

## Numerical modeling and Computational simulation of the SDE using C++ & FOTRAN 90 : Application in Fisheries management

S.Kapoor,

*Department of Mathematics,  
Indian Institute of Technology, Roorkee,  
Roorkee (INDIA)*

D.Pandey

*Department of Mathematics,  
Pt . L.M.S. Govt P.G.College Rishikesh  
Rishikesh (INDIA)*

V.Dabral

*Department of Mathematics,  
Roorkee Institute of Technology, Roorkee,  
Roorkee (INDIA)*

### Abstract:

In the present paper an attempt is made for the solution of SDE (Stochastic differential equation) using different numerical simulation. Here the four different technique has been adopt for the two test problem for the verification process. Main emphasis is given on the RKM (Runge kutta Method) in which the solution has minimum number of absolute error .i.e more accurate then other. some of the solution has been given in the literature which adopt directly. The RKM (Runge kutta Method) code for the generalized IVP problem in C++ and FORAN 90 is also presented given for the sake of convince to understand the simulation process. The comparative study is also shows the accuracy of the RKM (Runge kutta Method) simulation process. The application of the specific problem is also found in the field of Fisheries Management

**Keyword:** SDE, RKM, simulation, IVP, Error

### Introduction

Stochastic differential equations (SDEs) have become standard models for financial quantities such as asset prices, interest rates, and their derivatives. Unlike deterministic models such as ordinary differential equations, which have a unique solution for each appropriate initial condition, SDEs have solutions that are continuous-time stochastic processes. Methods for the computational solution of stochastic differential equations are based on similar techniques for ordinary differential equations, but generalized to provide support for stochastic dynamics. We will begin with a quick survey of the most fundamental concepts from stochastic calculus that are needed to proceed with our description of numerical methods. For full details, the reader may consult [16, 20]. A set of random variables  $X_t$  indexed by real numbers  $t \geq 0$  is called a continuous-time stochastic process. Each instance, or realization of the stochastic process is a choice from the random variable  $X_t$  for each  $t$ , and is therefore a function of  $t$ . Any (deterministic) function  $f(t)$  can be trivially considered as a stochastic process, with variance  $V(f(t)) = 0$ . An archetypal example that is ubiquitous in models from physics, chemistry, and finance is the Wiener process  $W_t$ , a continuous-time stochastic process with the following three properties:

Property 1. For each  $t$ , the random variable  $W_t$  is normally distributed with mean 0 and variance  $t$ .

Property 2. For each  $t_1 < t_2$ , the normal random variable  $W_{t_2} - W_{t_1}$  is independent of the random variable  $W_{t_1}$ , and in fact independent of all  $W_t$ ;

Property 3. The Wiener process  $W_t$  can be represented by continuous paths. The Wiener process, named after Norbert Wiener, is a mathematical construct that formalizes random behavior characterized by the botanist Robert Brown in 1827, commonly called Brownian motion. It can be rigorously defined as the scaling limit of random walks as the step size and time interval between steps both go to zero. Brownian motion is crucial in the modeling of stochastic processes since it represents the integral of idealized noise that is independent of frequency, called white noise. Often, the Wiener process is called upon to represent random, external influences on an otherwise deterministic system, or more generally, dynamics that for a variety of reasons cannot be deterministically modeled. SDE (Stochastic Differential Equations) have many applications in areas like Biology, Ecology, Physics, Population Dynamics, Finance and Economics. Methods of numerical solution usually extend corresponding methods applying in the numerical solution of ordinary differential equations (ODE) and they are combined with simulation of the stochastic term. Some of the methods are explicit and some are implicit. Methods are also distinguished with respect to whether they are ‘strongly’ or ‘weakly’ convergent. For an introduction to the subject we refer to the books by Kloeden and Platen (1999), Gard (1988). Some survey papers also exist (Kloeden and Platen (1989)).

The type of stochastic differential equation considered in this paper is of the form:

$$dX(t) = f(x, X(t))dt + g(t, X(t))dW(t), \quad (1)$$

$$X(0) = X_0, \quad 0 \leq t \leq T, \quad (2)$$

Where  $f(t, X(t))$  and  $g(t, X(t))$  are scalar functions and the initial condition is a random variable. The coefficient  $f(t, X(t))$  is known as the *drift* and the coefficient  $g(t, X(t))$  is known as the *diffusion* coefficient.  $dW(t)$  denotes a standard Wiener process with

$$E(W(t)) = 0, E(W(s), W(t)) = \min(s, t). \quad (3)$$

When  $g(t, X(t))$  does not depends upon  $X$ , the SDE is known as one with *additive* noise and as one with *multiplicative* noise otherwise.

It is convention to rewrite the SDE in integral form:

$$X(t) = X_0 + \int_0^t f(s, X(s))ds + \int_0^t g(s, X(s))dW(s) \quad (4)$$

The integral term involving the diffusion coefficient is interpreted in the so called Itô or Stochastic sense. Numerical methods developed in the literature include: Euler-Maruyama, Adams, EM, linear multistep methods etc. For convergence proofs and the simulation of the stochastic terms of the equation, some stochastic calculus theory is also usually required and we refer to the books by Oksendal (1998), Gard(1988). Published software may be found in the books by Kloeden, platen and Schurz (1997) (Pascal), Cyganowski, Kloeden, Ombach (2001) (Maple). Some Matlab programs have been included in the paper by Higham (2001).

In this paper, we are going to study the numerical solution of the stochastic differential equations of the form (1) using Rk method. To explain some of the basics of the numerical treatment of the problem we also describe a C++ and F 90 Simulation for generalized IVP problem

### **Mathematical Models and Development**

Fishery is one kind of renewable resources, that means the resources can be regenerated in finite time. According to (Clark [4], p.81) the history of Mathematical methods in renewable resources may be retrospect to an early period: in 1931 a mathematical economist Hotelling employed calculus of variations to analyze the exhaustible resources and then he published an article on The Economics of Exhaustible Resources. Until the 1973 energy crisis, lots of papers including of Hotelling's article which are related to natural resources were ignored (Clark [4] p.81). A number of deterministic economic models of renewable resources have been developed in the literature. In this section, some selected classes of models for fishery are introduced in the form of ordinary and partial differential equations according to (Clark [3], p.9-20, 36-42, 331-340) and ([4]).

#### **Schaefer's fishery model** (Clark [4] and [3], p.9-21, 37)

According to (Clark [3], p.37), the general form of the ODE model for exploited fish population is

$$\frac{dx}{dt} = G(x) - h(t), \quad (5), \quad x(0) = x_0 \quad (6)$$

Where  $x=x(t)$  is the size of the population at time  $t$ ,  $G(x)$  is a given function modeling the natural growth of the population and  $h(t)$  is the rate of harvesting. Models differ with respect to choice of functions  $G(x)$  and  $h(t)$ .

The following model was originally developed by M.B.Schaefer as a management tool for the Eastern Tropical pacific Tuna Fishery based on the general form of the ODE model (6.1)-(6.2) (clark, [4]):

$$\frac{dx}{dt} = G(x) - qEx \quad (7)$$

Where  $x$  represents biomass of fish population,  $t$  is time,  $G(x)$  is biological net growth rate,  $q$  is catchability coefficient,  $E$  is fishing effort.  $G(x)$  incorporates certain generalized assumptions of density dependent rates of birth, growth and mortality in the population.  $G(x)$  was specified in logistic form

$$G(x) = rx(1 - x/K) \quad (8)$$

Where  $r, K$  are positive parameters.  $K$  is called the carrying capacity. The term

$$h = qEx \quad (9)$$

represents the rate of mortality imposed by the fishery. Effort is defined as a certain standardized measure of the number of vessels operating at a given time.

The Schaefer model depends upon three parameters ( $r, K, q$ ). the Schaefer model has been employed in the management of other commercial fisheries (Clark [4]).

#### **Maximization of profit in fishery management** (Clark [4], p.83)

Let us assume a private firm which owns full rights to exploitation of the fish population. The goal of the firm is to maximize the present value of discounted net economic revenue. Letting  $E=E(t)$  denote the effort input, this present value can be expressed as (Clark [4], p.83):

$$PV = \int_0^{\infty} \alpha(t)\pi(x, E, t)dt \quad (10)$$

Where  $\alpha(t) = \text{discount factor}$ ,  $\pi(x, E, t)dt = \text{net revenue flow}$  and PV stands for present value.

Let  $\alpha(t) = \exp(-\delta t)$  and  $\pi(x, E, t) = ph - cE = pqEx - cE$ , where  $\delta, p$  and  $c$  are positive constants representing the discount rate, the price of fish and the cost of effort respectively, obtain from (6.6)

$$PV = \int_0^{\infty} e^{-\delta t}(pqx - c)Edt. \quad (11)$$

To find maximum of PV, calculus of variations and optimal control techniques are used.

#### **A Diffusion Model:** The Inshore-Offshore Fishery (Clark [3], p.331-339)

It consider a fishery model in which the spatial distribution and movement of the fish are taken into account. Assume a straight shoreline, and let  $x$  denote the distance from shore. The density  $u=u(x,t)$  of a fish population at time  $t$  is assumed to depend upon the distance  $x$ . The partial differential equation modeling growth and diffusion of the population is:

$$\frac{\partial u}{\partial t} = \sigma^2 \frac{\partial^2 u}{\partial x^2} + F(x, u) \quad \text{For } 0 \leq x \leq S \quad (12)$$

Where  $F(x,u)$  represents the natural growth rate of the population at distance  $x$  from shore, and  $S$  denotes the outer limit of the population's habitat. The condition of  $F(x,u)$  is  $F(x, u) > 0$  for  $0 < u < K(x)$ .  $K(x)$  is the carrying capacity at location  $x$ . The term  $u_{xx}$  is the diffusion term. The rate of diffusion of fish is assumed to be proportional to the gradient of the density, (12) is a nonlinear partial differential equation of parabolic type. The corresponding model for harvested populations is

$$\frac{\partial u}{\partial t} = \sigma^2 \frac{\partial^2 u}{\partial x^2} + F(x, u) - E(x)u$$

Where  $E(x)$  is the fisheries effort, which is assumed to depend on the distance  $x$  from the shore. The model is for Open-access Fishery inshore ( $0 \leq x \leq a$ ), offshore ( $0 \leq x \leq S$ ) which implies everyone can share the resources without paying barrier cost.

An example of an inshore-offshore model involving two sub-populations is the following (CLARK [3], p.332). Let  $x_1$  and  $x_2$  denote the respective biomasses of inshore and offshore sub- populations. The model is

$$\begin{cases} \frac{dx_1}{dt} = F_1(x_1) + \sigma(x_2 - x_1) - E_1 x_1 \\ \frac{dx_2}{dt} = F_2(x_2) + \sigma(x_1 - x_2) - E_2 x_2 \end{cases}$$

If  $x_2 > x_1$ , the offshore population diffuses into the inshore area at a rate proportioned to  $x_2 - x_1$ . Assume that the price is the same for both two populations, but the cost of fishery supplies effort is different. The objective functional for the combined fishery is given by

$$J[E_1, E_2] = \int_0^{\infty} e^{-\delta t} \{(px_1 - c_1)E_1 + (px_2 - c_2)E_2\} dt.$$

In the next section, we will discuss the fisheries management in stochastic environment. Some stochastic models will be given.

### SDE in Fisheries Management

Fisheries management is full of uncertainty. The SDE is a useful tool to model some behavior in Fisheries Management. Lots of stochastic models are employed in Fisheries Management. In this section, some models which are based on population dynamics will be presented and discussed. The material is researched mainly from the books by (Gard [10], p.157-172) and (Clark[3], p.9-20).

The simplest population model is:

$$dX_t = aXdt + bXdW_t \quad (13)$$

(13) is an extension of a deterministic model of the form

$$\frac{dX}{dt} = aX \quad , \text{ Where } a \text{ denotes intrinsic growth rate.} \quad (14)$$

Stochastic analogues can be developed from the deterministic logistic model

$$\frac{dX(t)}{dt} = rX \left(1 - \frac{X}{K}\right) \quad (15)$$

In various ways, as for example,

1. By considering the deterministic logistic model (15) and adding a stochastic term, thus obtaining

$$dX_t = rX \left(1 - \frac{X}{K}\right) dt + \sigma X dW_t \quad (16)$$

2. By considering the deterministic logistic model (15) and replacing r by  $r + \sigma W_t$ , thus obtaining

$$dX_t = rX \left(1 - \frac{X}{K}\right) dt + \sigma X \left(1 - \frac{X}{K}\right) dW_t \quad (17)$$

3. By considering the deterministic logistic model (15) and replacing  $1/K$  by  $\frac{1}{K} + \sigma W_t$ , thus obtaining

$$dX_t = rX \left(1 - \frac{X}{K}\right) dt + \sigma r X^2 dW_t \quad (18)$$

In terms of harvest behavior, Eqn. (13) and (16)-(18) can be modified, thus obtaining:

$$dX_t = (aX - h)dt + bXdW_t \quad (19)$$

$$dX_t = \left(rX \left(1 - \frac{X}{K}\right) - h(t)\right) dt + \sigma X dW_t \quad (20)$$

$$dX_t = \left(rX \left(1 - \frac{X}{K}\right) - h(t)\right) dt + \sigma X \left(1 - \frac{X}{K}\right) dW_t \quad (21)$$

$$dX_t = \left( rX \left( 1 - \frac{X}{K} \right) - h(t) \right) dt + \sigma r X^2 dW_t \quad (22)$$

### Numerical results:

In the present study we looking at the numerical solution of the present model. The basic of the paper to simulate the above problem in RKM (Runge kutta Method) scheme using some programming technique. C++ and Fotran-90 being used for it . The two test problem has been introduce for the solution. The analytical result are also presented for the test problems. And comparison being made here in Table-1 to 10. we will provide numerical results which are absolute values of the error at the end point  $T=1$  for different step sizes ( $\Delta=1/N$ ). In table 1 to 10, we will also provide ten series of numerical results for  $\Delta=0.0001$  ( $N=1000$ ). In all tables, EM denotes the EM method, Milstein denotes the Milstein mthod, Adams denotes the Adams method with fourth order and RKM (Runge kutta Method)denotes fourth order accurate difference method.

### Test problem 1

$$dX_t = -(\alpha + \beta^2 X)(1 - X^2)dt + \beta(1 - X^2)dW_t \quad (23)$$

Where  $\alpha = 1, \beta = 2, t \in [0,1]$  and  $X_0 = 0$ .

The exact solution is:

$$X_t = \frac{(1+X_0)e^{(-2\alpha t+2\beta W_t)}+X_0-1}{(1+X_0)e^{(-2\alpha t+2\beta W_t)}+1-X_0}.$$

The test problem is from (burrage and burrage [2]).

### Test problem 2

$$dX_t = (rX - h)dt + \sigma X dW_t \quad (24)$$

Where  $r = 0.2246, h = 4.9028e + 004, \sigma = 0.1, t \in [0,1]$  and  $X_0 = 200000$ .

The exact solution is:

$$X_t = e^{\left( \left( r - \frac{1}{2}\sigma^2 \right) t + \sigma W_t \right)} \left( X_0 - h \int_0^t \frac{1}{\left( \left( r - \frac{1}{2}\sigma^2 \right) s + \sigma W_s \right)} ds \right).$$

The analytical solution can be derived from (17). the part of coefficients are from (McDonald, Sandal and Steinshamn[21]). Eqn (24) can be employed in fisheries management to model fish population changes in terms of constant harvest.

The numerical results for both cases are showns in tables given below.

### Numrical simulation and Grid Generation:

The RKM (Runge kutta Method) has been adopted for the computational solution.. It is vary difficult to give the phenomena of computer programming for all the differential equation our main motivation to simulate over problem using RKM (Runge kutta Method)scheme so a generalized simulation process in C++ and FOTRAN 90 to the simulation algorithm corresponding to the linear differential equation (IVP) taking as counter example is shown here .There are two different kind of scheme can be used for the simplicity on is Implicit and other is semi implicit scheme . The simulation process is given below

#### FOTRAN

```
SUBROUTINE DOPRI5(N,FCN,X,Y,XEND,
& RTOL,ATOL,ITOL,
& SOLOUT,IOUT,
&
WORK,LWORK,IWORK,LIWORK,RPAR,IPAR,IDLID)
```

```
C -----
C  NUMERICAL SOLUTION OF A SYSTEM OF FIRST
0RDER
C  ORDINARY DIFFERENTIAL EQUATIONS Y'=F(X,Y).
C  THIS IS AN EXPLICIT RUNGE-KUTTA METHOD OF
0RDER (4)5
```

```

C DUE TO DORMAND & PRINCE (WITH STEPSIZE
CONTROL AND
C DENSE OUTPUT).
C
C
C INPUT PARAMETERS
C -----
C N      DIMENSION OF THE SYSTEM
C
C FCN    NAME (EXTERNAL) OF SUBROUTINE
COMPUTING THE
C      VALUE OF F(X,Y):
C      SUBROUTINE FCN(N,X,Y,F,RPAR,IPAR)
C      DOUBLE PRECISION X,Y(N),F(N)
C      F(1)=... ETC.
C
C X      INITIAL X-VALUE
C
C Y(N)   INITIAL VALUES FOR Y
C
C XEND   FINAL X-VALUE (XEND-X MAY BE
POSITIVE OR NEGATIVE)
C
C RTOL,ATOL RELATIVE AND ABSOLUTE ERROR
TOLERANCES. THEY
C CAN BE BOTH SCALARS OR ELSE BOTH
VECTORS OF LENGTH N.
C
C ITOL   SWITCH FOR RTOL AND ATOL:
C ITOL=0: BOTH RTOL AND ATOL ARE
SCALARS.
C THE CODE KEEPS, ROUGHLY, THE LOCAL
ERROR OF
C Y(I) BELOW RTOL*ABS(Y(I))+ATOL
C ITOL=1: BOTH RTOL AND ATOL ARE
VECTORS.
C THE CODE KEEPS THE LOCAL ERROR OF Y(I)
BELOW
C RTOL(I)*ABS(Y(I))+ATOL(I).
C
C SOLOUT  NAME (EXTERNAL) OF SUBROUTINE
PROVIDING THE
C NUMERICAL SOLUTION DURING
INTEGRATION.
C IF IOUT.GE.1, IT IS CALLED AFTER EVERY
SUCCESSFUL STEP.
C SUPPLY A DUMMY SUBROUTINE IF IOUT=0.
C IT MUST HAVE THE FORM
C      SUBROUTINE SOLOUT
(NR,XOLD,X,Y,N,CON,ICOMP,ND,
C      RPAR,IPAR,IRTRN)
C      DIMENSION Y(N),CON(5*ND),ICOMP(ND)
C      ...
C      SOLOUT FURNISHES THE SOLUTION "Y" AT
THE NR-TH
C GRID-POINT "X" (THEREBY THE INITIAL
VALUE IS
C THE FIRST GRID-POINT).
C "XOLD" IS THE PRECEEDING GRID-POINT.
C "IRTRN" SERVES TO INTERRUPT THE
INTEGRATION. IF IRTRN
C IS SET <0, DOPRI5 WILL RETURN TO THE
CALLING PROGRAM.
C IF THE NUMERICAL SOLUTION IS ALTERED
IN SOLOUT,
C SET IRTRN = 2
C
C ----- CONTINUOUS OUTPUT: -----
C DURING CALLS TO "SOLOUT", A CONTINUOUS
SOLUTION

```

```

C      FOR THE INTERVAL [XOLD,X] IS AVAILABLE
THROUGH
C      THE FUNCTION
C      >>> CONTD5(I,S,CON,ICOMP,ND) <<<
C      WHICH PROVIDES AN APPROXIMATION TO
THE I-TH
C      COMPONENT OF THE SOLUTION AT THE
POINT S. THE VALUE
C      S SHOULD LIE IN THE INTERVAL [XOLD,X].
C
C IOUT   SWITCH FOR CALLING THE SUBROUTINE
SOLOUT:
C      IOUT=0: SUBROUTINE IS NEVER CALLED
C      IOUT=1: SUBROUTINE IS USED FOR OUTPUT.
C      IOUT=2: DENSE OUTPUT IS PERFORMED IN
SOLOUT
C      (IN THIS CASE WORK(5) MUST BE
SPECIFIED)
C
C WORK   ARRAY OF WORKING SPACE OF LENGTH
"LWORK".
C WORK(1),...,WORK(20) SERVE AS PARAMETERS
FOR THE CODE.
C FOR STANDARD USE, SET THEM TO ZERO
BEFORE CALLING.
C "LWORK" MUST BE AT LEAST
8*N+5*NRDENS+21
C WHERE NRDENS = IWORK(5)
C
C LWORK  DECLARED LENGTH OF ARRAY "WORK".
C
C IWORK  INTEGER WORKING SPACE OF LENGTH
"LIWORK".
C IWORK(1),...,IWORK(20) SERVE AS
PARAMETERS FOR THE CODE.
C FOR STANDARD USE, SET THEM TO ZERO
BEFORE CALLING.
C "LIWORK" MUST BE AT LEAST NRDENS+21 .
C
C LIWORK DECLARED LENGTH OF ARRAY "IWORK".
C
C RPAR, IPAR REAL AND INTEGER PARAMETERS (OR
PARAMETER ARRAYS) WHICH
C CAN BE USED FOR COMMUNICATION
BETWEEN YOUR CALLING
C PROGRAM AND THE FCN, JAC, MAS, SOLOUT
SUBROUTINES.
C -----
C
C SOPHISTICATED SETTING OF PARAMETERS
C -----
C SEVERAL PARAMETERS
(WORK(1),...,IWORK(1),...) ALLOW
C TO ADAPT THE CODE TO THE PROBLEM AND TO
THE NEEDS OF
C THE USER. FOR ZERO INPUT, THE CODE
CHOISES DEFAULT VALUES.
C
C WORK(1) UROUND, THE ROUNDING UNIT, DEFAULT
2.3D-16.
C
C WORK(2) THE SAFETY FACTOR IN STEP SIZE
PREDICTION,
C DEFAULT 0.9D0.
C
C WORK(3), WORK(4) PARAMETERS FOR STEP SIZE
SELECTION
C THE NEW STEP SIZE IS CHOSEN SUBJECT TO
THE RESTRICTION

```

```

C      WORK(3) <= HNEW/HOLD <= WORK(4)
C      DEFAULT VALUES: WORK(3)=0.2D0,
WORK(4)=10.D0
C
C   WORK(5) IS THE "BETA" FOR STABILIZED STEP SIZE
CONTROL
C      (SEE SECTION IV.2). LARGER VALUES OF BETA (
<= 0.1 )
C      MAKE THE STEP SIZE CONTROL MORE STABLE.
DOPR15 NEEDS
C      A LARGER BETA THAN HIGHAM & HALL.
NEGATIVE WORK(5)
C      PROVOKE BETA=0.
C      DEFAULT 0.04D0.
C
C   WORK(6) MAXIMAL STEP SIZE, DEFAULT XEND-X.
C
C   WORK(7) INITIAL STEP SIZE, FOR WORK(7)=0.D0 AN
INITIAL GUESS
C      IS COMPUTED WITH HELP OF THE FUNCTION
HINIT
C
C   IWORK(1) THIS IS THE MAXIMAL NUMBER OF
ALLOWED STEPS.
C      THE DEFAULT VALUE (FOR IWORK(1)=0) IS
100000.
C
C   IWORK(2) SWITCH FOR THE CHOICE OF THE
COEFFICIENTS
C      IF IWORK(2).EQ.1 METHOD DOPR15 OF
DORMAND AND PRINCE
C      (TABLE 5.2 OF SECTION II.5).
C      AT THE MOMENT THIS IS THE ONLY POSSIBLE
CHOICE.
C      THE DEFAULT VALUE (FOR IWORK(2)=0) IS
IWORK(2)=1.
C
C   IWORK(3) SWITCH FOR PRINTING ERROR MESSAGES
C      IF IWORK(3).LT.0 NO MESSAGES ARE BEING
PRINTED
C      IF IWORK(3).GT.0 MESSAGES ARE PRINTED
WITH
C      WRITE (IWORK(3),*) ...
C      DEFAULT VALUE (FOR IWORK(3)=0) IS
IWORK(3)=6
C
C   IWORK(4) TEST FOR STIFFNESS IS ACTIVATED
AFTER STEP NUMBER
C      J*IWORK(4) (J INTEGER), PROVIDED
IWORK(4).GT.0.
C      FOR NEGATIVE IWORK(4) THE STIFFNESS TEST
IS
C      NEVER ACTIVATED; DEFAULT VALUE IS
IWORK(4)=1000
C
C   IWORK(5) = NRDENS = NUMBER OF COMPONENTS,
FOR WHICH DENSE OUTPUT
C      IS REQUIRED; DEFAULT VALUE IS IWORK(5)=0;
C      FOR 0 < NRDENS < N THE COMPONENTS (FOR
WHICH DENSE
C      OUTPUT IS REQUIRED) HAVE TO BE SPECIFIED
IN
C      IWORK(21),...,IWORK(NRDENS+20);
C      FOR NRDENS=N THIS IS DONE BY THE CODE.
C
C-----.
C
C   OUTPUT PARAMETERS
C -----

```

```

C X      X-VALUE FOR WHICH THE SOLUTION HAS
BEEN COMPUTED
C           (AFTER SUCCESSFUL RETURN X=XEND).
C
C Y(N)    NUMERICAL SOLUTION AT X
C
C H      PREDICTED STEP SIZE OF THE LAST
ACCEPTED STEP
C
C IDID   REPORTS ON SUCCESSFULNESS UPON
RETURN:
C           IDID= 1 COMPUTATION SUCCESSFUL,
C           IDID= 2 COMPUT. SUCCESSFUL
(INTERRUPTED BY SOLOUT)
C           IDID=-1 INPUT IS NOT CONSISTENT,
C           IDID=-2 LARGER NMAX IS NEEDED,
C           IDID=-3 STEP SIZE BECOMES TOO SMALL.
C           IDID=-4 PROBLEM IS PROBABLY STIFF
(INTERRUPTED).
C
C IWORK(17) NFCN  NUMBER OF FUNCTION
EVALUATIONS
C IWORK(18) NSTEP  NUMBER OF COMPUTED STEPS
C IWORK(19) NACCPT NUMBER OF ACCEPTED STEPS
C IWORK(20) NREJCT NUMBER OF REJECTED STEPS
(DUE TO ERROR TEST),
C           (STEP REJECTIONS IN THE FIRST STEP ARE
NOT COUNTED)
C-----
C *** * * * * * * * * * * * * * * * * * * * *
C DECLARATIONS
C *** * * * * * * * * * * * * * * * * * * * *
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
DIMENSION
Y(N),ATOL(*),RTOL(*),WORK(LWORK),IWORK(LIWORK)
DIMENSION RPAR(*),IPAR(*)
LOGICAL ARRET
EXTERNAL FCN,SOLOUT
C *** * * * * * * * * * * * * * *
C SETTING THE PARAMETERS
C *** * * * * * * * * * * * * * *
NFCN=0
NSTEP=0
NACCPT=0
NREJCT=0
ARRET=.FALSE.
C ----- IPRINT FOR MONITORING THE PRINTING
IF(IWORK(3).EQ.0)THEN
  IPRINT=6
ELSE
  IPRINT=IWORK(3)
END IF
C ----- NMAX , THE MAXIMAL NUMBER OF STEPS -----
IF(IWORK(1).EQ.0)THEN
  NMAX=100000
ELSE
  NMAX=IWORK(1)
  IF(NMAX.LE.0)THEN
    IF (IPRINT.GT.0) WRITE(IPRINT,*)
    &      'WRONG INPUT IWORK(1)=',IWORK(1)
    ARRET=.TRUE.
  END IF
END IF
C ----- METH COEFFICIENTS OF THE METHOD
IF(IWORK(2).EQ.0)THEN
  METH=1
ELSE
  METH=IWORK(2)
  IF(METH.LE.0.OR.METH.GE.4)THEN
    IF (IPRINT.GT.0) WRITE(IPRINT,*)

```

```

&      ' CURIOUS INPUT IWORK(2)='IWORK(2)
     ARRET=.TRUE.
    END IF
   END IF
C ----- NSTIFF  PARAMETER FOR STIFFNESS
DETECTION
  NSTIFF=IWORK(4)
  IF (NSTIFF.EQ.0) NSTIFF=1000
  IF (NSTIFF.LT.0) NSTIFF=NMAX+10
C ----- NRDENS  NUMBER OF DENSE OUTPUT
COMPONENTS
  NRDENS=IWORK(5)
  IF(NRDENS.LT.0.OR.NRDENS.GT.N)THEN
    IF (IPRINT.GT.0) WRITE(IPRINT,*)
    &      ' CURIOUS INPUT IWORK(5)='IWORK(5)
    ARRET=.TRUE.
  ELSE
    IF(NRDENS.GT.0.AND.IOUT.LT.2)THEN
      IF (IPRINT.GT.0) WRITE(IPRINT,*)
    & ' WARNING: PUT IOUT=2 FOR DENSE OUTPUT '
    END IF
    IF (NRDENS.EQ.N) THEN
      DO 16 I=1,NRDENS
16      IWORK(20+I)=I
    END IF
  END IF
C ----- UROUND  SMALLEST NUMBER SATISFYING
1.D0+UROUND>1.D0
  IF(WORK(1).EQ.0.D0)THEN
    UROUND=2.3D-16
  ELSE
    UROUND=WORK(1)
    IF(UROUND.LE.1.D-35.OR.UROUND.GE.1.D0)THEN
      IF (IPRINT.GT.0) WRITE(IPRINT,*)
    & ' WHICH MACHINE DO YOU HAVE? YOUR
UROUND WAS:',WORK(1)
    ARRET=.TRUE.
  END IF
  END IF
C ----- SAFETY FACTOR -----
  IF(WORK(2).EQ.0.D0)THEN
    SAFE=0.9D0
  ELSE
    SAFE=WORK(2)
    IF(SAFE.GE.1.D0.OR.SAFE.LE.1.D-4)THEN
      IF (IPRINT.GT.0) WRITE(IPRINT,*)
    &      ' CURIOUS INPUT FOR SAFETY FACTOR
WORK(2)='IWORK(2)
    ARRET=.TRUE.
  END IF
  END IF
C ----- FAC1,FAC2  PARAMETERS FOR STEP SIZE
SELECTION
  IF(WORK(3).EQ.0.D0)THEN
    FAC1=0.2D0
  ELSE
    FAC1=WORK(3)
  END IF
  IF(WORK(4).EQ.0.D0)THEN
    FAC2=10.D0
  ELSE
    FAC2=WORK(4)
  END IF
C ----- BETA FOR STEP CONTROL STABILIZATION -----
  IF(WORK(5).EQ.0.D0)THEN
    BETA=0.04D0
  ELSE
    IF(WORK(5).LT.0.D0)THEN
      BETA=0.D0
    ELSE
      BETA=WORK(5)
      IF(BETA.GT.0.2D0)THEN
        IF (IPRINT.GT.0) WRITE(IPRINT,*)
        &      ' CURIOUS INPUT FOR BETA:
      WORK(5)='IWORK(5)
      ARRET=.TRUE.
    END IF
    END IF
  END IF
C ----- MAXIMAL STEP SIZE
  IF(WORK(6).EQ.0.D0)THEN
    HMAX=XEND-X
  ELSE
    HMAX=WORK(6)
  END IF
C ----- INITIAL STEP SIZE
  H=WORK(7)
C ----- PREPARE THE ENTRY-POINTS FOR THE ARRAYS
IN WORK -----
  IEY1=21
  IEK1=IEY1+N
  IEK2=IEK1+N
  IEK3=IEK2+N
  IEK4=IEK3+N
  IEK5=IEK4+N
  IEK6=IEK5+N
  IEYS=IEK6+N
  IECO=IEYS+N
C ----- TOTAL STORAGE REQUIREMENT -----
  ISTORE=IEYS+5*NRDENS-1
  IF(ISTORE.GT.LWORK)THEN
    IF (IPRINT.GT.0) WRITE(IPRINT,*)
    & ' INSUFFICIENT STORAGE FOR WORK, MIN.
LWORK='ISTORE
    ARRET=.TRUE.
  END IF
  ICOMP=21
  ISTORE=ICOMP+NRDENS-1
  IF(ISTORE.GT.LIWORK)THEN
    IF (IPRINT.GT.0) WRITE(IPRINT,*)
    & ' INSUFFICIENT STORAGE FOR IWORK, MIN.
LIWORK='ISTORE
    ARRET=.TRUE.
  END IF
C ----- WHEN A FAIL HAS OCCURED, WE RETURN WITH
IDID=-1
  IF (ARRET) THEN
    IDID=-1
    RETURN
  END IF
C ----- CALL TO CORE INTEGRATOR -----
  CALL
  DOPCOR(N,FCN,X,Y,XEND,HMAX,H,RTOL,ATOL,ITOL,IPRI
NT,
  &
  SOLOUT,IOUT,IDL,NMAX,UROUND,METH,NSTIFF,SAFE,
  BETA,FAC1,FAC2,
  &
  WORK(IEY1),WORK(IEK1),WORK(IEK2),WORK(IEK3),WOR
K(IEK4),
  &
  WORK(IEK5),WORK(IEK6),WORK(IEYS),WORK(IECO),IWO
RK(ICOMP),
  & NRDENS,RPAR,IPAR,NFCN,NSTEP,NACCPT,NREJCT)
  WORK(7)=H
  IWORK(17)=NFCN
  IWORK(18)=NSTEP
  IWORK(19)=NACCPT
  IWORK(20)=NREJCT

```

```

C ----- RETURN -----
    RETURN
END

C
C
C
C ----- ... AND HERE IS THE CORE INTEGRATOR -----
C
    SUBROUTINE
DOPCOR(N,FCN,X,Y,XEND,HMAX,H,RTOL,ATOL,ITOL,IPRI
NT,
    &
SOLOUT,IOUT,IDID,NMAX,UROUND,METH,NSTIFF,SAFE,
BETA,FAC1,FAC2,
    &
Y1,K1,K2,K3,K4,K5,K6,YSTI,CONT,ICOMP,NRD,RPAR,IPAR,
    & NFCN,NSTEP,NACCPT,NREJCT)
C -----
C CORE INTEGRATOR FOR DOPRI5
C PARAMETERS SAME AS IN DOPRI5 WITH
WORKSPACE ADDED
C -----
C DECLARATIONS
C -----
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
DOUBLE PRECISION
K1(N),K2(N),K3(N),K4(N),K5(N),K6(N)
DIMENSION
Y(N),Y1(N),YSTI(N),ATOL(*),RTOL(*),RPAR(*),IPAR(*)
DIMENSION CONT(5*NRD),ICOMP(NRD)
LOGICAL REJECT, LAST
EXTERNAL FCN
COMMON /COND05/XOLD,HOUT
C **** * * * * * * * * * *
C INITIALISATIONS
C **** * * * * * * * * * *
    IF (METH.EQ.1) CALL
CDOPRI(C2,C3,C4,C5,E1,E3,E4,E5,E6,E7,
    & A21,A31,A32,A41,A42,A43,A51,A52,A53,A54,
    & A61,A62,A63,A64,A65,A71,A73,A74,A75,A76,
    & D1,D3,D4,D5,D6,D7)
FACOLD=1.D-4
EXPO1=0.2D0-BETA*0.75D0
FACC1=1.D0/FAC1
FACC2=1.D0/FAC2
POSNEG=SIGN(1.D0,XEND-X)
C --- INITIAL PREPARATIONS
ATOLI=ATOL(1)
RTOLI=RTOL(1)
LAST=.FALSE.
HLAMB=0.D0
IASTI=0
CALL FCN(N,X,Y,K1,RPAR,IPAR)
HMAX=ABS(HMAX)
IORD=5
IF (H.EQ.0.D0)
H=HINIT(N,FCN,X,Y,XEND,POSNEG,K1,K2,K3,IORD,
    & HMAX,ATOL,RTOL,ITOL,RPAR,IPAR)
NFCN=NFCN+2
REJECT=.FALSE.
XOLD=X
IF (IOUT.NE.0) THEN
    IRTRN=1
    HOUT=H
    CALL
SOLOUT(NACCPT+1,XOLD,X,Y,N,CONT,ICOMP,NRD,
    & RPAR,IPAR,IRTRN)
    IF (IRTRN.LT.0) GOTO 79
ELSE
    IRTRN=0
END IF
C --- BASIC INTEGRATION STEP
1 CONTINUE
IF (NSTEP.GT.NMAX) GOTO 78
IF (0.1D0*ABS(H).LE.ABS(X)*UROUND)GOTO 77
IF ((X+1.0D0*H-XEND)*POSNEG.GT.0.D0) THEN
    H=XEND-X
    LAST=.TRUE.
END IF
NSTEP=NSTEP+1
C --- THE FIRST 6 STAGES
IF (IRTRN.GE.2) THEN
    CALL FCN(N,X,Y,K1,RPAR,IPAR)
END IF
DO 22 I=1,N
22 Y1(I)=Y(I)+H*A21*K1(I)
    CALL FCN(N,X+C2*H,Y1,K2,RPAR,IPAR)
    DO 23 I=1,N
23 Y1(I)=Y(I)+H*(A31*K1(I)+A32*K2(I))
    CALL FCN(N,X+C3*H,Y1,K3,RPAR,IPAR)
    DO 24 I=1,N
24 Y1(I)=Y(I)+H*(A41*K1(I)+A42*K2(I)+A43*K3(I))
    CALL FCN(N,X+C4*H,Y1,K4,RPAR,IPAR)
    DO 25 I=1,N
25 Y1(I)=Y(I)+H*(A51*K1(I)+A52*K2(I)+A53*K3(I)+A54*K4(I))
    CALL FCN(N,X+C5*H,Y1,K5,RPAR,IPAR)
    DO 26 I=1,N
26 YSTI(I)=Y(I)+H*(A61*K1(I)+A62*K2(I)+A63*K3(I)+A64*K4(I)
    +A65*K5(I))
    XPH=X+H
    CALL FCN(N,XPH,YSTI,K6,RPAR,IPAR)
    DO 27 I=1,N
27 Y1(I)=Y(I)+H*(A71*K1(I)+A73*K3(I)+A74*K4(I)+A75*K5(I)+A76*K6(I))
    CALL FCN(N,XPH,Y1,K2,RPAR,IPAR)
    IF (IOUT.GE.2) THEN
        DO 40 J=1,NRD
            I=ICOMP(J)
CONT(4*NRD+J)=H*(D1*K1(I)+D3*K3(I)+D4*K4(I)+D5*K5(I)
    )+D6*K6(I)+D7*K2(I))
40 CONTINUE
    END IF
    DO 28 I=1,N
28 K4(I)=(E1*K1(I)+E3*K3(I)+E4*K4(I)+E5*K5(I)+E6*K6(I)+E7*
    K2(I))*H
    NFCN=NFCN+6
C --- ERROR ESTIMATION
ERR=0.D0
IF (ITOL.EQ.0) THEN
    DO 41 I=1,N
        SK=ATOL(I)+RTOL(I)*MAX(ABS(Y(I)),ABS(Y1(I)))
41 ERR=ERR+(K4(I)/SK)**2
    ELSE
        DO 42 I=1,N
            SK=ATOL(I)+RTOL(I)*MAX(ABS(Y(I)),ABS(Y1(I)))
42 ERR=ERR+(K4(I)/SK)**2
    END IF
    ERR=SQRT(ERR/N)
C --- COMPUTATION OF HNEW
FAC11=ERR**EXPO1
C --- LUND-STABILIZATION
FAC=FAC11/FACOLD**BETA
C --- WE REQUIRE FAC1 <= HNEW/H <= FAC2
FAC=MAX(FACC2,MIN(FACC1,FAC/SAFE))

```

```

HNEW=H/FAC
IF(ERR.LE.1.D0)THEN
C --- STEP IS ACCEPTED
    FACOLD=MAX(ERR,1.0D-4)
    NACCPT=NACCPT+1
C ----- STIFFNESS DETECTION
    IF (MOD(NACCPT,NSTIFF).EQ.0.OR.IASTI.GT.0) THEN
        STNUM=0.D0
        STDEN=0.D0
        DO 64 I=1,N
            STNUM=STNUM+(K2(I)-K6(I))**2
            STDEN=STDEN+(Y1(I)-YSTI(I))**2
64    CONTINUE
    IF (STDEN.GT.0.D0)
        HLAMB=H*SQRT(STNUM/STDEN)
        IF (HLAMB.GT.3.25D0) THEN
            NONSTI=0
            IASTI=IASTI+1
            IF (IASTI.EQ.15) THEN
                IF (IPRINT.GT.0) WRITE (IPRINT,*)
                &          ' THE PROBLEM SEEMS TO BECOME STIFF
AT X = ',X
                IF (IPRINT.LE.0) GOTO 76
            END IF
        ELSE
            NONSTI=NONSTI+1
            IF (NONSTI.EQ.6) IASTI=0
        END IF
    END IF
    IF (IOUT.GE.2) THEN
        DO 43 J=1,NRD
            I=ICOMP(J)
            YD0=Y(I)
            YDIFF=Y1(I)-YD0
            BSPL=H*K1(I)-YDIFF
            CONT(J)=Y(I)
            CONT(NRD+J)=YDIFF
            CONT(2*NRD+J)=BSPL
            CONT(3*NRD+J)=-H*K2(I)+YDIFF-BSPL
43    CONTINUE
    END IF
    DO 44 I=1,N
        K1(I)=K2(I)
44    Y(I)=Y1(I)
        XOLD=X
        X=XPH
        IF (IOUT.NE.0) THEN
            HOUT=H
            CALL
SOLOUT(NACCPT+1,XOLD,X,Y,N,CONT,ICOMP,NRD,
&           RPAR,IPAR,IRTRN)
            IF (IRTRN.LT.0) GOTO 79
        END IF
C ----- NORMAL EXIT
        IF (LAST) THEN
            H=HNEW
            IDID=1
            RETURN
        END IF
        IF(ABS(HNEW).GT.HMAX)HNEW=POSNEG*HMAX
        IF(REJECT)HNEW=POSNEG*MIN(ABS(HNEW),ABS(H))
        REJECT=.FALSE.
    ELSE
C --- STEP IS REJECTED
        HNEW=H/MIN(FACC1,FAC11/SAFE)
        REJECT=.TRUE.
        IF(NACCPT.GE.1)NREJECT=NREJECT+1
        LAST=.FALSE.
    END IF
H=HNEW
GOTO 1
C --- FAIL EXIT
76 CONTINUE
    IDID=-4
    RETURN
77 CONTINUE
    IF (IPRINT.GT.0) WRITE(IPRINT,979)X
    IF (IPRINT.GT.0) WRITE(IPRINT,' STEP SIZE TOO
SMALL, H=',H
    IDID=-3
    RETURN
78 CONTINUE
    IF (IPRINT.GT.0) WRITE(IPRINT,979)X
    IF (IPRINT.GT.0) WRITE(IPRINT,*)'
&      ' MORE THAN NMAX '=',NMAX,STEPS ARE
NEEDED'
    IDID=-2
    RETURN
79 CONTINUE
    IF (IPRINT.GT.0) WRITE(IPRINT,979)X
979 FORMAT(' EXIT OF DOPRI5 AT X=',E18.4)
    IDID=2
    RETURN
END
C
FUNCTION
HINIT(N,FCN,X,Y,XEND,POSNEG,F0,F1,Y1,IORD,
&           HMAX,ATOL,RTOL,ITOL,RPAR,IPAR)
C -----
C ---- COMPUTATION OF AN INITIAL STEP SIZE GUESS
C -----
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
DIMENSION Y(N),Y1(N),F0(N),F1(N),ATOL(*),RTOL(*)
DIMENSION RPAR(*),IPAR(*)
C ---- COMPUTE A FIRST GUESS FOR EXPLICIT EULER AS
C ---- H = 0.01 * NORM (Y0) / NORM (F0)
C ---- THE INCREMENT FOR EXPLICIT EULER IS SMALL
C ---- COMPARED TO THE SOLUTION
    DNF=0.0D0
    DNY=0.0D0
    ATOLI=ATOL(1)
    RTOLI=RTOL(1)
    IF (ITOL.EQ.0) THEN
        DO 10 I=1,N
            SK=ATOLI+RTOLI*ABS(Y(I))
            DNF=DNF+(F0(I)/SK)**2
10    DNY=DNY+(Y(I)/SK)**2
    ELSE
        DO 11 I=1,N
            SK=ATOL(I)+RTOL(I)*ABS(Y(I))
            DNF=DNF+(F0(I)/SK)**2
11    DNY=DNY+(Y(I)/SK)**2
    END IF
    IF (DNF.LE.1.D-10.OR.DNY.LE.1.D-10) THEN
        H=1.0D-6
    ELSE
        H=SQRT(DNY/DNF)*0.01D0
    END IF
    H=MIN(H,HMAX)
    H=SIGN(H,POSNEG)
C ---- PERFORM AN EXPLICIT EULER STEP
    DO 12 I=1,N
12    Y1(I)=Y(I)+H*F0(I)
        CALL FCN(N,X+H,Y1,F1,RPAR,IPAR)
C ---- ESTIMATE THE SECOND DERIVATIVE OF THE
SOLUTION
        DER2=0.0D0
        IF (ITOL.EQ.0) THEN
            DO 15 I=1,N

```

```

SK=ATOLI+RTOLI*ABS(Y(I))
15 DER2=DER2+((F1(I)-F0(I))/SK)**2
ELSE
DO 16 I=1,N
SK=ATOL(I)+RTOL(I)*ABS(Y(I))
16 DER2=DER2+((F1(I)-F0(I))/SK)**2
END IF
DER2=SQRT(DER2)/H
C ---- STEP SIZE IS COMPUTED SUCH THAT
C ---- H**IORD * MAX ( NORM ( F0 ), NORM ( DER2 ) ) = 0.01
DER12=MAX(ABS(DER2),SQRT(DNF))
IF (DER12.LE.1.D-15) THEN
H1=MAX(1.0D-6,ABS(H)*1.0D-3)
ELSE
H1=(0.01D0/DER12)**(1.D0/IORD)
END IF
H=MIN(100*ABS(H),H1,HMAX)
HINIT=SIGN(H,POSNEG)
RETURN
END
C
FUNCTION CONTD5(II,X,CON,ICOMP,ND)
C -----
C THIS FUNCTION CAN BE USED FOR CONTINUOUS
OUTPUT IN CONNECTION
C WITH THE OUTPUT-SUBROUTINE FOR DOPRI5. IT
PROVIDES AN
C APPROXIMATION TO THE II-TH COMPONENT OF THE
SOLUTION AT X.
C -----
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
DIMENSION CON(5*ND),ICOMP(ND)
COMMON /CONDOS5/XOLD,H
C ---- COMPUTE PLACE OF II-TH COMPONENT
I=0
DO 5 J=1,ND
IF (ICOMP(J).EQ.II) I=J
5 CONTINUE
IF (LEQ.0) THEN
WRITE (6,*) ' NO DENSE OUTPUT AVAILABLE FOR
COMP.',II
RETURN
END IF
THETA=(X-XOLD)/H
THETA1=1.D0-THETA

CONTD5=CON(I)+THETA*(CON(ND+I)+THETA1*(CON(2*N
D+I)+THETA*
& (CON(3*ND+I)+THETA1*CON(4*ND+I))))
RETURN
END
C
SUBROUTINE CDOPRI(C2,C3,C4,C5,E1,E3,E4,E5,E6,E7,
& A21,A31,A32,A41,A42,A43,A51,A52,A53,A54,
& A61,A62,A63,A64,A65,A71,A73,A74,A75,A76,
& D1,D3,D4,D5,D6,D7)
C -----
C RUNGE-KUTTA COEFFICIENTS OF DORMAND AND
PRINCE (1980)
C -----
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
C2=0.2D0
C3=0.3D0
C4=0.8D0
C5=8.D0/9.D0
A21=0.2D0
A31=3.D0/40.D0
A32=9.D0/40.D0
A41=44.D0/45.D0
A42=-56.D0/15.D0
A43=32.D0/9.D0
A51=19372.D0/6561.D0
A52=-25360.D0/2187.D0
A53=64448.D0/6561.D0
A54=-212.D0/729.D0
A61=9017.D0/3168.D0
A62=-355.D0/33.D0
A63=46732.D0/5247.D0
A64=49.D0/176.D0
A65=-5103.D0/18656.D0
A71=35.D0/384.D0
A73=500.D0/1113.D0
A74=125.D0/192.D0
A75=-2187.D0/6784.D0
A76=11.D0/84.D0
E1=71.D0/57600.D0
E3=-71.D0/16695.D0
E4=71.D0/1920.D0
E5=-17253.D0/339200.D0
E6=22.D0/525.D0
E7=-1.D0/40.D0
C ---- DENSE OUTPUT OF SHAMPINE (1986)
D1=-12715105075.D0/11282082432.D0
D3=87487479700.D0/32700410799.D0
D4=-10690763975.D0/1880347072.D0
D5=701980252875.D0/199316789632.D0
D6=-1453857185.D0/822651844.D0
D7=69997945.D0/29380423.D0
RETURN
END

/*****
***** */

A C++ Program to estimate the Differential value of a given
function at given point from given data using Runge-
Kutta Methods.

***** */

//----- Header Files -----//

***** */
#include <iostream.h>
#include <stdlib.h>
#include <string.h>
#include <stdio.h>
#include <conio.h>
#include <math.h>

***** */

//----- Global Variables -----//

***** */

```

```

const int max_size=13;
int n=0;
int top=-1;
long double h=0;
long double x=0;
long double x0=0;
long double y0=0;
long double two_stage_result=0;
long double three_stage_result=0;
long double xx[max_size]={0};
long double yy2[max_size]={0};
long double yy3[max_size]={0};
char Fx[100]={NULL};
char Stack[30][30]={NULL};
char Postfix_expression[30][30]={NULL};

/***************** Funcion Prototypes *******/
void push(const char *);
void convert_ie_to_pe(const char *);
const char* pop();
const long double evaluate_pe(const long double,const long double);
void show_screen();
void clear_screen();
void get_input();
void apply_runge_kutta_method();
void show_result();

/************* main() *****/
int main()
{
    clrscr();
    textmode(C4350);
    show_screen();
    get_input();
    apply_runge_kutta_method();
    show_result();
    getch();
    return 0;
}

/***************** Funcion Definitions *******/
//----- show_screen() -----
void show_screen()
{
    cprintf("\n*****");
    cprintf("*****-");
    cprintf("*-----");
    textbackground(1);
    cprintf(" Numerical Integration ");
    textbackground(8);
    cprintf(" -----*");
    cprintf("*-----*-");
    cprintf("*****-----*");
    for(int count=0;count<42;count++)
        cprintf("*-*");
    gotoxy(1,46);
    cprintf("*****-----*");
    cprintf("*-----*");
    cprintf("*****-----*");
    gotoxy(1,2);
}

}

```

```

***** ----- pop( ) ----- *****
//----- clear_screen( ) -----// *****
***** ----- const char* pop( ) -----
void clear_screen( )
{
    for(int count=0;count<37;count++)
    {
        gotoxy(5,8+count);
        cout<<"";
    }
    gotoxy(1,2);
}
***** ----- push(const char*) -----// *****
//----- void push(const char* Operand) -----
if(top==(max_size-1))
{
    cout<<"Error : Stack is full."<<endl;
    cout<<"\n      Press any key to exit.";
    getch();
    exit(0);
}
else
{
    strcpy(Operand,Stack[top]);
    strset(Stack[top],NULL);
    top--;
    return Operand;
}
***** ----- convert_ie_to_pe(const char*) ----- *****
//----- void convert_ie_to_pe(const char* Expression) -----
char Infix_expression[100]={NULL};
char Symbol_scanned[30]={NULL};
push("(");
strcpy(Infix_expression,Expression);
strcat(Infix_expression,"+0)");
int flag=0;
int count_1=0;
int count_2=0;

```

```

int equation_length=strlen(Infix_expression);
                                Symbol_scanned[0]=Infix_expression[count_1];
if(Infix_expression[0]=='(')
{
    flag=1;
                                count_1++;
do
{
    if(Infix_expression[count_1]!='(' &&
        Infix_expression[count_1]!='^' &&
        Infix_expression[count_1]!='*' &&
        Infix_expression[count_1]!='/')
    {
        int count_3=0;
        do
        {
            Infix_expression[count_1]!='+' &&
            Infix_expression[count_1]!='-' &&
            Infix_expression[count_1]!='.' &&
            Infix_expression[count_1]==')')
        {
            count_1++;
            count_3++;
            flag=0;
        }
        if(!Infix_expression[count_1-1]=='(' &&
            (Infix_expression[count_1]=='-'
            || Infix_expression[count_1]=='+'))
        {
            Infix_expression[count_1]!='+'
            && Infix_expression[count_1]!='-' &&
            Infix_expression[count_1]!='*' &&
            Infix_expression[count_1]!='/'
            && Infix_expression[count_1]!='^'
            && Infix_expression[count_1]!=')');
            flag=0;
        }
        if(strcmp(Symbol_scanned,"(")==0)
        {
            push("(");
            else if(strcmp(Symbol_scanned,")")==0)
            {
                while(strcmp(Stack[top],"(")!=0)
                {
                    strcpy(Postfix_expression[count_2],pop());
                    count_2++;
                }
                pop();
            }
            else if(strcmp(Symbol_scanned,"^")==0 ||
```
```



```

long double result=0;
char *endptr;
for(int count_2=0;count_2<2;count_2++)
{
    int flag=0;
    if(Operand[count_2][0]=='.')
    {
        int
length=strlen(Operand[count_2]);
        for(int
count_3=0;count_3<(length-1);count_3++)
            Operand[count_2][count_3]=Operand[count_2][(count_3+1)];
        Operand[count_2][count_3]=NULL;
        flag=1;
    }
    if(strcmp(Operand[count_2],"x")==0)
        operand[count_2]=x;
    else
if(strcmp(Operand[count_2],"y")==0)
        operand[count_2]=y;
    else
if(strcmp(Operand[count_2],"e")==0)
        operand[count_2]=2.718282;
    else
if(strcmp(Operand[count_2],"sinx")==0)
        operand[count_2]=sinl(x);
    else
if(strcmp(Operand[count_2],"cosx")==0)
        operand[count_2]=cosl(x);
    else
if(strcmp(Operand[count_2],"tanx")==0)
        operand[count_2]=tanl(x);
    else
if(strcmp(Operand[count_2],"lnx")==0)
        operand[count_2]=logl(x);
    else
if(strcmp(Operand[count_2],"logx")==0)
        operand[count_2]=log10l(x);
}
else
if(strcmp(Operand[count_2],"siny")==0)
    operand[count_2]=sinl(y);
else
if(strcmp(Operand[count_2],"cosy")==0)
    operand[count_2]=cosl(y);
else
if(strcmp(Operand[count_2],"tany")==0)
    operand[count_2]=tanl(y);
else
if(strcmp(Operand[count_2],"lny")==0)
    operand[count_2]=logl(y);
else
if(strcmp(Operand[count_2],"logy")==0)
    operand[count_2]=log10l(y);
else
    operand[count_2]=strtod(Operand[count_2],&endptr);
if(flag)
    operand[count_2]*=-1;
}
switch(Symbol_scanned[0])
{
    case '^':
result=powl(operand[1],operand[0]);
break;
    case '*':
result=operand[1]*operand[0];
break;
    case '/':
result=operand[1]/operand[0];
break;
    case '+':
result=operand[1]+operand[0];
break;
    case '-':
result=operand[1]-operand[0];
break;
}
gcvt(result,25,Result);

```

```

        push(Result);
    }

    else if(strcmp(Symbol_scanned,"")!=0)
        push(Symbol_scanned);

    }

while(strcmp(Symbol_scanned,"")!=0);

char Function_value[30]={NULL};

char *endptr;

strcpy(Function_value,pop( ));

function_value=strtod(Function_value,&endptr);

return function_value;

}

*****
//----- get_input( ) -----
*****


void get_input()
{
    clear_screen();

    gotoxy(6,10);

    cout<<"Input :";

    gotoxy(6,11);

    cout<<"IiiiiiiI";

    gotoxy(6,37);

    cout<<"Note : Write the function with proper Braces ( ) e.g;
2x+3 as (2*x)+3";

    gotoxy(6,40);

    cout<<"Available Operators : ^ (raised to power) , * , / , + , -
";
    gotoxy(6,42);

    cout<<"Available Operands : x , e , sinx , cosx , tanx , lnx ,
logx , ";

    gotoxy(6,44);

    cout<<"any number and all above functions
with y .";

    gotoxy(6,14);

    cout<<"Enter the Function : f(x,y) = ";
    cin>>Fx;

    convert_ie_to_pe(Fx);

    gotoxy(6,17);

    cout<<"Enter the value of xn = ";
    cin>>x;

    gotoxy(6,19);

    cout<<"Enter the value of x0 = ";
    cin>>x0;

    gotoxy(6,21);

    cout<<"Enter the value of y0 = ";
    cin>>y0;

    gotoxy(6,23);

    cout<<"Enter the value of h = ";
    cin>>h;

    }

    //----- apply_ruunge_kutta_method( ) -----
    *****

void apply_ruunge_kutta_method()
{
    xx[0]=x0;
    yy2[0]=y0;
    yy3[0]=y0;
    do
    {
        xx[(n+1)]=(xx[n]+h);
        n++;
    }
    while(xx[n]<x);

    long double k1=0;
    long double k2=0;
    long double ynp1=0;
    long double xn=x0;
}

```



```

        }

        cout<<"Estimated Value :";

        gotoxy(6,26);

        cout<<"|||||||||||||||||||||||";

        gotoxy(8,28);

        cout<<"Estimated Differential Value (using 2-Stage Method)
= "<<two_stage_result;

        gotoxy(8,30);

        cout<<"Estimated Differential Value (using 3-Stage Method)
= "<<three_stage_result;

        gotoxy(1,2);

    }

***** THE END *****

*****



MATLAB

function [wi, ti] = rk4 ( RHS, t0, x0, tf, N )

%RK4      approximate the solution of the initial value problem
%
%          x'(t) = RHS( t, x ),   x(t0) = x0
%
%          using the classical fourth-order Runge-Kutta method - this
%          routine will work for a system of first-order equations as
%          well as for a single equation
%
%          calling sequences:
%          [wi, ti] = rk4 ( RHS, t0, x0, tf, N )
%          rk4 ( RHS, t0, x0, tf, N )
%
%          inputs:
%          RHS  string containing name of m-file defining the
%          right-hand side of the differential equation; the
%          m-file must take two inputs - first, the value of
%          the independent variable; second, the value of the
%          dependent variable
%          t0  initial value of the independent variable
%          x0  initial value of the dependent variable(s)
%          if solving a system of equations, this should be a
%          row vector containing all initial values
%          tf  final value of the independent variable
%          N  number of uniformly sized time steps to be taken
%
%          to      advance the solution from t = t0 to t = tf
%
%          output:
%          wi  vector / matrix containing values of the
%          approximate
%          solution to the differential equation
%          ti  vector containing the values of the independent
%          variable at which an approximate solution has been
%          obtained
%
neqn = length ( x0 );
ti = linspace ( t0, tf, N+1 );
wi = [ zeros( neqn, N+1 ) ];

```

```

wi(1:neqn, 1) = x0';
k4 = h * feval ( RHS, t0 + h, x0 + k3 );
x0 = x0 + ( k1 + 2*k2 + 2*k3 + k4 ) / 6;
t0 = t0 + h;

```

```

for i = 1:N
    k1 = h * feval ( RHS, t0, x0 );
    k2 = h * feval ( RHS, t0 + h/2, x0 + k1/2 );
    k3 = h * feval ( RHS, t0 + h/2, x0 + k2/2 );

```

### Results of computation process

| steps   | True value | EM     | Milstein    | Adams  | RKM          |
|---------|------------|--------|-------------|--------|--------------|
| N=100   | 9.5675     | 0.7260 | 0.0488      | 0.5895 | 0.2065       |
| N=500   | 8.6583     | 0.0888 | 0.0119      | 0.1151 | 0.0207       |
| N=1000  | 10.0209    | 0.1474 | 0.0135      | 0.1650 | 0.0087       |
| N=2000  | 2.4827     | 0.0432 | 0.0018      | 0.0437 | 0.0044       |
| N=5000  | 2.0101     | 0.0185 | 7.0979e-004 | 0.0185 | 5.0137e-004  |
| N=10000 | 2.3081     | 0.0098 | 3.1517e-004 | 0.0099 | 2.9791e-004  |
| N=20000 | 2.9541     | 0.101  | 1.4628e-004 | 0.0102 | 1.3213e-004s |

Table 1: Test Problem 1

| steps   | True value | EM          | Milstein    | Adams       | RKM         |
|---------|------------|-------------|-------------|-------------|-------------|
| N=100   | 0.6401     | 0.0692      | 0.0030      | 0.0706      | 0.0067      |
| N=500   | 0.5774     | 0.0505      | 0.0012      | 0.0508      | 0.0037      |
| N=1000  | 0.6666     | 0.0212      | 4.8459e-004 | 0.0213      | 3.6176e-004 |
| N=2000  | -0.5304    | 0.0068      | 3.5143e-004 | 0.0068      | 1.9034e-004 |
| N=5000  | -0.6650    | 0.0038      | 1.5352e-004 | 0.0038      | 1.3588e-004 |
| N=10000 | -0.5807    | 6.9244e-004 | 7.7647e-005 | 7.0242e-004 | 4.7715e-005 |
| N=20000 | -0.3942    | 0.0015      | 6.3925e-005 | 0.0015      | 534108e-005 |

Table 2: Test Problem 2

| t       | True Solution | Numerical solution | Error    |
|---------|---------------|--------------------|----------|
| 0.00100 | 0.056400      | 0.056460           | 0.000060 |
| 0.10100 | 0.435042      | 0.415514           | 0.019528 |
| 0.20100 | -0.386871     | -0.452388          | 0.065468 |
| 0.30100 | 0.192301      | 0.077100           | 0.115201 |
| 0.40100 | 0.387541      | 0.274242           | 0.113298 |
| 0.50100 | 0.451076      | 0.314208           | 0.136868 |
| 0.60100 | 0.788344      | 0.742095           | 0.046249 |
| 0.70100 | 0.859078      | 0.810227           | 0.048859 |
| 0.80100 | 0.862022      | 0.827293           | 0.034729 |
| 0.90100 | 0.728410      | 0.637463           | 0.090947 |

Table 3: Numerical Results for Test problem 1-EM method

| t       | True Solution | Numerical solution | Error    |
|---------|---------------|--------------------|----------|
| 0.00100 | 0.056400      | 0.056460           | 0.000060 |
| 0.10100 | 0.435042      | 0.433431           | 0.001611 |
| 0.20100 | -0.386871     | -0.391438          | 0.004567 |
| 0.30100 | 0.192301      | 0.188191           | 0.004110 |
| 0.40100 | 0.387541      | 0.381777           | 0.005763 |
| 0.50100 | 0.451076      | 0.445563           | 0.005513 |
| 0.60100 | 0.788344      | 0.783381           | 0.004963 |
| 0.70100 | 0.859078      | 0.854925           | 0.004152 |
| 0.80100 | 0.862022      | 0.857197           | 0.004825 |
| 0.90100 | 0.728410      | 0.719102           | 0.009308 |

Table 4: Numerical Results for Test problem 1- Milsetne method

| t       | True Solution | Numerical solution | Error    |
|---------|---------------|--------------------|----------|
| 0.00100 | 0.056400      | 0.056460           | 0.000060 |
| 0.10100 | 0.435042      | 0.414970           | 0.020072 |
| 0.20100 | -0.386871     | -0.451468          | 0.064597 |
| 0.30100 | 0.192301      | 0.077443           | 0.114858 |
| 0.40100 | 0.387541      | 0.274027           | 0.113513 |
| 0.50100 | 0.451076      | 0.313985           | 0.137091 |
| 0.60100 | 0.788344      | 0.742675           | 0.045669 |
| 0.70100 | 0.859078      | 0.810882           | 0.048196 |
| 0.80100 | 0.862022      | 0.828288           | 0.033734 |
| 0.90100 | 0.728410      | 0.639080           | 0.089330 |

Table 5: Numerical Results for Test problem 1-Adams method

| t       | True Solution | Numerical solution | Error    |
|---------|---------------|--------------------|----------|
| 0.00100 | 0.056400      | 0.056360           | 0.000160 |
| 0.10100 | 0.435042      | 0.432942           | 0.002100 |
| 0.20100 | -0.386871     | -0.390630          | 0.003759 |
| 0.30100 | 0.192301      | 0.186226           | 0.002075 |
| 0.40100 | 0.387541      | 0.381169           | 0.004172 |
| 0.50100 | 0.451076      | 0.443172           | 0.003904 |
| 0.60100 | 0.788344      | 0.782031           | 0.002314 |
| 0.70100 | 0.859078      | 0.853644           | 0.001434 |
| 0.80100 | 0.862022      | 0.856262           | 0.001760 |
| 0.90100 | 0.728410      | 0.720592           | 0.007218 |

Table 6: Numerical Results for Test problem 1- RKM

| t       | True Solution | Numerical solution | Error  |
|---------|---------------|--------------------|--------|
| 0.00100 | 200570.451    | 200570.493         | 0.043  |
| 0.10100 | 205205.404    | 205192.581         | 12.824 |
| 0.20100 | 197007.965    | 197002.640         | 5.325  |
| 0.30100 | 203475.055    | 203459.580         | 15.475 |
| 0.40100 | 206240.448    | 206234.344         | 6.104  |
| 0.50100 | 207760.025    | 207770.608         | 10.582 |
| 0.60100 | 214772.536    | 214769.459         | 3.076  |
| 0.70100 | 218094.651    | 218109.320         | 14.669 |
| 0.80100 | 219173.738    | 219175.419         | 1.680  |
| 0.90100 | 216120.052    | 216138.085         | 18.033 |

Table 7: Numerical Results for Test problem 2-EM method

| t       | True Solution | Numerical solution | Error |
|---------|---------------|--------------------|-------|
| 0.00100 | 200570.451    | 200570.319         | 0.132 |
| 0.10100 | 205205.404    | 205203.676         | 1.728 |
| 0.20100 | 197007.965    | 197007.536         | 0.429 |
| 0.30100 | 203475.055    | 203472.498         | 2.557 |
| 0.40100 | 206240.448    | 206236.637         | 3.811 |
| 0.50100 | 207760.025    | 207755.177         | 4.848 |
| 0.60100 | 214772.536    | 214765.228         | 7.307 |
| 0.70100 | 218094.651    | 218085.764         | 8.887 |
| 0.80100 | 219173.738    | 219163.817         | 9.922 |
| 0.90100 | 216120.052    | 216110.190         | 9.862 |

Table 8: Numerical Results for Test problem 2-Milesten method

| t       | True Solution | Numerical solution | Error  |
|---------|---------------|--------------------|--------|
| 0.00100 | 200570.451    | 200570.493         | 0.043  |
| 0.10100 | 205205.404    | 205193.180         | 12.225 |
| 0.20100 | 197007.965    | 197002.259         | 5.706  |
| 0.30100 | 203475.055    | 203459.913         | 15.142 |
| 0.40100 | 206240.448    | 206235.183         | 5.264  |
| 0.50100 | 207760.025    | 207771.615         | 11.590 |
| 0.60100 | 214772.536    | 214771.361         | 1.174  |
| 0.70100 | 218094.651    | 218111.597         | 16.946 |
| 0.80100 | 219173.738    | 219177.797         | 4.058  |
| 0.90100 | 216120.052    | 216140.302         | 20.251 |

Table 9: Numerical Results for Test problem 2-Adams method

| t       | True Solution | Numerical solution | Error |
|---------|---------------|--------------------|-------|
| 0.00100 | 200570.451    | 200570.309         | 0.122 |
| 0.10100 | 205205.404    | 205204.176         | 1.029 |
| 0.20100 | 197007.965    | 197007.136         | 0.790 |
| 0.30100 | 203475.055    | 203471.832         | 1.223 |
| 0.40100 | 206240.448    | 206236.476         | 1.972 |
| 0.50100 | 207760.025    | 207755.163         | 2.822 |
| 0.60100 | 214772.536    | 214765.123         | 3.399 |
| 0.70100 | 218094.651    | 218086.038         | 4.613 |
| 0.80100 | 219173.738    | 219164.173         | 5.526 |
| 0.90100 | 216120.052    | 216112.403         | 7.649 |

Table 10: Numerical Results for Test problem 2- RKM

### Conclusion of evaluated numerical results

From the results presented in tables 1-10, we may notice that the Milstein and RKM (Runge kutta Method) methods out perform the other two in accuracy. For RKM (Runge kutta Method) method the errors decrease as N increases for all N values and all 2tests problems (with the exception of N=10000 and test problem 1). The same is more or less true for Milstein's method. For the other two methods (EM and Adams) the error increase for N = 20000 for test problem 1-3. Looking at tables 1-10where absolute values of the errors at different t values are given we notice the following. For Test 1, Milstein and Adams2 methods are the best and of comparable accuracy. For Test problem 2, all 4 methods are of similar accuracy. We may notice though that the EM and Adams method have oscillating errors.

In most cases, the Adams method and EM does not provide a better performance than the RKM (Runge kutta Method) method, in other words, for SDE with multiplicative noise, the Adams method is the same as the FDM method. Meanwhile the construction of code is much more complicated than the RKM (Runge kutta Method). The results confirms that when the stochastic error becomes more dominant, the adams method loses its power (compare also(Denk and schaffler [7])).According to (Denk and schaffler [7]). The main reason for the low accuracy of Adams method in the multiplicative case is that the high order of the deterministic convergence is surpassed by the low order of stochastic convergence. The RKM (Runge kutta Method)can overcome this problem well. The higher order term employed in stochastic part can benefit the order of convergence. The numerical results can confirm that the performance of RKM (Runge kutta Method) is much better than the original one.Overall, it appears that the RKM (Runge kutta Method) performed consistently better than all other methods with respect to accuracy with Milstein method second best.

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