

# HW #3 Report

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## 1 Problem 1

### 1.1 Part 1

Derive the stability function  $\Phi(z)$  for the following

- Predictor-corrector (Heun's)

$$u_{n+1} = u_n + \frac{h}{2} (f(u_n, t_n) + f(u_n + hf(u_n, t_n), t_n)).$$

For the model problem,  $u'(t) = \gamma u(t)$ ,  $f(u_n, t_n) = \gamma u_n$ , and  $f(u_n + h\gamma u_n) = \gamma(u_n + h\gamma u_n)$ . So we have

$$\begin{aligned} u_{n+1} &= u_n + \frac{h}{2} (\gamma u_n + \gamma(u_n + h\gamma u_n)) \\ &= \left(1 + h\gamma + \frac{(h\gamma)^2}{2}\right) u_n \end{aligned}$$

Define  $z = h\gamma$ , then the stability function  $\Phi(z)$  is

$$\boxed{\Phi(z) = 1 + z + \frac{z^2}{2}}.$$

- 4-th order Runge-Kutta

For any Runge-Kutta method. Where  $\vec{e}$  is a vector of all ones, we can write the general method in vector notation

$$\begin{aligned} \vec{k} &= f(u_n \vec{e} + hA\vec{k}) \\ u_{n+1} &= u_n + h\vec{b}^T \vec{k} \end{aligned}$$

Then for the model problem, we have

$$\begin{aligned}
\vec{k} &= \gamma u_n \vec{e} + \gamma h A \vec{k} \\
(I - zA) \vec{k} &= \gamma (I - zA)^{-1} u_n \vec{e} \\
\implies \vec{k} &= \gamma (I - zA)^{-1} u_n \vec{e}
\end{aligned}$$

and then,

$$\begin{aligned}
u_{n+1} &= u_n + h \vec{b}^T \vec{k} \\
&= u_n + z \vec{b}^T (I - zA)^{-1} u_n \vec{e} \\
&= \left(1 + z \vec{b}^T (I - zA)^{-1} \vec{e}\right) u_n
\end{aligned}$$

So the stability function  $\Phi(z)$  for any RK method is

$$\boxed{\Phi(z) = 1 + z \vec{b}^T (I - zA)^{-1} \vec{e}}.$$

## 1.2 Part 2

Study the zero-stability for each of the two LMMs below

•

$$u_{n+2} - 2u_{n+1} + u_n = hf(u_{n+1}, t_{n+1}) - f(u_n, t_n).$$

This has the following characteristic polynomial ( $= 0$ )

$$\begin{aligned}
\rho(\omega) &= \omega^2 - 2\omega + 1 = 0 \\
&= (\omega - 1)^2 = 0
\end{aligned}$$

With root  $\omega = 1$  of multiplicity 2. Since this is not a simple root, this method is **not** zero stable.

•

$$u_{n+2} - u_n = h \left( \frac{1}{3} f(u_{n+2}, t_{n+2}) + \frac{4}{3} f(u_{n+1}, t_{n+1}) + \frac{1}{3} f(u_n, t_n) \right).$$

This method has characteristic polynomial

$$\rho(\omega) = \omega^2 - 1.$$

with roots

$$\omega = \pm 1.$$

Since  $|w_i| \leq 1$ , this method satisfies the root condition, and **is** therefore zero stable.

## 2 Problem 2

### 2.1 Part 1

Show that 2s-DIRK is second order for  $\alpha = 1 - \frac{1}{\sqrt{2}}$

First order consistency condition

$$\begin{aligned}\sum_{i=1}^{p=2} b_i &= 1 \\ \implies (1 - \alpha) + \alpha &= 1\end{aligned}$$

Which is satisfied regardless of the value of  $\alpha$ .

Second order consistency condition

$$\begin{aligned}\sum_{i=1}^{p=2} b_i c_i &= \frac{1}{2} \\ &= \underbrace{(1 - \alpha)}_{b_1} \underbrace{\alpha}_{c_1} + \underbrace{\alpha}_{b_2} \underbrace{(1)}_{c_2} \\ &= \frac{1}{\sqrt{2}} \left( 1 - \frac{1}{\sqrt{2}} \right) + 1 - \frac{1}{\sqrt{2}} \\ &= \frac{1}{2}\end{aligned}$$

So for  $\alpha = 1 - \frac{1}{\sqrt{2}}$ , 2s-DIRK is at least second order.

### 2.2 Part 2

For the model problem,  $u' = \gamma u$ , derive the expressions for  $k_1$ ,  $k_2$ , and the stability function  $\Phi(z)$

$$\begin{aligned}k_1 &= f(u_n + \alpha h k_1, t_n + \alpha h) \\ &= \gamma u_n + \alpha \gamma h k_1 \\ \implies \boxed{k_1 &= \frac{\gamma}{1 - \alpha z} u_n}\end{aligned}$$

$$\begin{aligned}k_2 &= f(u_n + h((1 - \alpha)k_1 + \alpha k_2), t_n + h) \\ &= \gamma u_n + z \left( (1 - \alpha) \underbrace{\frac{\gamma}{1 - \alpha z} u_n}_{k_1} + \alpha k_2 \right) \\ \implies \boxed{k_2 &= \frac{(1 - \alpha z)\gamma + \gamma z(1 - \alpha)}{(1 - \alpha z)^2} u_n}\end{aligned}$$

Then when  $h$  is distributed in  $u_{n+1} = u_n + \underbrace{h}_{((1-\alpha)k_1 + \alpha k_2)}$ , we get  $k_1$ , and  $k_2$  as desired.

The stability function  $\Phi$  follows from these results. Plugging in  $k_1$  and  $k_2$

$$u_{n+1} = \left( 1 + (1-\alpha) \frac{z}{1-\alpha z} + \alpha \left( \frac{(1-\alpha z)z + z^2(1-\alpha)}{(1-\alpha z)^2} \right) \right) u_n.$$

So

$$\Phi(z) = 1 + (1-\alpha) \frac{z}{1-\alpha z} + \alpha \left( \frac{(1-\alpha z)z + z^2(1-\alpha)}{(1-\alpha z)^2} \right).$$

Finding the common denominator and simplifying, we get the result as desired

$$\boxed{\Phi(z) = \frac{1 + (1-2\alpha)z}{(1-\alpha z)^2}}.$$

### 2.3 Part 3

**Suppose 2s-DIRK is A-stable for  $\alpha = 1 - \frac{1}{\sqrt{2}}$ . Show that it satisfies the second condition of L-stability**

We want to show

$$\lim_{z \rightarrow \infty} \Phi(z) = 0.$$

Make the change of variable  $w = \frac{1}{z}$ , then we can equivalently take the limit as  $w \rightarrow 0$

$$\begin{aligned} \lim_{w \rightarrow 0} \Phi\left(\frac{1}{w}\right) &= \lim_{w \rightarrow 0} \frac{1 + (1-2\alpha)\frac{1}{w}}{(1 - \alpha\frac{1}{w})^2} \\ &= \lim_{w \rightarrow 0} \frac{w + (1-2\alpha)}{w\frac{1}{w^2}(w - \alpha)^2} \\ &= \lim_{w \rightarrow 0} \frac{w^2 + (1-2\alpha)w}{(w - \alpha)^2} \\ &= 0 \end{aligned}$$

So this method is L-stable.

## 3 Problem 3

Consider the implicit 2-step method

$$u_{n+2} - u_n = h \left( \frac{1}{3}f(u_{n+2}, t_{n+2}) + \frac{4}{3}f(u_{n+1}, t_{n+1}) + \frac{1}{3}f(u_n, t_n) \right)$$

### 3.1 Part 1

**Show**  $e_n(h) = O(h^5)$

For the model problem

$$\begin{cases} u'(t) = u(t) \\ u(0) = 1 \end{cases} \implies u(t) = e^t.$$

the local truncation error

$$\begin{aligned} e_n &= \sum_{j=0}^r \alpha_j e^{(t_n+jh)} - h \sum_{j=1}^r \beta_j e^{(t_n+jh)} \\ &= e^{t_n} \left( \sum \alpha_j e^{jh} - \log(e^{jh}) \sum \beta_j e^{jh} \right) \end{aligned}$$

define  $z = e^h$ , then the order of the local truncation error can be written in terms of  $z$

$$e_n = O(h^{p+1}) = O(\log(z)^{p+1}) \underbrace{=}_{\text{by Taylor expansion}} O(|z-1|^{p+1}).$$

and more compactly, in terms of characteristic polynomials

$$e_n = \sum \alpha_j z^j - \log(z) \sum \beta_j z^j \tag{1}$$

$$= \rho(z) - \log(z)\sigma(z) \tag{2}$$

So now for this problem, we need to check the order of equation (2). The characteristic polynomials are

$$\begin{aligned} \rho(z) &= z^2 - 1 \\ \sigma(z) &= \frac{1}{3}z^2 + \frac{4}{3}z + \frac{1}{3} \end{aligned}$$

let  $\zeta = z - 1$ , then

$$\begin{aligned} e_n &= \rho(\zeta + 1) - \log(\zeta + 1)\sigma(\zeta + 1) \\ &= (\zeta + 1)^2 - 1 - \log(\zeta + 1) \left( \frac{1}{3}(\zeta + 1)^2 + \frac{4}{3}(\zeta + 1) + \frac{1}{3} \right) \end{aligned}$$

Taylor expanding  $\log(\zeta + 1)$

$$\begin{aligned} e_n &= \zeta^2 + 2\zeta - \left( \zeta - \frac{\zeta^2}{2} + \frac{\zeta^3}{3} - \frac{\zeta^4}{4} + \frac{\zeta^5}{5} + O(\zeta^6) \left( \frac{1}{3}\zeta^2 + 2\zeta + 2 \right) \right) \\ &= \zeta^2 + 2\zeta - \left( 2\zeta + \zeta^2(-1+2) + \zeta^3(\cancel{\frac{2}{3}-1+\frac{1}{3}}) + \zeta^4(\cancel{-\frac{1}{2}+\frac{2}{3}-\frac{1}{6}}) + \zeta^5(\frac{2}{5}-\frac{1}{2}+\frac{1}{9}) + O(\zeta^6) \right) \\ &= \frac{1}{90}\zeta^5 + O(\zeta^6) \\ &= \frac{1}{90}(z-1)^5 + O((z-1)^6) \\ &= O(h^5) \end{aligned}$$

This shows that this implicit 2-step method is of order 5.

## 3.2 Part 2

Find roots of the following for  $z = -\epsilon$ ,  $\epsilon > 0$

$$\pi(\xi, z) = (\xi^2 - 1) - z \left( \frac{1}{3}\xi^2 + \frac{4}{3}\xi + \frac{1}{3} \right).$$

Plugging in  $z = -\epsilon$  and simplifying

$$\begin{aligned} \xi^2 - 1 + \epsilon \left( \frac{1}{3}\xi^2 + \frac{4}{3}\xi + \frac{1}{3} \right) &= 0 \\ \xi^2 \left( 1 + \frac{\epsilon}{3} \right) + \xi \left( \frac{4}{3}\epsilon \right) + \frac{\epsilon}{3} - 1 &= 0 \end{aligned}$$

Using the quadratic formula and simplifying

$$\begin{aligned} \xi &= \frac{-\frac{4}{3}\epsilon \pm \sqrt{(\frac{4}{3}\epsilon)^2 - 4(\frac{\epsilon}{3} + 1)(\frac{\epsilon}{3} - 1)}}{2(1 + \frac{\epsilon}{3})} \\ &= -\frac{2}{3}\epsilon \left( \frac{1}{1 + \frac{\epsilon}{3}} \right) \pm \frac{\sqrt{\frac{12}{9}\epsilon^2 + 4}}{2(1 + \frac{\epsilon}{3})} \\ &= -\frac{2}{3}\epsilon \left( \frac{1}{1 + \frac{\epsilon}{3}} \right) \pm \frac{\sqrt{1 + \frac{\epsilon}{3}}}{1 + \frac{\epsilon}{3}} \\ &= -\frac{2}{3}\epsilon \left( \frac{1}{1 + \frac{\epsilon}{3}} \right) \pm \frac{1}{\sqrt{1 + \frac{\epsilon}{3}}} \end{aligned}$$

Now taylor expanding,

$$\xi = -\frac{2}{3}\epsilon \left( 1 - \frac{\epsilon}{3} + O(\epsilon^2) \right) \pm 1 - \frac{\epsilon}{6} + O(\epsilon^2)$$

Gives us the roots

$$\xi_1(\epsilon) = 1 - \frac{5}{6}\epsilon + O(\epsilon^2), \quad \xi_2 = -\left( 1 + \frac{1}{3} \right) + O(\epsilon^2).$$

Since  $|\xi_i| \not\leq 1$ , the root condition is not satisfied, and so  $z = -\epsilon$  is **not** in the region of absolute stability.

## 4 Including required packages

```
using Plots
using LaTeXStrings
theme(:mute)

using Pkg
```

```
Pkg.activate("RAS")
include("code/RAS.jl") # Makes sure the module is run before using it
using .RAS: RAS_stabf, RASrk

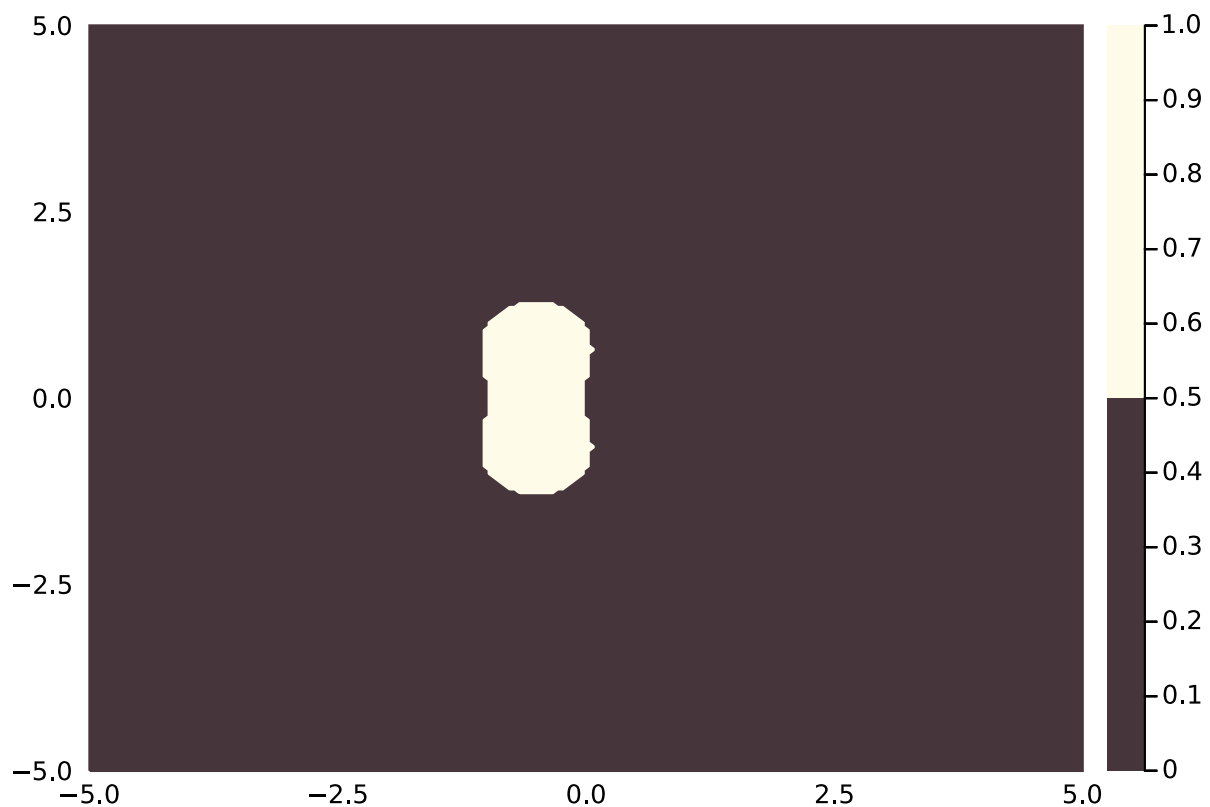
Pkg.activate("DiffyQ")
include("code/DiffyQ.jl") # Makes sure the module is run before using it
using .DiffyQ: s2_DIRK, BackwardEuler_n
```

## 5 Problem 4: Plot Region of Absolute Stability

Note that "light" color is the region of stability

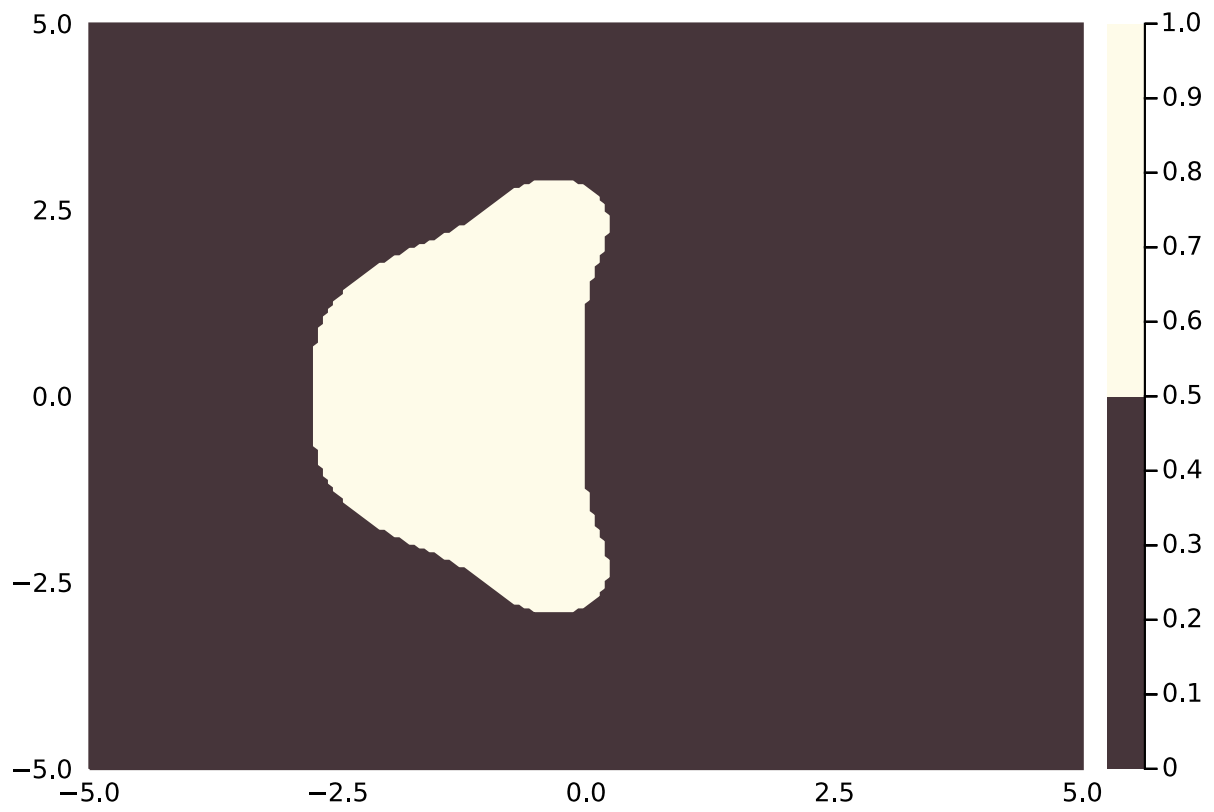
```
# stability function for Heun's
 $\Phi(z) = 1.0 + z + z^2$ 
xs, Z = RAS_stabf( $\Phi$ )

contourf(xs, xs, Z, levels = 1)
```



```
# RK4
A = [0 0 0 0
     1/2 0 0 0
     0 1/2 0 0
     0 0 1 0]
b = [1/6, 1/3, 1/3, 1/6]

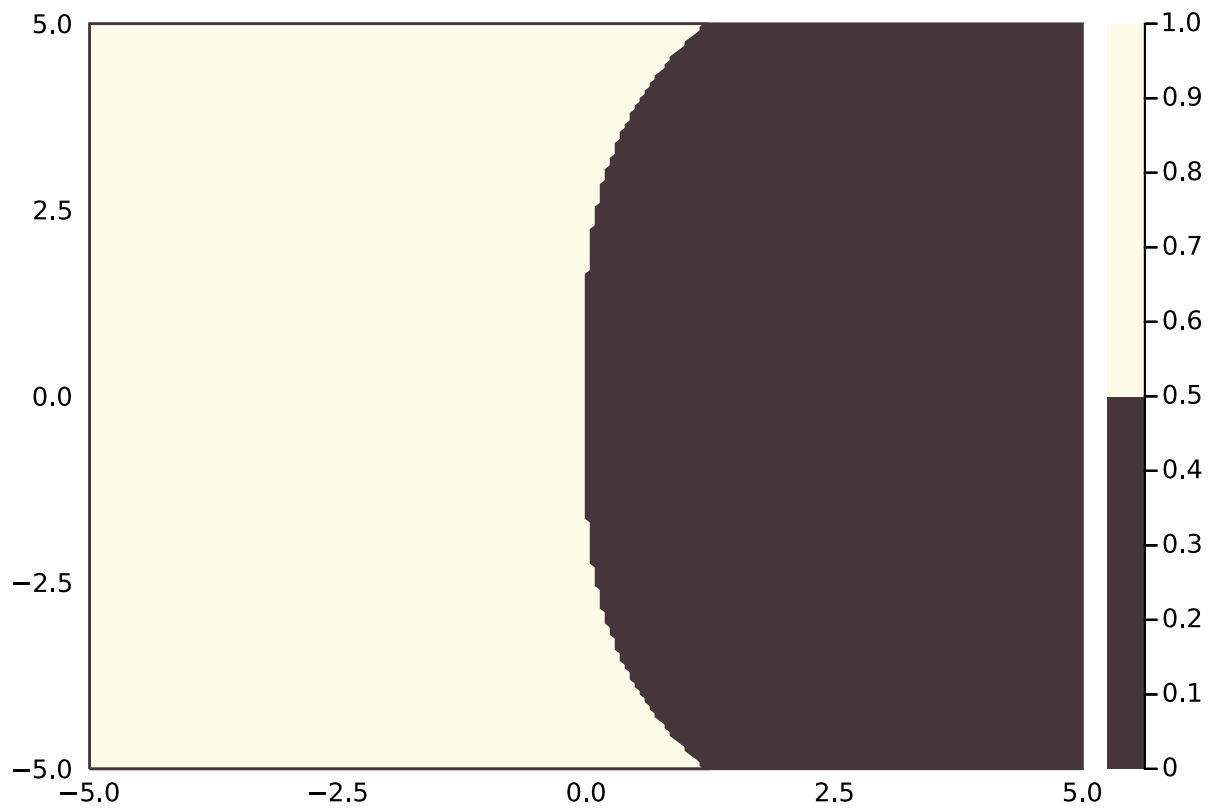
xs, Z = RASrk(A,b)
# plotly()
contourf(xs,xs,Z, levels = 1)
```



```
# 2s-DIRK
α = 1-1/sqrt(2)
A = [α 0
     1-α α]
b = [1-α, α]

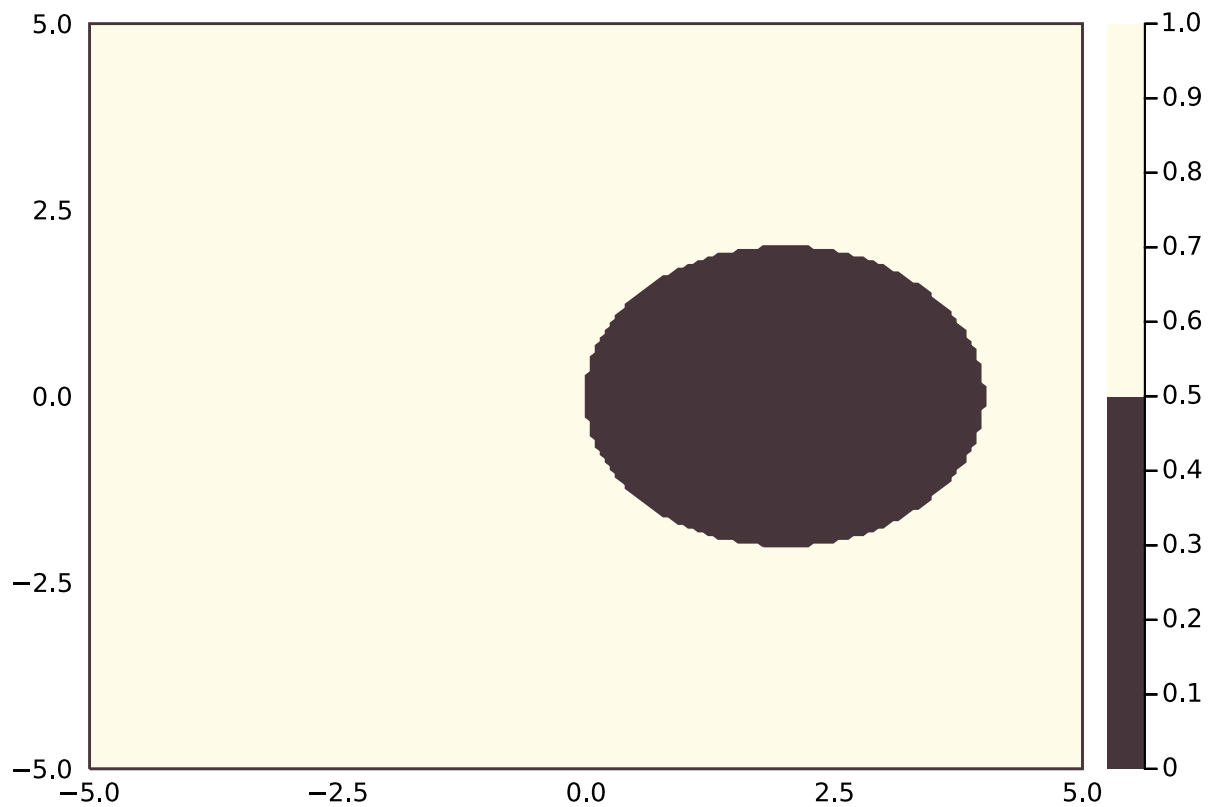
xs, Z = RASrk(A,b)
contourf(xs,xs,Z, levels = 1)
```





```
# 2s-Dirk
 $\alpha$  = 0.5
A = [ $\alpha$  0
      1- $\alpha$   $\alpha$ ]
b = [1- $\alpha$ ,  $\alpha$ ]

xs, Z = RASrk(A,b)
contourf(xs,xs,Z, levels = 1)
```



## 6 Problem 5

```
f(u,t,μ) = -(0.5*exp(20*cos(1.3*t)) * sinh(u-cos(t)));
```

```
α = 1 - 1/sqrt(2);
```

```
T = 30.0; h = 2.0^(-5); N = Int(T/h);
```

```
u0 = 0.0;
```

```
u = s2_DIRK(f, N, T, u0, α);
```

### 6.1 Part 1

```
tList = collect(0:N)*(T/N)
```

```
plot(tList, u, label = L"u(t)", thickness_scaling = 1.25)
```

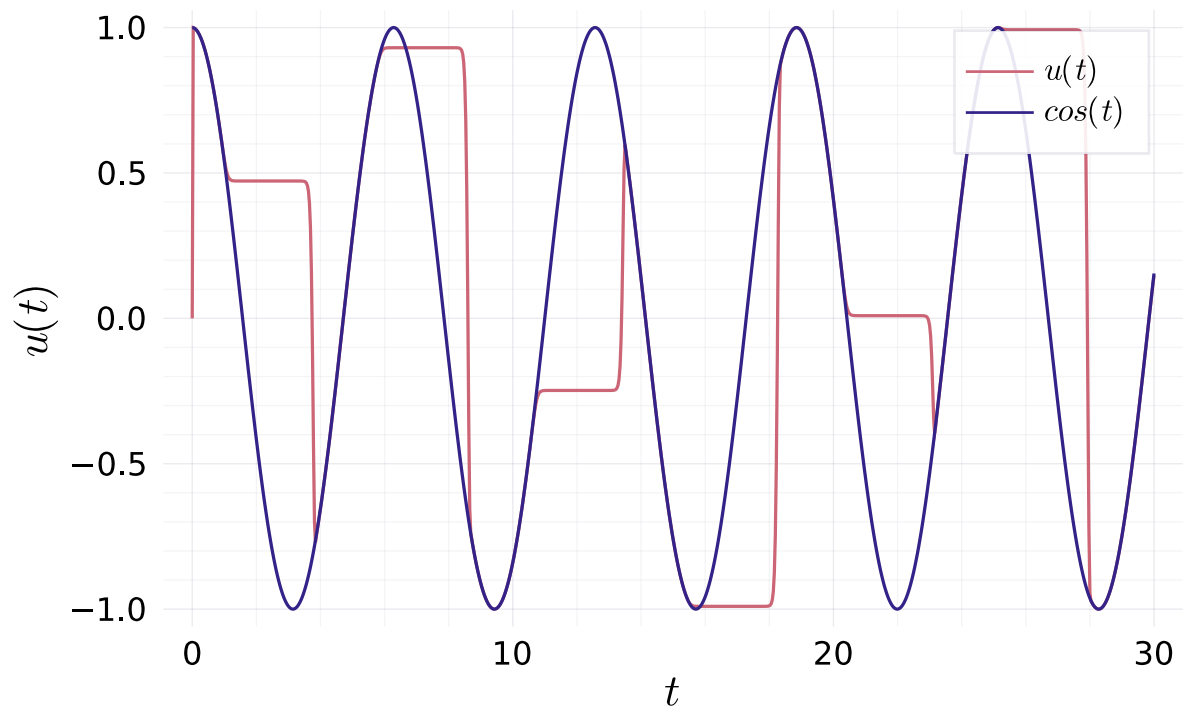
```
xlabel!(L"t")
```

```
ylabel!(L"u(t)")
```

```
title!(latexstring("2s-Dirk, h=2^{-5}, T=", T))
```

```
plot!(tList, cos.(tList), label = L"cos(t)")
```

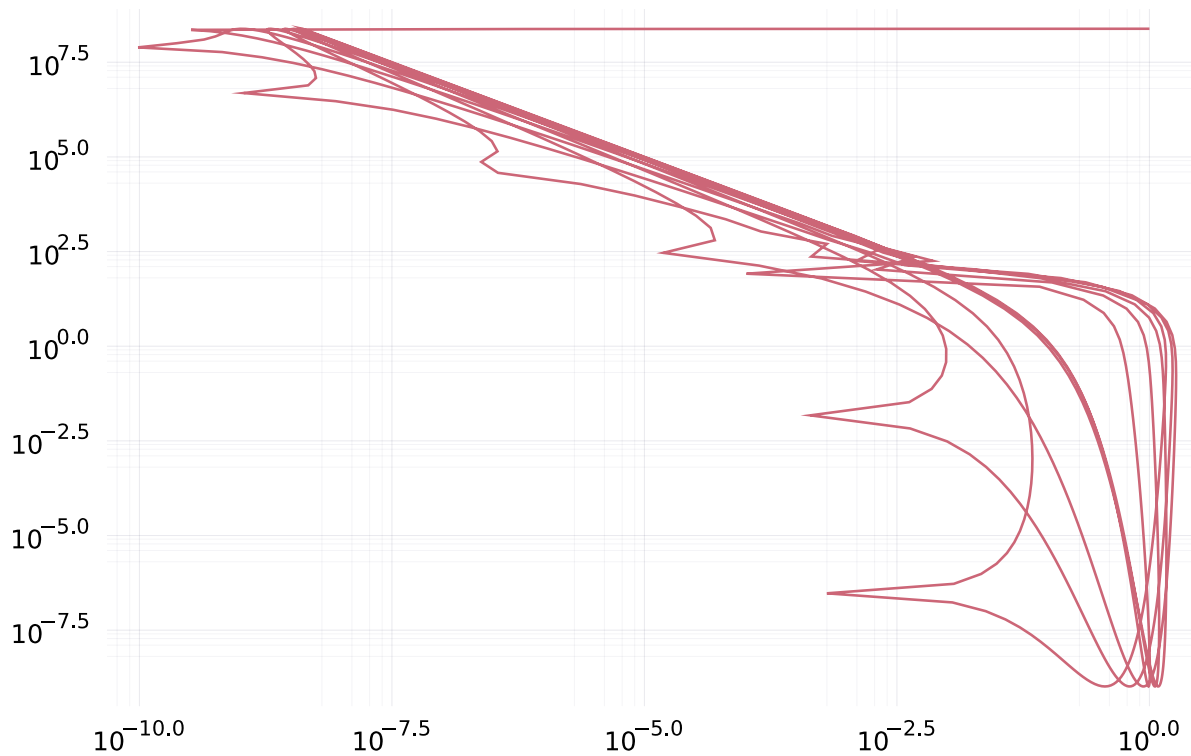
$2s - DIRK, h = 2^{-5}, T = 30.0$



## 6.2 Part 2

```
g(t) = 0.5*exp(20*cos(1.3*t))
plot(abs.(u - cos.(tList)), g.(tList), xaxis=:log, yaxis=:log, legend = false)
title!("loglog plot")
```

## loglog plot



## 7 Problem 6

```

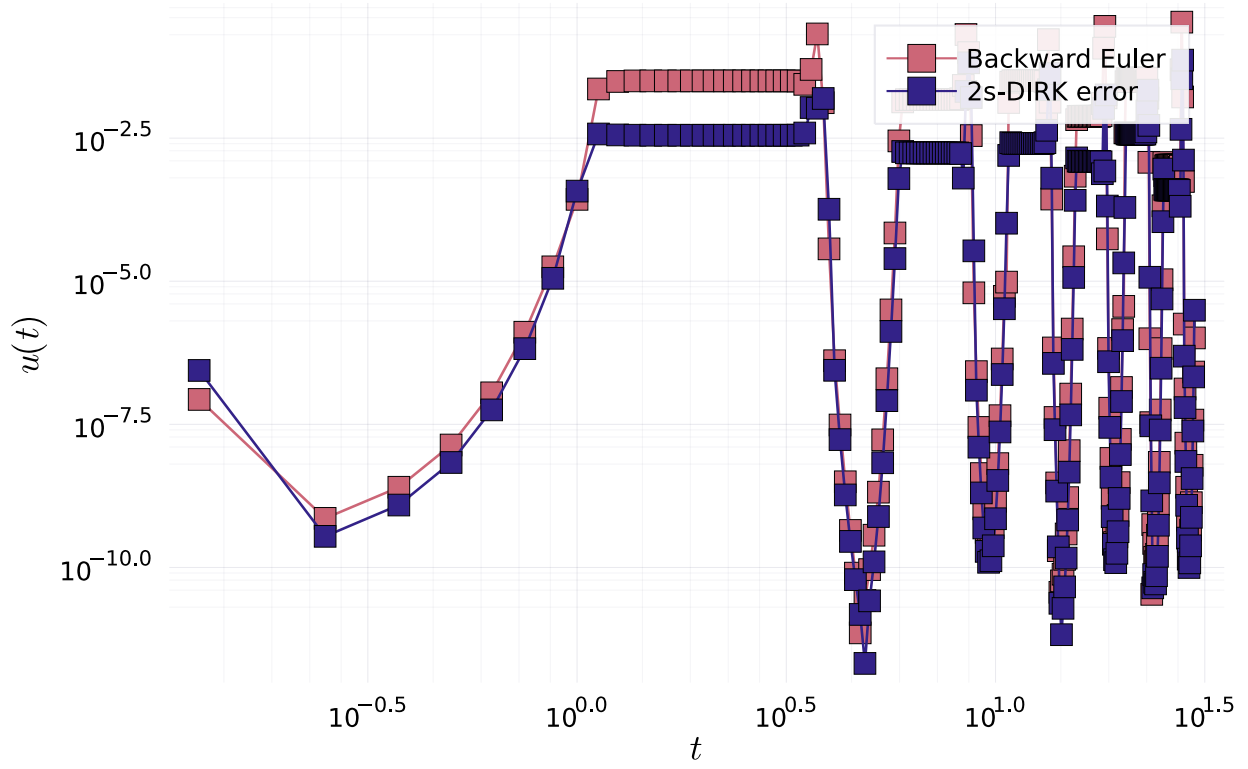
hs = 1 ./ (2 .^(3:8))
for i = 1 : length(hs)
    N = Int(T/hs[i])
    tList = collect(0:N)*(T/N)

    ## Backward Euler
    u_euler = BackwardEuler_n(f, N, T, u0)
    u_exact = BackwardEuler_n(f, 2*N, T, u0)
    euler_error = abs.(u_euler[1:N] - u_exact[1:2:2*N])./(1-0.5^1) # first order method
    p1 = plot(tList[2:N], euler_error[2:N], label = "Backward Euler", xaxis=:log,
    yaxis=:log, marker = (:square,5))
    xaxis!(L"t")
    yaxis!(L"u(t)")
    title!(latexstring("Error Estimate,h=",hs[i]))

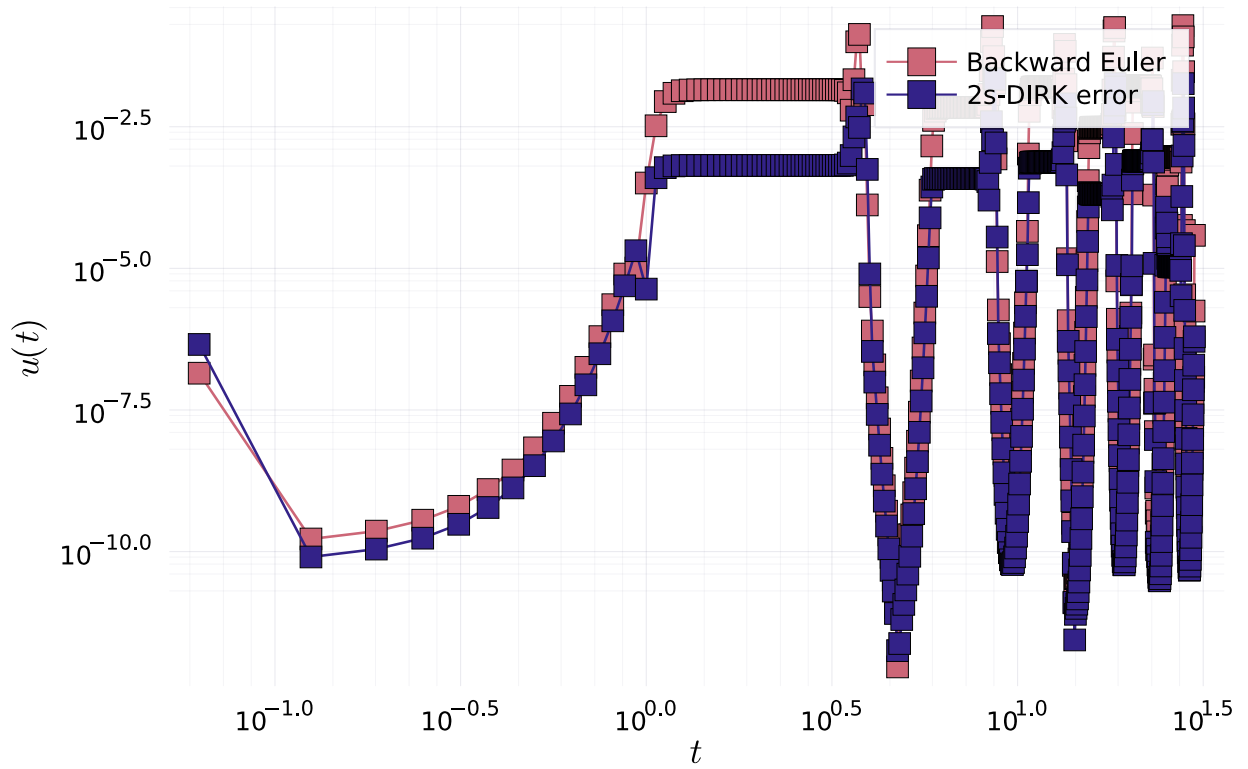
    ## s2-DIRK
    u_s2_DIRK = s2_DIRK(f, N, T, u0, α)
    u_Dexact = s2_DIRK(f, 2*N, T, u0, α)
    DIRK_error = abs.(u_s2_DIRK[1:N] - u_Dexact[1:2:2*N])./(1-0.5^2) # second order
    method
    p2 = plot!(tList[2:N], DIRK_error[2:N], label = "2s-DIRK error", xaxis=:log,
    yaxis=:log, marker = (:square,5))
    display(p2)
end

```

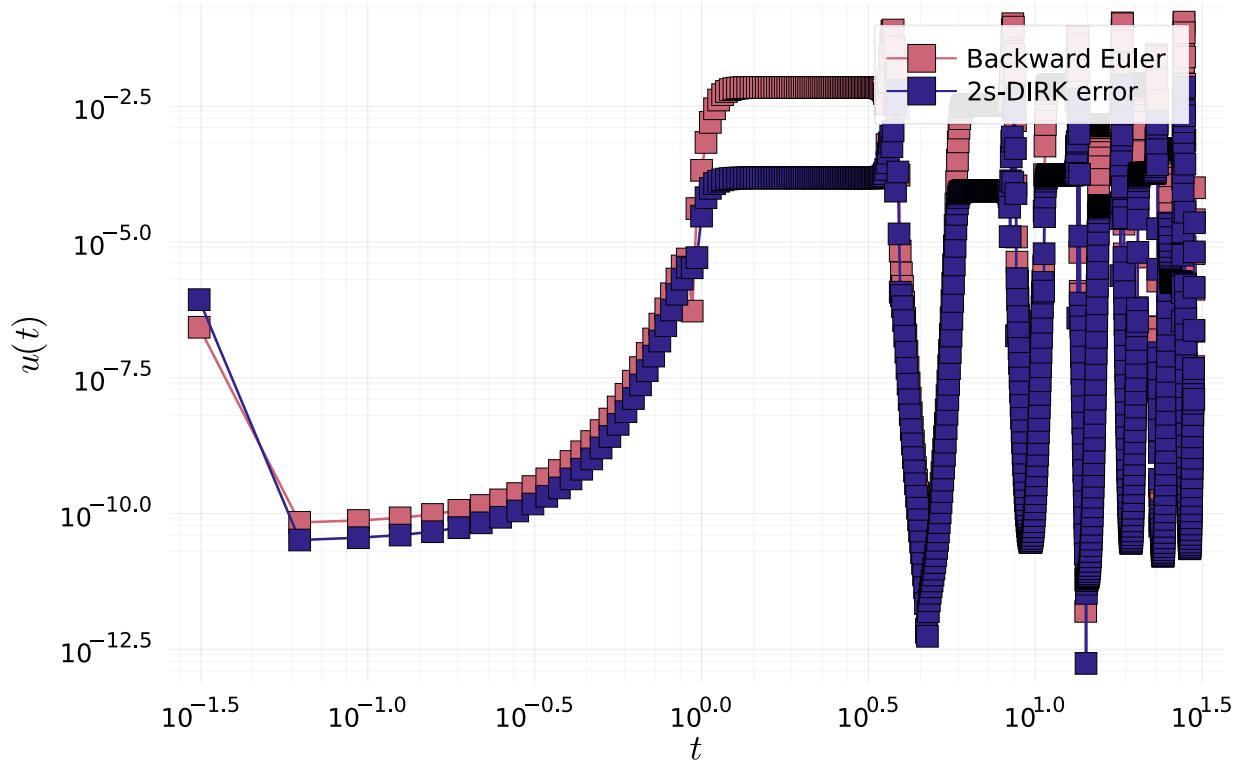
*ErrorEstimate,  $h = 0.125$*



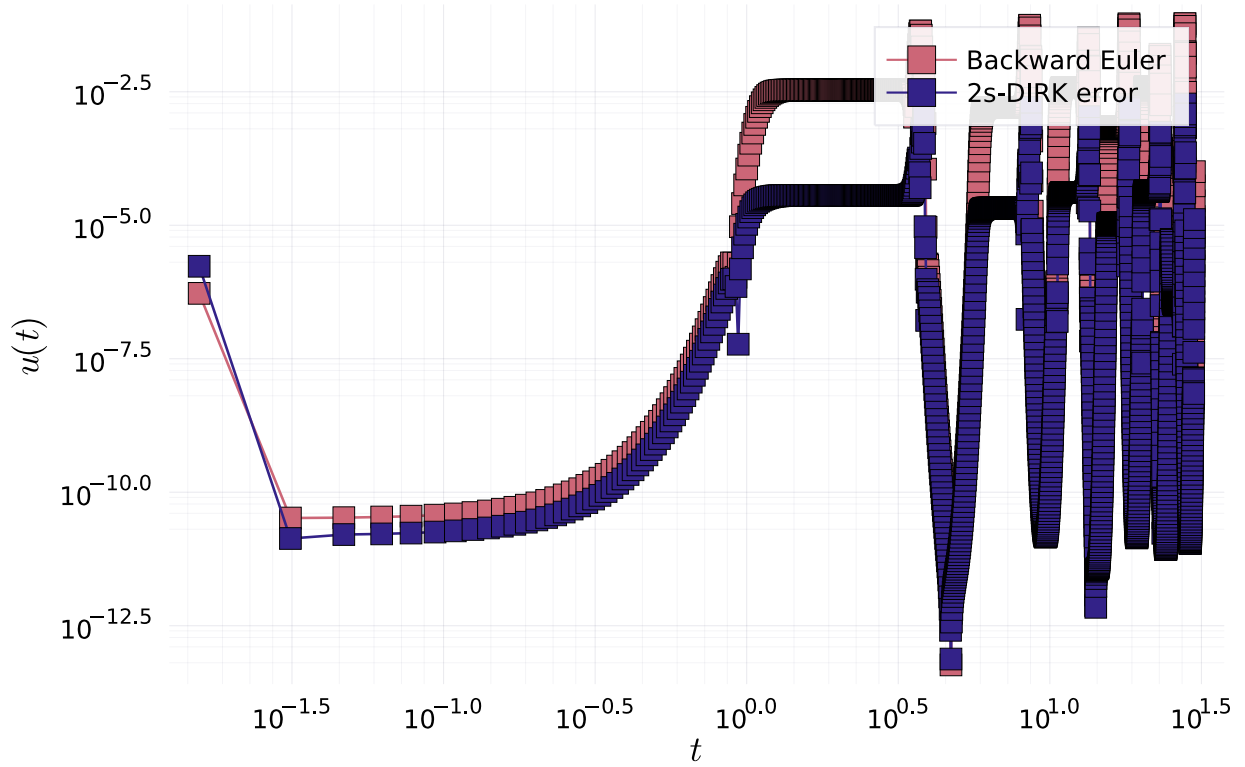
*ErrorEstimate,  $h = 0.0625$*



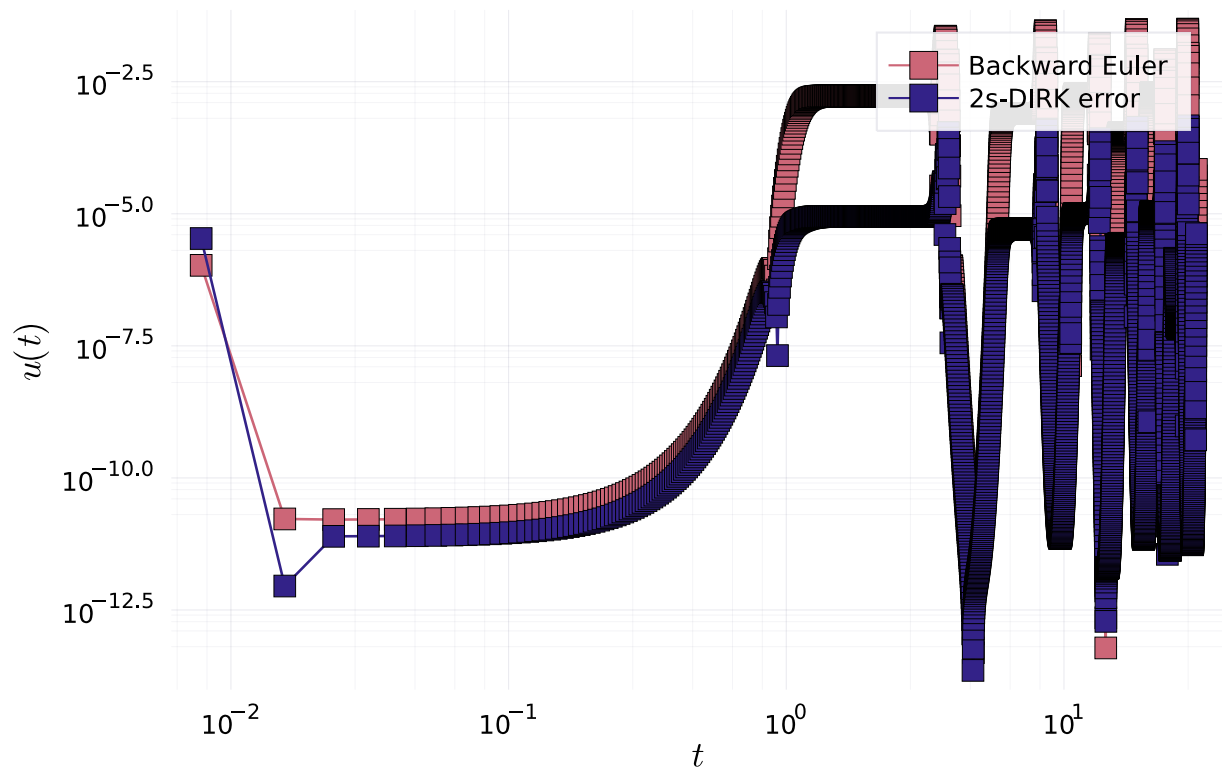
*Error Estimate,  $h = 0.03125$*



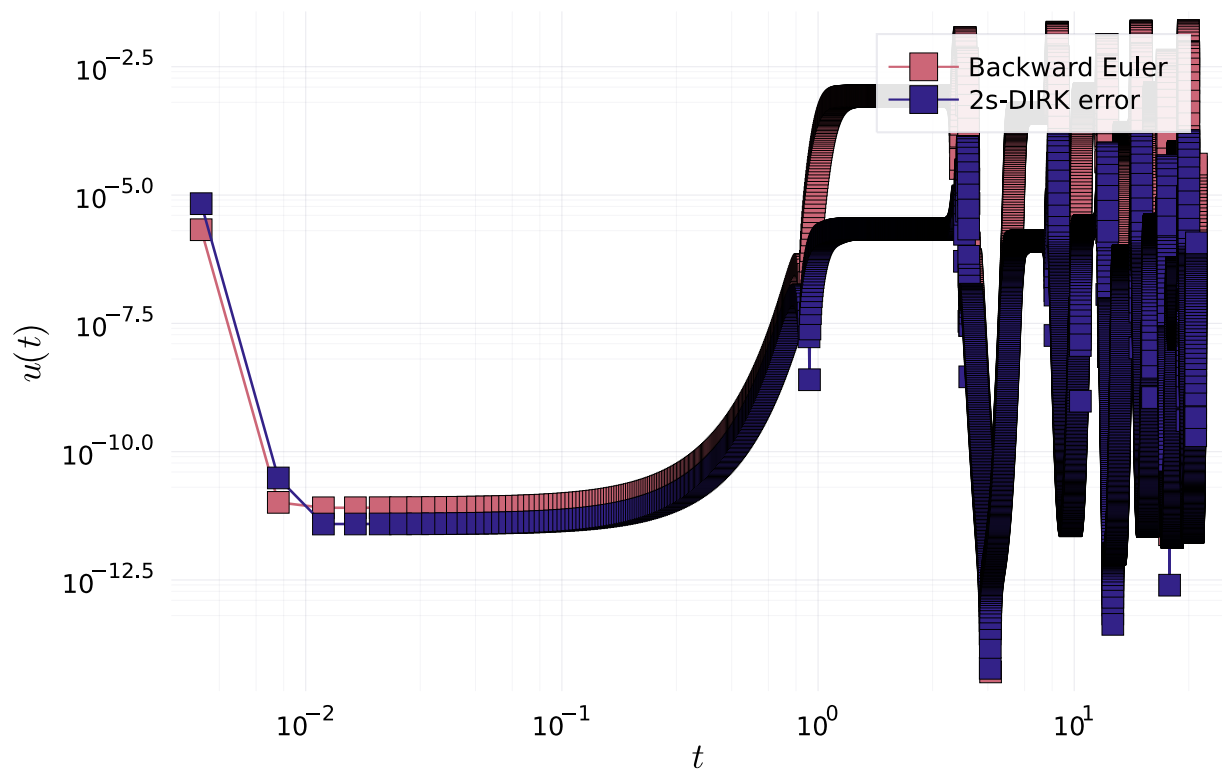
*Error Estimate,  $h = 0.015625$*



*ErrorEstimate,  $h = 0.0078125$*



*ErrorEstimate,  $h = 0.00390625$*



### 7.0.1 Part 2

```
h = 2.0^(-7)
N = Int(T/h)
tList = collect(0:N)*(T/N)
```

```

## Backward Euler
u_euler = BackwardEuler_n(f, N, T, u0)
u_eexact = BackwardEuler_n(f, 2*N, T, u0)
euler_error = abs.(u_euler[1:N] - u_eexact[1:2:2*N])./(1-0.5^1) # first order method
p1 = plot(tList[2:N], euler_error[2:N], label = "Backward Euler", xaxis=:log,
yaxis=:log, marker = (:square,5))
xaxis!(L"t")
yaxis!(L"u(t)")
title!(latexstring("Error Estimate,h=",h))

## s2-DIRK
u_s2_DIRK = s2_DIRK(f, N, T, u0,  $\alpha$ )
u_Dexact = s2_DIRK(f, 2*N, T, u0,  $\alpha$ )
DIRK_error = abs.(u_s2_DIRK[1:N] - u_Dexact[1:2:2*N])./(1-0.5^2) # second order method
p2 = plot!(tList[2:N], DIRK_error[2:N], label = "2s-DIRK error", xaxis=:log, yaxis=:log,
marker = (:square,5))
display(p2)

```

*Error Estimate,  $h = 0.0078125$*

