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## DIFFERENTIAL EQOATION MODEL

FOR YIELD CURVES

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Abstract: This paper examines a differential equation model, whose solutions have yield curve shapes. The paper also examines the relevance of the model with historical monthly U. S. Treasury nominal rates.

In this paper $m$ will represent the maturity of a yield curve and $i(m)$ will denote a yield curve, of interest rates $i$.

In the study of yield curve behavior, yield curve models should have at least the the following list of properities:
(1) Short term interest rates should be more volatile than long term rates. See [11]. That is, the short term rates have little influence on the later maturity yield values.
(2) Yield curves should converge to finite (and reasonable) values at infinity and zero. see [17]. Reasonable here will mean interest rates that range from $0.5 \%$ to $30 \%$.
(3) Yield curves should be continuous over all positive maturities.
(4) Yield curves should be bounded over all positive maturities, and again be reasonable as defined in (2).
(5) A specific interest rate at a specific maturity should be influenced by adjoining rates. This influence increases as the maturity $m$ increases. This property is modeled as a feedback component.
(6) A model of yield curves should replicate the model yield curves used by Jetton and Pedersen et al. See [8] and [12].
(7) A model of yield curves should correspond with historical yield curves.
(8) A model of yield curves over time should have high autocorrelation at lag one between the interest rates at infinity (and zero). otherwise stated, this period's interest rate should be closely related to last period's rate and to next period's rate.
(9) Ninety day interest rates should be highly correlated with the zero maturity interest rate.

Properties (2), (3), (4) will be restated as:
(4') Yield curves should be continuous and bounded over all nonnegative maturities (including infinity), and the bound should be reasonable.

Several models of yield curves have been studied. An incomplete list of them follows:

## I Polynomial fit of yield curves

The benefit of this type of model is that you will have a model that is reproducing, but it will not have a finite value at infinity. See [17].

## II Sieqel's truncated model of yield curves

Here the yield curve is approximated by a form $b+c / m$, Where $b$ is the interest rate at infinity and $c$ is a constant. See [17]. This model does satisfy property (4') except at zero, and it satisfies property (1). This model even replicates the yield curves in [8]; however, this model does not replicate the special humped yield curve mentioned in Figure 16 of [12].

III power series in $(1 / m)$, and Siegel general model
The Siegel general model approximates yield curves by
$b+c / m+f(m)$, where $f(m)$ tends to zero faster than $1 / m$. See [17]. It appears that power series in (1/m) form a subset of Siegel general model. If $f(m)$ is not continuous or bounded at a positive maturity other than infinity, the siegel general model will not satisfy (4'). Also $f(m)$ must be carefully chosen to replicate the humped yield curve of [12].

## IV Rational approximation of yield curves

Here a yield curve is represented as a quotient of two polynomials in $m$, say $P(m) / Q(m)$. Prior research using an $N$ point Pade approximate method called the Thiele Reciprocal Difference Method (see [3]) generated a collection of Pade approximates of the monthly U.S. Treasury nominal yield curves. These approximations were reproducing, however one was not able to guarantee that the result was continuous and bounded over the nonnegative reals. This was especially a problem on yield curves that were almost flat. Possibly there are other algorithms other than the Thiele algorithm that will allow control over the denominator $Q(m)$.

Also, the rational approximation converged to a finite value at infinity only when $P(m)$ and $Q(m)$ had the same degree. This occurs only when there are exactly an even number of known points. See [3]. When there are an odd number of points, ane can guarantee a finite value at infinity by an introducing an additional element obtained by interpolation.

## $V$ Differential Equation model of yield curves

The remaining part of the paper will examine the family of coupled linear differential equations of the following form

$$
\begin{align*}
& \frac{d i}{d u}+\epsilon i=\epsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right)  \tag{0.1a}\\
& \frac{d m}{d u}=a+h(\ln (m+q))
\end{align*}
$$

where $m$ represents the maturity, $i(m)$ represents a yield curve and $u$ is a parameter. This model is not necessarily the simplest model, and may very well be a subset of the siegel general form. It is not even reproducing as in the polynomial or rational approximation methods, but it does attempt to address the yield curve properties (1) through (9). Also note the right hand side of (0.1a) can written as

$$
\epsilon^{\prime}\left(\frac{\left(m-a_{1}\right)\left(m-a_{2}\right)}{\left(m-a_{3}\right)\left(m-a_{4}\right)}\right.
$$

so the six parameters can be reduced to five. However, we will
continue to use the six parameter model, because of the interpretation of the $b$ coefficent that follows.

Differential equations ( 0.1 a ) and ( 0.1 b ) were obtained by using various formulas and numerically solving the equations. These solutions were then graphed and examined. This approach is called geometric modeling. See [2] for further discussion and references on geometric modeling.

One can reduce ( $0.1 a$ ) and ( 0.1 b ) to the following form

$$
\begin{equation*}
\frac{d i}{d m}+\epsilon i r(I)=\epsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right) r(I) \tag{0.1}
\end{equation*}
$$

where $r(m)=1 /(a+h(\ln (m+q)))$.
Since (0.2) is a linear differential equation of the form

$$
\frac{d i}{d m}+P(m) i=Q(m)
$$

the general solutions are of the form

$$
\begin{equation*}
i(m)=\frac{\int Q(m) \exp \left(\int P(m) d m\right) d m}{\exp \left(\int P(m) d m\right)}+c \tag{0.3}
\end{equation*}
$$

See [13]. When attempting to solve (0.1), the integration of the numerator in (0.3) is prohibitive. To simplify a Pade [1/1] rational approximation to the function $r(m)$ is used. This resulted in the approximation $I(m)=(j m+k) /(m+p)$, where $j, k, p$ are constants. See [4] for details to make this rational approximation. With these restictions and approximations (0.1) becomes:

$$
\begin{equation*}
\frac{d i}{d m}+\epsilon i \frac{(j m+k)}{(m+p)}=\epsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right) \frac{(j m+k)}{(m+p)} \tag{0.4}
\end{equation*}
$$

In Appendix I, an analytic solution of a restriction of (0.4) will be derived with the use of (0.3). This derivation shows that the b coefficent above in (0.4) corresponds to the interest rate at infinity.

Now if

$$
\begin{equation*}
\frac{b m^{2}+c m+d}{m^{2}+f m+g} \tag{0.6}
\end{equation*}
$$

is expressed as a power series in (1/m) by an argument similar to Appendix $I_{\text {, }} i(m)$ will be a power series in (1/m). But by expressing the model as a rational expression of two second degree polynomials, one can easily control nonsingularity in the differential equation. By controlling noningularity one can assure the continuity of the solutions over the nonnegative reals. This is because if a derivative of a function exists at a point the function is continous at that point. However this does not allow one to assume that the values are bounded or reasonable as stated in property (4) above.

To control nonsingularity, chose a value $p$ that is positive and $r(m)$ will be well defined on all nonnegative $m$. To assure (0.6) will be nonsingular observe the following argument:

$$
\begin{aligned}
& \text { First assume } f>0, \text { and } 0<g<f^{2} / 4 \text {. So } \\
& 0<4 g<f^{2} \\
& 0<f^{2}-4 g<f^{2} \\
& 0<\sqrt{f^{2}-4 g}<f \\
& -f+\sqrt{f^{2}-4 g}<0 \\
& \frac{-f+\sqrt{f^{2}-4 g}}{2}<0
\end{aligned}
$$

This shows the right most root of $\mathrm{m}^{2}+f m+g$ is less than zero.
Next assume that $f \leq 0$, and $0 \leq f^{2} / 4<g$. So

$$
\begin{aligned}
& 0 \leq f^{2}<4 g \\
& f^{2}-4 g<0 \\
& \sqrt{f^{2}-4 g} \text { is not real. }
\end{aligned}
$$

This shows that $m^{2}+f m+g$ has no real roots. So by placing either of these restrictions on $f$ and $g$, and assuming $p$ is strictly positive, we can guarantee nonsingularity in (0.4).

To guarantee that the solutions to (0.4) are bounded, one must use the union of two arguments. By expressing (0.6) as a power series in $(1 / m)$, one can show that the solution is finite and bounded by an identical argument in Appendix $I$. However, to show the solutions are bounded at and near zero will require one

$$
\begin{equation*}
\frac{d i}{d m}=-\varepsilon i \frac{(j m+k)}{(m+p)}+\varepsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right) \frac{(j m+k)}{(m+p)} \tag{0.7}
\end{equation*}
$$

Note if the choice of $p, f$, and $g$ have been chosen as above, to control nonsingularity, this will guarantee that the partial derivative of the right hand side of (0.7) with respect to is continuous in a neighborhood of zero. This allows one to conclude by the existence theorem of order one differential equations, that the solution is bounded. See [14].

Also, one is be able to obtain the humped yield curve, by forcing the derivative of $i$ with respect to $m$ to be zero at a certain point, possibly allowing one to generate a humped yield curve. For instance, if the interest rate at maturity $m$ is a local maximum, you can force the value of $d$ to equal

$$
\begin{equation*}
i(m)\left(m^{2}+f m+g\right)-b m^{2}-c m \tag{0.8}
\end{equation*}
$$

and this will assure $i^{\prime}(m)$ is zero.
The concept of feedback is modeled as a system that takes a current value of the solution and multiplies it by a constant and places that value back into the system. This is usually written as a differential equation of the form:

$$
\frac{d y}{d x}+c y=E(x, y)
$$

As one can see (0.4) has this form. So property (5) is satisfied.

If one multiplies ( 0.4 ) through by $(m+p)$, the following is obtained

$$
\begin{equation*}
(m+p) \frac{d i}{d m}+\epsilon i(j m+k)=\epsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right)(j m+k) \tag{0.9}
\end{equation*}
$$

Note, that for maturities near $-p$, the effect of the derivative diminishes, and this allows for a greater volatility of a curve's values near that point. So if $p$ is a positive number near zero, property (1) is satisfied.

The coefficent $\epsilon$, has two major influences on the solutions of (0.4). First, observe what occurs when $\in=0$. This produces the simple differential equation

$$
\frac{d i}{}=0
$$

which has constant solutions $i(m)=c$. This corresponds to the flat yield curve. Secondly, $\epsilon$ is used to guarantee that solutions of (0.4) are constrained. Here $\epsilon$ will be used in the example that follows to guarantee the reasonableness of the solutions.

The other coefficents $c, d, f$, and $g$ do not have an interpretation, within the model.

The differential equation method has no means to guarantee that property (8) or (9) holds. All that one can hope is that these properties will hold after modeling. This will be addressed in the historical study in section II.

The following section will discuss the method of solution that is taken to examine the historical yield curves.

## I Boundary Value Problem and Multistage Monte Carlo Algorithm


#### Abstract

Solving for exact solutions to ( 0.4 ) is called boundary value problems. However boundary value problems for linear differential equations, require matching the solution to at most two points on the curve. For the historical model one desires an exact fit with every point on a given yield curve, i.e. reproducing. Since the analytical solution is a complex infinite series in ( $1 / \mathrm{m}$ ), it is prohibitive to attempt this approach. So, the approach was to check for historical validity by minimizing $$
\begin{equation*} \sum_{m \in I}\left|i_{m}-i_{m}\right| \tag{1.1} \end{equation*}
$$


where the index set $I$ is the set of all known maturities for the historical yield curve $i_{m}$. Here $i_{m}$ would be the corresponding values generated by solutions to (0.4). The sum of absolute differences, instead of least squares was used, to reduce the influence of outliers. See the discussion in [10] about this issue.

The equation (0.4) is arbitarily restricted to the following model for the historical study:

$$
\begin{equation*}
\frac{d i}{d m}+\epsilon i \frac{(2 m+2)}{(3 m+2)}=\epsilon\left(\frac{b m^{2}+c m+d}{m^{2}+f m+g}\right) \frac{(2 m+2)}{(3 m+2)} \tag{1.2}
\end{equation*}
$$

This assures that (1.2) is nonsingular at zero. This also allowed the problem to determine only six coefficents instead of
nine. So this assumes that there is an universal $r(m)$ for a specific family of yield curves.

The approach taken to provide coefficents to (0.4) was to use the Multistage Monte Carlo Optimization algorithm (see [6] and [7]) to select the coefficents, making sure that the choice of $f$ and $g$ controls nonsingularity. Also, when the interest rate at one year exceeds the surrounding two rates, the coefficent $d$ is constrained by equation (0.8) to model the historical humped yield curves.

Once the coefficents are determined, the four point RungeKutta algorithm is used to generate the numerical solution to the specific differential equation. See [5]. The starting point for the numerical solution is at the highest maturity known for the historical curve which was 20 for the curves from January 1, 1970 to Febuary 1, 1977, and 30 for all subsequent curves. The step size is set at 0.25 , so the solutions pass through the maturities of the indexed set 1 .

For each yield curve, there is between 15000 to 75000 random samples taken to generate the optimal solution.

See the Appendix II for a copy of the Quick Basic program that was used for the study.

## II Historical Results

The historical yield curves are the nominal monthly Treasury curves from January 1970 through December 1990. See [18] and [19]. There were 252 yield curves modeled. Table 1 indicates which rates are available. This table also describes the indexed set I.

| Date | $\begin{aligned} & 90 \\ & \text { day } \end{aligned}$ | $\begin{aligned} & 1 \\ & \mathrm{yr} \end{aligned}$ | $\begin{aligned} & 2 \\ & \mathrm{yr} \end{aligned}$ | $\begin{aligned} & 3 \\ & y r \end{aligned}$ | $\begin{aligned} & 5 \\ & y r \end{aligned}$ | $\begin{aligned} & 7 \\ & y r \end{aligned}$ | $\begin{aligned} & 10 \\ & Y \mathrm{r} \end{aligned}$ | $\begin{aligned} & 20 \\ & y r \end{aligned}$ | $\begin{aligned} & 30 \\ & \mathrm{yr} \end{aligned}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & 1 / 1 / 70 \\ & 2 / 1 / 77 \end{aligned}$ | $\mathbf{x}$ | x | NA | x | X | NA | $\mathbf{x}$ | X | NA |
| $\begin{aligned} & 3 / 1 / 77 \\ & 3 / 1 / 84 \end{aligned}$ | x | x | NA | x | x | NA | $\mathbf{x}$ | x | X |
| $\begin{array}{r} 4 / 1 / 84 \\ 12 / 1 / 86 \end{array}$ | $\mathbf{x}$ | X | x | x | X | x | x | X | X |
| $\begin{aligned} & 1 / 1 / 87 \\ & \text { current } \end{aligned}$ | $\mathbf{x}$ | x | x | $\mathbf{x}$ | X | x | x | NA | x |

TABLE 1

The error term (1.1) for the yield curves has a mean of 0.00211 , and a standard deviation of 0.00113 . Skewness is 0.406 , and kurtosis is 2.849 . The lowest value for ( 1.1 ) is 0.0002 , and the highest is 0.00601 . A frequency distribution of the error is in Figure 1. Assuming that the error term fits a gamma distribution, using the mean and variance above one obtains an estimate for $a$ and $\beta$. These are 3.441 and 1634.4 respectively. A gamma distibution with these parameter values for $\alpha$ and $\beta$ has $5 \%$ of its area above the value of .00425 . This model has 3.57\% above this value. So assuming a gamma distribution as the hypothesis, the hypothesis is accepted at 5\%.


Figure 1

Treating the sum

$$
\begin{equation*}
\sum_{m \in I} i_{m}-i_{m} \tag{4.1}
\end{equation*}
$$

as another error term, it has mean -0.00016 , standard deviation of 0.00139 , skewness of -0.264 and kurtosis of 3.648. The minimum value of (4.1) is -0.04727 , and the maximum is .003749. A frequency distribution of the error is in Figure 2. This error term appears to fit a normal distribution with the above sample statistics. A hypothesis that this error is normal is rejected at the 57 level because there are 15 out of 252 terms outside the second deviation from the mean. So $6 \%$ of the sample are outliers. Now the hypothesis can be accepted at the if level because 0.8 of the sample are outliers from the third deviation.

Let $z$ be the interest rate at maturity zero, and $q$ be the interest rate at 90 days. The statistics of the coefficents are in table 2.

Note the high autocorrelation for the 2 . This demonstrates property (8) for the zero maturity. However for the interest


Figura 2

| mean | standard deviation | skewness | kurtosis | $\min$ | $\max$ | lag one autocorrelation coefficent |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| b 0.0951 | 0.0289 | 1.462 | 8.753 | 10.0301 | $0.262 \mid$ | 0.3692 |
| c 0.1203 | 0.4540 | -2.495 | 18.071 | 1-3.33 | 1.321 | 0.1875 |
| d -0.0236 | 1.0283 | -0.673 | 8.39 | $\|-4.94\|$ | 3.4701 | 0.0661 |
| $\pm 2.0060$ | 2.998 | 0.108 | 1.684 | $\|-3.51\|$ | $7.824 \mid$ | 0.5515 |
| g 2.5250 | 1.788 | 0.608 | 2.366 | 0.031 | $7.182 \mid$ | 0.3488 |
| e -0.6575 | 0.8265 | -1.231 | 3.531 | $\|-3.20\|$ | $0.184 \mid$ | 0.0983 |
| q 0.0758 | 0.0266 | 1.105 | 4.104 | 10.032 | 0.1631 | 0.9658 |
| 20.0693 | 0.0260 | 1.181 | 4.694 | 10.0201 | 0.1621 | 0.8696 |

## TABLE 2

rate $b$ at infinity, the autocorrelation coefficent is disappointing. This may be due the fact that the approximation to $5($ m $)$ is not optimal. Also, it may be improved by increased computer runs to reduce the randomness. This may alsa be solved by modifying the differential equation ( 0.4 ) to clamp the values at infinity.

The correlation between $q$ and $z$ above is found to be 0.882 . So, property (9) is satistied.

The joint distribution of these coefficents through time will be the content of a future paper.

As one can see, the use of differential equations can aid one in the process of fitting data. Even though the process is somewhat prohibitive in the fact that random Monte Carlo techniques must be used to solve for various coefficents in the differential equation, there is an increased control over the process, where one can restrain the solutions to take on behavior that best satisfies the data. This paper showed that properties (1) through (9) were satisfied with the model (except for property (8) at infinity). Hopefully some of the concepts of this research may find uses in other topics such as graduation and other statistical fitting problems.

## Other topics for future study are as follows:

(1) Find the optimal choice of the linear rational term $r(m)=(j m+k) /(m+p)$. Possibly this will clear up the disappointing result concerning property (8) at infinity, and improve other results.
(2) Determine the marginal and joint distributions of the coefficents $b, c, d, f, g, \epsilon$, and $z$ through time. If this is completed, one may be able to use (0.4) in scenario generation as in [9]. The use of these distributions, may also lead to futher analysis of ( 0.4 ) as a random differential equation. There is the possibility that this model may have influence on multidimensional immunization as referred in [15] and [16].
(3) Improve the historical accuracy of this model, by examining the behavior of daily yield information.
(4) Determine a method to test for a coefficent's relevance in a differential equation by some statistical measure.
(5) Examine the model (0.4) for bifurcation as in [2] and for chaotic behavior. The author is especially interested in chaotic behavior of ( 0.4 ) possibly being exhibited historically in the early 1980's. This study may result in chaotic indicators to help prepare the industry for extreme swings in the value of short term rates.

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## APPENDIX I

Following is the proof that the differential equation

$$
\begin{equation*}
\frac{d i}{d m}+\frac{\epsilon i(j m+k)}{m+p}=\epsilon(b) \frac{(j m+k)}{m+p} . \tag{A.1}
\end{equation*}
$$

has a solution of the form $b+O(1 / m)$.
The proof of (A.1) will allow us to demonstrate, that as m approaches infinity, the solutions will approach a constant function, as referred to in [17].

Proof. Now using equation (0.3),

$$
\begin{equation*}
P(m)=\frac{\epsilon(j m+k)}{m+p} \tag{A.2}
\end{equation*}
$$

and

$$
\begin{equation*}
\exp \left(\int P(\mathbb{m}) d m\right)=\exp (j \in \mathbb{m})(m+p)(k-j p) \epsilon \tag{A.3}
\end{equation*}
$$

Now $Q(m)$ corresponds to

$$
\begin{equation*}
\epsilon b \frac{(j m+k)}{m+p} \tag{A.4}
\end{equation*}
$$

So
 the first term. Now expand $(m+p)\{(k-j p) \epsilon-1)$ by the binomial expansion theorem one obtains the following formula

$$
j b \epsilon \int m \exp (j \epsilon m)\left[(m)^{(k-j p) \epsilon-1}+((k-j p) \epsilon-1) p(m)^{(k-j p) \epsilon-2}\right.
$$

$$
+\ldots] \mathrm{dm}
$$

Discarding the $((k-j p) \epsilon-2)$ th and higher terms of $m$, we obtain

$$
\begin{align*}
& j b \epsilon \int m \exp (j \epsilon \mathbb{I})\left[(m)^{(k-j p) \epsilon-1}\right] d m \\
& =\quad j b \epsilon \int m^{(k-j p) \epsilon} \exp (j \epsilon \mathbb{I}) d m \tag{A.9}
\end{align*}
$$

Using integration by parts, (A.9) reduces to

$$
\begin{align*}
& \int Q(m) \exp \left(\int P(m) d m\right) d m  \tag{A.5}\\
& =\int \frac{b \epsilon(j m+k)}{m+p} \exp (j \epsilon m)(m+p)^{(k-j p) \epsilon} d m \\
& =j b \epsilon \int{\underset{j}{j}}_{(m+k) \exp (j \epsilon m)(m+p)}^{(k-j p) \epsilon-1} d m \quad \text { (A.6) } \\
& =j b \epsilon \int(m) \exp (j \epsilon m)(m+p)^{(k-j p) \epsilon-1} d m  \tag{A.7}\\
& +\quad \mathrm{kb} \epsilon \int \exp (\mathrm{j} \epsilon \mathrm{~m})(\mathrm{m}+\mathrm{p})^{(\mathrm{k}-\mathrm{jp}) \epsilon-1} \mathrm{dm}
\end{align*}
$$

$$
\begin{aligned}
& j b \epsilon\left[\frac{m^{(k-j p) \epsilon}}{(j \epsilon)} \quad \exp (j \in m)\right. \\
& -\frac{((k-j p) \epsilon) \mathrm{m}^{(k-j p) \epsilon-1}}{(j \epsilon)^{2}} \exp (j \epsilon \mathrm{I}) \quad \text { (A.10) } \\
& +\ldots] \\
& =j b \epsilon \exp (j \epsilon m)\left[\frac{1}{j \epsilon} m^{(k-j p) \epsilon}\right. \\
& -\frac{(k-j p) \epsilon}{(j \varepsilon)^{2}} \quad m^{(k-j p) \epsilon-1} \\
& +\cdots] \\
& \text { Expanding (A.3) by the binomial theorem we obtain } \\
& \exp (j \in m)\left[m^{(k-j p) \epsilon}+((k-j p) \epsilon) p m(k-j p) \epsilon-1\right. \\
& +((k-j p) \epsilon)((k-j p) \epsilon-1) p^{2}{ }^{(k-j p) \epsilon-2} \\
& +\ldots]
\end{aligned}
$$

$$
\begin{equation*}
j b \epsilon\left[\frac{1}{j \epsilon}+o(1 / m)\right]=b+0(1 / m) \tag{A.12}
\end{equation*}
$$

Since in each yield curve $\epsilon$ is fixed. The linear term above dominates when m approaches infinity. By a similar argument if $Q(m)$ was replaced by

$$
\begin{equation*}
\epsilon(b+c / m) \frac{(j m+k)}{m+p} \tag{A.4b}
\end{equation*}
$$

the solution would be of the form

$$
\begin{equation*}
b+\frac{c^{\prime}}{m}+o\left(1 / m^{2}\right) \tag{A.12b}
\end{equation*}
$$

In fact, if a power series of ( $1 / \mathrm{m}$ ) is used one will obtain a power series of ( $1 / \mathrm{m}$ ) with the leading constant term the same.

```
DECLARE SLB RK4 (W!(), W!, H!, X!, A!, E!, Cl, D!, F!, Gl, E!, Q!, R1, S!, L!())
DECLARE FUWCTIOW XDEFZ! (XI, Y!, A!, El, CI, D!, F!, G!, E!)
DECLARE FUNCT!ON XDEFI! (X!)
OIM W(2)
SCREEN 2
1 the variable xx is the treas.oat fon a specifie yiELD durve
DIM xx(9, 2)
I the variable table gives the fielos tMAt parse as in the treas.dat
DIM TABLE(9, 2)
TABLE(1, 1)=7: TABLE(1, 2)=12
TABLE(2, 1) = 13: TABLE(2, 2) = 18
TABLE(3, l) = 19: TABLE(3, 2) = 24
TABLE(4, 1) = 25: TABLE(4, 2) = 30
TABLE(5, 1) = 31: TAgLE(5, 2) = 36
TABLE(6, 1) = 37: TABLE(6, 2) = 42
TABLE(7, 1) = 43: TABLE(7, 2) = 48
TABLE(B, 1) = 49: TABLE(8, 2) = 54
TABLE(9, 1) = 55: TABLE(9, 2) =60
'VARIblES O,R,5 aRE USED TO PROPERLY PLOT ON TME SCREEN a SOLUTION OF DE
O = 20
R=100
S = 100
```

OPEN COMMANDS FOR IMPUT AS :4
TOP:
GET A YIELD CURVE
gosue freas
DDS = LEFTS(AS, 6)

```
SET UP SIEVE for rumge dutta saSEd dm umether 30 year or 20 year is the
'laSt yIELD CURVE valuE. THE SIEvE MUST BE DESIGNED 50 TO PASS THROUGH
KNOU YIELD CLRVE MATURITIES, TO TEST THE FIT OF THE SOLUTION TO THE
'ACTUAL YIELD CURVE.
IF XX(9, 2) = 0 THEM
    N=20 * 4
    INTERVAL = 20
ELSE
    N=30 * 6
    INTERVAL * 30
END IF
'RUNGE KUTYA SOLUTION [S PLACED INTO THE VARIABLE L
DIM L(M, 2)
DIM PL(N, 2), OX(5)
```

```
OIM A(7), N(7), x(7), B(7)
OIM LL(7), AU(7)
KOR = INSTR(COMTANOS," M)
IF KOR = O THEN KOR=LEM(COMMANOS)
XRES=LEFTS(COMMANDS,KOR)
```

SCRITS=(TRIMS (RIGHTS (COMANOS, LEN(COMHANDS)-KOR))
OPEN "R" - XRES FOR APPEIN AS *2
PRINT \#2, "NEWTEST6 " * XRES
CLOSE

KXXRRE:

MOLD $=.19$
CEILING = 25
$x(1)=0$
IF OX(2) $=0$ THEN CEILIMG $=10$
IF OX(1) $=0$ THEN CEILJNG $=5$

FOR X8 = i TO CEILING
CLS
RANOMMIEE TIMER
IF SCR11Sx*M THEN PRINT AS; WMEWTESTS * + XRBS
$m=.19$
MERR $=0$

If $\times \times(9,2)=0$ THEN
$B(1)=0 * 0 x(2)$
$A(1)=0$
$N(1)=0 * Q \times(2)$
$8(2)=.015$
$A(2)=x \times(8,2)$
$N(2)=.40$
$\theta(3)=-8 * \operatorname{cx}(3)$
$A(3)=0$
$W(3)=8 \cdot \operatorname{OX}(3)$
$B(6)=-8 * 0 \times(4)$
$A(6)=0$
$M(6)=8 * \operatorname{OX}(4)$
$\theta(5)=-8 * 0 \times(4)$
$a(5)=0 \cdot a x(4)$
$w(5)=8 *$ ox(6)
$B(6)=0 * \operatorname{ex}(5)$
$A(6)=4 \cdot \operatorname{ex}(5)$
$N(6)=8$ * OX(5)
$\theta(7)=-6$
$A(7)=0$
$N(7)=4$
END JF
If $X X(8,2)=0$ THEN
$B(1)=0 \cdot O \times(2)$
$A(1)=0$
$N(1)=0 * \operatorname{ex}(2)$
$B(2)=.015$
$A(2)=x \times(9,2)$
$N(2)=.40$
$8(3)=-8 * \operatorname{ex}(3)$
$A(3)=0$
$N(3)=8 * \operatorname{ax}(3)$
$B(4)=-8 * 0 x(6)$
$A(4)=0$
$W(4)=8 * \operatorname{OX}(4)$
$B(5)=-8 * \operatorname{ax}(4)$
$a(5)=0 \quad \operatorname{ax}(4)$
$W(5)=8=\operatorname{ax}(6)$
$B(6)=0$ ex(5)
$A(6)=4 * \operatorname{ax}(5)$
$N(6)=\operatorname{ex}(5)$
$B(7)=-8$
$A(7)=0$
$N(7)=8$
ENO :F
If $X X(8,2) 《 0$ AND $X X(9,2) \ll 0$ THEN
$B(1)=0 * \operatorname{ex}(2)$
$A(1)=0$
$M(1)=0$ - $0 \times(2)$
$B(2)=.015$
$\mathrm{A}(2)=\mathrm{x}(9,2)$
$M(2)=.60$
$B(3)=-8 \quad \operatorname{cx}(5)$
$A(3)=0$
$N(3)=8 * \operatorname{ax}(3)$
$B(4)=-8 * 0 x(4)$
$A(4)=0$
$M(4)=8 * 0 \times(4)$
$B(5)=-8 * 0 \times(4)$
$A(5)=0 \cdot \operatorname{ex}(6)$
$M(5)=8$ - ax( 6 )
$B(6)=0$ ax(5)
$A(6)=4$ Qx(5)
$N(6)=8$ e $\operatorname{ex}(5)$
$B(7)=-4$
A(7) $=0$
$\mathrm{M}(7)=6$
END IF
MTRS:

FOR $J=11015$
$Z=40$
IF M < . 003 THEM $2=100$
If $M>.01$ ANO $\perp>13$ THEN $2=10$
FOR 1 = 1 IO 2
TRIES $=0$
RETRY:
TRIES $=$ TRIES +1
FORK $x 2$ 10 7
IF (A(K) - $M(K) /(2) J \leqslant B(K))$ THEN $L(K)=B(K)$
ELSE
$L(X)=A(X)=N(X) / 2 d$
Emo if
IF (A(K) + $N(K) / 2 \quad J>M(K))$ THEM $A U(K)=W(K)-L L(K)$
ELSE

```
        MU(K) =A(K)+M(K)/2 J-LL(K)
    ENO IF
    L = RND
    x(K)=LL(K) * U* AU(K)
MEXT K
    IF TRIES > 150 TMEN GOTO ROUTE
    If }x(5)<0\mathrm{ ND X(6)<EX(5) 2/6 THEN COTO RETRY
    If }X(5)>=0\mathrm{ and }X(6)>=X(5)2/4 TMEN COTO RETR
    IF XX(3,2) O D TNEW
    IF XX(2,2) > XX(1,2) AND XX(2,2) > XX(3,2) TMEN
        x(6) = xx(2,2)*(1+x(5)+x(6)) - x(2) - x(3)
        END IF
        ELSE
    IF xx(2,2) > XX(1,2) AND XX(2,2) > XX(4,2) THEN
        x(4) = xx(2,2)*(1+x(5)+x(6)) - x(2) - x(3)
        ExO If
        END IF
-start
'h determines the step size of the rX algorithm
H = - [nterval / M
'let the starting value for the rk algorithm to de tme yield curve
ivalue for the 1 yenr maturity.
1F XX(9,2)=0 THEN
xxx = 20
u(2) = xx(8, 2)
ELSE
xxx = 30
W(2) = xx(9, 2)
END :F
L(0, 1) = XXX: L(0, 2) = W(2)
If SCRITS =*H THEM
locate 2, s
PRINT USING".4.*)}x(1); x(2); x(3); x(6); x(5); x(6); x(7
END IF
CALL RKL(W(), N, H, XXX, X(9), X(2), X(3), X(4), X(5), X(6), x(7), a, R, S, L())
'accumulate the variable ssum to test to see mow far tme actual
'YIELO CURVE IS AMAY from the de solution.
'NOTE: USED ABS IMSTEAD OF LEAST SOLARE.
    SSUN =0
    CERR = 0
IF SCRITS=WM THEN
LOCATE S, 1
```



```
END If
    fOR lx = O TO M
- this select case is set up to properly dbtaim the ssum variable
    select case l(IX, 1)
        case .zs
            If XX{1, 2) O O THEM
                SSUM = SSUK * ABS(L(1X, 2) - KO(1, 2))
                CERR = CERR - (L(1X, 2) - XX(1, 2))
            EMD IF
        CASE 1
```

```
            If XX(2, 2) > O THEN
                    SSUM = SSUM + ABS(L(IX, 2) - XX(2, 2))
                    CERR = CERR + L(IX, 2) - XX(2, 2)
            END 1F
        CASE 2
            IF xX(3, 2) O O them
            SSMM = SSUM * ABS(L(IX, 2) - xX(3, 2))
            CERR = CERR + (L(1X, 2) - KX(3, 2))
            ENO IF
        CASE 3
        IF xX(4, 2)>0 Them
        SSLM = SSLN - AES(L(IX, 2) - MX(4, 2))
        CERR = CERR + (L(1X, 2) - XO(4, 2))
        ENO IF
        CASE 5
        IF XX(5, 2) <> 0 TMEN
        SSLM = SSLM + ABS(L(1X, 2) - xX(5, 2))
        CERR = CERR + (L(IX, 2) - XX(5, 2))
        EmO IF
        CASE 7
        If xX(6, 2) <> O THEN
        SSUM = SSUM * ABS(L(1X, 2) - XX(6, 2))
        CERR = CERR + (L(IX, 2) - XX(6, 2))
        END :F
        case 10
        IF xx(7, 2) <> 0 THEN
        ssum = Ssum + ABS(L(IX, 2) - xx(7, 2))
        CERR = CERR + (L(IX, 2) - XX(7, 2))
        ENO :F
        CASE 20
        IF XX(B, 2) O O THEN
        SSUM = SSMm + AES(L(1X, 2) - xX(%, 2))
        CERR = CERR + (b(1X, 2) - XX(B, 2))
        END IF
        CASE 30
        1F XX(9, 2) > 0 them
        SSUM = SSUM + ABS(L(1X, 2) - XX(9, 2))
        CERR = CERR + (L(IX, 2) - XX(9, 2))
        ENO IF
        case else
    end select
    mEXT IX
```

```
PF = SSUM
```

PF = SSUM
If PF >= M THEN COTO LOCS
If PF >= M THEN COTO LOCS
M=PF
M=PF
MERR = CERR
MERR = CERR
fOR N1 = { rO 7
fOR N1 = { rO 7
A(N1) = X(N1)
A(N1) = X(N1)
mexy w1
mexy w1
2ED = 2(N,2)
2ED = 2(N,2)
If MOLO > M THEN MOLD = M

```
If MOLO > M THEN MOLD = M
```

LOC3: MEXT 1
MEXT d
CPEN MR" + xRES FOR APPEID AS $\$ 2$



CLOSE 12
IF MOLD \& . COOS TMEM EOTO COTOF HERE
ROUTE: MEXT KB
1F MOLD * . 002 THEN
IF OX(1) - 0 THEN
ax(1) - 1
coto kocres
EmD if
IF OX(2) - 0 TXEM
ox(2) - 1
COIO XOCRR
EnD If
If ax(3) - 0 THEM
ox(3) $=1$ coto worral
END If
IF OX(6) = D Them
$a x(4)=1$
coto coorrs
EnP IF

END IF

OUTOFHERE: COSUE TREAS
DDS = LEFTS (AS, 6)
COTO KXXRRE

TREAS:
ax(1) $=1$
$a x(2)=1$
$0 \times(3)=1$
$0 \times(4)=1$
$0 \times(5)=1$
If EOF (l) THEN COIO EMDUP
LIME IMPUT *1, AS
FOR $1=1$ TO


SELECT CASE I
CASE 1
x×(1, 1) $=.25$
case 2
x×(1, 1) =1
case 3
$x \times(1,1)=2$
CASE 4
$\mathrm{xx}(1,1)=3$
CASE 5
$\mathrm{xx}(1,1)=5$
CASE 6
$\mathrm{xx}(1,1)=T$
Case 7
$\mathrm{xx}(1,1)=10$
CASE 8
$\mathrm{xx}(1,1)=20$
CASE 9
$\mathbf{x x}(1,1)=30$
end select
mext 1
RETURN

ENDUP: CLOSE \#1
Eno

SUB RX4 (HC), N, H, X, A, B, G, D, F, G, E, D, R, S, L())
DIM $K(4,2)$
START $=X$
for : = 1 10 N
$K(1,2)=H * \operatorname{DEF} 2(X, U(2), A, B, C, D, F, G, E)$
$K(2,2)=N * \operatorname{XDFZ}(X+1 / 2 * H, U(2)+1 / 2 * K(4,2), A, B, C, 0, F, G, E)$
$K(3,2)=H * \operatorname{XDEF2(X}+1 / 2 * H, H(2) * 1 / 2 * K(2,2), A, E, C, D, F, G, E)$
$K(4,2)=N=X D E F(X+N, H(2)+K(3,2), A, E, C, D, F, E, E)$
$W(2)=W(2)+(K(1,2)+2 * K(2,2)+2 * K(3,2)+K(6,2)) / 6$
$X=$ START $+1 * H$
( 1,1 ) $=x:(1,2)=W(2)$

MEXT I

END Sus

FUNCTION XDEFY ( X )
XDEFT $=.0001+\operatorname{LOG}(X)$
EMO FUNETIOM

FUNCTIOM XDEF2 (X, Y, A, B, C, D, $F, G, E)$
ORS $=X * X * F * X * G$
If ORS = 0 THEM GAS $=.0001$
XDEF2 $=-E *\left(Y-\left(E^{* *} x+c * x+0\right) /\right.$ ORS $) *\left(2^{*} x+2\right) /(3 * x+2)$

ENO FUNCTION

