

# Fast recognition and application of Web user's behavioral patterns\*

T. Chovaňák  
Istituto Officina dei Materiali  
P.O. Box 1212  
Italy  
gubbiotti@corporation.com

O. Kašák  
Dipartimento di Fisica e Geologia  
P.O. Box 6221  
Italy  
malago@affiliation.org

M. Bieliková  
Dipartimento di Fisica e Scienze  
P.O. Box 5000  
Italy  
fin@affiliation.org

## ABSTRACT

Behavioral patterns can be understood as typical and repeating features of user's behavior during their visit of website. In this work we represent behavioral patterns as frequent itemsets of actions frequently taken by user's in their sessions. Frequent source of knowledge about behavior of users are web logs and actions taken during their visits to website aggregated to sessions. Whole process of processing web logs, finding behavioral patterns and their analysis is also known as Web Usage Mining. Found behavioral patterns may be used to create recommendations, predict user's intentions (which can be used to cache predicted pages), as support for website design change or complex understanding of website users' behavior. Existing methods of Web Usage Mining usually search for behavioral patterns common for whole set of web site users in static web logs. This work responds to actual trend of Web personalization and focusing on needs of individual users and also challenge to mine knowledge from fast streaming data. We propose solution that is able to process data about user sessions as streaming data and search for behavioral patterns telling us not only about behavior of global community of users, but also about actual behavior and changes in behavior of smaller user communities. We evaluate contribution of combining global and group behavioral patterns in their application to recommendation task. We also observe way this method is able to detect unique behavior of specific groups of users in domain of e-learning system and newspapers web portal.

## CCS CONCEPTS

• **Computer systems organization** → **Embedded systems**; *Redundancy*; Robotics • **Networks** → Network reliability

## KEYWORDS

ACM proceedings, text tagging

## 1 INTRODUCTION

Understanding website users' behavior precisely is crucial for better personalization and adaptation of website content and structure. Every user is less or more different from others but often there are groups of users with similar behavior in some specific situations and time. Detecting groups of users with similar

behavior and their behavioral patterns may lead to better understanding of users' intentions.

Web logs and users' actions stored there are often source of implicit information about users' behavior. Whole process of processing web logs, finding behavioral patterns and their analysis is known as Web Usage Mining (WUM). WUM is process consisting of 5 phases:

1. Collecting data from data sources
2. Preprocessing data
3. Patterns discovery
4. Analysis, validation and use of discovered patterns

One of main advantages of using WUM process to search for behavioral patterns is that data inputs consists of objective user feedback - actions taken by users. We can represent user session as set of actions taken by user (items) and behavioral patterns as frequent itemsets of actions gathered from preprocessed logs.

One of important challenges nowadays is fast in-memory processing of data generated as potentially infinite stream. Data streaming algorithms are built upon models that are incrementally updated with incoming data instances. Traditional data mining algorithms usually require more than one scan of all instances in database. In streaming data every instance should be processed only once. With data stream evolution conceptual drift can appear. Conceptual drift is important problem when performing data mining with streaming data. It is derived from changes in real environment where streaming data instances are created. Factors activating such changes are usually hidden and unknown. Conceptual drift is problem mainly with models created from old data before the drift happened which are consequently improper.

Goal of this work is to respond to current trends of Web personalization and focusing on needs of individual users by discovering behavioral patterns not only from global community of website users but also from smaller communities of users. We propose process with regard to requirements of data streaming algorithms, that is integrating data stream clustering algorithm used for segmenting active users to groups according their actual behavior with algorithm mining frequent closed itemsets over data stream that is able to respond quickly to conceptual drifts and is used to discover actual behavioral patterns represented as frequent closed itemsets from global community of users and from detected groups of users. Proposed method is able to detect recent behavior of global community and smaller communities and also changes of their behavior in time.

There are many applications of such discovered knowledge. For Web personalization as such, for predicting users' behavior, for caching web pages to memory, as important input when changing design of website or making some business decisions such e.g. about market segmentation.

We apply discovered knowledge about users' behavior to recommendation task and evaluate gain from combination of behavioral patterns for global community with patterns discovered for smaller communities against approach where we use only patterns discovered for global community in this specific task.

The paper is organized as follows. In §2, we review three related research directions. In §3, we describe method for mining global and group behavioral patterns over data stream and use them in recommendation task. In §4, we describe methodology of experiments and data we used to evaluate proposed method and simultaneously results we observed. We conclude the paper in §5.

## 2 Related Work

In this section we review few related research works. Our proposed process is combination of existing stream data mining algorithms from which every has its own research area. It's new application of WUM process that is able to run as whole in online environment. We apply extracted knowledge to recommend items to users, but there are other possible applications to it.

### 2.1 Existing applications of WUM to predict users' behavior and recommend items

Found knowledge about users' behavior with some of different representations of behavioral patterns can be applied to predict next steps of user or recommend items or products.

In (Jalali, Mustapha, Nasir Sulaiman, & Mamat, 2010) WebPUM method was proposed which represents behavioural patterns as partitions resulting from graph partitioning algorithm applied on user navigation graph where nodes are web pages, connection between them have weight computed from time and frequency information from user's sessions. Weight of connection rises as more frequently two pages appear, less timegap between them and more time user's spent on pages.

They use gained patterns to recommend next pages to user according their actual behaviour.

Next interesting method predicting user's next steps was proposed in (Liraki & Harounabadi, 2015). It is sophisticated system using fuzzy c-means clustering to find behavioural patterns represented as association rules.

In (Anandhi & Irfan Ahmed, 2014) 3 new contributions to enhance WUM process. First they proposed heuristic to identify user's sessions. Second they used DBScan to cluster user's sessions. DBScan is able to reveal otherwise ignored patterns because of their low support but high confidence when represented as association rules.

Their last proposed contribution is using inverted index to effectively predict user's behaviour online.

## 2.2 Quasi-Static Measurements: MOKE and MFM

**2.2.1 Component Structures.** Hysteresis loops were measured in the longitudinal configuration by magneto-optical Kerr effect (MOKE) using a Photo-Elastic Modulator operating at 50 kHz and lock-in amplification. The magnetic field, applied along the major axis of the elliptical elements, was swept between +1.0 and -1.0 kOe. To gain further information on the magnetization reversal process and the switching fields of the distinct dots, longitudinal minor loops were also measured.

**2.2.2 Magnetization.** To complement the above analysis of the magnetization evolution under an external field, the magnetic states of the bi-component structures were directly imaged at different point of the hysteresis cycle using in-field magnetic force microscopy (MFM) [5],[1]

**Eavesdropping.** MFM images were recorded by a Digital Instruments Nanoscope IIIa, using the phase detection mode, i.e., monitoring the cantilever's phase of oscillation while the magnetic tip was scanning the sample surface at a distance of 120 nm on the average (lift mode). Commercially available ferromagnetic CoCr-coated tips, magnetized to be a north pole, were used. In order to exclude the influence of the tip on the magnetic state of the sample, we used different scanning directions and tip to sample distances, obtaining the same results with different operating conditions.

## 2.3 Dynamic Measurements: BLS

BLS spectra of the thermal magnetic excitations were measured at room temperature in the back-scattering geometry by using a (3+3)-pass tandem Fabry-Perot interferometer. About 200 mW of monochromatic ( $\lambda = 532$  nm) laser were focused onto a spot having a diameter of about 30 microns. An external magnetic field with intensity ranging between  $-1.0 \text{ kOe} < H < +1.0 \text{ kOe}$  was applied parallel to the sample surface along the dots length and perpendicular to the incidence plane of light (Voigt geometry).

## 2.4 Ground-State Magnetization Determination and DMM Micromagnetic Simulations

**2.4.1 Determined.** The magnetization ground-states as well as the hysteresis loops were determined by using the OOMMF code. To reproduce the exact shape of the dots, a bitmap image of the basic unit of the bi-component dots was created from the SEM image of Fig. 1, and used as input for the simulations. Periodic boundary conditions have been applied to account for the chain arrangement of the Py/Co dots in the investigated sample.

**2.4.2 Micromagnetic.** For each micromagnetic cell the reduced magnetization takes the form where the magnetization (saturation magnetization) in the  $k$ -th cell; note that the saturation magnetization now depends on the ferromagnetic material through the index  $k$ . Hence, in a polar reference frame

$$(x + a)^n = \sum_{k=0}^n \binom{n}{k} x^k a^{n-k} \quad (1)$$

where  $K$  is the azimuthal (polar) angle of the magnetization (the time dependence is omitted). The second derivatives of the energy density depend on the micromagnetic cell indexes, and through them on the material index corresponding either to Py or Co. The expressions of  $E_{\text{ext}}$ ,  $E_{\text{exch}}$ ,  $E_{\text{dmg}}$  and  $E_{\text{ani}}$  are the same as the ones of the single-component system apart from the explicit dependence of the magnetic parameters on the given ferromagnetic material. Moreover, the uniaxial anisotropy energy density of Co is neglected.

It is possible to write the following periodicity rule valid for the dynamic magnetization  $\delta m(r)$  of each collective mode, a version of the Bloch theorem, viz. Note that, exchange contribution is set equal to zero, because in each unit cell the two elliptical dots are separated. Moreover, the uniaxial anisotropy energy density of Co is neglected Table 1.

**Table 1: Frequency of Special Characters**

Non-English or Math	Frequency	Comments
Ø	1 in 1,000	For Swedish names
\$	4 in 5	Used in business

$$x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} \quad (2)$$

Therefore one can observe either an in-phase (acoustic) or an out-of-phase (optical) character of the modes, with respect to the precession of the in-plane magnetization components in adjacent Py and Co dots.

We would like to mention that the DMM presents several advantages with respect to OOMMF for calculating the spectrum of magnetic eigenmodes for the following reasons: *a)* There is no need to excite the system by any magnetic field pulse, *b)* A single calculation allows to determine the frequencies and eigenvectors of all spin-wave modes of any symmetry, *c)* The spectrum is computed directly in the frequency domain, *d)* The mode degeneracy is successfully solved, *e)* The spatial profiles of the spin-wave modes are directly determined as eigenvectors and, finally, *f)* The differential scattering cross-section can be calculated accurately from the eigenvectors associated to each spin-wave mode. This is a clear indication that both the Py and Co sub-elements are in a single domain state where Py and Co magnetizations are all oriented with their magnetic moment along the chain and field direction. At point  $\beta$  ( $H = -372$  Oe) of the hysteresis loop, where the plateau is observed in the  $M$ - $H$  loop, the dark and bright spots of the Py dots are reversed with respect to those of Co, accounting for an antiparallel relative alignment of magnetization.

### 3 RESULTS AND DISCUSSION

#### 3.1 Magnetization Curves and MFM Characterization

The major hysteresis loop measured by MOKE, plotted in Fig. 1, displays a two-step switching process due to the distinct magnetization reversal of the Py and Co sub-elements, characterized by a different coercivity. As the field is reduced from positive saturation (upper branch of the  $M$ - $H$  loop), a 100% remanence is attained. Within each bi-component unit (about 36%) in good agreement with experimental result (about 40%).

To directly visualize the evolution of the magnetization in the Py and Co subunits of our bi-component dots during the reversal process, we performed a field-dependent MFM analysis whose main results are reported in Fig. 2. At large positive field ( $H = +800$  Oe, not shown here) and at remanence ( $\alpha$  point of the hysteresis loop of Fig. 1), the structures are characterized by a strong dipolar contrast due to the stray fields emanated from both the Py and Co dots.



**Figure 2: MFM images of the bi-component Py/Co dots for different values of the applied magnetic field which are indicated by greek letters along both the major and minor hysteresis loop.**

This is a clear indication that both the Py and Co sub-elements are in a single domain state where Py and Co magnetizations are all oriented with their magnetic moment along the chain and field direction. At point  $\beta$  ( $H = -372$  Oe) of the hysteresis loop, where the plateau is observed in the  $M$ - $H$  loop, the dark and bright spots of the Py dots are reversed with respect to those of Co, accounting for an antiparallel relative alignment of magnetization.

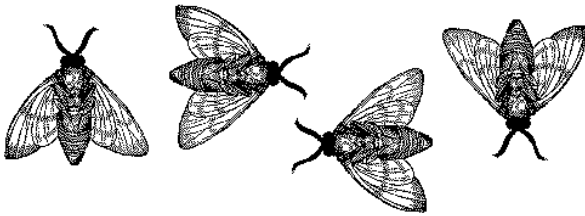
At relatively large negative fields (point  $\gamma$ ,  $H = -770$  Oe) the magnetization reversal is completed and the magnetization of the two adjacent sub-elements are saturated in the negative direction. The ground state remains unchanged when the field is now reduced to zero, i.e. remanent state coming from negative saturation, as confirmed by the MFM image taken at point  $\delta$  of Fig. 1.

We have also used MFM to measure the magnetic configurations along the minor hysteresis loop, described above. Once the AP ground state has been generated at  $H = -500$  Oe, the applied field is increased in the positive direction. The MFM image taken at point  $\alpha'$  of Fig. 2, remanent state of the minor loop ( $H = 0$ ), shows that the AP state is stable and remains unchanged until the magnetic field is increased up to +300 Oe where the Py magnetization reverses its orientation and returns to be aligned

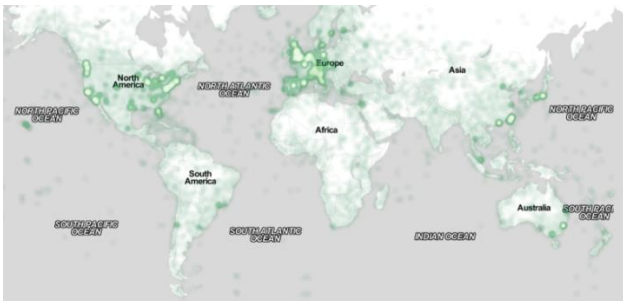
with that of Co dots. On the basis of the above MFM investigation, one can say that the structures are always in a single domain state, while the relative magnetization orientation between the adjacent Py and Co elements depends on both the field value and the sample history.

### 3.2 Field Dependent BLS Measurements and DMM Calculations

Fig. 3 displays the frequencies of BLS peaks plotted as a function of the applied field magnitude starting from positive values. The field is then decreased and reversed following the upper branch of the hysteresis loop, shown in the same figure. Up to five peaks are measured in the spectra, as shown in spectrum measured at  $H = 0$  Oe in the Fig. 3 inset, and their field evolution analyzed over the whole field range investigated. The detected modes are identified and labeled on the basis of their calculated spatial profiles, shown in Fig. 4 for  $H = 500$  and  $-500$  Oe.



**Figure 3:** Dependence of the magnetic eigenmode wave frequency on the applied field strength.



**Figure 4:** Calculated spatial distribution of the in-plane dynamic magnetization.

They exhibit marked localization into either the Co or the Py dots, as stated at the end of the previous Section, where it was introduced the labelling notation containing the dominant localization region (either Py or Co) and the spatial symmetry (EM, F, DE, etc).

When the dots are in the P state, up to five modes were detected in BLS spectra. On the basis of the calculated profiles (right panel of Fig. 4), we identified in the P state the two modes at lowest frequencies as the EM(Py) and the F(Py), with a very small spin precession amplitude into the Co dot. This is because for this material we are below the frequency threshold for the

existence of spin waves. A similar effect has been observed in periodic array of alternating Permalloy and Co nanostripes

Note that the nodal lines present in the spatial profile of the F (Co) mode perpendicular to the long axis of the ellipse do not correspond to a real change of sign of the dynamic magnetization and are due to the partial hybridization of the F mode with higher-order modes having frequencies close to the one of the F mode. Interestingly, the frequency slope of modes localized into the Co dots is larger than that of Py modes, due to larger values of the Co magnetization and gyromagnetic ratio. An overall good agreement between the calculated (dotted curves) and measured frequency (full points) has been achieved (see Fig. 3) even if some discrepancies are observed for the frequency of the EM and 1DE (Py) modes.

The corresponding spatial profiles of the modes are shown in the left panels of Fig. 4. Here one can see that the only mode which is purely localized in one dot is the EM of Co, because now it is sub-threshold for Py. A further reduction of  $H$ , which is sufficient to cause the Co magnetization reversal, produces a P state at negative fields and the frequency starts to increase.



**Figure 5:** Full point are the frequencies measured along the minor hysteresis.

Again as a function of the applied field. In this field range the frequencies of modes in the Py dots monotonously increase in a way similar to that measured in the P state for positive field values while an abrupt change in the frequency of Co modes occurs.

Notice that if one stops increasing the negative field to about  $-300$  Oe and comes back towards positive applied fields, BLS measurements can be performed following the minor hysteresis loop. This method permits to study, for example, the magnetization dynamics at remanence (without any external applied magnetic field) when the system is in the AP state (see MFM image  $\alpha'$  in Fig. 2), a configuration which cannot be achieved at remanence along the major M-H loop. In Fig. 5 we show the modes frequency measured along the minor loop (full points) and compare them with values measured along the major M-H loop (open points).

By inspection of the frequency slope of the modes, one can immediately understand the localization of modes into dots of different materials looking at their slope.

In particular, for three (two) modes we measure a negative (positive) frequency slope with an almost linear dependence on  $H$ . It is evident that modes with negative frequency slope are modes localized into the Py dot (EM, F and 1DE) while the two with positive slope are the F(Co) and the EM(Co) modes.



### 3.3 Analysis of the Dynamic Coupling as a Function of the Gap Size

One interesting point which emerges from analysis of Figs. 3 and 4 is that the frequency values of the eigenmodes are not the same at +500 Oe and at -500 Oe. This is expected for modes localized into the Co elements, since the external field is either parallel or antiparallel to their magnetization. However, for those mode localized into the Py sub-element. One could have predicted to find the same frequency values at  $\pm 500$  Oe, unless the dipolar coupling arising from the adjacent Co dot plays a significant role. In fact, as seen in Figs. 3 and 4, reversing the field from +500 to -500 Oe, the frequencies of EM(Py) and IDE(Py) modes increase by about 0.2 GHz and 0.6 GHz, respectively, while that of F(Py) decreases by 0.25 GHz. The reason of this complex behavior will be addressed in the following, analyzing the interplay of both static and dynamic dipolar coupling between the adjacent Py and Co dots Table 2.

This is a clear indication that both the Py and Co sub-elements are in a single domain state where Py and Co magnetizations are all oriented with their magnetic moment along the chain and field direction.

**Table 2: Comparison of Coefficients from Atomistic**

Atm	MS-CG	MS-CG/DPD
1.78	14.32	1.74 (-2%)
0.43	31.00	0.40 (-7%)
0.062	15.61	0.048 (-23%)
0.032	9.76	0.024 (-24%)
0.020	4.66	0.015 (-25%)
0.012	2.32	-"
0.0076	0.016	-"



**Figure 6: Calculated frequency evolution of modes detected in the BLS spectra.**

In Fig. 6 the calculated frequencies of the most representative eigenmodes at +500 Oe (FM state) and -500 Oe (AP state) are plotted as a function of the gap size  $d$  between the Py and Co sub units (please remind that in the real sample studied here,  $d = 35$  nm). As a general comment, it can be seen that the frequencies for the system in the AP state are more sensitive to  $d$  than those of the P state. In particular, the lowest three frequency modes of the AP state (EM(Co), EM(Py) and F(Py)) are downshifted with respect

to the case of isolated elements (dotted lines) and show a marked decrease with reducing  $d$ , while the two modes at higher frequencies (F(Co) and IDE(Py)) have an opposite behavior even though they exhibit a reduced amplitude. In the P state (right panel), the modes concentrated into the Py dots exhibit a moderate decrease with reducing  $d$ , while an opposite but less pronounced behavior is exhibited by the F(Co) mode.

1. Never, ever use vertical rules.
2. Never use double rules.

## 4 CONCLUSIONS

In summary, we have performed both an experimental and theoretical study of the spin eigenmodes in dipolarly coupled bi-component cobalt and permalloy elliptical nanodots. Several eigenmodes have been identified and their frequency evolution as a function of the intensity of the applied magnetic field has been measured by Brillouin light scattering technique, encompassing the ground states where the cobalt and permalloy dots magnetizations are parallel or anti-parallel, respectively. In correspondence to the transition between the two different ground states, the mode frequency undergoes an abrupt variation and more than that, in the anti-parallel state, the frequency is insensitive to the applied field strength. The experimental results have been successfully interpreted by the dynamic matrix method which permits to calculate both the mode frequencies and the spatial profiles.

## A HEADINGS IN APPENDICES

The rules about hierarchical headings discussed above for the body of the article are different in the appendices. In the appendix environment, the command section is used to indicate the start of each Appendix, with alphabetic order designation (i.e., the first is A, the second B, etc.) and a title (if you include one). So, if you need hierarchical structure within an Appendix, start with subsection as the highest level. Here is an outline of the body of this document in Appendix-appropriate form:

### A.1 Introduction

### A.2 Experimental and Computational Details

#### A.2.1 Sample Fabrication

#### A.2.2 Quasi-Static Measurements: MOKE and MFM

#### Component Structures

#### Magnetization.

#### A.2.3 Dynamic Measurements: BLS

#### A.2.4 Ground-State Magnetization Determination and DMM Micromagnetic Simulations

#### Determined.

#### Micromagnetic

### A.3 Results and Discussion

### A.3.1 Magnetization Curves and MFM Characterization

### A.3.2 Field Dependent BLS Measurements and DMM Calculations

### A.3.3 Analysis of the Dynamic Coupling as a Function of the Gap Size

## A.4 Conclusions

## A.5 References

## ACKNOWLEDGMENTS

This work was partially supported by the MIUR-PRIN 2010–11 Project 2010ECA8P3 “DyNanoMag” and by the National Research Foundation, Prime Minister's office, Singapore under its Competitive Research Programme (CRP Award No. NRF-CRP 10-2012-03).

## REFERENCES

- [1] Patricia S. Abril and Robert Plant. 2007. The patent holder's dilemma: Buy, sell, or troll? *Commun. ACM* 50, 1 (Jan. 2007), 36–44. DOI: <http://dx.doi.org/10.1145/1188913.1188915>
- [2] I. F. Akyildiz, W. Su, Y. Sankarasubramaniam, and E. Cayirci. 2002. Wireless Sensor Networks: A Survey. *Comm. ACM* 38, 4 (2002), 393–422.
- [3] David A. Anisi. 2003. *Optimal Motion Control of a Ground Vehicle*. Master's thesis. Royal Institute of Technology (KTH), Stockholm, Sweden.
- [4] P. Bahl, R. Chancre, and J. Dungeon. 2004. SSCH: Slotted Seeded Channel Hopping for Capacity Improvement in IEEE 802.11 Ad-Hoc Wireless Networks. In *Proceeding of the 10th International Conference on Mobile Computing and Networking (MobiCom'04)*. ACM, New York, NY, 112–117.
- [5] Kenneth L. Clarkson. 1985. *Algorithms for Closest-Point Problems (Computational Geometry)*. Ph.D. Dissertation. Stanford University, Palo Alto, CA. UMI Order Number: AAT 8506171.
- [6] Jacques Cohen (Ed.). 1996. Special Issue: Digital Libraries. *Commun. ACM* 39, 11 (Nov. 1996).
- [7] Bruce P. Douglass. 1998. Statecarts in use: structured analysis and object-orientation. In *Lectures on Embedded Systems*, Grzegorz Rozenberg and Frits W. Vaandrager (Eds.). Lecture Notes in Computer Science, Vol. 1494. Springer-Verlag, London, 368–394. DOI: <http://dx.doi.org/10.1007/3-540-65193-429>
- [8] Ian Editor (Ed.). 2008. *The title of book two* (2nd. ed.). University of Chicago Press, Chicago, Chapter 100. DOI: <http://dx.doi.org/10.1007/3-540-09237-4>