Improvement of the Automated Orientation Criteria for New Graduates

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Summary

Education can be defined as the action of instilling in individuals a set of knowledge, moral and intellectual values, considered necessary to reach the desired level of culture. In Côte d'Ivoire, for several years, the orientation of new high school graduation as remained a major challenge. Indeed, the plethoric number of students in Ivorian universities the insufficiency of the classrooms, the insufficiency of the teaching staff and trainers, etc. make orientation a very selective mechanism. Today, these assignments are made almost automatically. It is therefore imperative that these automatic assignment devices become increasingly sophisticated in order to minimize academic failure. It is in this context that this article is written situate the objective of which is to propose an automatic classification of new hig school graduates in order to identify homogeneous profiles. The interest of this work lies in the understanding of the criteria necessary for an effective personalization of the automatic assignment system for new graduates.

Keywords:

Educational system, Automatic assignment, Automatic classification.

1. Introduction

The Ivorian educational system has been facing difficulties for years. It is crossed by a series of crises and no level escapes it (primary, secondary, and higher) [1]. According to Seydou OUATTARA, the high unemployment rate explains the presence of less passionate and motivated teachers in the educational system [2]. An important element of this crisis is the phenomenon of university dropouts and complaints of bad orientation, especially at the level of baccalaureate holders. Given the very high number of baccalaureate holders to be assigned each year in relation to the reception capacities, it is not obvious that these assignments are in line with the cognitive abilities of future students. In such a situation, assignments are made by digital platforms that automatically assign students based on static criteria.

Indeed, since 2016 Côte d'Ivoire has started a phase of integration of ICTs in its schools in order to make its educational system one of the most competitive on the continent [3], [4]. Despite all these efforts, we are

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witnessing the case of failures, even drop-outs in some Universities. One of the arguments commonly put forward to explain this situation is the inadequate orientation of high school graduates in the courses. Indeed, each year, many graduates present themselves at the doors of the Universities. Faced with this plethoric number of candidates, orientation becomes digital. The problem therefore arises from the performance of automated orientation criteria in order to limit or even avoid university failures. In view of all the above, one wonders how to improve the school orientation system for Ivorian baccalaureate holders.

The objective of this article is to use classification methods in order to determine the relevant indicators of orientation and success of the baccalaureate holders. It will be about using, in a complementary way, multivariate statistical analysis methods Multiple Correspondence Analysis (MCA), Principal Component Analysis (PCA) to end up with an Ascending Hierarchical Classification (HAC).

The first step is to present a literature review of guidance methods. Then present the methodological approach and the results of the classification method. Finally, we will end up end with a conclusion.

State of the art

The orientation process is characterized by several factors and differs from one country to another. The works of Marion Erouart of the University of Lyon stated that this orientation process must be integrated upstream and then into higher education training. Indeed, it suggests that the first two years constitute a core curriculum then the students specialize when they enter the third year in one of the nine specialties offered by Erouart and al.. Hubert, and al. proposes in his work a recommendation system for orientation towards higher education [4]. This system is based on the operation of an ontology in order to specify the application domain of the student in a set of classes (the training) linked together by relations. GRENET and al., and IEHLE and al. offer a critical analysis of assignment procedures such as the SIGEM software and

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registrations used in France in French secondary and higher education [6], [7]. Combe and al. proposes an assignment procedure. This procedure not only makes it possible to quantify the impact that its adoption could have compared to the current system. But also, is based on the specificities of the least attractive academies to increase the rate of movement of tenured teachers up to 30% [8]. Omar and al. offers an orientation system based on the neural network algorithm. This system is able to classify the academic orientations according to the professional category (Activities, Aptitudes, Abilities, Personality) [9]. Sakers and al. is interested in finding a match between the data characterizing candidates for higher education courses (candidate and class marks, high school of origin, etc.) and the marks obtained by these same candidates during their first year at university [10].

Material and method

As part of the study, we used an analytical and statistical processing mechanism for a database containing a mixture of secondary school leavers from 2000 to 2021. To implement our typology, we made the choice of a multivariate analysis. The data for baccalaureate holders is recorded in Table 1 below. This table includes 303 high school diploma holders and 32 variables describing information on them.

Table 1 : List of information about bachelors

	D	E	F G	н	1	J KLMNOPQRST	U	V	W	х	Y Z	AA	AB	AC	AD	AE	AF	
1	date_naissar	genre	matricule serie_bac	c note_mat moy_e	class_n not	te_phy_ch m nc m nc m nc m nc m nc ma	y_class_a n	ote_allema m	ioy_class_a not	e_espagn mo	y_class_e date_bac	filiere	categorie_filiere	choix	categorie_fil ch	wix_respect	diplome	
2	8/24/2009	Male	686-45-5488 A2	10	11	0000#######	14	14	15	11	12 02/07/2000	Allemand	Litterature	Anglais	Litterature	0		1
3	08/07/2007	Female	671-11-6066 C	13	14	14 # 5 8 7 # # 9 8 # #	18	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Maths-Infor	MATHS-INFC	1		1
4	8/21/1999	Female	535-27-6330 A2	6	11	0000####8##	15	13	14	12	14 02/07/2001	Anglais	Litterature	Anglais	Litterature	1		1
5	11/19/2005	Male	631-09-0269 D	15	14	15 # # # 5 8 # # 9 8 #	9	0	0	0	0 02/07/2010	Maths-Informatiq	MATHS-INFO	Maths-Infor	MATHS-INFC	1		0
6	06/02/2001	Female	412-72-7361 D	12	13	16 # # # # # # # # # #	15	0	0	0	0 02/07/2001	Sociologie	Scs Sociologie	Sociologie	Scs Sociologi	1		0
7	11/01/2001	Male	827-71-4365 C	14	15	16 # # # # # # # # 9 #	7	0	0	0	0 02/07/2001	Sciences Economi	Scs Sociologie	Lettres mod	Litterature	0		0
B	3/18/2005	Male	734-83-2627 A2	10	11	00########	14	14	15	12	13 02/07/2001	Psychologie	Scs Sociologie	Psychologie	Scs Sociologi	1		1
9	07/03/2003	Female	693-20-3439 D	13	14	16	12	0	0	0	0 02/07/2001	Psychologie	Scs Sociologie	Psychologie	Scs Sociologi	1		1
0	06/05/2002	Male	808-77-7670 C	13	15	14	11	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Sciences Eco	Scs Sociologi	0		0
1	10/27/2001	Female	720-79-2174 C	13	14	15	15	0	0	0	0 02/07/2001	Sciences Econom	Scs Sociologie	Sciences Eco	Scs Sociologi	1		1
2	1/31/1998	Female	850-96-6434 D	11	13	14	15	0	0	0	0 02/07/2006	Maths-Informatiq	MATHS-INFO	Maths-Infor	MATHS-INFC	1		0
3	05/10/2002	Male	614-81-9785 C	9	14	14	14	0	0	0	0 02/07/2006	Biologie	Biologie	Biologie	Biologie	1		1
4	10/21/2000	Male	596-84-4902 A2	7	10	0000#######	15	12	11	14	13 02/07/2010	Espagnol	Litterature	Anglais	Litterature	0		0
5	11/17/2003	Female	140-80-2563 A2	8	10	0000#######	12	12	13	14	13 02/07/2001	Espagnol	Litterature	Espagnol	Litterature	1		1
6	3/30/2002	Male	394-38-6123 A2	9	11	0000#######	9	13	12	14	13 02/07/2001	Lettres modernes	Litterature	Lettres mod	Litterature	1		1
7	8/24/2008	Female	743-73-7240 A2	9	11	0000#######	13	11	12	13	13 02/07/2001	Lettres modernes	Litterature	Anglais	Litterature	0		0
8	2/26/2003	Male	560-28-2791 D	14	15	15 # # # # # # # # # #	14	0	0	0	0 02/07/2001	Maths-Informatig	MATHS-INFO	Maths-Infor	MATHS-INFC	1		0
9	3/25/2015	Male	459-34-4072 C	12	12	14 # # # # # # 8 # # #	15	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Physique	Scs Physique	0		0
0	6/25/2008	Male	735-71-8349 C	12	14	14	12	0	0	0	0 02/07/2001	Maths-Informatig	MATHS-INFO	Physique	Scs Physique	0		1
1	1/16/2000	Female	371-81-0229 D	13	12	14	14	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Maths-Infor	MATHS-INFC	1		1
2	09/01/2001	Female	615-45-4956 A2	10	11	0 0 0 0 # # # # # # #	12	14	15	12	13 02/07/2001	Allemand	Litterature	Allemand	Litterature	1		1
3	1/26/2002	Female	652-25-9615 A2	7	12	0000##9#97#	14	14	13	12	13 02/07/2001	Allemand	Litterature	Allemand	Litterature	1		1
4	02/02/2012	Male	693-68-7331 A2	9	11	0000####76#	15	12	11	13	14 02/07/2021	Anglais	Litterature	Lettres mod	Litterature	0		0
5	6/23/2003	Male	391-93-6036 A2	7	10	0000#######	16	13	12	12	13 02/07/2001	Anglais	Litterature	Anglais	Litterature	1		1
6	5/23/2001	Male	815-95-6387 C	13	14	15 # # # # # 9 6 # # 8	9	0	0	0	0 02/07/2021	Maths-Informatig	MATHS-INFO	Sciences Eco	Scs Sociologi	0		1
7	05/09/2002	Male	710-23-6160 C	12	13	14	7	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Sciences Eco	Scs Sociologi	0		1
8	04/04/2003	Male	816-96-3267 D	14	15	15 # # # # # # # 7 # #	17	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Sciences Eco	Scs Sociologi	0		0
9	06/12/1999	Male	809-98-0047 D	13	14	14 # # # # # # 8 # 5 #	11	0	0	0	0 02/07/2001	Maths-Informatiq	MATHS-INFO	Sciences Eco	Scs Sociologi	0		0
0	01/08/2005	Male	443-64-3282 D	12	13	14	14	0	0	0	0 02/07/2001	Physique	Scs Physique	Physique	Scs Physique	1		1
1	1/23/2000	Female	260-07-3007 C	12	14	14 # # # # # 9 # # # #	14	0	0	0	0 02/07/2001	Physique	Scs Physique	Physique	Scs Physique	1		1
2	09/05/1998	Male	885-65-8344 A2	9	10	0 0 0 0 # # # # # # #	16	12	13	11	12 02/07/2001	Anglais	Litterature	Anglais	Litterature	1		1
3	10/20/2002	Female	572-74-4659 A2	7	9	0000#######	14	11	10	14	15 02/07/2001	Anglais	Litterature	Anglais	Litterature	1		1
4	4/25/2004	Female	331-61-6761 A2	5	9	0 0 0 0 # # # # # # #	15	12	13	10	11 02/07/2001	Anglais	Litterature	Lettres mod	Litterature	0		0
5	12/13/2004	Male	567-46-6802 C	12	13	15	13	0	0	0	0 02/07/2001	Maths-Informatio	MATHS-INFO	Sciences Eco	Scs Sociologi	0		1
6	02/10/2000	Male	444-50-5852 C	13	14	12	12	0	0	0	0 02/07/2001	Maths-Informatio	MATHS-INFO	Sciences Eco	Scs Sociologi	0		1
7	7/20/2004	Male	357-33-0239 D	13	14	15	13	0	0	0	0 02/07/2001	Maths-Informatig	MATHS-INFO	Sciences Eco	Scs Sociologi	0		0
8	8/27/2005	Male	188-77-6979 D	12	13	14	12	0	0	0	0 02/07/2001	Physique	Scs Physique	Physique	Scs Physique	1		1
9	02/12/2005	Female	189-32-2304 D	13	14	13	13	0	0	0	0 02/07/2001	Physique	Scs Physique	Physique	Scs Physique	1		0
0	12/26/2000	Female	802-78-0668 C	12	13	14	13	0	0	0	0 02/07/2001	Physique	Scs Physique	Maths-Infor	MATHS-INFC	0		0
ń	10/18/2000	Male	164-84-4934 C	13	13	12	15	0	0	0	0 02/07/2001	Sciences Economi	Scs Sociologie	Sciences Fco	Srs Sociologi	1		1

As part of the classification, the R 4.2.0 software was used for the data analyses with the FactoMineR packages for the data analysis and factoextra for the data visualization. One- and two-dimensional descriptive analyses were performed. In order to highlight the linear links between the individuals of our sample, a multidimensional analysis, the Principal Component Analysis (PCA) was carried out. Its objective is not only to reduce the number of variables assigned to individuals and facilitate observations while saving as much student information as possible but also to determine homogeneous groups of observations or atypical observations. Among the variables available to our sample, some are qualitative, others quantitative. Their joint analysis uses an MCA. This analysis makes it possible to create clusters (groupings or distances) of objects. Subsequently, the information from this analysis was used to carry out a CAH. The Ascending Hierarchical Classification (AHC) made it possible not to have to choose a priori provisional class centers or the number of classes. This method makes it possible to reveal the "true" classes if they exist since they are determined

automatically and do not depend on an a priori choice as in the case of the Classification method around Mobile Centers (CCM) [11]. Kaiser's criterion was used to determine the axes to be retained for PCA and MCA. According to this Kaiser criterion, it is advisable to retain the axes whose relevant variables are clearly > 1 [12], [13].

Results

- Results of the first PCA

A first Principal Component Analysis (PCA) was performed to determine the relevant variables as shown in the following figure



Fig 1 : Graph of the variables of the first PCA

This Principal Component Analysis allowed us to observe that the variables "grade_allemand", "mean_german", "grade_spanish" and "mean_spanish', are well correlated with each other on the second dimension. These variables are negatively correlated. While the variables "notes_svt", "avg_svt", avg_phy_chemistry", "note_phy_chemistry", are correlated with each other on the first dimension; the correlation is positive.

The fact that the variables note_español and note_allemand are correlated and the variables note_phy_chimie and note_svt are correlated, shows that the pupils of the literary series do not have the subjects physics chemistry and svt in their school program and the pupils of the scientific series do not have the subjects Spanish and German in their school curriculum.

At the end of this analysis, we conclude that class averages are not necessary in the orientation of baccalaureate holders; therefore, we will remove the averages from our data.

Results of the second PCA

A second presentation analysis was carried out as shown in the following figure 2:



Fig 2 : Graph of the variables of the second PCA

This analysis shows that the variables "note_svt", note_phy_chimie", "note_math" are correlated with each other, but the correlation is negative on the first dimension; it also shows that the variables "note_french", note_hist_geo", "note_allemand", "note_aespañol", "note_philo", are correlated with each other and the correlation is positive on the first dimension.

Subsequently, a CAH was carried out as shown in the following figures 3.4 and 5:



Fig 4 : Dendrogram of the classification on the result of the PCA carried out

\$`1`				
	v.test	Mean in category	overall mean s	d in category
note_phy_chimie	5.763254	14.3461538	9.347682	0.7566275
note_svt	5.660722	13.1730769	8.685430	1.6257679
note_math	5.617443	12.9615385	11.380795	0.7585801
note_espagnol	-5.064723	0.4615385	4.410596	2.3076923
note_allemand	-5.109715	0.5384615	4.579470	2.6923077
note_hist_geo	-5.192397	12.5576923	13.440397	1.5861596
note_francais	-5.298705	12.5769231	13.483444	1.9547393
age	-10.790064	20.3269231	21.619205	0.5084348
-	overall sd	p.value		
note_phv_chimie	6.8625323	8,250750e-09		
note svt	6.2728037	1.507371e-08		
note math	2.2265768	1.938036e-08		
note espagnol	6.1695380	4.089936e-07		
note allemand	6.2576027	3.226445e-07		
note hist deo	1.3451251	2.076041e-07		
note francais	1.3537014	1.166268e=07		
age	0.9476506	3.835224e-27		
9-				
\$`2`				
	v.test	Mean in category	overall mean s	d in category
note_phy_chimie	12.090748	14.32414	9.347682	0.9386914
note_svt	11.779774	13.11724	8.685430	0.9210447
note_math	9.646132	12.66897	11.380795	1.1145135
age	8.762539	22.11724	21.619205	0.5446093
note_francais	-3.151322	13.22759	13.483444	0.9153468
note_anglais	-4.128334	14.26207	14.519868	1.1080937
note_philo	-8.077474	11.61379	12,241722	1.0251069
noto oceanol	11 010610	0.00000	4 410506	0 000000

Fig 5 : Quantitative variables describing the classes

This classification gave three classes, namely:

Class1:

For the first class, the most significant variables are "note_svt", "note_phy_chimie", "note_math". For the "note_svt" variable, the individuals of class 1 take values that are significantly different from zero and above the average. They are nonzero because the test value (6.355075) is greater than 2; and they are above average because the test value is positive. Class 1 individuals have an average of 13,300,000 in SVT. For the variable "note_phy_chimie", the individuals of class 1 also take values significantly different from zero and higher than the average. The test value is 6.339025. Individuals in class 1

have an average of 14.3833333 in Physics Chemistry. For the "note_math" variable, the individuals of class 1 also take values different from zero and higher than the average. The test value is 5.506286. Individuals in class 1 have an average of 12.8000000 in Mathematics.

Class 2:

For the second class, the most important variables are "note phy chimie", "note svt", "note math". For the "note phy chime" variable, the individuals of class 2 take values that are significantly different from zero and above the average. They are nonzero because the test value (12.090748) is greater than 2; and they are above average because the test value is positive. Individuals in class 2 have an average of 14.32414 in Physics and Chemistry. In this class, the test and average values of the "note phy chimie" variable are higher than those of class 1. With regard to the "note svt" variable, the individuals of class 2 also take significantly different values zero and above average. The test value is 11.779774. Class 2 individuals have an average of 13.11724 in SVT. In this class, the test value of the "note svt" variable is greater than that of class 1; but the average value is lower than that of class 1 (13.11724 <13.300000).

For the "note_math" variable, the individuals of class 2 also take values different from zero and higher than the average. The test value is 9.646132. Individuals in class 2 have an average of 12.66897 in Mathematics. We then observe that the test value of the "note_math" variable is higher than that of class 1, while the average is lower than that of class 1. We note that in class 2, the variable 'age" has the positive test value, which is not the case in class 1 where it has a negative value. This variable is thus a significant variable in class 2, which is not the case in class 1.

Class 3

Only the variables of the literature category characterize the third class; in particular "note allemand", "note spanish", "note philo", "note french", "note hist geo". For "note english", the variable "note allemand", the individuals of class 3 take specific values different from zero and higher than the average. The test value is 16.851410. Individuals in class 3 have an average of 12.9047619 in German. With regard to the variable "note español", the individuals of class 3 also take significant values different from zero and higher than the average. The test value is 16.519682. Individuals in class 3 have an average of 12.4571429 in Spanish. For the variable "note philo", the individuals of class 3 also take values different from zero and higher than the average.

The test value is 9.550628. Individuals in class 3 have an average of 13.2190476 in Philosophy.

As for the "note_francais" variable, the individuals of class 3 also take significant values different from zero and above the average. The test value is 7.506592. Individuals in class 3 have an average of 14.2857143 in French.

As for the variable "note_anglais", the individuals of class 3 also take values different from zero and higher than the average. The test value is 5.493361. Individuals in class 3 have an average of 14.9714286 in English. For the variable "note_hist_geo", the individuals of class 3 have values different from zero and higher than the average. The test value is 4.193256. Individuals in class 3 have an average of 13.8857143 in History-Geography.

Results of the second ACM

The figure below represents the visualization of the biplot of individuals and variables. In this figure 6, individuals are in blue and variables are in red



Fig 1: Biplot of individuals and variables

The variables filiere Scs Physique, categorie_filiere_Physique and choice_Scs Physique, are close on the graph; therefore these variables have a similar true profile. same is for The the variables and filiere Economics Management Sciences. categorie filiere Scs Sociologie, which are close on the graph.

Let us now proceed to the verification of the existence of correlation between the variables and the main axes of the MCA. Figure 7 below specifies the existence of correlation between the variables and the main axes of the MCA.



Fig 2: Correlation between variables

The graphic representation above indicates that the variables "categorie_filiere", "choice", "categorie_filier" and "filiere" are correlated with each other. And this correlation is positive on the first axis and negative on the second axis.

At the end of this multiple correspondence analysis, we obtained significant results; it will now be necessary to apply the ascending hierarchical classification to these results.

Hierarchical ascending classification on MCA results

The classification on the results of the ACM gives 4 classes, namely: class1, class 2, class 3 and class 4. The first axis is associated with class 1. As for classes 2 and 3, they are associated with the axis 2 and axis 3. While class 4 is associated with axis 1. Let's visualize the dendrogram highlighting the four classes. Figures 8 and 9 below present the dendrogram and the 3D representation generated by the classification:



Fig 3: Dendrogram of the classification carried out on the results of the MCA



Fig 4: 3D graph combining hierarchical classification and factor plan

Conclusion

The work carried out in this article made it possible to make the classification of baccalaureate holders from data comprising the marks obtained in the Baccalaureate examination and information concerning their orientations. To do this, a Principal Component Analysis (PCA) was first used to determine the relevant variables in the orientation process. It made it possible to understand that the marks in the Baccalaureate examination were very strongly correlated with the averages obtained in class. This result proves that it is not necessary to take class averages into account in the orientation of new baccalaureate holders. Then, a Multiple Correspondence Analysis (MCA) was carried out to be able to take into account the qualitative variables, in particular the chosen sector and the sector of assignment of the high school graduate, in automatic classification. Finally, а Hierarchical Ascending Classification (HAC) was performed. It made it possible to have three (3) groups of graduates with different characteristics. Therefore, if one wishes to avoid failures, the specificities of each group should be taken into account in the automated criteria used in the online orientations of new graduates.

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