

# 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 high 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

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



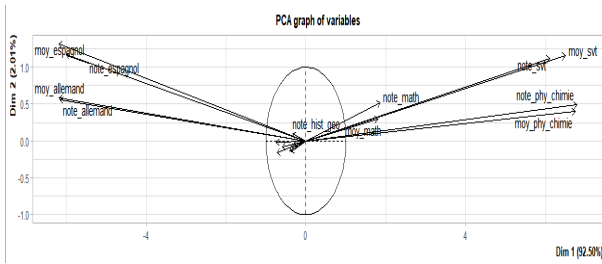


Fig 1 : Graph of the variables of the first PCA

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\_espagnol 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:

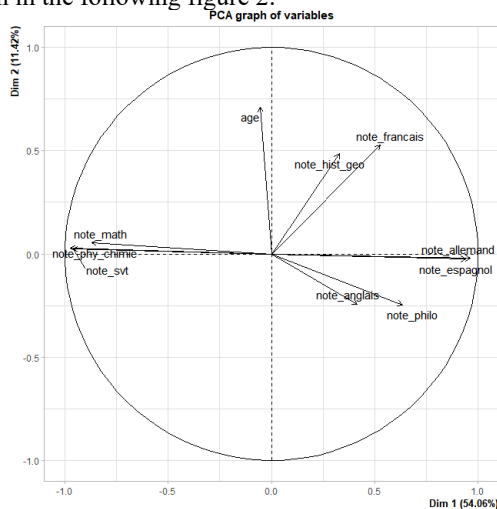


Fig 2 : Graph of the variables of the second PCA

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:

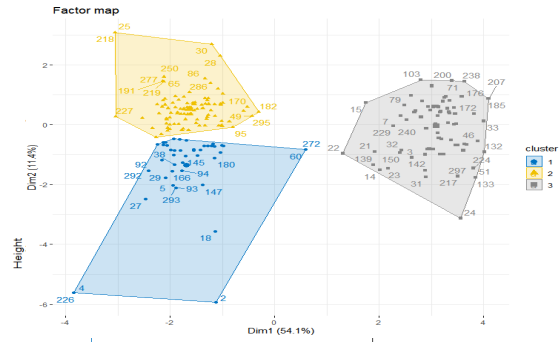


Fig 3 : Visualization of the individuals of each group



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

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$*1*
      v.test Mean in category overall mean sd 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      0.3384615  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      6.8625323  8.250750e-09
note_phy_chimie  6.2728037  1.507371e-08
note_svt        2.2265768  1.938035e-08
note_math       6.1695380  4.089936e-07
note_espagnol  6.2576027  3.226445e-07
note_allemand  1.3451251  2.076041e-07
note_hist_geo  1.3537014  1.166268e-07
note_francais  0.9476506  3.855224e-27
$*2*
      v.test Mean in category overall mean sd in category
note_phy_chimie  12.090748      14.32414  9.347682  0.9386914
note_svt        11.779774      13.11724  8.685430  1.9210447
note_math       9.646132      12.66897  11.380795  1.1145135
age            8.762539      22.11724  21.619205  0.3446093
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
note_espagnol  11.010610      0.00000  4.410596  0.0000000
    
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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 < 13.3000000). 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\_english", "note\_hist\_geo". For 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

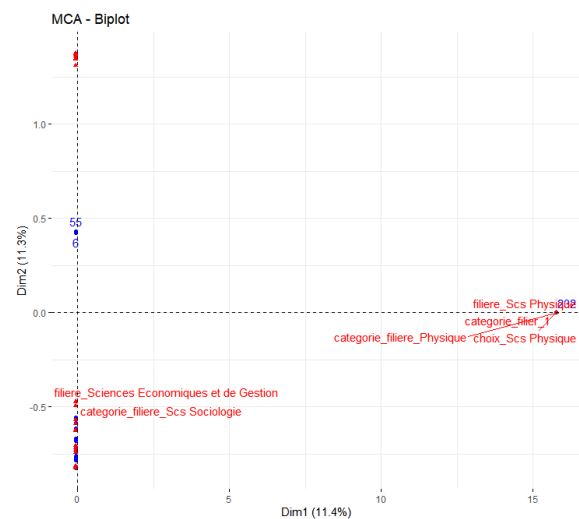


Fig 1: Biplot of individuals and variables

The variables filiere\_Scs Physique, categorie\_filier\_e\_Physique and choix\_Scs Physique, are close on the graph; therefore these variables have a similar profile. The same is true for the variables filier\_e\_Economics and Management Sciences, categorie\_filier\_e\_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.

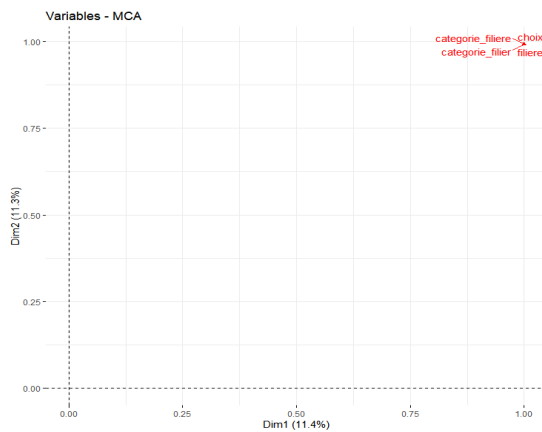


Fig 2: Correlation between variables

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:

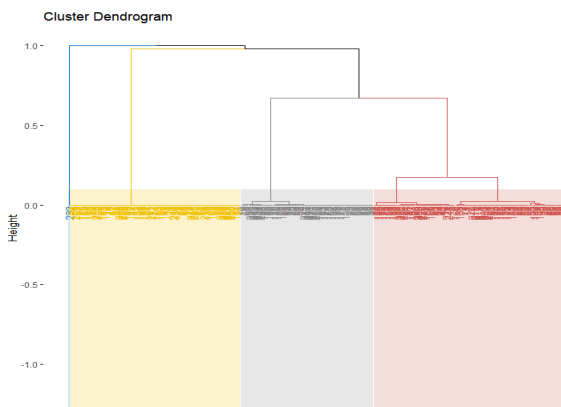


Fig 3: Dendrogram of the classification carried out on the results of the MCA

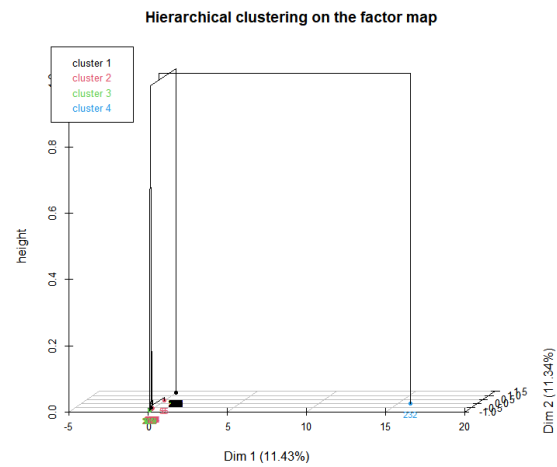


Fig 4: 3D graph combining hierarchical classification and factor plan

### 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, a 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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