#### Name: Hongda Li AMATH 585 2022 WINTER HW7

## Problem 1

The conjugate gradient algorithm is given as:

$$p_0 = b - Ax_0$$
For  $i = 0, 1, \cdots$ 

$$a_i = \frac{\|r_i\|^2}{\|p_i\|_A^2}$$

$$x_{i+1} = x_i + a_i p_i$$

$$r_{i+1} = r_i - a_i A p_i$$

$$b_i = \frac{\|r_{i+1}\|_2^2}{\|r_i\|_2^2}$$

$$p_{i+1} = r_{i+1} + b_i p_i$$

The algorithm has been rephrased. And we make the assumption that the matrix A is symmetric poisitive definite. In addition, observe that  $a_i, b_i$  are non-negative real numbers. This means that we can move then around even if the vector in the inner products can be complex. We wish to prove 3 hypothesis about the algorithm inductively:

$$\mathcal{H}_1(k) \equiv \forall \ 0 \le j \le k - 1 : \langle r_k, p_j \rangle = 0$$

$$\mathcal{H}_2(k) \equiv \forall \ 0 \le j \le k - 1 : \langle p_k, Ap_j \rangle = 0$$

$$\mathcal{H}_3(k) \equiv \forall \ 0 \le j \le k - 1 : \langle r_k, r_j \rangle = 0$$

$$(1.1)$$

First we verify the basecase by considering:  $\mathcal{H}_1(1), \mathcal{H}_2(1), \mathcal{H}_3(1)$ .

$$\langle r_{1}, r_{0} \rangle = \langle r_{0} - a_{0} A p_{0}, r_{0} \rangle$$

$$= \langle r_{0}, p_{0} \rangle - a_{0} \langle r_{0}, A p_{0} \rangle$$

$$= \langle r_{0}, r_{0} \rangle - a_{0} \langle r_{0}, A r_{0} \rangle$$

$$= 0$$

$$\Longrightarrow \mathcal{H}_{3}(1) \text{ is true}$$

$$\langle p_{1}, A p_{0} \rangle = \langle r_{1}, A p_{0} \rangle + \frac{\langle r_{1}, r_{1} \rangle}{\langle r_{0}, r_{0} \rangle} \langle p_{0}, A p_{0} \rangle$$

$$= \langle r_{1}, a_{0}^{-1} (r_{0} - r_{1}) \rangle + a_{0}^{-1} \langle r_{1}, r_{1} \rangle$$

$$\mathcal{H}_{3}(1) \Longrightarrow = -a_{0}^{-1} \langle r_{1}, r_{1} \rangle + a_{0}^{-1} \langle r_{1}, r_{1} \rangle$$

$$\Longrightarrow \mathcal{H}_{2}(1) \text{ is true}$$

$$\langle r_{1}, p_{0} \rangle = \langle r_{1}, r_{0} \rangle$$

$$\mathcal{H}_{3}(1) \Longrightarrow = 0$$

$$\Longrightarrow \mathcal{H}_{1}(1) \text{ is true}$$

Basecase is asserted by the definition of the starting conditions and the coefficient  $a_0$ . next we assume that  $\mathcal{H}_1(k), \mathcal{H}(k), \mathcal{H}(k)$  are all true, and then we wish to prove inductively that they remainds to be true. First, we establish some equalities that are not obvious to simplify the proof, and then we prove it.

$$\langle p_k, Ap_k \rangle = \langle r_k + b_{k-1}p_{k-1}, Ap_k \rangle$$

$$= \langle r_k, Ap_k \rangle \quad \text{by: } \mathcal{H}_2(k)$$

$$\langle r_k, p_k \rangle = \langle r_k, r_k + b_{k-1}p_{k-1} \rangle$$

$$= \langle r_k, r_k \rangle \quad \text{by: } \mathcal{H}_1(k)$$

$$(1.3)$$

The first is implied by  $\mathcal{H}_2(k)$  and the second one is asserted by  $\mathcal{H}_1(k)$ . Next, we prove that  $\mathcal{H}_3(k+1)$  is true.

$$\langle r_{k+1}, r_k \rangle = \langle r_k, r_k \rangle - a_k \langle r_k, Ap_k \rangle$$

$$= \langle r_k, r_k \rangle - a_k \langle p_k, Ap_k \rangle \quad \text{by (1.3)}$$

$$= \langle r_k, r_k \rangle - \langle r_k, r_k \rangle$$

$$= 0$$

$$\forall 0 \le j \le k-1: \quad \langle r_{k+1}, r_j \rangle = \langle r_k - a_k Ap_k, r_j \rangle$$

$$= \langle r_k, r_j \rangle - a_k \langle Ap_k, r_j \rangle$$

$$= -a_k \langle Ap_k, r_j \rangle$$

$$= -a_k \langle Ap_k, p_j - b_{j-1}p_{j-1} \rangle$$

$$= 0 \quad \text{by } \mathcal{H}_2(k)$$

$$\implies \mathcal{H}_3(k+1) \text{ is true.}$$

Next, we consider:

$$\langle r_{k+1}, p_k \rangle = \langle r_k, p_k \rangle - a_k \langle Ap_k, p_k \rangle$$

$$= \langle r_k, r_k \rangle - a_k \langle Ap_k, p_k \rangle$$
 By: (1.3)
$$= 0$$

$$\forall \ 0 \le j \le k-1 : \quad \langle r_{k+1}, p_j \rangle = \langle r_k - a_k Ap_k, p_j \rangle$$

$$= \langle r_k, p_j \rangle - a_k \langle Ap_k, p_j \rangle$$

$$= 0$$
 by:  $\mathcal{H}_1(k) \wedge \mathcal{H}_2(k)$ 

$$\implies \mathcal{H}_1(k+1)$$
 is true

One last hyphothesis to prove. Consider:

$$\langle p_{k+1}, Ap_k \rangle = \langle r_{k+1}, Ap_k \rangle + b_k \langle p_k, Ap_k \rangle$$

$$= \langle r_{k+1}, Ap_k \rangle + \frac{\langle r_{k+1}, r_{k+1} \rangle}{\langle r_k, r_k \rangle} \langle p_k, Ap_k \rangle$$

$$= \langle r_{k+1}, Ap_k \rangle + a_k^{-1} \langle r_{k+1}, r_{k+1} \rangle$$

$$= \langle r_{k+1}, a_k^{-1} (r_k - r_{k+1}) \rangle + a_k^{-1} \langle r_{k+1}, r_{k+1} \rangle$$

$$= \langle r_{k+1}, -a_k^{-1} r_{k+1} \rangle + a_k^{-1} \langle r_{k+1}, r_{k+1} \rangle$$
 by  $\mathcal{H}_3(k+1)$  from: (1.4)
$$= 0$$

$$\forall 0 \leq j \leq k-1: \quad \langle p_{k+1}, Ap_j \rangle = \langle r_{k+1} + b_k p_k, Ap_j \rangle$$

$$= \langle r_{k+1}, a_j^{-1} (r_j - r_{j+1}) \rangle$$

$$= 0$$
 by:  $\mathcal{H}_3(k+1)$ 

$$\implies \mathcal{H}_2(k+1)$$
 is true.

All hypotheses fall through, and the base case is true by the algorithm. The proof has been completed.

## Problem 2

### Problem 2 Code

Make a finite difference method for the following system:  $u''(x) = f(x) \ u(0) = a, \ u(1) = b.$  Dirichilet boundary conditions Parameter:

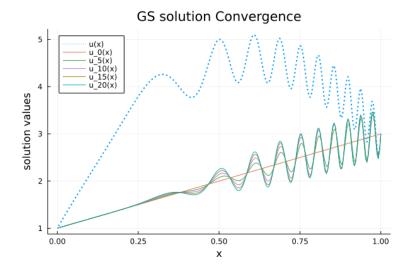
```
n: grid point including the boundary point. The number of interval is n-1, and the interval size is h=1/(n-1).
{\color{red} \textbf{function}} \ \ \textbf{MakeFiniteDiffProblem(}
    n;
f::T1=nothing,
     a::T2=0,
    b::T2=0
) where {T1<:Union{Function, Nothing}, T2<:Number}
     @assert n >= 3 "The number of grid point including"*
          "the boundary conditions has to be at least 3
     if isnothing(f)
         f = (x) -> 0
     end
    end
h = 1/(n - 1)
gridPoints = LinRange(0, 1, n) |> collect
SymM = SymTridiagonal(fill(-2.0, n - 2), fill(1.0, n - 3))
SymM ./= h^2
     RHS = f.(gridPoints[2:end - 1])
RHS[1] -= a/h^2
RHS[end] -= b/h^2
return SymM, RHS end
    Unpreconditioned Conjugate Gradient Algorithm.
function CGS(
    Α,
     b;
     x0=nothing,
     eps=1e-10,
     maxitr=20,
     eps2=Inf
     if isnothing(x0)
         x0 = similar(b)
     end
     r = b - A*x0
    p = \tilde{r}
     v = Vector{typeof(x0)}()
    push!(v, x0)
Ap = similar(b)
     for _ in 1:maxitr
          Ap .= A*p
          a = dot(r, r)/dot(p, Ap)
          if a < 0
              @warn ("Matrix is not symmetric positive definite for CGS. a=$(a)")
          end
          push!(v, v[end] + a*p)
         b = 1/dot(r, r)
         r = r - a*Ap
          if norm(r, Inf) < eps && norm(v[end] - v[end - 1], Inf) < eps2
    println("norm small as: $(norm(r, Inf))")</pre>
          end
          b *= dot(r,r)
         p = r + b*p
return hcat(v...) end
    Performs the Guass Sediel Iterations.
function GS(
    Α,
    b;
    x0=nothing,
     eps=1e-10,
    maxitr=20,
     eps2=Inf
     U = triu(A, 1)
     L = tril(A)
     if isnothing(x0)
```

```
x0 = zeros(size(b))
    end
    v = Vector\{typeof(x0)\}()
    push! (v, x0)
    for _ in 1:maxitr
        push!(v, L(b - U*v[end]))
        res = norm(b - A*v[end], Inf)
if norm(v[end] - v[end - 1], Inf) < eps2 &&
            res < eps
            break
        end
    end
return hcat(v...) end
### Using the code above to answer the questiosn we have ------
using LinearAlgebra, Plots
function f(x)
    phi(x) = 20*pi*x^3
    Dphi(x) = 60*pi*x^2
    DDphi(x) = 129*pi*x
    a = 0.5
return -20 + a*DDphi(x)*cos(phi(x)) - a*(Dphi(x)^2)*sin(phi(x)) end
function f(x::AbstractVecOrMat)
return f.(x) end
function u(x)
    a = 0.5
    phi(x) = 20*pi*x^3
return 1 + 12*x - 10*x^2 + a*sin(phi(x)) end
function u(x::AbstractVecOrMat)
return u.(x) end
function Problem2GS(m=257)
    lbd = 1; rbd = 3
    A, b = MakeFiniteDiffProblem(m, f=f, a=lbd, b=rbd)
    # A |> display
# b |> display
    GridPointsBd = (LinRange(0, 1, m) |> collect)
    IntGridPoints = GridPointsBd[2:end - 1]
    solns = GS(A, b, x0=1 .+ 2*IntGridPoints, maxitr=20)
    # Plotting the solutions along with the correct solutions.
    fig = plot(
        IntGridPoints,
        u(IntGridPoints),
        legend=:topleft,
        label="u(x)
        line=(:dot, 2),
        xlabel="x"
        ylabel="solution values",
title="GS solution Convergence"
    for J in 1:5:size(solns, 2)
        plot!(
            fig, GridPointsBd, vcat(lbd, solns[:, J], rbd),
             label="u_{s}(J-1)(x)"
    end
    fig |> display
savefig(fig, "p2_gs_solns.png")
        legend=:topleft,
        label="u(x)
        line=(:dot, 2),
        xlabel="x",
        ylabel="errors",
title="GS Error Oscillation Patterns"
    @info "Problem 2 GS Errors: "
    for J in 1:5:size(solns, 2)
        err = vcat(lbd, solns[:, J], rbd) - u(GridPointsBd)
        plot!(
```

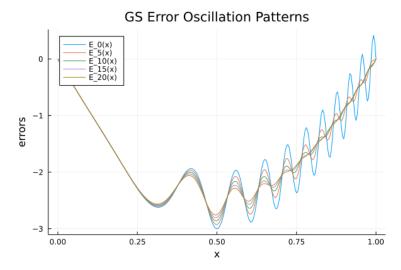
```
fig, GridPointsBd,
            err.
            label="E_$(J-1)(x)"
        err = norm(err, Inf)/sqrt(m - 1)
println("err: $(err)")
    end
    fig |> display
    savefig(fig, "p2_gs_errors.png")
function Problem2CG(m=257)
    lbd = 1; rbd = 3
    A, b = MakeFiniteDiffProblem(m, f=f, a=lbd, b=rbd)
    # A |> display
# b |> display
    GridPointsBd = (LinRange(0, 1, m) | > collect)
    IntGridPoints = GridPointsBd[2:end - 1]
    solns = CGS(A, b, x0=1 .+ 2*IntGridPoints, maxitr=20)
    # Plotting the solutions along with the correct solutions.
    fig = plot(
        IntGridPoints,
        u(IntGridPoints),
        legend=:topleft,
        label="u(x)
        line=(:dot, 2),
        xlabel="x";
        ylabel="solution values",
        title="CG solution Convergence"
    for J in 1:5:size(solns, 2)
        plot!(
            fig, GridPointsBd, vcat(lbd, solns[:, J], rbd),
            label="u_{s}(J-1)(x)
    end
    fig |> display
    savefig(fig, "p2_cg_solns.png")
    fig = plot(
        legend=:topleft,
        label="u(x)
        line=(:dot, 2),
        xlabel="x",
        ylabel="errors",
title="CG Error Oscillation Patterns"
    @info "Problem 2 CG Errors: "
    for J in 1:5:size(solns, 2)
        err = vcat(lbd, solns[:, J], rbd) - u(GridPointsBd)
        plot!(
            fig, GridPointsBd,
            err,
label="E_$(J-1)(x)"
        err = norm(err, Inf)/sqrt(m - 1)
        println("err: $(err)")
    end
    fig |> display
    savefig(fig, "p2_cg_errors.png")
return end
Problem2CG();
Problem2GS();
```

#### **GS** Solutions

Below is a plot of solutions generated by the Gauss Sediel method compare to the correct solution. 20 Iterations of GS is performed 0, 5, 10, 16, 20'th iterations are plotted:



And the plot of the errors for each of the solution is:



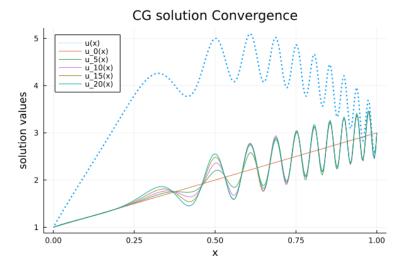
Please observe that the error is getting smoother and smoother as higher frequencies of oscilations are wiped away by the iterative methods. Finally, this is a list for the Inf Norm of the errors:

err: 0.1875

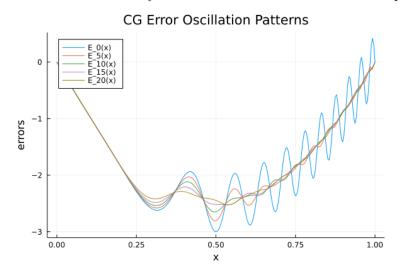
err: 0.1827841933736352 err: 0.1787560123586121 err: 0.1753054397389111 err: 0.17228048081525071

## **CG** Solutions

The plot of the solution found by CG during iterations: 0, 5, 10, 15, 20 is:



The plot of the error of the solution found by CG for solution at the same iteration is plotted:



Observe that the same type of smoothing of the high frequencies oscilations of the solution exists for CG as well. Here is the list of errors for the solution found by cg at these key iterations:

err: 0.1875

err: 0.1753017032952856 err: 0.16554214673565795 err: 0.15765193211388995 err: 0.15774519163539247

# Problem 3

## Problem 4

**4(a)** 

**4(b)**