

# PSI Numerical Methods

Winter 2024

## Homework Assignment 2: Partial Differential Equations

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The main goal of this homework is to solve the scalar wave equation

$$\partial_t^2 \phi = \Delta \phi$$

on the surface of a sphere. To discretize  $\phi(t, \theta, \varphi)$  in space, we recommend using spherical harmonic functions  $Y_{lm}$ . The Laplace operator for spherical harmonics is a diagonal operator given by the factor  $-l(l+1)$  for each  $c^{lm}$  (i.e., it is independent of the mode number  $m$ ). This will lead to a system of ODEs for the coefficient vector  $c^{lm}$ .

To discretize  $\phi$  in time, we want to use standard ODE methods. These standard methods assume that there is only a first derivative in time. We thus introduce a new function  $\psi = \partial_t \phi$ , arriving at the system

$$\partial_t \phi = \psi$$

$$\partial_t \psi = \Delta \phi$$

Both functions  $\phi$  and  $\psi$  need to be expanded in spherical harmonics, i.e., the coefficient vector is now twice as long.

Note: You can use the package `SphericalFunctions` to obtain the spherical harmonics functions  $Y_{lm}$

## Questions

(a) Numerically implement the discretization of  $\phi$  in terms of spherical harmonics.

To start, we decompose the function  $\phi$  in terms of spherical harmonics, which form a complete orthonormal basis:

$$\phi(t, \theta, \varphi) = c^{lm} Y_{lm}(\theta, \varphi)$$

Inserting this ansatz into the Poisson equation, we end up with the second order ordinary differential equation (ODE):

$$\begin{aligned} \ddot{c}^{lm}(t) Y_{lm}(\theta, \varphi) &= -l(l+1) c^{lm}(t) Y_{lm}(\theta, \varphi) \\ \implies \ddot{c}^{lm}(t) &= -l(l+1) c^{lm}(t) \end{aligned}$$

This appears to be a much nicer equation to solve. But this hides quite a bit of complexity. For instance, we need infinitely many  $l, m$  values in principle to get a perfect decomposition of  $\phi$ . These correspond to  $l \times m$  sets of differential equations. In our code, we cap the number of terms in the expansion.

```
In [1]: # The most miserable part of the whole code (loading packages)
# using Pkg
# Pkg.add("SphericalHarmonics")
# Pkg.add("HCubature")
# Pkg.add("DifferentialEquations")
# Pkg.add("PyCall")
# using PyCall
# Pkg.add("PyPlot")
using Plots; pyplot()
# using PyPlot
using SphericalHarmonics
using HCubature
using DifferentialEquations
```

```
In [2]: # Spherical coordinate definition
N = 100 # Spherical grid size
θ = LinRange(0, π, N);
φ = LinRange(0, 2π, N);
# Cartesian projections of spherical coordinate (θ, φ')
X = sin.(θ) .* cos.(φ');
Y = sin.(θ) .* sin.(φ');
Z = cos.(θ) .* ones(N)';
```

(b) Use an initial condition that is peaked around the North Pole, i.e., that looks similar to a Gaussian with a width equal to 0.2. (The exact initial condition does not matter).

```
In [3]: # Function to extract Y_lm coefficients given an initial distribution
# Init = Initial Condition, the Gaussian in our case
# l_max = Number of terms l in C_lm expansion
function C_lm(Init, l_max)
    coefficients = hcubature(x -> computeYlm(x[1], x[2], l_max) * Init(x[1], x[2])
    return coefficients
end

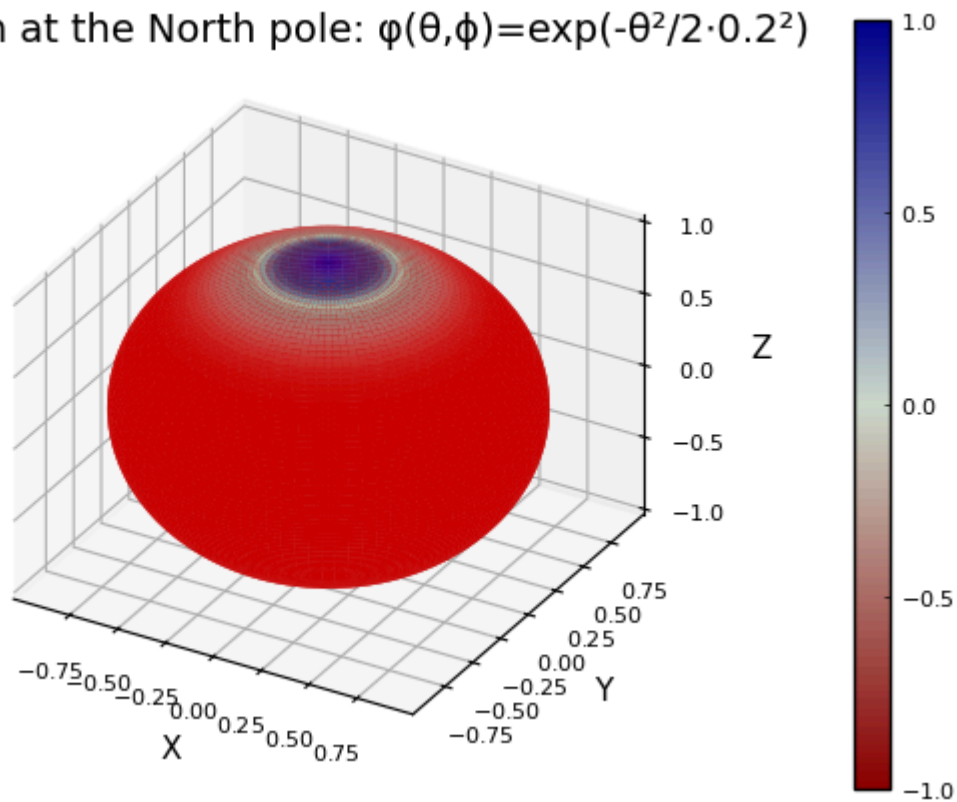
C_lm (generic function with 1 method)
```

```
In [4]: # Generates the Gaussian initial condition for a given ( $\theta, \phi$ ) coordinate  
function Gaussian( $\theta, \phi$ )  
    return  $\exp(-\theta^2 / (2 * 0.2)^2)$   
end  
  
Gaussian (generic function with 1 method)
```

```
In [5]: Phi_gauss = zeros(Complex{Float64}, 100, 100)  
  
for i in 1:100  
    for j in 1:100  
        Phi_gauss[i,j] = Gaussian( $0 + \pi/100 * i$ ,  $0 + \pi/50 * j$ )  
    end  
end
```

```
In [6]: surface(X,Y,Z,fill_z=real.(Phi_gauss),c=:redsblues,xlabel="X",ylabel="Y",zlabel=
```

Gaussian at the North pole:  $\phi(\theta,\phi)=\exp(-\theta^2/2\cdot 0.2^2)$



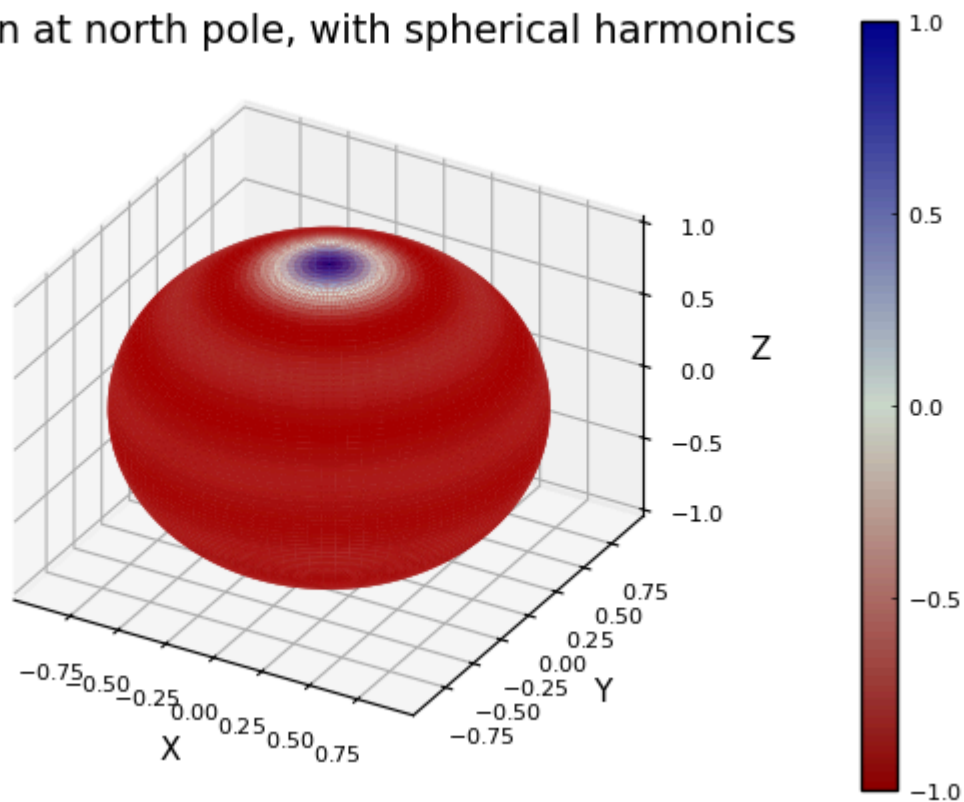
The above plot shows what a Gaussian (with azimuthal symmetry) looks like on a sphere. We will now attempt to recreate this using the  $Y_{lm}$  (spherical harmonics) decomposition.

```
In [7]: Phi_gauss = C_lm(Gaussian,10)

Phi_t = zeros(Complex{Float64}, 100, 100)
for l in 1:100
    for m in 1:100
        Phi_t[l,m] = Phi_gauss' * computeYlm(0 + pi/100 * l, 0 + pi/50 * m, 10) /
    end
end
```

```
In [8]: surface(X,Y,Z,fill_z=real.(Phi_t),c=:redsblues,xlabel="X",ylabel="Y",zlabel="Z",
```

Gaussian at north pole, with spherical harmonics



Comparing the 2 figures above, we find that using  $l_{max} = 10$ , we can successfully recreate the Gaussian initial condition.

**(c) Evolve the system in time to see from  $t = 0$  to  $t = 10$  using your favorite ODE integrator. The resulting evolution should look similar to water waves moving on the surface of a pond, except that the pond is the surface of a sphere.**

For a given value of  $l$ , we have to consider  $(2l + 1)$   $m$  values. This corresponds to  $2l + 1$  system of ODEs.

```
In [9]: function lm(l_max)
        llmm = []
        for i in 1:l_max + 1
            append!(llmm, fill(i - 1, 2i - 1))
        end
        return llmm
    end
```

lm (generic function with 1 method)

Next, we define the set of ODEs that have to be solved. We have the second-order equation

$$\ddot{c}^{lm}(t) = -l(l + 1)c^{lm}(t)$$

Which we write as 2 coupled first order equations

$$\dot{c}^{lm} = u^{lm}$$

$$\dot{u}^{lm} = -l(l + 1)c^{lm}$$

We specified the Gaussian initial condition and set the initial velocities randomly.

```
In [10]: function ODE_system!(dc, c, p, t)
          l = p[1]
          dc[1] = c[2] # First derivative of c
          dc[2] = - l * (1 + 1) * c[1] # Second derivative of c
        end

        ODE_system! (generic function with 1 method)
```

```
In [11]: l_max = 20;

          # Initial conditions in terms of coefficients C_lm
          C0 = C_lm(Gaussian, l_max);
          Der_C0 = rand(length(C0));

          # Time range for solution
          trange = (0.0, 10.0);

          # Defining l,m values for each c
          lm_ = lm(l_max);
```

```
In [12]: # Array to store solutions
          SolnArray = []

          # Solve the ODE for each l,m value
          for i in 1:length(lm_)
            l = lm_[i]
            prob = ODEProblem(ODE_system!, [C0[i], Der_C0[i]], trange, [1])
            Sol = solve(prob)
            push!(SolnArray, Sol)
          end
```



```
In [13]: # Ordering the solutions in time with time-step dt
dt=0.1
C_t = []

for i in 1:length(lm_)
    push!(C_t,[x[1] for x in real.(SolnArray[i](0:dt:10))])
end

C_t = hcat(C_t...);
```

```
In [14]: # Converting coefficients C_lm(t) into function  $\phi(t)$ 
Solutions = zeros(101,100,100)
for k in 1:101
    for l in 1:100
        for m in 1:100
            Solutions[k,l,m] = C_t[k,:]' * real.(computeYlm(0 + pi/100 * l, 0 + pi/100 * m))
        end
    end
end
```

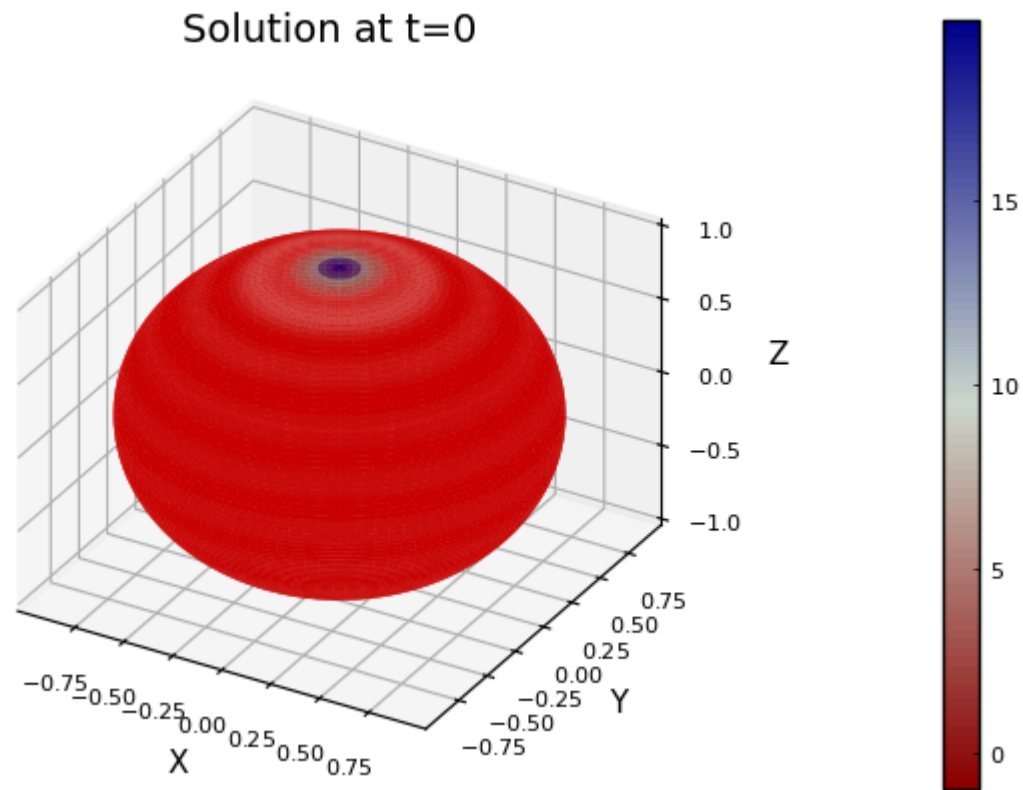
```
In [15]: # Plotting solutions at different times

# plots_array = []

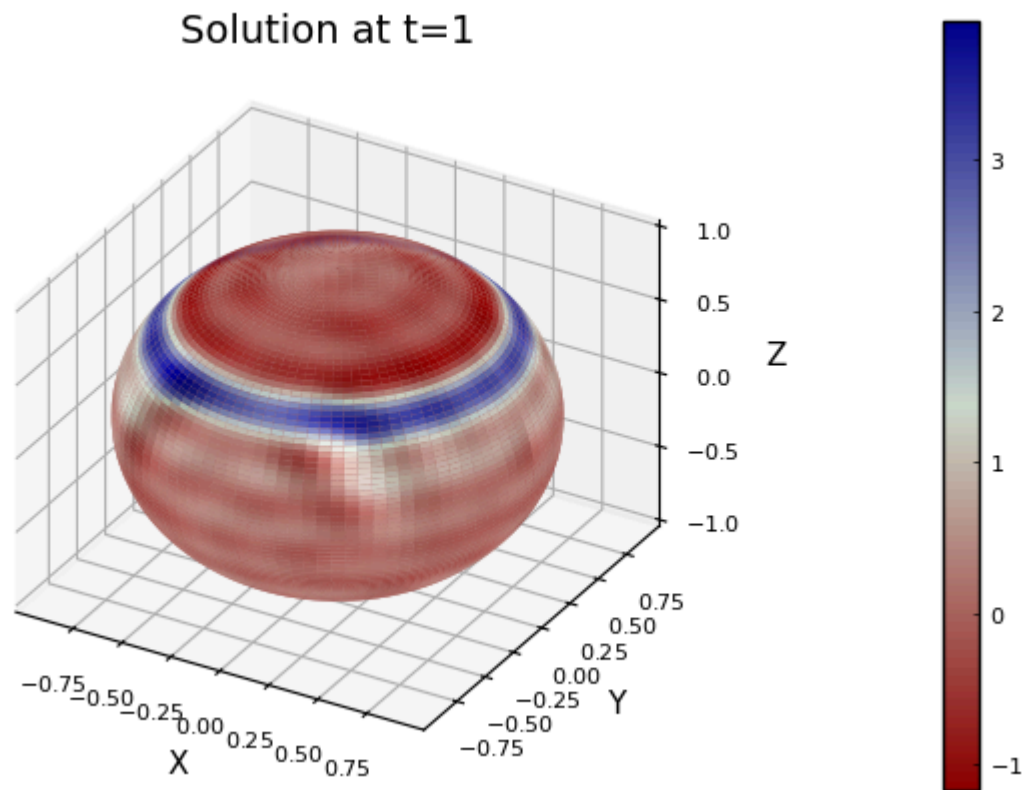
# push!(plots_array, surface(X,Y,Z,fill_z=real.(Solutions[1,:,:]),c=:redsblues,xl=0,yl=0,zl=0))
# push!(plots_array, surface(X,Y,Z,fill_z=real.(Solutions[10,:,:]),c=:redsblues,xl=0,yl=0,zl=0))
# push!(plots_array, surface(X,Y,Z,fill_z=real.(Solutions[50,:,:]),c=:redsblues,xl=0,yl=0,zl=0))
# push!(plots_array, surface(X,Y,Z,fill_z=real.(Solutions[100,:,:]),c=:redsblues,xl=0,yl=0,zl=0))

# plot(plots_array..., layout=(2, 2))
```

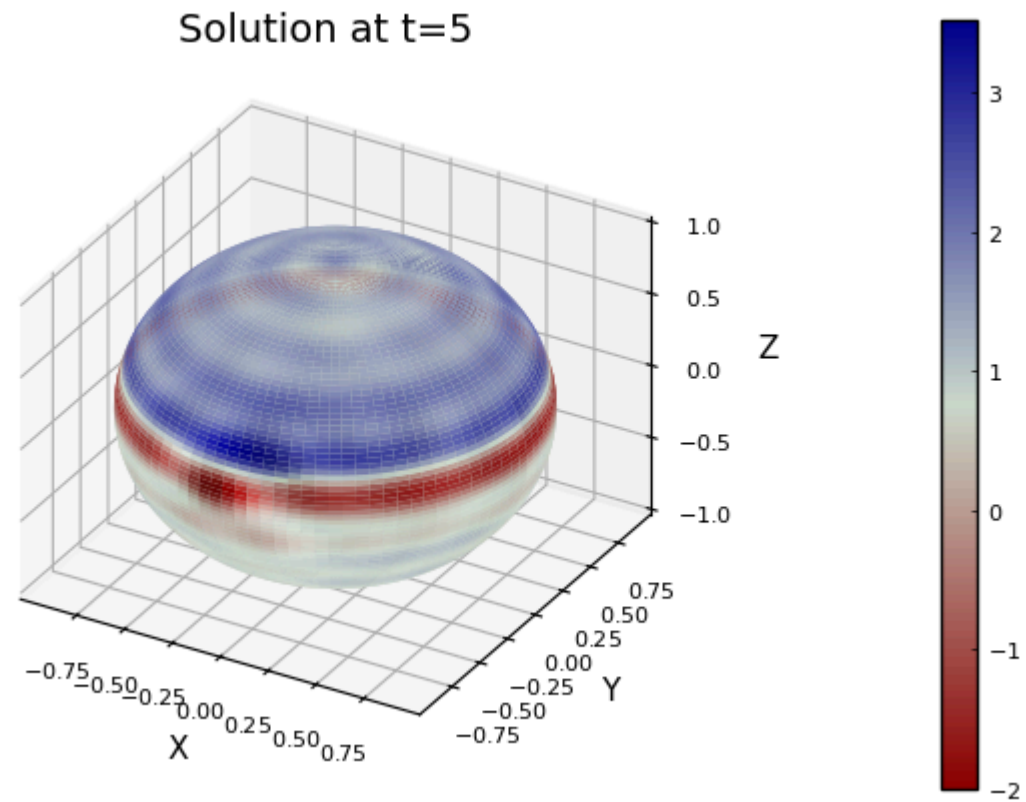
```
In [16]: surface(X,Y,Z,fill_z=real.(Solutions[1,:,:]),c=:redsblues,xlabel="X",ylabel="Y",
```



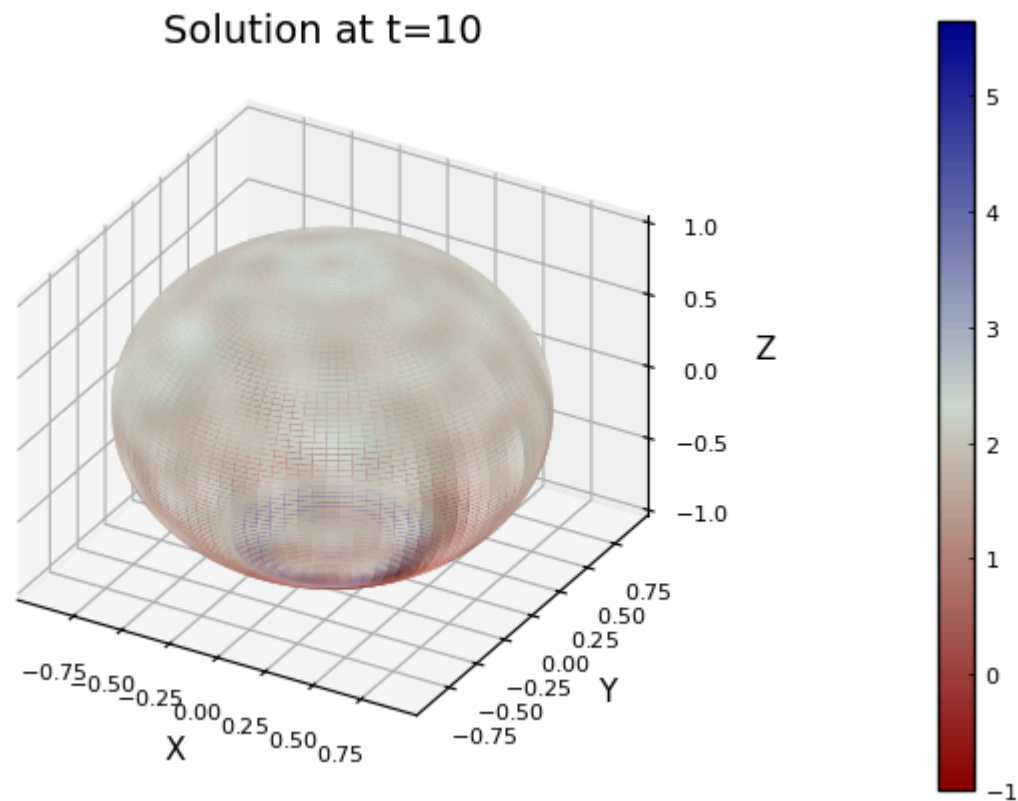
```
In [17]: surface(X,Y,Z,fill_z=real.(Solutions[10,:,:]),c=:redsblues,xlabel="X",ylabel="Y"
```



```
In [18]: surface(X,Y,Z,fill_z=real.(Solutions[50,:,:]),c=:redsblues,xlabel="X",ylabel="Y"
```



```
In [19]: surface(X,Y,Z,fill_z=real.(Solutions[100,:,:]),c=:redsblues,xlabel="X",ylabel="Y"
```



(d) Create a series of figures or a movie that shows how the solution  $\phi$  evolves in time. Perform the simulation three times with different choices of  $l_{max}$ , and at least one of these with a small  $l_{max}$  (e.g.,  $l_{max} = 4$ ) to study the influence of the cut-off  $l_{max}$ .

```
In [39]: # Initializing l_max values
l_max1 = 4;
l_max2 = 10;
l_max3 = 20;

# Initial conditions in terms of coefficients C_lm (with random first derivatives)
C01 = C_lm(Gaussian, l_max1);
Der_C01 = rand(length(C01));

C02 = C_lm(Gaussian, l_max2);
Der_C02 = rand(length(C02));

C03 = C_lm(Gaussian, l_max3);
Der_C03 = rand(length(C03));

# Defining l,m values for each c
lm_1 = lm(l_max1);
lm_2 = lm(l_max2);
lm_3 = lm(l_max3);

# C_lm arrays
C1_t = []
C2_t = []
C3_t = []

Any[]
```

```
In [40]: # Solving the differential equations for C_lm(t)\

# Arrays to store solutions
SolnArray1 = []
SolnArray2 = []
SolnArray3 = []

# L_max=4
# Solve the ODE for each l,m value
for i in 1:length(lm_1)
    l = lm_1[i]
    prob = ODEProblem(ODE_system!, [C01[i], Der_C01[i]], trange, [1])
    Sol1 = solve(prob)
    push!(SolnArray1, Sol1)
end

# L_max=10
# Solve the ODE for each l,m value
for i in 1:length(lm_2)
    l = lm_2[i]
    prob = ODEProblem(ODE_system!, [C02[i], Der_C02[i]], trange, [1])
    Sol2 = solve(prob)
    push!(SolnArray2, Sol2)
end

# L_max=20
# Solve the ODE for each l,m value
for i in 1:length(lm_3)
    l = lm_3[i]
    prob = ODEProblem(ODE_system!, [C03[i], Der_C03[i]], trange, [1])
    Sol3 = solve(prob)
    push!(SolnArray3, Sol3)
end
```

```

In [41]: # Converting coefficients  $C_{lm}(t)$  into function  $\varphi(t)$ 

# L_max = 4
for i in 1:length(lm_1)
    push!(C1_t,[x[1] for x in real.(SolnArray1[i](0:dt:10))])
end

C1_t = hcat(C1_t...);

# L_max = 10
for i in 1:length(lm_2)
    push!(C2_t,[x[1] for x in real.(SolnArray2[i](0:dt:10))])
end

C2_t = hcat(C2_t...);

# L_max = 20
for i in 1:length(lm_3)
    push!(C3_t,[x[1] for x in real.(SolnArray3[i](0:dt:10))])
end

C3_t = hcat(C3_t...);

Solutions1 = zeros(101,100,100)
Solutions2 = zeros(101,100,100)
Solutions3 = zeros(101,100,100)
for k in 1:101
    for l in 1:100
        for m in 1:100
            Solutions1[k,l,m] = C1_t[k,:]' * real.(computeYlm(0 + pi/100 * l, 0 +
            Solutions2[k,l,m] = C2_t[k,:]' * real.(computeYlm(0 + pi/100 * l, 0 +
            Solutions3[k,l,m] = C3_t[k,:]' * real.(computeYlm(0 + pi/100 * l, 0 +
        end
    end
end

```



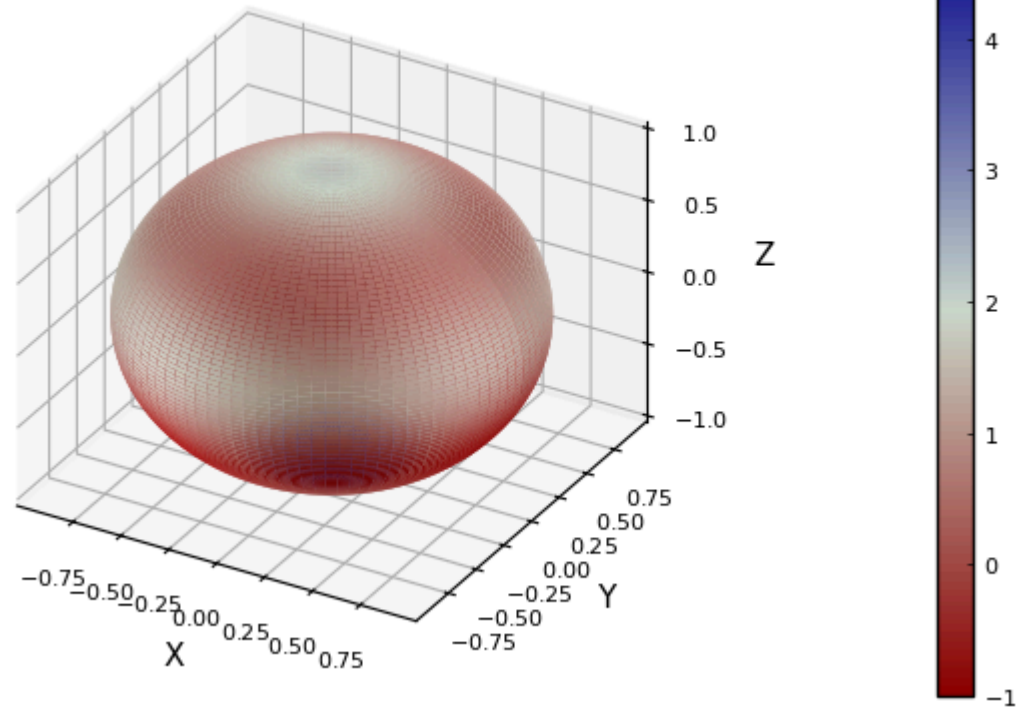
## Animations

$$l_{max} = 4$$

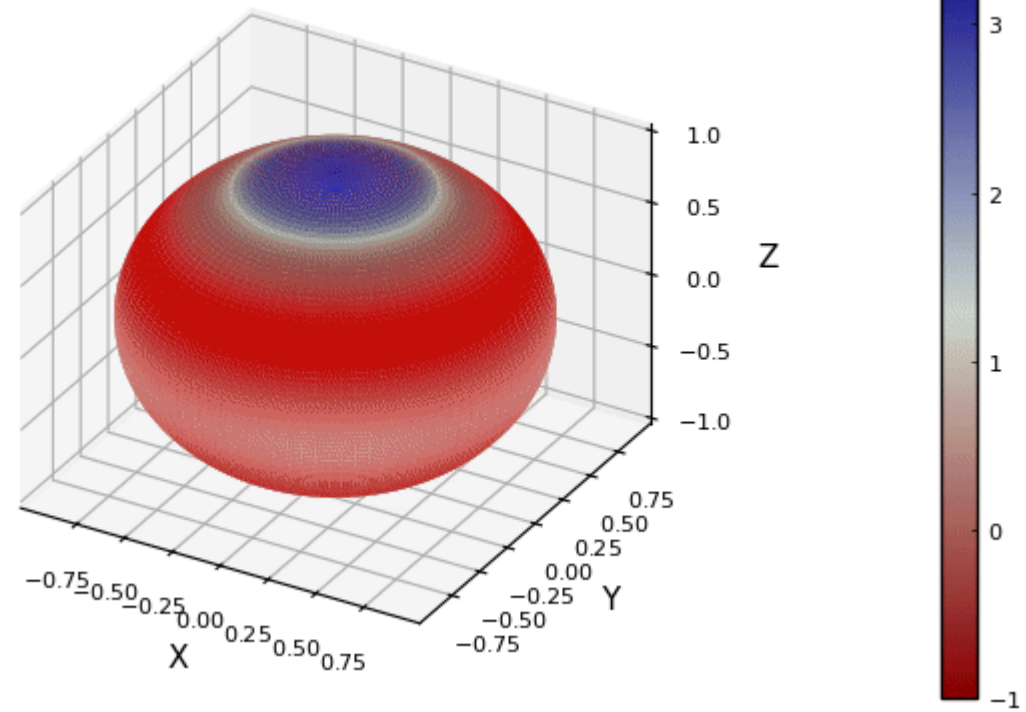
```
In [42]: @gif for i in 1:100
          surface(X, Y, Z, fill_z = Solutions1[i,:,:], c=:redsblues, xlabel = "X", ylab
          end
```

[ Info: Saved animation to C:\Users\mbrnovic\Downloads\tmp.gif

Solution for  $l_{\max} = 4$ ,  $t = 10.0$



Solution for  $l_{\max} = 4$ ,  $t = 0.1$

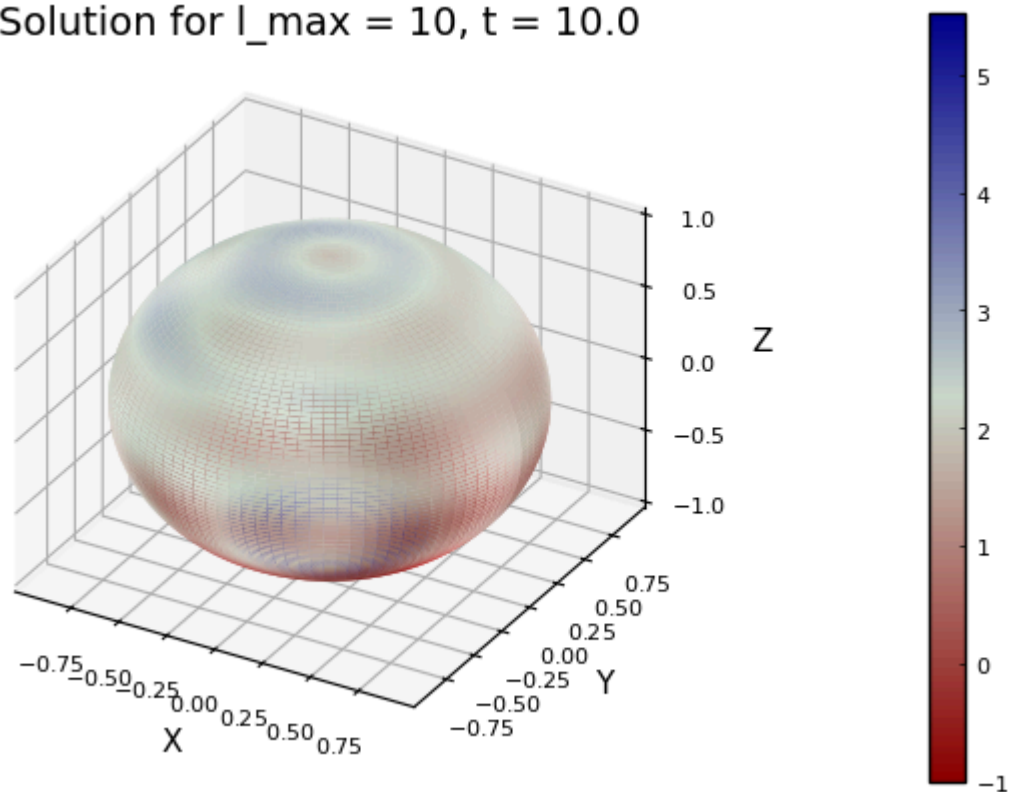


$$l_{max} = 10$$

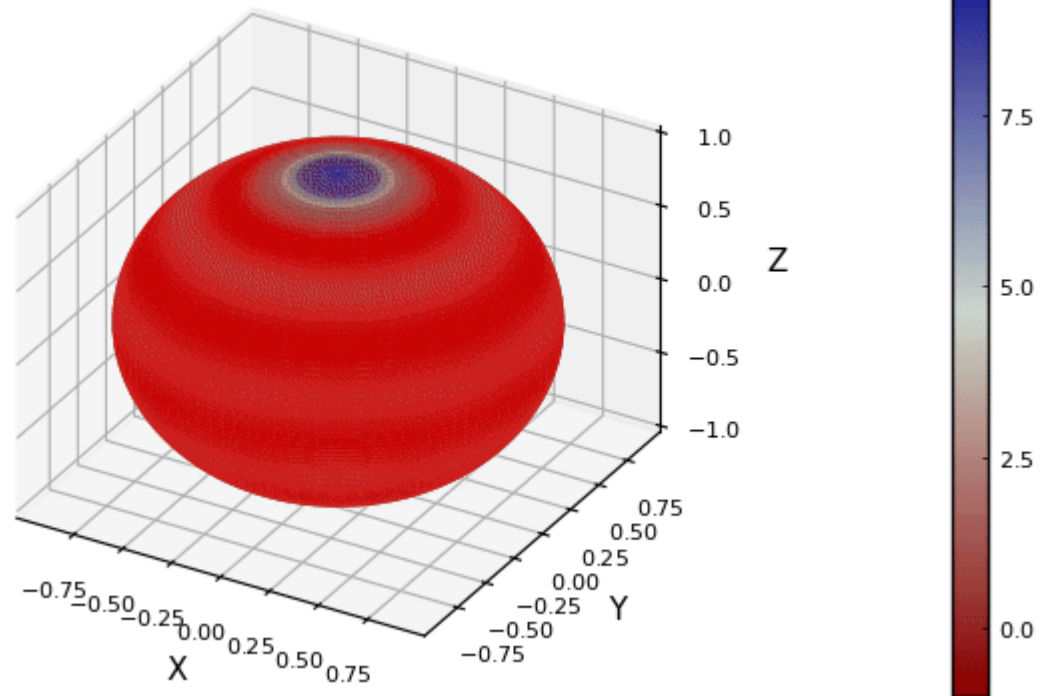
```
In [43]: @gif for i in 1:100
          surface(X, Y, Z, fill_z = Solutions2[i,:,:], c=:redsblues, xlabel = "X", ylabel = "Y", zlabel = "Z")
          end
```

[ Info: Saved animation to C:\Users\mbrnovic\Downloads\tmp.gif

Solution for  $l_{max} = 10, t = 10.0$



Solution for  $I_{\max} = 10$ ,  $t = 0.1$

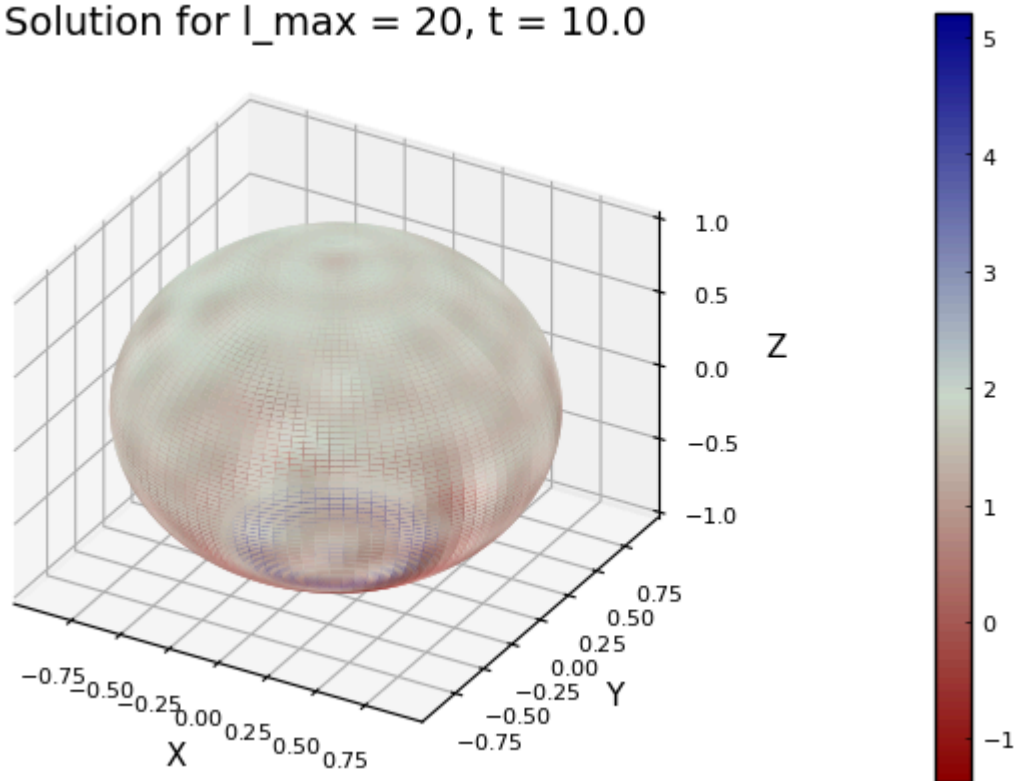


$$l_{max} = 20$$

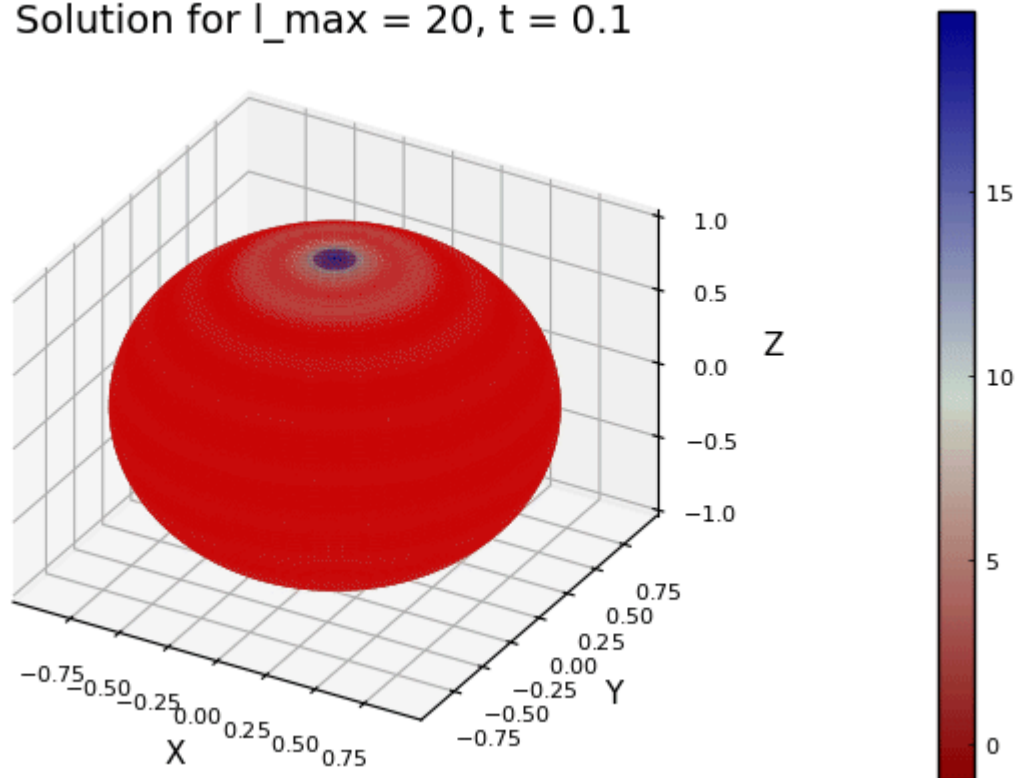
```
In [44]: @gif for i in 1:100
          surface(X, Y, Z, fill_z = Solutions3[i,:,:], c=:redsblues, xlabel = "X", ylab
          end
```

[ Info: Saved animation to C:\Users\mbrnovic\Downloads\tmp.gif

Solution for  $l_{max} = 20$ ,  $t = 10.0$



Solution for  $l_{\max} = 20$ ,  $t = 0.1$



## Acknowledgements

- I would like to thank **Marko** for sharing his approach to plotting the results
- I've used ChatGPT at various points in this homework to get help with the syntax in Julia