Structuring Portfolio Selection Criteria for Interactive Decision Support

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Abstract1

A trichotomic evaluation system for portfolio selection support is proposed through this paper. The methodology works in two phases: First, Arbitrage Pricing Theory (APT) is used to estimate portfolios’ expected return and to identify influence factors and risk origins. ELECTRE TRI method aggregates all the common risk criteria into a unique one, which is more understandable by real investors or portfolio managers. By this way each alternative portfolio is evaluated on three criteria only including return, residual risk and common risk. In the second phase, the MINORA multicriteria interactive system based on preference disaggregation is proposed to select attractive portfolios. The whole methodological framework is illustrated by an application to the French stock market.

1. Introduction

To model the relation between risk and return, there are two well-known equilibrium models: Capital Asset Pricing Model (CAPM, Sharpe, 1963) with a one-dimensional conception of risk and the APT (Ross, 1976) with a multidimensional one.

An analysis of risk nature in portfolio management shows that it comes from various origins. Hence, the risk is multidimensional and the APT can be efficiently used to determine the expected re-

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1 We are grateful for the pertinent remarks from our two referees.
turn of portfolio and the sensibilities of common risk factors. Furthermore, APT does not impose a restrictive comportment on the investor as in the CAPM. Then, the question is how is it possible to manage portfolio selection?

Then, Multi Criteria Decision Aid (MCDA) is an attractive solution to manage portfolio selection on this basis. Indeed, MCDA proceed in the resolution of decision problems where the decision-maker is confronted to various conflicting criteria. Furthermore, MCDA presents the advantage of being able to take into account the preferences of any particular decision-maker (portfolio manager, investor).

As, APT is conceived to take into account various sources of risk, this equilibrium model is well adapted to the nature of multicriteria portfolio selection problem. The APT is used for evaluating portfolios and MCDA to proceed to the selection of some of them.

Several authors have already proposed to use MCDA for portfolio selection; but as far as we know, except our first study (Hurson and Ricci-Xella, 1997, 2000), none of them have linked it to APT.

In our first paper\(^2\) we propose to rank portfolios following the characteristics of their betas. This paper was essentially an analysis of the risk attached to the portfolio. We propose now a more global analysis by adding two other criteria: the return and the specific risk estimated by the APT model.

The difficulty is that, if the return and the residual risk are simply added to several common risk criteria, the importance of the later is artificially augmented. This is why we decide to construct a synthesis criterion to obtain a measure of common risk. A natural and common way to do this is to assign portfolios to different class of risk. This is done using ELECTRE TRI (Yu, 1992) method, because it is a MCDA method conceived to sort alternatives to different classes. Finally, the MINORA multicriteria interactive system (Cf. Siskos et al., 1993; Spyridakos and Yannacopoulos, \(^2\) Hurson Ch. and Ricci-Xella N., 1997.)
1995) is used to rank the portfolios from the best to the worst according to the investor preferences.

The rest of this paper is organised as follows. Section 2 presents the methodological issue. The third section develops the method used to construct a common risk criterion. The fourth section proposes a trichotomic decision support system. Hence, fifth section gives an application to the French market. Finally, the conclusion summarises the obtained results and gives some future direction.

2. Methodological Issues

The basic components of the proposed methodological framework are presented in Figure 1. The first paragraph presents an argumentation to use three criteria to manage portfolio selection: expected return, common risk and specific risk. The second paragraph exposes the implementation of the APT model; the third discusses the links between APT and MCDM.

2.1. Trichotomic portfolio management

Corporate manager uses risk and expected return to manage stock portfolios. Concerning risk, firms are clearly exposed to a variety of risks, some of which are market wide and others, which are specific to the firm.

- Market wide risks are macroeconomic or macro factors. These factors affect most all firm’s earning and stock prices when relevant macro risks turn generally favourable, stock price rise and investors do well; when same variables go the other way, investors suffer.

- Specific risk arises because many of the perils that surround an individual company of peculiar to that company and perhaps its direct competitors.

Theoretically, the specific risk can be eliminated by diversification. Several studies have shown that total elimination of the specific risk require the construction of portfolios containing a minim-
um of 10 stocks, although the degree of impact may vary across market. Theoretically, because the specific risk wash out in diversified portfolio, it is not remunerated.

However, the argument that diversification reduces risk is incontestable, but the assumption that theory makes – that diversifiable risk does not matter– is debatable. As in Damodoran\(^3\) this argument is based on the presumption that the marginal investor in the market, who sets prices and therefore determines how risk is viewed, is well diversified and expects to be rewarded only for the nondiversifiable risk. If the marginal investor is not well diversified, however, diversifiable risk may affect prices and expected returns.

Moreover in practice, market imperfections can lead investors to accept specific risk not remunerated. This is due to the fact that the marginal benefits of diversification decrease with the addition of each new asset in the portfolio. Given that diversification can be costly, in terms of transactions and information costs, this provides a rational for why most investors do not carry diversification to its logical limit and hold as many assets as they are available. Thus, managers worry about both macro and micro risks.

This is why three criteria should be used to manage portfolio selection. The one-factor model provides a simple description of stock returns, but unfortunately, it is unrealistic. When stocks are sensitive to interest rate risk as well as to market risk and firm-specific risks, the interest rate risk generate correlation between the market model residuals implying that more that one common factor generates stocks returns. Thus, as stated in the introduction, APT model is an efficient way to evaluate these three criteria, but the interpretation of the betas as risk measures is difficult for a non-specialist of APT.

Then, to obtain a more comprehensible methodology, applicable by any investors, it is necessary to translate the betas in criteria easy to understand.

\(^3\) Damodoran A.,1997, p.97.
Furthermore, if a return criterion and a residual risk criterion are simply added to several common risk criteria, the importance of the later is artificially increased. So we suggest creating a synthesis criterion for all the risks’ origin. A natural and common way to do it, is to assign each portfolio to a class of risk and to associate to this classification a discrete scale, ELECTRE TRI is well adapted to do so (section 3).

Then MCDA for portfolio selection can be performed with three criteria: return, common risk and residual risk, we will see in section below (section 4) that MINORA seems to be a good way to do it.

Concerning the decision set, it is best to use portfolio rather than securities because the securities evaluation with APT does not give reliable result, which is not the case with portfolios. To construct this decision set the portfolios were generated on the basis of the variance to obtain various risk levels.

2.2. Arbitrage Pricing Theory

The APT proposes a multifactor relation between return and risk under less restrictive hypotheses than CAPM. It supposes that the return of any asset, \( R_i \) \((i=1,2,\ldots,n)\), is equal to the expected return, \( E(R_i) \), plus an unexpected return \( (\sum_{k=1}^{K} b_{ik} F_k + e_i) \), where \( F_k \) is a common factor, \( b_{ik} \) is the sensibility coefficient of asset \( i \) to factor \( k \) and, \( e_i \) the specific risk component of the asset \( i \):

\[
R_i = E(R_i) + \sum_{k=1}^{K} b_{ik} F_k + e_i \text{ namely equation [I].}
\]

The common risk factors represent the surprises that meet the investors and that they cannot avoid. One estimates that on a long term, on average, these surprises compensate themselves.

To determine the APT fundamental relation, Ross develops an economic argument that is, at equilibrium we cannot have any arbitrage portfolio. This arbitrage portfolio must verify three condi-
tions: 1) no change in wealth, 2) no additional systematic risk, 3) no complementary return.

It follows from this reasoning combined with linear algebra that the expected return can be expressed as a linear combination. Hence, we obtain the APT fundamental relation called equation [II]:

\[ E(R_A) = \lambda_0 + \sum_{k=1}^{K} \lambda_k k \]

Where \( \lambda_0 \) is the intercept of the pricing relationship (zero–beta rate) and \( \lambda_k \) the risk premium on the k–Th. factor.

The above equality works because Ross assumes that the arbitrage portfolio error term variances become negligible when the number of assets increases. Here, Ross applies the law of large numbers to justify the convergence of a series of random variables to a constant; but this law is not sufficient and the residual term is not necessarily equal to zero. Taking into account these difficulties, several authors tried to use other derivations to replace the approximation with equality; this is done using two types of argument: an arbitrage argument or an equilibrium argument.

The APT based on equilibrium, linking itself to competitive market, offers a conceptual base stronger than the APT based on arbitrage. Nevertheless, as said before, the model needs enough securities without knowing both the number and the nature of risk factors (Cf. Ricci–Xella 1987, 1994, Batteau 1999).

Notice that the existence of factor structure in the equation [I] is a hypothesis while the equation [II] is an implication of the APT.

In this paper, we choose to use static APT, which is easier to implement than a conditional APT. The use of a conditional APT can be see as preferable. However, for a first proposal of combination of APT and MCDM, it is easier to suppose that the market is efficient in its informational sense than to create ex–ante data difficult to realise. Then, anticipation became realisation and then it is possible to use ex–post data. One of the conditions for this kind of APT is the non–variability of the parameters. Therefore, to test APT
on ex-post data, we suppose that reliable information does not vary from one period to another and is sufficient and available to determine the real anticipation. Nevertheless, it is true that this hypothesis can be rejected to choose a dynamic APT, that will probably give better result (for the most recent literature on this topic see Hwang and Satchell, 2000; Kan and Wang, 2000; Wang, 2000). This fact should be considered for future work.

The development of an empirical test for APT is presented in section 4.1 and 5.1.

2.3. The link between apt and MCDA

According to several empirical tests, APT seems to be a better model than CAPM because in APT there are several risk sources. Note that the APT is a normative model that imposes restrictive assumptions, even if they are less restrictive than in the CAPM. Then, it would be particularly interesting to adapt the model to a real investor preferences. Effectively, any investor is confronted with a given risk in a particular situation. Then, he has objectives and a risk attitude that are specifics to him, which should be taken into account.

Identifying several sources of risk, the APT pleads for a portfolio selection policy able to manage a multicriteria choice. Then, multicriteria decision making constitutes an interesting methodological framework to our study. Linking the multicriteria evaluation of asset portfolio and the research of a satisfactory solution for the investor, the MCDA methods allow us to take into account the investors’ specific objectives. Furthermore, these methods do not impose any normative scheme to the attitude of the investors. The use of multicriteria methods allows us to synthesise in a single procedure the theoretical and practical aspects of portfolio selection; thus it offers a non-normative use of the theory. Multicriteria decision making will facilitate and favour the analysis of compromise between the criteria. It equally permits to manage the hetero-
geneity of criteria's scale and the fuzzy and imprecise\textsuperscript{4} nature of the evaluation that it will contribute to clarify. The originality of MCDA equally offers, systematising the decision process, the possibility to obtain a gain of time and/or to increase the number of assets considered by the practitioner.

MCDA is not only a set of methods conceived to manage multicriteria decision problems. It is an activity that begins with the problem formulation (choice of a set of criteria, or construction of a multiobjective problem under constraints) and finish with the suggestion of a solution. Then, it is easy to see, again, the complementarity of APT and MCDA. Identifying the set of independent factors that best explain the return, the APT brings to MCDA the coherent family that it needs.

There are two approaches of multicriteria decision making.

- The descriptive approach supposes that the decision-maker has a stable and “rational” preference system that can be described by a utility function. Then, the utility function must be a good approximation of the decision-maker's preference system to find the “best solution”. There is no place in this approach for hesitations and incomparabilities.

- The constructive approach considers that in a decision process the intervention of human judgement makes the descriptive approach not appropriated (Cf. Roy 1987, 1992). The decision-maker's preferences are not stable, not very structured. The model used must accept hesitations, incomparabilities and in-transitivities.

We consider that the constructive approach is more close to reality and can help the decision-maker to solve his problem

\textsuperscript{4} Here the words fuzzy and imprecise refer to: (a) on one hand the delicacy of an investor’s judgement (the human nature and the lack of information), that will not always allow to discriminate between two close situations, and (b) on the other, the use of a representation model, which is a simplification of reality that expresses itself in an error term.
without normativity. Then, the multicriteria decision methods we choose to use, the MINORA interactive system and the ELECTRE TRI outranking method, belong to the constructive approach.

3. Structuring a Common Risk Criterion

It is necessary to construct a common risk criterion based on the betas issued from APT as shown in figure 1.

A natural and common way to do is to assign each portfolio to a class of risk and to associate to this classification a qualitative scale. ELECTRE TRI is well adapted to do so, because it is a method conceived to sort alternatives to different classes. Furthermore, for sorting problems an outranking method as ELECTRE TRI seems to be well adapted and present the advantage to admit incomparabilities and intransitivities.

This paragraph presents firstly the ELECTRE TRI method and secondly the way to construct a synthesis criterion for common risk.

3.1. The ELECTRE TRI method

ELECTRE TRI is a method especially conceived for sorting problems, it is used to sort the portfolio in risk categories. The categories are defined by some reference profiles that are fictitious portfolios defined by their values on the criteria. Let denote the categories \( C_i \), \( i=1,...,c \), and the profile \( r_i \). Where \( C_1 \) is the worst category, \( C_c \) the best one and \( r_i \) the theoretical limit between the categories \( C_i \) and \( C_{i+1} \). In ELECTRE TRI, the information asked from the decision-maker about his preferences takes the form, for each criterion and each profile, of a relative weight and indifference, preference and veto thresholds.

To sort the portfolios, ELECTRE TRI compare each of them to the profiles in order to construct a valued outranking relation, \( \sigma_s(a,b) \). It measures from 0 to 1 the strength of the relation “\( a \) outranks \( b \)” (\( a \) is at least as good as \( b \)). This valued relation is transformed in a
“net” outranking relation in the following way: $\sigma_s(a,b) \geq l \Leftrightarrow a \S b$, where $S$ represents the net outranking relation and $l$ a “cut level” ($0.5 \leq l \leq 1$) above which the relation $a$ outranks $b$ is considered as valid. ELECTRE TRI presents the advantage to accept intransitivity and incomparability\textsuperscript{5}.

In ELECTRE TRI there are two sorting procedures (the pessimistic one and the optimistic one) to assign each alternative into one of the categories defined beforehand. Generally the pessimistic procedure is applied when prudence is necessary. The optimistic procedure is applied when the decision-maker desires to promote alternatives that present some attractive characteristics. However, the best way to use ELECTRE TRI is certainly to compare the results of the two procedures. Indeed, in cases where some alternatives belong to different categories in both procedures, the conclusion is that they are incomparable with one or more reference profiles. This is explained by the fact that these alternatives have good values for some criteria and, simultaneously, bad values for other criteria. In this way the notion of incomparability included in ELECTRE TRI brings an important information to the decision-maker and for this reason the best way to employ ELECTRE TRI is to use the two assignment procedures and to compare the results.

The methods from the ELECTRE family are very popular, they have been used with success in a great number of studies and in portfolio selection by Martel et al. (1988), Szala 1990, Khoury et al. (1993) Hurson and Zopounidis (1995, 1996, 2000), Zopounidis et al. (1995).

3.2. Construction of a synthesis criterion for common risk

Five different classes of risk will be used to construct the common risk criterion (for betas):

As stated above the best use of ELECTRE TRI consists to use simultaneously the two available procedures (pessimistic and op-

\textsuperscript{5} For more details see appendix 1 or Yu 1992.
timistic procedures) because the comparison of their results brings an important information to the investor. Applying this principle we will use ELECTRE TRI sorting the portfolios in three categories to obtain finally five classes of risk. Let $(s)$ denote the best category $C_3$ (high risk) and the worst $C_1$ (low risk). The five risk classes are conceived using the following principles:

- The portfolios that belong to the best category ($C_3$) in both optimistic and pessimistic sorting can be recommended without hesitation to the portfolio manager. Then, the value of the synthesis criterion will be low risk, these portfolio constitute the top class for a risk averter.
- The portfolio belonging to the categories 2 and 3 can be considered as relatively attractive. Then, the value of the synthesis criterion will be relatively low risk.
- When the portfolios belong to the middle category ($C_2$) for both optimistic and pessimistic sorting, this means that these have moderate values on all criteria and, consequently, they present a medium risk.
- The portfolio belonging to the categories 1 and 2 can be considered as relatively not attractive. Then, the value of the synthesis criterion will be relatively high risk.
- The portfolios that belong to the worst category ($C_1$) in both optimistic and pessimistic sorting will not be recommend to the portfolio manager. Then, the value of the synthesis criterion will be high risk; these portfolios constitute the worst class for a risk averter.
- The portfolios belonging to the categories 1 and 3 are incomparable with the two profiles. This means that these portfolios have good values for some criteria and, simultaneously, bad values for other criteria. In this case, for the purpose of the illustration we will assign in section 4 these portfolios to third class by default because it is the medium category. However, it is rather arbitrary to assign the portfolio to a class of risk on
this basis. The solution in this case is to let the decision-maker affect himself the portfolios to a category. Effectively, in this case, considering the information at its disposal, the method cannot give an affectation beyond all doubt, it encounter some difficulties to conclude because the portfolios are incomparable with the profiles. Another solution is also to create a category of particular portfolios "unclassable". The use of complete aggregation method as UTADIS or PREFDIS (UTA-related method applied for sorting problems, c.f. Doumpos and Zopounidis 1998, Zopounidis and Doumpos 1998) is also possible. This method will be probably easier to use on this context because it requires only a sorting of some reference projects. However, we loose the advantage of the acceptation of incomparabilities, which allow highlighting particular portfolios that can make problem. Note also that UTADIS will probably affect those portfolios in a medium category, joining the solution chosen by default in the section 4. Finally, the choice of the method depends of the investor and his ability to manage the information necessary to use ELECTRE TRI, especially for the determination of the profile. A solution could be that the investor assigns himself these portfolios to class of risk depending on its risk behaviour. In others words an optimistic investor will assign the portfolio to the best classes and conversely for a pessimistic one.

As was indicated, the objective is to sort the portfolios in three categories: attractive portfolio \( (C_3) \), uncertain portfolio to be studied further \( (C_2) \), non-attractive portfolio \( (C_1) \).

4. **Towards a Trichotomic Decision Support**

After the construction of the common risk criterion with ELECTRE TRI, MCDA for portfolio selection can be performed with three criteria: return, common risk and residual risk. To do so, we decide to rank portfolios from the “best” to the “worst”, because it is a common and natural preoccupation of analysts. The assump-
tion of existence of an order is the same as the existence of a utility function. Then, the outranking methods that reject the axiomatic of utility do not seem to be well adapted to solve ranking problems, the use of an interactive method, as MINORA seems to be a better way.

This section begins presenting the implementation of APT that gives return, betas and residual risk that are necessary to obtain the three criteria. Then, the MINORA system used to rank the portfolio is developed.

4.1. The implementation of APT

The tests on APT consist in the identification of the number and the nature of risk factors.

Concerning the determination of the number of factors:

In APT, two methods are generally proposed:
- either one determines the sufficient number of factors, using data analysis like Principal Component Analysis or using Maximum Likelihood Factor Analysis as in Roll and Ross (1980) or Dhrymes, Friend, Gultekin and Gultekin (1984, 1985), ...
- or to prespecify the number of factors to test APT validity as in Chen (1983), Oldfield and Rogalski (1981).

Concerning the identification of the factors:
- One often uses the Fama and MacBeth’s techniques (1973) about CAPM relating these factors to exogenous variables. Effectively in the first version of the APT the nature of the factors was unknown, then without any attractive sense for commercial practitioner. Roll and Ross (1980) were among the first to look specifically for APT factors.
- Afterwards, the following version of APT gave an economic interpretation to the factors easily comprehensible, then accept—

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6 Or endogenous with completude tests, in order to study if reflected risk in the covariance matrix of return are valuable and no other anomalies appear (size effect, week–end effect, moments of distribution,...).
able, by portfolio managers. Following Roll and Ross' paper (1983), because more than half of the realised return is the result of non anticipated variations, the systematic forces which influence returns are those that cause variations of interest rate and revenues. Chen, Roll and Ross (1983), who derived the common factors from a set of data and then tested them for their relationship to fundamental macroeconomic variables made the most famous study.

- Finally, the mesoeconomic\textsuperscript{7} APT is an interesting solution but it is expensive to collect monthly accounting and financial data.

This is why we use the macroeconomic version of the APT to determine a pre-specified number of factors.

In the first step, the test consists in determining the number of significant factors. The matrix of sensibilities is calculated using time-series to regress odd asset returns on unexpected macroeconomic variables plus a market index portfolio. In the second step, a cross-sectional regression on even portfolio returns using the sensibilities as independent variables (we assume the beta constant over the period of time) in which the parameters being estimated are $\lambda_0$ (zero-beta or the return on the riskless asset) and $\lambda_k$ (the risk premiums). Finally, the third step suggests to use a battery test (F-test, t-student, Root Mean Square Error, $R^2$...) to determine the number of pertinent factors.

4.2. The MINORA System

MINORA is an interactive system that ranks, from the best to the worst, a set of alternatives. On the basis of a ranking made by the decision-maker on a subset of well-known alternatives, MINORA

\textsuperscript{7} The term mesoeconomic signifies that the risk factors are both micro with financial or accounting components and macroeconomic variables (Cf. Ricci-Xella, 1994).
uses ordinal regression to estimate a set of separable additive value functions of the following form:

\[ u(g) = u_1(g_1) + u_2(g_2) + ... + u_c(g_c) \]

Where \( g=(g_1, ..., g_c) \) is the performance vector of an alternative and \( u_i(g_i) \) is the marginal value function of criteria i, normalised between 0 and 1).

In MINORA the interaction takes the form of an analysis of inconsistencies between the ranking established by the decision-maker and the ranking issue from the additive value function. The interaction is organised around the following questions presented to the decision-maker:
- Is he ready to modify his ranking?
- Does he wish to modify the relative importance of a criterion, its scale or the marginal utilities (trade off analysis)?
- Does he wish to modify the family of criteria used: to add, cancel, modify, divide or join some criteria
- Does he wish to modify the whole formulation of the problem?

These questions send back to the corresponding stages of MINORA and the method stops when an acceptable compromise is determined. Then, the result (a value function) is extrapolated to the whole set of alternative to give a ranking of them.

The interactive estimation of the additive value function in MINORA helps the decision-maker to construct his model in a non-normative way. It organises, in a unique procedure, all the activity of decision making, from the model formulation to the result. In the same time the decision-maker is constantly integrated to the resolution processes and can control its evolution at any moment. Finally, notice that MINORA method had been used successfully to

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8 The term utility used in a stochastic context is replaced by the term value in a determinist context.

9 See appendix 1 for a presentation of the estimation program and Despotis et al. (1990) for more details.


5. An Application to French Stock Market

In this section the application of the methodology to the French Stock Market is presented, first the implementation of APT formulation to obtain the criteria values and then the applications of ELECTRE TRI and MINORA using those criteria.

5.1. The APT

The methodology to obtain persuasive (reliable) exogenous risk factors is the following.

Our sample consists of return on Paris Stock Exchange firms (SDIB–SBF) from November 1983 through September 1991. The alterations of capital are taken into account and the prices corrected. Macroeconomic variables came from O.E.C.D. and I.N.S.E.E\(^\text{11}\). We determine unexpected changes in those 28 variables normalised factors and some will be eliminated because they are too correlated between them. The others were used to create eleven unanticipated or non-expected macroeconomic variables that we employed (Cf. Table 2) and the French market index CAC240 (SDIB–SBF that contains at the maximum the 250 most important French securities) is added to this to obtain an equality in the pricing relationship. One can assert that in selecting the number of factors in an APT, there may be a problem of mis-specification plus "simultaneity bias" when the market portfolio is included as a factor. However, we tested versions of APT containing CAC 40 and versions not containing CAC 40, the results indicate that this factor is significant (see Ricci 1994). The riskless asset is represented by short-term interest rate in end

\(^{11}\) For brevity, we don’t present the preliminary statistic analysis of the data set used in order to concentrate the paper to its problematic, this statistic analysis is available upon request (Ricci 1994).
of period (or PIBOR–3months). We adopt 28 portfolios of 6 assets each as advised by Gibbons–Ross–Shanken (1989). These portfolios are generated using monthly logarithmic returns (Cf. Ricci–Xella 1994).

The APT version used in this study regresses 11 normalised un–expected and/or unanticipated macro–economic variables plus the market index. These variables were regressed on the return of a set of portfolios that was generated following their capitalisation. In this version we take into consideration only the test of APT on 28 portfolio of 6 stocks (from 16 possibilities), that have the lower value of the Root Mean Square Error (RMSE), as one can see on table 3. This valuation of APT is issue versus those from alternative static APT versions (APT with unknown factors, macro–economic APT, etc.) and CAPM (cf. Ricci 1994).

In the majority of the APT tests, the variables that best identify the risk factors were, in decreasing order: the constant component, the consumption price index (Macro11) and the market portfolio (CAC240). The presence of consumption price index was surprising, because its beta is significant in opposition to he result obtained in the studies performed on the US market. Then, another test of the APT have been performed with data from INSEE, instead OCDE, and the result was the same. This result can be explicated by the particularity of the French stock market at that time (cf. Ricci 1994). Then, came some variables that were valuable with a significant level between 0 and 5% (in decreasing order): The risk premium (Macro6), the monthly growth of industrial production (Macro1), and the growth rate of money (Macro9). The other variables were not significant enough. This is why this application uses the following macro–economic variables: macro 1, 6, 9, 11 and the CAC240, as in Hurson and Ricci–Xella's previous study (1997). The sensibilities factors called beta (thus beta 1 correspond to the macro 1), obtained with the APT model are presented in table 4.
After the evaluation of the portfolio on the chosen criteria (betas, return and specific risk), it is now possible to perform the multicriteria analysis.

5.2. Application of ELECTRE TRI

The first step of the multicriteria analysis, as stated above, is the construction of a synthesis criterion for common risk with ELECTRE TRI.

At the moment of the study, the French stock market was a bull market. Then, the investors choose to maximise the beta of the non-anticipated factors (growth of industrial production, risk premium variations, growth rate of money, CAC240) for which the growth is good for investors because they give more return. Conversely, the beta associated to the variations of consumption price index (Beta 11) is minimised; indeed the growth of unexpected inflation is bad for investor because it gives less return. Then, as it is explained in the presentation of the methodology, ELECTRE TRI is used to create a synthesis criterion for common risk. The criteria weight, the reference profiles and the thresholds used to compute the discordance and concordance indexes in ELECTRE TRI are presented in the table 5.

It belongs to investor to set the value of profiles in order to define the three categories. Those values must be fixed considering the risk attitude of the investor. We suppose that this investor decides to give the above values to the profile.

The criteria's weights are all equal to 1 because all the criteria have a priori the same importance. The indifference and the preference profiles are perception thresholds. The indifference threshold gives the value below which the decision-maker considers that the difference between two portfolios is not significant. Then, in our application, a difference lower than 0.1 on the criterion beta 1 is considered as not significant. The preference threshold gives the value above which, a difference between two portfolios imply a certain (strong) preference for one of them, considering the criterion examined. For
example, in our study, a difference greater than 0.5 between two portfolios on the criterion beta 1, imply, considering this criterion alone, a strong preference for one of these two portfolios. Here, the values of these thresholds are the same for the four first criteria because their scales have similar ranges of values. The veto threshold has a different nature. It gives the value above which a difference on the criterion between two portfolios a and b, in favour of a, imply the rejection of the outranking of b by a (“b is at least as good as a”), even if b has better values on the others criteria. In ELECTRE TRI, the portfolios are compared to the preference profiles. Then, the veto threshold has for effect to forbid, in the pessimistic procedure, the sorting of a portfolio in a category if at least one criterion is in favour of the low profile of this category, with a difference superior to the veto threshold. If this situation appears, the criterion responsible of the veto becomes decisive. Then, considering that all our criteria have the same type of significance (sensitivity to a macro-economic variable), and that none of them must take a particular importance, we decide to note use veto threshold. This is done fixing the veto threshold to their default value that corresponds to the maximum of the criterion (then the vetoes are not active).

In ELECTRE TRI, the default cut level is $l=0.67$, that corresponds to a majority level of $2/3$ (various studies and the experience show that this majority level seems to be often appropriate). Table 6 presents ELECTRE TRI’s results and the synthesis criterion. Several values of $l$ have been tested; the result obtained was stable for values of $l$ from 0.61 to 0.80. Then, the results present good properties of stability.

The portfolio P7, P9, P22 and P23 are affected to category 3 in optimistic procedure and to category 1 in pessimistic one. That means, as stated above, that these portfolios are considered as good following some criteria and bad following others. Thus, those portfolios are affected a priori to the medium class of risk.
A sensitivity analysis has been done concerning the parameters used in ELECTRE TRI and in MINORA. The results of this analysis show that the proposed solutions present good properties of stability\(^\text{12}\).

Then, a synthesis criterion for common risk is constructed. It will be used, with the return and the residual risk, to perform a ranking of the portfolios with the help of MINORA.

5.3. Application of MINORA

Table 7 presents the evaluation of the portfolios on the three criteria used in MINORA.

To apply the MINORA system, it is necessary to have a reference set of portfolios and a ranking expressed by the portfolio manager on this reference set. The reference set must respect two principles: 1) include well-known stocks by the portfolio manager; 2) the portfolios of this set must cover all the range of possibilities. For this study, a set of 9 not linked portfolios have been chosen and ranked with the help of some utilities in MINORA software. Then MINORA system, through the UTA STAR method, provided the following additive value model:

\[
u(g) = 0.26372u_{\text{risk}}(\text{Risk}) + 0.34831u_{\text{return}}(\text{Return}) + 0.38796u_{\text{resid}}(\text{Residual})\]

This value function is the most appropriate, since rank correctly all the portfolios of the reference set, indicating complete agreement between the portfolio manager and the additive value model. Table 8 presents the reference set, the ranking of the decision maker, the global value of the portfolios and the ranking issue from the additive value model.

The marginal value functions corresponding to this model are presented in the figures 2 to 4. Figure 5 shows how the global value is decomposed in marginal utilities for each portfolio of the

\(^{12}\) For brevity these results are not reported in this paper, they are available upon request.
Structuring Portfolio Selection Criteria for Interactive Decision Support

The observation of the results shows that all the criteria have almost the same weight. Nevertheless, this is not sufficient to appreciate the relative importance of a criterion; the later depends also on the discriminatory power of the criteria. The discriminatory power of a criterion depends on the shape of its marginal value function. This is all the more important since the slope of the marginal value curve is high (if the curve is flat this means that all the portfolios have the same value on the criterion and, then, this criterion has no effect). Then, observing the figures, 2 to 4, one can see that the three criteria have a strong discriminatory power on all their scale. This finding is confirmed examining the figure 5, one can see that the marginal values for a same criterion vary from one portfolio to the other.

The results of MINORA after extrapolation of the value function to the whole set of portfolio are presented in table 9.

The four first portfolios have good values on the criteria: return superior to 2.45 % (the mean is 2.43%), low or very low residual risk and medium, good or very good level of common risk. The best portfolio is the portfolio 24, this portfolio presents one of the bests returns (2.93%, only 6 portfolios present a better return), a very low residual risk and a medium common risk. In second position one can found the portfolio 23; this portfolio has the highest return (4.02%) but its residual risk is superior to the portfolio 24 one. Portfolios 21 and 28, in third and fourth positions have lower values of return but are better on common risk.

6. Concluding Remarks

In this paper, we have presented a methodology for portfolio selection that exploits the complementarity and the respective advantages of APT and MCDA. The APT enables us to efficiently evaluate the return of portfolios and, by identifying the relevant common factors of influence, gives the way to perform a multicriteria management of risk. First, in order to make the proposed methodology
accessible to any investor, the five common risk criteria of origin (betas issue from the APT) are translated in simple and easily comprehensible criterion with a concrete and economic sense. This allows the investors and the portfolio managers to manage easily these criteria and the proposed multicriteria methods. In this way, the investor or the portfolio manager will be able to analyze easily the consequences of his choices on the level of taken risk. Furthermore, ELECTRE TRI allows creating a synthesis criterion for all the common risk origins. Then, MCDA provides an original and efficient framework to conduct a portfolio selection using the three criteria portfolio selection approach identified by the APT. This point is obtained by adding to the synthesis common risk criterion two other criteria to take into account the return and the residual risk, this is important to perform an efficient portfolio selection. Finally, these three criteria are used to rank the portfolios from the best to the worst using the MINORA interactive system.

MCDA allows to take into consideration the portfolio manager preferences and all the relevant criteria, whatever their origins, for portfolio selection. Done this way, the portfolio selection model is undertaken without any normative constraints. Finally, this methodological framework brings a new knowledge to portfolio selection and helps with the improvement of a scientific and active portfolio management.

This paper constitute a new and original direction of research, to the best of our knowledge there is no others propositions linking PAT and MCDM. Nevertheless, this first proposition is perfectible for future research:

- To propose a dynamic extension of the present study using a dynamic APT.
- To compare the ranking issue from our methodology versus ranking (even heuristic) from others experts.
- To test the present methodology on real cases (with the cooperation of a real portfolio manager).
References


Hurson Ch. and Ricci–Xella N., “Multicriteria Decision Making and Portfolio Management with Asset Pricing Theory” in C. Zo–


Hurson Ch., La gestion de portefeuilles boursiers et l'aide multicritère à la décision, Thèse de Doctorat, GREQAM., Université d'Aix-Marseille II, 1995.


Siskos J., Spiridakos A. and Yannacopoulos D., "MINORA: “A multicriteria decision aiding system for discrete alternatives”, in J. Siskos and C. Zopounidis (ed.), *Special Issue on Multicriteria De-


Table 1: Definition of risk classes

<table>
<thead>
<tr>
<th>Class of risk</th>
<th>Preference order*</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Structuring Portfolio Selection Criteria for Interactive Decision Support

Low risk 1  
Relatively low risk 2  
Medium risk 3  
Relatively high risk 4  
High risk 5  

* These ranking correspond to a risk averter: class 1 is better than class 5 (the ranking will be inverted for a risk lover). This is a very common and not restrictive hypothesis, in all the theoretical models (APT, CAPM, etc.) the investor is suppose to be risk averse because this correspond to the great majority of investors. Then the hypothesis of risk aversion is a condition for the use of APT.

Table 2: Definition of unexpected or unanticipated macro-economic variables

<table>
<thead>
<tr>
<th>N°</th>
<th>Symbol</th>
<th>Name</th>
</tr>
</thead>
<tbody>
<tr>
<td>MACRO1</td>
<td>MP&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Monthly growth of industrial production</td>
</tr>
<tr>
<td>MACRO2</td>
<td>UP&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Annual growth of industrial production</td>
</tr>
<tr>
<td>MACRO3</td>
<td>UEI&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Non expected inflation</td>
</tr>
<tr>
<td>MACRO4</td>
<td>DEI&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Variations of expected inflation</td>
</tr>
<tr>
<td>MACRO5</td>
<td>UPR&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Risk premium variation (by bond)</td>
</tr>
<tr>
<td>MACRO6</td>
<td>CACEXC&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Risk premium variation (by CAC240)</td>
</tr>
<tr>
<td>MACRO7</td>
<td>TS&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Variation of the interest rate term structure</td>
</tr>
<tr>
<td>MACRO8</td>
<td>XM&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Commercial balance variation</td>
</tr>
<tr>
<td>MACRO9</td>
<td>HT&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Growth rate of money</td>
</tr>
<tr>
<td>MACRO10</td>
<td>ATL&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Market index variation (logarithmic)</td>
</tr>
<tr>
<td>MACRO11</td>
<td>VCI&lt;sub&gt;t&lt;/sub&gt;</td>
<td>Variation of consumption prices index</td>
</tr>
</tbody>
</table>

Table 3: RMSE test

<table>
<thead>
<tr>
<th>grouping type</th>
<th>Value of RMSE</th>
</tr>
</thead>
<tbody>
<tr>
<td>RR012: 14 portfolios of 12 stocks</td>
<td>0.0001827</td>
</tr>
<tr>
<td>RR008: 21 portfolios of 8 stocks</td>
<td>0.0005599</td>
</tr>
<tr>
<td>RR007: 24 portfolios of 7 stocks</td>
<td>0.0005946</td>
</tr>
<tr>
<td>RR006: 28 portfolios of 6 stocks</td>
<td>0.0004834</td>
</tr>
</tbody>
</table>
Table 4: 

The chosen criteria sensibility (in order to facilitate the reading the betas presented below are multiplied by 1000, hence for P2 beta CAC240 = 1.413).

<table>
<thead>
<tr>
<th>Portfolio</th>
<th>Beta 1 (growth of industrial production)</th>
<th>Beta 6 (risk premium variations)</th>
<th>Beta 9 (growth rate of money)</th>
<th>Beta 11 (consumption price index)</th>
<th>Beta CAC240</th>
</tr>
</thead>
<tbody>
<tr>
<td>P1</td>
<td>2.65</td>
<td>9.34</td>
<td>0.86</td>
<td>-0.69</td>
<td>-154.05</td>
</tr>
<tr>
<td>P2</td>
<td>4.15</td>
<td>-2.66</td>
<td>4.54</td>
<td>-1.85</td>
<td>1210.42</td>
</tr>
<tr>
<td>P3</td>
<td>4.60</td>
<td>3.74</td>
<td>3.92</td>
<td>-1.68</td>
<td>593.63</td>
</tr>
<tr>
<td>P4</td>
<td>-2.89</td>
<td>-5.98</td>
<td>-1.99</td>
<td>1.62</td>
<td>1413.65</td>
</tr>
<tr>
<td>P5</td>
<td>3.13</td>
<td>3.86</td>
<td>2.33</td>
<td>-2.91</td>
<td>449.65</td>
</tr>
<tr>
<td>P6</td>
<td>0.91</td>
<td>4.86</td>
<td>0.53</td>
<td>0.18</td>
<td>292.95</td>
</tr>
<tr>
<td>P7</td>
<td>1.44</td>
<td>-9.07</td>
<td>2.07</td>
<td>2.91</td>
<td>1941.50</td>
</tr>
<tr>
<td>P8</td>
<td>1.18</td>
<td>3.99</td>
<td>2.69</td>
<td>-1.63</td>
<td>361.50</td>
</tr>
<tr>
<td>P9</td>
<td>1.56</td>
<td>11.64</td>
<td>2.15</td>
<td>2.34</td>
<td>-314.95</td>
</tr>
<tr>
<td>P10</td>
<td>0.14</td>
<td>0.55</td>
<td>-0.25</td>
<td>-0.28</td>
<td>402.25</td>
</tr>
<tr>
<td>P11</td>
<td>-1.12</td>
<td>-1.78</td>
<td>-1.47</td>
<td>-0.42</td>
<td>1356.25</td>
</tr>
<tr>
<td>P12</td>
<td>-1.97</td>
<td>1.76</td>
<td>-1.89</td>
<td>0.90</td>
<td>682.01</td>
</tr>
<tr>
<td>P13</td>
<td>-2.50</td>
<td>4.65</td>
<td>-0.45</td>
<td>3.49</td>
<td>526.68</td>
</tr>
<tr>
<td>P14</td>
<td>0.27</td>
<td>3.21</td>
<td>-0.05</td>
<td>-0.98</td>
<td>269.82</td>
</tr>
<tr>
<td>P15</td>
<td>-0.16</td>
<td>6.48</td>
<td>-1.55</td>
<td>0.87</td>
<td>54.17</td>
</tr>
<tr>
<td>P16</td>
<td>0.31</td>
<td>17.22</td>
<td>2.52</td>
<td>0.04</td>
<td>-252.06</td>
</tr>
<tr>
<td>P17</td>
<td>0.01</td>
<td>1.47</td>
<td>0.36</td>
<td>0.65</td>
<td>829.00</td>
</tr>
<tr>
<td>P18</td>
<td>0.59</td>
<td>2.35</td>
<td>3.09</td>
<td>2.69</td>
<td>747.38</td>
</tr>
<tr>
<td>P19</td>
<td>-0.01</td>
<td>-4.5</td>
<td>2.16</td>
<td>-0.66</td>
<td>1376.23</td>
</tr>
<tr>
<td>P20</td>
<td>0.35</td>
<td>7.08</td>
<td>0.59</td>
<td>1.92</td>
<td>77.84</td>
</tr>
<tr>
<td>P21</td>
<td>0.99</td>
<td>-0.04</td>
<td>0.97</td>
<td>-1.54</td>
<td>931.67</td>
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</table>
Table 5: Suggested parameters in ELECTRE-TRI

<table>
<thead>
<tr>
<th>Criteria weight</th>
<th>Beta 1</th>
<th>Beta 6</th>
<th>Beta 9</th>
<th>Beta 11</th>
<th>Beta CAC240</th>
</tr>
</thead>
<tbody>
<tr>
<td>High profile</td>
<td>0.56</td>
<td>3.8</td>
<td>0.9</td>
<td>0.7</td>
<td>760</td>
</tr>
<tr>
<td>Low profile</td>
<td>-0.1</td>
<td>1</td>
<td>-0.3</td>
<td>-0.8</td>
<td>380</td>
</tr>
<tr>
<td>Indifference threshold</td>
<td>0.1</td>
<td>0.1</td>
<td>0.1</td>
<td>0.1</td>
<td>10</td>
</tr>
<tr>
<td>Preference threshold</td>
<td>0.5</td>
<td>0.5</td>
<td>0.5</td>
<td>0.5</td>
<td>50</td>
</tr>
<tr>
<td>Veto threshold for high profile</td>
<td>4.6</td>
<td>17.22</td>
<td>4.54</td>
<td>3.49</td>
<td>1941.5</td>
</tr>
</tbody>
</table>

Table 6: Result of ELECTRE TRI and synthesis criterion

<table>
<thead>
<tr>
<th>Portfolio</th>
<th>Pessimist</th>
<th>Optimist</th>
<th>Synthesis</th>
<th>Portfolio</th>
<th>Pessimist</th>
<th>Optimist</th>
<th>Synthesis</th>
</tr>
</thead>
<tbody>
<tr>
<td>P1</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
<td>P15</td>
<td>1</td>
<td>1</td>
<td>High risk</td>
</tr>
<tr>
<td>P2</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
<td>P16</td>
<td>2</td>
<td>3</td>
<td>Relatively low risk</td>
</tr>
<tr>
<td>P3</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
<td>P17</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
</tr>
<tr>
<td>P4</td>
<td>1</td>
<td>1</td>
<td>High risk</td>
<td>P18</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
</tr>
<tr>
<td>P5</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
<td>P19</td>
<td>2</td>
<td>3</td>
<td>Relatively low risk</td>
</tr>
<tr>
<td>P6</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
<td>P20</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
</tr>
<tr>
<td>P7</td>
<td>1</td>
<td>3</td>
<td>Medium risk</td>
<td>P21</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
</tr>
<tr>
<td>P8</td>
<td>3</td>
<td>3</td>
<td>Low risk</td>
<td>P22</td>
<td>1</td>
<td>3</td>
<td>Medium risk</td>
</tr>
<tr>
<td>P9</td>
<td>1</td>
<td>3</td>
<td>Medium risk</td>
<td>P23</td>
<td>1</td>
<td>3</td>
<td>Medium risk</td>
</tr>
<tr>
<td>P10</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
<td>P24</td>
<td>2</td>
<td>2</td>
<td>Medium risk</td>
</tr>
</tbody>
</table>
Table 7: Portfolio evaluation criteria used in MINORA

<table>
<thead>
<tr>
<th>Portfolio</th>
<th>Residual risk</th>
<th>Return (%)</th>
<th>Common risk synthesis criterion</th>
</tr>
</thead>
<tbody>
<tr>
<td>P1</td>
<td>1.44</td>
<td>3.36</td>
<td>Low common risk</td>
</tr>
<tr>
<td>P2</td>
<td>1.33</td>
<td>1.80</td>
<td>Low common risk</td>
</tr>
<tr>
<td>P3</td>
<td>1.91</td>
<td>2.42</td>
<td>Low common risk</td>
</tr>
<tr>
<td>P4</td>
<td>1.31</td>
<td>2.117</td>
<td>High common risk</td>
</tr>
<tr>
<td>P5</td>
<td>0.66</td>
<td>1.75</td>
<td>Low common risk</td>
</tr>
<tr>
<td>P6</td>
<td>0.76</td>
<td>1.63</td>
<td>Medium common risk</td>
</tr>
<tr>
<td>P7</td>
<td>1.68</td>
<td>3.46</td>
<td>Relatively high common risk</td>
</tr>
<tr>
<td>P8</td>
<td>0.45</td>
<td>1.22</td>
<td>Low common risk</td>
</tr>
<tr>
<td>P9</td>
<td>1.59</td>
<td>1.62</td>
<td>Relatively high common risk</td>
</tr>
</tbody>
</table>
Table 8: Ranking of the reference set portfolios

<table>
<thead>
<tr>
<th>Decision maker ranking</th>
<th>1</th>
<th>2</th>
<th>2</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
</tr>
</thead>
<tbody>
<tr>
<td>portfolio</td>
<td>P21</td>
<td>P8</td>
<td>P22</td>
<td>P3</td>
<td>P20</td>
<td>P27</td>
<td>P2</td>
<td>P4</td>
<td>P9</td>
</tr>
<tr>
<td>Global value</td>
<td>0.46</td>
<td>0.36</td>
<td>0.36</td>
<td>0.35</td>
<td>0.24</td>
<td>0.17</td>
<td>0.16</td>
<td>0.15</td>
<td>0.14</td>
</tr>
<tr>
<td></td>
<td>45</td>
<td>80</td>
<td>80</td>
<td>89</td>
<td>04</td>
<td>56</td>
<td>65</td>
<td>74</td>
<td>16</td>
</tr>
</tbody>
</table>

| Value model ranking | 1 | 2 | 2 | 4 | 5 | 6 | 7 | 8 | 9 |

| P10 | 0.33 | 0.74 | Medium common risk |
| P11 | 1.2 | 3.59 | Medium common risk |
| P12 | 1.03 | 2.36 | Relatively high common risk |
| P13 | 1.01 | 2.08 | High common risk |
| P14 | 0.74 | 1.10 | Medium common risk |
| P15 | 0.28 | 1.47 | Medium common risk |
| P16 | 1.38 | 3.97 | Relatively low common risk |
| P17 | 0.68 | 2.42 | Medium common risk |
| P18 | 0.91 | 2.61 | Medium common risk |
| P19 | 0.93 | 2.86 | Medium common risk |
| P20 | 0.56 | 1.70 | Medium common risk |
| P21 | 0.62 | 2.75 | Low common risk |
| P22 | 0.38 | 3.54 | Relatively high common risk |
| P23 | 0.46 | 4.02 | Medium common risk |
| P24 | 0.24 | 2.93 | Medium common risk |
| P25 | 0.15 | 2.98 | Medium common risk |
| P26 | 0.31 | 2.78 | Medium common risk |
| P27 | 0.51 | 2.30 | High common risk |
| P28 | 0.49 | 2.45 | Low common risk |
Table 9: Results of MINORA (final ranking of portfolios)

<table>
<thead>
<tr>
<th>Portfolio</th>
<th>Value</th>
<th>Ranking</th>
<th>Portfolio</th>
<th>Value</th>
<th>Ranking</th>
</tr>
</thead>
<tbody>
<tr>
<td>P24</td>
<td>0.6086</td>
<td>1</td>
<td>P17</td>
<td>0.2897</td>
<td>15</td>
</tr>
<tr>
<td>P23</td>
<td>0.5428</td>
<td>2</td>
<td>P19</td>
<td>0.2646</td>
<td>16</td>
</tr>
<tr>
<td>P21</td>
<td>0.4645</td>
<td>3</td>
<td>P18</td>
<td>0.2579</td>
<td>17</td>
</tr>
<tr>
<td>P28</td>
<td>0.4495</td>
<td>4</td>
<td>P7</td>
<td>0.2519</td>
<td>18</td>
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<tr>
<td>P16</td>
<td>0.4436</td>
<td>5</td>
<td>P20</td>
<td>0.2403</td>
<td>19</td>
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<tr>
<td>P25</td>
<td>0.4308</td>
<td>6</td>
<td>P12</td>
<td>0.2377</td>
<td>20</td>
</tr>
<tr>
<td>P26</td>
<td>0.4298</td>
<td>7</td>
<td>P6</td>
<td>0.2350</td>
<td>21</td>
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<tr>
<td>P5</td>
<td>0.4011</td>
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<td>P1</td>
<td>0.2291</td>
<td>22</td>
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<td>P8</td>
<td>0.3680</td>
<td>9</td>
<td>P14</td>
<td>0.2069</td>
<td>23</td>
</tr>
<tr>
<td>P22</td>
<td>0.3680</td>
<td>9</td>
<td>P27</td>
<td>0.1756</td>
<td>24</td>
</tr>
<tr>
<td>P3</td>
<td>0.3589</td>
<td>11</td>
<td>P2</td>
<td>0.1665</td>
<td>25</td>
</tr>
<tr>
<td>P15</td>
<td>0.3449</td>
<td>12</td>
<td>P4</td>
<td>0.1574</td>
<td>26</td>
</tr>
<tr>
<td>P11</td>
<td>0.3422</td>
<td>13</td>
<td>P9</td>
<td>0.1415</td>
<td>27</td>
</tr>
<tr>
<td>P10</td>
<td>0.3167</td>
<td>14</td>
<td>P13</td>
<td>0.1187</td>
<td>28</td>
</tr>
</tbody>
</table>
Figure 1. Tri-criteria ranking method

DATA BASE
APT

RETURN
RESIDUAL RISK

COMMON RISK
(Beta)

Creation of a synthesis criteria for common risk with ELECTRE TRI

Ranking with MINORA

Figure 2: Value curve, return

Figure 3: Value curve, common risk

Figure 4: Value curve, residual risk

Figure 5: Reference set portfolio
APPENDIX 1: ELECTRE-TRI

Firstly, a partial concordance index is calculated for each criterion and each couple (portfolio, profile). The partial concordance index $C(a \ S \ r_i)$ represents between 0 and 1 the strength of the relation "a is at least as good as $r_i$"; this index is constructed as the following schema.

$q_i$ is an indifference threshold and $p_i$ is a preference threshold.

Then a global concordance index is calculated as follow:

$$ C(a, r_j) = \frac{\sum_{i=1}^{k} k_i c_i(a, r_j)}{\sum_{i=1}^{k} k_i} $$

$k_i$ represents the importance of criterion $i$. 
Therefore, a partial discordance index is calculated. It measures the opposition to the affirmation "a is at least as good as r_i", this index is constructed as in the following schema.

\[ D_i(a, r_j) \]

\[ g(i) - p_i \]

\[ g(i) - v_i \]

\[ G_i(a) \]

\[ D_i(a, r_j) \]

\[ G_i \]

\[ G \]

\[ G_i(a, r_j) = \begin{cases} C(a, r_j) & \text{si } G_i(a, r_j) = \emptyset \\ C(a, r_j) \prod_{i \in G} \frac{1 - D_i(a, r_j)}{1 - C(a, r_j)} & \text{si } G_i(a, r_j) \neq \emptyset \end{cases} \]

\( G \) is the set of criteria and:

\[ G_i(a, r_j) = \{ i \in G : D_i(a, r_j) > C(a, r_j) \} \]

This valued outranking relation is transformed in a net outranking relation \( S \) as stated in the section 3.1, and this net relation in a preference, indifference or incomparability relation as follows:

\[ a \sim r_j \quad a \equiv r_j \quad a \not\equiv r_j \quad a \not\sim r_j \quad a \not\equiv r_j \]

\( \not\equiv \) represents incomparability.

The pessimistic and optimistic affectation are performed as in the two following schemas:
No incomparabilities, identical affectations

Presence of incomparabilities, different affectations
APPENDIX 2: MINORA

\[
\text{Min } F = \sum_{a \in \mathcal{P}} (\sigma_a^+ + \sigma_a^-)
\]

\[
\sum_{i=1}^{k} \left[ u_i(g(a)) - u_i(g(b)) \right] + \sigma_a^+ - \sigma_a^- - \sigma_b^+ + \sigma_b^- \geq d \ \forall (a,b) \in \mathcal{P} \times \mathcal{P} / a \sim b
\]

\[
\sum_{i=1}^{k} \left[ u_i(g(a)) - u_i(g(b)) \right] + \sigma_a^+ - \sigma_a^- - \sigma_b^+ + \sigma_b^- = 0 \ \forall (a,b) \in \mathcal{P} \times \mathcal{P} / a \approx b
\]

\[
u_i(g_a) = 0, \sum_{i=1}^{k} u_i(g_i^+) = 1, \forall i = 1..k
\]

\[
\sigma_a^+ \geq 0, \sigma_a^- \leq 0, \forall a \in \mathcal{P}
\]

\(\mathcal{P}\) represents the reference set, 
a and b two portfolios,
a \sim b means "a is preferred to b", 
a \approx b means "a is indifferent to b",
the variable \(\sigma_a^+\) and \(\sigma_a^-\) are error variables, representing potential errors concerning the utility of a,
Constraints 1, the parameter \(d\), is the result used to ensure the strict preference of a portfolio a over a portfolio b,
Constraints 3 are normalisation constraint, \(g_i^r\) and \(g_i^s\) are respectively minimum and maximum value of criterion \(g_i\),
Constraints 4 are the classical non-negativity constraint of linear programming,
The marginal utility functions are piecewise linear approximation of the real marginality utility.