An empirical investigation <u>of-into</u> the arbitrage pricing <u>theory theory</u> o<u>f</u>n China A share market after financial crisis

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## Abstract

The purpose of tThis paper is to examines the applicability of Arbitrage Pricing Theory APT in-to China's A Share Market after the financial crisis; using monthly data of 100 stocks from Shanghai Stock Exchange A <u>Sshare</u> as proxy variables to the whole market. Above all, <u>itthis paper</u> obtains four priced common factors from twenty-two22 macroeconomic variables by Exploratory Factor Analysis and builds relevant APT model for further empirical test. Then through multiple linear regressions, we test the explanatory power of APT in China's A Share Market. According to the<u>Using</u> empirical test results and the problem of poor goodness of fitting, we <u>offer probable give possible</u>-reasons (for what?) and <u>make relevantsome</u> suggestions.

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#### [ .Introduction

Arbitrage Pricing Theory (APT)<sub>a</sub> formulated by Ross (1976)<sub>a</sub> is proposed to be an alternative to the mean variance Capital Asset Pricing Model (CAPM) which was introduced by Sharp (1969) and<sub>7</sub> Lintner (1965). In the analysis of APT, Ross-(1976) suggest<u>s</u>ed that the market portfolio d<u>oes\_idn</u><sup>2</sup>ot play a specific role any-more. It prove<u>s</u>d that <u>the</u> expected return to assets was systematically influenced by multiple exogenous macroeconomic factors. Depending on the assumption of a well-diversified, perfectly competitive and frictionless economy, APT specifie<u>s</u>d the linear relationship between expected return to assets and a series of macroeconomic factors.

The paper of Chen, Roll and Ross (1986) selected the systematic forces that haved significant effect on the two determinants of stock returns: discount factors and expected cash flows. Through a vector autoregressive model, shock to real industrial production, changes in the risk premium, twist in the yield curve, unexpected inflation and change in expected inflation were explored to have systematic influences on stock market returns. In addition, by using a multi-factor model repeated in every subsample of each year, Chen, Roll and Ross obtained for each macroeconomic variable a time series of estimates of associated risk premium, which indicatesd the explanatory power to the multi-factor model. This kind of analytical method has been adopted till now for examining the validity of APT in the stock market.

During the recent decades, APT is considered to be the most significant analytic tool for to explaining the pricing behaviour in the assets market, mainly in some relatively developed markets which more than satisfy the assumptions of well-diversified, perfectly competitive and frictionless economy. In order tTo provide a better fitting tool to explain the market phenomenon, more mature commercial factor models like BIRR Model<sup>1</sup>, RAM Model<sup>2</sup> are derived. While the emerging markets in developing countries under imperfect system are not mature and unhealthy, assumptions of APT cannot be satisfied completely. Pproper models for emerging markets are still being explored. China's stock market, the most typical emerging market. Since foundationestablishment, A Share Market has always been the main driving force of China's stock market and guides the overall direction of stock-the market, in both aspects of capitalization, and outstanding shares, which can be (shown in Figure 1 and Figure 2).

<sup>&</sup>lt;sup>1</sup> Burmeister, Roll and Ross (2003) introduced BIRR Model in the paper "Using Macroeconomic Factors to Control Portfolio Risk", <<u>http://www.ftse.com/Analytics/BIRR</u>>

<sup>&</sup>lt;sup>2</sup> RAM Model (Risk Attribution Model) was introduced by Salomon Brothers in 1986



Source: Wind



Figure 2: Total outstanding shares and outstanding shares in A Share market.

## Source: Wind

A Share Market has always been developing <u>duringin</u> the unrest <u>period\_and faces</u> strict <u>regulatory</u> controls. Not <u>uUntil 1996</u>, because of increasingly<u>the</u> opened capital market, the IPO system and reverse merger, were not open <u>did\_and</u> when it <u>did</u> the number of listings <u>by</u> private enterprises increased rapidly. The year of 2007 wais considered as the most significant periodplendid time in the history of China's market development\_as also, the same \_\_to the A Share Market. <u>Reforms in</u> Nnon-tradable shares reform and reform in the Fund Industry pushed the A Share Market to <u>be a</u> bull <u>runmarket</u>. Both of the Shanghai A Share Index and the Shenzhen A Share Index reached their peaks, respectively, to <u>be</u> 6429.68<sup>3</sup> and 1629.7546<sup>4</sup>, increasing by over twice the size asof 2006. As shown in Figure 3, the closing point in both the exchanges got toreached an unprecedented peaklevel.

Figure 3: A share closing point in Shanghai Stock Exchange and Shenzhen Stock Exchanges.

#### Comment [.3]: What's unrest?

 <sup>&</sup>lt;sup>3</sup> Statistical yearbook 2012 of Shanghai Stock Exchange
 <sup>4</sup> Statistical yearbook 2008 of Shenzhen Stock Exchange

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# Source: Wind

In 2008, the global financial crisis spilled over toenveloped China\_too5 leading to a deep correction in China's equity market and a sharp contraction in terms of market value. China's equity market shrank dramatically, which was embodied in the depressedion of the \_-A share market, mainly because of the Pprice Comparison effect between the A-shares and H shares. The year-of 2008 is also saw no doubt to be another turning point into China's in history. After the crisis, the on the one hand, A share market remained grim-. The A Share Index swung back and forth around 2000 points during the recent five years and event reached 1806.79–<sup>5</sup>\_points in 2008. To strengthenreinforce the A share market, RQFII mechanism<sup>6</sup>\_-came into force in August-of 2011, with the pumping of starting from RMB\_20 billion RMB. Even though the effect of RQFII has been hanging ion firedoubt so far, it is still-be the focus of investors. On the other hand, Tthe prices of shares on A Share Mmarket are has serious bubble displayed in the overestimated prices of A shares without reference to and-price\_-earnings ratio. Investors firmly believe that there exists a great arbitrage opportunity chance in the A Share market.

The financial crisis forced the stock market to change, no matter <u>whether</u> in terms of structure or system. The China's A share market <u>has</u> becomes further open and mature. Economists <u>wishinsist onto</u> examinging the validity of APT in China's market while the <u>outcomeresult</u> is controversial. Opponents-<u>believeconsider that</u> the-great changes in China's stock market <u>have not didn't-improved</u> the <u>situation at</u> all. <u>fitting degree of APT apparently</u>. Nevertheless, supporters <u>feel considerthat</u> investors can <u>use the opportunity arising out of the crisis in the combine the situation</u> of capital market and <u>their own people's</u>-preferences, and useing APT in order to make rational choices-in the crisis. From the perspective of the current market, macroeconomic factors like inflation, real-estate market and consumer price index play a dominant part.

The <u>objective</u> of this paper is to examine the performance of APT on China's stock market after the financial crisis. The paper is organized as follows. Section II reviews the literature of APT. In Section III, we briefly introduce the revised model

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<sup>&</sup>lt;sup>5</sup> Hexun A share index

<sup>&</sup>lt;sup>6</sup> RQFII is the RMB qualified foreign institutional investor scheme. Launched in August 2011 it allowed a small number of Chinese financial firms to establish RMB-denominated funds in Hong Kong for investment in the mainland. The aim being to allow overseas investors to use offshore RMB deposits to invest in mainland securities markets.

based on China's physical situation and methodology in Exploratory Factor Analysis, to select economic state variables which have systematic impact on <u>returns from</u> stock<u>s</u>-return. Section IV introduces the data. Section V describes the economic characteristics of the macroeconomic variables selected and conducts a cross-sectional regression <u>analysis among between</u> the factors and stock return to examine the validity of APT in China. Section VI-provides the \_\_concludes the worksion.

### II .Literature Review

Arbitrage means taking advantage of the price variance and trading <u>amongbetween</u> different markets to <u>getobtain</u> riskless profit. The <u>aA</u>rbitrage<u>behavior</u> is realized by buying high-return securities and selling low-return <u>onesecurities</u> at the same time. As a consequence, the price of high-return securities will be pushed up while the rate of return will fall down and the low-return securities will be pulled down while the rate of return will <u>gorise</u> up. This process finally <u>brings equilibrium makes\_to</u> the security market<u>reached its equilibrium</u>. The <u>Arbitrage</u> Pricing <u>Theory</u> <u>showsindicates</u>, under unbalanced state, that there exists no arbitrage chance in the market. Moreover, the return to risky assets can be explained by multiple factors.

### (1) Theoretical development of APT

Though the Arbitrage Pricing Theory<u>APT</u> has become one of the mainstream models in the study of assets pricing issue, doubts are always heard, especially when compared with the Capital Assets Pricing Model. Roll and Ross (1980), Chen (1983), Chen, Roll and Ross (1986), Fama and French (1992), Groeneworld and Fraser (1997) are the most famous APT supporters. For one thing, the \_\_APT requires less but more realistic assumptions: a. multiple investment terms; b. taxes exist; c. investors cannot borrow money at the riskless rate. For another thing, APT shows better explanatory power than CAPM as it is a multi-factor model. Researchers like Dhrymes, Friends and Gultekin (1984) questioned the testability both of APT and CAPM.

As mentioned in Roll and Ross (1980), "APT has clear advantages over CAPM. APT is based on a linear return generating process and requires no utility assumptions beyond monotonicity and concavity." In addition, APT allows multiple factors and the equilibrium is characterized by-the linear relationship between return to assets and the priced common factors. APT's empirical test indicates its performance by cross-sectional regressions.

the empirical evidence, the APT cannot be <u>disregarded\_rejected\_in</u> favo<u>u</u>r of any <u>otheralternative</u> hypothesis and the APT performs better over CAPM <u>in regard to as</u> implemented by three market indices: S&P Index, Value Weighted Stock Index and Equally Weighted Stock Index. The APT was proved to beis a better-<u>\_</u>fitting model <u>into</u> explaining the pricing behavio<u>u</u>r in stock market.

Following Chen's methodology, Cagnetti (2002) tested and compared CAPM and APT in the Italian stock market with pre-set factors. All of the Davidson and Mackinnon Equation, the Posterior Odds Ratio and Residual Analysis strongly supported the performance of APT. In addition, through factor analysis, Cagnetti <u>obtainedgot</u> eight common factors: market portfolio factor, fixed income securities, market index, foreign variables, production and monetary factor, inflation factor, industrial factor, people's expectation factor, which follow closely the logic of economic activity.

McElroy and Burmeister (1986) and Burmeister and McElroy (1988) recast the APT as a multivariate nonlinear regression model with across-equation restriction. They replaced the unknown random factors with observed macroeconomic variables. The nonlinear time--series methodology made the parametric APT testable rather than assuming thate priceing restrictions hold. Antoniou, Garrett and Priestley (1998) followed this method to investigate the performance of APT on London Stock Exchange. The paper randomly divided the securities into two subsamples to test whether-exist the unique return-generating process across two subsamples exists as well as risk prices in different subsamples. Finally, Antoniou, Garrett and Priestley (1998) got the concluded sion that to get develop a unique return-generating process, the model should be robust across different samples. Moreover, tThey got five common factors to price securities, among which are: unexpected inflation, money supply and excess returns on market portfolio carrying the same prices of risk in different samples, and the other two factors are: default risk and exchange rate only offerred as marginal improvement in the performance of APT. Azeez and Yonezawa (2006) adopted the same way method to examine the validity of APT in the Japanese stock market during the bubble period. They also and compared the robustness of priced factors over-during the bubble period with pre- and post-bubble period datas. Four different types of risk factors-money supply, inflation, exchange rate and industrial production-were identified with having that had significant influence on the returns to stocks turned out to be money supply, inflation, exchange rate and industrial production.

Fama and French (1992) compared the CAPM and APT from several aspects. Besides showing the weakness that had been found in Chen (1983) and Chen, Roll and Ross (1986), this paper also implied that CAPM overestimated the risk-free rate and underestimated the market risk premium. Moreover, the returns to high-beta assets will be overestimated and the returns to low-beta assets will be underestimated to low-beta assets. In the aspect of empirical test, Fama and French had different

ideas from traditional two-step method, which might have the errors-in-variables problem. Using the monthly data for of 19 sectors from the Australian Stock Exchange, Fama and French consider thated the beta of market cannot explain the variance among different returns to on stocks, while this can be interpreted by three classes of factors. Any changes in the real domestic activity, nominal domestic influences and foreign variables are supposed to change investors' expectations of future cash flows and afterwards influence current stock prices. The test results indicated that inflation rate wais more an empirical support than other macroeconomic factors.

In recent decades, with the improvement of in assets pricing theory, several commercial assets pricing models have been derived from the APT multi-factor model; for instance, BIRR Model, RAM (Risk Attribution –Model). BIRR Model was raised developed by Burmeister, Ibbotson, Roll and Ross (2003), who introduced five macroeconomic factors: confidence risk, time horizon risk, inflation risk, business cycle risk, and market timing risk. All the factors were divided into leading indicator, coincident indicator and lagging indicator and determined the order to build a better fitting model. As a consequence, four factors that indicated greater influence on return to  $stock_7$  respectivelyare: to be macro-economic climate index, the CRB metal stock index, difference between the speed of deposit and loan, and one-year treasury yields. Among all the above, CRB metal stock index representing the commercial cycle had greater sensibility. RAM (Risk Attribution Model) considered adopted six factors to explain the pricing behaviour in the stock market: long-term bond yields, short-term treasury yields, inflation and exchange rate.

#### (2) The study results about the vV alidity of APT in China's stock market

For-In the period of unrestfurmoil and rapid development of China's stock market, plentiful-several\_economists have\_showned great interest in the pricing issue of China's stock market, no matter before or after the financial crisis in 2008. As one of the most crucial assumptions of APT is a well-diversified, perfectly competitive and frictionless economy, most of the earlier\_studies in-using it <u>APT before were to</u> examined\_the\_its performance of <u>APT</u>-in developed stock markets. However, the rapid rise in emerging markets or transition economy is obvious to all, especially in China, and hence- Chinese economists devoted themselves to examineplain the market behaviour in China with a revised model.

Li (2002) revised the multi-factor model in his paper *Experimental Test about Shenzhen Stock Market by APT* according to changes in market system. Price Limit System and T-+-1 Trading Rules were added into the regression model as two dummy variables. In the first step, by factor analysis, Li got\_listed\_four macroeconomic factors which which reflected significantly impact that-various industries listed have-on the Shenzhen stock exchangemarket, as well-asalso the

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impact of earning performance and financial <u>conditions</u> <u>status</u> <u>onf</u> their own companies. In the second step, Li-respectively</u> examined the <u>correctness accuracy</u> of APT with the non-modified and modified models<u>in Shenzhen market</u>. The <u>study test</u> results showsed that both of the dummy variables had significant effect on returns to from stock, which means that unsystematic risk from policies had great effect on investors' behaviour and market direction. The physical significance of Li's suggests paper is that the most key point to promote the more radical marketing approach to improve China's stock market<u></u> is to speed up the marketization. A similar method will be adopted in this paper, too but by adding a different dummy variable, RQFII System, will be added into the multi-factor model.

Liu and Qin (2004) adopted the test methodology <u>ofin</u> Brown and Weinstein\_(1984), respectively selecting two sample groups in two periods and examining <u>if\_whether</u> the generation process was <u>the</u> same among different subsample groups <u>or as well as</u> whether different sample groups faced the same factor structure. Before the test of APT<sub>7</sub> by Maximum Likelihood Method the author<u>s</u> proved the<u>re</u> existenced <u>of</u> at least nine factors affecting the China's market and examined the <u>relevancevalidity</u> of the <u>nine factor</u>-model <u>to\_in</u>-China-market. The test results <u>suggested implied</u> that, at that time, the APT model might not explain the economic behaviour in China's market. The pricing <u>issue</u> of China's stock market was influenced <u>too</u> much by policies and it was hard to <u>considerform</u> the real\_value investment philosophy. However, this phenomenon coincided with China's emerging market or transition economy.

Tian (2006) followed the methodology of BIRR <u>Mm</u>odel to select macro risk factors in China's stock market. The procedure <u>wais</u> as follows: firstly, divide all the alternative factors into leading, consistent and lagging factors through Variance Inflation Factor method and robust regression; secondly, build the best model with selected factors; last, to avoid the heteroscedasticity problems, use the GARCH <u>Mm</u>odel to <u>obtainget</u> risk sensitivity matrix. In this paper, factors including total energy production, 7-day interbank interest rates, M2, CPI, long-term loan, FDI, outstanding of deposits, were selected as macro risk factors.

After the financial crisis, China's stock market <u>underwent several experienced great</u> changes. Though <u>badly affected by suffering a lot from</u> the crisis, <u>the</u> Chinese market was opened by a relatively passively and <u>slowly regained its became more</u> normal tive<u>status</u>. Hu and Chen (2009) considered th<u>e</u> at one assumption of APT was that risk-averse individual investor always preferred increasing income without <u>exposing</u> to higher risk, which could be achieved by arbitrage. This assumption wais in accord<u>ance</u> with the investors' psychology during the depression. Hence, APT model was <u>well received infurther appreciated for</u> China's stock market after the Ffinancial <u>Cc</u>risis. Yin (2008) did a similar study on the validity of APT in China's stock market. The majority of preveious studies <u>examined of</u> the validity of APT in China and <u>suggested indicated</u> that policies had greater influence <u>on the market</u> than the market force and too overmuch regulation supervision might increase the risk.

III. The Model and Methodology

(1) (1) The APT model starts with the <u>following assumptions</u>: Asset markets are perfectly competitive and frictionless. No arbitrage chance <u>exists</u> in the market. Individuals have homogeneous beliefs that assets returns are generated by a linear k-factor model;

The k factors that have effect on assets returns are stochastic and unexpected. Here we assume the mean of every factor <u>to be</u> zero;

All the investors <u>arehave the</u> risk\_-averse and monotone increasing utility function.

Under <u>the</u> above assumptions, Ross in 1976 generalized a multi-factor  $F_1$ ,  $F_2$ ,  $F_3$ ..... $F_k$  systematically influencing the returns <u>from</u> on-assets

$$\mathbf{R}_{i} = \mathbf{E}(\mathbf{R}_{i}) + \mathbf{b}_{i1}\mathbf{F}_{1} + \mathbf{b}_{i2}\mathbf{F}_{2} + \dots + \mathbf{b}_{ik}\mathbf{F}_{k} + \mathbf{\varepsilon}_{i}, i=1, 2, 3..., m, m > k$$
(1)

We where  $R_i$  is random rate of return to asset I;  $F_i$  (j=1, 2...k) -is denoted as the

systematic common factor,  $E(F_iF_i)=0, i\neq j$ , impliesying that factors are independent of

between each other;  $b_{ik}$  is the unique coefficient of sensitivity that asset *i* has on factor j, measuring the systematic risk-;  $\varepsilon_i$  is considered as a disturbance term, which measures the idiosyncratic and unsystematic risk, including all the unrelated information about other assets.  $E(R_i)$ -refers to the expected return fromto stock *i* when other risks equal to zero. We assume  $E(\varepsilon_i) = 0$  and  $E(\varepsilon_i \varepsilon_i) = 0$ ,  $i \neq j$ .

Associated with no risk arbitrage and sufficient portfolio diversification, the equilibrium relationship is as follows:

$$\mathbf{E}(\mathbf{R}_{i}) = \mathbf{\theta}_{0} + \mathbf{\theta}_{1}\mathbf{b}_{i1} + \mathbf{\theta}_{2}\mathbf{b}_{i2} + \dots + \mathbf{\theta}_{k}\mathbf{b}_{ik}$$
(2)

where  $\theta_0$  is denoted as the expected return to riskless asset;  $\theta_i$  can be interpreted as

risk premiums corresponding to risk factor  $F_i(j=1,2...k)$ .

In consideration of the RQFII, we intend to test whether the fluctuations in policy have significant influence on our regression, so we introduce this mechanism into APT model as dummy variable. The revised model is shown as follows:

$$\mathbf{R}_{\mathbf{i}} = \mathbf{E}(\mathbf{R}_{\mathbf{i}}) + \mathbf{b}_{\mathbf{i}1}\mathbf{F}_{\mathbf{1}} + \mathbf{b}_{\mathbf{i}2}\mathbf{F}_{\mathbf{2}} + \dots + \mathbf{b}_{\mathbf{i}k}\mathbf{F}_{\mathbf{k}} + \mathbf{\varepsilon}_{\mathbf{i}} + \mathbf{c}\mathbf{D}$$
(3)

where  $D_i$  is dummy variable,  $D_{-} = -0$ , before the RQFII

t, after the RQFII c is the sensitivity of price to the dummy variable.

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Formatted: Font: (Default) Times New Roman, 12 pt, Complex Script Font: Times New Roman, 12 pt Therefore, the model can be <u>converted</u>translated to:

$$R_{i} = \begin{bmatrix} E(R_{i}) + c] + b_{i1}F_{1} + b_{i2}F_{2} + \dots + b_{ik}F_{k} + \varepsilon_{i} + cD, D=1 \\ \\ E(R_{i}) + b_{i1}F_{1} + b_{i2}F_{2} + \dots + b_{ik}F_{k} + \varepsilon_{i}, \qquad D=0 \end{bmatrix}$$

(2) Exploratory Factor analysis

However, the APT model itself doesn't have the access to-macroeconomic factors. In general, there exist two methods to determine the number of factors. One way is to run regressions with artificially stated factors on the rate of return to stock and to determine the final common factors, based on test of goodness of fit. Yasushi (1988) adopted parallel data used in Chen, Roll and Ross\_(1986) and mixed with international trade factors to discuss the different significances of different pricing effects of similar macroeconomic factors in a parallel but with a distinguishing economy. According to recent studies, to support the APT model, the artificially factors should be properly integrated. combined. Nevertheless, an agreement on how to correctly combine these factors which can persistently and completely explain the cross-sectional variation has not yet been arrived at. cannot be reached an agreement. In addition, artificially stated factors may lead to information overlap of or the number of variables to solve the above problem may lead to information loss and lack of explanatory power to model.

Another methodology is to pre-set several statistic variables as alternate common factors and use Principal Component Analysis or Maximum Likelihood Method to determine common factors. This procedure is called <u>the</u> Exploratory Factor Analysis (EFA), which is a common methodology of dimensionality reduction, making use of the correlation between original variables to best combine and simplify the multi-variable data. Interpreting most <u>of the</u> information of <u>the</u> original data can reduce difficulty in examining the validity of APT. The common factors after sifting mainly avoid linear correlation. The second method seems to provide more objective analysis and better fitness.

The starting point of EFA is to replace the majority information of original variables with fewer mutually independent factors. Assuming that there exist p variables,  $x_i(i = 1, 2, ..., p)$  are standardized variables;  $F_j(j = 1, 2, ..., m)$  are denoted as the systematic common factors, then,

$$x_1 = a_{11}F_1 + a_{12}F_2 + \dots + a_{1m}F_m + a_1\varepsilon_1 x_2 = a_{21}F_1 + a_{22}F_2 + \dots + a_{2m}F_m + a_2\varepsilon_2$$

. . . . . .

 $\mathbf{x}_{p-1} = a_{(p-1)1}F_1 + a_{(p-1)2}F_2 + \dots + a_{(p-1)m}F_m + a_{p-1}\varepsilon_{p-1}$  $\mathbf{x}_p = a_{p1}F_1 + a_{p2}F_2 + \dots + a_{pm}F_m + a_p\varepsilon_p$ 

This Also can also be expressed by the (m, p) matrix:

$$\mathbf{X} = \mathbf{A}\mathbf{F} + \mathbf{a}\mathbf{\varepsilon}$$

where A is denoted as factor loading matrix, and;  $a_{ij}$  is called factor loading, representing correlation coefficient between original variable *i* and common factor *j*. The bigger the absolute value of  $a_{ij}$  is, the more significant variable *x* is to factor *j*; The sum of squares of  $a_{ij}$  in line *i* in matrix A is defined as the C communalities between original variables; reflecting the contribution of common factors to original variables:

$$\mathbf{h}_i^2 = \sum_{j=1}^m a_{ij}^2$$

Below the premise of standardized variables, the variance of  $x_i$  can be expressed as

$$\mathbf{h}_{i}^{2}+\mathbf{\varepsilon}_{i}^{2}=\mathbf{1}_{:}$$

The sum of squares of  $a_{ij}$  in column *j* is  $\forall \underline{v}$  ariance contribution:

$$S_j = \sum_{i=1}^{P} a_{ij}^2$$

The higher  $S_j$  is, the more significant the common factor *j* is;  $\varepsilon$  is regarded as the part <u>that</u> cannot be explained by common factors, equivalent to the residual term.

Common factors prove to have <u>the</u> following characteristics: a. the number of factors m is less than the number of original variables p; b. common factors are not simple reduction of original variables, but effective recombination; c. there doesn't exist correlation between different factors, which means one variable cannot be explained by multiple common factors; d. common factors have named explanatory according to the combined information.

### (3) Procedure of EFA and cross-sectional regression

Basic steps of factor analysis focus on two central issues: how to construct, name and explain the common factors. <u>Here, This paperwe</u> will do <u>the factor analysis as</u> <u>perecording to</u> the following steps:

a. Judge the fitness of all the original variables to do EFA

The most crucial assumption of EFA is that there exist strong correlations between original variables. Therefore, correlation analysis will be carried <u>out</u> first<del>ly</del>. The simplest method is to conduct statistical test with the correlation coefficients matrix. If the majority of correlation coefficients is less than 0.3 and cannot pass the statistical test, factor analysis is inapplicable.

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In this paper, two methodologies will be adopted. The first one is Kaiser-Meyer-Olkin test (KMO). KMO statistics is an index used ftor compareing the correlation coefficients and partial correlation coefficients.

$$\mathrm{KMO} = \frac{\sum \sum_{i \neq j} r_{ij}^2}{\sum \sum_{i \neq j} r_{ij}^2 + \sum \sum_{i \neq j} p_{ij}^2}$$

 $r_{ij}$  is denoted as the simple correlation coefficient between variables *i* and *j*-, and *p*<sub>ij</sub>

is the partial correlation coefficients.and KMO statistics is an index between 0 and 1. The closer the value is to 1, the more appropriate it is <u>forto do</u> analysis. Kaiser even<del>r</del> gave the measurement level: 0.9 means pretty appropriate; 0.7—to 0.8 means-suitable; results under 0.6 are not recommended <u>forto do</u> factor analysis.

The second method is Bartlett Test of Sphericity, which starts from correlation coefficients matrix. The null hypothesis: correlation coefficients matrix is supposed to be an identity matrix. If the <u>datastatistic</u> is relatively large and p-value is less than significance level, then we reject the null hypothesis. This means the correlation coefficients matrix proves not to be an identity matrix which is appropriate for factor analysis; <u>otherwise</u>, i<u>ft is not, it is not suitable</u> recommended <u>for</u>to do factor analysis.

b. Construct the common factors and calculate the factor-loading matrix

After making sure the applicability of EFA, we have several traditional methodologies for constructing the common factors: Principal Component Analysis, Principal Axis Factoring, Maximum Likelihood Method and Least Square Method.

The key difference between <u>allthe</u> above methods is <u>the as</u>-following: Principal Component Analysis, just through rotating the coordinates, recombines the ordinary variables into a new linear combination, also called Principal Component; while Principal Axis Factoring, Maximum Likelihood Method and Least Square Method base on the Factor Model in which combination between potential variables and random effects variables represents <u>the</u> initial variables. This paper employs the most common method<del>;</del> Principal Component Analysis. All the variables in <u>the</u> new combination are uncorrelated, <u>and are</u> expressed as follows:

$$y_1 = u_{11}x_1 + u_{21}x_2 + \dots + u_{p1}x_p$$
  

$$y_2 = u_{12}x_1 + u_{22}x_2 + \dots + u_{p2}x_p$$
  

$$y_p = u_{1p}x_1 + u_{2p}x_2 + \dots + u_{pp}x_p$$

.....

We where  $y_i$  is uncorrelated with  $y_j$ ; the variances of  $y_1, y_2, \dots, y_p$  as well as the ability to combine variables are decreaseing.

The coefficients u are required to be

$$u_{1k}^2 + u_{2k}^2 + \dots + u_{pk}^2 = 1 \ (k = 1, 2, \dots p)$$
  
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Therefore,  $-_y_1$ ,  $y_2$ ,... $y_p$  are referred to as the first, the second...<u>and</u> the last principal component of <u>the</u> original variables. In general, only several components with highest variance will be selected as principal component.

During the Principal Component Analysis, we obtain p eigenvalues and related eigenvectors used for calculating the factor—loading matrix A. Based on the main purpose of Principal Component Analysis to decrease the number of variables, we usually select loading matrix of m common factors, m < p. Eigenvalues  $S_i$  (i = 1,2, ... p) of common factors are required to be larger than 1. What's more, the total

cumulative variance contribution rate of these m factors  $(S_1+S_2+...+S_m) / \sum S_i$ 

should be larger than 80%.  $S_i(i = 1, 2, ..., p) / \sum S_i$  is denoted as the percentage variance contribution of every factors.

#### c. Increase the interpretability of factors by rotation

The interpretation of common factors is another core issue of factor analysis. As  $e\underline{E}$  very common factor combines information from multiple variables<sub>1</sub>; in other words, one common factor <u>only enables to</u> explains small amount of information from every variable, a common factors is not a typically representative <u>offer</u> any single variable. The situation above leads to cryptic interpretation of new factors. The ideal access to better interpreting the common factors is rotation. In this paper, we use Varimax Rotation Method to realisze a clearer factor interpretation, implying that the loading of a certain variable on one factor closes to 1 and loadings on other factors close to 0. One common factor turns to be a typical representative <u>offer</u> one certain variable. But it is regrettable that the interpretations of factors <u>isare</u> quite subjective. We come across various interpretations <u>occur</u> in various combinations. So far, there is no clearity criteriona for this.

d. Calculate the factor scores.

The final step is to calculate the factor scores according to coefficients of the score function by Regression Method.

 $F_{j} = \beta_{j1}x_{1} + \beta_{j2}x_{2} + \dots + \beta_{jp}x_{p}$  (j=1,2,....m)

So far, <u>we have examined the work of factor analysis is done</u>. In the empirical part, we obtain the APT model by two-step multiple linear regressions. In the first part, the factor scores are used as independent variables and returns tofrom sample stocks in A Share market are used as dependent variables. In the second part, constant terms in the first regression representing the expected returns are used as dependent variables and coefficients are used as independent variables, getting the final APT model.

IV.\_Data

The data used are monthly closing prices of a randomly selected sample of 100

stocks in the Shanghai Stock Exchange A Share Index as proxy variables for China's A Share market from January 2008 to December 2012, provided by Wind Information. To avoid-the stock price distortion resulting from exclusion ofding dividend and rights, all price parameters will data-adopts backward adjusted price without missing data. For the monthly return to stocks cannot be obtained directly, we use the calculateing method as follows:

 $\frac{\text{Closing price in month } T - \text{Closing price in month } (T - 1)}{\text{Closing price in month } (T - 1)}$ 

In terms of choosing the appropriate group of economic factors, <u>Since</u> there is no formal theoretical guidance to choose the appropriate group of economic factors, we turn to- Chen, Roll and Ross (1983) who gave a group of macroeconomic factors including *industrial production index, unexpected inflation, changes in default risk, unexpected change in term structure*.; bBaseding on the conclusion of Chen, Roll and Ross, Antoniou, Garrett and Priestleyet al. (1998) added *real retail sales, commodity prices, exchange rate* also; Cagnetti (2002) obtained 8-eight factors by conducting factor analysis on thirty three33 variables, <u>risk premium, foreign variables, production and monetary, inflation, industrial production and people's expectations;</u> Yasushi (1988) added international variables, especially oil price to examine the impact of international factors on the returns from to stocks.

We have arrived at the following macroeconomic factors after Having learnt examining several previous works: others' study results, the macroeconomic factors we use are Currency in Circulation ( $M_0$ ), Narrow money supply ( $M_1$ ), Broad money supply ( $M_2$ ), Consumer Price Index (CPI), Producer Price Index (PPI), Foreign Direct Investment (FDI), Net Trade\_(NT), Gold Reserve\_(GR), Foreign Currency Reserve\_(FCR), NYMEX Oil Future Price\_(OP), Exchange Rate by indirect quotation\_(ER), 3-month Treasury Security Yield\_(SB), 10-year Government Bond Yield\_(LB), Inflation Rate\_(IFR), Benchmark Interest Rate\_(BIR), Ratio of Exports to Imports for China\_(REI), Macro-economic Climate Index\_(MECI), Import Price Index\_(IPI), Consumer Confidence Index\_(CCI), Business Confidence Index\_\_(BCI), Financial Institution Short-term Loan\_(FSL), Financial Institution Long-term Loan (FLL). Unfortunately, monthly data of GDP and Industrial Production awere not readily available.found. The date of macroeconomic variables is are selected based on monthly-level data during of the same period with data s of stock price and there exist no missing data is missing.

Comment [.6]: Concept not clear.

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## Table 1 : Correlation Matrix

Note: Circulation ( $M_0$ ), Narrow money supply ( $M_1$ ), Broad money supply ( $M_2$ ), Consumer Price Index (CPI), Producer Price Index (PPI), Foreign Direct Investment (FDI), Net Trade\_(NT), Gold Reserve\_(GR), Foreign Currency Reserve\_(FCR), NYMEX Oil Future Price(OP), Exchange Rate by indirect quotation(ER), 3-month Treasury Security Yield(SB), 10-year Government Bond Yield\_(LB), Inflation Rate\_(IFR), Benchmark Interest Rate\_(BIR), Ratio of Exports to Imports for China\_(REI), Macro-economic Climate Index\_(MECI), Import Price Index\_(IPI), Consumer Confidence Index\_(CCI), Business Confidence Index\_(BCI), Financial Institution Short-term Loan\_(FSL), Financial Institution Long-term Loan\_(FLL). Table 1 shows the correlation matrix of all the macroeconomic variables over 60 months from 2008 to 2012. Just taking several variables as <u>an</u> examples, the high correlations between (<u>i</u>4) money supply and CPI, (<u>2ii</u>) inflation and PPI, (<u>3iii</u>) net trade and ratio of exports to imports, and (<u>iv</u>4) foreign currency reserve and exchange rate show the information of related <u>to</u> variables can be explained by certain common factors. In addition, the low correlation between oil price and money supply implies the existence of no less than two common factors; and <u>single</u> variable alone <u>will is</u> not <u>be sufficient</u> enough to explain the macro economy.

V.Testing Results

(1) Exploratory Factor Analysis

The statistic of Kaiser-Meyer-Olkin test <u>yields a factor of is-</u>0.801; Bartlett Test of Sphericity gives Chi-square approximation <u>of 2835.029</u> with degree of freedom <u>of</u> 231; the p-value is 0.000, less than the significance level 0.05. All results above prove that original variables are suitable <u>for to do</u>-Exploratory Factor Analysis.

The following figure is the Screen Plot between common factors and eigenvalues, with horizontal axis representing the number of common factors and vertical axis representing the eigenvalues of the certain factor. According to the Screen Plot,  $t_{\rm T}$  he eigenvalues of first four factors show larger variations while from the fifth <u>one factor</u> eigenvalues start to<u>will</u> level off. It implies that the first four factors have significant effects on explaining original variables. To increase the accuracy in the test of APT, we <u>limit the study to decide to keep</u>-four common factors.



As shown in the Communalities Table 2, the initial communalities of all the variables in column 2 are 1.000; the extraction communalities of 80% variables are more than 0.900, close to 1, which means the information can be explained by common factors isby more than 90%, only with a smalla little information loss that Taking Currency in Circulation ( $M_0$ ) as an example, 0.918 means 91.8% of variance can be illustrated by

#### Comment [.7]:

common factors.

#### Table 2: Communalities

Method: Principal Component Analysis

(3) Table 3 ishows the Total Variance of explained of extraction factors as well as that after rotation. The second column in the table describes the variance contributions of all factors to initial variables in a decreasing order. In tThe third column, we get shows the percentage of variables variance of variables explained by factors, also in the decreasing order as well. The fourth column shows the accumulated percentages of variance explained by factors. The fifth to seventh columns specially show the description of the four extracted factors. The variance contributions of factors are respectively, to be 9.107, 5.898, 2.949, and 1.284, and: Tthe percentage of variance explained for extracted factors is are separately to be 41.396%, 26.808%, 13.404%, 5.834%. For example, 41.396% equals to 20 divided by 9.107. The accumulated percentages of first four extracted factors are 41.396%, 68.205%, 81.609%, 87.443%. The information in table proves the sresults shown on ame conclusion as Screen Plot, recommending four common factors. The eighth to tenth columns show the results after rotation. The eigenvalues of four factors now are: change to be 8.911, 5.538, 2.658, 2.130, totally explaining 87.443% of the total variance of initial variables.

(3)\_Table 4 <u>shows contains</u>-all the factor loadings before and after factor rotation, <u>showing implying</u> the relation between initial variables and the common factors. We <u>giveassign</u> names with better interpretability to <u>the</u> common factors depending on <u>the</u> following factor loading matrix so that we can explain the structure and main characteristics further. As shown in <u>the t</u>able 4, before rotation,  $M_0$ ,  $M_1$ ,  $M_2$ , CPI, FDI, GR, FCR, EX, FSL, FLL have larger loadings on the first factor; PPI, OP, IFL, LB, MECI, IPI,CCI have larger loadings on the second factor, <u>if and</u> NT, GR, BIR, REI, FCL have large loadings-on the third factor?<u>s</u>. FDI, NT, GR have similar loadings on different factors. Apparently, there exists <u>a common thread coincident information</u> <u>inamong all the</u> four factors, which <u>will goes against to explain</u> the common factor and provide and find proper economic direction.<u>implication</u>.

We now employ <u>After using Varimax Method,the</u> factor rotation <u>method which</u> makes factor loading matrix clearer and gives better interpreta<u>tion</u>bility than <u>earlier</u>before. Every variable can be explained by a single factor. The first factor mainly reflects the information concerning *Consumption and Investment Activity* 

*Factor*, including Currency in Circulation ( $M_0$ ), Narrow money supply ( $M_1$ ), Broad money supply ( $M_2$ ), Consumer Price Index (CPI), Foreign Direct Investment (FDI), Gold Reserve\_(GR), Foreign Currency Reserve\_(FCR), Exchange Rate by indirect quotation\_(EX), Financial Institution Short-term Loan\_(FSL), Financial Institution Long-term Loan\_(FLL); the second factor covers the information in Producer Price Index (PPI), NYMEX Oil Future Price\_(OP), Inflation Rate (IFL), 3-month Treasury Security Yield\_(SB), 10-year Government Bond Yield\_(LB), Macro-economic Climate Index\_(MECI), Import Price Index\_(IPI), and Consumer Confidence Index\_(CCI). The, we call the second factor is the Price Movement and Yield to Treasury Bonds Factor; the third factor includes Net Trade\_(NT), Ratio of Exports to Imports for China\_(REI), Business Confidence Index\_(BCI)\_s-is named as-International Trade Factor; then the fourth factor explains the left Benchmark Interest Rate\_(BIR), which can be defined as the Benchmark Interest Rate Factor. Every initial variable—has belongsed to a common factor.

Table 4: Factor Loadings

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Method: Orthogonal rotation method; Rotation converges after <u>five</u>5 iterations

Table 5 indicates the covariance matrix of common factors. Columns 2—to-5 express the covariance before rotation and column 6—to-9 show the covariance after rotation. According to the characteristics of factor analysis, factors are supposed to be orthogonal and unrelated to each other. Results in the covariance matrix after rotation prove this point, which is the appropriate to be the factor infor APT.

Table 5: Component Score Covariance Matrix

(4) Table 6 shows us the coefficients of the score function by Regression Method.

Table 6: Component Score Coefficient Matrix

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Extraction Method: Principal Component Analysis Rotation Method: Varimax with Kaiser Normalizaion. Component Scores

Therefore, we can write the score function based on the following: is.

$$\begin{split} \mathbf{F}_1 &= 0.112 M_0 + 0.115 M_1 + 0.113 M_2 \dots + 0.111 \mathrm{FSL} + 0.114 \mathrm{FLL} \\ \mathbf{F}_2 &= 0.005 M_0 + 0.031 M_1 - 0.006 M_2 \dots - 0.026 \mathrm{FSL} + 0.025 \mathrm{FLL} \\ \mathbf{F}_3 &= 0.040 M_0 + 0.050 M_1 + 0.026 M_2 \dots + 0.017 \mathrm{FSL} + 0.040 \mathrm{FLL} \\ \mathbf{F}_4 &= -0.011 M_0 - 0.108 M_1 - 0.022 M_2 \dots + 0.057 \mathrm{FSL} - 0.085 \mathrm{FLL} \end{split}$$

Finally, we can calculate all the factor scores (see Appendix A)<u>.</u> which will be shown in the Appendix.

## (2) Empirical Test of APT

Through the analysis above, we define four common factors: *Consumption and Investment Activity Factor, Price Movement and Yield to Treasury Bonds Factor, International Trade Factor, Benchmark Interest Rate Factor* and <u>obtainget</u> the factor scores. To build <u>the APT model</u>, we take the monthly returns <u>tof</u> the 100 sample stocks from 2008 to 2012 as explained variable and take the factor scores as explanatory variables for use in to do multiple linear regression. As a result, we obtain the constant terms as expected return <u>tofrom</u> each stock and coefficients <u>toof</u> four factors (see Appendix B). In all the results, about 35% <u>of</u> joint p-values are significant at the significance level of 5% and <u>there are an additionalfurther about 20% joint</u> p values are significant at the significance level of 10% <u>level</u>. Then we take constant terms as explained variables and take coefficients as explanatory variables <u>for do-a</u> <u>second</u> multiple linear regression. <u>again.</u>

As mentioned in Section\_III, we this paper-adds the RQFII as a dummy variable. Nevertheless, during all-the regressions, RQFII is not significant tfor all the sample stock returns, and even lowers the whole significance of or each regression. Hence, we will not perform the second regression with dummy variables. The test results suggests that the RQFII system hasn't played a significant role in improving China's A Share market so far.

Using the unrevised model, we get the regression equation as follows:

$$E(R_i) = -0.0069029 - 0.1520825\widehat{b_{i1}} - 0.2327547\widehat{b_{i2}} - 0.0027024\widehat{b_{i3}} + 0.012801\widehat{b_{i4}}$$

We-can notice the following points from the above regression equation: on the one hand, risky security *i* has negative sensitivity coefficients on the first three common
factors, implying that the larger the risk premiaums of *Consumption and Investment Activity Factor*, *Price Movement and Yield to Treasury Bonds Factor*, *International*

*Trade Factor*-are, the smaller <u>is</u> the expected return rate to the risky security i-is; on the other hand, risky security i has positive sensitivity coefficients on the last common factor, implying thate larger the risk premium of *Benchmark Interest Rate Factor* is, the larger is the expected return rate to security i-is.

The p-value to the total test is 0.0395, smaller than the significance level\_ $\alpha$ \_=\_0.05, which means that all the variables jointly have significant effect on expected return rates to stocks. However, the R-square which measures the goodness of fit is not satisfactory and there exists residual between the model and the real market. In this paper, we consider <u>that</u> this result may be caused by data deficiencies. Data of China's A Share Market has the problem of inadequate macroeconomic index and inconsistent statistical standard. Many macroeconomic factors only have annual data but not monthly data, such as *GDP*, *Industrial Production and Retail Sales*, which may cause the APT Model underspecified and further result in the low goodness of fit. Another reason is supposed to be that China's A Share market is not mature and capital market is not sufficiently opened. The macro-environment in China cannot satisfy the assumption of APT, especially the assumption of a well-diversified, perfectly competitive and frictionless economy. There may exist great arbitrage chances in China's A Share market, and the great bubble in stock market just proves this.

## VI. Conclusion

In this paper, we obtain four common factors with clear economic interpretations by Exploratory Factor Analysis, separately to be *Consumption and Investment Activity Factor*, *Price Movement and Yield to Treasury Bonds Factor*, *International Trade Factor and Benchmark Interest Rate Factor*. All these factors are combined and simplified from <u>22</u>twenty two initial variables, such as Money supply, Net Trade, Interest Rate and Inflation Rate, etc. This paper takes the factors of international trade, monetary policy, macro-economic climate index, commodity prices movements into account so that we can get significant macroeconomic factors as wide as possible and describe the current situation of the A Share market accurately. <u>Data Aa</u>nalysis-about statistics shows the extraction effects is very good. The only pity and shortcoming is that data like GDP, Industrial Production which have ever proved to perform significant roles in affecting the expected returns significantly ofto risky securities cannot be provided, which may have influenced on the result of factor analysis.

In the empirical part, we take 100 sample stocks from Shanghai Stock Exchange A Share Index as-the proxy variables to A Share market. Then this paper builds on the APT model based on four priced factors obtained by factor analysis and investigates the validity of the APT-model in the China's A Share market. According to the test results, we found the negative effect of *Consumption and Investment Activity Factor*, *Price Movement and Yield to Treasury Bonds Factor*, *International Trade Factor* have on the expected return tfrom risky securities and the positive effect of *Benchmark Interest Rate*. However, the APT model cannot be proved as a good fitting

model to current stock market, likely because of or two reasons: first, underspecified model because of data deficiencies; second, the incomplete and relatively closed and controlled market environment. Current China's stock market is in the transition stage, behaving as not sufficiently open and imperfect market system. Many measures aiming at improvinge the A Share market are just testing the water, just like the RQFII system which hasn't had too much positive effect, on the development of A Share market. Further, lagging of policy lagies means that it will take time for policies to work. Keeping\_Aiming atthe above problem\_in\_minds, this paper has several suggestions: for one thing, expand the data capacity for further study to increase the fitting degree of APT model; for another\_thing, promote\_the market to changes in macroeconomic environment.