner in our example source code?

References:

[1]<http://hadoop.apache.org/docs/current/hadoop-project-dist/hadoop-common/SingleCluster.html>

[2]Dean, Jeffrey and Sanjay Ghemawat. "MapReduce: simplified data processing on large clusters." Communications of the ACM 51.1 (2008): 107-113.
<http://static.googleusercontent.com/media/research.google.com/en/us/archive/mapreduce-osdi04.pdf>

[3]MapReduce Tutorial. <http://hadoop.apache.org/docs/current/hadoop-mapreduce-client/hadoop-mapreduce-client-core/MapReduceTutorial.html>

[4] Tom White."Hadoop: The Definitive Guide", 3rd Edition, Yahoo Press, 2012.

Stage 2:

- ***Title: Design of a Parallel Version of the Markov Model based Learning Phase (See Fig. 1)***

Objective: Students will use the background knowledge and jargon acquired in Stage 1 to design a parallel version of the Markov model's training phase. This design assumes students will use the Mapreduce programming model, thus a clear mapping of the algorithm in terms of map and reduce functions is expected. The deliverables for this phase include:

1. A working serial Java/C code, with comments, of the training algorithm using the training sequence in the Markov Chains tutorial's slides #10 as the input. The result is supposed to be the same as the one presented in the tutorial, and the output should include:
 - The count for each state transition.
 - The transition matrix.

2. A pseudo code of the training algorithm, using the MapReduce programming model. Thus, a clear mapping of the algorithm in terms of map and reduce functions is expected. The pseudo code should provide:
 - The map and reduce functions, which implement the Markov chain training algorithm.
 - The input and output key-value data type for each map and reduce function.
 - The design should be accompanied with a clear definition of the design, diagrams, and references.
3. Extra credit will be given if students make:
 - a. The pseudo code work with any given input file. A working Java code with comments is expected to submit.
 - b. A dataflow representation of the Mapreduce pseudo code. In this view, each node represents either a map or a reduce function, and the edges represent how the data flow across the nodes. One possible example can be found in the Tez apache project at <http://tez.apache.org/index.html>

HINTS:

- During the training you are provided with the set of states. For example, in the tutorial we knew that the states were (R)aining and (S)unny.
- The students will be provided with the input dataset for Mapreduce. Each file has one sequence per line, and each sequence is assumed to be generated by the same transition matrix.
- **MapReduce Installation:** It is easier to install Hadoop on a CentOS virtual machine. Students can follow the MapReduce tutorial slides page #23 to set up an all-in-one VM with Hadoop and Eclipse already installed. These are the steps:
 - Download and install VirtualBox on your machine: <http://virtualbox.org/wiki/Downloads>
 - Download the Cloudera Quickstart VM for virtualbox at <http://www.cloudera.com/content/cloudera/en/downloads.html>
 - Uncompress the VM archive. It is compressed with 7-Zip. If needed, you can download a tool to uncompress the archive at <http://www.7-zip.org/>
 - Start VirtualBox(sudo virtualbox) and click Import Appliance. Click the folder icon beside the location field. Browse to the uncompressed archive folder, select the .ovf file, and click the Open button. Click the Continue button. Click the Import button.

References:

Please refer to the slides for useful references.

- *Result due on **Due on November 7th at 23.59.***

Stage 3:

- **Title:** Implement your design on State 2 into a Fraud-Detection Platform 652-FDP
- **Objective:** Learn how to use 652-FDP for model training and use 652-FDP to integrate your own design.
- **Result due on** *Due on November 21st*: TBA.

Stage 4:

- **Title:** Project's Presentation and Report
- **Objective:** Present each group's results to the class.
- **Result due on** *Due on December 2nd and December 5th*: TBA.

Submission Details:

All reports will be submitted using an IEEE Paper Template (http://www.ieee.org/conferences_events/conferences/publishing/templates.html). Remember to cite all your sources and to include all the group's member's names in the reports and presentations.

Include any source files you wrote. Each program you write must be commented and have its own Makefile or equivalent.

Remember to include the HW/SW specifications of the machine(s) where you run your experiments.

All submissions will be made by the group's leader to Jaime Arteaga jaime@udel.edu as following: all files in a ZIP file with the name **group<YOUR_GROUPS_NUMBER>-project-stage<NUMBER_OF_STAGE>-eleg652-14f.zip** (e.g. **group2-project-stage3-eleg652-14f.zip**) with subject **ELEG652-14F Project Stage <NUMBER_OF_STAGE>** (e.g. **ELEG652-14F Project Stage 3**) before the specified deadline.