Homework 2: Radar Imaging

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1 Characteristic of the mission and Introduction

The characteristic and parameters of the mission are encoded at the beginning of our code as seen on figure 1.

```
1 %% initialization
2 - clc
3 - C = 3e8; % spead of light
4 - B = 170e6; % bandwidth
5 - kr = 1.597256e12; % chirp rate
6 - fc = 5.42876e9;
7 - lambda = C / fc;
8 - fs = 24485000; % sampling freq
9 - ts = 1/fs;
10 - beam_width = 0.1920;
11 - prf = 307.292; % pulse repetition frequency
```

Figure 1: Parameters

It is also asked to provide the resolution in range and azimuth, as dictated by the following equations $\rho_r = \frac{C}{2B}$ and $\rho_y = \frac{\lambda}{2L_s}R$. This range and azimuth resolution will dictate what is the minimal size of cell we should consider in our projection as seen on figure 2

2 Focusing by Time Domain Back Projection (TDBP)

2.1 Creating the back projection grid

As stated previously, our pixel size and therefore back projection grid must be generated by keeping in mind the range and azimuth resolution that limits our system.

 $^{^{1}}L_{s}$ is the length of the array , in our case , it is a synthetic array , therefore it is dictated by the speed of the plane and the number of sample generated.

Figure 2: We extract the final and initial positions of the plane and compute both resolutions. with that we create pixels

It is good as a rule of thumb to use 0.75 times the resolution in azimuth and range to determine the size of our pixels.

2.2 Change of spatial reference system

From then, as explain on appendix B in the provided document, we convert our initial reference system, the position data of the plane from latitude, longitude and altitude (LLA) to an intermediate referential that yields Northing, Easting and altitude (ENU). To do this transformation we needed to use a reference, we settled on using the mean value of longitude and latitude. Finally we convert this into our final referential by applying a rotation matrix to the ENU coordinates so that the East and North axis should is in a way that the East axis aligns in the direction of the trajectory. From there we determine an azimuth, range, altitude reference system. This procedure can be visualised on figure 3.

Figure 3: Transformation of our spacial reference system from LLA to ENU to Range/Azimuth/height.

2.3 Distance Computations

In this section, now that the coordinates system has been redefined , we can compute the distance from each pixel to each element of our virtual array (each slow time position of the plane) as

Figure 4: Calculating the distance from each pixel , to each of the 3884 "virtual antennas" created by the trajectory of our plane .

2.4 Focus pixel on the ground

This section is to artificially create the range compressed data that would have been receiving on the plane. Now that all the slow time distance to all pixel have been computed , we can find their corresponding double travel time (we are talking in fast time ,to go from the plane to the pixel and back) and map it to have the corresponding slow time. Also we can create the focus pixel as described on part 4.See figure 5

Figure 5: Creating Focused pixel

2.5 Frequency to distance convertion

The idea of this section , is to compute a FFT in fast time of the signal received by the plane , by doing so and coupled by that fact that we are using a chirp there is a clear correspondence between the frequency domain and the spacial (distance) domain². So to say,the values of a pixel at distance r_1 is mapped in our signal at a frequency $f_1 = K_r \frac{2r_1}{C}$, with K_r being the chip rate , the speed at which the chip change in frequency with respect to time. Therefore for each virtual antenna, we can compute the FFT in fast time and transform the frequency axis into a distance axis, by applying the equation stated previously , and project to each pixel the corresponding signal value. By doing so and by adding the contribution of each virtual antenna we get our final desired "picture". In order to do so we will also use some zero padding (adding zeros at the end of our signal) while doing the FFT to "artificially" increase the number of possible indices in the frequency domain. It is important to keep in mind that no matter what the highest index corresponds to the sampling frequency f_s .

 $^{^2{\}rm See}$ Appendix A of homework.

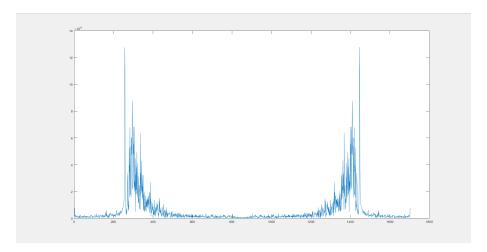


Figure 6: FFT in fast time of our received signal, without zero padding, the signal being real is symmetrical, so only the first half matters.

Figure 7: FFT in fast time with zero padding and range transformation.

2.6 Selecting a random pixel

To start off the TDBP we need to pick a target pixel in a random position on our grid, see figure 8

Figure 8: Picking a random pixel from the grid

2.7 Range compressed data simulation

Now, the range compressed data is "artificially" generated according to the formula on point 5.2. In this formula W is the antenna pattern, we decided to go for a sinc antenna pattern for our implementation. That sinc has as argument all the angle divided by the beam width. See figure 9.

```
** range compressed data computation

** calculating R0 = di[[stance from the random target to the trajectory of the stance in the stance from the random target to the trajectory of the stance from the random target to the trajectory of the stance from the stance from the random target to the trajectory of the stance from the stance
```

Figure 9: Generating the range compressed data

2.8 Time Domain back projection

For this section, we created two different implementation that also using different interpolation functions.

2.8.1 First implementation

As explain in section 2.5, the idea is that for each slow time we find the corresponding ranged compressed data with the frequency/range mapping. Knowing the distance, we know the range, we therefore also know at what frequency to look at. Then having found the corresponding value, we find all the other pixels at the same distance from the SAR as the current one. We repeat this for each 3884 slow time and add each contribution. Then we should have a 2D SINC function around the main pixel position.

Figure 10: TDBP1.1

```
comp.nei_dist "abs(neighbor_dis - target_dis(l,l)); % subtracting the distance of neighbor from the target distance % finding a minimum values (x1_min_min_l) = 10; comp_nei_dist(indi_min_l) = 10; comp_nei_dist(indi_min_l)
```

Figure 11: TDBP1.2

```
end
neighbor_pos_l; % new meighbor_pos_l; % vector representing the distance of each neighbor to the plane position
dif_l = plane_position(:,l)'-neighbor_pos_l; % vector representing the distance of each neighbor to the plane position
neighbor_dis_l = zeros(0,1);
for m = litls
neighbor_dis_l(m,l) = norm(dif_l(m,:)); % calculating the distance
end
if k == 1
comp_nei_dist_l =abs( neighbor_dis_l - target_dis(l,l)); % subtracting the distance of neighbor from the target distance

253 - comp_nei_dist_l =abs( neighbor_dis_l - temp_dis); % subtracting the distance of neighbor from the prevouse target distance

254 - comp_nei_dist_l =abs( neighbor_dis_l - temp_dis); % subtracting the distance of neighbor from the prevouse target distance

255 - distance
256 - distance
257 - ois plane_neighbor_dis_l - target_dis(l,l));
258 - ois plane_neighbor_dis_l - target_dis(l,l));
259 - break
260 - end
261 - end
262 - end
263 - find the the most similar distance
264 - [xl_min_ind_min] = min(comp_nei_dist_l);
265 - comp_nei_dist_l (ind_min_l);
266 - comp_nei_dist_l (ind_min_l);
267 - comp_nei_dist_l (ind_min_l) = nin(comp_nei_dist_l);
268 - comp_nei_dist_l (ind_min_l) = nin(comp_nei_dist_l);
270 - pred_neighbor_dis_l(ind_min_l);
271 - comp_nei_dist_l (ind_min_l) = nin(comp_nei_dist_l);
272 - comp_nei_dist_l (ind_min_l) = nin(comp_nei_dist_l);
273 - break
274 - break
275 - break
276 - comp_nei_dist_l (ind_min_l) = pixel_size_x
277 - break
278 - clseif new_neighbor(l,l) == pixel_size_y
279 - break
280 - clseif new_neighbor(l,2) == pixel_size_y
270 - break
271 - break
272 - clseif new_neighbor(l,2) == pixel_size_y
273 - break
274 - clseif new_neighbor(l,2) == pixel_size_y
275 - break
276 - clseif new_neighbor(l,2) == pixel_size_y
277 - break
278 - clseif new_neighbor(l,2) == pixel_size_y
279 - clseif new_neighbor(l,2) == pixel_size_y
270 - clseif new_neighbor(l,2) == pixel_size_y
271 - clseif new_neighbor(l,2) == pixel_size_y
272 - clseif new_neighbor(l,2) == pixel_size_y
273 - clseif new_neighbor(l,2) == pixel_size_y
274 - clseif new_
```

Figure 12: TDBP1.3

Figure 13: TDBP1.4

Figure 14: TDBP1.5

2.8.2 Second implementation

In this section we will use the interp1 function.

We start off by finding the slow time at which teh SAR see the selected pixel, therefore the angle of the target with respect to the positions of the SAR is computed. If the object is at a bigger angle than the beamwidth , we can consider that it is not seen by the plane.

Figure 15: Is the target seen by the SAR position.

The next step is to find all the pixels seen by the SAR in each individual slow time. By basing our calculation on the geometry we can easily extract the distance and azimuth that the plane observe at each slow time. Because we know the size of the pixel in the azimuth axis (Y) we can compute how many pixels the radar can pickup at each specific slow time but also the pixel that change in each consecutive slow time. We end up storing the totality of pixels seen by the radar along its trajectory.

Figure 16: Determining which pixels from the grid are seen by the SAR

From that we can reevaluate the matrix of distance by only containing the pixels that are actually seen by the SAR. This matrix is a sub matrix of our original distance matrix. Now that that new matrix contains only the pixels of interest, we can proceed with the TDBP, once again in each slow time and only for the pixels seen by the SAR at that slow time moment we interpolate the ranged compressed signal ³ onto a matrix containing all the distance between the radar and the pixels. We then proceed to do that for each slow time positions and sum the contribution on each pixel respectively.

 $^{^3}$ because after doing the FFT and making the transformation to a distance axis , this range compressed data has a X axis corresponding to a distance, if a specific pixel is in between two represented "x"axis distances , we need to interpolate between those two indices

```
## we distance matrix according to the rows

## temp_distance = zeros(1, sensor_num); % initialise matrix

## temp_distance = zeros(1, sensor_num); % initialise matrix

## temp_distance | zeros(pixel_size_x*pixel_size_y, sensor_num); % matrix of projecting ( size of the grid on the ground)

## teach row is pixel, each column is a distance to an atnenna

## counter = 1;

## to n = 1: 1: pixel_size_x

## to n = 1: 1: sensor_num

## to n = 1: 1: sensor_num

## to n = 1: 1: sensor_num

## temp_distance(1,k) = norm(pixel_position - plane_position(:,k)); % calculate distance between pixel and antenna

## distance_l(counter,:) = temp_distance; % each row is pixel, each column is a distance to an atnenna

## counter = counter + 1: % stops at teh end of the for loop

## end

## end

## end
```

Figure 17: Reevaluating the distance matrix.

Figure 18: Appliying the TDBP but only considering the "seen" pixels.

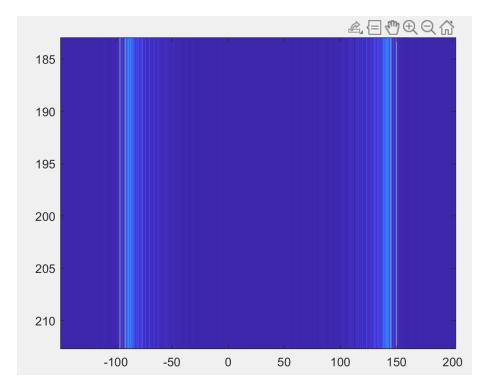


Figure 19: Range compressed pulse using imagesc.