Classifying Returns as Extreme: European Stock and Bond Markets

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Classifying Returns as Extreme: European Stock and Bond Markets*

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Abstract: I consider the stock and bond markets of 14 EU countries. I use two classification schemes for defining extreme returns: One, the existing univariate classification scheme which considers each market separately. Two, the new multivariate classification scheme that considers all the markets jointly whereby a shorter sample period is needed. For the bond markets the simultaneous extreme return variable (used for analyzing integration and contagion of financial markets) is not statistically different for the two schemes. For the stock markets there are differences, but they are disappearing in the most recent sample period.

Keywords: European stock markets; European bond markets; Extreme returns; Financial crisis; Integration of financial markets

JEL Classifications: G01; G12; G15
1 Introduction

The previous literature investigates when asset returns are extreme, that is when there are large negative returns on a given financial market. Also, the previous literature studies which explanatory variables explain the simultaneous occurrence of extreme returns across several financial markets. A long sample period is needed because only the 5% smallest returns are classified as being extreme. In this paper, I provide a new classification scheme for returns being extreme that requires a shorter sample period.

So far, returns are classified as being extreme by considering each asset market separately. I denote this for the univariate classification. That is, to classify whether a return on a particular stock market (say Germany) is extreme on a given day, only the time series of returns for that particular stock market (Germany) is investigated. Here, I advocate considering several similar asset markets simultaneously when classifying extreme returns. I denote this for the multivariate classification. For the German stock market example, I use all the stock markets of the old EU countries to classify German stock returns as extreme. In this way it is possible to apply data from a much shorter sample period compared to the univariate classification scheme.

to distinguish between local, regional, and global crashes where crashes are
deﬁned as simultaneous extreme returns.

In this paper I consider the stock and bond markets of 14 old EU member
states. I use two classiﬁcation schemes for deﬁning extreme returns, namely
ﬁrst the univariate classiﬁcation scheme of the previous literature and second
the new multivariate classiﬁcation scheme. I show that the two classiﬁcation
schemes give rise to statistically different simultaneous extreme return vari-
ables for stock markets and similar variables for bond markets. In the most
recent time period after the onset of the ﬁnancial crisis, the two classiﬁcation
schemes give similar results, also for the stock markets.

The remaining part of the paper is organized as follows. First, I introduce
the stock and bond data. Second, I provide details about the classiﬁcation
schemes. Third, I show the empirical results for the European stock markets
followed by the European bond markets. Finally, I conclude.

2 Data

I consider the following 14 old EU member states: Austria, Belgium, Den-
mark, Finland, France, Germany, Greece, Ireland, Italy, the Netherlands,
Portugal, Spain, Sweden, and the UK. I use local currency daily log-returns
from total return indices obtained from DataStream.

The DataStream stock indices cover the period from 1990 through 2012,
providing 5,999 daily stock return observations for each country or a total
of 83,986 daily stock return observations (14 times 5,999). The DataStream
benchmark 10-year government bond total return index cover a shorter period
from 2000 to 2012, providing 3,391 daily bond return observations for each
country or a total of 47,474 bond observations.

Table 1 shows the descriptive statistics for the two data sets. The mean
daily stock returns are small and the means are not signiﬁcantly different
across countries. The standard deviations are about the same size but sig-
niﬁcantly different. The stock return series are strongly leptokurtic but close
to being symmetric. Normality is always rejected by the Jarque and Bera (1980) test. For the bond returns the means are even smaller and still not significantly different across countries. There is more variability in the standard deviations. For a few countries, the returns have positive skewness and for most countries the returns are strongly leptokurtic. Normality is always rejected.

3 Classification Schemes

The classification schemes are exemplified for the stock markets. The bond market definitions follow straightforward.

3.1 Univariate Classification

First, I follow the previous literature beginning with Bae et al. (2003) and use their classification scheme. I denote this for the univariate classification scheme. A return for a given country (say Germany) is extreme if it is below the 5% threshold of the distribution of German stock returns for the entire sample period, i.e. the 5% lowest returns are classified as being extreme. The same procedure applies for the other 13 countries’ stock markets.

3.2 Multivariate Classification

Second, I introduce a new classification scheme, namely the multivariate classification scheme. Here the stock returns for all 14 countries are pooled. Then the extreme returns are those that are below the 5% threshold of the pooled distribution of the stock returns. Using the multivariate classification scheme implies that for some countries there will be more than 5% extreme returns and for others less than 5%. Extreme is seen from the perspective of the combined European stock markets.
3.3 Simultaneous Extreme Returns

I investigate the variable that counts the number of countries that have extreme stock returns simultaneously for each day in the sample period. The simultaneous extreme return variable takes on integer values between zero (no countries have extreme returns) and 14 (all countries have extreme returns). The simultaneous extreme return variable will be different for the two classification schemes. This variable has been the explained variable when analyzing integration and contagion of financial markets in the previous literature.

4 Extreme Stock Returns

Table 2 shows the number of extreme stock returns for both the univariate and multivariate classification. For most countries the number of extreme returns is significantly smaller using the multivariate classification scheme compared to the univariate scheme. For Finland, Greece, Italy, and Sweden the number of extreme returns is significantly larger. Only for four countries (France, Ireland, Netherlands, and Spain) is the number of extreme returns not significantly different for the two schemes.

Table 3 shows the frequency of the simultaneous extreme returns using both classification schemes. Using the multivariate classification, there are fewer days without any extreme returns and more days with only one extreme return. The proportion of days with two or more extreme returns is fairly close for the two classification schemes. Table 4 shows the cross tabulation of the simultaneous extreme returns for the two schemes. Here I combine days with at least three extreme returns into one category. As expected, the two schemes are not statistically independent. Importantly, the two schemes are significantly different.

I consider the sub-sample behavior of the simultaneous extreme return variables, namely by considering the period since the onset of the recent financial crisis separately (since 2008). Here, I cannot reject that the two
schemes are identical. For the period before the financial crisis (1990 to 2007), the two schemes are significantly different. So, it is mainly due to differences in the first part of the sample period that the two schemes lead to differences for the simultaneous extreme stock return variable.

5 Extreme Bond Returns

The corresponding results for extreme bond returns are shown in separate panels in Tables 2-4. Table 2 shows that only for Italy and the UK is the multivariate classification scheme not significantly different from the univariate scheme. For most countries the multivariate classification scheme defines fewer extreme returns (same as for stocks). For Greece, Ireland, Portugal, and Spain are there significantly more extreme returns using the multivariate classification scheme. Those are the countries which have experienced the sovereign debt crisis.

The simultaneous extreme bond returns in Table 3 show the same pattern as for the stock market. The cross tabulation in Table 4 of the simultaneous extreme bond returns shows that the two schemes are not independent (same as for stocks). In contrast to the stock markets, it cannot be rejected that the two schemes are identical. This also applies for the sub periods before and after the onset of the financial crisis. So, for bond markets it is of no consequence whether I use the univariate or the multivariate classification scheme for investigating integration and contagion issues. For bond markets it is thus feasible to use a shorter time period when classifying extreme returns, simply by pooling the information from all the European bond markets.

6 Conclusion

In this paper I investigate the classification of extreme returns for EU stock and bond markets. So far, the literature has considered the classification of extreme returns for each country separately. In this paper I show that in
particular for bond markets it is relevant to consider all the EU bond markets jointly, whereby a shorter sample period is called for. For the most recent time period (since 2008) the two classification schemes give rise to similar simultaneous extreme return variables, also for the EU stock markets.

Future research could investigate if the multivariate classification scheme is also feasible in other geographic markets than the European.

References


<table>
<thead>
<tr>
<th>Year</th>
<th>Authors</th>
<th>Title</th>
</tr>
</thead>
<tbody>
<tr>
<td>2013-19</td>
<td>Asger Lunde and Kasper V. Olesen</td>
<td>Modeling and Forecasting the Volatility of Energy Forward Returns - Evidence from the Nordic Power Exchange</td>
</tr>
<tr>
<td>2013-20</td>
<td>Anders Bredahl Kock</td>
<td>Oracle inequalities for high-dimensional panel data models</td>
</tr>
<tr>
<td>2013-21</td>
<td>Malene Kallestrup-Lamb, Anders Bredahl Kock and Johannes Tang Kristensen</td>
<td>Lassoing the Determinants of Retirement</td>
</tr>
<tr>
<td>2013-22</td>
<td>Johannes Tang Kristensen</td>
<td>Diffusion Indexes with Sparse Loadings</td>
</tr>
<tr>
<td>2013-24</td>
<td>Nima Nonejad</td>
<td>A Mixture Innovation Heterogeneous Autoregressive Model for Structural Breaks and Long Memory</td>
</tr>
<tr>
<td>2013-26</td>
<td>Nima Nonejad</td>
<td>Long Memory and Structural Breaks in Realized Volatility: An Irreversible Markov Switching Approach</td>
</tr>
<tr>
<td>2013-27</td>
<td>Nima Nonejad</td>
<td>Particle Markov Chain Monte Carlo Techniques of Unobserved Component Time Series Models Using Ox</td>
</tr>
<tr>
<td>2013-28</td>
<td>Ulrich Hounyo, Silvia Goncalves and Nour Meddahi</td>
<td>Bootstrapping pre-averaged realized volatility under market microstructure noise</td>
</tr>
<tr>
<td>2013-29</td>
<td>Jiti Gao, Shin Kanaya, Degui Li and Dag Tjøstheim</td>
<td>Uniform Consistency for Nonparametric Estimators in Null Recurrent Time Series</td>
</tr>
<tr>
<td>2013-30</td>
<td>Ulrich Hounyo</td>
<td>Bootstrapping realized volatility and realized beta under a local Gaussianity assumption</td>
</tr>
<tr>
<td>2013-31</td>
<td>Nektarios Aslanidis, Charlotte Christiansen and Christos S. Savva</td>
<td>Risk-Return Trade-Off for European Stock Markets</td>
</tr>
<tr>
<td>2013-32</td>
<td>Emilio Zanetti Chini</td>
<td>Generalizing smooth transition autoregressions</td>
</tr>
<tr>
<td>2013-33</td>
<td>Mark Podolskij and Nakahiro Yoshida</td>
<td>Edgeworth expansion for functionals of continuous diffusion processes</td>
</tr>
<tr>
<td>2013-34</td>
<td>Tommaso Proietti and Alessandra Luati</td>
<td>The Exponential Model for the Spectrum of a Time Series: Extensions and Applications</td>
</tr>
<tr>
<td>2013-35</td>
<td>Bent Jesper Christensen, Robinson Kruse and Philipp Sibbertsen</td>
<td>A unified framework for testing in the linear regression model under unknown order of fractional integration</td>
</tr>
<tr>
<td>2013-36</td>
<td>Niels S. Hansen and Asger Lunde</td>
<td>Analyzing Oil Futures with a Dynamic Nelson-Siegel Model</td>
</tr>
<tr>
<td>2013-37</td>
<td>Charlotte Christiansen</td>
<td>Classifying Returns as Extreme: European Stock and Bond Markets</td>
</tr>
</tbody>
</table>