

# 3D-Space Mapping via UAV-based RF sensing

P r o j e c t   R e a d i n e s s  
P r e s e n t a t i o n

12/08/2020

Team 4

Customer Mentor/Technical Manager:  
Dr. Honggang Zhang

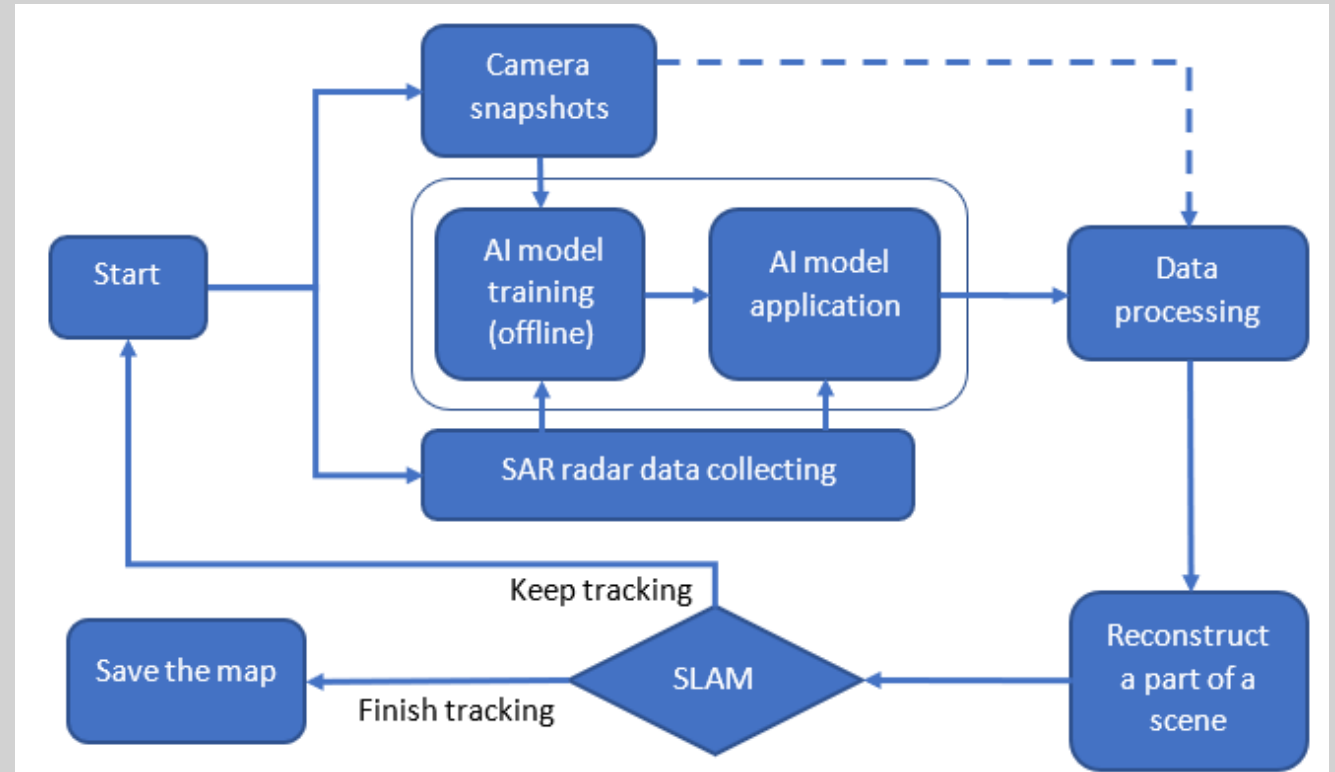
Zhuoming Huang  
Alinson Sanquintin (Junior)

# Team Roles

- SAR Experiment team:
  - Zhuoming Huang
    - SAR experiments, synthesize radar data for model training
  - Alinson Sanquintin
    - Assist with experiments and data processing in MATLAB, reviewing.
- Neural Network team (external):
  - Yue Sun and Lucas Lomba
    - Building Neural Network to train AI models and generate high-resolution images/depth images.

# Initial Design

- Use 64x8 virtual antennas to form a SAR (Synthetic-aperture Radar) structure, collect raw mmWave radar data, and generate raw intensity images
- Use depth images from a 3D camera as ground truth to train neural network models
- The neural network generates high-resolution intensity images or depth images from low-resolution raw images



# Design Improvement

- Use 2 sliders to form a SAR (Synthetic-aperture Radar) structure
  - In addition to 64 vertical positions with 8 virtual antennas in horizontal direction, move the radar sensor horizontally as well to improve the azimuth resolution.
  - Virtual antennas: 64x8 -> 64x24
    - -> 64x64 (need more time)
  - Azimuth resolution: 15° -> 5°
- Use CAD models to synthesize radar data and intensity images to train a deep neural network model.
  - We need huge number of data samples ( $\geq 3000$ ) for neural network training. But it needs significant time to construct a 64x24 SAR in real experiments (at least 1 hour).
  - Speed up data sample collection
  - Less than 1 minute for synthesizing one data sample
    - (not including time for generating CAD models)
  - 3000 samples: 3000hrs -> 50hrs

# Parts, Tool, Equipment, and Software Availability

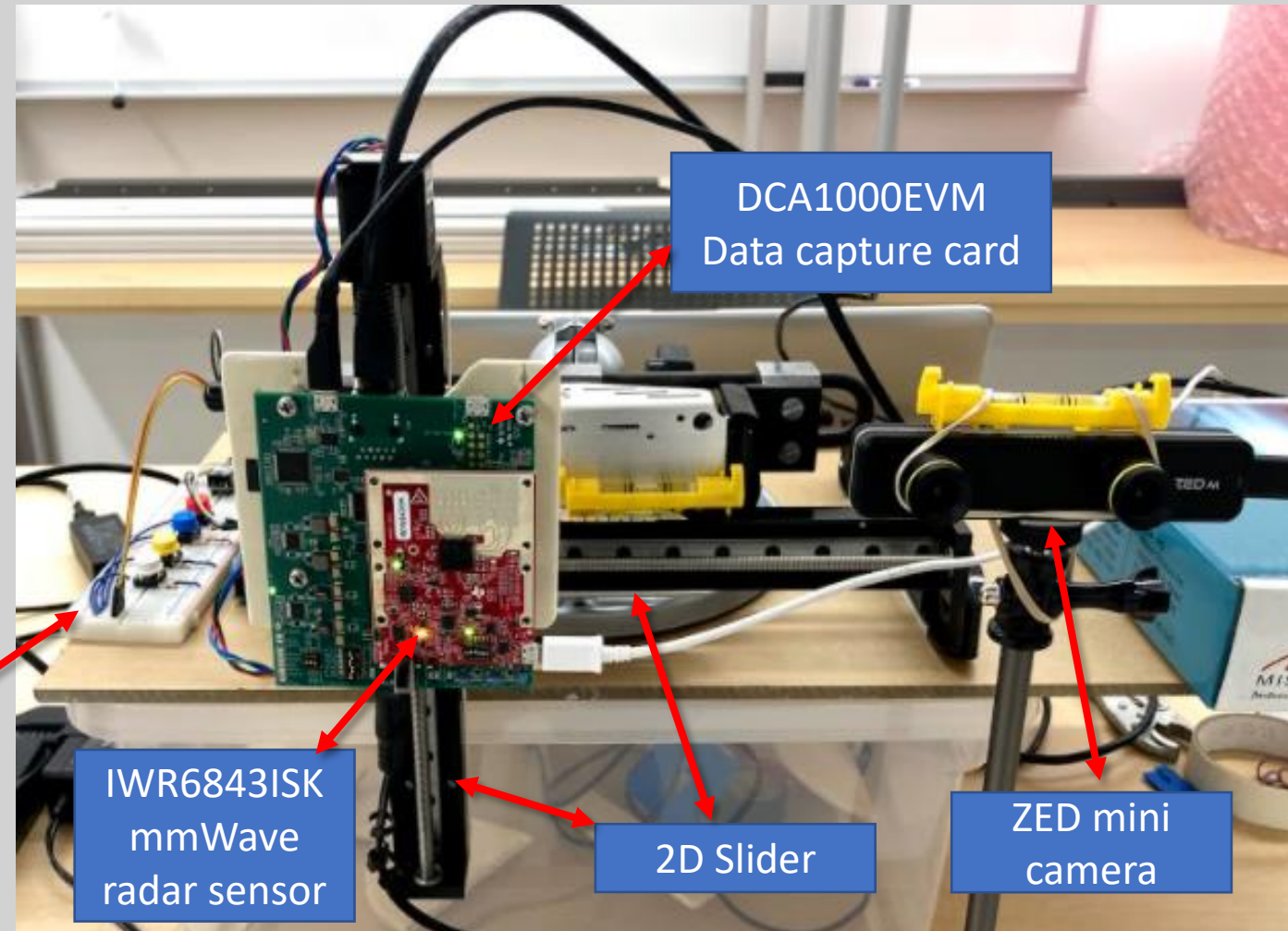
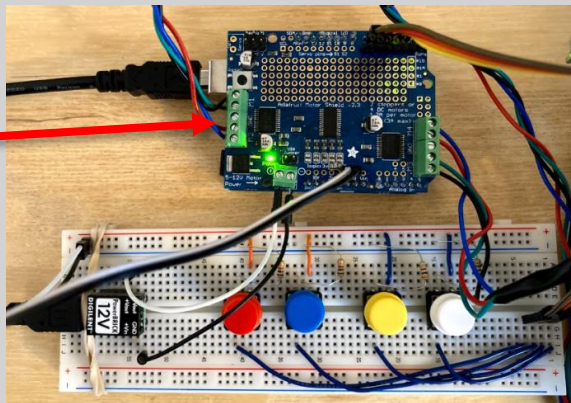
- TI IWR6843ISK mmWave radar sensor & Data capture card DCA1000EVM
  - Collect raw ADC radar data
- ZED mini 3D camera
  - Captures depth images as ground truth
- 2D Ball Screw Slider driven by Arduino
  - Moves the radar sensor to form a synthetic-aperture radar structure
- TI mmWave Studio
  - Configures and triggers radar sensor to collect data
- MATLAB
  - Process radar sensor data
  - Radar Data Synthesizer [1]
    - Synthesizes radar data from CAD models
  - Generates intensity images/radar heatmaps
- Python
  - Trains Neural Network models

# Subsystems: Hardware

- Sliders
  - Repeated positioning accuracy of  $\pm 0.05\text{mm}$
  - Allow us to accurately move the radar by half of wavelength  $\lambda/2 = 2.5\text{mm}$ , to form a SAR.

Arduino Uno with Adafruit motor shield

Slider controller



DCA1000EVM  
Data capture card

IWR6843ISK  
mmWave  
radar sensor

2D Slider

ZED mini  
camera

# Subsystem: mmWave Studio

The screenshot displays the mmWave Studio 2.1.1.0 software interface. The main window is divided into several sections:

- Sensor Configuration:** Includes fields for Profile Id (0), Start Freq (60.000000 GHz), Frequency Slope (29.982 MHz/us), Idle Time (100.00 us), TX Start Time (0.00 us), ADC Start Time (6.00 us), ADC Samples (256), Sample Rate (2047 kpsps), Ramp End Time (133.00 us), RX Gain (30 dB), RF Gain Target (30dB), VCO Select (VCO2), and Calib LUT Update options.
- Chirp Configuration:** Includes Profile Id (0), Frequency Slope Var (0.000 MHz/us), Start Chirp for Cfg (2), End Chirp for Cfg (2), Start Freq Var (0.000000 MHz), and TX Enable for current chirp options (TX0, TX1, TX2).
- Frame Configuration:** Includes Start Chirp TX (0), End Chirp TX (2), No of Frames (1), Trigger Select (SoftwareTrigger), and Duty Cycle (49.7%).
- Waveform Diagram:** A central plot showing the chirp cycle time, including Ramp Start, Start ADC Sampling, ADC Sampling Time, Ramp End, and Transmitter is ON.
- Output Log:** Displays a series of messages from the RadarAPI, including status updates, version information, and configuration commands.

Configure mmWave radar sensor:

- Start frequency
  - 60GHz
- Sweep slope
  - 29.982MHz/us
- Sample rate
  - 2047kpsps
- Number of samples
  - 256
- Bandwidth
  - $3987.61\text{MHz} \cong 4\text{GHz}$
- Transmitters and receivers
  - 2Tx + 4Rx

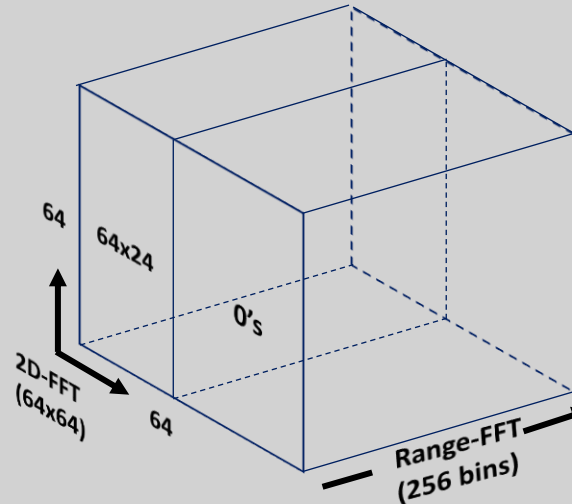
Trigger frames to send and capture radar signal



# Subsystem: MATLAB Program

## HeatmapGenerator.m

1. Apply range-FFT along 256 samples from each virtual antenna
2. Fit data from each range bin (64x24) into a 64x64 matrix, with zero-padding
3. Apply 2D-FFT on each 64x64 matrix
4. Generate heatmaps in each range bin



(screenshots of parts of the program)

```
for range_idx = 1:numADCSamples
    fft_data(:,1:NTX) = rf_data(:,range_idx,:);
    all_fft = fft2(fft_data);
    all_fft = fftshift(all_fft);
    all_fft = flip(flip(all_fft,2));
    temp = abs(all_fft);
    f = figure('visible', 'off');
    heatmap(temp);
    grid off
    colorbar;
    colormap default
    caxis([0 round(m/0.8)]);
    svaddr = '..\intensity_figs\el_az_rg_';
    saveas(f, strcat(svaddr,num2str(range_idx)), 'png')
    disp(range_idx)
end
```

## RadarDataSynthesizer.m [1]

1. Simulate reflector from CAD models of objects
2. Remove occlusion
3. Simulate radar signal
4. Generate radar heatmaps

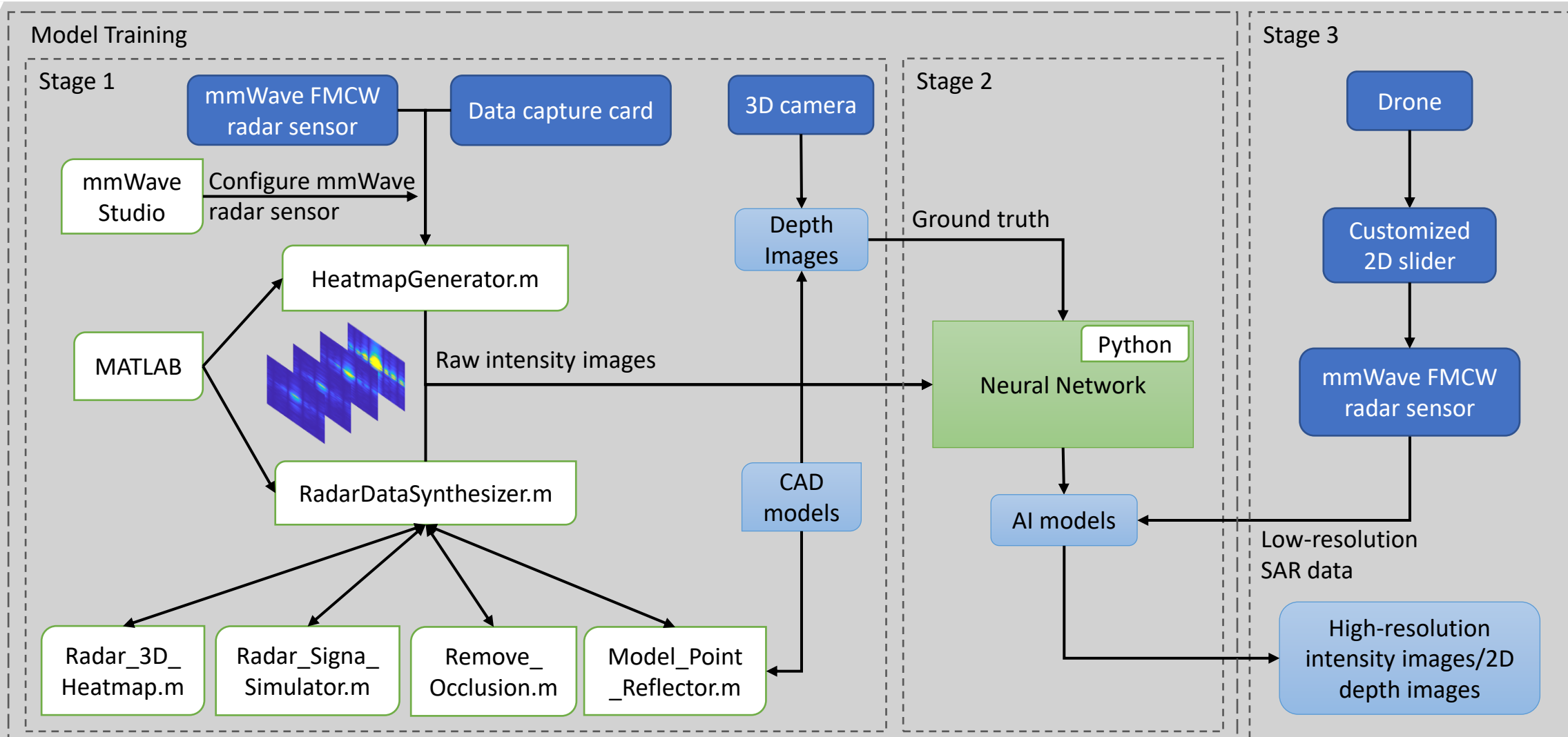
```
148
149
150 % Visualize the radar heatmap top view
151 radar_heatmap_top = squeeze(max(radar_heatmap,[],3));
152 f3 = figure('visible', sp);
153 imagesc(radar_heatmap_top);
154 set(gca,'XDir','reverse');
155 set(gca,'YDir','normal');
156 colormap jet; caxis([0 1e11]);
157 colorbar;
158 %xlabel('Range'); ylabel('Azimuth');
159 xlabel('Azimuth'); ylabel('Range'); % 20201129
160 set(gca,'FontSize',16) % Creates an axes and sets its FontSize to 18
161 svaddr = 'F:\3_Education\UMASS\Courses\droneSLAM\mmWave\hawkeye_synthesizer\figs\md_';
162 saveas(f3, strcat(svaddr,num2str(CAD_idx),"_pm_",num2str(ks),"3RadarHeatmapTopView"), 'png');
163
164 % Visualize the radar heatmap front view
165 radar_heatmap_front = squeeze(max(radar_heatmap,[],1));
166 f4 = figure('visible', sp);
167 imagesc(radar_heatmap_front);
168 set(gca,'XDir','reverse');
169 colormap jet; caxis([0 1e11]);
170 colorbar;
171 %xlabel('Azimuth'); ylabel('Elevation');
172 xlabel('Azimuth'); ylabel('Elevation'); % 20201129
173 set(gca,'FontSize',16) % Creates an axes and sets its FontSize to 18
174 svaddr = 'F:\3_Education\UMASS\Courses\droneSLAM\mmWave\hawkeye_synthesizer\figs\md_';
175 saveas(f4, strcat(svaddr,num2str(CAD_idx),"_pm_",num2str(ks),"4RadarHeatmapFrontView"), 'png');
```



# Updated Preliminary Budget

No.	Item	Qty	Unit Price (USD)	Cost (USD)	Description	Note
1	IWR6843ISK	1	135	135	mmWave radar sensor	Previously owned
2	ZED mini	1	399	399	3D camera	Previously owned
3	3D-printed connector	1	N/A	10	Holding radar and camera, extend to a drone	Plan to order
4	T6x1-200mm Ball Screw Slider	2	70	140	For moving the radar sensor horizontally/vertically in SAR experiments only	Recently purchased
5	Arduino Uno	1	23	23	Slider control	Previously owned
6	Adafruit motor shield	1	19.95	19.95	Motor driver	Previously owned
7	DCA1000EVM	1	499	499	Raw data capture card, for experiments only	Previously owned
8	Customized 2D slider	1	N/A	25	For forming a SAR on a drone	Plan to order
	Total			1250.95		
	Used budget			140	Out of the \$1000 budget	
	Final product estimated budget			569		

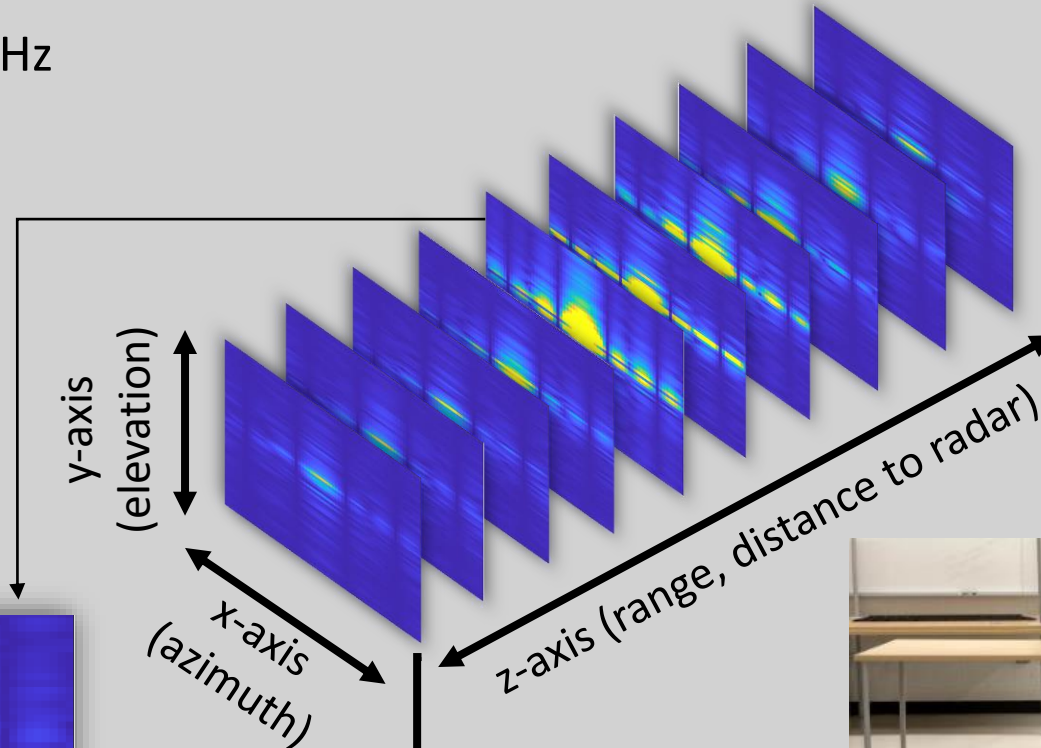
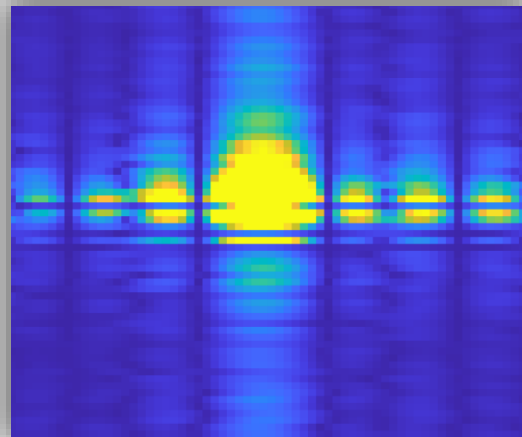
# System Architecture



# Preliminary Data: Heatmaps from Real Experiments

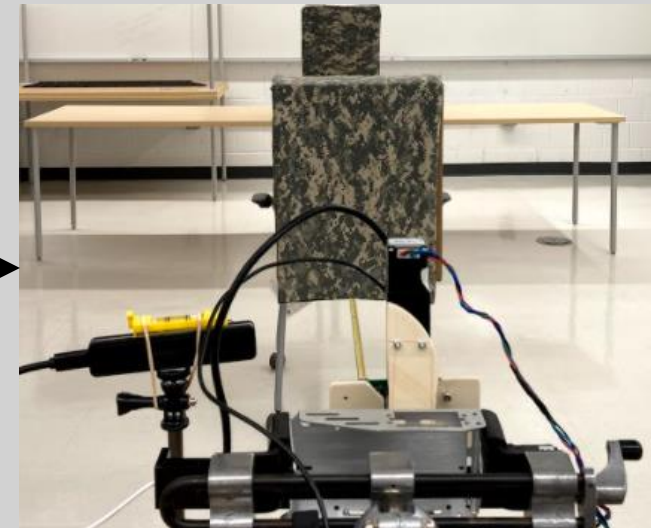
- Operating frequency: 60GHz-64GHz
- Number of samples: 256
- Size of virtual antennas: 64 x 24
- Antenna spacing: 2.5mm
- Range resolution: 4cm
- Angular resolution:
  - Azimuth = 5°, Elevation = 1.8°

Range bin: 51  
Corresponding  
to distance:  
 $(51-1) \times 4\text{cm} = 200\text{cm}$



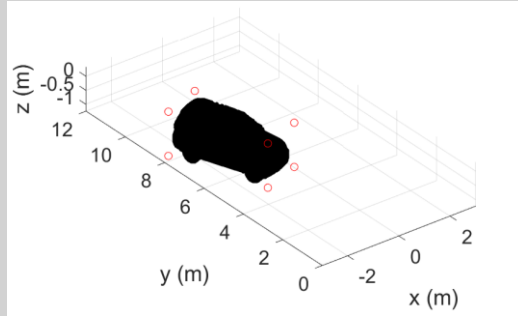
Total 256  
intensity images,  
only 10 are  
shown here.  
(range bins 47-56)

3D Radar intensity  
version of the scene

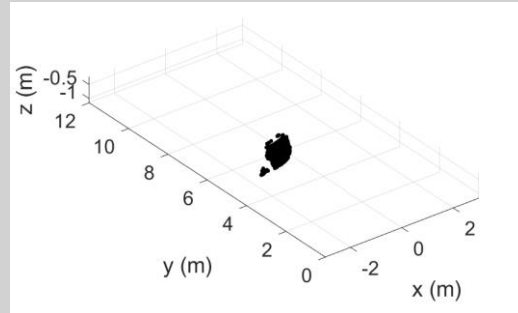


# Preliminary Data: Radar Data Synthesizer <sup>[1]</sup>

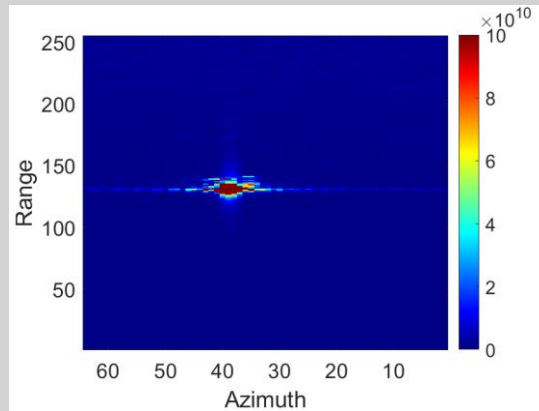
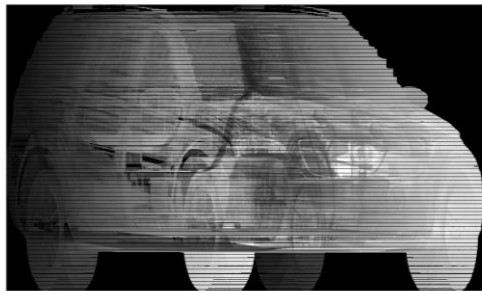
(1) CAD model



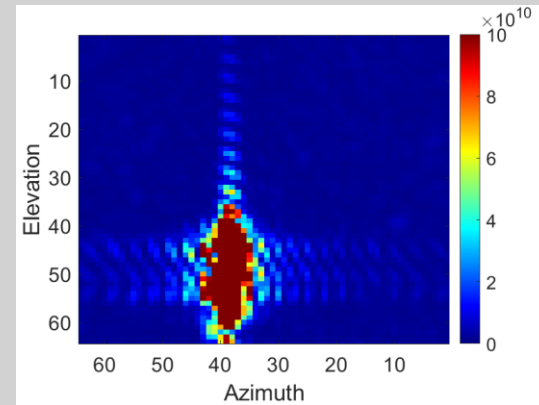
(2) Reflector



(3) Depth image

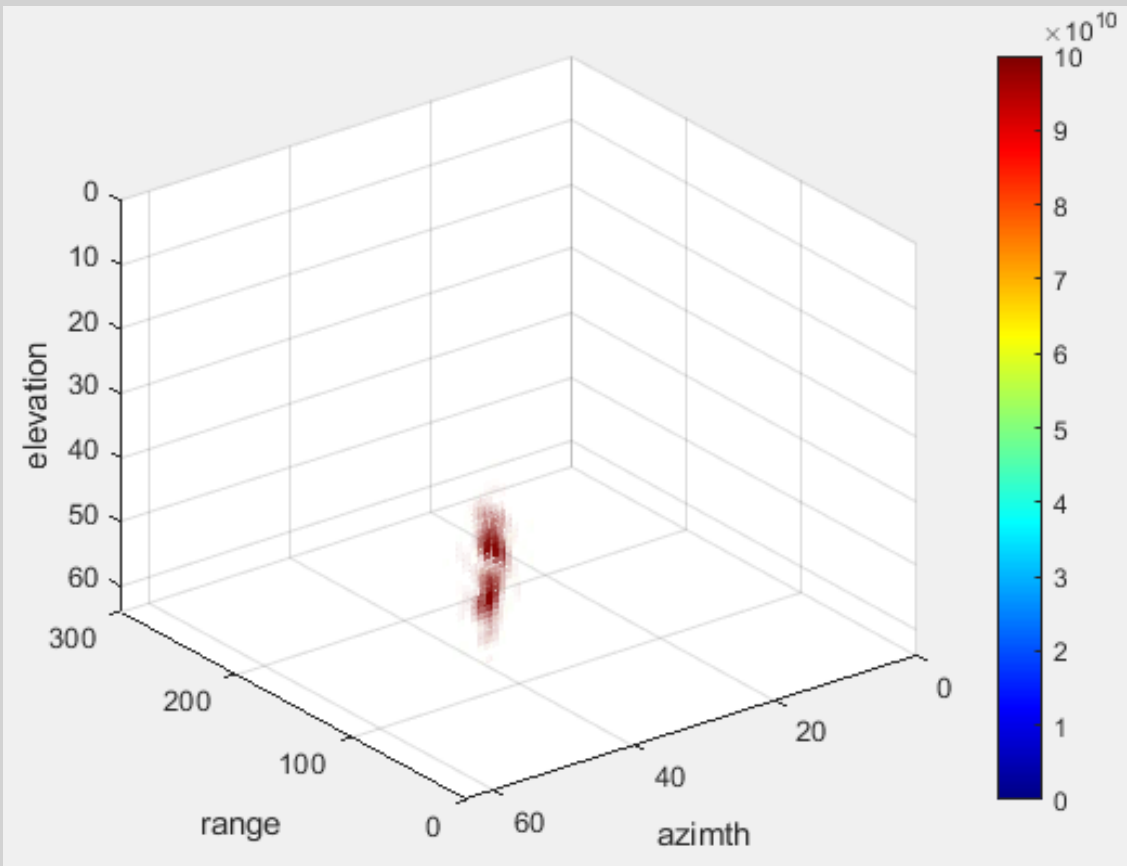


(4) Top view



(5) Front view

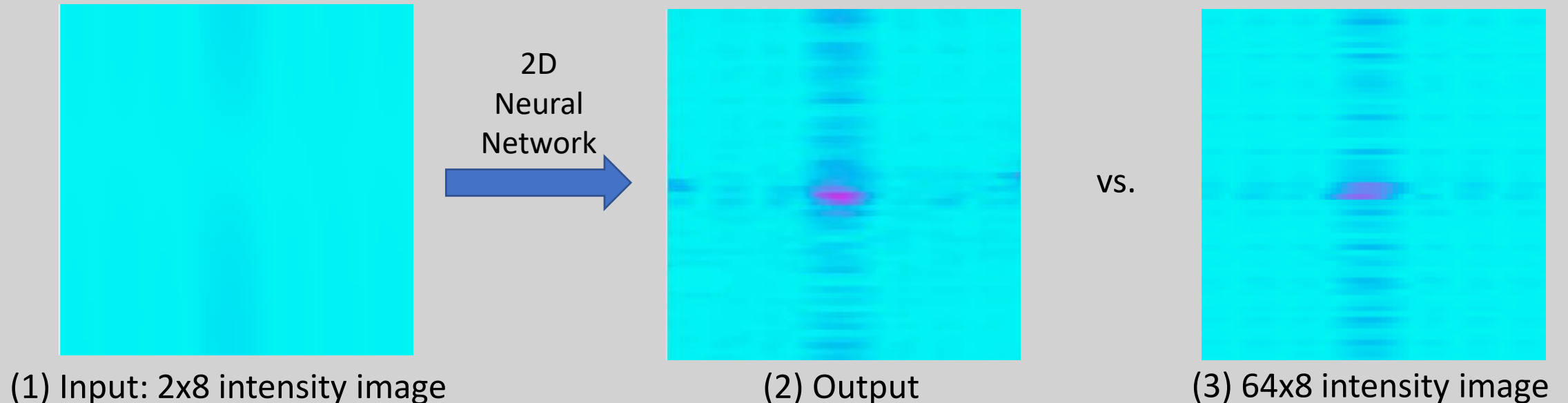
Synthesized radar data



(6) 3D point cloud

# Preliminary Data: Neural Network

- Low-resolution input -> high-resolution output
- 2D-to-2D, pix2pix GAN (Generative Adversarial Network)
- Using only 2x8 virtual antennas to achieve similar resolution from 64x8 virtual antennas.
- Example of our NN producing output from data provided by SuperRF [2]



# Metrics and Performance Measurement (preview of DVP)

Requirement	Measurement
• <b>Capability to outline shapes of objects in a scene</b>	Observe that the shapes of objects are successfully outlined in the 2D depth image output after processing raw radar data.
• <b>Accurately estimates positions of objects in 3D space</b>	Compare our result to depth images (ground truth) given by ZED mini camera.
• <b>Portable (small and lightweight)</b>	Measure physical dimension and weight ( $\approx 100\text{g}$ )
• <b>Mounted on a drone (connector)</b>	Confirm that a connector is given, and all hardware components can be secured on a drone through this connector.
• <b>Low-cost</b>	Calculate the budget for final product ( $\leq \$1000$ )
• <b>Easy to use</b>	Count the steps for configuration and operation ( $\leq 10$ )

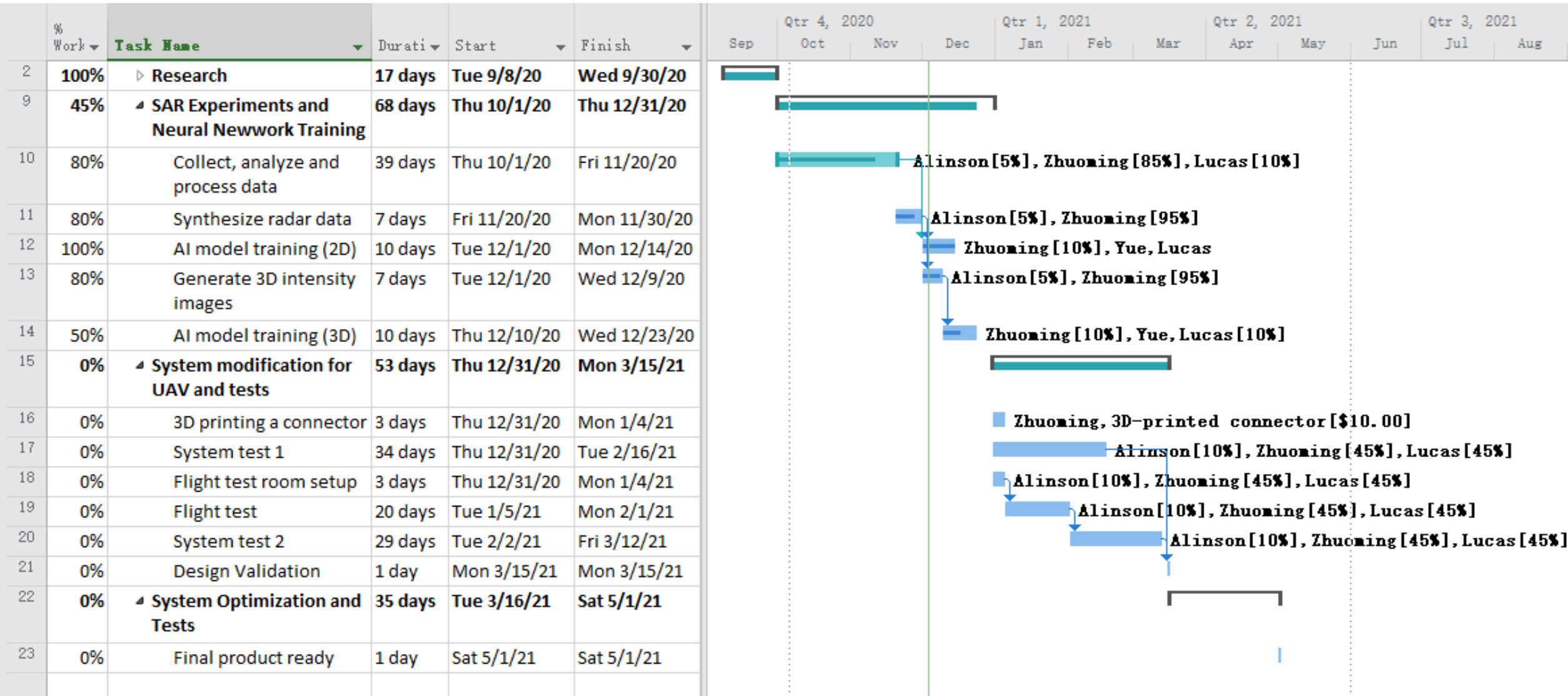
# Deliverables

- A 3D-space mapping system
  - Generates high-resolution depth images from sparse raw mmWave radar data
  - Model-training function
    - MATLAB program for generating raw intensity images
    - Neural network for training AI Models
  - 2D slider with control code
  - Modified radar sensor firmware
  - A holder connecting the radar sensor and camera, mounting the system on a drone



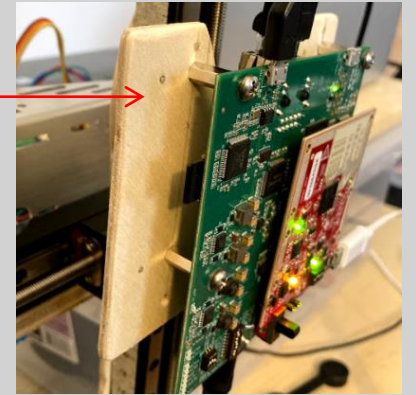
# Project Management (till Design Validation Presentation)

GANTT CHART



# Anticipated Problems and Solutions

- Access to 3D printers on campus
  - Temporarily using handcrafted wooden connectors
  - Order 3D-printing online or handcrafting with acrylic board.
- UAV flight tests
  - Safety issue
    - Wear eye protectors and gloves
  - Flight-test room availability
    - Outdoor public facility with net-covering
  - Hardware availability
    - Use another ZED mini (not the one on UAV) to capture depth images as ground truth for scene reference if Kinect or OptiTrack system are not available.



# Ethical Considerations

- Privacy issue when conducting outdoor experiments
  - Must ask for a permission if private properties are involved
  - Keep the camera off in public areas if it is not necessary to turn it on, or if someone doesn't feel comfortable in front of it.
- Covid-19 Guidelines
  - Masks on
  - Keep social distance

# Thank you

- Questions?
- Reference
  - [1] Junfeng Guan, Sohrab Madani, Suraj Jog, and Haitham Hassanieh. 2020. High Resolution Millimeter Wave Imaging For Self-Driving Cars. IEEE CVPR (2020).
  - [2] Shiwei Fang and Shahriar Nirjon. 2020. SuperRF: Enhanced 3D RF Representation Using Stationary Low-Cost mmWave Radar. In International Conference on Embedded Wireless Systems and Networks (EWSN)..., Vol. 2020. NIH Public Access, 120.