

# Post\_processing

April 11, 2023

## 1 Question (5) (a)

1.1 Perform numerical experiments with implicit Euler scheme. Use the same equation and parameters provided in questions 3 and 4. Solver the linear system at each time step using the Jacobi method (tolerance=  $10^{-4}$ ; do not use any libraries).

## 2 Answer (5) (a):

3 For the tolerance value 1e-4

4 Import the necessary libraries

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

## 5 Reading the name of the files generated

```
[2]: Output_files = pd.read_csv("Output_file_names.csv",delimiter=",",header=None).
    ↪to_numpy()
Output_files = np.squeeze(Output_files)
Output_files = Output_files.tolist()
```

### 5.1 Output files generated

```
[3]: Output_files
```

```
[3]: ['Question_5_Implicit_u_n_t_0.400000_N_32_rd_0.500000_.csv',
'Analytical_Solution_u_n_t_0.400000_N_32_rd_0.500000_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_32_rd_0.500000_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_32_rd_0.166667_.csv',
'Analytical_Solution_u_n_t_0.400000_N_32_rd_0.166667_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_32_rd_0.166667_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_64_rd_0.500000_.csv',
'Analytical_Solution_u_n_t_0.400000_N_64_rd_0.500000_.csv',
```

```

'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_64_rd_0.500000_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_64_rd_0.166667_.csv',
'Analytical_Solution_u_n_t_0.400000_N_64_rd_0.166667_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_64_rd_0.166667_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_128_rd_0.500000_.csv',
'Analytical_Solution_u_n_t_0.400000_N_128_rd_0.500000_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_128_rd_0.500000_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_128_rd_0.166667_.csv',
'Analytical_Solution_u_n_t_0.400000_N_128_rd_0.166667_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_128_rd_0.166667_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_256_rd_0.500000_.csv',
'Analytical_Solution_u_n_t_0.400000_N_256_rd_0.500000_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_256_rd_0.500000_.csv',
'Question_5_Implicit_u_n_t_0.400000_N_256_rd_0.166667_.csv',
'Analytical_Solution_u_n_t_0.400000_N_256_rd_0.166667_.csv',
'Question_5_Implicit_Average_Error_u_n_t_0.400000_N_256_rd_0.166667_.csv']

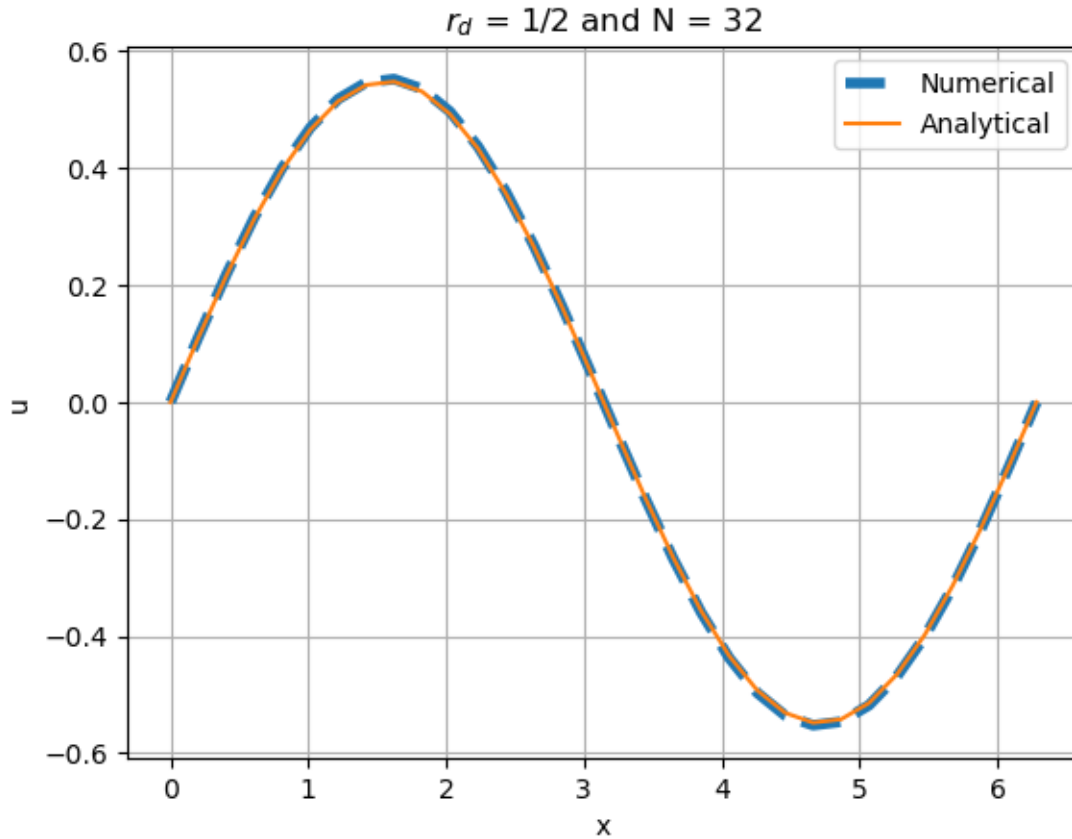
```

## 5.2 Plotting the results for $rd = 1/2$ and $N = 32$ at $t_{end} = 0.4$

```

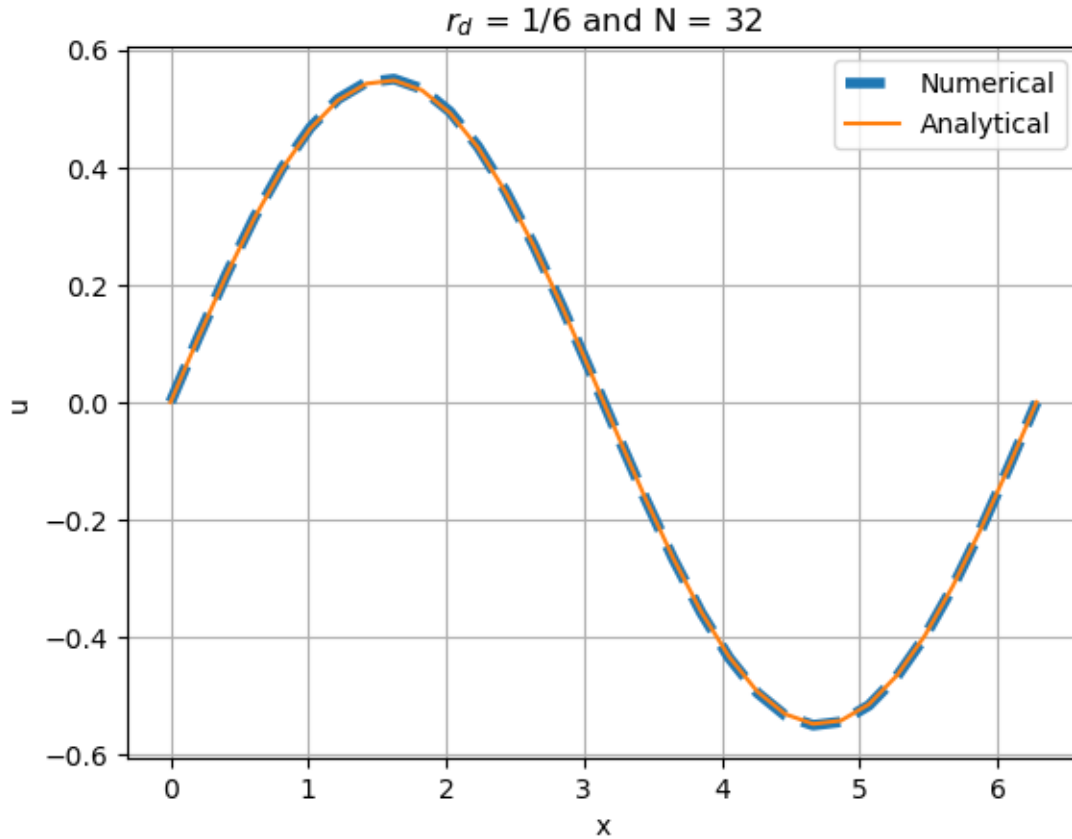
[4]: Numerical_Solution_N_32_rd_half = pd.read_csv(Output_files[0], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_32_rd_half = pd.read_csv(Output_files[1], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_32_rd_half.shape[0])
plt.plot(x, Numerical_Solution_N_32_rd_half, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_32_rd_half)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d = 1/2$ and $N = 32$")
plt.grid()
plt.savefig("r_d_1_2_and_N_32.png", dpi = 1000)
plt.show()

```



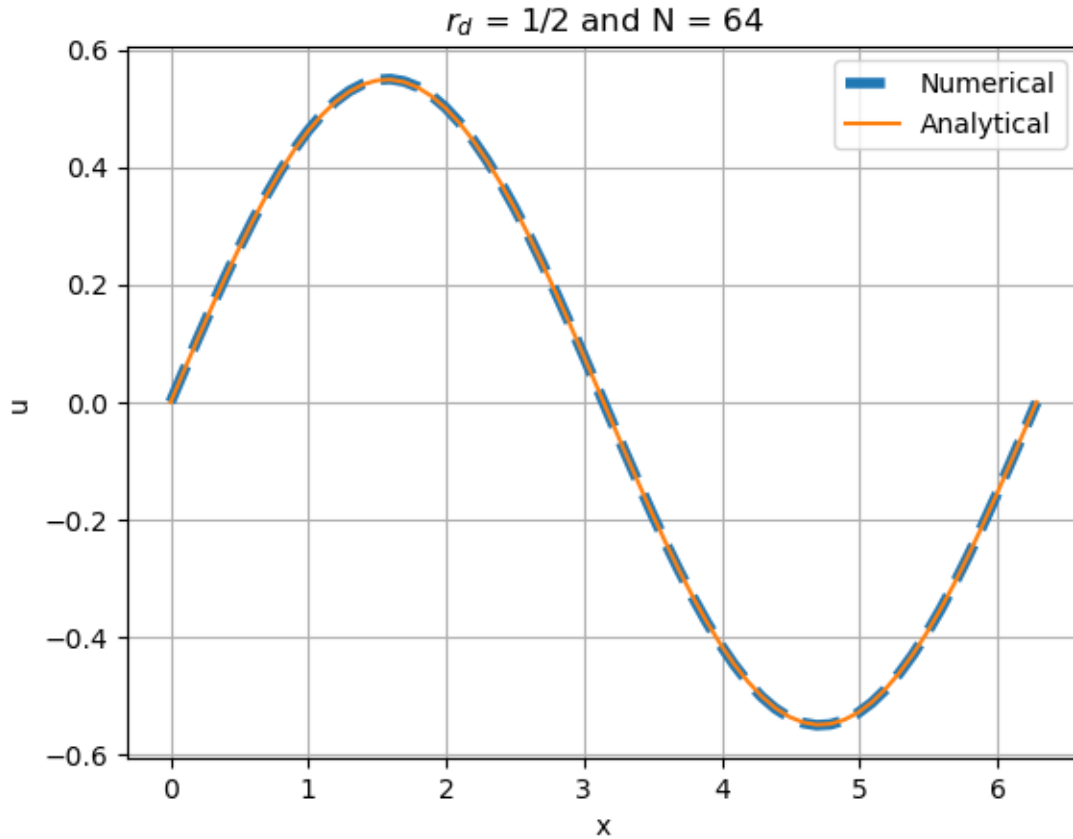
### 5.3 Plotting the results for $rd = 1/6$ and $N = 32$ at $t_{end} = 0.4$

```
[5]: Numerical_Solution_N_32_rd_1_6 = pd.read_csv(Output_files[3], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_32_rd_1_6 = pd.read_csv(Output_files[4], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_32_rd_1_6.shape[0])
plt.plot(x, Numerical_Solution_N_32_rd_1_6, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_32_rd_1_6)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d = 1/6$ and $N = 32$")
plt.grid()
plt.savefig("r_d_1_6_and_N_32.png", dpi = 1000)
plt.show()
```



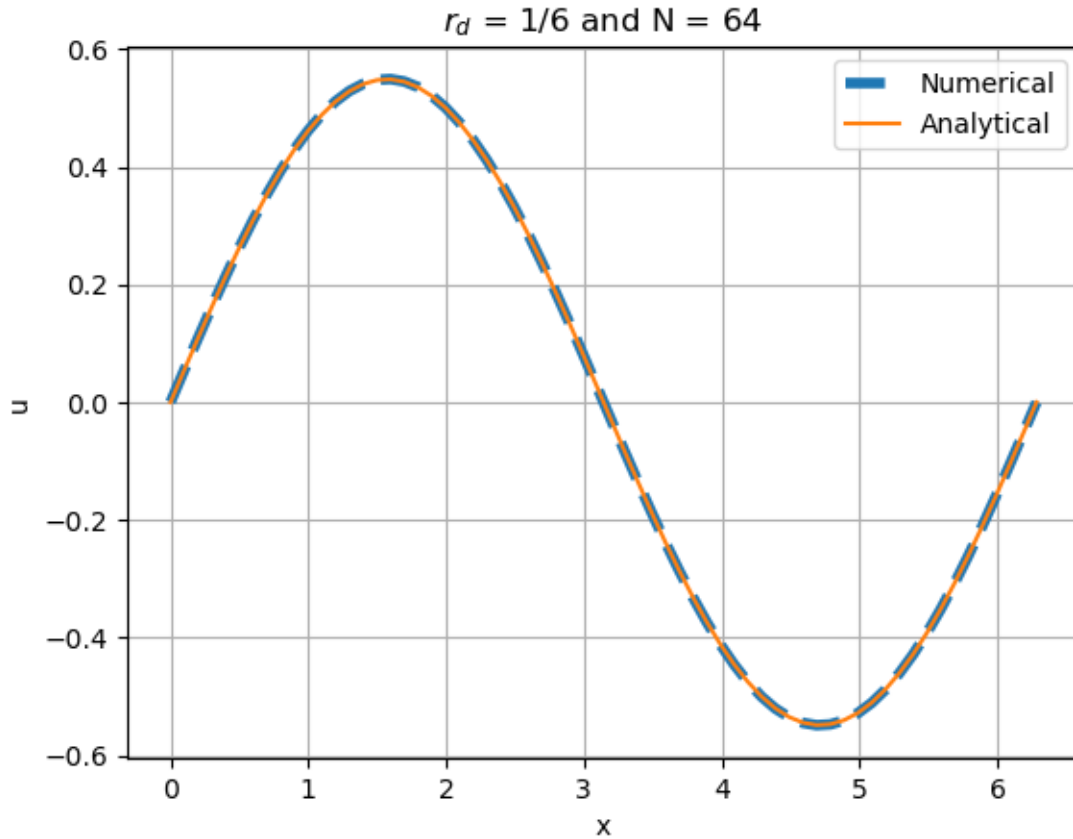
#### 5.4 Plotting the results for $r_d = 1/2$ and $N = 64$ at $t_{end} = 0.4$

```
[6]: Numerical_Solution_N_64_rd_half = pd.read_csv(Output_files[6], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_64_rd_half = pd.read_csv(Output_files[7], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_64_rd_half.shape[0])
plt.plot(x, Numerical_Solution_N_64_rd_half, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_64_rd_half)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d = 1/2$ and $N = 64$")
plt.grid()
plt.savefig("r_d_1_2_and_N_64.png", dpi = 1000)
plt.show()
```



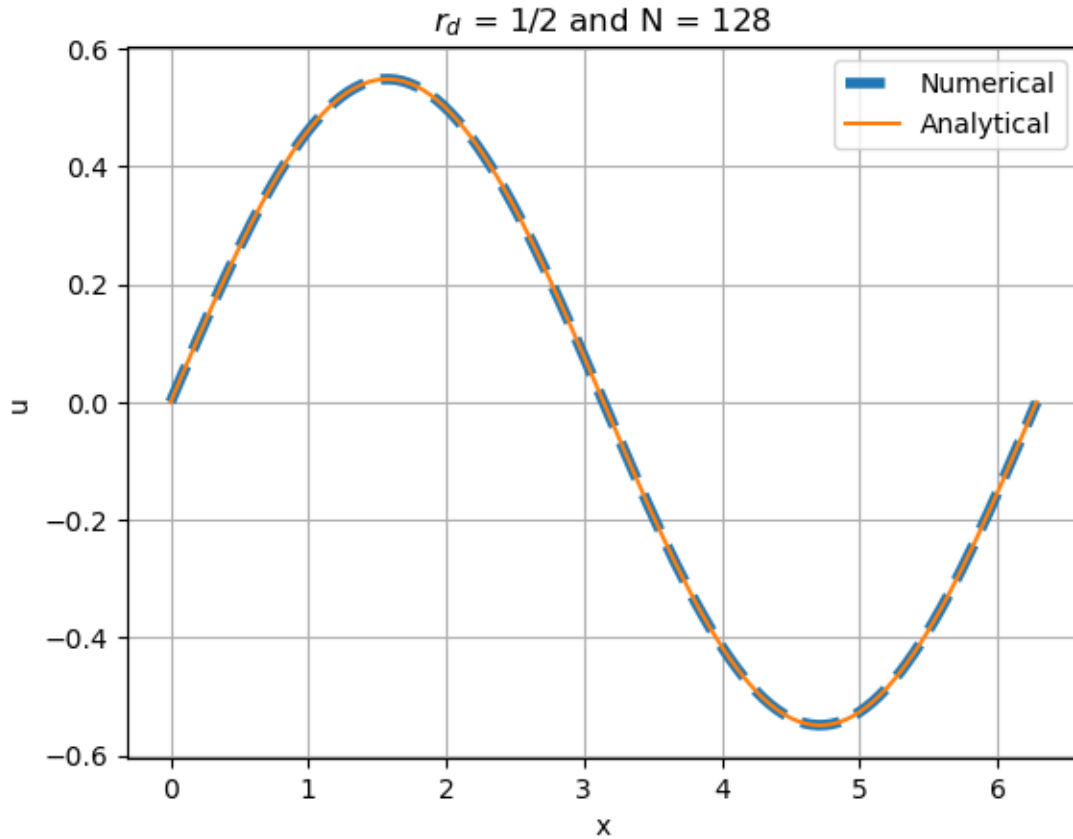
### 5.5 Plotting the results for $rd = 1/6$ and $N = 64$ at $t_{end} = 0.4$

```
[7]: Numerical_Solution_N_64_rd_1_6 = pd.read_csv(Output_files[9], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_64_rd_1_6 = pd.read_csv(Output_files[10], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_64_rd_1_6.shape[0])
plt.plot(x, Numerical_Solution_N_64_rd_1_6, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_64_rd_1_6)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d$ = 1/6 and N = 64")
plt.grid()
plt.savefig("r_d_1_6_and_N_64.png", dpi = 1000)
plt.show()
```



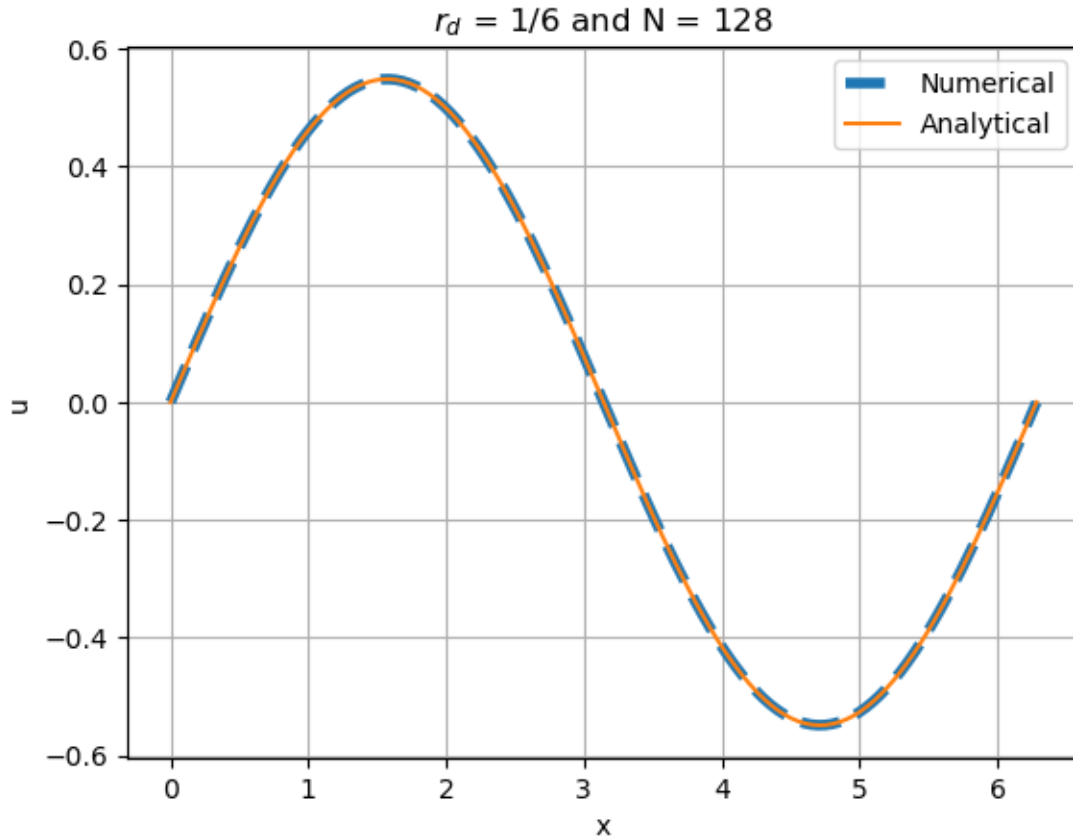
## 5.6 Plotting the results for $r_d = 1/2$ and $N = 128$ at $t_{end} = 0.4$

```
[8]: Numerical_Solution_N_128_rd_half = pd.read_csv(Output_files[12], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_128_rd_half = pd.read_csv(Output_files[13], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_128_rd_half.shape[0])
plt.plot(x, Numerical_Solution_N_128_rd_half, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_128_rd_half)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d = 1/2$ and $N = 128$")
plt.grid()
plt.savefig("r_d_1_2_and_N_128.png", dpi = 1000)
plt.show()
```



### 5.7 Plotting the results for $rd = 1/6$ and $N = 128$ at $t_{end} = 0.4$

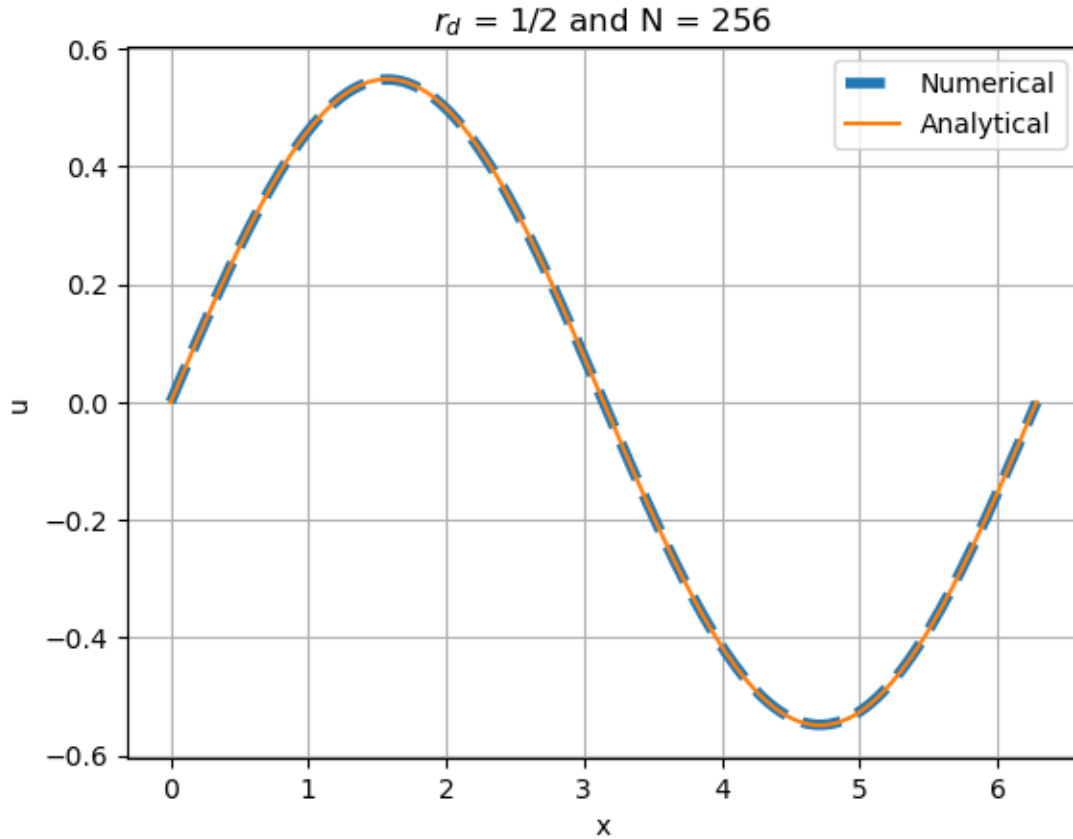
```
[9]: Numerical_Solution_N_128_rd_1_6 = pd.read_csv(Output_files[15], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_128_rd_1_6 = pd.read_csv(Output_files[16], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_128_rd_1_6.shape[0])
plt.plot(x, Numerical_Solution_N_128_rd_1_6, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_128_rd_1_6)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d$ = 1/6 and N = 128")
plt.grid()
plt.savefig("r_d_1_6_and_N_128.png", dpi = 1000)
plt.show()
```



### 5.8 Plotting the results for $rd = 1/2$ and $N = 256$ at $t_{end} = 0.4$

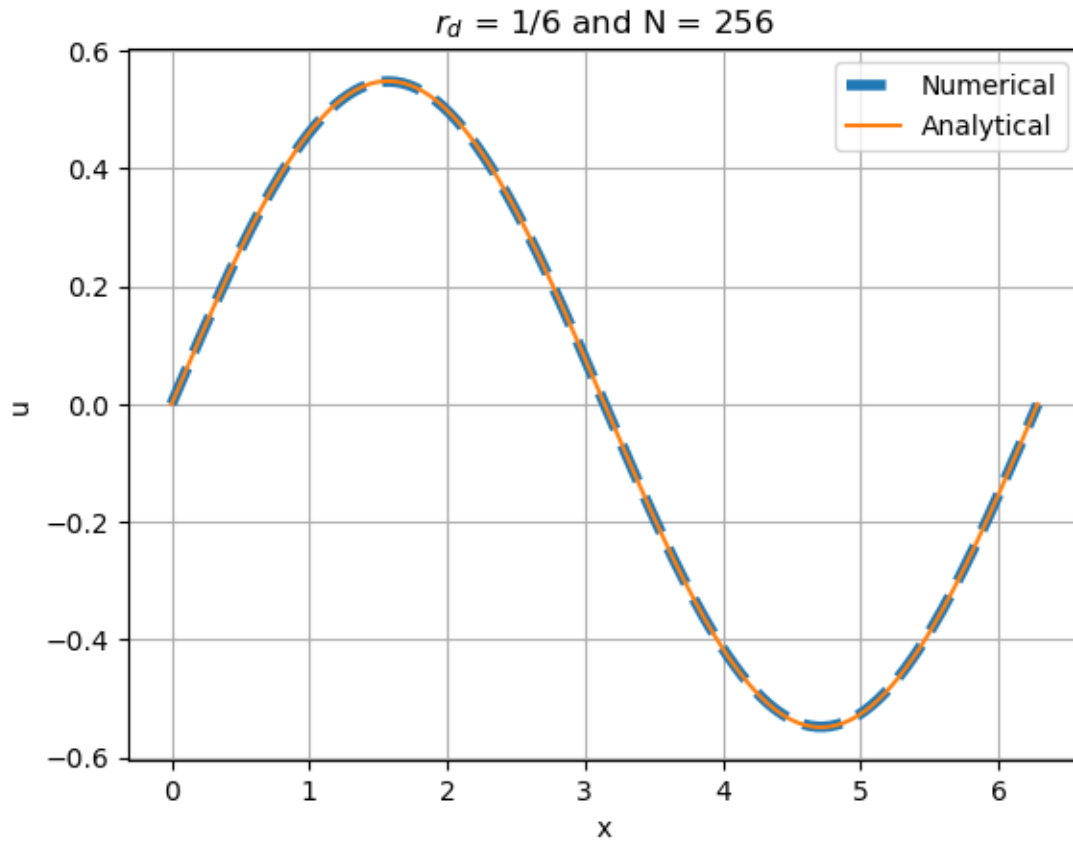
```
[10]: Numerical_Solution_N_256_rd_half = pd.read_csv(Output_files[18], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_256_rd_half = pd.read_csv(Output_files[19], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_256_rd_half.shape[0])
plt.plot(x, Numerical_Solution_N_256_rd_half, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_256_rd_half)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d$ = 1/2 and N = 256")
plt.grid()
plt.savefig("r_d_1_2_and_N_256.png", dpi = 1000)
plt.show()
```





### 5.9 Plotting the results for $rd = 1/6$ and $N = 256$ at $t_{end} = 0.4$

```
[11]: Numerical_Solution_N_256_rd_1_6 = pd.read_csv(Output_files[21], delimiter = "\t", header=None).to_numpy()
Analytical_Solution_N_256_rd_1_6 = pd.read_csv(Output_files[22], delimiter = "\t", header=None).to_numpy()
x = np.linspace(0, 2*np.pi, Numerical_Solution_N_256_rd_1_6.shape[0])
plt.plot(x, Numerical_Solution_N_256_rd_1_6, linestyle='dashed', linewidth=4)
plt.plot(x, Analytical_Solution_N_256_rd_1_6)
plt.xlabel("x")
plt.ylabel("u")
plt.legend(["Numerical", "Analytical"])
plt.title(R"$r_d$ = 1/6 and N = 256")
plt.grid()
plt.savefig("r_d_1_6_and_N_256.png", dpi = 1000)
plt.show()
```



## 6 Average error vs step size for different values of $r_d$ in log-log scale

```
[12]: rd = []
for i in range(2,len(Output_files),3):
    temp = pd.read_csv(Output_files[i], delimiter = ",",header=None).to_numpy().
    ↪squeeze()
    rd.append(temp.item())
rd_half = []
for i in range(0,len(rd),2):
    temp = rd[i]
    rd_half.append(temp)
rd_one_sixth = []
for i in range(1,len(rd),2):
    temp = rd[i]
    rd_one_sixth.append(temp)
```

```
[13]: rd
```

```
[13]: [0.00268342,  
      0.00135759,  
      0.000577935,  
      0.000241834,  
      3.84373e-05,  
      3.5107e-05,  
      0.000375391,  
      0.000379308]
```

```
[14]: rd_half
```

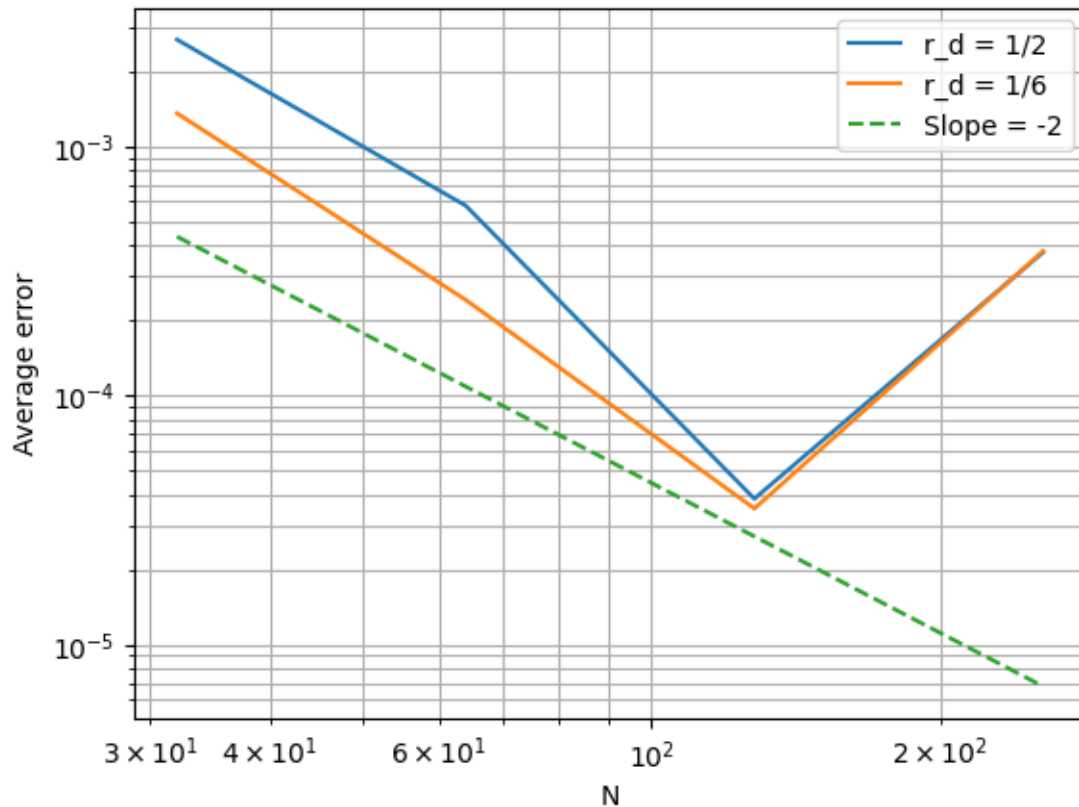
```
[14]: [0.00268342, 0.000577935, 3.84373e-05, 0.000375391]
```

```
[15]: rd_one_sixth
```

```
[15]: [0.00135759, 0.000241834, 3.5107e-05, 0.000379308]
```

```
[16]: N = [32,64,128,256]
```

```
[17]: fig = plt.figure()  
      plt.loglog(N,rd_half)  
      plt.loglog(N,rd_one_sixth)  
      plt.plot(N,(1.50*np.array(N))**(-2),"--")  
      plt.legend(["r_d = 1/2", "r_d = 1/6", "Slope = -2"])  
      plt.xlabel("N")  
      plt.ylabel("Average error")  
      plt.grid(which='both')  
      plt.show()  
      fig.savefig("Average_Error_log.png",dpi = 500, bbox_inches="tight")
```



## 7 Question (5) (b):

7.1 In a  $N$  vs  $E(N)$  graph, compare the errors with explicit method for  $rd = 1/2$ . Provide an explanation for your observations.

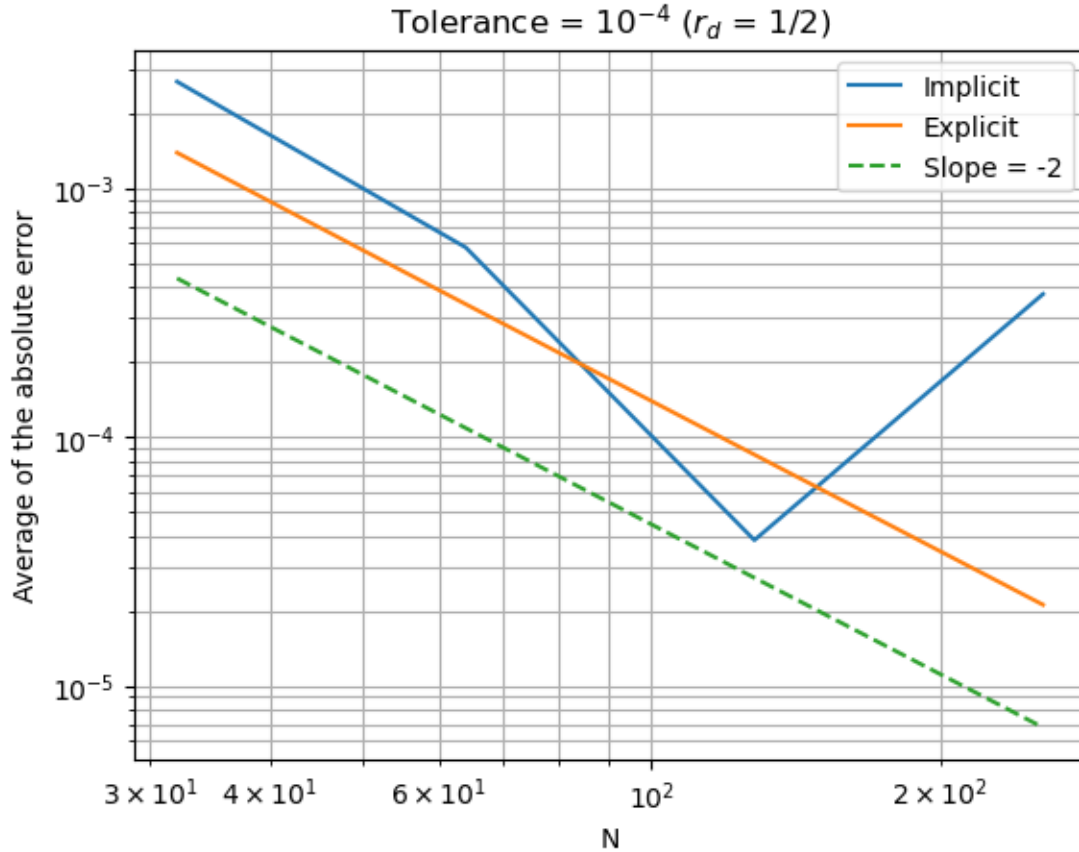
## 8 Answer (5) (b):

```
[18]: Q_4_Error_Data = pd.read_csv("Question_4_Error_Data.csv",header=None)
```

## 9 Average error vs step size for different methods in log-log scale

```
[19]: fig = plt.figure()
plt.loglog(N,rd_half)
plt.loglog(N,Q_4_Error_Data)
plt.plot(N,(1.50*np.array(N))**(-2),"--")
plt.legend(["Implicit", "Explicit", "Slope = -2"])
plt.xlabel("N")
plt.ylabel("Average of the absolute error")
plt.grid(which='both')
```

```
plt.title(R"Tolerance =  $10^{-4}$  ( $r_d = 1/2$ )")
plt.show()
fig.savefig("Average_Error_log_for_Different_Methods.png",dpi = 500,
           bbox_inches="tight")
```



- 9.1 In case of the implicit method, the truncation error dominates, when we are using lesser number of grid points but when we are using a finer grid the propagation error keep on accumulating over iterations and eventually dominates the truncation error. Hence, the error increases on using a finer grid, for a fixed tolerance.
- 9.2 We have found that on decreasing the tolerance from the  $10^{-4}$  to  $10^{-6}$  the error keeps on reducing.