The Expectations Hypothesis of the Term Structure: The Case of Ireland*

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Abstract: Using a number of short-term maturities and monthly data, 1984-1997, we provide a number of tests of the expectations hypothesis (EH) of the term structure. The paper draws on cointegration techniques and the methodological approach of Campbell and Shiller (1987,1991). On balance our results lend support to the EH and are broadly consistent with recent findings for the UK, but are in sharp contrast to those for the US.

I  INTRODUCTION

The expectations hypothesis (EH) of the term structure implies that the yield spread between the long rate and short rate is an optimal predictor of future changes in short rates over the life of the “long bond”.

There is a great deal of evidence on the EH for the US based on the Campbell and Shiller (1991) VAR methodology using monthly data on spot rates (e.g. Hardouvelis, 1994). In general, for a wide variety of maturities from 1 to 12 months and for 2,3,4 ... 10-year maturities, Campbell and Shiller (1991) reject the EH. The (long-short) interest rate spread does not predict the direction of changes in the long-term interest rate that is consistent with the EH, and future

*The views expressed in this paper are not necessarily those held by the Central Bank of Ireland and are the personal responsibility of the authors.
changes in short rates are not often closely correlated with the long-short spread (Campbell and Shiller, 1991).

Kugler (1988) using US, German and Swiss monthly data on one and three month Euromarket deposit rates found support for the EH only on German data (for the period of March 1974 to August 1986). Similarly, Engsted (1994) using Danish money market rates and (Engsted and Tanggaard, 1994) for longer maturity bonds find considerable support for the EH when the variation in interest rates is relatively large, such as in the post-1992 ERM “crisis period”. This is to be expected following the analysis of Mankiw and Miron (1986), for if interest rate stabilisation results in random walk behaviour for short rates, then the expected change in short rates is zero and the spread has no predictive power for future short rates (see also Rudebusch, 1995).

Using the Campbell-Shiller VAR methodology on data at the short end of the maturity spectrum (i.e. up to one year) Cuthbertson (1996a) finds reasonable support for the EH on UK data. This is in contrast to Taylor (1992) who uses maturities of 5, 10 and 15 years and finds strongly against the EH (see also MacDonald and Speight, 1991). To our knowledge, the only related paper using Irish data, is that of Hurley (1990).1 Using a number of interest rate combinations for the period 1979 to 1989, Hurley (1990) finds little evidence to support the EH, in that the spread does not forecast future changes in short rates. However, as noted by the author, the study suffers a number of drawbacks. First, the data used in the study are yields taken from the Central Bank Bulletin and so are not continuously compounded spot rates as required by (linear) tests of the EH. Second, the author uses OLS estimation, which is inappropriate, given the inclusion of overlapping observations.

The main aim of this paper is to present evidence on the behaviour of the term structure of Irish interest rates at the short end of the maturity spectrum. The paper applies co-integration techniques and the methodological approach of Campbell and Shiller (1987,1991).2 To our knowledge the expectations hypothesis (EH) using the VAR approach has not been examined using Irish data. We test parameter restrictions on the VAR models using a high quality data set. By using spot rates based on quoted discount rates we avoid having to use an approximation to zero coupon yields and the par yield approximation

1. McGettigan (1995) drawing on the approach of McCulloch (1971, 1975) “fits” yield curves to Irish rates using discount functions but does not test the EH, as is done in this paper. As noted by McGettigan, a similar “curve fitting” approach was used by Breen and Keogh (1990). Breen (1991) in his comment on Joyce (1991) also draws the distinction between “fitting” yield curves and explicit tests of the EH.

2. Earlier work may be found in Melino (1988) and Shiller (1989).
which are required when analysing coupon paying bonds (Shiller, 1979). We also assess the results in comparison to the previous evidence.3

II THE EXPECTATIONS HYPOTHESIS

The expectations hypothesis (EH) of the term structure posits that the return on an n-period bond $R_t^{(n)}$ is determined solely by expectations of (current and) future rates on a set of m-period short rates $r_t^{(m)}$ (where $n > m$). Using continuously compounded spot rates the “fundamental term structure” relationship is:

$$R_t^{(n)} = \frac{1}{k} \sum_{i=0}^{k-1} E_t r_t^{(m)}_{t+i}$$

where $k = n/m$ is an integer and $E_t$ is the expectations operator (with information up to and including time $t$). If there is a time invariant term premium, which is constant for given $(n,m)$ then (1) will also contain a constant term. The intuition behind (1) is easily seen by taking $n=3$ and $m=1$. If $\$1$ is invested at the 3-year spot rate, then the certain amount received after 3-years is $\$(1+R_t)^3$. Alternatively at $t=0$, the investor can consider investing $\$1$ at the one-period rate $r_t$ and then reinvesting at the one-period rates in years two and three (i.e. rolling over the one-period investment). The latter is a risky strategy and results in expected “dollar” receipts of $\$(1+r_t) (1+E_t r_{t+1}) (1+E_t r_{t+2})$. The EH assumes investors are risk neutral and that the market is efficient, hence in equilibrium $(1+R_t)^3 = (1+r_t) (1+E_t r_{t+1}) (1+E_t r_{t+2})$. Taking logarithms of both sides of the latter expression and noting that $\ln(1+r_t)$ is the continuously compounded interest rate, we obtain Equation (1) – see Cuthbertson (1996b) for further details.

We can re-arrange (1) in terms of the spread and the change in interest rates (since below we find that these are stationary variables) and (1) can then be seen to imply that the “long-short” spread is an optimal predictor of future changes in short rates, $r_t^{(m)}$:

$$S_t^{(n,m)} = E_t \sum_{i=1}^{k-1} (1 - i / k) \Delta^{m}r_t^{(m)}_{t+i} = E_t \left[ PFS_t^{(n,m)} \right]$$

where \( S_t^{(n,m)} = \left( R_t^{(n)} - r_t^{(m)} \right) \) is the yield spread. Equation (2) implies that if future short rates are expected to rise, then this will be accompanied by an increase in the spread. To see the intuition behind (2), again consider the case \( n=3, m=1 \). Suppose at \( t=0 \), investors believe that inflation in years two and three will be higher (than previously anticipated). Then they will revise upwards their forecasts of the one-period rates pertaining to years 2 and 3, that is \( E_t \Delta r_{t+1} \) and \( E_t \Delta r_{t+2} \), and hence \( E_t \Delta r_{t+1} \) and \( E_t \Delta r_{t+2} \) will also rise. Therefore, rolling over “one-period” investments will currently give a higher expected return than investing at the 3-year spot rate. Investors will therefore sell 3-year (zero coupon) bonds to invest in one-year bonds, and the price of 3-year bonds will consequently fall. But the latter implies that their yield \( R_t \) will rise, as will the spread \( S_t = \left( R_t - r_t \right) \). Arbitrage ensures that \( R_t \) increases until the higher spread just equals the (weighted average of) future expected increases in one-period rates, as summarised in (2). For our simple case, Equation (2) is \( S_t = \left( \frac{2}{3} \right) E_t \Delta r_{t+1} + \left( \frac{1}{3} \right) E_t \Delta r_{t+2} \).

The perfect foresight spread \( PFS_t \) in (2) is simply the (weighted average) of actual future changes in short term rates (which agents are trying to forecast). However, in the literature it is referred to as the “perfect foresight spread” because under the EH, it can also be interpreted as the spread that would ensue if agents had perfect foresight about future movements in interest rates (i.e. made no forecast errors).

A testable implication of Equation (2) is that the spread Granger causes future changes in short rates.\(^4\) If \((R_t^{(n)}, r_t^{(m)})\) are found to be \( I(1) \), then \( \Delta r_t^{(m)} \) is \( I(0) \), which from Equation (2) implies that the spread \( S_t^{(n,m)} = \left( R_t^{(n)} - r_t^{(m)} \right) \) should also be \( I(0) \). The latter implies that \((R_t^{(n)}, r_t^{(m)})\) should be co-integrated with a co-integrating vector \((1, -1)\).\(^5\) If we now add the assumption of rational expectations (RE):

\[
 r_{t+im}^{(m)} = E_t r_{t+im}^{(m)} + \varepsilon_{t+im} 
\]

we obtain the following single equation test of the null of the “expectations hypothesis plus rational expectations”, EH + RE:

\(^4\) Strictly, failure of Granger causality does not constitute a rejection of the EH, but a failure to confirm it.

\(^5\) Strictly, for this to hold, forecast errors and any term premia must also be \( I(0) \). Tzavalis and Wickens (1998) report results for a stationary term premia, using US data. An alternative finding is reported in Evans and Lewis (1994), however this may be due to the regime shift found by Tzavalis and Wickens not being considered. Shea (1992) examines the possibility of multiple co-integrating vectors, an issue not explored here since we concentrate on tests of bilateral relationships.
THE EXPECTATIONS HYPOTHESIS OF THE TERM STRUCTURE

\[
PFS_t^{(n,m)} = \alpha + \beta S_t^{(n,m)} + \gamma \Omega_t + \varepsilon_t^* \\
H_0: \alpha = \gamma = 0, \beta = 1
\]  

(4)

\(\varepsilon_t^*\) is a moving average error of order \((n-m-1)\) consisting of a weighted sum of future values of \(\varepsilon_{t+t+m}\), and \(\Omega_t\) represents the information available to agents at time \(t\), or earlier. Under RE, \(\varepsilon_t^*\) is independent of \(\Omega_t\), and in particular is independent of the yield spread. If there is a constant term premia or if there are differential yet constant transactions costs (between investing “long” and in a series of rolled-over short-term investments) then \(\alpha \neq 0\). Under RE the right hand side variables in Equation (4) are independent of \(\varepsilon_t^*\) and hence we do not require IV estimation. However a GMM estimator is employed to correct the covariance matrix for the moving-average error of order \((n-m-1)\) and possible heteroscedasticity (Hansen, 1982; Newey and West, 1987).

III  VAR METHODOLOGY

One of the problems in using Equation (2) is the correction needed for overlapping data and it may result in standard errors which are biased in finite samples, with the degree of bias being more severe, the greater the degree of “overlap”. The VAR methodology overcomes the latter defect and also allows forecasts of futures changes in interest rates to be influenced not only by lagged changes in interest rates (as in a single equation framework) but also by the spread, which according to the EH has predictive power for future interest rates (see Equation (2)). More formally, if \(Z_t = (S_t^{(n,m)}, \Delta r_t^{(m)})\) is stationary, then there exists a bivariate Wold representation (Hannan, 1970) which may be approximated by a vector autoregression (VAR) of order \(p\), which in companion form is:

\[
Z_t = AZ_{t-1} + v_t
\]

(5)

where \(A\) is the matrix of estimated VAR coefficients. Forecasts of future changes in interest rates can be obtained from (5) by noting that \(E_t(Z_{t+j}) = A^j Z_t\). However, to test the EH using (2) what we require is a time series of future forecasts of the change in interest rates. This is easily obtained if we create a “selection vector” \(e_2\) which has all zero entries except for a single entry of “unity” to “pick out” \(\Delta r_t^{(m)}\) from the \(Z\) vector. A forecast of \(E_t\Delta r_{t+j}^{(m)}\) is then given by \(e_2' (A^j Z_t)\).
The left hand side of (2) is the spread \(S_t^{(n,m)}\) which is the first entry in the \(Z\) vector, so \(e_1' Z_t = S_t^{(n,m)}\). Substituting \(E_t\Delta r_{t+j}^{(m)} = e_2' (A^j Z_t)\) for the interest rate forecasts in (2) and for \(e_1' Z_t = S_t^{(n,m)}\), we obtain the following restrictions on the estimated parameters of the \(A\)-matrix (see Campbell and Shiller, 1991; Taylor, 1992 and Cuthbertson 1996a).
\[ e_1' = e_2'A [I - (m/n)(I - A^n) (I-A^m)^{-1}] (I - A)^{-1} \] (6)

We apply Equation (6) on monthly data for \((n,m) = (6,3), (6,1), (3,1)\). The VAR methodology suggests several approaches to testing the EH + RE under weakly rational expectations. The restrictions in (6) can be shown to imply that information at time \(t\) other than that contained in \(S_t^{(n,m)}\) should not help to predict future changes in short rates (i.e. \(S_t^{(n,m)}\) must be an optimal predictor of future changes in short rates). The restrictions on the parameters of the VAR in Equation (6) are tested using a Wald test.\(^6\)

The forecasts from the VAR of the (weighted average) of future short rates on the right-hand-side of Equation (2) are a complex function of the estimated parameters of the \(A\)-matrix of the VAR. This forecast is known as the “theoretical spread”, denoted \(S_t^{(n,m)'}\) where,

\[ S_t^{(n,m)'} = e_2'A [I - (m/n)(I - A^n) (I-A^m)^{-1}] (I - A)^{-1} Z_t \] (7)

Using the VAR estimates of the \(A\)-matrix, Equation (7) can be used to give a time series of the (weighted average) of the forecasts of future changes in interest rates \(S_t^{(n,m)'}\). For example, for \(n=3, m=1\) the (weighted) forecast of future changes in one-period interest rates \(S_t^{(3,1)'} = (2/3) E_t \Delta r_{t+1} + (1/3) E_t \Delta r_{t+2}\) and using (7) is easily seen to be given by \(\{ (2/3) e_2'A + (1/3) e_2'A^2 \} Z_t\). If the EH is correct then we expect the time series of the theoretical spread \(S_t^{(n,m)'}\) to move the same as the actual spread \(S_t^{(n,m)}\). Indeed the formal test of the VAR restrictions in (6) are merely testing the hypothesis, \(H_0: S_t^{(n,m)} = S_t^{(n,m)'}\). Campbell and Shiller (1991) note that formal tests of the VAR restrictions may lead to rejection of the EH even though the deviations from the null are quite small from an economic perspective. They suggest computing the time series forecasts of interest rates in (7) (i.e. the “theoretical spread”), without imposing the VAR restrictions and comparing this with the time series behaviour of the actual spread. If the EH is true then we would expect a graph of the actual spread \(S_t^{(n,m)}\) and the theoretical spread \(S_t^{(n,m)'}\) to move together and hence for the standard deviation ratio SDR = \(\sigma(S_t^{(n,m)'})/\sigma(S_t^{(n,m)})\), and the correlation coefficient, Corr\((S_t^{(n,m)}, S_t^{(n,m)'}\) both to equal unity.\(^7\) If \(\sigma(S_t^{(n,m)}) > \sigma(S_t^{(n,m)'}\) then there is “excess volatility”, that is the actual spread is more volatile than the optimal predictor of future short rates.

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6. In testing the VAR restrictions we use a GMM correction to the covariance matrix of the VAR system.

7. The standard errors of SDR = \(\sigma(S_t^{(n,m)'})/\sigma(S_t^{(n,m)})\) and R (\(S_t^{(n,m)}, S_t^{(n,m)'}\) are non-linear functions of the estimated A matrix from the VAR and can be computed as \([f(\gamma)'\Psi f(\gamma)]\) where \(f(\gamma)\) are the statistics of interest and \(\Psi\) is the (GMM) variance-covariance matrix of the parameters \(\gamma\).
IV EMPIRICAL RESULTS

4.1 The Data

The data used consists of Irish short-term interest rates (spot rates) with a term to maturity of less than six months, that is money market rates (which were kindly provided by the Bank of Ireland from simultaneously observed screen-quoted rates). The data set is monthly, from January 1984 to October 1997 and the rates are converted to continuously compounded rates. The 1-month and the 6-month yields are graphed in Figure 1 and it can be seen that these appear to move together in the long run but in the short run there are divergences as the long-short spread varies over time.

Figure 1: 1-Month and 6-Month Interest Rates

4.2 Unit Roots and Co-integration

Table 1 gives the results of unit root tests, which indicate that we cannot reject the null hypothesis that changes in short rates \( \Delta r_t^{(m)} \) and the yield spread \( S_t^{(n,m)} \) are I(0). Table 2 shows the OLS co-integration regression results and as can be seen the \( \beta \) in \( R_t^{(n)} = \alpha + \beta r_t^{(m)} \) is “bounded” by the 2 regressions (i.e. \( R_t^{(n)} \) on \( r_t^{(m)} \) and vice-versa).\(^9\) This result provides weak evidence in favour of the EH under the assumption of a constant or stationary term premium and any expectation scheme that yields I(0) forecast errors.

8. We focus on tests of the EH in terms of 3 interest rate combinations; (1,3), (1,6) and (3,6) month.

9. Hall (1986) suggests that the co-integration regressions of \( y_t \) on \( x_t \) and \( x_t \) on \( y_t \) should provide upper and lower limits for the co-integration parameter.
Table 1: Unit Root Tests

<table>
<thead>
<tr>
<th>Variable</th>
<th>Maturity</th>
<th>ADF(5)</th>
<th>PP-Stat</th>
</tr>
</thead>
<tbody>
<tr>
<td>Interest Rate: ( R_t^{(n)} )</td>
<td>1 month</td>
<td>-2.64</td>
<td>-2.62</td>
</tr>
<tr>
<td></td>
<td>3 month</td>
<td>-2.35</td>
<td>-2.20</td>
</tr>
<tr>
<td></td>
<td>6 month</td>
<td>-2.20</td>
<td>-1.82</td>
</tr>
<tr>
<td>Change in interest rate: ( \Delta R_t^{(n)} )</td>
<td>1 month</td>
<td>-6.56</td>
<td>-11.49</td>
</tr>
<tr>
<td></td>
<td>3 month</td>
<td>-6.42</td>
<td>-11.49</td>
</tr>
<tr>
<td></td>
<td>6 month</td>
<td>-5.74</td>
<td>-12.07</td>
</tr>
<tr>
<td>Spread: ( S_t^{(n,m)} )</td>
<td>(3,1) month</td>
<td>-4.49</td>
<td>-6.33</td>
</tr>
<tr>
<td></td>
<td>(6,1) month</td>
<td>-4.40</td>
<td>-5.31</td>
</tr>
<tr>
<td></td>
<td>(6,3) month</td>
<td>-4.32</td>
<td>-4.89</td>
</tr>
</tbody>
</table>

Notes: The sample period is from January 1984 to October 1997. ADF(5) is the augmented Dickey-Fuller statistic with 5 lags, which ensures the regressions are free of serial correlation. PP is the Phillips-Perron (1988) statistic with correction for up to 5th order serial correlation. The critical value for both test statistics is –2.86 at the 5 per cent significance level.

Table 2: OLS Cointegration Tests: \( R_t^{(n)} = \alpha + \beta r_t^{(m)} + \epsilon_t \)

<table>
<thead>
<tr>
<th>Dep. Variable</th>
<th>Maturity of Expl. Variable</th>
<th>( \beta ) coeff.</th>
<th>ADF(1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>6 month</td>
<td>1 month</td>
<td>0.84</td>
<td>-4.94</td>
</tr>
<tr>
<td>1 month</td>
<td>6 month</td>
<td>1.08</td>
<td>-5.30</td>
</tr>
<tr>
<td>6 month</td>
<td>3 month</td>
<td>0.93</td>
<td>-4.63</td>
</tr>
<tr>
<td>3 month</td>
<td>6 month</td>
<td>1.05</td>
<td>-4.75</td>
</tr>
<tr>
<td>3 month</td>
<td>1 month</td>
<td>0.93</td>
<td>-6.21</td>
</tr>
<tr>
<td>1 month</td>
<td>3 month</td>
<td>1.05</td>
<td>-6.44</td>
</tr>
</tbody>
</table>

Notes: The augmented Dickey-Fuller (ADF) statistic for the residuals, \( \epsilon_t \), ensuring that enough lags are present to ensure no serial correlation remains. The critical value for the ADF statistic (at 5 per cent significance) is –2.88 (MacKinnon, 1991). The cointegrating regressions are estimated for the period January 1984 to October 1997.

We now turn to tests based on the Johansen (1988) procedure. The Johansen results, shown in Table 3, provide strong evidence that \( R_t^{(n)} \) and \( r_t^{(m)} \) are cointegrated and we cannot reject the hypothesis that the co-integrating vector is given by the theoretical value \((-1,1)\). The (normalised, unconstrained) point estimates for the co-integrating vectors from the Johansen procedure are \((-1, 0.99)\) for each case. The Johansen procedure being a systems estimator and including
lagged variables to remove any serial correlation in the residuals should give more informative results than the OLS cointegrating regressions, reported above.

Table 3: Johansen Procedure on $R_t^{(n)}$ and $r_t^{(m)}$

<table>
<thead>
<tr>
<th>Interest rates $(n,m)$</th>
<th>Lag length</th>
<th>Cointegrating Vector</th>
<th>LR test</th>
</tr>
</thead>
<tbody>
<tr>
<td>(6,1) months</td>
<td>2</td>
<td>$(-1, 0.99)$</td>
<td>0.51</td>
</tr>
<tr>
<td>(6,3) months</td>
<td>3</td>
<td>$(-1, 0.99)$</td>
<td>0.04</td>
</tr>
<tr>
<td>(3,1) months</td>
<td>3</td>
<td>$(-1, 0.99)$</td>
<td>2.32</td>
</tr>
</tbody>
</table>

Notes: In the Johansen procedure both the maximum eigenvalue test and the trace test do not reject the null of a unique co-integrating vector. The likelihood ratio (LR) statistic in column 4 tests the null that the co-integrating vector is $(-1,1)$. Under the null, the reported test statistic has a critical value (at 5 per cent significance level) of 3.84.

4.3 The Spread and the Predictability of Changes in Short Rates

The regression of the perfect foresight spread, $PFS_t^{(n,m)}$ on the actual spread $S_t^{(n,m)}$ and the limited information set $H_t$ (consisting of lags of $S_t^{(n,m)}$ and $\Delta r_t^{(m)}$) are shown in Table 4. In all cases we do not reject the null of $H_0: \beta=1$ or that information, available at time $t$ or earlier does not incrementally add to the predictions of future interest rates. We also do not reject the null that the constant term premium is zero (i.e. $\alpha = 0$). The results therefore do not reject the EH + RE.

Table 4: Does the Spread Predict Future Changes in Short-Rates?

Regression: $PFS_t^{(n,m)} = \alpha + \beta S_t^{(n,m)} + \gamma H_t$

<table>
<thead>
<tr>
<th>Spread $(n,m)$</th>
<th>$\alpha$</th>
<th>$\beta$</th>
<th>Wald Test</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>s.e.(\alpha)</td>
<td>s.e.(\beta)</td>
<td>$H_0: \beta=1$</td>
</tr>
<tr>
<td>(6,1)</td>
<td>-0.001</td>
<td>1.04</td>
<td>0.09</td>
</tr>
<tr>
<td></td>
<td>(0.001)</td>
<td>(0.15)</td>
<td>[0.77]</td>
</tr>
<tr>
<td>(6,3)</td>
<td>-0.0006</td>
<td>1.02</td>
<td>0.007</td>
</tr>
<tr>
<td></td>
<td>(0.0008)</td>
<td>(0.32)</td>
<td>[0.93]</td>
</tr>
<tr>
<td>(3,1)</td>
<td>-0.001</td>
<td>0.87</td>
<td>0.38</td>
</tr>
<tr>
<td></td>
<td>(0.001)</td>
<td>(0.21)</td>
<td>[0.54]</td>
</tr>
</tbody>
</table>

Notes: The regression coefficients reported in columns 2 and 3 are from the regression with $\gamma = 0$ imposed. The method of estimation is GMM with a correction for heteroscedasticity and moving average errors using the Newey-West (1987) declining weights. The last 3 columns report Wald statistics and marginal significance levels for the null hypothesis stated. For $H_0: \gamma = 0$ the reported results are for an information set which includes 4 lags of the change in the interest rates and the interest rate spread. The null $H_0: \beta=1$, is conditional on $\gamma=0$ while the null $H_1: \alpha=0$, $\beta=1$ is also conditional on $\gamma=0$. 
4.4 The Theoretical Spread and the VAR Results

Table 5 contains the results from the VAR models for $S_t(n,m)$ and $\Delta r_t(m)$. The lag length is chosen to minimise the Akaike Information Criterion (AIC), except for the rare occasions when additional lags are required to avoid any serial correlation in the residuals. A weak test of the EH is that the spread Granger-causes changes in short-term interest rates and this is not rejected for all maturities (Table 5, column 3). There is also Granger-causality from $\Delta r_t(m)$ to $S_t(n,m)$ for the (6m,3m) case, indicating feedback in the VAR regression.

Table 5: VAR Model for $(S_t(n,m), \Delta r_t(m))$

<table>
<thead>
<tr>
<th>Spread $(n,m)$</th>
<th>Lag</th>
<th>Granger Tests</th>
<th>Causality</th>
<th>Ljung-Box Q(26) S_t–eqn.</th>
<th>$\Delta r_t$–eqn.</th>
<th>R2–statistic S_t–eqn.</th>
<th>$\Delta r_t$–eqn.</th>
</tr>
</thead>
<tbody>
<tr>
<td>(6,1)</td>
<td>2</td>
<td>&lt;0.01</td>
<td>0.48</td>
<td>9.36</td>
<td>17.4</td>
<td>0.52</td>
<td>0.22</td>
</tr>
<tr>
<td>(6,3)</td>
<td>2</td>
<td>&lt;0.01</td>
<td>&lt;0.01</td>
<td>13.3</td>
<td>27.7</td>
<td>0.62</td>
<td>0.19</td>
</tr>
<tr>
<td>(3,1)</td>
<td>2</td>
<td>&lt;0.01</td>
<td>0.50</td>
<td>11.3</td>
<td>18.8</td>
<td>0.38</td>
<td>0.26</td>
</tr>
</tbody>
</table>

Notes: “Lag” denotes the lag length that minimises the Akaike Information Criterion (AIC). Where the latter (occasionally) results in an equation system with serial correlation, the AIC is overridden and extra lags added (back) until any residual serial correlation is eliminated. The critical value for Q(26) is 38.89 (5 per cent significance level). In columns 3 and 4 we report the marginal significance levels for the Granger-causality tests of $S_t(n,m)$ on $\Delta r_t(m)$ and vice versa (statistics are calculated after applying the GMM correction for heteroscedasticity used in Campbell and Shiller (1991)). The final 2 columns give the R2–statistic for each equation. The regressions are estimated for the whole sample period, January 1984 to October 1997.

For illustrative purposes the graph of the actual spread $S_t$ and the theoretical spread $S_t'$ are shown for $(n,m) = (6, 1)$ and they move closely together, Figure 2. In the regression of $S_t$ on $S_t'$ (Table 6) the point estimate of the slope coefficients are very close to unity, for all 3 maturity combinations. The intercepts in these regressions are not statistically significantly different from zero in each case. Table 7 provides the metrics for the relationship between the actual spread $S_t$ and the theoretical spread $S_t'$. The results indicate that the VAR restrictions are not rejected. For all maturities there is a strong correlation (column 4) between the actual spread $S_t(n,m)$ and the predicted (theoretical) spread $S_t'(n,m)$. The standard deviation ratio, $SDR = \sigma(S_t(n,m)')/\sigma(S_t(n,m))$ yields estimates (column 3) which are all within two standard errors of unity. The Wald test of the VAR cross equation parameter restrictions (Table 7, column 2), are not rejected for the (6m,3m) and (6m,1m) case, but are rejected for (3m,1m).

On balance our results would appear to lend support to the EH, but how do we interpret the rejection of the VAR restrictions? Campbell and Shiller (1987) show that rejection implies either (1) information $(H_t \subseteq \Omega_t)$ at time $t$ or earlier (other than $S_t(n,m)$) influences future changes in short rates or (2) the influence of
Table 6: Regression of the Actual Spread $S_t$ on the Theoretical Spread $S'_t$

<table>
<thead>
<tr>
<th>Interest Rate Maturity</th>
<th>$\alpha$ coeff.</th>
<th>s.e.</th>
<th>$\beta$ coeff.</th>
<th>s.e.</th>
<th>$R^2$ statistic</th>
</tr>
</thead>
<tbody>
<tr>
<td>(6,1)</td>
<td>-0.01</td>
<td>0.02</td>
<td>0.90</td>
<td>0.06</td>
<td>0.93</td>
</tr>
<tr>
<td>(6,3)</td>
<td>0.02</td>
<td>0.01</td>
<td>0.81*</td>
<td>0.07</td>
<td>0.87</td>
</tr>
<tr>
<td>(3,1)</td>
<td>0.02</td>
<td>0.03</td>
<td>0.83</td>
<td>0.13</td>
<td>0.80</td>
</tr>
</tbody>
</table>

Notes: The regressions are estimated for the whole sample period, January 1984 to October 1997. The estimated regression is $S_t = \alpha + \beta S'_t + \varepsilon_t$ which is estimated by GMM with heteroscedastic corrected errors. A star indicates the estimated coefficient is statistically different from that implied by the null hypothesis (at a 5 per cent significance level), which for $\alpha$ is $H_0: \alpha = 0$ and for $\beta$ is $H_0: \beta = 1$. The theoretical spread $S'_t$ is obtained from the predictions from the VAR using $z = [S_t, \Delta R_t]$.

The current spread $S_t^{(n,m)}$ on future changes in interest rates via the chain rule of forecasting is less than required by the EH. However in contrast to (1) our single equation perfect foresight regressions in Table 4 reject the null that $H_t$ influences future changes in interest rates. Hence rejection of the VAR restrictions is probably due to the “low weight” given to $S_t^{(n,m)}$ in the VAR regression. This may occur because over short forecasting horizons one might expect agents to utilise almost minute by minute observations of $S_t^{(n,m)}$ and $\Delta R_t^{(m)}$ and hence forecasts based on monthly data might not adequately mimic such behaviour for the (3m,1m) maturities.
Table 7: Tests of the EH Using Weakly Rational Expectations

<table>
<thead>
<tr>
<th>Spread (n,m)</th>
<th>Wald test $W$ [.] = p-value</th>
<th>$\sigma(S_t^{(n,m)})/\sigma(S_t^{(n,m')})$ (.) = std. Error</th>
<th>$\text{Corr}(S_t^{(n,m)}, S_t^{(n,m')})$ (.) = std. Error</th>
</tr>
</thead>
<tbody>
<tr>
<td>(6,1)</td>
<td>4.91 [0.30]</td>
<td>1.08(0.19)</td>
<td>0.96(0.05)</td>
</tr>
<tr>
<td>(6,3)</td>
<td>3.09 [0.54]</td>
<td>1.05(0.24)</td>
<td>0.93(0.10)</td>
</tr>
<tr>
<td>(3,1)</td>
<td>12.19 [0.02]</td>
<td>0.97(0.23)</td>
<td>0.89(0.12)</td>
</tr>
</tbody>
</table>

Notes  The regressions are estimated for the whole sample period, January 1984 to October 1997.

V  INTERPRETATION OF RESULTS

In this section we analyse our results and compare them with those from other studies. On balance our results favour the EH. The perfect foresight regressions (Table 4), the standard deviation ratios $\sigma(S_t^{(n,m)})/\sigma(S_t^{(n,m')})$, and the coefficient of determination $\text{Corr}(S_t^{(n,m)}, S_t^{(n,m')})$ (Table 5), are consistent with the EH and this may be contrasted with the rejection of the VAR cross-equation restrictions for (3m,1m) combination in Table 5.10

First, the perfect foresight spread regressions implicitly allow potential future events (known to agents but not to the econometrician) to influence expectations, whereas the VAR approach requires the explicit information set known both to agents and the econometrician. Hence, if the econometrician erroneously excludes variables affecting traders’ perceptions, then the estimated VAR coefficients may be biased, resulting in rejection of the VAR cross equation restrictions. Second, if agents actually do use the VAR methodology for forecasting, one would expect them to utilise almost minute by minute observations of $(S_t^{(n,m)}, \Delta r_t^{(m)})$: hence forecasts based on monthly data seem unlikely to adequately mimic such behaviour. Third, Campbell and Shiller (1991) have argued that rejection of the cross-equation parameter restrictions although statistically significant may not constitute a major departure from the EH on economic grounds, as long as the theoretical spread closely tracks the actual spread.11 Finally, the Monte Carlo evidence in Bekaert et al. (1997) and Bekaert and Hodrick (2000) shows that the Wald test suffers from severe size distortions and use of asymptotic critical values “results in gross over-rejection of the null”.

Our perfect foresight spread results are broadly consistent with the results in Cuthbertson (1996a); who found slope coefficients ranging from 0.73 to 1.3.

10. As has been mentioned earlier, this is the first known study using the VAR methodology for Irish interest rates, and as such comparison will be made with similar studies on US and UK data.

11. For example, suppose theory suggests an elasticity of unity between 2 variables and the estimated equation is $\ln y = 0.99 \ln x$ with a standard error 0.001. While we strongly reject the null of a unit elasticity, the predicted values of $\ln y$, will closely mirror the actual values.
The author could not reject the null, $H_0: \beta = 1, \gamma = 0$ which is consistent with the EH for the UK at the short end of the maturity spectrum. This is also consistent with Hurn et al. (1996) and Cuthbertson et al. (1996).\(^{12}\)

Campbell and Shiller (1992) use monthly data on US government bonds for the period 1946 to 1986, and their results are broadly consistent with our reported results for the perfect foresight spread equations. However Campbell and Shiller find that the $\text{Corr}(S_t^{(n,m)}, S_t^{(n,m)^*})$ are relatively low being in the range 0-0.7 and the values of the variance ratio are in the range 2-10 for maturities of less than 1 year. Campbell and Shiller do not directly test the VAR cross-equation restrictions but this has been done subsequently by Shea (1992) who in general finds they are rejected. Again our results are generally consistent with those of Cuthbertson (1996a) using UK data at short maturities. Cuthbertson’s results from the VAR models for $S_t(n,m)$ and $\Delta r_t(m)$ indicate that $S_t(n,m)$ Granger causes $S_t(n,m)$ and $\Delta r_t(m)$: a weak test of the EH.\(^{13}\) The author also finds that for all maturities there are strong correlations between the actual and theoretical spread, and that the variance ratios are close to unity. Hurn et al. (1996) using monthly LIBOR rates (1975-1991) find the VAR cross-equation restrictions are not rejected, while Cuthbertson (1996a) rejects these. Our results on Irish data are more informative than those of Hurley (1990), because of our longer and more accurate data set and the wider array of tests used. We find much more support for the EH than does Hurley (1990).

VI CONCLUSIONS

Using a number of short-term maturities on monthly Irish money market rates from 1982 to 1997, we perform a number of tests of the EH of the term structure of interest rates for Ireland. On balance our results would appear to lend support to the EH. Using the Johansen procedure on the 1-month, 3-month and 6-month interest rates we find that the cointegrating vector between any pair of interest rates is $\{1, -1\}$ — this provides a “weak” test in favour of the EH. A regression of the future change in short rates (i.e. the perfect foresight spread) on the actual spread gives a coefficient of unity on the latter variable, which is again consistent with the EH. Turning to the bivariate VAR we find that the forecast of future changes in short rates (i.e. the theoretical spread) moves closely with the actual spread. Both the standard deviations ratio and the correlation

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12. Campbell and Shiller (1991) use monthly US data from January 1952 to February 1987, and find little support for the EH at the short end of the maturity spectrum. The authors obtained slope coefficients $\beta$ ranging between 0 and 0.5. They do find beta coefficients of around 1 for maturities of 4.5 and 10 years.

13. Consistent with the results found in our study, Cuthbertson (1996a) finds Granger causality from $\Delta r_t^{(m)}$ to $S_t^{(n,m)}$ indicating substantial feedback in the VAR regressions.
coefficients for the theoretical spread relative to the actual spread give results in favour of the EH. However, the VAR cross-equation restrictions although favourable to the EH for the (6,3) and (6,1) interest rate combinations, are not so for the (3,1) combination. However, we do provide a number of reasons why the cross-equation restrictions may be rejected (including the small sample evidence in Bekaert et al. (1997) and Bekaert and Hodrick (2000)), even though the EH remains a good representation of the data. It is encouraging that our results are consistent with recent studies which also focus on the short end of the maturity spectrum (e.g. Cuthbertson (1996a) for the UK and Engsted (1994) for Denmark).

REFERENCES


Irish Case”, *The Economic and Social Review*, Vol. 22, No. 1, pp. 25-34.