

# Autonomous Underwater Vehicle: A Surveillance Protocol

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# Preface

This project is an attempt at combining all of my major skills into creating a simulation, imaging, perception and control pipeline for Autonomous Underwater Vehicles (AUV). As such, creating this project involves creating a number of pipelines.

The first pipeline is the signal simulation pipeline. The signal simulation pipeline involves sea-floor point-cloud creation and simulating the signals received by the sensor arrays of the AUV. The signals recorded by the sensor-arrays on the AUV contains information from the surrounding environment. The imaging pipeline performs certain operations on the recorded signals to obtain acoustic images of the surrounding environment. To that end, this pipeline involves the topics of signal processing, linear algebra, signals and systems.

As such, the second pipeline is the imaging pipeline. The inputs to the imaging pipeline is the signals recorded by the different sensor-arrays of the AUV, in addition to the parameters of the AUV and its components. This pipeline involves match-filtering, focussing and beamforming operations to create acoustic images of the surrounding environment. Depending on the number of ULAs present, the imaging pipeline is responsible for creating multiple acoustic images in real-time. Thus, this pipeline involves the topics of Digital Signal Processing, Match-Filtering, Estimation and Detection Theory and so on.

The images created by the imaging pipeline are fed to the perception-to-control pipeline. This pipeline takes in the image formed created from the ULA signals, parameters of AUV and its components, and some historical data, it provides instructions regarding the movement of the AUV. The mapping from the inputs to the controls is called policy. Learning policies is a core part of reinforcement learning. Thus, this pipeline mainly involves the topics of reinforcement learning. And since we'll be using convolutional neural nets and transformers for learning the policies, this pipeline involves a significant amount of machine and deep learning.

The final result is an AUV that is primarily trained to map an area of the sea-floor in a constant surveillance mode. The RL-trained policy will also be trained to deal with different kinds of sea-floor terrains: those containing hills, valleys, and path-obstructing features. Due to the resource constrained nature of the marine vessel, we also prioritize efficient policies in the policy-training pipeline.

The project is currently written in C++. And since there is non-trivial amount of training and adaptive features in the pipelines, we'll be using LibTorch (the C++ API of PyTorch) to enable computation graphs, backpropagation and thereby, learning in our AUV pipeline. However, for the sections where a computation graph is not required, such as signal simulation, we will be writing templated STL code.

# Introduction

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# Chapter 1

## Setup

### 1.1 Overview

- Clone the AUV repository: `https://github.com/vrsreeganesh/AUV.git`.
- This can be performed by entering the terminal, “cd”-ing to the directory you wish and then typing: `git clone https://github.com/vrsreeganesh/AUV.git` and press enter.
- Note that in case it has not been setup, ensure github setup in the terminal. If not familiar with the whole git work-routine, I suggest sticking to Github Desktop. Its a lot easier and the best to get started right away.
- Or if you do not wish to follow a source-control approach, just download the repository as a zip file after clicking the blue code button.

# Chapter 2

## Underwater Environment Setup

### Overview

All physical matter in this framework is represented using point-clouds. Thus, the sea-floor also is represented using a number of 3D points. In addition to the coordinates, the points also have the additional property of “reflectivity”. It is the impulse response of that point.

Sea-floors in real-life are rarely flat. They often contain valleys, mountains, hills and much richer geographical features. Thus, training an agent to function in such environments call for the creation of similar structures in our simulations. Even though there must be infinite variations in the structures found under water, we shall take a constrained and structured approach to creating these variations. To that end, we shall start with an additive approach. We define few types of underwater structure whos shape, size and what not can be parameterized to enable creation of random seafloors. The full-script for creating the sea-floor is available in section ??.

### 2.1 Sea “Floor”

The first entity that we will be adding to create the seafloor is the floor itself. This is set of points that are in the lowest ring of point-clouds in the point-cloud representation of the total sea-floor.

The most basic approach to creating this is to create a flat seafloor, where all the points have the same height. While this is a good place to start, it is good to bring in some realism to the seafloor. To that end, we shall have some rolling hills as the sea-floor. Each “hill ” is created using the method outlined in Algorithm ??. The method involves deciding the location of the hills, the dimension of the hills and then designing a hill by combining an exponential function and a cosine function. We’re aiming to essentially produce gaussian-looking sea-floor hills. After the creation, this becomes the set of points representing the lowest set of points in the overall seafloor structure.

**Algorithm 1** Hill Creation

---

```

1: Input: Mean vector  $\mathbf{m}$ , Dimension vector  $\mathbf{d}$ , 2D points  $\mathbf{P}$ 
2: Output: Updated  $\mathbf{P}$  with hill heights
3:  $\text{num\_hills} \leftarrow \text{numel}(\mathbf{m}_x)$ 
4:  $H \leftarrow$  Zeros tensor of size  $(1, \text{numel}(\mathbf{P}_x))$ 
5: for  $i = 1$  to  $\text{num\_hills}$  do
6:    $x_{\text{norm}} \leftarrow \frac{\frac{\pi}{2}(\mathbf{P}_x - \mathbf{m}_x[i])}{\mathbf{d}_x[i]}$ 
7:    $y_{\text{norm}} \leftarrow \frac{\frac{\pi}{2}(\mathbf{P}_y - \mathbf{m}_y[i])}{\mathbf{d}_y[i]}$ 
8:    $h_x \leftarrow \cos(x_{\text{norm}}) \cdot e^{\frac{|x_{\text{norm}}|}{10}}$ 
9:    $h_y \leftarrow \cos(y_{\text{norm}}) \cdot e^{\frac{|y_{\text{norm}}|}{10}}$ 
10:   $h \leftarrow \mathbf{d}_z[i] \cdot h_x \cdot h_y$ 
11:  Apply boundary conditions:
12:  if  $x_{\text{norm}} > \frac{\pi}{2}$  or  $x_{\text{norm}} < -\frac{\pi}{2}$  or  $y_{\text{norm}} > \frac{\pi}{2}$  or  $y_{\text{norm}} < -\frac{\pi}{2}$  then
13:     $h \leftarrow 0$ 
14:  end if
15:   $H \leftarrow H + h$ 
16: end for
17:  $\mathbf{P} \leftarrow \text{concatenate}([\mathbf{P}, H])$ 

```

---

## 2.2 Simple Structures

### 2.2.1 Boxes

These are apartment like structures that represent different kinds of rectangular pyramids. These don't necessarily correspond to any real-life structures but these are super simple structures that will help us assess the shadows that are created in the beamformed acoustic image.



**Algorithm 2** Generate Box Meshes on Sea Floor

---

**Require:** *across\_track\_length*, *along\_track\_length*, *box\_coordinates*, *box\_reflectivity*

- 1: **Initialize** min/max width, length, height, meshdensity, reflectivity, and number of boxes
- 2: Generate random center points for boxes:
- 3:  $midxypoints \leftarrow \text{rand}([3, num\_boxes])$
- 4:  $midxypoints[0] \leftarrow midxypoints[0] \times across\_track\_length$
- 5:  $midxypoints[1] \leftarrow midxypoints[1] \times along\_track\_length$
- 6:  $midxypoints[2] \leftarrow 0$
- 7: Assign random dimensions to each box:
- 8:  $boxwidths \leftarrow \text{rand}(num\_boxes) \times (max\_width - min\_width) + min\_width$
- 9:  $boxlengths \leftarrow \text{rand}(num\_boxes) \times (max\_length - min\_length) + min\_length$
- 10:  $boxheights \leftarrow \text{rand}(num\_boxes) \times (max\_height - min\_height) + min\_height$
- 11: **for**  $i = 1$  to  $num\_boxes$  **do**
- 12:   Generate mesh points along each axis:
- 13:    $xpoints \leftarrow \text{linspace}(-boxwidths[i]/2, boxwidths[i]/2, boxwidths[i] \times meshdensity)$
- 14:    $ypoints \leftarrow \text{linspace}(-boxlengths[i]/2, boxlengths[i]/2, boxlengths[i] \times meshdensity)$
- 15:    $zpoints \leftarrow \text{linspace}(0, boxheights[i], boxheights[i] \times meshdensity)$
- 16:   Generate 3D mesh grid:
- 17:    $X, Y, Z \leftarrow \text{meshgrid}(xpoints, ypoints, zpoints)$
- 18:   Reshape  $X, Y, Z$  into 1D tensors
- 19:   Compute final coordinates:
- 20:    $boxcoordinates \leftarrow \text{cat}(X, Y, Z)$
- 21:    $boxcoordinates[0] \leftarrow boxcoordinates[0] + midxypoints[0][i]$
- 22:    $boxcoordinates[1] \leftarrow boxcoordinates[1] + midxypoints[1][i]$
- 23:    $boxcoordinates[2] \leftarrow boxcoordinates[2] + midxypoints[2][i]$
- 24:   Generate reflectivity values:
- 25:    $boxreflectivity \leftarrow meshreflectivity + \text{rand}(1, \text{size}(boxcoordinates)) - 0.5$
- 26:   Append data to final tensors:
- 27:    $box\_coordinates \leftarrow \text{cat}(box\_coordinates, boxcoordinates, 1)$
- 28:    $box\_reflectivity \leftarrow \text{cat}(box\_reflectivity, boxreflectivity, 1)$
- 29: **end for**

---

## 2.2.2 Sphere

Just like boxes, these are structures that don't necessarily exist in real life. We use this to essentially assess the shadowing in the beamformed acoustic image.

**Algorithm 3** Sphere Creation

---

**num\_hills**  $\leftarrow$  Number of Hills

---

# Chapter 3

## Hardware Setup

### Overview

The AUV contains a number of hardware that enables its functioning. A real AUV contains enough components to make a victorian child faint. And simulating the whole thing and building pipelines to model their working is not the kind of project to be handled by a single engineer. So we'll only model and simulate those components that are absolutely required for the running of these pipelines.

### 3.1 Transmitter

Probing systems are those systems that send out a signal, listen to the reflection and infer qualitative and quantitative qualities of the environment, matter or object, it was trying to infer information about. The transmitter is one of the most fundamental components of probing systems. As the name suggests, the transmitter is the equipment responsible for sending out the probing signal into the medium.

Transmitters are of many kinds. But the ones that we will be considering will be directed transmitters, which means that these transmitters have an associated beampattern. To the uninitiated, this means that the power of the transmitted signal is not transmitted in all directions equally. A beampattern is a graphical representation of the power received by an ideal receiver when placed at different angles.

Transmitters made out of a linear-array of individual transmitters use beamforming to “direct” the major power of the transmitter. These kind of systems have well studied beampatterns which can be utilized in our simulations. These kind of studies and inculcating that in our pipelines produce accurate signal simulation pipelines.

For now, we stick to a very simple model of a transmitter. We assume that the transmitter sends out the power equally into a particular cone from the AUV position.

The full-script for the setup of the transmitter is given in section ?? and the class definition for the transmitter is given in section ??.

## 3.2 Uniform Linear Array

Perhaps the most important component of probing systems are the “listening” systems. After “illuminating” the medium with the signal, we need to listen to the reflections in order to infer properties. In fact, there are some probing systems that do not use transmitter. Thus, this easily makes the case for the simple fact that the “listening” components of probing systems are the most important components of the whole system.

Uniform arrays are of many kinds but the most popular ones are uniform linear arrays and uniform planar arrays. The arrays in this case contain a number of sensors arranged in a uniform manner across a line or a plane.

Linear arrays have the property that the information obtained from elevation,  $\phi$  is no longer available due to the dimensionality of the array-structure. Thus, the images obtained from processing the signals recorded by a uniform linear array will only have two-dimensions: the azimuth,  $\theta$  and the range,  $r$ .

Thus, for 3D imaging, we shall be working with planar arrays. However, due to the higher dimensionality of the output signal, the class of algorithms required to create 3D images are a lot more computationally efficient. In addition, due to the simpler nature of the protocols involved with our AUV, uniform linear arrays will work just fine.

## 3.3 Marine Vessel

“Marine Vessel” refers to the platform on which the previously mentioned components are mounted on. These usually range from ships to submarines to AUVs. In our context, since we’re working with the AUV, the marine vessel in our case is the AUV.

The standard AUV has four degrees of freedom. Unlike drones that has practically all six degrees of freedom, AUV’s are two degrees short. However, that is okay for the functionalities most drones are designed for. But for now, we’re allowing the simulation to create a drone that has all six degrees of freedom. This will soon be patched.

# Chapter 4

## Signal Simulation

### Overview

- Define LFM.
- Define shadowing.
- Simulate Signals (basic)
- Simulate Signals with additional effects (doppler)

### 4.1 Transmitted Signal

- In probing systems, which are systems which transmit a signal and infer qualitative and quantitative characteristics of the environment from the signal return, the ideal signal is the Dirac delta signal. However, Dirac-deltas are nearly impossible to create because of their infinite bandwidth structure. Thus, we need to use something else that is more practical but at the same time, gets us quite close to the Dirac-delta. So we use something of a watered-down delta-function, which is a bandlimited delta function, or the linear frequency-modulated signal. The LFM is a signal whose frequency increases linearly in its duration. This means that the signal has a flat magnitude spectrum but quadratic phase.
- The LFM is characterised by the bandwidth and the center-frequency. The higher the resolution required, the higher the transmitted bandwidth is. So bandwidth is a characterizing factor. The higher the bandwidth, the better the resolution obtained.
- The transmitted signals used in these cases depend highly on the kind of SONAR we're using it for. The systems we're using currently contain one FLS and two side-scan or 3 FLS (I'm yet to make up mind here).
- The signal is defined in setup-script of the transmitter. Please refer to section: ??;

## 4.2 Signal Simulation

1. The signals simulation is performed using simple ray-tracing. The distance travelled from the transmitted to scatterer and then the sensor is calculated for each scatter-sensor pair. And the transmitted signal is placed at the recording of each sensor corresponding to each scatterer.
2. First we obtain the set of scatterers that reflect the transmitted signal.
3. The distance between all the sensors and the scatterer distances are calculated.
4. The time of flight from the transmitter to each scatterer and each sensor is calculated.
5. This time is then calculated into sample number by multiplying with the sampling-frequency of the uniform linear arrays.
6. We then build a signal matrix that has the dimensions corresponding to the number of samples that are recorded and the number of sensors that are present in the sensor-array.
7. We place impulses in the points corresponding to when the signals arrives from the scatterers. The result is a matrix that has x-dimension as the number of samples and the y-dimension as the number of sensors.
8. Each column is then convolved (linearly convolved) with the transmitted signal. The resulting matrix gives us the signal received by each sensor. Note that this method doesn't consider doppler effects. This will be added later.

---

### Algorithm 4 Signal Simulation

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

## 4.3 Ray Tracing

- There are multiple ways for ray-tracing.
- The method implemented during the FBLS and SS SONARs weren't super efficient as it involved pair-wise dot-products. Which becomes an issue when the number of points are increased, which is the case when the range is super high or the beamwidth is super high.

### 4.3.1 Pairwise Dot-Product

- In this method, given the coordinates of all points that are currently in the illumination cone, we find the cosines between every possible pairs of points.
- This is where the computational complexity arises as the number of dot products increase exponentially with increasing number of points.

- This method is a liability when it comes to situations where the range is super high or when the angle-beamwidth is non-narrow.

### 4.3.2 Range Histogram Method

- Given the angular beamwidths: azimuthal beamwidth and elevation beamwidth, we quantize square cone into a number of different values (note that the square cone is not an issue as the step before ensures conical subsetting).
- We split the points into different "range-cells".
- For each range-cell, we make a 2D histogram of azimuths and elevations. Then within each range-cell and for each azimuth-elevation pair, we find the closest point and add it to the check-box.
- In the next range-cell, we only work with those azimuth-elevation pairs whose check-box has not been filled. Since, for the filled ones, the filled scatter will shadow the others in the following range cells.

---

#### Algorithm 5 Range Histogram Method

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

# Chapter 5

## Imaging

### Overview

- Present basebanding, low-pass filtering and decimation.
- Present beamforming.
- Present different synthetic-aperture concepts.

### 5.1 Decimation

1. Due to the large sampling-frequencies employed in imaging SONAR, it is quite often the case that the amount of samples received for just a couple of milliseconds make for non-trivial data-size.
2. In such cases, we use some smart signal processing to reduce the data-size without loss of information. This is done using the fact that the transmitted signal is non-baseband. This means that using a method known as quadrature modulation, we can maintain the information content without the humongous amount data.
3. After basebanding the signal, this process involves decimation of the signal respecting the bandwidth of the transmitted signal.

#### 5.1.1 Basebanding

1. Basebanding is performed utilizing the frequency-shifting property of the fourier transform

$$x(t)e^{j2\pi\omega_0 t} \leftrightarrow X(\omega - \omega_0)$$

2. Since we're working with digital signals, this is implemented in the following manner

$$x[n]e^{j\frac{2\pi k_0 n}{N}} \leftrightarrow X(k - k_0)$$

---

**Algorithm 6** Basebanding

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

**5.1.2 Lowpass filtering**

1. Now that we have the signal in the baseband, we lowpass filter the signal based on the bandwidth of the signal. Since we're perfectly centering the signal using  $f_c$ , we can have the cutoff-frequency of the lowpass filter to be just above half the bandwidth of the transmitted signal. Note that the signals should not be brought down back into the real-domain using `abs()` or `real()` functions since the negative frequencies are no longer symmetrical.
2. After low-pass filtering, we have a band-restricted signal that contains all of the data in the baseband. This allows for decimation, which is what we'll do in the next step.

**5.1.3 Decimation**

1. Now that we have the bandlimited signal, what we shall do is decimation. Decimation essentially involves just taking every  $n$ -th sample where  $n$  in this case is the decimation factor.
2. The resulting signal contains the same information as that of the real-sampled signal but with much less number of samples.

**5.2 Match-Filtering**

1. To understand why match-filtering is going on, it is important to understand pulse compression.
2. In "probing" systems, which are basically systems where we send out some signal, listen to the reflection and infer quantitative and qualitative aspects of the environment, the best signal is the impulse signal (see Dirac Delta). However, this signal is not practical to use. Primarily due to the very simple fact that this particular signal has a flat and infinite bandwidth. However, this signal is the idea.
3. So instead, we're left with using signals that have a finite length,  $T_{\text{Transmitted Signal}}$ . However, the issue with that is that a scatter of infinitesimal dimension produce a response that has a length of  $T_{\text{Transmitted Signal}}$ . Thus, it is important to ensure that the response of each object, scatter or what not has comparable dimensions. This is where pulse compression comes in. Using this technique, we transform the received signal to produce a signal that is as close as possible to the signal we'd receive if we were to send out a direct delta pulse.
4. Thus, this process involves something of a detection. The closest method is something of a correlation filter where we run a copy of the transmitted signal through the received recording and take inner-products at each time step (known as the cor-



relation operation). This method works great if we're in the real domain. However, thanks to the quadrature demodulation we performed, this process is now no longer valid. But the idea remains the same. The point of doing a correlation analysis is so that where there is a signal, a spike appears. The sample principle is used to develop the match-filter.

5. We want to produce a filter, which when convolved with the received signal produces a spike. Since we're trying to produce something similar to the response of an ideal transmission system, we want the output to be that of an ideal spike, which is the delta function. So we're essentially trying to find a filter, which when multiplied with the transmitted signal, produces the diract delta.
6. The answer can be found by analyzing the frequency domain. The frequency domain basis representation of the delta-function is a flat magnitude and linear phase. Thus, this means that the filter that we use on the transmitted signal must produce a flat magnitude and linear phase. The transmitted signal that we're working with, being an LFM, means that the magnitude is already flat. The phase, however, is quadratic. So we need the matched filter to have a flat magnitude and a quadratic phase that cancels away that of the transmitted signa's quadratic component. All this leads to the best candidate: the complex conjugate of the transmitted signal. However, since we're now working with the quadrature demodulated signal, the matched filter is the complex conjugate of the quadrature demodulated transmitted signal.
7. So once the filter is made, convolving that with the received signal produces a number of spikes in the processed signal. Note that due to working in the digital domain and some other factors, the spikes will not be perfect. Thus it is not safe to take the `abs()` or `real()` just yet. We'll do that after beamforming.
8. But so far, this marks the first step of the perception pipeline.

---

**Algorithm 7** Match-Filtering
 

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

## Beamforming

- Prior to imaging, we precompute the range-cell characteristics.
- In addition, we also calculate the delays given to each sensor for each of those range-azimuth combinations.
- Those are then stored as a look-up table member of the class.
- At each-time step, what we do is we buffer split the simulated/received signal into a 3D matrix, where each signal frame corresponds to the signals for a particular range-cell.
- Then for each range-cell, we beamform using the delays we precalculated. We perform this without loops in order to utilize CPU and reduce latency.

---

**Algorithm 8** Beamforming

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

# Chapter 6

## Control Pipeline

### Overview

1. The inputs to the control-pipeline is the images obtained from previous pipeline.
2. Currently the plan is to use DQN.

### DQN

1. Here we're essentially trying to create a control pipeline that performs the protocol that we need.
2. The aim of the AUV is to continuously map a particular area of the sea-floor and perform it despite the presence of sea-floor structures.
- 3.

---

#### Algorithm 9 DQN

---

**ScatterCoordinates**  $\leftarrow$   
**ScatterReflectivity**  $\leftarrow$   
**AngleDensity**  $\leftarrow$  Quantization of angles per degree.  
**AzimuthalBeamwidth**  $\leftarrow$  Azimuthal Beamwidth  
**RangeCellWidth**  $\leftarrow$  The range-cell width

---

### Artificial Acoustic Imaging

1. In order to ensure faster development, we shall start off with training the DQN algorithm with artificial acoustic images. This is rather important due to the fact that the imaging pipelines (currently) has some non-trivial latency. This means that using those pipelines to create the inputs to the DQN algorithm will skyrocket the training time.
2. So the approach that we shall be taking will be write functions to create artificial acoustic images directly from the scatterer-coordinates and scatterer-reflectivity values. The latency for these functions are negligible compared to that of beamforming-

based imaging algorithms. The function for this has been added and is available in section ?? under the function name, *nfdc\_createAcousticImage*. Please note that these functions are not to be directly called from the main function. Instead, it is expected that the main function calls the AUV classes's method, *create\_ArtificialAcousticImage*. This function calls the class ULA's method appropriately.

3. After the ULA's create their respective acoustic images, they are put together, either by dimension-wise concatenation or depth-wise concatenation and feed to the neural net to produce control sequences.
4. We need to work on the dimensions of these images though. The best thing to do right now is to finalize the transmitter and receiver parameters and then over-estimate the dimensions of the final beamforming-produced image. We shall then use these dimensions to create the artificial acoustic image and start training the policy.

---

**Algorithm 10** Artifical Acoustic Imaging

---

**ScatterCoordinates**  $\leftarrow$  Coordinates of points in the point-cloud.

**auvCoordinates**  $\leftarrow$  Coordinates of AUV/ULA.

---

# **Chapter 7**

## **Results**

# Chapter 8

## Software

### Overview

- 

## 8.1 Class Definitions

### 8.1.1 Class: Scatter

The following is the class definition used to encapsulate attributes and methods of the scatterers.

---

```
1 // // header-files
2 // #include <iostream>
3 // #include <ostream>
4 // #include <torch/torch.h>
5
6 // #pragma once
7
8 // // hash defines
9 // #ifndef PRINTSPACE
10 // #define PRINTSPACE    std::cout<<"\n\n\n\n\n\n\n\n"<<std::endl;
11 // #endif
12 // #ifndef PRINTSMALLLINE
13 // #define PRINTSMALLLINE std::cout<<"-----"<<std::endl;
14 // #endif
15 // #ifndef PRINTLINE
16 // #define PRINTLINE    std::cout<<"======"<<std::endl;
17 // #endif
18 // #ifndef DEVICE
19 //     #define DEVICE    torch::kMPS
20 //     // #define DEVICE    torch::kCPU
21 // #endif
22
23
24 // #define PI    3.14159265
25
26
27 // // function to print tensor size
28 // void print_tensor_size(const torch::Tensor& inputTensor) {
29 //     // Printing size
30 //     std::cout << "[";
31 //     for (const auto& size : inputTensor.sizes()) {
32 //         std::cout << size << ",";
33 //     }
34 //     std::cout << "\b]" <<std::endl;
```

```

35 // }
36
37 // // Scatterer Class = Scatterer Class
38 // // Scatterer Class = Scatterer Class
39 // // Scatterer Class = Scatterer Class
40 // // Scatterer Class = Scatterer Class
41 // // Scatterer Class = Scatterer Class
42 // class ScattererClass{
43 // public:
44
45 //     // public variables
46 //     torch::Tensor coordinates; // tensor holding coordinates [3, x]
47 //     torch::Tensor reflectivity; // tensor holding reflectivity [1, x]
48
49 //     // constructor = constructor
50 //     ScattererClass(torch::Tensor arg_coordinates = torch::zeros({3,1}),
51 //                   torch::Tensor arg_reflectivity = torch::zeros({3,1})):
52 //         coordinates(arg_coordinates),
53 //         reflectivity(arg_reflectivity) {}
54
55 //     // overloading output
56 //     friend std::ostream& operator<<(std::ostream& os, ScattererClass& scatterer){
57
58 //         // printing coordinate shape
59 //         os<<"\t> scatterer.coordinates.shape = ";
60 //         print_tensor_size(scatterer.coordinates);
61
62 //         // printing reflectivity shape
63 //         os<<"\t> scatterer.reflectivity.shape = ";
64 //         print_tensor_size(scatterer.reflectivity);
65
66 //         // returning os
67 //         return os;
68 //     }
69
70 //     // copy constructor from a pointer
71 //     ScattererClass(ScattererClass* scatterers){
72
73 //         // copying the values
74 //         this->coordinates = scatterers->coordinates;
75 //         this->reflectivity = scatterers->reflectivity;
76 //     }
77
78 // };
79
80 template <typename T>
81 class ScattererClass
82 {
83 public:
84     // members
85     std::vector<std::vector<T>> coordinates;
86     std::vector<T> reflectivity;
87
88     // Constructor
89     ScattererClass() {}
90
91     // Constructor
92     ScattererClass(std::vector<std::vector<T>> coordinates_arg,
93                   std::vector<T> reflectivity_arg):
94         coordinates(coordinates_arg),
95         reflectivity(reflectivity_arg) {}
96
97     // Save to CSV
98     void savetocsv(){
99         fWriteMatrix(this->coordinates, "../csv-files/coordinates.csv");
100         fWriteVector(this->reflectivity, "../csv-files/reflectivity.csv");
101     }
102 };

```

---

### **8.1.2 Class: Transmitter**

The following is the class definition used to encapsulate attributes and methods of the projectors used.





```

66 //      int decimation_factor;                      // the new decimation factor
67 //      float post_decimation_sampling_frequency;    // the new sampling frequency
68 //      torch::Tensor lowpassFilterCoefficientsForDecimation; //
69
70 //      // imaging related
71 //      float range_resolution;                      // theoretical range-resolution =  $\frac{c}{2B}$ 
72 //      float azimuthal_resolution;                  // theoretical azimuth-resolution =
73 //       $\frac{\lambda}{(N-1) \cdot \text{inter-element-distance}}$ 
74 //      float range_cell_size;                      // the range-cell quanta we're choosing for efficiency trade-off
75 //      float azimuth_cell_size;                    // the azimuth quanta we're choosing
76 //      torch::Tensor mulFFTMatrix;                 // the matrix containing the delays for each-element as a slot
77 //      torch::Tensor azimuth_centers;               // tensor containing the azimuth centers
78 //      torch::Tensor range_centers;                 // tensor containing the range-centers
79 //      int frame_size;                             // the frame-size corresponding to a range cell in a decimated signal
80
81 //      matrix
82 //      torch::Tensor matchFilter;                   // torch tensor containing the match-filter
83 //      int num_buffer_zeros_per_frame;              // number of zeros we're adding per frame to ensure no-rotation
84 //      torch::Tensor beamformedImage;               // the beamformed image
85 //      torch::Tensor cartesianImage;
86
87 //      // artificial acoustic-image related
88 //      torch::Tensor currentArtificialAcousticImage; // the acoustic image directly produced
89
90 //      // constructor
91 //      ULAClass(int numsensors = 32,
92 //               float inter_element_spacing = 1e-3,
93 //               torch::Tensor coordinates = torch::zeros({3, 2}),
94 //               float sampling_frequency = 48e3,
95 //               float recording_period = 1,
96 //               torch::Tensor location = torch::zeros({3,1}),
97 //               torch::Tensor signalMatrix = torch::zeros({1, 32}),
98 //               torch::Tensor lowpassFilterCoefficientsForDecimation = torch::zeros({1,10})):
99 //      num_sensors(numsensors),
100 //      inter_element_spacing(inter_element_spacing),
101 //      coordinates(coordinates),
102 //      sampling_frequency(sampling_frequency),
103 //      recording_period(recording_period),
104 //      location(location),
105 //      signalMatrix(signalMatrix),
106 //      lowpassFilterCoefficientsForDecimation(lowpassFilterCoefficientsForDecimation){
107 //          // calculating ULA direction
108 //          torch::Tensor sensorDirection = coordinates.slice(1, 0, 1) - coordinates.slice(1, 1, 2);
109
110 //          // normalizing
111 //          float normvalue = torch::linalg_norm(sensorDirection, 2, 0, true,
112 //          torch::kFloat).item<float>();
113
114 //          if (normvalue != 0){
115 //              sensorDirection = sensorDirection / normvalue;
116 //          }
117
118 //          // copying direction
119 //          this->sensorDirection = sensorDirection.to(DATATYPE);
120 //      }
121
122 //      // overriding printing
123 //      friend std::ostream& operator<<(std::ostream& os, ULAClass& ula){
124 //          os<<"\t number of sensors : "<<ula.num_sensors <<std::endl;
125 //          os<<"\t inter-element spacing: "<<ula.inter_element_spacing <<std::endl;
126 //          os<<"\t sensor-direction " <<torch::transpose(ula.sensorDirection, 0, 1)<<std::endl;
127 //          PRINTSMALLLINE
128 //          return os;
129 //      }
130
131 //      /* =====
132 //      Aim: Init
133 //      ===== */
134 //      void init(TransmitterClass* transmitterObj){
135 //          // calculating range-related parameters
136 //          this->range_resolution = 1500/(2 * transmitterObj->fc);
137 //          this->range_cell_size = 40 * this->range_resolution;
138 //          if (DEBUG_ULA) std::cout << "\t ULAClass::init: line 136" << std::endl;

```

```

138 //      // status printing
139 //      if (DEBUG_ULA) {
140 //          std::cout << "\t\t ULAClass::init(): this->range_resolution = " \
141 //              << this->range_resolution \
142 //              << std::endl;
143 //          std::cout << "\t\t ULAClass::init(): this->range_cell_size = " \
144 //              << this->range_cell_size \
145 //              << std::endl;
146 //      }
147 //      if (DEBUG_ULA) std::cout << "\t ULAClass::init: line 147" << std::endl;
148
149 //      // calculating azimuth-related parameters
150 //      this->azimuthal_resolution = \
151 //          (1500/transmitterObj->fc) \
152 //          /((this->num_sensors-1)*this->inter_element_spacing);
153 //      this->azimuth_cell_size = 2 * this->azimuthal_resolution;
154 //      if (DEBUG_ULA) std::cout << "\t ULAClass::init: line 154" << std::endl;
155
156 //      // creating and storing the match-filter
157 //      this->nfdc_CreateMatchFilter(transmitterObj);
158 //      if (DEBUG_ULA) std::cout << "\t ULAClass::init: line 158" << std::endl;
159 //  }
160
161 //  // Create match-filter
162 //  void nfdc_CreateMatchFilter(TransmitterClass* transmitterObj){
163
164 //      // creating matrix for basebanding the signal
165 //      torch::Tensor basebanding_vector = \
166 //          torch::linspace( \
167 //              0, \
168 //              transmitterObj->Signal.numel()-1, \
169 //              transmitterObj->Signal.numel() \
170 //          ).reshape(transmitterObj->Signal.sizes());
171 //      basebanding_vector = \
172 //          torch::exp( \
173 //              -1 * COMPLEX_1j * 2 * PI \
174 //              * (transmitterObj->fc/this->sampling_frequency) \
175 //              * basebanding_vector);
176 //      if (DEBUG_ULA) std::cout << "\t\t ULAClass::nfdc_createMatchFilter: line 176" << std::endl;
177
178 //      // multiplying the signal with the basebanding vector
179 //      torch::Tensor match_filter = \
180 //          torch::mul(transmitterObj->Signal, \
181 //              basebanding_vector);
182 //      if (DEBUG_ULA) std::cout << "\t\t ULAClass::nfdc_createMatchFilter: line 182" << std::endl;
183
184 //      // low-pass filtering to get the baseband signal
185 //      fConvolve1D(match_filter, this->lowpassFilterCoefficientsForDecimation);
186 //      if (DEBUG_ULA) std::cout << "\t\t ULAClass::nfdc_createMatchFilter: line 186" << std::endl;
187
188 //      // creating sampling-indices
189 //      int decimation_factor = \
190 //          std::floor((static_cast<float>(this->sampling_frequency)/2) \
191 //              /(static_cast<float>(transmitterObj->bandwidth)/2));
192 //      int final_num_samples = \
193 //          std::ceil(static_cast<float>(match_filter.numel())/static_cast<float>(decimation_factor));
194 //      torch::Tensor sampling_indices = \
195 //          torch::linspace(1, \
196 //              (final_num_samples-1) * decimation_factor, \
197 //              final_num_samples).to(torch::kInt) - torch::tensor({1}).to(torch::kInt);
198 //      if (DEBUG_ULA) std::cout << "ULAClass::nfdc_createMatchFilter: line 197" << std::endl;
199
200 //      // sampling the signal
201 //      match_filter = match_filter.index({sampling_indices});
202
203 //      // taking conjugate and flipping the signal
204 //      match_filter = torch::flipud( match_filter);
205 //      match_filter = torch::conj( match_filter);
206
207 //      // storing the match-filter to the class member
208 //      this->matchFilter = match_filter;
209 //  }
210
211 //  // overloading the "=" operator
212 //  ULAClass& operator=(const ULAClass& other){

```

```

213 //      // checking if copying to the same object
214 //      if(this == &other){
215 //          return *this;
216 //      }
217
218 //      // copying everything
219 //      this->num_sensors      = other.num_sensors;
220 //      this->inter_element_spacing = other.inter_element_spacing;
221 //      this->coordinates      = other.coordinates.clone();
222 //      this->sampling_frequency = other.sampling_frequency;
223 //      this->recording_period  = other.recording_period;
224 //      this->sensorDirection   = other.sensorDirection.clone();
225
226 //      // new additions
227 //      // this->location          = other.location;
228 //      this->lowpassFilterCoefficientsForDecimation = other.lowpassFilterCoefficientsForDecimation;
229 //      // this->sensorDirection   = other.sensorDirection.clone();
230 //      // this->signalMatrix      = other.signalMatrix.clone();
231
232
233 //      // returning
234 //      return *this;
235 //  }
236
237 //  // build sensor-coordinates based on location
238 //  void buildCoordinatesBasedOnLocation(){
239
240 //      // length-normalize the sensor-direction
241 //      this->sensorDirection = torch::div(this->sensorDirection, torch::linalg_norm(this->sensorDirection,
242 // \
243 //                                     2, 0, true, \
244 //                                     DATATYPE));
245
246 //      if(DEBUG_ULA) std::cout<<"\t ULAClass: line 105 \n";
247
248 //      // multiply with inter-element distance
249 //      this->sensorDirection = this->sensorDirection * this->inter_element_spacing;
250 //      this->sensorDirection = this->sensorDirection.reshape({this->sensorDirection.numel(), 1});
251 //      if(DEBUG_ULA) std::cout<<"\t ULAClass: line 110 \n";
252
253 //      // create integer-array
254 //      // torch::Tensor integer_array = torch::linspace(0, \
255 //      //                                     this->num_sensors-1, \
256 //      //                                     this->num_sensors).reshape({1,
257 //      //                                     this->num_sensors}).to(DATATYPE);
258 //      torch::Tensor integer_array = torch::linspace(0, \
259 //      //                                     this->num_sensors-1, \
260 //      //                                     this->num_sensors).reshape({1, \
261 //      //                                     this->num_sensors});
262 //      std::cout<<"integer_array = "; fPrintTensorSize(integer_array);
263 //      if(DEBUG_ULA) std::cout<<"\t ULAClass: line 116 \n";
264
265 //      //
266 //      torch::Tensor test = torch::mul(torch::tile(integer_array, {3, 1}).to(DATATYPE), \
267 //      //      torch::tile(this->sensorDirection, {1,
268 //      //      this->num_sensors}).to(DATATYPE));
269 //      this->coordinates = this->location + test;
270 //      if(DEBUG_ULA) std::cout<<"\t ULAClass: line 120 \n";
271
272 //  }
273
274 //  // signal simulation for the current sensor-array
275 //  void nfdc_simulateSignal(ScattererClass* scatterers,
276 //  //      TransmitterClass* transmitterObj){
277
278 //      // creating signal matrix
279 //      int numsamples      = std::ceil((this->sampling_frequency * this->recording_period));
280 //      this->signalMatrix = torch::zeros({numsamples, this->num_sensors}).to(DATATYPE);
281
282
283 //      // getting shape of coordinates
284 //      std::vector<int64_t> scatterers_coordinates_shape = \
285 //      scatterers->coordinates.sizes().vec();
286
287 //      // making a slot out of the coordinates
288 //      torch::Tensor slottedCoordinates = \

```

```

285 //      torch::permute(scatterers->coordinates.reshape({
286 //          scatterers_coordinates_shape[0],          \
287 //          scatterers_coordinates_shape[1],          \
288 //          1})                                     \
289 //      ), {2, 1, 0}).reshape({
290 //          1,                                          \
291 //          (int)(scatterers->coordinates.numel()/3), \
292 //          3});
293
294
295 //      // repeating along the y-direction number of sensor times.
296 //      slottedCoordinates =
297 //          torch::tile(slottedCoordinates,          \
298 //              {this->num_sensors, 1, 1});
299 //      std::vector<int64_t> slottedCoordinates_shape = \
300 //          slottedCoordinates.sizes().vec();
301
302
303 //      // finding the shape of the sensor-coordinates
304 //      std::vector<int64_t> sensor_coordinates_shape = \
305 //          this->coordinates.sizes().vec();
306
307 //      // creating a slot tensor out of the sensor-coordinates
308 //      torch::Tensor slottedSensors = \
309 //          torch::permute(this->coordinates.reshape({
310 //              sensor_coordinates_shape[0],          \
311 //              sensor_coordinates_shape[1],          \
312 //              1}), {2, 1, 0}).reshape({(int)(this->coordinates.numel()/3), \
313 //              1,                                     \
314 //              3});
315
316
317 //      // repeating slices along the x-coordinates
318 //      slottedSensors = \
319 //          torch::tile(slottedSensors,          \
320 //              {1, slottedCoordinates_shape[1], 1});
321
322 //      // slotting the coordinate of the transmitter and duplicating along dimensions [0] and [1]
323 //      torch::Tensor slotted_location = \
324 //          torch::permute(this->location.reshape({3, 1, 1}), \
325 //              {2, 1, 0}).reshape({1,1,3});
326 //      slotted_location = \
327 //          torch::tile(slotted_location, {slottedCoordinates_shape[0], \
328 //              slottedCoordinates_shape[1], \
329 //              1});
330
331
332
333 //      // subtracting to find the relative distances
334 //      torch::Tensor distBetweenScatterersAndSensors = \
335 //          torch::linalg_norm(slottedCoordinates - slottedSensors, \
336 //              2, 2, true, torch::kFloat).to(DATATYPE);
337
338 //      // subtracting distance between relative fields
339 //      torch::Tensor distBetweenScatterersAndTransmitter = \
340 //          torch::linalg_norm(slottedCoordinates - slotted_location, \
341 //              2, 2, true, torch::kFloat).to(DATATYPE);
342
343 //      // adding up the distances
344 //      torch::Tensor distOfFlight = \
345 //          distBetweenScatterersAndSensors + distBetweenScatterersAndTransmitter;
346 //      torch::Tensor timeOfFlight = distOfFlight/1500;
347 //      torch::Tensor samplesOfFlight = \
348 //          torch::floor(timeOfFlight.squeeze() \
349 //              * this->sampling_frequency);
350
351
352
353 //      // Adding pulses
354 //      #pragma omp parallel for
355 //      for(int sensor_index = 0; sensor_index < this->num_sensors; ++sensor_index){
356 //          for(int scatter_index = 0; scatter_index < samplesOfFlight[0].numel(); ++scatter_index){
357
358 //              // getting the sample where the current scatter's contribution must be placed.
359 //              int where_to_place = \

```

```

360 //          samplesOfFlight.index({sensor_index, \
361 //                                scatter_index \
362 //                                }).item<int>());
363
364 //          // checking whether that point is out of bounds
365 //          if(where_to_place >= numsamples) continue;
366
367 //          // placing a reflectivity-scaled impulse in there
368 //          this->signalMatrix.index_put_({where_to_place, sensor_index}, \
369 //          this->signalMatrix.index({where_to_place, \
370 //          sensor_index}) + \
371 //          scatterers->reflectivity.index({0, \
372 //          scatter_index})));
373 //      }
374 //  }
375
376
377
378 //  // Adding pulses
379 //  // for(int sensor_index = 0; sensor_index < this->num_sensors; ++sensor_index){
380
381 //      // indices associated with current index
382 //      torch::Tensor tensor_containing_placing_indices = \
383 //      samplesOfFlight[sensor_index].to(torch::kInt);
384
385 //      // calculating histogram
386 //      auto uniqueOutputs = at::_unique(tensor_containing_placing_indices, false, true);
387 //      torch::Tensor bruh = std::get<1>(uniqueOutputs);
388 //      torch::Tensor uniqueValues = std::get<0>(uniqueOutputs).to(torch::kInt);
389 //      torch::Tensor uniqueCounts = torch::bincount(bruh).to(torch::kInt);
390
391 //      // placing values according to histogram
392 //      this->signalMatrix.index_put_({uniqueValues.to(torch::kLong), sensor_index}, \
393 //      uniqueCounts.to(DATATYPE));
394
395 //  }
396
397 //  // Creating matrix out of transmitted signal
398 //  torch::Tensor signalTensorAsArgument = \
399 //  transmitterObj->Signal.reshape({
400 //  transmitterObj->Signal.numel(), \
401 //  1});
402 //  signalTensorAsArgument = \
403 //  torch::tile(signalTensorAsArgument, \
404 //  {1, this->signalMatrix.size(1)});
405
406
407
408 //  // convolving the pulse-matrix with the signal matrix
409 //  fConvolveColumns(this->signalMatrix, \
410 //  signalTensorAsArgument);
411
412
413 //  // trimming the convolved signal since the signal matrix length remains the same
414 //  this->signalMatrix = \
415 //  this->signalMatrix.index({
416 //  torch::indexing::Slice(0, numsamples), \
417 //  torch::indexing::Slice()});
418
419
420 //  // returning
421 //  return;
422 //  }
423
424 //  /* =====
425 //  Aim: Decimating basebanded-received signal
426 //  ===== */
427 //  void nfdc_decimateSignal(TransmitterClass* transmitterObj){
428
429 //      // creating the matrix for frequency-shifting
430 //      torch::Tensor integerArray = torch::linspace(0, \
431 //      this->signalMatrix.size(0)-1, \
432 //
433 //      this->signalMatrix.size(0)).reshape({this->signalMatrix.size(0), 1});
434 //      integerArray

```

```

434 //      integerArray          = torch::exp(COMPLEX_1j * transmitterObj->fc * integerArray);
435
436 //      // storing original number of samples
437 //      int original_signalMatrix_numsamples = this->signalMatrix.size(0);
438
439 //      // producing frequency-shifting
440 //      this->signalMatrix      = torch::mul(this->signalMatrix, integerArray);
441
442 //      // low-pass filter
443 //      torch::Tensor lowpassfilter_impulseresponse = \
444 //          this->lowpassFilterCoefficientsForDecimation.reshape( \
445 //              {this->lowpassFilterCoefficientsForDecimation.numel(), \
446 //              1});
447 //      lowpassfilter_impulseresponse = \
448 //          torch::tile(lowpassfilter_impulseresponse, \
449 //              {1, this->signalMatrix.size(1)});
450
451 //      // low-pass filtering the signal
452 //      fConvolveColumns(this->signalMatrix,
453 //          lowpassfilter_impulseresponse);
454
455 //      // Cutting down the extra-samples from convolution
456 //      this->signalMatrix = \
457 //          this->signalMatrix.index({torch::indexing::Slice(0, original_signalMatrix_numsamples), \
458 //              torch::indexing::Slice()});
459
460 //      // // Cutting off samples in the front.
461 //      // int cutoffpoint = lowpassfilter_impulseresponse.size(0) - 1;
462 //      // this->signalMatrix = \
463 //      //      this->signalMatrix.index({ \
464 //      //          torch::indexing::Slice(cutoffpoint, \
465 //      //              torch::indexing::None), \
466 //      //          torch::indexing::Slice() \
467 //      //      });
468
469 //      // building parameters for downsampling
470 //      int decimation_factor      = std::floor(this->sampling_frequency/transmitterObj->bandwidth);
471 //      this->decimation_factor      = decimation_factor;
472 //      this->post_decimation_sampling_frequency = \
473 //          this->sampling_frequency / this->decimation_factor;
474 //      int numsamples_after_decimation = std::floor(this->signalMatrix.size(0)/decimation_factor);
475
476 //      // building the samples which will be subsetted
477 //      torch::Tensor samplingIndices = \
478 //          torch::linspace(0, \
479 //              numsamples_after_decimation * decimation_factor - 1, \
480 //              numsamples_after_decimation).to(torch::kInt);
481
482 //      // downsampling the low-pass filtered signal
483 //      this->signalMatrix = \
484 //          this->signalMatrix.index({samplingIndices, \
485 //              torch::indexing::Slice()});
486
487 //      // returning
488 //      return;
489 //  }
490
491 //  /* =====
492 //  Aim: Match-filtering
493 //  ----- */
494 //  void nfdc_matchFilterDecimatedSignal(){
495
496 //      // Creating a 2D matrix out of the signal
497 //      torch::Tensor matchFilter2DMatrix = \
498 //          this->matchFilter.reshape({this->matchFilter.numel(), 1});
499 //      matchFilter2DMatrix = \
500 //          torch::tile(matchFilter2DMatrix, \
501 //              {1, this->num_sensors});
502
503
504 //      // 2D convolving to produce the match-filtering
505 //      fConvolveColumns(this->signalMatrix, \
506 //          matchFilter2DMatrix);
507
508

```

```

509 //      // Trimming the signal to contain just the signals that make sense to us
510 //      int startingpoint = matchFilter2DMatrix.size(0) - 1;
511 //      this->signalMatrix = \
512 //          this->signalMatrix.index({ \
513 //              torch::indexing::Slice(startingpoint, \
514 //                                      torch::indexing::None), \
515 //              torch::indexing::Slice()});
516
517 //      // // trimming the two ends of the signal
518 //      // int startingpoint = matchFilter2DMatrix.size(0) - 1;
519 //      // int endingpoint = this->signalMatrix.size(0) \
520 //          - matchFilter2DMatrix.size(0) \
521 //          + 1;
522 //      // this->signalMatrix = \
523 //          //      this->signalMatrix.index({ \
524 //              //          torch::indexing::Slice(startingpoint, \
525 //              //              endingpoint), \
526 //              //          torch::indexing::Slice()});
527
528
529 //  }
530
531 //  /* =====
532 //  Aim: precompute delay-matrices
533 //  ----- */
534 //  void nfdc_precomputeDelayMatrices(TransmitterClass* transmitterObj){
535
536 //      // calculating range-related parameters
537 //      int number_of_range_cells = \
538 //          std::ceil(((this->recording_period * 1500)/2)/this->range_cell_size);
539 //      int number_of_azimuths_to_image = \
540 //          std::ceil(transmitterObj->azimuthal_beamwidth / this->azimuth_cell_size);
541
542 //      // creating centers of range-cell centers
543 //      torch::Tensor range_centers = \
544 //          this->range_cell_size * \
545 //          torch::arange(0, number_of_range_cells) \
546 //          + this->range_cell_size/2;
547 //      this->range_centers = range_centers;
548
549 //      // creating discretized azimuth-centers
550 //      torch::Tensor azimuth_centers = \
551 //          this->azimuth_cell_size * \
552 //          torch::arange(0, number_of_azimuths_to_image) \
553 //          + this->azimuth_cell_size/2;
554 //      this->azimuth_centers = azimuth_centers;
555 //      this->azimuth_centers = this->azimuth_centers - torch::mean(this->azimuth_centers);
556
557 //      // finding the mesh values
558 //      auto range_azimuth_meshgrid = \
559 //          torch::meshgrid({range_centers, \
560 //                          azimuth_centers}, "ij");
561 //      torch::Tensor range_grid = range_azimuth_meshgrid[0]; // the columns are range_centers
562 //      torch::Tensor azimuth_grid = range_azimuth_meshgrid[1]; // the rows are azimuth-centers
563
564 //      // going from 2D to 3D
565 //      range_grid = \
566 //          torch::tile(range_grid.reshape({range_grid.size(0), \
567 //                                          range_grid.size(1), \
568 //                                          1}), {1,1,this->num_sensors});
569 //      azimuth_grid = \
570 //          torch::tile(azimuth_grid.reshape({azimuth_grid.size(0), \
571 //                                           azimuth_grid.size(1), \
572 //                                           1}), {1, 1, this->num_sensors});
573
574 //      // creating x_m tensor
575 //      torch::Tensor sensorCoordinatesSlot = \
576 //          this->inter_element_spacing * \
577 //          torch::arange(0, this->num_sensors).reshape({
578 //              1, 1, this->num_sensors
579 //          }).to(DATATYPE);
580
581 //      sensorCoordinatesSlot = \
582 //          torch::tile(sensorCoordinatesSlot, \
583 //                      {range_grid.size(0),\

```



```

584 //         range_grid.size(1),
585 //         1});
586
587 // // calculating distances
588 // torch::Tensor distanceMatrix = \
589 //     torch::square(range_grid - sensorCoordinatesSlot) + \
590 //     torch::mul((2 * sensorCoordinatesSlot), \
591 //         torch::mul(range_grid, \
592 //             1 - torch::cos(azimuth_grid * PI/180)));
593 // distanceMatrix = torch::sqrt(distanceMatrix);
594
595 // // finding the time taken
596 // torch::Tensor timeMatrix = distanceMatrix/1500;
597 // torch::Tensor sampleMatrix = timeMatrix * this->sampling_frequency;
598
599 // // finding the delay to be given
600 // auto bruh390 = torch::max(sampleMatrix, 2, true);
601 // torch::Tensor max_delay = std::get<0>(bruh390);
602 // torch::Tensor delayMatrix = max_delay - sampleMatrix;
603
604 // // now that we have the delay entries, we need to create the matrix that does it
605 // int decimation_factor = \
606 //     std::floor(static_cast<float>(this->sampling_frequency)/transmitterObj->bandwidth);
607 // this->decimation_factor = decimation_factor;
608
609
610 // // calculating frame-size
611 // int frame_size = \
612 //     std::ceil(static_cast<float>((2 * this->range_cell_size / 1500) * \
613 //         static_cast<float>(this->sampling_frequency)/decimation_factor));
614 // this->frame_size = frame_size;
615
616 // // calculating the buffer-zeros to add
617 // // int num_buffer_zeros_per_frame = \
618 // //     static_cast<float>(this->num_sensors - 1) * \
619 // //     static_cast<float>(this->inter_element_spacing) * \
620 // //     this->sampling_frequency / 1500;
621
622 // int num_buffer_zeros_per_frame = \
623 //     std::ceil((this->num_sensors - 1) * \
624 //         this->inter_element_spacing * \
625 //         this->sampling_frequency \
626 //         / (1500 * this->decimation_factor));
627
628 // // storing to class member
629 // this->num_buffer_zeros_per_frame = \
630 //     num_buffer_zeros_per_frame;
631
632 // // calculating the total frame-size
633 // int total_frame_size = \
634 //     this->frame_size + this->num_buffer_zeros_per_frame;
635
636 // // creating the multiplication matrix
637 // torch::Tensor mulFFTMMatrix = \
638 //     torch::linspace(0, \
639 //         total_frame_size-1, \
640 //         total_frame_size).reshape({1, \
641 //             total_frame_size, \
642 //             1}).to(DATATYPE); // creating an array 1,...,frame_size
643 // of shape [1,frame_size, 1];
644 // mulFFTMMatrix = \
645 //     torch::div(mulFFTMMatrix, \
646 //         torch::tensor(total_frame_size).to(DATATYPE)); // dividing by N
647 // mulFFTMMatrix = mulFFTMMatrix * 2 * PI * -1 * COMPLEX_1j; // creating tenosr values for -1j * 2pi *
648 // k/N
649 // mulFFTMMatrix = \
650 //     torch::tile(mulFFTMMatrix, \
651 //         {number_of_range_cells * number_of_azimuths_to_image, \
652 //             1, \
653 //             this->num_sensors}); // creating the larger tensor for it
654
655 // // populating the matrix
656 // for(int azimuth_index = 0; \
657 //     azimuth_index < number_of_azimuths_to_image; \

```

```

657 //      ++azimuth_index){
658 //      for(int range_index = 0; \
659 //          range_index < number_of_range_cells; \
660 //          ++range_index){
661 //          // finding the delays for sensors
662 //          torch::Tensor currentSensorDelays = \
663 //              delayMatrix.index({range_index, \
664 //                              azimuth_index, \
665 //                              torch::indexing::Slice()});
666 //          // reshaping it to the target size
667 //          currentSensorDelays = \
668 //              currentSensorDelays.reshape({1, \
669 //                                          1, \
670 //                                          this->num_sensors});
671 //          // tiling across the plane
672 //          currentSensorDelays = \
673 //              torch::tile(currentSensorDelays, \
674 //                          {1, total_frame_size, 1});
675 //          // multiplying across the appropriate plane
676 //          int index_to_place_at = \
677 //              azimuth_index * number_of_range_cells + \
678 //              range_index;
679 //          mulFFTMMatrix.index_put_({index_to_place_at, \
680 //                                   torch::indexing::Slice(),
681 //                                   torch::indexing::Slice()}, \
682 //                                   currentSensorDelays);
683 //      }
684 //  }
685
686 //  // storing the mulFFTMMatrix
687 //  this->mulFFTMMatrix = mulFFTMMatrix;
688 //  }
689
690 //  /* =====
691 //  Aim: Beamforming the signal
692 //  ----- */
693 //  void nfdc_beamforming(TransmitterClass* transmitterObj){
694
695 //      // ensuring the signal matrix is in the shape we want
696 //      if(this->signalMatrix.size(1) != this->num_sensors)
697 //          throw std::runtime_error("The second dimension doesn't correspond to the number of sensors \n");
698
699 //      // adding the batch-dimension
700 //      this->signalMatrix = \
701 //          this->signalMatrix.reshape({
702 //              1,
703 //              this->signalMatrix.size(0),
704 //              this->signalMatrix.size(1)});
705
706
707 //      // zero-padding to ensure correctness
708 //      int ideal_length = \
709 //          std::ceil(this->range_centers.numel() * this->frame_size);
710 //      int num_zeros_to_pad_signal_along_dimension_0 = \
711 //          ideal_length - this->signalMatrix.size(1);
712
713
714 //      // printing
715 //      if (DEBUG_ULA) PRINTSMALLLINE
716 //      if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | this->range_centers.numel()   =
717 //      "<<this->range_centers.numel() <<std::endl;
718 //      if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | this->frame_size         =
719 //      "<<this->frame_size <<std::endl;
720 //      if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | ideal_length           =
721 //      "<<ideal_length <<std::endl;
722 //      if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | this->signalMatrix.size(1)   =
723 //      "<<this->signalMatrix.size(1) <<std::endl;
724 //      if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming |
725 //      num_zeros_to_pad_signal_along_dimension_0 = "<<num_zeros_to_pad_signal_along_dimension_0 <<std::endl;
726 //      if (DEBUG_ULA) PRINTSPACE
727
728 //      // appending or slicing based on the requirements
729 //      if (num_zeros_to_pad_signal_along_dimension_0 <= 0) {
730
731 //          // sending out a warning that slicing is going on

```

```

727 //         if (DEBUG_ULA) std::cerr <<"\t\t ULAClass::nfdc_beamforming | Please note that the signal
matrix has been sliced. This could lead to loss of information"<<std::endl;
728
729 //         // slicing the signal matrix
730 //         if (DEBUG_ULA) PRINTSPACE
731 //         if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | this->signalMatrix.shape (before
slicing) = "<< this->signalMatrix.sizes().vec() <<std::endl;
732 //         this->signalMatrix = \
733 //             this->signalMatrix.index({torch::indexing::Slice(), \
734 //                                     torch::indexing::Slice(0, ideal_length), \
735 //                                     torch::indexing::Slice()});
736 //         if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming | this->signalMatrix.shape (after
slicing) = "<< this->signalMatrix.sizes().vec() <<std::endl;
737 //         if (DEBUG_ULA) PRINTSPACE
738
739 //     }
740 //     else {
741 //         // creating a zero-filled tensor to append to signal matrix
742 //         torch::Tensor zero_tensor = \
743 //             torch::zeros({this->signalMatrix.size(0), \
744 //                           num_zeros_to_pad_signal_along_dimension_0, \
745 //                           this->num_sensors}).to(DATATYPE);
746
747 //         // appending to signal matrix
748 //         this->signalMatrix = \
749 //             torch::cat({this->signalMatrix, zero_tensor}, 1);
750 //     }
751
752 //     // breaking the signal into frames
753 //     fBuffer2D(this->signalMatrix, frame_size);
754
755
756 //     // add some zeros to the end of frames to accomodate delaying of signals.
757 //     torch::Tensor zero_filled_tensor = \
758 //         torch::zeros({this->signalMatrix.size(0), \
759 //                       this->num_buffer_zeros_per_frame, \
760 //                       this->num_sensors}).to(DATATYPE);
761 //     this->signalMatrix = \
762 //         torch::cat({this->signalMatrix, \
763 //                     zero_filled_tensor}, 1);
764
765
766 //     // tiling it to ensure that it works for all range-angle combinations
767 //     int number_of_azimuths_to_image = this->azimuth_centers.numel();
768 //     this->signalMatrix = \
769 //         torch::tile(this->signalMatrix, \
770 //                     {number_of_azimuths_to_image, 1, 1});
771
772 //     // element-wise multiplying the signals to delay each of the frame accordingly
773 //     this->signalMatrix = torch::mul(this->signalMatrix, \
774 //                                     this->mulFFTMMatrix);
775
776 //     // summing up the signals
777 //     // this->signalMatrix = torch::sum(this->signalMatrix, \
778 //                                     2, \
779 //                                     true);
780 //     this->signalMatrix = torch::sum(this->signalMatrix, \
781 //                                     2, \
782 //                                     false);
783
784
785 //     // printing some stuff
786 //     if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming: this->azimuth_centers.numel() =
"<<this->azimuth_centers.numel() <<std::endl;
787 //     if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming: this->range_centers.numel() =
"<<this->range_centers.numel() <<std::endl;
788 //     if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming: total number =
"<<this->range_centers.numel() * this->azimuth_centers.numel() <<std::endl;
789 //     if (DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_beamforming: this->signalMatrix.sizes().vec() =
"<<this->signalMatrix.sizes().vec() <<std::endl;
790
791 //     // creating a tensor to store the final image
792 //     torch::Tensor finalImage = \
793 //         torch::zeros({this->frame_size * this->range_centers.numel(), \
794 //                       this->azimuth_centers.numel()}).to(torch::kComplexFloat);

```

```

795
796
797 //      // creating a loop to assign values
798 //      for(int range_index = 0; range_index < this->range_centers.numel(); ++range_index){
799 //          for(int angle_index = 0; angle_index < this->azimuth_centers.numel(); ++angle_index){
800
801 //              // getting row index
802 //              int rowindex = \
803 //                  angle_index * this->range_centers.numel() \
804 //                  + range_index;
805
806 //              // getting the strip to store
807 //              torch::Tensor strip = \
808 //                  this->signalMatrix.index({rowindex, \
809 //                                          torch::indexing::Slice()});
810
811 //              // taking just the first few values
812 //              strip = strip.index({torch::indexing::Slice(0, this->frame_size)});
813
814 //              // placing the strips on the image
815 //              finalImage.index_put_({\
816 //                  torch::indexing::Slice((range_index)*this->frame_size, \
817 //                                          (range_index+1)*this->frame_size), \
818 //                  angle_index}, \
819 //                  strip);
820
821 //          }
822 //      }
823
824 //      // saving the image
825 //      this->beamformedImage = finalImage;
826
827
828
829 //      // converting image from polar to cartesian
830 //      nfdc_PolarToCartesian();
831
832
833 //  }
834
835 //  /* =====
836 //  Aim: Converting Polar Image to Cartesian
837 //  .....
838 //  Note:
839 //      > For now, we're assuming that the r value is one.
840 //  ----- */
841 //  void nfdc_PolarToCartesian(){
842
843
844 //      // deciding image dimensions
845 //      int num_pixels_width = 128;
846 //      int num_pixels_height = 128;
847
848
849 //      // creating query points
850 //      torch::Tensor max_right = \
851 //          torch::cos(\
852 //              torch::max(\
853 //                  this->azimuth_centers \
854 //                  - torch::mean(this->azimuth_centers) \
855 //                  + torch::tensor({90}).to(DATATYPE)) \
856 //              * PI/180);
857 //      torch::Tensor max_left = \
858 //          torch::cos(\
859 //              torch::min(this->azimuth_centers \
860 //                  - torch::mean(this->azimuth_centers) \
861 //                  + torch::tensor({90}).to(DATATYPE)) \
862 //              * PI/180);
863 //      torch::Tensor max_top = torch::tensor({1});
864 //      torch::Tensor max_bottom = torch::min(this->range_centers);
865
866
867 //      // creating query points along the x-dimension
868 //      torch::Tensor query_x = \
869 //          torch::linspace(

```

```

870 //          max_left,          \
871 //          max_right,         \
872 //          num_pixels_width   \
873 //          ).to(DATATYPE);
874
875 //      torch::Tensor query_y =          \
876 //      torch::linspace(                \
877 //          max_bottom.item<float>(),    \
878 //          max_top.item<float>(),        \
879 //          num_pixels_height           \
880 //          ).to(DATATYPE);
881
882
883 //      // converting original coordinates to their corresponding cartesian
884 //      float delta_r = 1/static_cast<float>(this->beamformedImage.size(0));
885 //      float delta_azimuth =          \
886 //      torch::abs(                    \
887 //          this->azimuth_centers.index({1}) \
888 //          - this->azimuth_centers.index({0}) \
889 //          ).item<float>();
890
891
892
893 //      // getting query points
894 //      torch::Tensor range_values = \
895 //      torch::linspace(            \
896 //          delta_r,                \
897 //          this->beamformedImage.size(0) * delta_r, \
898 //          this->beamformedImage.size(0)           \
899 //          ).to(DATATYPE);
900 //      range_values = \
901 //      range_values.reshape({range_values.numel(), 1});
902 //      range_values = \
903 //      torch::tile(range_values, \
904 //          {1, this->azimuth_centers.numel()});
905
906
907 //      // getting angle-values
908 //      torch::Tensor angle_values =          \
909 //      this->azimuth_centers                 \
910 //      - torch::mean(this->azimuth_centers) \
911 //      + torch::tensor({90});
912 //      angle_values =                      \
913 //      torch::tile(                        \
914 //          angle_values,                   \
915 //          {this->beamformedImage.size(0), 1});
916
917
918 //      // converting to cartesian original points
919 //      torch::Tensor query_original_x = \
920 //      range_values * torch::cos(angle_values * PI/180);
921 //      torch::Tensor query_original_y = \
922 //      range_values * torch::sin(angle_values * PI/180);
923
924
925
926 //      // converting points to vector 2D format
927 //      torch::Tensor query_source =          \
928 //      torch::cat({                          \
929 //          query_original_x.reshape({1, query_original_x.numel()}), \
930 //          query_original_y.reshape({1, query_original_y.numel()}), \
931 //          0;
932
933
934 //      // converting reflectivity to corresponding 2D format
935 //      torch::Tensor reflectivity_vectors = \
936 //      this->beamformedImage.reshape({1, this->beamformedImage.numel()});
937
938
939 //      // creating image
940 //      torch::Tensor cartesianImageLocal = \
941 //      torch::zeros(                        \
942 //          {num_pixels_height,             \
943 //          num_pixels_width}               \
944 //          ).to(torch::kComplexFloat);

```

```

945
946 //      /*
947 //      Next Aim: start interpolating the points on the uniform grid.
948 //      */
949 //      #pragma omp parallel for
950 //      for(int x_index = 0; x_index < query_x.numel(); ++x_index){
951 //          // if(DEBUG_ULA) std::cout << "\t\t\t x_index = " << x_index << " ";
952 //          #pragma omp parallel for
953 //          for(int y_index = 0; y_index < query_y.numel(); ++y_index){
954 //              // if(DEBUG_ULA) if(y_index%16 == 0) std::cout<<". ";
955 //
956 //              // getting current values
957 //              torch::Tensor current_x = query_x.index({x_index}).reshape({1, 1});
958 //              torch::Tensor current_y = query_y.index({y_index}).reshape({1, 1});
959 //
960 //
961 //
962 //
963 //              // getting the query value
964 //              torch::Tensor query_vector = torch::cat({current_x, current_y}, 0);
965 //
966 //
967 //              // copying the query source
968 //              torch::Tensor query_source_relative = query_source;
969 //              query_source_relative = query_source_relative - query_vector;
970 //
971 //
972 //              // subsetting using absolute values and masks
973 //              float threshold = delta_r * 10;
974 //              // PRINTDOTS
975 //              auto mask_row = \
976 //                  torch::abs(query_source_relative[0]) <= threshold;
977 //              auto mask_col = \
978 //                  torch::abs(query_source_relative[1]) <= threshold;
979 //              auto mask_together = torch::mul(mask_row, mask_col);
980 //
981 //
982 //
983 //
984 //              // calculating number of points in threshold neighbourhood
985 //              int num_points_in_threshold_neighbourhood = \
986 //                  torch::sum(mask_together).item<int>();
987 //              if (num_points_in_threshold_neighbourhood == 0){
988 //                  continue;
989 //              }
990 //
991 //
992 //
993 //              // subsetting points in neighbourhood
994 //              torch::Tensor PointsInNeighbourhood = \
995 //                  query_source_relative.index({
996 //                      torch::indexing::Slice(), \
997 //                      mask_together});
998 //              torch::Tensor ReflectivitiesInNeighbourhood = \
999 //                  reflectivity_vectors.index({torch::indexing::Slice(), mask_together});
1000 //
1001 //
1002 //              // finding the distance between the points
1003 //              torch::Tensor relativeDistances = \
1004 //                  torch::linalg_norm(PointsInNeighbourhood, \
1005 //                      2, 0, true, \
1006 //                      torch::kFloat).to(DATATYPE);
1007 //
1008 //
1009 //              // calculating weighing factor
1010 //              torch::Tensor weighingFactor = \
1011 //                  torch::nn::functional::softmax( \
1012 //                      torch::max(relativeDistances)- relativeDistances, \
1013 //                      torch::nn::functional::SoftmaxFuncOptions(1));
1014 //
1015 //
1016 //              // combining intensities using distances
1017 //              torch::Tensor finalIntensity = \
1018 //                  torch::sum(
1019 //                      torch::mul(weighingFactor, \

```

```

1020 //                      ReflectivitiesInNeighbourhood));
1021
1022 //          // assigning values
1023 //          cartesianImageLocal.index_put_({x_index, y_index}, finalIntensity);
1024
1025 //      }
1026 //      // std::cout<<std::endl;
1027 //  }
1028
1029 //      // saving to member function
1030 //      this->cartesianImage = cartesianImageLocal;
1031
1032 //  }
1033
1034 //  /* =====
1035 //  Aim: create acoustic image directly
1036 //  ----- */
1037 //  void nfdc_createAcousticImage(ScattererClass* scatterers, \
1038 //                               TransmitterClass* transmitterObj){
1039
1040 //      // first we ensure that the scatterers are in our frame of reference
1041 //      scatterers->coordinates = scatterers->coordinates - this->location;
1042
1043 //      // finding the spherical coordinates of the scatterers
1044 //      torch::Tensor scatterers_spherical = fCart2Sph(scatterers->coordinates);
1045
1046 //      // note that its not precisely projection. its rotation. So the original lengths must be
1047 //      // maintained. but thats easy since the operation of putting the elevation to be zero works just fine.
1048 //      scatterers_spherical.index_put_({1, torch::indexing::Slice()}, 0);
1049
1050 //      // converting the points back to cartesian
1051 //      torch::Tensor scatterers_acoustic_cartesian = fSph2Cart(scatterers_spherical);
1052
1053 //      // removing the z-dimension
1054 //      scatterers_acoustic_cartesian = \
1055 //          scatterers_acoustic_cartesian.index({torch::indexing::Slice(0, 2, 1), \
1056 //                                              torch::indexing::Slice()});
1057
1058 //      // deciding image dimensions
1059 //      int num_pixels_x = 512;
1060 //      int num_pixels_y = 512;
1061 //      torch::Tensor acousticImage = \
1062 //          torch::zeros({num_pixels_x, \
1063 //                      num_pixels_y}).to(DATATYPE);
1064
1065 //      // finding the max and min values
1066 //      torch::Tensor min_x = torch::min(scatterers_acoustic_cartesian[0]);
1067 //      torch::Tensor max_x = torch::max(scatterers_acoustic_cartesian[0]);
1068 //      torch::Tensor min_y = torch::min(scatterers_acoustic_cartesian[1]);
1069 //      torch::Tensor max_y = torch::max(scatterers_acoustic_cartesian[1]);
1070
1071 //      // creating query grids
1072 //      torch::Tensor query_x = torch::linspace(0, 1, num_pixels_x);
1073 //      torch::Tensor query_y = torch::linspace(0, 1, num_pixels_y);
1074
1075 //      // scaling it up to image max-point spread
1076 //      query_x = min_x + (max_x - min_x) * query_x;
1077 //      query_y = min_y + (max_y - min_y) * query_y;
1078 //      float delta_queryx = (query_x[1] - query_x[0]).item<float>();
1079 //      float delta_queryy = (query_y[1] - query_y[0]).item<float>();
1080
1081 //      // creating a mesh-grid
1082 //      auto queryMeshGrid = torch::meshgrid({query_x, query_y}, "ij");
1083 //      query_x = queryMeshGrid[0].reshape({1, queryMeshGrid[0].numel()});
1084 //      query_y = queryMeshGrid[1].reshape({1, queryMeshGrid[1].numel()});
1085 //      torch::Tensor queryMatrix = torch::cat({query_x, query_y}, 0);
1086
1087 //      // printing shapes
1088 //      if(DEBUG_ULA) PRINTSMALLLINE
1089 //      if(DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_createAcousticImage: query_x.shape =
1090 //      "<<query_x.sizes().vec()<<std::endl;
1091 //      if(DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_createAcousticImage: query_y.shape =
1092 //      "<<query_y.sizes().vec()<<std::endl;
1093 //      if(DEBUG_ULA) std::cout<<"\t\t ULAClass::nfdc_createAcousticImage: queryMatrix.shape =
1094 //      "<<queryMatrix.sizes().vec()<<std::endl;

```

```

1091
1092 //      // setting up threshold values
1093 //      float threshold_value = \
1094 //          std::min(delta_queryx, \
1095 //              delta_queryy); if(DEBUG_ULA) std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage:
line 711"<<std::endl;
1096
1097 //      // putting a loop through the whole thing
1098 //      for(int i = 0; i<queryMatrix[0].numel(); ++i){
1099 //          // for each element in the query matrix
1100 //          if(DEBUG_ULA) std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line 716"<<std::endl;
1101
1102 //          // calculating relative position of all the points
1103 //          torch::Tensor relativeCoordinates = \
1104 //              scatterers_acoustic_cartesian - \
1105 //              queryMatrix.index({torch::indexing::Slice(), i}).reshape({2, 1}); if(DEBUG_ULA)
std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line 720"<<std::endl;
1106
1107 //          // calculating distances between all the points and the query point
1108 //          torch::Tensor relativeDistances = \
1109 //              torch::linalg_norm(relativeCoordinates, \
1110 //                  1, 0, true, \
1111 //                  DATATYPE);if(DEBUG_ULA) std::cout<<"\t\t\t
ULAClass::nfdc_createAcousticImage: line 727"<<std::endl;
1112 //          // finding points that are within the threshold
1113 //          torch::Tensor conditionMeetingPoints = \
1114 //              relativeDistances.squeeze() <= threshold_value;if(DEBUG_ULA) std::cout<<"\t\t\t
ULAClass::nfdc_createAcousticImage: line 729"<<std::endl;
1115
1116 //          // subsetting the points in the neighbourhood
1117 //          if(torch::sum(conditionMeetingPoints).item<float>() == 0){
1118
1119 //              // continuing implementation if there are no points in the neighbourhood
1120 //              continue; if(DEBUG_ULA) std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line
735"<<std::endl;
1121 //          }
1122 //          else{
1123 //              // creating mask for points in the neighbourhood
1124 //              auto mask = (conditionMeetingPoints == 1);if(DEBUG_ULA) std::cout<<"\t\t\t
ULAClass::nfdc_createAcousticImage: line 739"<<std::endl;
1125
1126 //              // subsetting relative distances in the neighbourhood
1127 //              torch::Tensor distanceInTheNeighbourhood = \
1128 //                  relativeDistances.index({torch::indexing::Slice(), mask});if(DEBUG_ULA)
std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line 743"<<std::endl;
1129
1130 //              // subsetting reflectivity of points in the neighbourhood
1131 //              torch::Tensor reflectivityInTheNeighbourhood = \
1132 //                  scatterers->reflectivity.index({torch::indexing::Slice(), mask});if(DEBUG_ULA)
std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line 747"<<std::endl;
1133
1134 //              // assigning intensity as a function of distance and reflectivity
1135 //              torch::Tensor reflectivityAssignment = \
1136 //                  torch::mul(torch::exp(-distanceInTheNeighbourhood), \
1137 //                      reflectivityInTheNeighbourhood);if(DEBUG_ULA) std::cout<<"\t\t\t
ULAClass::nfdc_createAcousticImage: line 752"<<std::endl;
1138 //              reflectivityAssignment = \
1139 //                  torch::sum(reflectivityAssignment);if(DEBUG_ULA) std::cout<<"\t\t\t
ULAClass::nfdc_createAcousticImage: line 754"<<std::endl;
1140
1141 //              // assigning this value to the image pixel intensity
1142 //              int pixel_position_x = i%num_pixels_x;
1143 //              int pixel_position_y = std::floor(i/num_pixels_x);
1144 //              acousticImage.index_put_({pixel_position_x, \
1145 //                  pixel_position_y}, \
1146 //                  reflectivityAssignment.item<float>());if(DEBUG_ULA)
std::cout<<"\t\t\t ULAClass::nfdc_createAcousticImage: line 761"<<std::endl;
1147 //          }
1148
1149 //      }
1150
1151 //      // storing the acoustic-image to the member
1152 //      this->currentArtificialAcousticImage = acousticImage;
1153
1154 //      // // saving the torch::tensor

```



```

1155 //      // torch::save(acousticImage, \
1156 //      //      "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/Assets/acoustic_image.pt");
1157
1158
1159
1160 //      // // bringing it back to the original coordinates
1161 //      // scatterers->coordinates = scatterers->coordinates + this->location;
1162 //  }
1163
1164
1165
1166 // };
1167
1168
1169 template <typename T>
1170 class ULAClass
1171 {
1172 public:
1173     // intrinsic parameters
1174     int num_sensors; // number of sensors
1175     T inter_element_spacing; // space between sensors
1176     std::vector<std::vector<T>> coordinates; // coordinates of each sensor
1177     T sampling_frequency; // sampling frequency of the sensors
1178     T recording_period; // recording period of the ULA
1179     std::vector<T> location; // location of first coordinate
1180
1181     // derived
1182     std::vector<T> sensor_direction;
1183     std::vector<std::vector<T>> signalMatrix;
1184
1185     // decimation related
1186     int decimation_factor; // the new decimation factor
1187     T post_decimation_sampling_frequency; // the new sampling frequency
1188     std::vector<T> lowpass_filter_coefficients_for_decimation; // filter-coefficients for filtering
1189
1190     // imaging related
1191     T range_resolution; // theoretical range-resolution =  $\frac{c}{2B}$ 
1192     T azimuthal_resolution; // theoretical azimuth-resolution =  $\frac{\lambda}{(N-1) \cdot \text{inter-element-distance}}$ 
1193     T range_cell_size; // the range-cell quanta we're choosing for efficiency trade-off
1194     T azimuth_cell_size; // the azimuth quanta we're choosing
1195     std::vector<T> azimuth_centers; // tensor containing the azimuth centers
1196     std::vector<T> range_centers; // tensor containing the range-centers
1197     int frame_size; // the frame-size corresponding to a range cell in a decimated signal
1198     matrix
1199
1200     std::vector<std::vector<complex<T>>> mulFFTMMatrix; // the matrix containing the delays for each-element
1201     // as a slot
1202     std::vector<complex<T>> matchFilter; // torch tensor containing the match-filter
1203     int num_buffer_zeros_per_frame; // number of zeros we're adding per frame to ensure
1204     // no-rotation
1205     std::vector<std::vector<T>> beamformedImage; // the beamformed image
1206     std::vector<std::vector<T>> cartesianImage; // the cartesian version of beamformed image
1207
1208     // Artificial acoustic-image related
1209     std::vector<std::vector<T>> currentArtificialAcousticImage; // acoustic image directly produced
1210
1211     // Basic Constructor
1212     ULAClass() = default;
1213
1214     // constructor
1215     ULAClass(const int num_sensors_arg,
1216             const auto inter_element_spacing_arg,
1217             const auto& coordinates_arg,
1218             const auto& sampling_frequency_arg,
1219             const auto& recording_period_arg,
1220             const auto& location_arg,
1221             const auto& signalMatrix_arg,
1222             const auto& lowpass_filter_coefficients_for_decimation_arg):
1223         num_sensors(num_sensors_arg),
1224         inter_element_spacing(inter_element_spacing_arg),
1225         coordinates(std::move(coordinates_arg)),
1226         sampling_frequency(sampling_frequency_arg),
1227         recording_period(recording_period_arg),

```

```
1226         location(std::move(location_arg)),
1227         signalMatrix(std::move(signalMatrix_arg)),
1228         lowpass_filter_coefficients_for_decimation(std::move(lowpass_filter_coefficients_for_decimation_arg))
1229     {
1230
1231         // calculating ULA direction
1232         sensor_direction = std::vector<T>{coordinates[1][0] - coordinates[0][0],
1233                                           coordinates[1][1] - coordinates[0][1],
1234                                           coordinates[1][2] - coordinates[0][2]};
1235
1236         // normalizing
1237         auto norm_value_temp {std::inner_product(sensor_direction.begin(), sensor_direction.end(),
1238                                                  sensor_direction.begin(),
1239                                                  0.00)};
1240
1241         // dividing
1242         if (norm_value_temp != 0) {sensor_direction = sensor_direction / norm_value_temp;}
1243     }
1244
1245     // deleting copy constructor/assignment
1246     // ULAClass(const ULAClass& other)           = delete;
1247     // ULAClass& operator=(const ULAClass& other) = delete;
1248
1249
1250 };
```

---





```

129
130 // initializing all the ULAs
131 this->ULA_fls.init(      &this->transmitter_fls);
132 this->ULA_port.init(     &this->transmitter_port);
133 this->ULA_starboard.init( &this->transmitter_starboard);
134 if (DEBUGMODE_AUV) cout << "AUVClass::init: line 134" << endl;
135
136
137 // precomputing delay-matrices for the ULA-class
138 thread ULA_fls_precompute_weights_t(&ULAClass::nfdc_precomputeDelayMatrices, \
139                                     &this->ULA_fls, \
140                                     &this->transmitter_fls);
141 thread ULA_port_precompute_weights_t(&ULAClass::nfdc_precomputeDelayMatrices, \
142                                     &this->ULA_port, \
143                                     &this->transmitter_port);
144 thread ULA_starboard_precompute_weights_t(&ULAClass::nfdc_precomputeDelayMatrices, \
145                                             &this->ULA_starboard, \
146                                             &this->transmitter_starboard);
147 if (DEBUGMODE_AUV) cout << "AUVClass::init: line 145" << endl;
148
149 // joining the threads back
150 ULA_fls_precompute_weights_t.join();
151 ULA_port_precompute_weights_t.join();
152 ULA_starboard_precompute_weights_t.join();
153
154 }
155
156
157
158 /*=====
159 Aim: stepping motion
160 -----*/
161 void step(float timestep){
162
163     // updating location
164     this->location = this->location + this->velocity * timestep;
165     if(DEBUGMODE_AUV) cout<<"\t AUVClass: page 81 \n";
166
167     // updating attributes of members
168     this->syncComponentAttributes();
169     if(DEBUGMODE_AUV) cout<<"\t AUVClass: page 85 \n";
170 }
171
172
173
174 /*=====
175 Aim: updateAttributes
176 -----*/
177 void syncComponentAttributes(){
178
179     // updating ULA attributes
180     if(DEBUGMODE_AUV) cout<<"\t AUVClass: page 97 \n";
181
182     // updating locations
183     this->ULA_fls.location      = this->location;
184     this->ULA_port.location     = this->location;
185     this->ULA_starboard.location = this->location;
186
187     // updating the pointing direction of the ULAs
188     Tensor ula_fls_sensor_direction_spherical = \
189         fCart2Sph(this->pointing_direction); // spherical coords
190     ula_fls_sensor_direction_spherical[0] = \
191         ula_fls_sensor_direction_spherical[0] - 90;
192     Tensor ula_fls_sensor_direction_cart = \
193         fSph2Cart(ula_fls_sensor_direction_spherical);
194
195     this->ULA_fls.sensorDirection      = ula_fls_sensor_direction_cart; // assigning sensor directionf or
196                                     ULA-FLS
197     this->ULA_port.sensorDirection     = -this->pointing_direction; // assigning sensor direction for
198                                     ULA-Port
199     this->ULA_starboard.sensorDirection = -this->pointing_direction; // assigning sensor direction for
200                                     ULA-Starboard
201
202     // // calling the function to update the arguments
203     // this->ULA_fls.buildCoordinatesBasedOnLocation(); if(DEBUGMODE_AUV) cout<<"\t AUVClass: line 109

```

```

201     \n";
202     // this->ULA_port.buildCoordinatesBasedOnLocation(); if(DEBUGMODE_AUV) cout<<"\t AUVClass: line 111
203     \n";
204     // this->ULA_starboard.buildCoordinatesBasedOnLocation(); if(DEBUGMODE_AUV) cout<<"\t AUVClass: line
205     113 \n";
206
207     // updating transmitter locations
208     this->transmitter_fls.location = this->location;
209     this->transmitter_port.location = this->location;
210     this->transmitter_starboard.location = this->location;
211
212     // updating transmitter pointing directions
213     this->transmitter_fls.updatePointingAngle( this->pointing_direction);
214     this->transmitter_port.updatePointingAngle( this->pointing_direction);
215     this->transmitter_starboard.updatePointingAngle( this->pointing_direction);
216 }
217
218 /*=====
219 Aim: operator overriding for printing
220 -----*/
221 friend ostream& operator<<(ostream& os, AUVClass &auv){
222     os<<"\t location = "<<transpose(auv.location, 0, 1)<<endl;
223     os<<"\t velocity = "<<transpose(auv.velocity, 0, 1)<<endl;
224     return os;
225 }
226
227 /*=====
228 Aim: Subsetting Scatterers
229 -----*/
230 void subsetScatterers(ScattererClass* scatterers,\
231     TransmitterClass* transmitterObj,\
232     float tilt_angle){
233
234     // ensuring components are synced
235     this->syncComponentAttributes();
236     if(DEBUGMODE_AUV) cout<<"\t AUVClass: page 120 \n";
237
238     // calling the method associated with the transmitter
239     if(DEBUGMODE_AUV) {cout<<"\t\t scatterers.shape = "; fPrintTensorSize(scatterers->coordinates);}
240     if(DEBUGMODE_AUV) cout<<"\t\t tilt_angle = "<<tilt_angle<<endl;
241     transmitterObj->subsetScatterers(scatterers, tilt_angle);
242     if(DEBUGMODE_AUV) cout<<"\t AUVClass: page 124 \n";
243 }
244
245 // yaw-correction matrix
246 Tensor createYawCorrectionMatrix(Tensor pointing_direction_spherical, \
247     float target_azimuth_deg){
248
249     // building parameters
250     Tensor azimuth_correction = tensor({target_azimuth_deg}).to(DATATYPE).to(DEVICE) - \
251         pointing_direction_spherical[0];
252     Tensor azimuth_correction_radians = azimuth_correction * PI / 180;
253
254     Tensor yawCorrectionMatrix = \
255         tensor({cos(azimuth_correction_radians).item<float>(), \
256             cos(tensor({90}).to(DATATYPE).to(DEVICE)*PI/180 +
257                 azimuth_correction_radians).item<float>(), \
258             (float)0, \
259             sin(azimuth_correction_radians).item<float>(), \
260             sin(tensor({90}).to(DATATYPE).to(DEVICE)*PI/180 +
261                 azimuth_correction_radians).item<float>(), \
262             (float)0, \
263             (float)0, \
264             (float)1}).reshape({3,3}).to(DATATYPE).to(DEVICE);
265
266     // returning the matrix
267     return yawCorrectionMatrix;
268 }
269
270 // pitch-correction matrix
271 Tensor createPitchCorrectionMatrix(Tensor pointing_direction_spherical, \

```

```

271         float target_elevation_deg){
272
273     // building parameters
274     Tensor elevation_correction      = tensor({target_elevation_deg}).to(DATATYPE).to(DEVICE) - \
275         pointing_direction_spherical[1];
276     Tensor elevation_correction_radians = elevation_correction * PI / 180;
277
278     // creating the matrix
279     Tensor pitchCorrectionMatrix = \
280         tensor({(float)1, \
281             (float)0, \
282             (float)0, \
283             (float)0, \
284             cos(elevation_correction_radians).item<float>(), \
285             cos(tensor({90}).to(DATATYPE).to(DEVICE)*PI/180 + \
286                 elevation_correction_radians).item<float>()), \
287             (float)0, \
288             sin(elevation_correction_radians).item<float>(), \
289             sin(tensor({90}).to(DATATYPE).to(DEVICE)*PI/180 + \
290                 elevation_correction_radians).item<float>())}).reshape({3,3}).to(DATATYPE);
291
292     // returning the matrix
293     return pitchCorrectionMatrix;
294 }
295
296 // Signal Simulation
297 void simulateSignal(ScattererClass& scatterer){
298
299     // printing status
300     cout << "\t AUVClass::simulateSignal: Began Signal Simulation" << endl;
301
302     // making three copies
303     ScattererClass scatterer_fls      = scatterer;
304     ScattererClass scatterer_port     = scatterer;
305     ScattererClass scatterer_starboard = scatterer;
306
307     // finding the pointing direction in spherical
308     Tensor auv_pointing_direction_spherical = fCart2Sph(this->pointing_direction);
309
310     // asking the transmitters to subset the scatterers by multithreading
311     thread transmitterFLSSubset_t(&AUVClass::subsetScatterers, this, \
312         &scatterer_fls, \
313         &this->transmitter_fls, \
314         (float)0);
315     thread transmitterPortSubset_t(&AUVClass::subsetScatterers, this, \
316         &scatterer_port, \
317         &this->transmitter_port, \
318         auv_pointing_direction_spherical[1].item<float>());
319     thread transmitterStarboardSubset_t(&AUVClass::subsetScatterers, this, \
320         &scatterer_starboard, \
321         &this->transmitter_starboard, \
322         - auv_pointing_direction_spherical[1].item<float>());
323
324     // joining the subset threads back
325     transmitterFLSSubset_t.join();
326     transmitterPortSubset_t.join();
327     transmitterStarboardSubset_t.join();
328
329     // multithreading the saving tensors part.
330     thread savetensor_t(fSaveSeafloorScatterers, \
331         scatterer, \
332         scatterer_fls, \
333         scatterer_port, \
334         scatterer_starboard);
335
336     // asking ULAs to simulate signal through multithreading
337     thread ulaflds_signalsim_t(&ULAClass::nfdc_simulateSignal, \
338         &this->ULA_fls, \
339         &scatterer_fls, \
340         &this->transmitter_fls);
341     thread ulaport_signalsim_t(&ULAClass::nfdc_simulateSignal, \
342         &this->ULA_port, \
343         &scatterer_port, \

```

```

344         &this->transmitter_port);
345     thread ulastarboard_signalsim_t(&ULAClass::nfdc_simulateSignal, \
346                                     &this->ULA_starboard, \
347                                     &scatterer_starboard, \
348                                     &this->transmitter_starboard);
349
350     // joining them back
351     ulafpls_signalsim_t.join(); // joining back the thread for ULA-FLS
352     ulaport_signalsim_t.join(); // joining back the signals-sim thread for ULA-Port
353     ulastarboard_signalsim_t.join(); // joining back the signal-sim thread for ULA-Starboard
354     savetensor_t.join(); // joining back the signal-sim thread for tensor-saving
355
356
357 }
358
359 // Imaging Function
360 /* =====
361 ----- */
362 void image(){
363
364     // asking ULAs to decimate the signals obtained at each time step
365     thread ULA_fls_image_t(&ULAClass::nfdc_decimateSignal, \
366                             &this->ULA_fls, \
367                             &this->transmitter_fls);
368     thread ULA_port_image_t(&ULAClass::nfdc_decimateSignal, \
369                             &this->ULA_port, \
370                             &this->transmitter_port);
371     thread ULA_starboard_image_t(&ULAClass::nfdc_decimateSignal, \
372                                  &this->ULA_starboard, \
373                                  &this->transmitter_starboard);
374
375     // joining the threads back
376     ULA_fls_image_t.join();
377     ULA_port_image_t.join();
378     ULA_starboard_image_t.join();
379
380     // saving the decimated signal
381     if (SAVE_DECIMATED_SIGNAL_MATRIX) {
382         cout << "\t AUVClass::image: saving decimated signal matrix" \
383              << endl;
384         save(this->ULA_fls.signalMatrix, \
385             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/decimated_signalMatrix_fls.pt");
386         save(this->ULA_port.signalMatrix, \
387             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/decimated_signalMatrix_port.pt");
388         save(this->ULA_starboard.signalMatrix, \
389             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/decimated_signalMatrix_starboard.pt");
390     }
391
392     // asking ULAs to match-filter the signals
393     thread ULA_fls_matchfilter_t( \
394         &ULAClass::nfdc_matchFilterDecimatedSignal, \
395         &this->ULA_fls);
396     thread ULA_port_matchfilter_t( \
397         &ULAClass::nfdc_matchFilterDecimatedSignal, \
398         &this->ULA_port);
399     thread ULA_starboard_matchfilter_t( \
400         &ULAClass::nfdc_matchFilterDecimatedSignal, \
401         &this->ULA_starboard);
402
403     // joining the threads back
404     ULA_fls_matchfilter_t.join();
405     ULA_port_matchfilter_t.join();
406     ULA_starboard_matchfilter_t.join();
407
408
409     // saving the decimated signal
410     if (SAVE_MATCHFILTERED_SIGNAL_MATRIX) {
411
412         // saving the tensors
413         cout << "\t AUVClass::image: saving match-filtered signal matrix" \
414              << endl;
415         save(this->ULA_fls.signalMatrix, \
416             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/matchfiltered_signalMatrix_fls.pt");
417         save(this->ULA_port.signalMatrix, \
418             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/matchfiltered_signalMatrix_port.pt");
419     }

```



```

419     save(this->ULA_starboard.signalMatrix, \
420          "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/matchfiltered_signalMatrix_starboard.pt");
421
422     // running python-script
423
424 }
425
426
427 // performing the beamforming
428 thread ULA_fls_beamforming_t(&ULAClass::nfdc_beamforming, \
429                             &this->ULA_fls, \
430                             &this->transmitter_fls);
431 thread ULA_port_beamforming_t(&ULAClass::nfdc_beamforming, \
432                             &this->ULA_port, \
433                             &this->transmitter_port);
434 thread ULA_starboard_beamforming_t(&ULAClass::nfdc_beamforming, \
435                                   &this->ULA_starboard, \
436                                   &this->transmitter_starboard);
437
438 // joining the filters back
439 ULA_fls_beamforming_t.join();
440 ULA_port_beamforming_t.join();
441 ULA_starboard_beamforming_t.join();
442
443
444 }
445
446
447
448
449 /* =====
450 Aim: directly create acoustic image
451 ----- */
452 void createAcousticImage(ScattererClass* scatterers){
453
454     // making three copies
455     ScattererClass scatterer_fls    = scatterers;
456     ScattererClass scatterer_port   = scatterers;
457     ScattererClass scatterer_starboard = scatterers;
458
459     // printing size of scatterers before subsetting
460     PRINTSMALLLINE
461     cout<< "\t > AUVClass::createAcousticImage: Beginning Scatterer Subsetting"<<endl;
462     cout<< "\t AUVClass::createAcousticImage: scatterer_fls.coordinates.shape (before) = ";
463         fPrintTensorSize(scatterer_fls.coordinates);
464     cout<< "\t AUVClass::createAcousticImage: scatterer_port.coordinates.shape (before) = ";
465         fPrintTensorSize(scatterer_port.coordinates);
466     cout<< "\t AUVClass::createAcousticImage: scatterer_starboard.coordinates.shape (before) = ";
467         fPrintTensorSize(scatterer_starboard.coordinates);
468
469     // finding the pointing direction in spherical
470     Tensor auv_pointing_direction_spherical = fCart2Sph(this->pointing_direction);
471
472     // asking the transmitters to subset the scatterers by multithreading
473     thread transmitterFLSSubset_t(&AUVClass::subsetScatterers, this, \
474                                 &scatterer_fls, \
475                                 &this->transmitter_fls, \
476                                 (float)0);
477     thread transmitterPortSubset_t(&AUVClass::subsetScatterers, this, \
478                                   &scatterer_port, \
479                                   &this->transmitter_port, \
480                                   auv_pointing_direction_spherical[1].item<float>());
481     thread transmitterStarboardSubset_t(&AUVClass::subsetScatterers, this, \
482                                         &scatterer_starboard, \
483                                         &this->transmitter_starboard, \
484                                         - auv_pointing_direction_spherical[1].item<float>());
485
486     // joining the subset threads back
487     transmitterFLSSubset_t.join( );
488     transmitterPortSubset_t.join( );
489     transmitterStarboardSubset_t.join( );
490
491     // asking the ULAs to directly create acoustic images
492     thread ULA_fls_acoustic_image_t(&ULAClass::nfdc_createAcousticImage, this->ULA_fls, \

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```

491         &scatterer_fls, &this->transmitter_fls);
492     thread ULA_port_acoustic_image_t(&ULAClass::nfdc_createAcousticImage, &this->ULA_port, \
493         &scatterer_port, &this->transmitter_port);
494     thread ULA_starboard_acoustic_image_t(&ULAClass::nfdc_createAcousticImage, &this->ULA_starboard, \
495         &scatterer_starboard, &this->transmitter_starboard);
496
497     // joining the threads back
498     ULA_fls_acoustic_image_t.join( );
499     ULA_port_acoustic_image_t.join( );
500     ULA_starboard_acoustic_image_t.join();
501
502 }
503
504 };
505
506
507
508
509
510
511
512
513
514
515
516
517
518
519
520
521
522
523
524
525
526
527
528 // 0.0000,
529 // 0.0000,
530 // 0.0000,
531 // 0.0000,
532 // 0.0000,
533 // 0.0000,
534 // 0.0000,
535 // 0.0000,
536 // 0.0000,
537 // 0.0000,
538 // 0.0000,
539 // 0.0000,
540 // 0.0000,
541 // 0.0000,
542 // 0.0000,
543 // 0.0000,
544 // 0.0000,
545 // 0.0000,
546 // 0.0000,
547 // 0.0000,
548 // 0.0000,
549 // 0.0000,
550 // 0.0000,
551 // 0.0000,
552 // 0.0000,
553 // 0.0000,
554 // 0.0000,
555 // 0.0000,
556 // 0.0000,
557 // 0.0000,
558 // 0.0000,
559 // 0.0001,
560 // 0.0001,
561 // 0.0002,
562 // 0.0003,
563 // 0.0006,
564 // 0.0009,
565 // 0.0014,

```

```
566 // 0.0022, 0.0032, 0.0047, 0.0066, 0.0092, 0.0126, 0.0168, 0.0219, 0.0281, 0.0352, 0.0432, 0.0518, 0.0609,  
      0.0700, 0.0786, 0.0861, 0.0921, 0.0958, 0.0969, 0.0950, 0.0903, 0.0833, 0.0755, 0.0694, 0.0693, 0.0825,  
      0.1206
```

---

## 8.2 Setup Scripts

### 8.2.1 Seafloor Setup

Following is the script to be run to setup the seafloor.

---

```

1  /* =====
2  Aim: Setup sea floor
3  =====*/
4
5  // including headerfiles
6  #include <torch/torch.h>
7  #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/ScattererClass.h"
8
9  // including functions
10 #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Functions/fCreateHills.cpp"
11 #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Functions/fCreateBoxes.cpp"
12
13 #ifndef DEVICE
14     #define DEVICE          kCPU
15     // #define DEVICE      kMPS
16     // #define DEVICE      kCUDA
17 #endif
18
19 // adding terrain features
20 #define BOXES                false
21 #define HILLS                true
22 #define DEBUG_SEAFLOOR      false
23 #define SAVETENSORS_Seafloor false
24 #define PLOT_SEAFLOOR        false
25
26 // functin that setups the sea-floor
27 void SeafloorSetup(ScattererClass& scatterers) {
28
29     // sea-floor bounds
30     int bed_width = 100; // width of the bed (x-dimension)
31     int bed_length = 100; // length of the bed (y-dimension)
32
33     // creating some tensors to pass. This is put outside to maintain scope
34     Tensor box_coordinates = zeros({3,1}).to(DATATYPE).to(DEVICE);
35     Tensor box_reflectivity = zeros({1,1}).to(DATATYPE).to(DEVICE);
36
37     // creating boxes
38     if (BOXES)
39         fCreateBoxes(bed_width, \
40                     bed_length, \
41                     box_coordinates, \
42                     box_reflectivity);
43
44     // scatter-intensity
45     // int bed_width_density = 100; // density of points along x-dimension
46     // int bed_length_density = 100; // density of points along y-dimension
47     int bed_width_density = 10; // density of points along x-dimension
48     int bed_length_density = 10; // density of points along y-dimension
49
50     // setting up coordinates
51     auto xpoints = linspace(0, \
52                             bed_width, \
53                             bed_width * bed_width_density).to(DEVICE);
54     auto ypoints = linspace(0, \
55                             bed_length, \
56                             bed_length * bed_length_density).to(DEVICE);
57
58     // creating mesh
59     auto mesh_grid = meshgrid({xpoints, ypoints}, "ij");
60     auto X = mesh_grid[0];
61     auto Y = mesh_grid[1];
62     X = reshape(X, {1, X.numel()});
63     Y = reshape(Y, {1, Y.numel()});
64
65     // creating heights of scattereres
66     if(HILLS == true){

```

```

67
68 // setting up hill parameters
69 int num_hills = 10;
70
71 // setting up placement of hills
72 Tensor points2D = cat({X, Y}, 0);
73 Tensor min2D = std::get<0>(min(points2D, 1, true));
74 Tensor max2D = std::get<0>(max(points2D, 1, true));
75 Tensor hill_means = \
76     min2D \
77     + mul(rand({2, num_hills}), \
78         max2D - min2D);
79
80 // setting up hill dimensions
81 Tensor hill_dimensions_min = \
82     tensor({5, \
83         5, \
84         2}).reshape({3,1});
85 Tensor hill_dimensions_max = \
86     tensor({30, \
87         30, \
88         10}).reshape({3,1});
89 Tensor hill_dimensions = \
90     hill_dimensions_min + \
91     mul(hill_dimensions_max - hill_dimensions_min, \
92         rand({3, num_hills}));
93
94 // calling the hill-creation function
95 fCreateHills(hill_means, \
96     hill_dimensions, \
97     points2D);
98
99 // setting up floor reflectivity
100 Tensor floorScatter_reflectivity = \
101     ones({1, Y.numel()}).to(DEVICE);
102
103 // populating the values of the incoming argument.
104 scatterers.coordinates = points2D; // assigning coordinates
105 scatterers.reflectivity = floorScatter_reflectivity; // assigning reflectivity
106 }
107 else{
108
109 // assigning flat heights
110 Tensor Z = zeros({1, Y.numel()}).to(DEVICE);
111
112 // setting up floor coordinates
113 Tensor floorScatter_coordinates = cat({X, Y, Z}, 0);
114 Tensor floorScatter_reflectivity = ones({1, Y.numel()}).to(DEVICE);
115
116 // populating the values of the incoming argument.
117 scatterers.coordinates = floorScatter_coordinates; // assigning coordinates
118 scatterers.reflectivity = floorScatter_reflectivity; // assigning reflectivity
119 }
120
121 // combining the values
122 if(DEBUG_SEAFLOOR) std::cout<<"\t SeafloorSetup: line 166 \n";
123 if(DEBUG_SEAFLOOR) {std::cout<<"\t scatterers.coordinates.shape = ";
124     fPrintTensorSize(scatterers.coordinates);}
125 if(DEBUG_SEAFLOOR) {std::cout<<"\t box_coordinates.shape = "; fPrintTensorSize(box_coordinates);}
126 if(DEBUG_SEAFLOOR) {std::cout<<"\t scatterers.reflectivity.shape = ";
127     fPrintTensorSize(scatterers.reflectivity);}
128 if(DEBUG_SEAFLOOR) {std::cout<<"\t box_reflectivity = "; fPrintTensorSize(box_reflectivity);}
129
130 // assigning values to the coordinates
131 scatterers.coordinates = cat({scatterers.coordinates, box_coordinates}, 1);
132 scatterers.reflectivity = cat({scatterers.reflectivity, box_reflectivity}, 1);
133
134 // saving tensors
135 if(SAVETENSORS_Seafloor){
136     save(scatterers.coordinates, \
137         "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/SeafloorScatter_gt.pt");
138     std::cout<<"SeafloorSetup: Saved Seafloor "<<std::endl;
139 }

```



## 8.2.2 Transmitter Setup

Following is the script to be run to setup the transmitter.

---

```

1  /* =====
2  Aim: Setup sea floor
3  =====*/
4  // #include <torch/torch.h>
5  // #include <cmath>
6
7  #ifndef DEVICE
8      // #define DEVICE      torch::kMPS
9      #define DEVICE      torch::kCPU
10 #endif
11
12
13
14 // function to calibrate the transmitters
15 void TransmitterSetup(TransmitterClass& transmitter_fls,
16                      TransmitterClass& transmitter_port,
17                      TransmitterClass& transmitter_starboard) {
18
19     // Setting up transmitter
20     float sampling_frequency = 160e3;           // sampling frequency
21     float f1                 = 50e3;           // first frequency of LFM
22     float f2                 = 70e3;           // second frequency of LFM
23     float fc                 = (f1 + f2)/2;     // finding center-frequency
24     float bandwidth          = std::abs(f2 - f1); // bandwidth
25     float pulselength        = 5e-2;           // time of recording
26
27     // building LFM
28     torch::Tensor timearray = torch::linspace(0, \
29                                             pulselength, \
30                                             floor(pulselength * sampling_frequency)).to(DEVICE);
31     float K                  = (f2 - f1)/pulselength; // calculating frequency-slope
32     torch::Tensor Signal     = K * timearray;         // frequency at each time-step, with f1 = 0
33     Signal                   = torch::mul(2*PI*(f1 + Signal), \
34                                     timearray);      // creating
35     Signal                   = cos(Signal);           // calculating signal
36
37
38     // Setting up transmitter
39     torch::Tensor location   = torch::zeros({3,1}).to(DEVICE); // location of transmitter
40     float azimuthal_angle_fls = 0; // initial pointing direction
41     float azimuthal_angle_port = 90; // initial pointing direction
42     float azimuthal_angle_starboard = -90; // initial pointing direction
43
44     float elevation_angle    = -60; // initial pointing direction
45
46     float azimuthal_beamwidth_fls = 20; // azimuthal beamwidth of the signal cone
47     float azimuthal_beamwidth_port = 20; // azimuthal beamwidth of the signal cone
48     float azimuthal_beamwidth_starboard = 20; // azimuthal beamwidth of the signal cone
49
50     float elevation_beamwidth_fls = 20; // elevation beamwidth of the signal cone
51     float elevation_beamwidth_port = 20; // elevation beamwidth of the signal cone
52     float elevation_beamwidth_starboard = 20; // elevation beamwidth of the signal cone
53
54     int azimuthQuantDensity = 10; // number of points, a degree is split into quantization density
55                                     along azimuth (used for shadowing)
56     int elevationQuantDensity = 10; // number of points, a degree is split into quantization density
57                                     along elevation (used for shadowing)
58     float rangeQuantSize = 10; // the length of a cell (used for shadowing)
59
60     float azimuthShadowThreshold = 1; // azimuth threshold (in degrees)
61     float elevationShadowThreshold = 1; // elevation threshold (in degrees)
62
63     // transmitter-fls
64     transmitter_fls.location = location; // Assigning location
65     transmitter_fls.Signal   = Signal;   // Assigning signal
66     transmitter_fls.azimuthal_angle = azimuthal_angle_fls; // assigning azimuth angle
67     transmitter_fls.elevation_angle = elevation_angle; // assigning elevation angle
68     transmitter_fls.azimuthal_beamwidth = azimuthal_beamwidth_fls; // assigning azimuth-beamwidth

```

```

69 transmitter_fls.elevation_beamwidth = elevation_beamwidth_fls; // assigning elevation-beamwidth
70 // updating quantization densities
71 transmitter_fls.azimuthQuantDensity = azimuthQuantDensity; // assigning azimuth quant density
72 transmitter_fls.elevationQuantDensity = elevationQuantDensity; // assigning elevation quant density
73 transmitter_fls.rangeQuantSize = rangeQuantSize; // assigning range-quantization
74 transmitter_fls.azimuthShadowThreshold = azimuthShadowThreshold; // azimuth-threshold in shadowing
75 transmitter_fls.elevationShadowThreshold = elevationShadowThreshold; // elevation-threshold in shadowing
76 // signal related
77 transmitter_fls.f_low = f1; // assigning lower frequency
78 transmitter_fls.f_high = f2; // assigning higher frequency
79 transmitter_fls.fc = fc; // assigning center frequency
80 transmitter_fls.bandwidth = bandwidth; // assigning bandwidth
81
82
83
84 // transmitter-portside
85 transmitter_port.location = location; // Assigning location
86 transmitter_port.Signal = Signal; // Assigning signal
87 transmitter_port.azimuthal_angle = azimuthal_angle_port; // assigning azimuth angle
88 transmitter_port.elevation_angle = elevation_angle; // assigning elevation angle
89 transmitter_port.azimuthal_beamwidth = azimuthal_beamwidth_port; // assigning azimuth-beamwidth
90 transmitter_port.elevation_beamwidth = elevation_beamwidth_port; // assigning elevation-beamwidth
91 // updating quantization densities
92 transmitter_port.azimuthQuantDensity = azimuthQuantDensity; // assigning azimuth quant density
93 transmitter_port.elevationQuantDensity = elevationQuantDensity; // assigning elevation quant density
94 transmitter_port.rangeQuantSize = rangeQuantSize; // assigning range-quantization
95 transmitter_port.azimuthShadowThreshold = azimuthShadowThreshold; // azimuth-threshold in shadowing
96 transmitter_port.elevationShadowThreshold = elevationShadowThreshold; // elevation-threshold in shadowing
97 // signal related
98 transmitter_port.f_low = f1; // assigning lower frequency
99 transmitter_port.f_high = f2; // assigning higher frequency
100 transmitter_port.fc = fc; // assigning center frequency
101 transmitter_port.bandwidth = bandwidth; // assigning bandwidth
102
103
104
105 // transmitter-starboard
106 transmitter_starboard.location = location; // assigning location
107 transmitter_starboard.Signal = Signal; // assigning signal
108 transmitter_starboard.azimuthal_angle = azimuthal_angle_starboard; // assigning azimuthal signal
109 transmitter_starboard.elevation_angle = elevation_angle;
110 transmitter_starboard.azimuthal_beamwidth = azimuthal_beamwidth_starboard;
111 transmitter_starboard.elevation_beamwidth = elevation_beamwidth_starboard;
112 // updating quantization densities
113 transmitter_starboard.azimuthQuantDensity = azimuthQuantDensity;
114 transmitter_starboard.elevationQuantDensity = elevationQuantDensity;
115 transmitter_starboard.rangeQuantSize = rangeQuantSize;
116 transmitter_starboard.azimuthShadowThreshold = azimuthShadowThreshold;
117 transmitter_starboard.elevationShadowThreshold = elevationShadowThreshold;
118 // signal related
119 transmitter_starboard.f_low = f1; // assigning lower frequency
120 transmitter_starboard.f_high = f2; // assigning higher frequency
121 transmitter_starboard.fc = fc; // assigning center frequency
122 transmitter_starboard.bandwidth = bandwidth; // assigning bandwidth
123
124 }

```

---



### 8.2.3 Uniform Linear Array

Following is the script to be run to setup the uniform linear array.

---

```

1  /* =====
2  Aim: Setup sea floor
3  NOAA: 50 to 100 KHz is the transmission frequency
4  we'll create our LFM with 50 to 70KHz
5  =====*/
6
7
8  // Choosing device
9  #ifndef DEVICE
10     // #define DEVICE      kMPS
11     #define DEVICE      kCPU
12 #endif
13
14
15 // the coefficients for the low-pass filter.
16 #define LOWPASS_DECIMATE_FILTER_COEFFICIENTS 0.0000, 0.0000, 0.0000, 0.0000, 0.0000, 0.0000, 0.0001, 0.0003,
    0.0006, 0.0015, 0.0030, 0.0057, 0.0100, 0.0163, 0.0251, 0.0364, 0.0501, 0.0654, 0.0814, 0.0966, 0.1093,
    0.1180, 0.1212, 0.1179, 0.1078, 0.0914, 0.0699, 0.0451, 0.0192, -0.0053, -0.0262, -0.0416, -0.0504,
    -0.0522, -0.0475, -0.0375, -0.0239, -0.0088, 0.0057, 0.0179, 0.0263, 0.0303, 0.0298, 0.0253, 0.0177,
    0.0086, -0.0008, -0.0091, -0.0153, -0.0187, -0.0191, -0.0168, -0.0123, -0.0065, -0.0004, 0.0052, 0.0095,
    0.0119, 0.0125, 0.0112, 0.0084, 0.0046, 0.0006, -0.0031, -0.0060, -0.0078, -0.0082, -0.0075, -0.0057,
    -0.0033, -0.0006, 0.0019, 0.0039, 0.0051, 0.0055, 0.0050, 0.0039, 0.0023, 0.0005, -0.0012, -0.0025,
    -0.0034, -0.0036, -0.0034, -0.0026, -0.0016, -0.0004, 0.0007, 0.0016, 0.0022, 0.0024, 0.0023, 0.0018,
    0.0011, 0.0003, -0.0004, -0.0011, -0.0015, -0.0016, -0.0015
17
18
19
20 void ULASetup(ULAClass& ula_fls,
21               ULAClass& ula_port,
22               ULAClass& ula_starboard) {
23
24     // setting up ula
25     int num_sensors      = 64;                // number of sensors
26     float sampling_frequency = 160e3;         // sampling frequency
27     float inter_element_spacing = 1500/(2*sampling_frequency); // space between samples
28     float recording_period   = 10e-2;         // sampling-period
29
30
31     // building the direction for the sensors
32     Tensor ULA_direction = tensor({-1,0,0}).reshape({3,1}).to(DATATYPE).to(DEVICE);
33     ULA_direction        = ULA_direction/linalg_norm(ULA_direction, 2, 0, true, DATATYPE).to(DEVICE);
34     ULA_direction        = ULA_direction * inter_element_spacing;
35
36
37     // building the coordinates for the sensors
38     Tensor ULA_coordinates = mul(linspace(0, num_sensors-1, num_sensors).to(DEVICE), \
39                                  ULA_direction);
40
41     // the coefficients for the decimation filter
42     Tensor lowpassfiltercoefficients = tensor({LOWPASS_DECIMATE_FILTER_COEFFICIENTS}).to(DATATYPE);
43
44
45     // assigning values
46     ula_fls.num_sensors      = num_sensors;    // assigning number of sensors
47     ula_fls.inter_element_spacing = inter_element_spacing; // assigning inter-element spacing
48     ula_fls.coordinates      = ULA_coordinates; // assigning ULA coordinates
49     ula_fls.sampling_frequency = sampling_frequency; // assigning sampling frequencys
50     ula_fls.recording_period  = recording_period; // assigning recording period
51     ula_fls.sensorDirection   = ULA_direction; // ULA direction
52     ula_fls.lowpassFilterCoefficientsForDecimation = lowpassfiltercoefficients;
53
54
55     // assigning values
56     ula_port.num_sensors      = num_sensors;    // assigning number of sensors
57     ula_port.inter_element_spacing = inter_element_spacing; // assigning inter-element spacing
58     ula_port.coordinates      = ULA_coordinates; // assigning ULA coordinates
59     ula_port.sampling_frequency = sampling_frequency; // assigning sampling frequencys
60     ula_port.recording_period  = recording_period; // assigning recording period
61     ula_port.sensorDirection   = ULA_direction; // ULA direction
62     ula_port.lowpassFilterCoefficientsForDecimation = lowpassfiltercoefficients;

```

```
63
64
65 // assigning values
66 ula_starboard.num_sensors      = num_sensors;           // assigning number of sensors
67 ula_starboard.inter_element_spacing = inter_element_spacing; // assigning inter-element spacing
68 ula_starboard.coordinates      = ULA_coordinates;       // assigning ULA coordinates
69 ula_starboard.sampling_frequency = sampling_frequency;  // assigning sampling frequency
70 ula_starboard.recording_period  = recording_period;     // assigning recording period
71 ula_starboard.sensorDirection  = ULA_direction;        // ULA direction
72 ula_starboard.lowpassFilterCoefficientsForDecimation = lowpassfiltercoefficients;
73
74 }
```

---

## 8.2.4 AUV Setup

Following is the script to be run to setup the vessel.

---

```

1  /* =====
2  Aim: Setup sea floor
3  NOAA: 50 to 100 KHz is the transmission frequency
4  we'll create our LFM with 50 to 70KHz
5  =====/
6
7  #ifndef DEVICE
8      #define DEVICE      torch::kMPS
9      // #define DEVICE    torch::kCPU
10 #endif
11
12 // =====
13 void AUVSetup(AUVClass* auv) {
14
15     // building properties for the auv
16     torch::Tensor location      = torch::tensor({0,50,30}).reshape({3,1}).to(DATATYPE).to(DEVICE); //
17         starting location of AUV
18     torch::Tensor velocity      = torch::tensor({5,0, 0}).reshape({3,1}).to(DATATYPE).to(DEVICE); //
19         starting velocity of AUV
20     torch::Tensor pointing_direction = torch::tensor({1,0, 0}).reshape({3,1}).to(DATATYPE).to(DEVICE); //
21         pointing direction of AUV
22
23     // assigning
24     auv->location      = location;          // assigning location of auv
25     auv->velocity      = velocity;          // assigning vector representing velocity
26     auv->pointing_direction = pointing_direction; // assigning pointing direction of auv
27 }

```

---

## 8.3 Function Definitions

### 8.3.1 Cartesian Coordinates to Spherical Coordinates

---

```

1  /* =====
2  Aim: Setup sea floor
3  =====*/
4  #include <torch/torch.h>
5  #include <iostream>
6
7  // hash-defines
8  #define PI          3.14159265
9  #define DEBUG_Cart2Sph false
10
11 #ifndef DEVICE
12     #define DEVICE      torch::kMPS
13     // #define DEVICE    torch::kCPU
14 #endif
15
16
17 // bringing in functions
18 #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Functions/fPrintTensorSize.cpp"
19
20 #pragma once
21
22 torch::Tensor fCart2Sph(torch::Tensor cartesian_vector){
23
24     // sending argument to the device
25     cartesian_vector = cartesian_vector.to(DEVICE);
26     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 26 \n";
27
28     // splatting the point onto xy plane
29     torch::Tensor xysplat = cartesian_vector.clone().to(DEVICE);
30     xysplat[2] = 0;
31     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 31 \n";
32
33     // finding splat lengths
34     // torch::Tensor xysplat_lengths = torch::linalg_norm(xysplat, 2, 0, true, DATATYPE).to(DEVICE);
35     torch::Tensor xysplat_lengths = torch::linalg_norm(xysplat, 2, 0, true, torch::kFloat).to(DATATYPE);
36     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 35 \n";
37
38     // finding azimuthal and elevation angles
39     torch::Tensor azimuthal_angles = torch::atan2(xysplat[1], xysplat[0]).to(DEVICE) * 180/PI;
40     azimuthal_angles = azimuthal_angles.reshape({1, azimuthal_angles.numel()});
41     torch::Tensor elevation_angles = torch::atan2(cartesian_vector[2], xysplat_lengths).to(DEVICE) * 180/PI;
42     // torch::Tensor rho_values = torch::linalg_norm(cartesian_vector, 2, 0, true, DATATYPE).to(DEVICE);
43     torch::Tensor rho_values = torch::linalg_norm(cartesian_vector, \
44                                                  2, 0, true, torch::kFloat).to(DATATYPE);
45     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 42 \n";
46
47
48     // printing values for debugging
49     if (DEBUG_Cart2Sph){
50         std::cout<<"azimuthal_angles.shape = "; fPrintTensorSize(azimuthal_angles);
51         std::cout<<"elevation_angles.shape = "; fPrintTensorSize(elevation_angles);
52         std::cout<<"rho_values.shape = "; fPrintTensorSize(rho_values);
53     }
54     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 51 \n";
55
56     // creating tensor to send back
57     torch::Tensor spherical_vector = torch::cat({azimuthal_angles, \
58                                                  elevation_angles, \
59                                                  rho_values}, 0).to(DEVICE);
60     if (DEBUG_Cart2Sph) std::cout<<"\t fCart2Sph: line 57 \n";
61
62     // returning the value
63     return spherical_vector;
64 }

```

---

### 8.3.2 Spherical Coordinates to Cartesian Coordinates

---

```

1  /* =====
2  Aim: Setup sea floor
3  =====*/
4  #include <torch/torch.h>
5
6  #pragma once
7
8  // hash-defines
9  #define PI          3.14159265
10 #define MYDEBUGFLAG false
11
12 #ifndef DEVICE
13     // #define DEVICE      torch::kMPS
14     #define DEVICE      torch::kCPU
15 #endif
16
17 torch::Tensor fSph2Cart(torch::Tensor spherical_vector){
18
19
20
21
22     // sending argument to device
23     spherical_vector = spherical_vector.to(DEVICE);
24
25     // creating cartesian vector
26     torch::Tensor cartesian_vector =
27         torch::zeros({3,(int)(spherical_vector.numel()/3)}).to(DATATYPE).to(DEVICE);
28
29     // populating it
30     cartesian_vector[0] = spherical_vector[2] * \
31         torch::cos(spherical_vector[1] * PI/180) * \
32         torch::cos(spherical_vector[0] * PI/180);
33     cartesian_vector[1] = spherical_vector[2] * \
34         torch::cos(spherical_vector[1] * PI/180) * \
35         torch::sin(spherical_vector[0] * PI/180);
36     cartesian_vector[2] = spherical_vector[2] * \
37         torch::sin(spherical_vector[1] * PI/180);
38
39     // returning the value
40     return cartesian_vector;
41 }

```

---

### 8.3.3 Column-Wise Convolution

---

```

1  /* =====
2  Aim: Convolving the columns of two input matrices
3  =====*/
4  #include <ratio>
5  #include <stdexcept>
6  #include <torch/torch.h>
7
8  #pragma once
9
10 // hash-defines
11 #define PI          3.14159265
12 #define MYDEBUGFLAG false
13
14 #ifndef DEVICE
15     // #define DEVICE      torch::kMPS
16     #define DEVICE      torch::kCPU
17 #endif
18
19
20 void fConvolveColumns(torch::Tensor& inputMatrix, \
21                     torch::Tensor& kernelMatrix){
22
23

```

```

24 // printing shape
25 if(MYDEBUGFLAG) std::cout<<"inputMatrix.shape =
    [<<inputMatrix.size(0)<<","<<inputMatrix.size(1)<<std::endl;
26 if(MYDEBUGFLAG) std::cout<<"kernelMatrix.shape =
    [<<kernelMatrix.size(0)<<","<<kernelMatrix.size(1)<<std::endl;
27
28 // ensuring the two have the same number of columns
29 if (inputMatrix.size(1) != kernelMatrix.size(1)){
30     throw std::runtime_error("fConvolveColumns: arguments cannot have different number of columns");
31 }
32
33
34 // calculating length of final result
35 int final_length = inputMatrix.size(0) + kernelMatrix.size(0) - 1; if(MYDEBUGFLAG) std::cout<<"\t\t\t
    fConvolveColumns: 27"<<std::endl;
36
37 // converting the two arguments to float since fft doesn't work with halves
38 inputMatrix = inputMatrix.to(torch::kFloat);
39 kernelMatrix = kernelMatrix.to(torch::kFloat);
40
41 // calculating FFT of the two matrices
42 torch::Tensor inputMatrix_FFT = torch::fft::fftn(inputMatrix, \
43     {final_length}, \
44     {0}); if(MYDEBUGFLAG) std::cout<<"\t\t\t fConvolveColumns:
        32"<<std::endl;
45 torch::Tensor kernelMatrix_FFT = torch::fft::fftn(kernelMatrix, \
46     {final_length}, \
47     {0}); if(MYDEBUGFLAG) std::cout<<"\t\t\t fConvolveColumns:
        35"<<std::endl;
48
49 // element-wise multiplying the two matrices
50 torch::Tensor MulProduct = torch::mul(inputMatrix_FFT, kernelMatrix_FFT); if(MYDEBUGFLAG)
    std::cout<<"\t\t\t fConvolveColumns: 38"<<std::endl;
51
52 // finding the inverse FFT
53 torch::Tensor convolvedResult = torch::fft::ifftn(MulProduct, \
54     {MulProduct.size(0)}, \
55     {0}); if(MYDEBUGFLAG) std::cout<<"\t\t\t fConvolveColumns:
        43"<<std::endl;
56
57 // bringing them back to the pipeline datatype
58 kernelMatrix = kernelMatrix.to(DATATYPE);
59
60 // over-riding the result with the input so that we can save memory
61 inputMatrix = convolvedResult.to(DATATYPE); if(MYDEBUGFLAG) std::cout<<"\t\t\t fConvolveColumns:
    46"<<std::endl;
62
63 }

```

---

### 8.3.4 Buffer 2D

---

```

1 /* =====
2 Aim: Convolve the columns of two input matrices
3 =====*/
4 #include <stdexcept>
5 #include <torch/torch.h>
6
7 #pragma once
8
9 // hash-defines
10 #ifndef DEVICE
11     // #define DEVICE      torch::kMPS
12     #define DEVICE      torch::kCPU
13 #endif
14
15 // #define DEBUG_Buffer2D true
16 #define DEBUG_Buffer2D false
17
18
19 void fBuffer2D(torch::Tensor& inputMatrix,
20     int frame_size){

```

```

21
22 // ensuring the first dimension is 1.
23 if(inputMatrix.size(0) != 1){
24     throw std::runtime_error("fBuffer2D: The first-dimension must be 1 \n");
25 }
26
27 // padding with zeros in case it is not a perfect multiple
28 if(inputMatrix.size(1)%frame_size != 0){
29     // padding with zeros
30     int numberofzeroestoadd = frame_size - (inputMatrix.size(1) % frame_size);
31     if(DEBUG_Buffer2D) {
32         std::cout << "\t\t\t fBuffer2D: frame_size = " << frame_size <<
33         std::endl;
34         std::cout << "\t\t\t fBuffer2D: inputMatrix.sizes().vec() = " << inputMatrix.sizes().vec() <<
35         std::endl;
36         std::cout << "\t\t\t fBuffer2D: numberofzeroestoadd = " << numberofzeroestoadd << std::endl;
37     }
38
39     // creating zero matrix
40     torch::Tensor zeroMatrix = torch::zeros({inputMatrix.size(0), \
41         numberofzeroestoadd, \
42         inputMatrix.size(2)});
43     if(DEBUG_Buffer2D) std::cout<<"\t\t\t fBuffer2D: zeroMatrix.sizes() =
44         "<<zeroMatrix.sizes().vec()<<std::endl;
45
46     // adding the zero matrix
47     inputMatrix = torch::cat({inputMatrix, zeroMatrix}, 1);
48     if(DEBUG_Buffer2D) std::cout<<"\t\t\t fBuffer2D: inputMatrix.sizes().vec() =
49         "<<inputMatrix.sizes().vec()<<std::endl;
50 }
51
52 // calculating some parameters
53 // int num_frames = inputMatrix.size(1)/frame_size;
54 int num_frames = std::ceil(inputMatrix.size(1)/frame_size);
55 if(DEBUG_Buffer2D) std::cout << "\t\t\t fBuffer2D: inputMatrix.sizes = " << inputMatrix.sizes().vec() <<
56     std::endl;
57 if(DEBUG_Buffer2D) std::cout << "\t\t\t fBuffer2D: framesize = " << frame_size << std::endl;
58 if(DEBUG_Buffer2D) std::cout << "\t\t\t fBuffer2D: num_frames = " << num_frames << std::endl;
59
60 // defining target shape and size
61 std::vector<int64_t> target_shape = {num_frames, \
62     frame_size, \
63     inputMatrix.size(2)};
64 std::vector<int64_t> target_strides = {frame_size * inputMatrix.size(2), \
65     inputMatrix.size(2), \
66     1};
67 if(DEBUG_Buffer2D) std::cout << "\t\t\t fBuffer2D: STATUS: created shape and strides"<< std::endl;
68
69 // creating the transformation
70 inputMatrix = inputMatrix.as_strided(target_shape, target_strides);
71 }

```

---

### 8.3.5 fAnglesToTensor

```

1 #include <torch/torch.h>
2 // function: angles to vector
3 torch::Tensor fAnglesToTensor(float azimuthal_angle,
4     float elevation_angle)
5 {
6     // calculating tensor
7     torch::Tensor coordinateTensor = torch::tensor({cos(elevation_angle) * cos(azimuthal_angle),
8         cos(elevation_angle) * sin(azimuthal_angle),
9         sin(elevation_angle)}).view({3,1});
10
11     // returning value
12     return coordinateTensor;
13 }

```

---

### 8.3.6 fCalculateCosine

---

```
1 // including headerfiles
2 #include <torch/torch.h>
3
4 // function to calculate cosine of two tensors
5 torch::Tensor fCalculateCosine(torch::Tensor inputTensor1,
6                               torch::Tensor inputTensor2)
7 {
8     // column normalizing the the two signals
9     inputTensor1 = fColumnNormalize(inputTensor1);
10    inputTensor2 = fColumnNormalize(inputTensor2);
11
12    // finding their dot product
13    torch::Tensor dotProduct = inputTensor1 * inputTensor2;
14    torch::Tensor cosineBetweenVectors = torch::sum(dotProduct,
15                                                    0,
16                                                    true);
17
18    // returning the value
19    return cosineBetweenVectors;
20
21 }
```

---



## 8.4 Main Scripts

### 8.4.1 Signal Simulation

1.

---

```

1  /*=====
2  Aim: Signal Simulation
3  =====*/
4
5
6  // including
7  #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/packages.h"
8  #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/config.h" // hash-defines
9  #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/classes.h" // class defs
10 #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/setupscripts.h" // setup-scripts
11 #include "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/include/functions.h" // functions
12
13
14 // main-function
15 int main() {
16
17     // Ensuring no-gradients are built
18     NoGradGuard no_grad;
19
20     // Building Sea-floor
21     ScattererClass SeafloorScatter;
22     thread scatterThread_t(SeafloorSetup, \
23                           ref(SeafloorScatter));
24
25     // Building ULA
26     ULAClass ula_fls, ula_port, ula_starboard;
27     thread ulaThread_t(ULASetup, \
28                       ref(ula_fls), \
29                       ref(ula_port), \
30                       ref(ula_starboard));
31
32     // Building Transmitter
33     TransmitterClass transmitter_fls, transmitter_port, transmitter_starboard;
34     thread transmitterThread_t(TransmitterSetup,
35                               ref(transmitter_fls),
36                               ref(transmitter_port),
37                               ref(transmitter_starboard));
38
39     // recombining threads
40     scatterThread_t.join(); // making the scattetr population thread join back
41     ulaThread_t.join();    // making the ULA population thread join back
42     transmitterThread_t.join(); // making the transmitter population thread join back
43
44     // building AUV
45     AUVClass auv; // instantiating class object
46     AUVSetup(&auv); // populating
47
48     // attaching components to the AUV
49     auv.ULA_fls = ula_fls; // attaching ULA-FLS to AUV
50     auv.ULA_port = ula_port; // attaching ULA-Port to AUV
51     auv.ULA_starboard = ula_starboard; // attaching ULA-Starboard to AUV
52     auv.transmitter_fls = transmitter_fls; // attaching Transmitter-FLS to AUV
53     auv.transmitter_port = transmitter_port; // attaching Transmitter-Port to AUV
54     auv.transmitter_starboard = transmitter_starboard; // attaching Transmitter-Starboard to AUV
55
56     // storing
57     ScattererClass SeafloorScatter_deepcopy = SeafloorScatter;
58
59     // pre-computing the data-structures required for processing
60     auv.init();
61
62     // mimicking movement
63     int number_of_stophops = 4;
64     // if (true) return 0;
65     for(int i = 0; i<number_of_stophops; ++i){

```

```

66
67 // time measuring
68 auto start_time = high_resolution_clock::now();
69
70 // printing some spaces
71 PRINTSPACE; PRINTSPACE; PRINTLINE; cout<<"i = "<<i<<endl; PRINTLINE
72
73 // making the deep copy
74 ScattererClass SeafloorScatter = SeafloorScatter_deepcopy;
75
76 // signal simulation
77 auv.simulateSignal(SeafloorScatter);
78
79 // saving simulated signal
80 if (SAVETENSORS) {
81
82     // saving the signal matrix tensors
83     save(auv.ULA_fls.signalMatrix, \
84         "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/signalMatrix_fls.pt");
85     save(auv.ULA_port.signalMatrix, \
86         "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/signalMatrix_port.pt");
87     save(auv.ULA_starboard.signalMatrix, \
88         "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/signalMatrix_starboard.pt");
89
90     // running python script
91     string script_to_run = \
92         "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/Python/Plot_SignalMatrix.py";
93     thread plotSignalMatrix_t(fRunSystemScriptInSeperateThread, \
94         script_to_run);
95     plotSignalMatrix_t.detach();
96
97 }
98
99
100 if (IMAGING_TOGGLE) {
101
102     // creating image from signals
103     auv.image();
104
105     // saving the tensors
106     if(SAVETENSORS){
107         // saving the beamformed images
108         save(auv.ULA_fls.beamformedImage, \
109             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/ULA_fls_image.pt");
110         // save(auv.ULA_port.beamformedImage, \
111             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/ULA_port_image.pt");
112         // save(auv.ULA_starboard.beamformedImage, \
113             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/ULA_starboard_image.pt");
114
115         // saving cartesian image
116         save(auv.ULA_fls.cartesianImage, \
117             "/Users/vrsreeganesh/Documents/GitHub/AUV/Code/C++/Assets/ULA_fls_cartesianImage.pt");
118
119         // // running python file
120         // system("python
121             /Users/vrsreeganesh/Documents/GitHub/AUV/Code/Python/Plot_BeamformedImage.py");
122         system("python /Users/vrsreeganesh/Documents/GitHub/AUV/Code/Python/Plot_cartesianImage.py");
123     }
124
125
126
127 // measuring and printing time taken
128 auto end_time = high_resolution_clock::now();
129 duration<double> time_duration = end_time - start_time;
130 PRINTDOTS; cout<<"Time taken (i = "<<i<<" = "<<time_duration.count()<<" seconds"<<endl; PRINTDOTS
131
132 // moving to next position
133 auv.step(0.5);
134
135 }
136
137
138

```

```
139
140
141
142
143     // returning
144     return 0;
145 }
```

---

# Chapter 9

## Reading

### 9.1 Primary Books

- 1.

### 9.2 Interesting Papers

# Chapter 10

## General Purpose Templated Functions

### 10.1 Concatenate Functions

```
1 // input = [vector, vector],
2 // output = [vector]
3 template <std::size_t axis, typename T>
4 auto concatenate(const std::vector<T>& input_vector_A,
5                 const std::vector<T>& input_vector_B) -> std::enable_if_t<axis == 1, std::vector<T> >
6 {
7     // creating canvas vector
8     auto num_elements {input_vector_A.size() + input_vector_B.size()};
9     auto canvas {std::vector<T>(num_elements, (T)0) };
10
11     // filling up the canvas
12     std::copy(input_vector_A.begin(), input_vector_A.end(),
13               canvas.begin());
14     std::copy(input_vector_B.begin(), input_vector_B.end(),
15               canvas.begin()+input_vector_A.size());
16
17     // moving it back
18     return std::move(canvas);
19 }
20
21 // =====
22 // input = [vector, vector],
23 // output = [matrix]
24 template <std::size_t axis, typename T>
25 auto concatenate(const std::vector<T>& input_vector_A,
26                 const std::vector<T>& input_vector_B) -> std::enable_if_t<axis == 0,
27                 std::vector<std::vector<T>> >
28 {
29     // throwing error dimensions
30     if (input_vector_A.size() != input_vector_B.size())
31         std::cerr << "concatenate:: incorrect dimensions \n";
32
33     // creating canvas
34     auto canvas {std::vector<std::vector<T>>(
35                 2, std::vector<T>(input_vector_A.size())
36                 )};
37
38     // filling up the dimensions
39     std::copy(input_vector_A.begin(), input_vector_A.end(), canvas[0].begin());
40     std::copy(input_vector_B.begin(), input_vector_B.end(), canvas[1].begin());
41
42     // moving it back
43     return std::move(canvas);
44 }
45 // =====
46 // input = [vector, vector, vector],
47 // output = [matrix]
```

```

48 template <std::size_t axis, typename T>
49 auto concatenate(const std::vector<T>& input_vector_A,
50                 const std::vector<T>& input_vector_B,
51                 const std::vector<T>& input_vector_C) -> std::enable_if_t<axis == 0,
                    std::vector<std::vector<T>>> >
52 {
53     // throwing error dimensions
54     if (input_vector_A.size() != input_vector_B.size() ||
55         input_vector_A.size() != input_vector_C.size())
56         std::cerr << "concatenate:: incorrect dimensions \n";
57
58     // creating canvas
59     auto canvas {std::vector<std::vector<T>>>(
60         3, std::vector<T>(input_vector_A.size())
61     )};
62
63     // filling up the dimensions
64     std::copy(input_vector_A.begin(), input_vector_A.end(), canvas[0].begin());
65     std::copy(input_vector_B.begin(), input_vector_B.end(), canvas[1].begin());
66     std::copy(input_vector_C.begin(), input_vector_C.end(), canvas[2].begin());
67
68     // moving it back
69     return std::move(canvas);
70 }
71
72 // =====
73 // input = [matrix, vector],
74 // output = [matrix]
75 template <std::size_t axis, typename T>
76 auto concatenate(const std::vector<std::vector<T>>& input_matrix,
77                 const std::vector<T> input_vector) -> std::enable_if_t<axis == 0,
                    std::vector<std::vector<T>>> >
78 {
79     // creating canvas
80     auto canvas {input_matrix};
81
82     // adding to the canvas
83     canvas.push_back(input_vector);
84
85     // returning
86     return std::move(canvas);
87 }

```

---

## 10.2 Data Structures

---

```
1 struct TreeNode {
2     int val;
3     TreeNode *left;
4     TreeNode *right;
5     TreeNode() : val(0), left(nullptr), right(nullptr) {}
6     TreeNode(int x) : val(x), left(nullptr), right(nullptr) {}
7     TreeNode(int x, TreeNode *left, TreeNode *right) : val(x), left(left), right(right) {}
8 };
9
10
11 struct ListNode {
12     int val;
13     ListNode *next;
14     ListNode() : val(0), next(nullptr) {}
15     ListNode(int x) : val(x), next(nullptr) {}
16     ListNode(int x, ListNode *next) : val(x), next(next) {}
17 };
```

---

## 10.3 Linspace

---

```

1  // in-place
2  template <typename T>
3  auto linspace(auto&          input,
4                auto          startvalue,
5                auto          endvalue,
6                auto          numpoints) -> void
7  {
8      auto stepsize = static_cast<T>(endvalue - startvalue)/static_cast<T>(numpoints-1);
9      for(int i = 0; i<input.size(); ++i) {input[i] = startvalue + i*stepsize;}
10 };
11 // in-place
12 template <typename T>
13 auto linspace(vector<complex<T>>& input,
14               auto          startvalue,
15               auto          endvalue,
16               auto          numpoints) -> void
17 {
18     auto stepsize = static_cast<T>(endvalue - startvalue)/static_cast<T>(numpoints-1);
19     for(int i = 0; i<input.size(); ++i) {
20         input[i] = startvalue + static_cast<T>(i)*stepsize;
21     }
22 };
23
24 // return-type
25 template <typename T>
26 auto linspace(T          startvalue,
27               T          endvalue,
28               size_t      numpoints)
29 {
30     vector<T> input(numpoints);
31     auto stepsize = static_cast<T>(endvalue - startvalue)/static_cast<T>(numpoints-1);
32
33     for(int i = 0; i<input.size(); ++i) {input[i] = startvalue + static_cast<T>(i)*stepsize;}
34
35     return input;
36 };
37
38 // return-type
39 template <typename T, typename U>
40 auto linspace(T          startvalue,
41               U          endvalue,
42               size_t      numpoints)
43 {
44     vector<double> input(numpoints);
45     auto stepsize = static_cast<double>(endvalue - startvalue)/static_cast<double>(numpoints-1);
46
47     for(int i = 0; i<input.size(); ++i) {input[i] = startvalue + i*stepsize;}
48
49     return input;
50 };

```

---



## 10.4 Max

---

```
1 template <std::size_t axis, typename T>
2 auto max(const std::vector<std::vector<T>>> input_matrix) -> std::enable_if_t<axis == 1,
   std::vector<std::vector<T>>> >
3 {
4     // setting up canvas
5     auto canvas {std::vector<std::vector<T>>>(input_matrix.size(), std::vector<T>(1))};
6
7     // filling up the canvas
8     for(auto row = 0; row < input_matrix.size(); ++row)
9         canvas[row][0] = *(std::max_element(input_matrix[row].begin(), input_matrix[row].end()));
10
11     // returning
12     return std::move(canvas);
13 }
```

---

## 10.5 Meshgrid

---

```

1 // =====
2 template <typename T>
3 auto meshgrid(const std::vector<T>& x,
4               const std::vector<T>& y)
5 {
6
7     // creating and filling x-grid
8     std::vector<std::vector<T>> xcanvas(y.size(), std::vector<T>(x.size(), 0));
9     for(auto row = 0; row < y.size(); ++row)
10         std::copy(x.begin(), x.end(), xcanvas[row].begin());
11
12     // creating and filling y-grid
13     std::vector<std::vector<T>> ycanvas(y.size(), std::vector<T>(x.size(), 0));
14     for(auto col = 0; col < x.size(); ++col)
15         for(auto row = 0; row < y.size(); ++row)
16             ycanvas[row][col] = y[row];
17
18     // returning
19     return std::move(std::pair{xcanvas, ycanvas});
20 }
21
22 // =====
23 template <typename T>
24 auto meshgrid(std::vector<T>&& x,
25               std::vector<T>&& y)
26 {
27
28     // creating and filling x-grid
29     std::vector<std::vector<T>> xcanvas(y.size(), std::vector<T>(x.size(), 0));
30     for(auto row = 0; row < y.size(); ++row)
31         std::copy(x.begin(), x.end(), xcanvas[row].begin());
32
33     // creating and filling y-grid
34     std::vector<std::vector<T>> ycanvas(y.size(), std::vector<T>(x.size(), 0));
35     for(auto col = 0; col < x.size(); ++col)
36         for(auto row = 0; row < y.size(); ++row)
37             ycanvas[row][col] = y[row];
38
39     // returning
40     return std::move(std::pair{xcanvas, ycanvas});
41 }
42

```

---

## 10.6 Minimum

---

```
1 template <std::size_t axis, typename T>
2 auto min(std::vector<std::vector<T>> input_matrix) -> std::enable_if_t<axis == 1,
   std::vector<std::vector<T>> >
3 {
4     // creating canvas
5     auto canvas {std::vector<std::vector<T>>(input_matrix.size(), std::vector<T>(1))};
6
7     // storing the values
8     for(auto row = 0; row < input_matrix.size(); ++row)
9         canvas[row][0] = *(std::min_element(input_matrix[row].begin(), input_matrix[row].end()));
10
11     // returning the value
12     return std::move(canvas);
13 }
```

---

## 10.7 Division

---

```

1 // =====
2 // matrix division with scalars
3 template <typename T>
4 auto operator/((const std::vector<T>& input_vector,
5                const T& input_scalar)
6 {
7     // creating canvas
8     auto canvas {input_vector};
9
10    // filling canvas
11    std::transform(canvas.begin(), canvas.end(),
12                  canvas.begin(),
13                  [&input_scalar](const auto& argx){
14                      return static_cast<double>(argx) / static_cast<double>(input_scalar);
15                  });
16
17    // returning value
18    return std::move(canvas);
19 }
20 // =====
21 // matrix division with scalars
22 template <typename T>
23 auto operator/=(const std::vector<T>& input_vector,
24                const T& input_scalar)
25 {
26    // creating canvas
27    auto canvas {input_vector};
28
29    // filling canvas
30    std::transform(canvas.begin(), canvas.end(),
31                  canvas.begin(),
32                  [&input_scalar](const auto& argx){
33                      return static_cast<double>(argx) / static_cast<double>(input_scalar);
34                  });
35
36    // returning value
37    return std::move(canvas);
38 }

```

---

## 10.8 Addition

---

```

1 // =====
2 // y = vector + vector
3 template <typename T>
4 std::vector<T> operator+(const std::vector<T>& a,
5                          const std::vector<T>& b)
6 {
7     // Identify which is bigger
8     const auto& big = (a.size() > b.size()) ? a : b;
9     const auto& small = (a.size() > b.size()) ? b : a;
10
11     std::vector<T> result = big; // copy the bigger one
12
13     // Add elements from the smaller one
14     for (size_t i = 0; i < small.size(); ++i) {
15         result[i] += small[i];
16     }
17
18     return result;
19 }
20 // =====
21 template <typename T>
22 std::vector<T>& operator+=(std::vector<T>& a,
23                          const std::vector<T>& b) {
24
25     const auto& small = (a.size() < b.size()) ? a : b;
26     const auto& big = (a.size() < b.size()) ? b : a;
27
28     // If b is bigger, resize 'a' to match
29     if (a.size() < b.size()) {a.resize(b.size());}
30
31     // Add elements
32     for (size_t i = 0; i < small.size(); ++i) {a[i] += b[i];}
33
34     // returning elements
35     return a;
36 }
37 // =====
38 template <typename T>
39 std::vector<std::vector<T>> operator+(const std::vector<std::vector<T>>& a,
40                                     const std::vector<std::vector<T>>& b)
41 {
42     // fetching dimensions
43     const auto& num_rows_A {a.size()};
44     const auto& num_cols_A {a[0].size()};
45     const auto& num_rows_B {b.size()};
46     const auto& num_cols_B {b[0].size()};
47
48     // choosing the three different metrics
49     if (num_rows_A != num_rows_B && num_cols_A != num_cols_B){
50         cout << format("a.dimensions = [{},{}], b.shape = [{},{}]\n",
51                        num_rows_A, num_cols_A,
52                        num_rows_B, num_cols_B);
53         std::cerr << "dimensions don't match\n";
54     }
55
56     // creating canvas
57     auto canvas {std::vector<std::vector<T>>>(
58         std::max(num_rows_A, num_rows_B),
59         std::vector<T>(std::max(num_cols_A, num_cols_B), (T)0.00)
60     )};
61
62     // performing addition
63     if (num_rows_A == num_rows_B && num_cols_A == num_cols_B){
64         for(auto row = 0; row < num_rows_A; ++row){
65             std::transform(a[row].begin(), a[row].end(),
66                            b[row].begin(),
67                            canvas[row].begin(),
68                            std::plus<T>());
69         }
70     }
71     else if(num_rows_A == num_rows_B){
72

```

```

73 // if number of columns are different, check if one of the cols are one
74 const auto min_num_cols {std::min(num_cols_A, num_cols_B)};
75 if (min_num_cols != 1) {std::cerr<< "Operator+: unable to broadcast\n";}
76 const auto max_num_cols {std::max(num_cols_A, num_cols_B)};
77
78 // using references to tag em differently
79 const auto& big_matrix {num_cols_A > num_cols_B ? a : b};
80 const auto& small_matrix {num_cols_A < num_cols_B ? a : b};
81
82 // Adding to canvas
83 for(auto row = 0; row < canvas.size(); ++row){
84     std::transform(big_matrix[row].begin(), big_matrix[row].end(),
85                   canvas[row].begin(),
86                   [&small_matrix,
87                    &row](const auto& argx){
88                         return argx + small_matrix[row][0];
89                     });
90 }
91 }
92 else if(num_cols_A == num_cols_B){
93
94     // check if the smallest column-number is one
95     const auto min_num_rows {std::min(num_rows_A, num_rows_B)};
96     if(min_num_rows != 1) {std::cerr<< "Operator+: unable to broadcast\n";}
97     const auto max_num_rows {std::max(num_rows_A, num_rows_B)};
98
99     // using references to differentiate the two matrices
100    const auto& big_matrix {num_rows_A > num_rows_B ? a : b};
101    const auto& small_matrix {num_rows_A < num_rows_B ? a : b};
102
103    // adding to canvas
104    for(auto row = 0; row < canvas.size(); ++row){
105        std::transform(big_matrix[row].begin(), big_matrix[row].end(),
106                      small_matrix[0].begin(),
107                      canvas[row].begin(),
108                      [](const auto& argx, const auto& argy){
109                          return argx + argy;
110                      });
111    }
112 }
113 else {
114     PRINTLINE PRINTLINE PRINTLINE PRINTLINE PRINTLINE
115     cout << format("check this again \n");
116 }
117
118 // returning
119 return std::move(canvas);
120 }

```

---

## 10.9 Multiplication (Element-wise)

---

```

1 // scalar * vector =====
2 template <typename T>
3 auto operator*(T scalar,
4               const vector<T>& inputvector){
5     vector<T> temp(inputvector.size());
6     std::transform(inputvector.begin(),
7                   inputvector.end(),
8                   temp.begin(),
9                   [&scalar](T x){return scalar * x;});
10    return temp;
11 }
12 // scalar * vector =====
13 template <typename T1, typename T2>
14 auto operator*(T1 scalar,
15               const vector<T2>& inputvector){
16     using T3 = decltype(std::declval<T1>() * std::declval<T2>());
17     vector<T3> temp(inputvector.size());
18     std::transform(inputvector.begin(),
19                   inputvector.end(),
20                   temp.begin(),
21                   [&scalar](auto x){return static_cast<T3>(scalar) * static_cast<T3>(x);});
22    return temp;
23 }
24 // vector * scalar =====
25 template <typename T>
26 auto operator*(const vector<T>& inputvector,
27               T scalar)
28 {
29     vector<T> temp(inputvector.size());
30     std::transform(inputvector.begin(), inputvector.end(), temp.begin(), [&scalar](T x){return scalar * x;});
31     return temp;
32 }
33 // vector * vector =====
34 template <typename T>
35 auto operator*(const std::vector<T>& input_vector_A,
36               const std::vector<T>& input_vector_B)
37 {
38     // throwing error: size-disparity
39     if (input_vector_A.size() != input_vector_B.size()) {std::cerr << "operator*: size disparity \n";}
40
41     // creating canvas
42     auto canvas {std::vector<T>(input_vector_A)};
43
44     // element-wise multiplying
45     std::transform(input_vector_A.begin(), input_vector_A.end(),
46                   input_vector_B.begin(),
47                   canvas.begin(),
48                   [](const auto& argx, const auto& argy){
49                       return argx * argy;
50                   });
51
52     // moving it back
53     return std::move(canvas);
54 }
55 // scalar * matrix =====
56 template <typename T>
57 auto operator*(T scalar,
58               const std::vector<std::vector<T>>& inputMatrix)
59 {
60     std::vector<std::vector<T>> temp {inputMatrix};
61     for(int i = 0; i<inputMatrix.size(); ++i){
62         std::transform(inputMatrix[i].begin(),
63                       inputMatrix[i].end(),
64                       temp[i].begin(),
65                       [&scalar](T x){return scalar * x;});
66     }
67     return temp;
68 }
69 // scalar * matrix =====
70 template <typename T1, typename T2>
71 auto operator*(T1 scalar,
72               const std::vector<std::vector<T2>>& inputMatrix)

```

```

73 {
74     std::vector<std::vector<T2>> temp {inputMatrix};
75     for(int i = 0; i<inputMatrix.size(); ++i){
76         std::transform(inputMatrix[i].begin(),
77             inputMatrix[i].end(),
78             temp[i].begin(),
79             [&scalar](T2 x){return static_cast<T2>(scalar) * x;});
80     }
81     return temp;
82 }
83 // matrix * matrix =====
84 template <typename T>
85 auto operator*(const std::vector<std::vector<T>>& A,
86     const std::vector<std::vector<T>>& B) -> std::vector<std::vector<T>>
87 {
88     // Case 1: element-wise multiplication
89     if (A.size() == B.size() && A[0].size() == B[0].size()) {
90         std::vector<std::vector<T>> C(A.size(), std::vector<T>(A[0].size()));
91         for (std::size_t row = 0; row < A.size(); ++row) {
92             std::transform(A[row].begin(), A[row].end(),
93                 B[row].begin(),
94                 C[row].begin(),
95                 [](const auto& x, const auto& y){ return x * y; });
96         }
97         return C;
98     }
99
100     // Case 2: broadcast column vector
101     else if (A.size() == B.size() && B[0].size() == 1) {
102         std::vector<std::vector<T>> C(A.size(), std::vector<T>(A[0].size()));
103         for (std::size_t row = 0; row < A.size(); ++row) {
104             std::transform(A[row].begin(), A[row].end(),
105                 C[row].begin(),
106                 [&](const auto& x){ return x * B[row][0]; });
107         }
108         return C;
109     }
110
111     // case 3: when second matrix contains just one row
112     // case 4: when first matrix is just one column
113     // case 5: when second matrix is just one column
114
115     // Otherwise, invalid
116     else {
117         throw std::runtime_error("operator* dimension mismatch");
118     }
119 }
120 // matrix-multiplication =====
121 template <typename T1, typename T2>
122 auto matmul(const std::vector<std::vector<T1>>& matA,
123     const std::vector<std::vector<T2>>& matB)
124 {
125
126     // throwing error
127     if (matA[0].size() != matB.size()) {std::cerr << "dimension-mismatch \n";}
128
129     // getting result-type
130     using ResultType = decltype(std::declval<T1>() * std::declval<T2>() + \
131         std::declval<T1>() * std::declval<T2>() );
132
133     // creating aliases
134     auto finalnumrows {matA.size()};
135     auto finalnumcols {matB[0].size()};
136
137     // creating placeholder
138     auto rowcolproduct = [&](auto rowA, auto colB){
139         ResultType temp {0};
140         for(int i = 0; i < matA.size(); ++i) {temp += static_cast<ResultType>(matA[rowA][i]) +
141             static_cast<ResultType>(matB[i][colB]);}
142         return temp;
143     };
144
145     // producing row-column combinations
146     std::vector<std::vector<ResultType>> finaloutput(finalnumrows, std::vector<ResultType>(finalnumcols));

```



```
146     for(int row = 0; row < finalnumrows; ++row){for(int col = 0; col < finalnumcols;
147         ++col){finaloutput[row][col] = rowcolproduct(row, col);}}
148
149     // returning
150     return finaloutput;
151 }
152 // scalar operators =====
153 auto operator*(const std::complex<double> complexscalar,
154               const double doublescalar){
155     return complexscalar * static_cast<std::complex<double>>(doublescalar);
156 }
157 auto operator*(const double doublescalar,
158               const std::complex<double> complexscalar){
159     return complexscalar * static_cast<std::complex<double>>(doublescalar);
160 }
161 auto operator*(const std::complex<double> complexscalar,
162               const int scalar){
163     return complexscalar * static_cast<std::complex<double>>(scalar);
164 }
165 auto operator*(const int scalar,
166               const std::complex<double> complexscalar){
167     return complexscalar * static_cast<std::complex<double>>(scalar);
168 }
```

---

## 10.10 Subtraction

---

```

1 // =====
2 // Aim: subtracting scalar from a vector
3 template <typename T>
4 std::vector<T> operator-(const std::vector<T>& a, const T scalar){
5     std::vector<T> temp(a.size());
6     std::transform(a.begin(),
7                   a.end(),
8                   temp.begin(),
9                   [scalar](T x){return (x - scalar);});
10    return temp;
11 }
12 // =====
13 template <typename T>
14 auto operator-(const std::vector<std::vector<T>>& input_matrix_A,
15               const std::vector<std::vector<T>>& input_matrix_B)
16 {
17     // throwing error in case of dimension differences
18     if (input_matrix_A.size() != input_matrix_B.size() ||
19         input_matrix_A[0].size() != input_matrix_B[0].size())
20         std::cerr << "operator- dimension mismatch\n";
21
22     // setting up canvas
23     auto canvas {std::vector<std::vector<T>>>(
24         input_matrix_A.size(),
25         std::vector<T>(input_matrix_A[0].size())
26     )};
27
28     // subtracting values
29     for(auto row = 0; row < input_matrix_B.size(); ++row)
30         std::transform(input_matrix_A[row].begin(), input_matrix_A[row].end(),
31                       input_matrix_B[row].begin(),
32                       canvas[row].begin(),
33                       [](const auto& x, const auto& y){
34                           return x - y;
35                       });
36
37     // returning
38     return std::move(canvas);
39 }
40 
```

---

## 10.11 Printing Containers

---

```

1  // vector printing function
2  template<typename T>
3  void fPrintVector(vector<T> input){
4      for(auto x: input) cout << x << ", ";
5      cout << endl;
6  }
7
8  template<typename T>
9  void fpv(vector<T> input){
10     for(auto x: input) cout << x << ", ";
11     cout << endl;
12 }
13 // =====
14 template<typename T>
15 void fPrintMatrix(const std::vector<std::vector<T>> input_matrix){
16     for(const auto& row: input_matrix)
17         cout << format("{}\n", row);
18 }
19 template <typename T>
20 void fPrintMatrix(const string& input_string,
21                  const std::vector<std::vector<T>> input_matrix){
22     cout << format("{} = \n", input_string);
23     for(const auto& row: input_matrix)
24         cout << format("{}\n", row);
25 }
26
27
28 template<typename T, typename T1>
29 void fPrintHashMap(unordered_map<T, T1> input){
30     for(auto x: input){
31         cout << format("{}{} | ", x.first, x.second);
32     }
33     cout << endl;
34 }
35
36 void fPrintBinaryTree(TreeNode* root){
37     // sending it back
38     if (root == nullptr) return;
39
40     // printing
41     PRINTLINE
42     cout << "root->val = " << root->val << endl;
43
44     // calling the children
45     fPrintBinaryTree(root->left);
46     fPrintBinaryTree(root->right);
47
48     // returning
49     return;
50 }
51
52
53 void fPrintLinkedList(ListNode* root){
54     if (root == nullptr) return;
55     cout << root->val << " -> ";
56     fPrintLinkedList(root->next);
57     return;
58 }
59
60 template<typename T>
61 void fPrintContainer(T input){
62     for(auto x: input) cout << x << ", ";
63     cout << endl;
64     return;
65 }
66 // =====
67 template <typename T>
68 auto size(std::vector<std::vector<T>> inputMatrix){
69     cout << format("{}{} \n", inputMatrix.size(), inputMatrix[0].size());
70 }
71
72 template <typename T>

```

```
73 auto size(const std::string inputstring, std::vector<std::vector<T>> inputMatrix){  
74     cout << format("{} = [{} , {}]\n", inputstring, inputMatrix.size(), inputMatrix[0].size());  
75 }
```

---

## 10.12 Random Number Generation

---

```

1 // =====
2 template <typename T>
3 auto rand(const T min, const T max) {
4     static std::random_device rd; // Seed
5     static std::mt19937 gen(rd()); // Mersenne Twister generator
6     std::uniform_real_distribution<> dist(min, max);
7     return dist(gen);
8 }
9 // =====
10 template <typename T>
11 auto rand(const T min,
12           const T max,
13           const size_t numelements)
14 {
15     static std::random_device rd; // Seed
16     static std::mt19937 gen(rd()); // Mersenne Twister generator
17     std::uniform_real_distribution<> dist(min, max);
18
19     // building the finaloutput
20     vector<T> finaloutput(numelements);
21     for(int i = 0; i<finaloutput.size(); ++i) {finaloutput[i] = static_cast<T>(dist(gen));}
22
23     return finaloutput;
24 }
25 // =====
26 template <typename T>
27 auto rand(const T argmin,
28           const T argmax,
29           const vector<int> dimensions)
30 {
31
32     // throwing an error if dimension is greater than two
33     if (dimensions.size() > 2) {std::cerr << "dimensions are too high\n";}
34
35     // creating random engine
36     static std::random_device rd; // Seed
37     static std::mt19937 gen(rd()); // Mersenne Twister generator
38     std::uniform_real_distribution<> dist(argmin, argmax);
39
40     // building the finaloutput
41     vector<vector<T>> finaloutput;
42     for(int i = 0; i<dimensions[0]; ++i){
43         vector<T> temp;
44         for(int j = 0; j<dimensions[1]; ++j) {temp.push_back(dist(gen));}
45         // cout << format("\t\t temp = {}\n", temp);
46
47         finaloutput.push_back(temp);
48     }
49
50     // returning the finaloutput
51     return finaloutput;
52 }
53 // =====
54 auto rand(const vector<int> dimensions)
55 {
56
57     using ReturnType = double;
58
59     // throwing an error if dimension is greater than two
60     if (dimensions.size() > 2) {std::cerr << "dimensions are too high\n";}
61
62     // creating random engine
63     static std::random_device rd; // Seed
64     static std::mt19937 gen(rd()); // Mersenne Twister generator
65     std::uniform_real_distribution<> dist(0.00, 1.00);
66
67     // building the finaloutput
68     vector<vector<ReturnType>> finaloutput;
69     for(int i = 0; i<dimensions[0]; ++i){
70         vector<ReturnType> temp;
71         for(int j = 0; j<dimensions[1]; ++j) {temp.push_back(dist(gen));}

```

```

73     finaloutput.push_back(std::move(temp));
74 }
75
76 // returning the finaloutput
77 return std::move(finaloutput);
78
79 }
80 // =====
81 template <typename T>
82 auto rand_complex_double(const T          argmin,
83                          const T          argmax,
84                          const vector<int>& dimensions)
85 {
86
87     // throwing an error if dimension is greater than two
88     if (dimensions.size() > 2) {std::cerr << "dimensions are too high\n";}
89
90     // creating random engine
91     static std::random_device rd; // Seed
92     static std::mt19937 gen(rd()); // Mersenne Twister generator
93     std::uniform_real_distribution<> dist(argmin, argmax);
94
95     // building the finaloutput
96     vector<vector<complex<double>>> finaloutput;
97     for(int i = 0; i<dimensions[0]; ++i){
98         vector<complex<double>> temp;
99         for(int j = 0; j<dimensions[1]; ++j) {temp.push_back(static_cast<double>(dist(gen)));}
100        finaloutput.push_back(std::move(temp));
101    }
102
103    // returning the finaloutput
104    return finaloutput;
105
106 }

```

---

## 10.13 Reshape

---

```

1 // =====
2 // reshaping a matrix into another matrix
3 template <std::size_t M, std::size_t N, typename T>
4 auto reshape(const std::vector<std::vector<T>>& input_matrix){
5
6     // verifying size stuff
7     if (M*N != input_matrix.size() * input_matrix[0].size())
8         std::cerr << "Dimensions are quite different\n";
9
10    // creating canvas
11    auto canvas {std::vector<std::vector<T>>(
12        M, std::vector<T>(N, (T)0)
13    )};
14
15    // writing to canvas
16    size_t tid {0};
17    size_t target_row {0};
18    size_t target_col {0};
19    for(auto row = 0; row<input_matrix.size(); ++row){
20        for(auto col = 0; col < input_matrix[0].size(); ++col){
21            tid = row * input_matrix[0].size() + col;
22            target_row = tid/N;
23            target_col = tid%N;
24            canvas[target_row][target_col] = input_matrix[row][col];
25        }
26    }
27
28    // moving it back
29    return std::move(canvas);
30 }
31 // =====
32 // reshaping a matrix into a vector
33 template<std::size_t M, typename T>
34 auto reshape(const std::vector<std::vector<T>>& input_matrix){
35
36     // checking element-count validity
37     if (M != input_matrix.size() * input_matrix[0].size())
38         std::cerr << "Number of elements differ\n";
39
40     // creating canvas
41     auto canvas {std::vector<T>(M, 0)};
42
43     // filling canvas
44     for(auto row = 0; row < input_matrix.size(); ++row)
45         for(auto col = 0; col < input_matrix[0].size(); ++col)
46             canvas[row * input_matrix.size() + col] = input_matrix[row][col];
47
48     // moving it back
49     return std::move(canvas);
50 }
51
52 // =====
53 // Matrix to matrix
54 // =====
55 template<typename T>
56 auto reshape(const std::vector<std::vector<T>>& input_matrix,
57             const std::size_t M,
58             const std::size_t N){
59
60     // checking element-count validity
61     if (M * N != input_matrix.size() * input_matrix[0].size())
62         std::cerr << "Number of elements differ\n";
63
64     // creating canvas
65     auto canvas {std::vector<std::vector<T>>(
66        M, std::vector<T>(N, (T)0)
67    )};
68
69    // writing to canvas
70    size_t tid {0};
71    size_t target_row {0};
72    size_t target_col {0};

```

```

73     for(auto row = 0; row<input_matrix.size(); ++row){
74         for(auto col = 0; col < input_matrix[0].size(); ++col){
75             tid      = row * input_matrix[0].size() + col;
76             target_row = tid/N;
77             target_col = tid%N;
78             canvas[target_row][target_col] = input_matrix[row][col];
79         }
80     }
81
82     // moving it back
83     return std::move(canvas);
84 }
85
86 // =====
87 // converting a matrix into a vector
88 // =====
89 template<typename T>
90 auto reshape(const std::vector<std::vector<T>>& input_matrix,
91             const size_t M){
92
93     // checking element-count validity
94     if (M != input_matrix.size() * input_matrix[0].size())
95         std::cerr << "Number of elements differ\n";
96
97     // creating canvas
98     auto canvas = std::vector<T>(M, 0);
99
100    // filling canvas
101    for(auto row = 0; row < input_matrix.size(); ++row)
102        for(auto col = 0; col < input_matrix[0].size(); ++col)
103            canvas[row * input_matrix.size() + col] = input_matrix[row][col];
104
105    // moving it back
106    return std::move(canvas);
107 }

```

---



## 10.14 Transpose

---

```
1  template <typename T>
2  auto transpose(const std::vector<T> input_vector){
3
4      // creating canvas
5      auto canvas {std::vector<std::vector<T>>>{
6          input_vector.size(),
7          std::vector<T>(1)
8      }};
9
10     // filling canvas
11     for(auto i = 0; i < input_vector.size(); ++i){
12         canvas[i][0] = input_vector[i];
13     }
14
15     // moving it back
16     return std::move(canvas);
17 }
```

---

## 10.15 CSV File-Writes

---

```

1 // =====
2 template <typename T>
3 void fWriteVector(const vector<T>&          inputvector,
4                  const string&            filename){
5
6     // opening a file
7     std::ofstream fileobj(filename);
8     if (!fileobj) {return;}
9
10    // writing the real parts in the first column and the imaginary parts in the second column
11    if constexpr(std::is_same_v<T, std::complex<double>> ||
12                std::is_same_v<T, std::complex<float>> ||
13                std::is_same_v<T, std::complex<long double>>){
14        for(int i = 0; i<inputvector.size(); ++i){
15            // adding entry
16            fileobj << inputvector[i].real() << "+" << inputvector[i].imag() << "i";
17
18            // adding delimiter
19            if(i!=inputvector.size()-1) {fileobj << ",";}
20            else {fileobj << "\n";}
21        }
22    }
23    else{
24        for(int i = 0; i<inputvector.size(); ++i){
25            fileobj << inputvector[i];
26            if(i!=inputvector.size()-1) {fileobj << ",";}
27            else {fileobj << "\n";}
28        }
29    }
30
31    // return
32    return;
33 }
34 // Matrix writing =====
35 template <typename T>
36 auto fWriteMatrix(const std::vector<std::vector<T>>> inputMatrix,
37                  const string                        filename){
38
39    // opening a file
40    std::ofstream fileobj(filename);
41
42    // writing
43    if (fileobj){
44        for(int i = 0; i<inputMatrix.size(); ++i){
45            for(int j = 0; j<inputMatrix[0].size(); ++j){
46                fileobj << inputMatrix[i][j];
47                if (j!=inputMatrix[0].size()-1) {fileobj << ",";}
48                else {fileobj << "\n";}
49            }
50        }
51    }
52    else{
53        cout << format("File-write to {} failed\n", filename);
54    }
55 }
56
57
58 template <>
59 auto fWriteMatrix(const std::vector<std::vector<std::complex<double>>>> inputMatrix,
60                  const string                        filename){
61
62    // opening a file
63    std::ofstream fileobj(filename);
64
65    // writing
66    if (fileobj){
67        for(int i = 0; i<inputMatrix.size(); ++i){
68            for(int j = 0; j<inputMatrix[0].size(); ++j){
69                fileobj << inputMatrix[i][j].real() << "+" << inputMatrix[i][j].imag() << "i";
70                if (j!=inputMatrix[0].size()-1) {fileobj << ",";}
71                else {fileobj << "\n";}
72            }

```

```
73     }  
74 }  
75 else{  
76     cout << format("File-write to {} failed\n", filename);  
77 }  
78 }
```

---

## 10.16 abs

---

```
1 // =====
2 // y = abs(vector)
3 template <typename T>
4 auto abs(const std::vector<T>& input_vector)
5 {
6     // creating canvas
7     auto canvas {input_vector};
8
9     // calculating abs
10    std::transform(canvas.begin(),
11                  canvas.end(),
12                  canvas.begin(),
13                  [](auto& argx){return std::abs(argx);});
14
15    // returning
16    return std::move(canvas);
17 }
18 // =====
19 // y = abs(matrix)
20 template <typename T>
21 auto abs(const std::vector<std::vector<T>> input_matrix)
22 {
23     // creating canvas
24     auto canvas {input_matrix};
25
26     // applying element-wise abs
27    std::transform(input_matrix.begin(),
28                  input_matrix.end(),
29                  input_matrix.begin(),
30                  [](auto& argx){return std::abs(argx);});
31
32    // returning
33    return std::move(canvas);
34 }
```

---