

GOKHALE EDUCATION SOCIETY'S



**N. B. MEHTA (V) SCIENCE COLLEGE**

ACHARYA BHISE VIDYANAGAR, BORDI

TAL - DAHANU, DIST. PALGHAR, 401701, Tel. 02528 - 254357



**DEPARTMENT OF INFORMATION TECHNOLOGY  
AND  
COMPUTER SCIENCE**

# Certificate

Class \_\_\_\_\_

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Uni. Exam No. \_\_\_\_\_ has satisfactorily completed the required number of  
practical and worked for the 1st term / 2nd term/ both the terms of the Year  
\_\_\_\_\_ in the college laboratory as laid down by the university.

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## Practical 1

Aim :- Implement Contour Plots

Code :-

using PlotlyJS

```
plot(contour(
```

```
z=[
```

```
10 10.625 12.5 15.625 20
```

```
5.625 6.25 8.125 11.25 15.625
```

```
2.5 3.125 5. 8.125 12.5
```

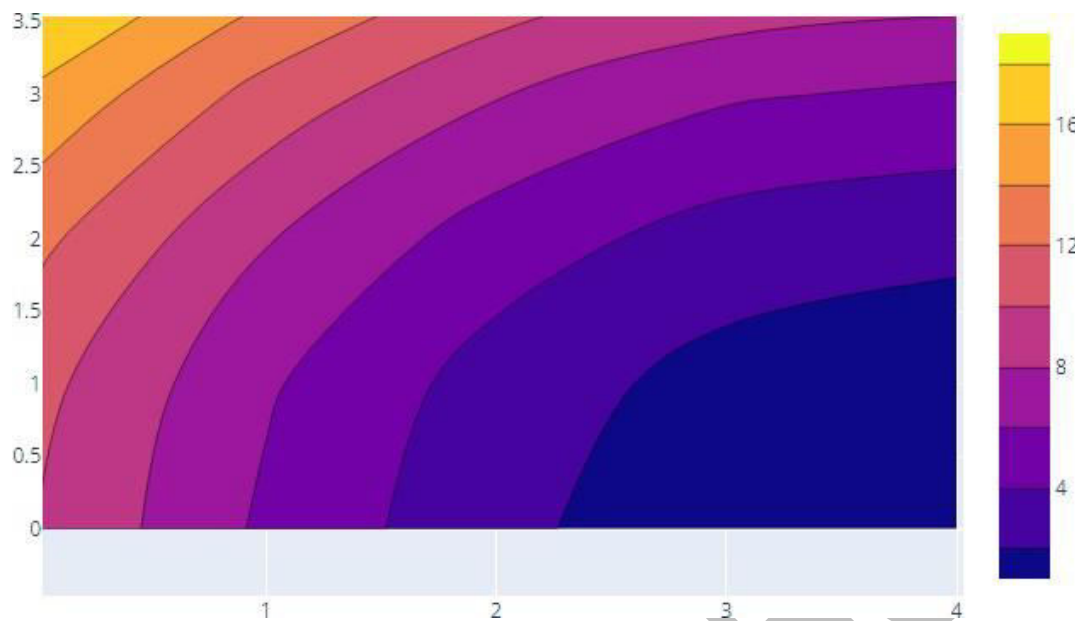
```
0.62 1.25 3.125 6.25 10.625
```

```
0 0.625 2.5 5.625 10
```

```
]'
```

```
))
```

OutPut:-



## Practical 2

Aim :- Implement Fibonacci and Golden section search.

### # Fibonacci section search

Code:-

```
function fibonacci_section_search(f, a, b, n, g=0.01)
    s=(1-0.5)/(1+0.5)
    p=1/(1.618*(1-s^(n+1))/(1-s^n))
    d=p*b+(1-p)*a
    yd=f(d)
    for i in 1:n-1
        print(a)
```

```

    print("/n")
    print(b)
    print("/n")
    if i==n-1
        c=g*a+(1-g)*d
    else
        c=p*a+(1-p)*b
    end
    yc=f(c)
    if yc<yd
        b,d,yd=d,c,yc
    else
        a,b=b,c
    end
    p=1(1.618*(1-s^(n-i+1))/(1-s^(n-i)))
end
return a < b ? (a, b) : (b, a)end

```

**# fibonacci\_search (generic function with 3 methods)**

```

function f(x)
    return x*x-x+1
end1
end

```

**# f (generic function with 1 method)**

Result=fibonacci\_section\_search(f,-1,1,10)

**OutPut:-**

(-0.011235955056179792 , 0.011235955056179796)

**# Golden section search****Code:-**

```
function golden_section_search(f, a, b, n)
```

```
    p=1.618-1
```

```
    d=p*b+(1-p)*a
```

```
    yd=f(d)
```

```
    for i = 1 : n - 1
```

```
        c=p*a+(1-p)*b
```

```
        yc=f(c)
```

```
        if yc<yd
```

```
            b, d, yd=d, c, yc
```

```
        else
```

```
            a, b=b, c
```

```
        end
```

```
    end
```

```
    return a < b ? (a, b) : (b, a)
```

```
end
```

**# golden\_section\_search (generic function with 1 method)**

```
function f(x)
```

```
    return x*x
```

```
end
```

**# f (generic function with 1 method)**

```
result=golden_section_search(f,2,8,5)
```

**Output :-**

(2, 2.8755439999999997)

**Practical 3**

**Aim:- Implement Quadratic Fit Search**

**Code:-**

```
function quadratic_fit_search(f,a,b,c,n)
    ya,yb,yc=f(a),f(b),f(c)
    for i in 1:n-3
        print(a,"/n",b,"/n",c,"/n")
        x=0.5*(ya*(b^2-c^2)+yb*(c^2-a^2)+yc*(a^2 -
b^2))/(ya*(b-c)+yb*(c-a)+yc*(a-b))
        yx=f(x)
        if x > b
            if yx > yb
                c, yc = x, yx
            else
                a, ya, b, yb = b, yb, x, yx
            end
        elseif x < b
            if yx > ya
                a, ya = x, yx
            else
```

```

        c, yc, b, yb = b, yb, x, yx
    end
end
end
return (a, b, c)
end
# quadratic_fit_search (generic function with 1 method)

function f(n)
    return n*n+2*n-1
end
# f (generic function with 1 method)

result=quadratic_fit_search(f,1,6,10,5)

```

**Output :-**

```
1/n6/n10/n1/n-1.0/n6/n(1, -1.0, 6)
```

## Practical 4

**Aim :-** Implement Gradient descent

**Code:-**

```

function gradient_descent(p, q, x1; a=0.1,
maxiter=1000, g=1e-5)
    x-copy(x1)

```



```
f=x -> 0 * x + q
x2= -f(x)
iter = 0
while norm(x2) > g || iter <= maxiter
    iter +=1
    x += a * x2
    x2 = -f(x)
end
return x
end
# gradient_descent (generic function with 1 method)
```

```
p=[10.0 -1.0;
   -1.0 1.0 ];
q=[0; -10.0];
x2=zeros(2);
```

**Output :-**

```
1.1111111111111103
11.111111111111104
```

## Practical 5

**Aim :-** Implement quasi-Newton methods to find the local maxima

Code:-

```
function newtonsMethodForUnivariate (x_guess,
max_iter)
    f_1=2 * x_guess
    f_2=2
    converged = false
    iter = 0
    while converged == false
        x_optimum = x_guess - (f_1/f_2)
        x_guess = x_optimum
        println("Iteration : $iter, Current Guess:
        $x_guess")
        if x_guess - 1 < 0.01
            converged = true
        end
        if iter > max_iter
```

```
        converged = true
    end
    iter=iter+1
end
end
# newtonsMethodForUnivariate (generic
function with 1 method)
newtonsMethodForUnivariate (3,100)
```

## Output :-

Iteration : 0, Current Guess: 0.0

## Practical 6

**Aim :-** Implement the Adagrad method with application, RMSprop and Adadelta.

### # Adagrad method with application

Code:-

```
function ada_grad(x_guess, max_iter, alpha)
    fd=2 * x__guess - 2
```

```
Converged = false
Iter = 0
prev_sgs = 0
while converged == false
    delta = alpha * fd
    sgs = prev_sgs + (fd)^2

    x_optimum = x_guess - delta/sqrt(sgs)
    x_guess = x_optimum
    prev_sgs = sgs
    println("Iteration : $iter, current Guess: $x_guess")
    if x_guess - 1 < 0.01
        converged = true
    end
    if iter > max_iter
        converged = true
    end
    iter = iter+1
end
end
```

**# ada\_grad (generic function with 1 method)**

```
ada_grad(3,20000,0.01)
```

### **Output :-**

Iteration : 10043, current Guess : 1.0101584845215472

Iteration : 10044, current Guess : 1.0100587087650088

Iteration : 10045, current Guess : 1.0099589379745384

## #RMSprop

Code :-

```
function rms_prop(x_guess, max_iter, alpha, beta)

    fd = 2 * x_guess - 2
    converged = false
    iter = 0
    prev_sgs = 0
    while converged == false
        delta = alpha * fd
        sgs = (prev_sgs * beta) + ((fd)^2) * (1-beta)
        x_optimum = x_guess - delta/sqrt(sgs)
        x_guess = x_optimum

        prev_sgs = sgs
        println("Iteration : $iter, Current_Guess : $x_guess")
        if x_guess - 1 < 0.01
            converged = true
        end
        if iter > max_iter
            converged = true
        end
        iter = iter + 1
    end
end
```

**# rms\_prop (generic function with 1 method)**

```
rms_prop(3, 4000, 0.01, 0.99)
```

**Output:-**

Iteration:95, Current\_Guess: 1.0244024028520233

Iteration:96, Current\_Guess: 1.0117305980069258

Iteration:97, Current\_Guess: 0.9990969991751104

**# Adadelta****Code:-**

```
function ada_delta(x_guess, max_iter, gamma)
    fd= 2 * x_guess - 2
    converged = false
    iter = 0
    prev_sgs = 0
    Ex = 0
    ep = 1e - 5
    while converged == false
        sgs = (gamma * prev_sgs) + ((1-gamma) * (fd^2))
        rms_g = sqrt(sgs + ep)
        rms_x = sqrt(Ex + ep)
        x = (rms_x/rms_g) * fd
        Ex = (gamma * Ex) + ((1-gamma) * (x^2))

        prev_sgs = sgs
    x_optimum = x_guess - x
```

```
x_guess = x_optimum
println("Iteration : $iter, Current_Guess : $x_guess")

if x_guess - 1 < 0.00001
    converged = true
end
if iter>max_iter
    converged = true
end
iter = iter+1
end
end
```

**# ada\_delta (generic function with 1 method)**

```
ada_delta(3, 400, 0.9)
```

**Output :-**

```
Iteration :143 , Current_Guess :1.03482265980986482
Iteration :144, Current_Guess :1.0186375926957296
Iteration :145, Current_Guess :1.002417732057021
Iteration :146, Current_Guess :1.0861670747700315
```

## Practical 7

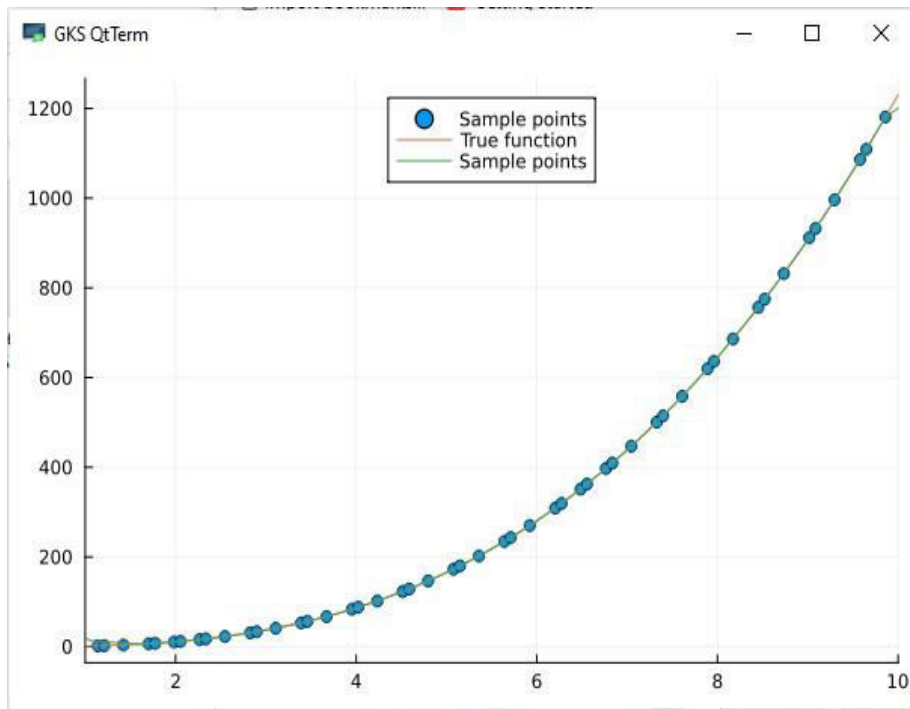
**Aim :-** Implement radial basis functions using surrogate modelling.

**Code:-**

```
using Surrogates
using Plots
f=x -> log(x) * x^2 + x^3
ib=1.0
ub=10.0
x = sample(50, ib, ub, SobolSample())
y = f.(x)
my_radial_basis = RadialBasis(x,y,ib,ub)
approx = my_radial_basis(5.4)
using Plots
plot(x, y, seriestype=:scatter, label="Sample points",
xlims=(ib, ub), legend=:top)
plot!(f, label="True function", xlims=(ib,ub), legend=:top)
plot!(my_radial_basis, label="Sample points", xlims=(ib, ub),
legend=:top)
```

**Output:-**





## Practical 8

**Aim:-** Apply Random Forest in surrogate Model.

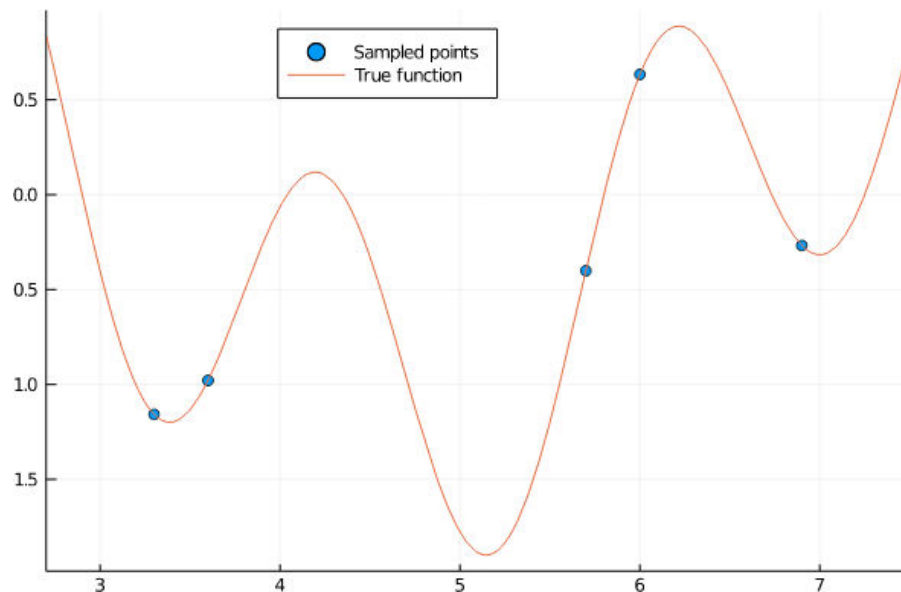
**Code:-**

```
using Surrogates
using Plots
```

```
f(x) = sin(x) + sin(10/3 * x)
n_samples = 5
lower_bound = 2.7
upper_bound = 7.5
x = sample(n_samples, lower_bound,
upper_bound, SobolSample())
y = f.(x)
scatter(x, y, label="Sampled points",
xlims=(lower_bound, upper_bound))
```

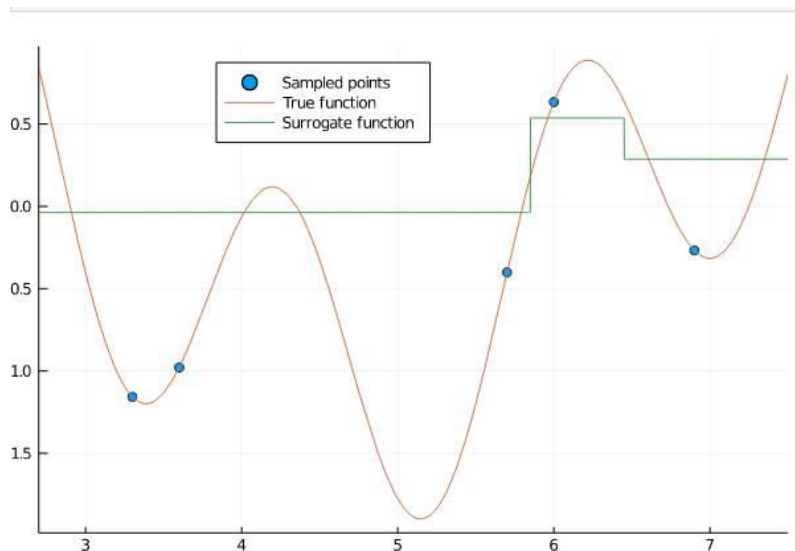
```
plot!(f, label="True function",
xlims=(lower_bound, upper_bound),
legend=:top)
```

Output: -



```
num_round = 2
randomforest_surrogate = RandomForestSurrogate(x ,y ,lower_bound,
upper_bound, num_round = 2)
plot(x, y, seriestype=:scatter, label="Sampled points",
xlims=(lower_bound, upper_bound), legend=:top)
plot!(f, label="True function", xlims=(lower_bound, upper_bound),
legend=:top)
plot!(randomforest_surrogate, label="Surrogate function",
xlims=(lower_bound, upper_bound), legend=:top)
```

**Output:**



## Practical 9

**Aim:** - Implement Gaussian Process and its application.

Code:-

```
using GaussianProcesses
using Random
Random.seed!(20140430)
n=10
x=2*3.14*rand(n)
y=sin.(x)+0.05*rand(n);
mZero=MeanZero()
kern=SE(0.0, 0.0)
logObsNoise=-1.0
gp=GP(x, y, mZero, kern, logObsNoise)
x=0:0.1:2*3.14
```

### Method1

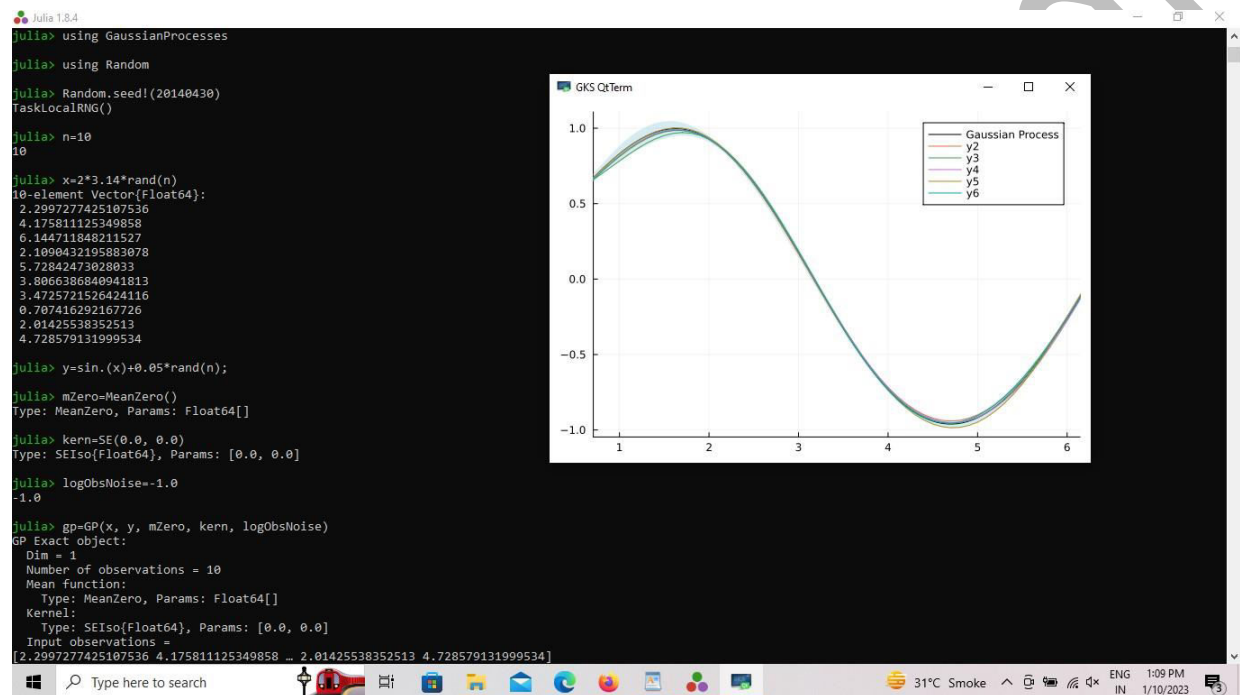
```
using Plots
plot(gp; obsv=false)
```

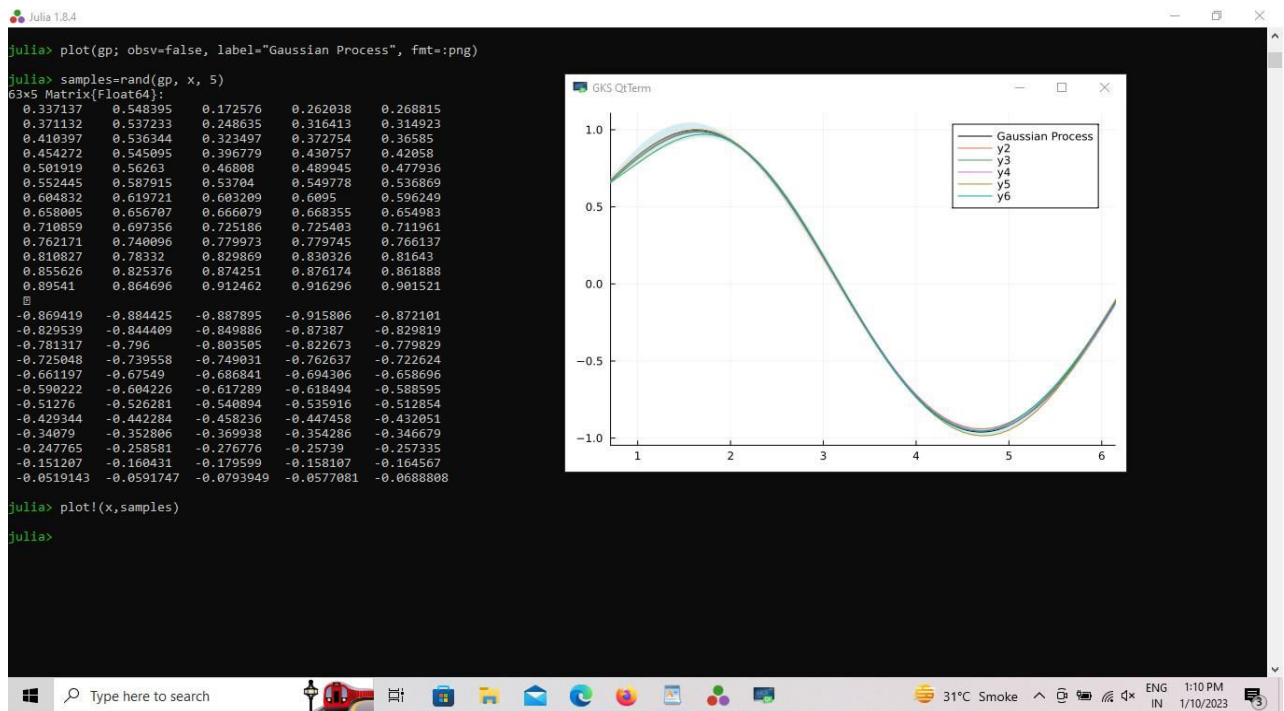
```

optimize!(gp)
plot(gp; obsv=false, label="Gaussian Process", fmt=:png)
samples=rand(gp, x, 5)
plot!(x,samples)

```

**Output:-**



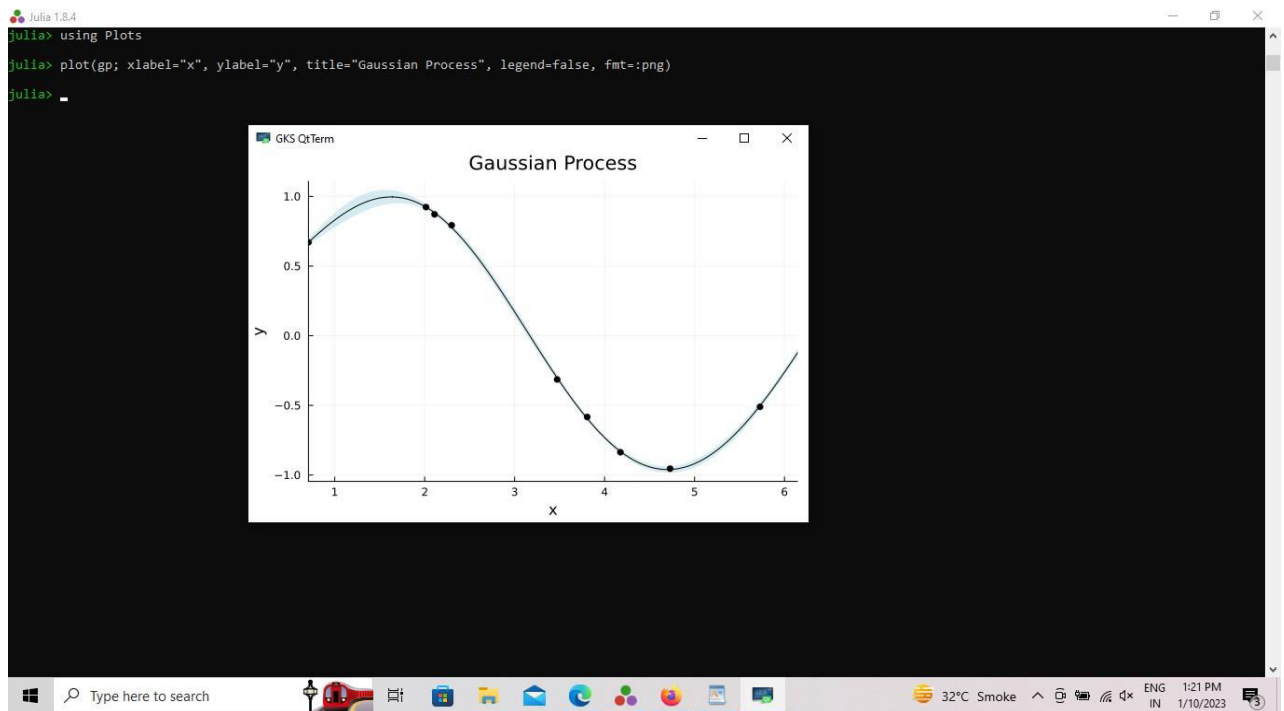


## Method2

Code:-

using Plots

plot(gp; xlabel="x", ylabel="y", title="Gaussian Process",  
legend=false, fmt=:png)



## Practical 10

**Aim :-** Path finding using Ant Colony Optimization with an application

**Code:-**

```
using AntColony
distance_matrix = rand(10,10)
aco(distance_matrix, is_tour=true)
aco(distance_matrix, start_node=1, end_node=5)
```

**Output:-**

```
Julia 1.8.4
1 dependency successfully precompiled in 20 seconds. 211 already precompiled.

julia> using AntColony

julia> distance_matrix = rand(10,10)
10x10 Matrix{Float64}:
0.0670739 0.627131 0.788468 0.00377026 0.944017 0.84528 0.764806 0.538912 0.534518 0.421503
0.898901 0.715016 0.384268 0.903681 0.208158 0.991979 0.934606 0.00386285 0.0428881 0.553988
0.14423 0.533598 0.245043 0.342453 0.785317 0.201394 0.842878 0.267902 0.234743 0.0332262
0.361415 0.511839 0.0604465 0.169706 0.526274 0.0525583 0.517447 0.427529 0.909036 0.215229
0.164391 0.326585 0.652173 0.581036 0.0157661 0.413838 0.182987 0.119006 0.299029 0.760263
0.211952 0.367053 0.258278 0.464581 0.612962 0.0916417 0.631333 0.0639033 0.33035 0.260968
0.641993 0.797584 0.413586 0.923831 0.392246 0.104878 0.470376 0.512995 0.784756 0.795832
0.410777 0.641526 0.520102 0.843813 0.308037 0.0502845 0.100142 0.599046 0.294033 0.68932
0.589147 0.858768 0.065627 0.465361 0.681022 0.379947 0.473732 0.021474 0.342582 0.695088
0.962903 0.810556 0.803052 0.200556 0.149126 0.462516 0.143565 0.424315 0.947835 0.118322

julia> aco(distance_matrix, is_tour=true)
10-element Vector{Int64}:
6
4
1
5
10
3
9
2
7
8

julia> aco(distance_matrix, start_node=1, end_node=5)
10-element Vector{Int64}:
1
4
10
3
9
2
8
6
7
5

julia>
```