

Computer_Homework1

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1 Computer Homework 1	1
1.1 Requirements	1
1.2 Installation	1
1.3 How To Use	1
1.4 Copyright	2
1.5 License	2
2 Finite difference Method	3
2.1 Complexity	4
2.2 Accuracy	4
2.3 Convergence	4
3 Theta	7
3.1 Complexity	7
3.2 Accuracy	7
3.3 Convergence	8
4 Trajectory	9
4.1 Trajectory Plot	9
5 Trajectory Error analysis	11
5.1 Error Analysis: Single precision	11
5.2 Error Analysis: Double precision	12
5.3 Numerical Error Analysis: single precision	14
6 File Index	15
6.1 File List	15
7 File Documentation	17
7.1 hw1.hpp File Reference	17
7.1.1 Detailed Description	18
7.1.2 Function Documentation	18
7.1.2.1 HW1()	18
7.2 main.cpp File Reference	19
7.2.1 Detailed Description	19
Index	21

Chapter 1

Computer Homework 1

Solve Kepler problem via finite difference Method

1.1 Requirements

To install this program, you should have

- C++ compiler like g++
- gnu make or cmake

1.2 Installation

- gnu make
 - Type make, then you can see hw1 executable file in bin directory
- cmake
 1. make build directory
 2. go to build directory and type cmake .. -DPRECISION_LEVEL precision_level
 - precision_level 0: float
 - precision_level 1: double
 3. Type make then you can see hw1 executable in build directory

1.3 How To Use

Execute hw1 then, it will interactively read

- initial condition
- number of grid points to evaluate
- output file name

Then it computes and saves solution to file. You can plot the result using usual plotting software like gnuplot

1.4 Copyright

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1.5 License

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Chapter 2

Finite difference Method

To solve Kepler problem, we need to solve

$$\frac{d^2\zeta}{dt^2} = \frac{1}{\zeta^3} - \frac{1}{\zeta^2} \quad (2.1)$$

with initial condition

$$\begin{aligned}\zeta(t_0) &= \zeta_0 \\ \zeta'(t_0) &= \zeta'_0\end{aligned}$$

To solve above 2nd order ordinary differential equation (2.1) numerically, we need to approximate 2nd derivative as finite difference. Suppose that the solution $\zeta(t)$ has continuous 4th order derivative in the Domain $[t_0, t_f]$. then

$$\zeta(x+h) = \zeta(x) + \zeta'(x)h + \frac{1}{2!}\zeta''(x)h^2 + \frac{1}{3!}\zeta'''(x)h^3 + \frac{1}{4!}\zeta^{(4)}(\eta)h^4 \quad (2.2)$$

for some $\eta(x, h) \in (t_0, t_f)$. Using 4th order Taylor approximation (2.2), we can derive following equation

$$\zeta(x-h) - 2\zeta(x) + \zeta(x+h) = h^2\zeta''(x) + O(h^4) \quad (2.3)$$

Next uniformly divide the domain $[t_0, t_f]$ into n sub intervals. let x_i be the end points of the sub intervals then for $0 \leq i \leq n$,

$$x_i = t_0 + ih \quad (2.4)$$

, where $h = (t_f - t_0)/n$. Now for $0 \leq i \leq n$, define ζ_i as following

$$\zeta_i = \zeta(x_i) \quad (2.5)$$

Then we can rewrite finite difference equation (2.3) as following

$$\zeta_{i-1} - 2\zeta_i + \zeta_{i+1} = h^2\zeta''_i + O(h^4) \quad (2.6)$$

for $1 \leq i \leq n-1$. Plug this equation (2.6) into 2nd order ode (2.1), then we have following recurrence relation

$$\zeta_{i-1} - 2\zeta_i + \zeta_{i+1} = h^2 \left(\frac{1}{\zeta_i^3} - \frac{1}{\zeta_i^2} \right) \quad (2.7)$$

To estimate global truncation error, consider $\phi_i = \zeta_{i+1} - \zeta_i$, for $i \geq 1$. Then we can rewrite above recurrence relation (2.7), as

$$\phi_i - \phi_{i-1} = h^2 \left(\frac{1}{\zeta_i^3} - \frac{1}{\zeta_i^2} \right) \quad (2.8)$$

In above equation (2.8), we truncate $O(h^4)$ terms, so local truncation error of ϕ_i is $O(h^4) = O(n^{-4})$. Therefore global truncation error of ϕ_i can be roughly estimated to $O(n^{-3})$. Note that $\zeta_i = \zeta_0 + \sum_{j=0}^{i-1} \phi_j$. Hence, the global truncation error of ζ_i could be estimated to

$$O(n^{-2}) \quad (2.9)$$

To solve recurrence relation of ζ (2.7), we need to know both ζ_0 and ζ_1 . However only ζ_0 is explicitly given by the initial condition. To approximate ζ_1 with $O(n^{-3})$ error bound, I use 2nd order Taylor expansion.

$$\zeta_1 \approx \zeta_0 + \zeta'_0 h + \frac{1}{2!} \zeta''_0 h^2 \quad (2.10)$$

ζ''_0 can be derived by 2nd order ode (2.1)

$$\zeta''_0 = \frac{1}{\zeta_0^3} - \frac{1}{\zeta_0^2} \quad (2.11)$$

2.1 Complexity

Clearly

$$O(n).$$

2.2 Accuracy

Global truncation error is roughly estimated by

$$O(n^{-2}).$$

2.3 Convergence

- Initial Condition

$$\zeta(0) = 0.9$$

$$\zeta'(0) = 0$$

- Initial time: 0
- Final time: 10

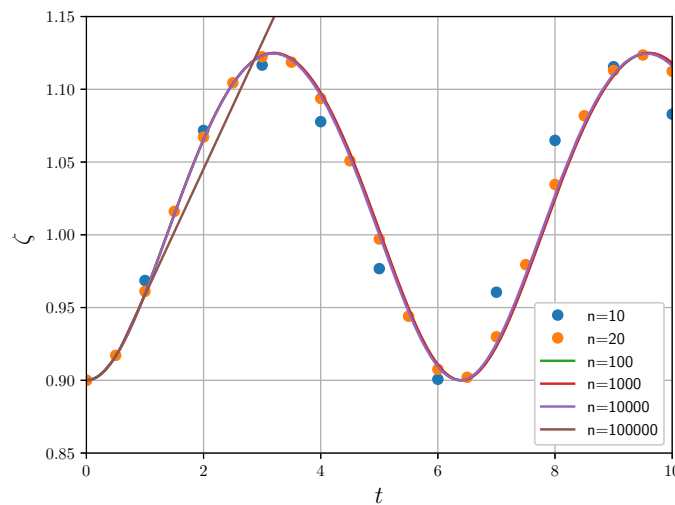


Figure 2.1 Convergence plot: single precision

Since at $n = 10^5$, difference of two successive ζ value reaches machine epsilon of single precision, roughly 10^{-7} , evaluated ζ diverges at $n = 10^5$.

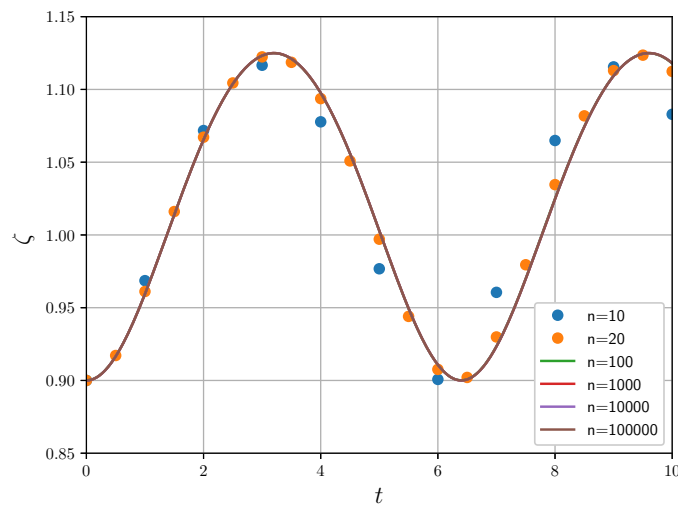


Figure 2.2 Convergence plot: double precision

However in double precision, machine epsilon, roughly 10^{-16} , is far lower than difference of two successive ζ , so calculated ζ converges to exact solution.

Chapter 3

Theta

By conservation of angular momentum, Angle θ satisfies following relation

$$\frac{d\theta}{dt} = \frac{1}{\zeta^2} \quad (3.1)$$

Integrate both side then we can deduce

$$\theta(t) = \theta_0 + \int_{t_0}^t \frac{1}{\zeta^2} dt \quad (3.2)$$

Let $\theta_i = \theta(t_i)$ as in [Finite difference Method](#), then for $1 \leq i$,

$$\theta_i = \theta_{i-1} + \int_{t_{i-1}}^{t_i} \frac{1}{\zeta^2} dt \quad (3.3)$$

Next approximate the integral using trapezoidal rule then

$$\theta_i \approx \theta_{i-1} + \frac{t_i - t_{i-1}}{2} \left(\frac{1}{\zeta_{i-1}^2} + \frac{1}{\zeta_i^2} \right) \quad (3.4)$$

θ_i has $O(n^{-3})$ local turncation error for trapezoidal rule and additional $O(n^{-3})$ for the multiplication of $t_i - t_{i-1} = (t_f - t_0)/n = O(n^{-1})$ and global turncation error of ζ (see [Finite difference Method Accuracy](#)). So the global turncation error of θ can be estimated to $O(n^{-2})$

3.1 Complexity

Clearly

$$O(n) \quad (3.5)$$

3.2 Accuracy

The global turncation error of θ is rouhtly estimated to

$$O(n^{-2}) \quad (3.6)$$

3.3 Convergence

- Initial Condition

$$\zeta(0) = 0.9$$

$$\zeta'(0) = 0$$

$$\theta(0) = 0$$

- Initial time: 0
- Final time: 10

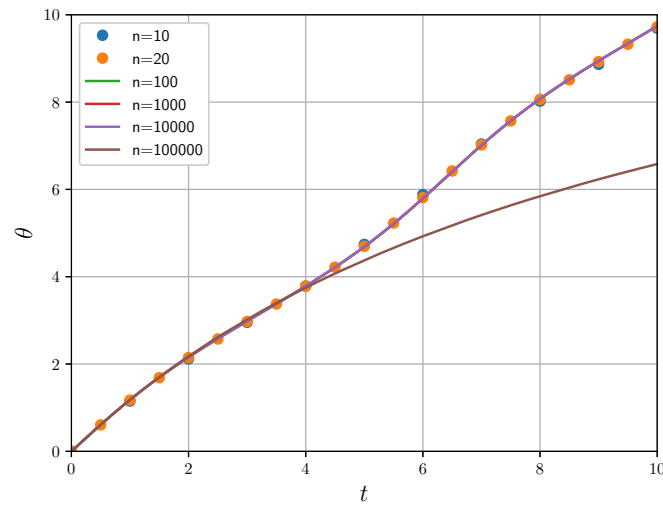


Figure 3.1 Convergence plot: single precision

Since at $n = 10^5$, difference of two successive θ value reaches machine epsilon of single precision, roughly 10^{-7} , evaluated θ diverges at $n = 10^5$.

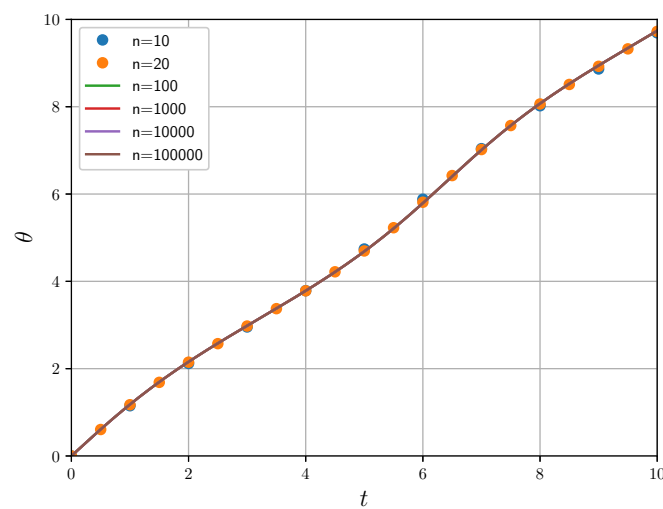


Figure 3.2 Convergence plot: double precision

However in double precision, machine epsilon, roughly 10^{-16} , is far lower than difference of two successive θ , so calculated θ converges to exact solution.

Chapter 4

Trajectory

We know that

$$\begin{aligned}x(t) &= \zeta(t) \cos \theta(t) \\y(t) &= \zeta(t) \sin \theta(t)\end{aligned}\tag{4.1}$$

Using above relation (4.1) , we can draw trajectory plot.

4.1 Trajectory Plot

- Initial Condition

$$\begin{aligned}\zeta(0) &= 0.9 \\ \zeta'(0) &= 0 \\ \theta(0) &= 0\end{aligned}$$

- Initial time: 0
- Final time: 10

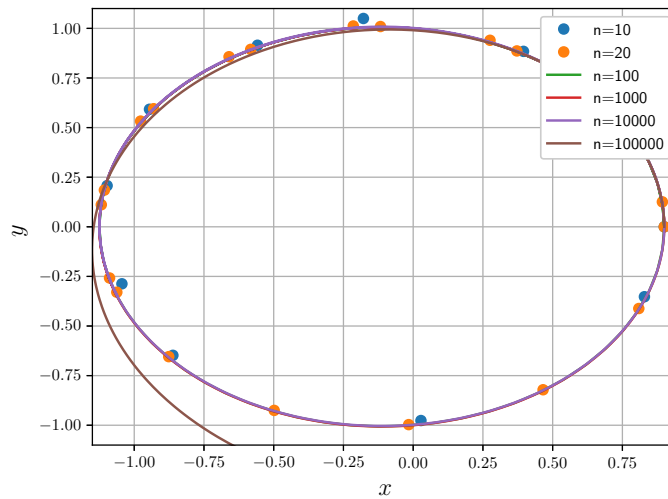


Figure 4.1 Convergence plot: single precision

In single precision, due to the numerical error, both calculated ζ and θ diverge at $n = 10^5$. So, trajectory also diverges at $n = 10^5$.

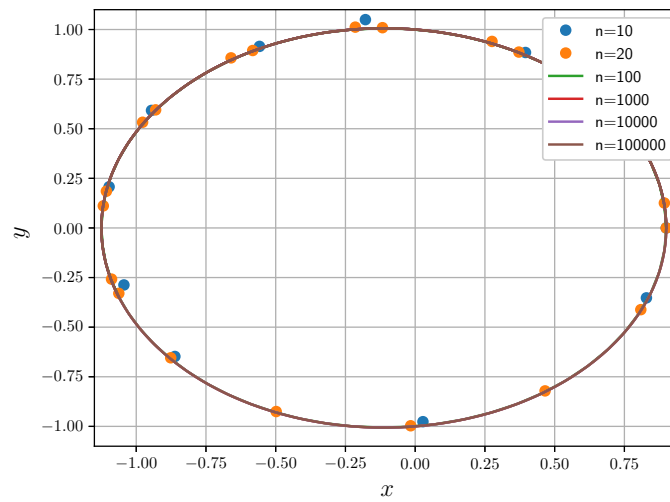


Figure 4.2 Convergence plot: double precision

However, in double precision, both calculated ζ and θ converges to exact solution. Thus, in double precision, trajectory also converges to exact ellipse trajectory.

Chapter 5

Trajectory Error analysis

Trajectory approximated by finite difference method converges to exact ellipse trajectory in double precision. Thus, I assume that double precision trajectory calculated with 10^5 steps is the exact path. Then we can estimate error at the selected points ($t = 1, 2, \dots, 10$) as

$$\text{error} = \sqrt{(x(t) - x_{ref}(t))^2 + (y(t) - y_{ref}(t))^2} \quad (5.1)$$

Error can be separated to numerical and truncation part. Usually numerical part of error increases and truncation part of error decreases as n increases.

5.1 Error Analysis: Single precision

Error Analysis: Single precision

t	$n = 10$	$n = 20$	$n = 10^2$	$n = 10^3$	$n = 10^4$
1	3.0436e-2	6.6311e-3	2.5783e-4	2.2096e-7	9.6580e-6
2	4.1441e-2	9.4500e-3	3.7512e-4	8.5674e-6	7.9277e-5
3	3.2106e-2	7.7771e-3	3.0825e-4	6.3610e-6	2.4799e-4
4	2.0137e-2	4.0393e-3	1.5794e-4	2.4904e-6	2.4253e-3
5	6.0738e-2	1.2223e-2	4.5770e-4	2.8016e-5	6.4120e-3
6	8.1767e-2	1.8988e-2	7.4039e-4	5.4708e-5	1.0189e-2
7	4.6725e-2	1.1119e-2	4.4611e-4	6.2943e-5	9.5221e-3
8	6.6307e-2	1.4042e-2	5.3708e-4	6.7596e-5	5.2319e-3
9	8.8691e-2	2.1295e-2	8.4331e-4	9.3236e-5	1.7729e-3
10	6.2815e-2	1.7533e-2	7.1144e-4	1.2318e-4	3.0507e-3

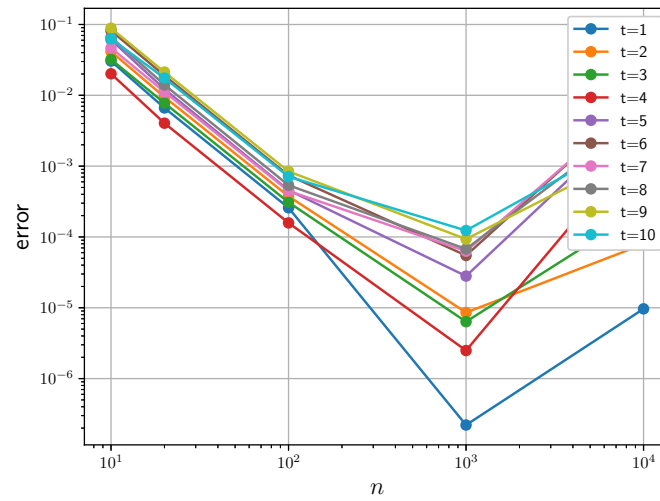


Figure 5.1 Error Analysis: single precision

In single precision, numerical error is dominant when $n > 10^3$. So, above $n = 10^3$, accuracy is worsen as n increases.

5.2 Error Analysis: Double precision

Error Analysis: Double precision

t	$n = 10$	$n = 20$	$n = 10^2$	$n = 10^3$	$n = 10^4$
1	3.0435e-2	6.6281e-3	2.5612e-4	2.5585e-6	2.5331e-8
2	4.1443e-2	9.4516e-3	3.7178e-4	3.7147e-6	3.6782e-8
3	3.2105e-2	7.7820e-3	3.1125e-4	3.1122e-6	3.0826e-8
4	2.0132e-2	4.0413e-3	1.5700e-4	1.5682e-6	1.5529e-8
5	6.0736e-2	1.2223e-2	4.6248e-4	4.6140e-6	4.5660e-8
6	8.1766e-2	1.8989e-2	7.4523e-4	7.4457e-6	7.3689e-8
7	4.6723e-2	1.1193e-2	4.5363e-4	4.5387e-6	4.4904e-8
8	6.6301e-2	1.4040e-2	5.3646e-4	5.2527e-6	5.3030e-8
9	8.8670e-2	2.1230e-2	8.4269e-4	8.4218e-6	8.3409e-8
10	6.2816e-2	1.7533e-2	7.1995e-4	7.2055e-6	7.1345e-8

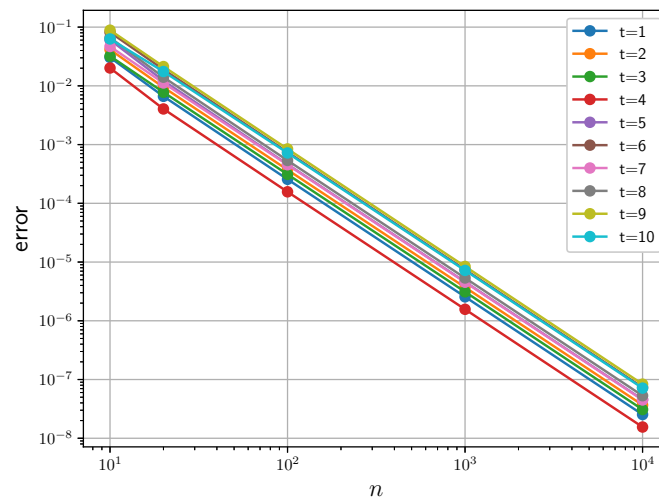


Figure 5.2 Error Analysis: double precision

However, when double precision, accuracy is better and better as n increases. Thus, we can conclude that in double precision, truncation error is much more dominant than numerical error. Therefore we can estimate the order of truncation error using above plot. Order of error is estimated to

$$O(n^{-2}) \quad (5.2)$$

Practical error bound is same as estimated error bound $O(n^{-2})$. Combining double precision error analysis table and order of error $O(n^{-2})$, we can estimate prefactor $C \leq 10$. Thus, for $t \in [0, 10]$,

$$\text{error} \leq \frac{10}{n^2} \quad (5.3)$$

5.3 Numerical Error Analysis: single precision

Since, in double precision, numerical error is neglectable in $n < 10^5$, we can estimate the numerical error of single precision results as following equation.

$$\text{error}_{num} = \sqrt{(x_{float}(t) - x_{double}(t))^2 + (y_{float}(t) - y_{double}(t))^2} \quad (5.4)$$

Numerical Error: Single precision

t	$n = 10$	$n = 20$	$n = 10^2$	$n = 10^3$	$n = 10^4$
1	1.4799e-6	2.9999e-6	1.6165e-6	2.5254e-6	9.6329e-6
2	3.9058e-6	4.5055e-6	6.2816e-6	5.2916e-6	7.9308e-5
3	1.6573e-6	4.9166e-6	4.3752e-6	4.7693e-6	2.4801e-4
4	5.4706e-6	2.5806e-6	1.0410e-6	1.8560e-6	2.4252e-3
5	2.1820e-6	5.9421e-7	6.0163e-6	2.3731e-5	6.4120e-3
6	8.9125e-7	1.4110e-6	4.8574e-6	4.7471e-5	1.0189e-2
7	3.0391e-6	1.6316e-6	8.4827e-6	5.8825e-5	9.5221e-3
8	5.8355e-6	4.3674e-6	5.9646e-6	7.1031e-5	5.2319e-3
9	6.7675e-6	2.6497e-6	6.5098e-7	1.0148e-4	1.7730e-3
10	1.7299e-6	2.4663e-6	9.2152e-6	1.2992e-4	3.0507e-3

In single precision, numerical error is less than 10^{-5} when $n < 10^3$. However, at $n = 10^3$, numerical error is comparable to truncation error and at $n = 10^4$, numerical error is dominant and increases as n greater. Therefore, in single precision, we need to concern about numerical error and to find optimal step number n_{opt} .

Chapter 6

File Index

6.1 File List

Here is a list of all documented files with brief descriptions:

main.cpp	Main program for homework1 of Computer1 class in Yonsei University Interactively reads initial condition, number of grid points to evaluate and output file name then computes and saves solution	19
hw1.hpp	Header file for homework1 of Computer1 class in Yonsei University Use finite difference method to solve Kepler problem	17

Chapter 7

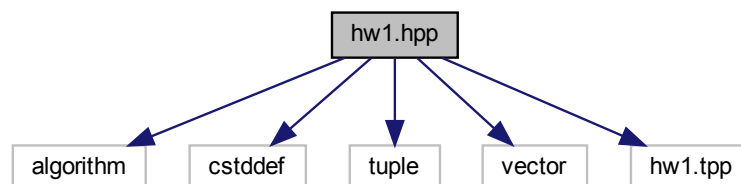
File Documentation

7.1 hw1.hpp File Reference

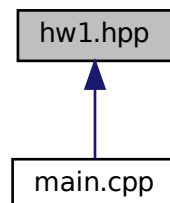
Header file for homework1 of Computer1 class in Yonsei University Use finite difference method to solve Kepler problem.

```
#include <algorithm>
#include <cstdint>
#include <tuple>
#include <vector>
#include "hw1.hpp"
```

Include dependency graph for hw1.hpp:



This graph shows which files directly or indirectly include this file:



Functions

- `template<typename T >`
`std::tuple< std::vector< T >, std::vector< T >, std::vector< T > > HW1 (T t0, T t1, std::size_t n, T y0, T y0p, T theta0)`

HW1: Solve Kepler problem via finite difference Method Behavior of HW1 is undefined when type T is not equal to one of float, double, long double.

7.1.1 Detailed Description

Header file for homework1 of Computer1 class in Yonsei University Use finite difference method to solve Kepler problem.

Author

pistack (Junho Lee)

Date

2021. 11. 10.

7.1.2 Function Documentation

7.1.2.1 HW1()

```
template<typename T >
std::tuple<std::vector<T>, std::vector<T>, std::vector<T> > HW1 (
    T t0,
    T t1,
    std::size_t n,
    T y0,
    T y0p,
    T theta0 )
```

HW1: Solve Kepler problem via finite difference Method Behavior of HW1 is undefined when type T is not equal to one of float, double, long double.

Parameters

<i>t0</i>	initial time
<i>t1</i>	final time
<i>n</i>	number of gird points to evaluate
<i>y0</i>	initial condition for zeta
<i>y0p</i>	intial condition for derivative of zeta
<i>theta0</i>	initial condition for theta

Returns

tuple of time, zeta and theta

See also

[Finite difference Method](#)

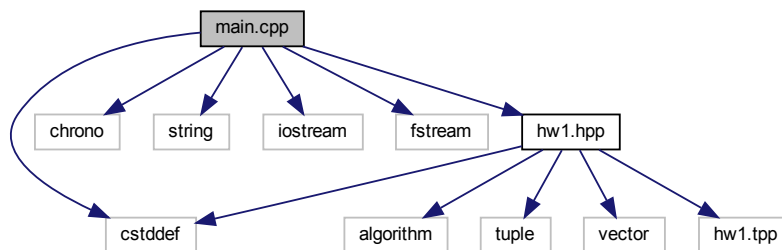
[Theta](#)

7.2 main.cpp File Reference

main program for homework1 of Computer1 class in Yonsei University Interactively reads initial condition, number of grid points to evaluate and output file name then computes and saves solution.

```
#include <cstdint>
#include <chrono>
#include <string>
#include <iostream>
#include <fstream>
#include "hw1.hpp"
```

Include dependency graph for main.cpp:



Macros

- `#define PRECISION float`
- `#define DIGITS 6`

Functions

- `int main (void)`

7.2.1 Detailed Description

main program for homework1 of Computer1 class in Yonsei University Interactively reads initial condition, number of grid points to evaluate and output file name then computes and saves solution.

Author

pistack (Junho Lee)

Date

2021. 11. 10.

Index

HW1
 hw1.hpp, [18](#)
hw1.hpp, [17](#)
 HW1, [18](#)

main.cpp, [19](#)