## Homework 3, Computational Physics

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October 24, 2020

#### Abstract

This goal of this work was to analyze systems for which curve fittings and approximations were essential. The first problem we fit a curve to a specified function, the second, we used a summation relationship and its partial sums to approximate values for which the sum diverges.

### 1 Introduction

#### 1.1 Harmonic Oscillator Fit

We used a potential function V(x)

$$V(x) = -2 + 2(1 - \frac{1}{2}exp(-\frac{x}{2} + 2))^2$$

The objective for this problem was to use the potential function for a harmonic oscillator to fit a curve to the absolute minimum of the given function V(x).

#### 1.2 Sum Approximation

Given the following sum  $\zeta(s)$ 

$$\zeta(s) = \sum_{n=1}^{\infty} n^{-s}$$

The objective of this task was to calculate a few values of  $\zeta$  and use these to predict the value of  $\zeta$  when s=1.

## 1.3 Setup and General Methods

Fortran was used to analyze each system. To implement common mathematical terms and define the precision to which values were calculated a file was created. This file, named numtype, is shown below.

```
integer,parameter :: dp = selected_real_kind(15,307)
    integer,parameter :: qp = selected_real_kind(33,4931)
    real(dp), parameter :: pi = 4*atan(1._dp)
    complex(dp), parameter :: iic = (0._dp,1._dp)
    end module numtype
A Makefile, figure 3, was used to compile the fortran code and create and exe-
cutable file.
OBJS1 = numtype.o thielecf.o prob2.o
PROG1 = approx
F90 = gfortran
F90FLAGS = -03 -funroll-loops -fexternal-blas
LIBS = -framework Accelerate
LDFLAGS = $(LIBS)
all: $(PROG1)
$(PROG1): $(OBJS1)
$(F90) $(LDFLAGS) -o $@ $(OBJS1)
clean:
rm -f $(PROG1) *.{o,mod} fort.*
.SUFFIXES: $(SUFFIXES) .f90
.f90.o:
$(F90) $(F90FLAGS) -c $<
```

module numtype

## 2 Solutions

#### 2.1 Harmonic Oscillator Fit

First, the potential function V(x) was plotted to find the minimum.

$$V(x) = -2 + 2(1 - \frac{1}{2}exp(-\frac{x}{2} + 2))^{2}$$

The initial plot of the function did not make the minima immediately obvious, so we tooke the derivative of V(x) giving

$$dV(x)/dx = 7.389e^{-x}(e^{x/2} - 3.69453)$$

Finding the zero's of dV/dx narrowed our parameters to find the minimum of V(x) The following is the plot of the minimum of V(x)

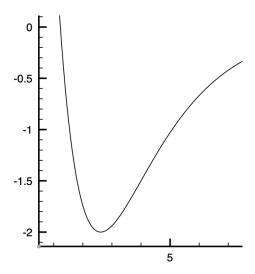


Figure 1: Minimum of V(x)

The goal was then to fit a standard harmonic oscillator potential function U(x) to the minimum of V(x).

$$U(x) = \frac{1}{2}a(x-b)^{2} + c$$

a, b, c are parameters that define and shape the harmonic potential function. An intial guess was made for each, a, b, and c based on the plot of V(x). These parameters were then used to iterate through a fitting algorithm to fit the harmonic oscillator potential, U(x), to the given potential function, V(x).

Below is the code used to genarte the plot, initialize and call the fitting

```
algorithim, and asses the strength of fit using a chi squared calculation.
```

```
module setup
    use numtype
    implicit none
    integer, parameter :: npmax = 400, npar = 3
    integer, parameter :: nspmin = 1, nspmax = 250
    real(dp) :: xx(1:npmax), yy(1:npmax)
    integer :: nsp, ical, iprint
end module setup
program problem_1
    use setup
    implicit none
    integer :: stat, i, itmin, itmax, in
    real(dp), external :: chi2
    real(dp) :: xstart(npar), fstart, stepi, epsf, dx
    xx(1) = 1.0_dp
                                                                        !-----
    dx = .02_dp
                                                                        !This loop populat
    do in = 1, 250
                                                                        !given function
        yy(in) = -2.0_dp + (2 * ((1 - (0.5*(exp((-xx(in)/2)+2))))**2))
        write(11,*) xx(in), yy(in)
                                                                        !used to graph the
        xx(in+1) = xx(in) + dx
    end do
    xstart(1:npar) = (/ 2.0_dp , 2.5_dp, -2.0_dp /)
```

 $xstart(1:npar) = (/ 0.5573_dp , 3.2226_dp, -2.0384_dp /)$ 

```
ical = 0
   iprint = 7
    fstart = chi2 (xstart)
   stepi= 0.05_dp
    epsf = 0.001_dp
    itmin = 100
    itmax = 1000
    iprint = 0
    call downhill(npar,chi2,xstart,fstart,stepi,epsf,itmin,itmax)
    iprint = 17
    fstart = chi2 (xstart)
   print *, xstart(1:npar)
end program problem_1
function chi2( par )
   use setup
    implicit none
   real(dp) :: chi2
   real(dp) :: par(npar)
    real(dp) :: K, mid, x, fi, height
    integer :: i
   ical = ical + 1
   K = par(1); mid = par(2); height = par(3)
                                                     !using harmonic function V(x) = 1/2
    chi2 = 0
                   ! chi^2
   do i = nspmin, nspmax
       x = xx(i)
       fi = 0.5_dp * K * (x - mid)**2 + height
                                                !harmonic function V(x) = 1/2 K (x)
        chi2 = chi2 + (yy(i) - fi)**2 * 1/sqrt( 2._dp + yy(i) )
    end do
    chi2 = chi2 / abs(nspmax-nspmin)
```

#### end function chi2

The fitting algorithm used the inital parameters and iterates through to optimize these parameters. It is called the downhill method and was used as follows

```
subroutine downhill(n,func,xstart,fstart,stepi,epsf,itmin,iter)
ı
   n
                dimension of the problem
!
   func
                function
   xstart
                starting values
   fstart
                conrespoding function value
   stepi
                relative stepsize for initial simplex
   epsf
                epsilon for termination
   itmin
                termination is tested if itmin < it
   iter
                maximum number of iterations
   use numtype
   implicit none
   integer :: n, iter, itmin
   real(dp), external :: func
   real(dp) :: xstart(1:n), fstart, stepi, epsf
   real(dp), parameter :: alph=1._dp, gamm=2._dp, &
                            rho=0.5_dp, sig=0.5_dp
```

```
real(dp) :: xi(1:n,1:n+1), x(1:n,1:n+1), &
    fi(1:n+1), f(1:n+1), &
    x0(1:n), xr(1:n), xe(1:n), xc(1:n), &
    fxr, fxe, fxc, deltaf
integer :: i, ii, it
xi(1:n,1) = xstart(1:n); fi(1) = fstart
do i = 2, n+1
    xi(1:n,i)=xi(1:n,1)
    xi(i-1,i)=xi(i-1,i)*(1+stepi)
    fi(i)=func(xi(1:n,i))
end do
do it = 1, iter
    do i = 1, n+1
                                            ! ordering
        ii = minloc(fi(1:n+1),dim=1)
        x(1:n,i) = xi(1:n,ii); f(i) = fi(ii)
        fi(ii) = huge(0._dp)
    end do
    xi(1:n,1:n+1) = x(1:n,1:n+1)
    fi(1:n+1) = f(1:n+1)
    x0(1:n) = sum(x(1:n,1:n),dim=2)/n
                                       ! central
    if ( itmin < it ) then
                                        ! condition for exit
        deltaf = (f(n)-f(1))
        !write(777,*) it,deltaf
        if(deltaf < epsf ) exit</pre>
    end if
    xr(1:n) = x0(1:n)+alph*(x0(1:n)-x(1:n,n+1))
    fxr = func(xr)
    if( fxr < f(n) .and. &
                                        ! reflection
            f(1) \le fxr ) then
        xi(1:n,n+1) = xr(1:n); fi(n+1) = fxr
        cycle
    else if (fxr < f(1)) then
                                        ! expansion
        xe(1:n) = x0(1:n)+gamm*(x0(1:n)-x(1:n,n+1))
```

```
fxe = func(xe)
        if( fxe < fxr ) then
            xi(1:n,n+1) = xe(1:n); fi(n+1) = fxe
            cycle
        else
            xi(1:n,n+1) = xr(1:n); fi(n+1) = fxr
            cycle
        end if
    else if (fxr >= f(n)) then
                                      ! contraction
        xc(1:n) = x(1:n,n+1)+rho*(x0(1:n)-x(1:n,n+1))
        fxc = func(xc)
        if( fxc \le f(n+1) ) then
            xi(1:n,n+1) = xc(1:n); fi(n+1) = fxc
            cycle
        else
                                             ! reduction
            do i = 2, n+1
                xi(1:n,i) = x(1:n,1)+sig*(x(1:n,i)-x(1:n,1))
               fi(i) = func(xi)
            end do
            cycle
        end if
    end if
end do
xstart(1:n)=xi(1:n,1); fstart = fi(1)
```

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end subroutine downhill

## 2.2 Sum Approximation

We analyzed the function

$$\zeta(s) = \sum_{n=1}^{\infty} n^{-s}$$

The sum  $\zeta(1)$  diverges, so we used other local convergent s values to make an approximation for  $\zeta(1)$  by means of a continuous fraction algorithm.

We used s values from 2 to 4 and calculated each corresponding  $\zeta(s)$  sum. The s and  $\zeta(s)$  values were input into the continuos fraction algorithm to then evaluate  $\zeta(1)$ .

The sums were calculated as follows:

```
program prob_2
use numtype
use thiele_approx
implicit none
integer, parameter :: np = 11
integer :: inmax, in, i
real(dp), dimension(1:np) :: sums, xx, yy
real(dp) :: x
inmax = 10**7
sums = 0
do in = 1,inmax
                                             !itereate sums for function chi(s)
    sums(1) = sums(1) + 1/(in**2.0_dp)
                                             !chi(s) = sum from 1 to inf. of n^-s
    sums(2) = sums(2) + 1/(in**2.2_dp)
                                             !at various values of s
    sums(3) = sums(3) + 1/(in**2.4_dp)
    sums(4) = sums(4) + 1/(in**2.6_dp)
    sums(5) = sums(5) + 1/(in**2.8_dp)
    sums(6) = sums(6) + 1/(in**3.0_dp)
    sums(7) = sums(7) + 1/(in**3.2_dp)
    sums(8) = sums(8) + 1/(in**3.4_dp)
    sums(9) = sums(9) + 1/(in**3.6_dp)
    sums(10) = sums(10) + 1/(in**3.8_dp)
    sums(11) = sums(11) + 1/(in**4.0_dp)
```

#### end do

```
xx(1) = 2.0_dp
                                            !chosen values for s to analyze sum
xx(2) = 2.2_dp
                                             !to input into thiele cf approximation
xx(3) = 2.4_dp
xx(4) = 2.6_dp
xx(5) = 2.8_dp
xx(6) = 3.0_dp
xx(7) = 3.2_dp
xx(8) = 3.4_dp
xx(9) = 3.6_dp
xx(10) = 3.8_dp
xx(11) = 4.0_dp
yy(1) = sums(1)
                                             !resulting sums of corresponding s values
yy(2) = sums(2)
                                            !yy is a function of xx
yy(3) = sums(3)
yy(4) = sums(4)
yy(5) = sums(5)
yy(6) = sums(6)
yy(7) = sums(7)
yy(8) = sums(8)
yy(9) = sums(9)
yy(10) = sums(10)
yy(11) = sums(11)
write(10,*) xx, yy
                                            !make sure s values and partial sums are g
call thiele_coef( np, xx, yy, an )
                                            !generate continued fraction coefficients
x = 1.0_dp
print *, x, thiele_cf (x, np, xx, an)
                                       !use cf coeffs to evalutae the function at
```

#### end program

The continued fraction method used was the Thiele method and this was the code used:

# module thiele\_approx

```
use numtype
implicit none
integer, parameter :: maxpt = 50
    real(dp), dimension(maxpt) :: zn, fn, an
contains
        subroutine thiele_coef( nn, zn, fn, an )
        ! coefficients of Thiele continued fraction
            use numtype
            implicit none
            real(dp), dimension(maxpt) :: zn, fn, an
            real(dp), dimension(maxpt,maxpt) :: gn
            integer :: nn, n, nz
            gn(1,1:nn) = fn(1:nn)
            do n = 2, nn
                do nz = n, nn
                    gn(n,nz) = (gn(n-1,n-1) - gn(n-1,nz)) / &
                        ((zn(nz)-zn(n-1))*gn(n-1,nz))
                end do
            end do
            forall ( n = 1:nn ) an(n) = gn(n,n)
        end subroutine thiele_coef
        function thiele_cf (z, nn, zn, an) result(cfrac)
        ! evaluate the Thiele continued fraction
            use numtype
            implicit none
            real(dp) :: z
            real(dp), dimension(maxpt) :: zn, an
            integer :: nn, n
            real(dp) :: cf0(2), cf1(2), cf(2), cfrac
            cf0(1) = 0._dp; cf0(2) = 1._dp
```

```
cf1(1) = an(1); cf1(2) = 1._dp
do n = 1, nn-1
    cf = cf1 + (z - zn(n)) * an(n+1) * cf0
    cf0 = cf1; cf1 = cf
end do
cfrac = cf(1)/cf(2)
```

end function thiele\_cf

end module thiele\_approx

## 3 Results

#### 3.1 Harmonic Oscillator Fit

With x ranging from 1 to 6 the resulting  $Chi^2$  value was 0.1650 and the plots are shown in figure 2.

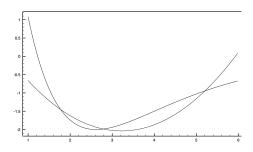


Figure 2: U(x) fit to V(x)

To get a better fit, the range of x would need to be shrunk, closer to the absolute minimum of V(x).

## 3.2 Sum Approximation

Using  $10^8$  terms in each sum and eleven pairs of s and  $\zeta(s)$  values,  $\zeta(1)$  was approximated to be

$$\zeta(1) = 542548.10289$$