

Re-entry Analysis from Serendipitous Radar data (RASR)

Technical Presentation

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Table of Contents

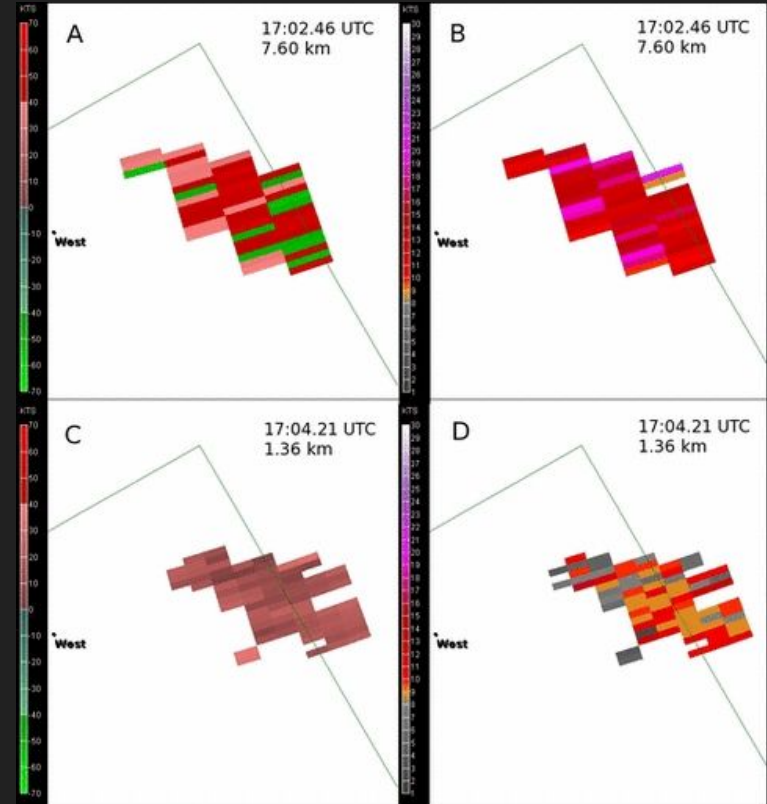
- Glossary
- Background
- Purpose
- Setup
- First Attempt
- Current Solution
- Testing
- Output
- References

Glossary

- TACC - Texas Advanced Computing Center, home to Stampede2 supercomputer and Corral file storage server
- NEXRAD - Next Generation Weather Radar
- NOAA - National Oceanic and Atmospheric Association
- CNN - Convolutional Neural Network
- ASTRIA - Advanced Science and Technology Research in Astronautics, an astrodynamics research program led by Dr. Moriba Jah
- ARES - Astromaterials Research and Exploration Science

Background

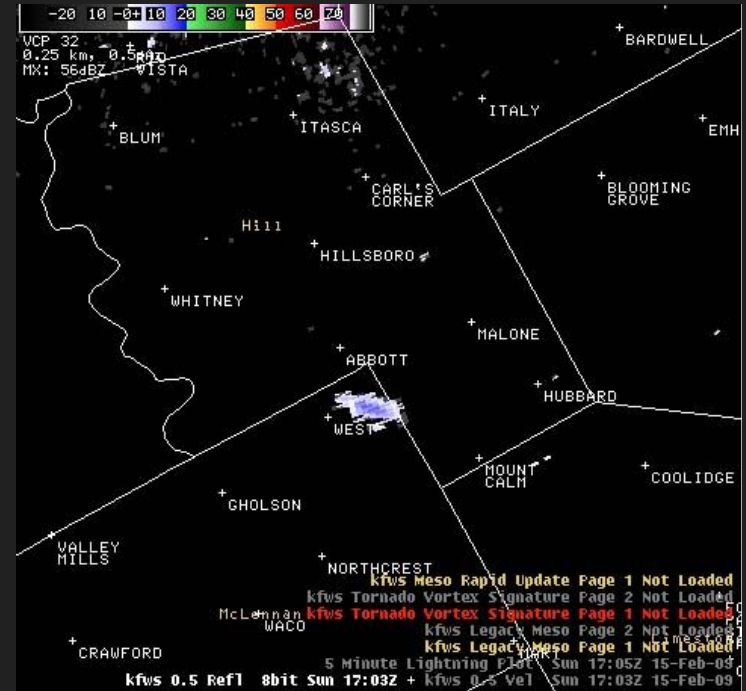
- Re-entering objects become lost quite frequently
- Most US Govt. radar (Military) loses objects somewhere between 200 km and ground
- NEXRAD weather data can be co-opted to catch meteors, as shown by Fries [1]
- NASA ARES keeps a manually updated database of visual sightings with corresponding velocity signature



Radar signature examples at two sweeps
in West, TX, from Fries [1]

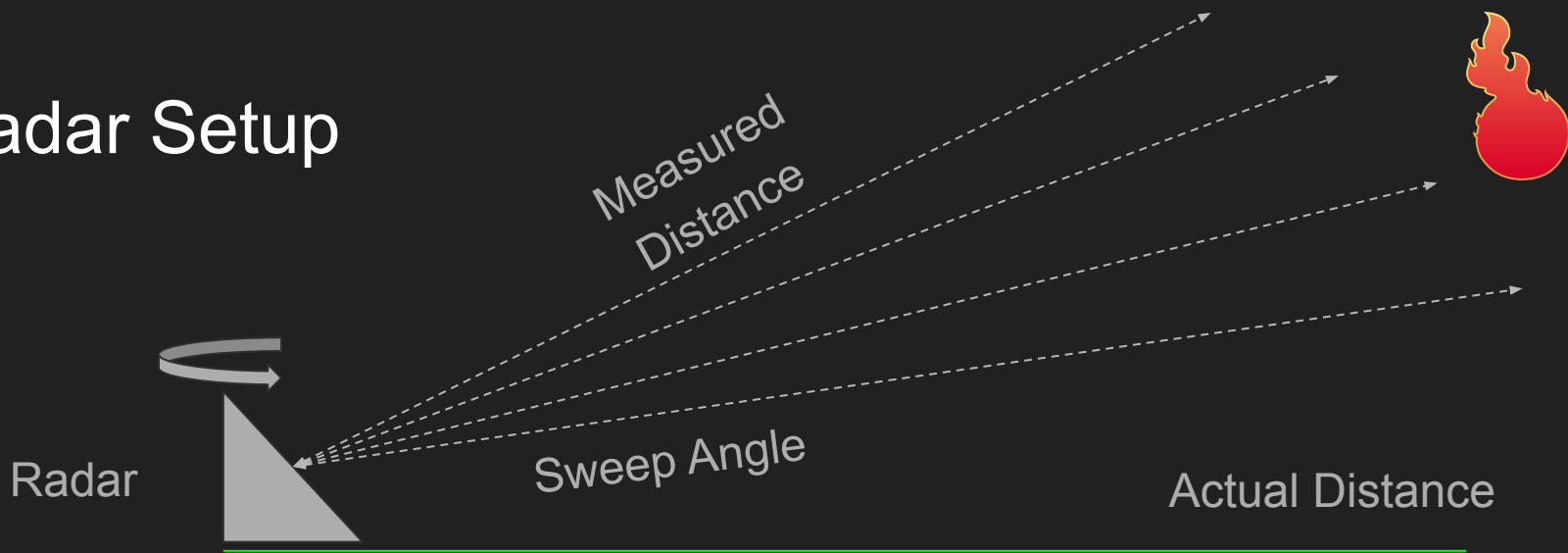
Purpose

1. Automate the process of detecting and locating these falls
2. Determine if the object is true meteor or orbital debris
3. Kinematically back-calculate trajectory to determine correlation with known orbital bodies



Larger example of same detection, from
Fries [1]

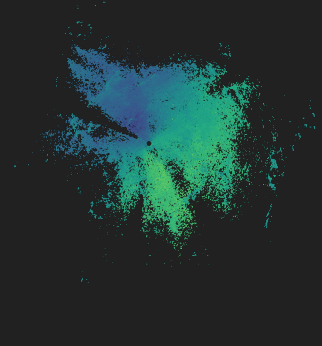
Radar Setup



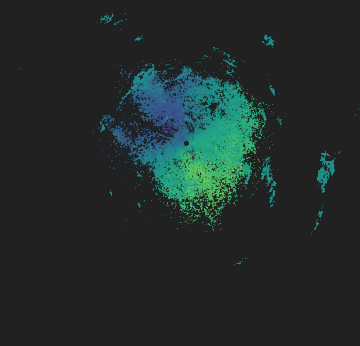
Sweep 1



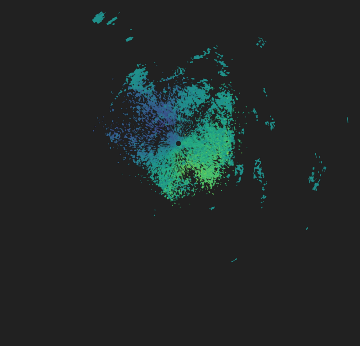
Sweep 2



Sweep 3



Sweep 4



First Attempt - Automation

1. Developed Python NEXRAD website parser with Graduate student
2. Data files unpacked with Py-ART, NOAA's open source Python package
3. OpenCV package used to process sweeps with colormaps
4. Computations parallelized using Pool

```
def getMaps():  
    #Set velocity colormap  
    cmaplist = np.array([[255,0,0],#~70  
                          [255,0,0],#~60  
                          [210,0,0],#~50  
                          [180,0,0],#~40  
                          [150,0,0],#~30  
                          [115,0,0],#~20  
                          [70,40,40],#~10  
                          [70,70,70],#0  
                          [40,70,40],#10  
                          [0,115,0],#20  
                          [0,150,0],#30  
                          [0,180,0],#40  
                          [0,210,0],#50  
                          [0,255,0],#60  
                          [0,255,0]])/255#70  
  
    cmaplist = np.flip(cmaplist,axis=0)  
    velMap = color.LinearSegmentedColormap.from_list('velMap',cmaplist,N=256)  
  
    #Set spectrum width colormap  
    cmaplist = np.array([[20,20,20],#0  
                          [40,40,40],#6  
                          [40,150,40],#12 40 150  
                          [150,40,40],#18 150 40  
                          [150,70,0],#24  
                          [255,255,0]])/255#30  
  
    spwMap = color.LinearSegmentedColormap.from_list('spwMap',cmaplist,N=256)  
    return velMap, spwMap  
  
def getData(filename, velMap, spwMap, writeImgs):  
    cutoff = 100  
    edgeFilter = 8#12  
    areaFraction = 1*(10**-4)  
    circRatio = 0.3  
    fillFilt = 30  
    colorIntensity = cutoff  
    date=filename
```

OpenCV colormaps from our code

First Attempt - Issues

- Color maps and tuned filters were specific to one example fall
 - Hard to generalize to all examples
 - **Solution:** More flexible algorithm for image processing
-
- Memory issues with Matplotlib, TACC resources inefficiently allocated
 - **Solution:** Bypass Matplotlib or fix memory issue

Current Solution

- Decided to move to convolutional neural networks, chose Tensorflow Keras
 - Computationally efficient and complex-filtering capable
- Fixed Matplotlib issues with figure management
- Inherited much of the I/O functionality from first iteration

```
This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (on
eDNN)to use the following CPU instructions in performance-critical operations:
AVX2 FMA
To enable them in other operations, rebuild TensorFlow with the appropriate comp
iler flags.
2020-10-07 15:48:23.822981: I tensorflow/core/platform/profile_utils/cpu_utils.c
c:104] CPU Frequency: 1999965000 Hz
2020-10-07 15:48:23.823418: I tensorflow/compiler/xla/service/service.cc:168] XL
A service 0x563a8abdc4e0 initialized for platform Host (this does not guarantee
that XLA will be used). Devices:
2020-10-07 15:48:23.823440: I tensorflow/compiler/xla/service/service.cc:176]
StreamExecutor device (0): Host, Default Version
Checking KCBX at 07/18/2020 23:53:05
```

Current Solution - CNN

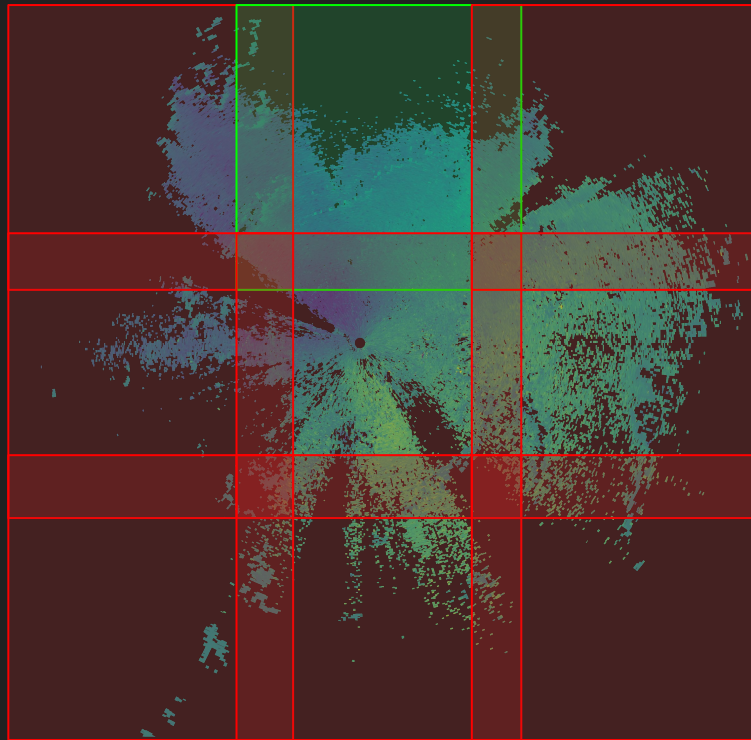
- 8 layer CNN
 - 3 Convolution
 - 2 Pooling
 - 1 Flattening
 - 2 Dense
- Designed with default structure
- Next steps:
 - Sensitivity analysis
 - Optimization

```
#####  
  
#start_time = time.time()  
thresh = 0.98  
nd = 10  
dim = 2500  
h = int(dim/nd)  
cpath = os.getcwd()  
dirname = cpath + '/data/'  
  
model = models.Sequential()  
model.add(layers.Conv2D(h, (4, 4), activation='relu', input_shape=(h, h, 4)))  
model.add(layers.MaxPooling2D((2, 2)))  
model.add(layers.Conv2D(500, (4, 4), activation='relu'))  
model.add(layers.MaxPooling2D((2, 2)))  
model.add(layers.Conv2D(64, (3, 3), activation='relu'))  
model.add(layers.Flatten())  
model.add(layers.Dense(32, activation='relu'))  
model.add(layers.Dense(1, activation = "sigmoid"))  
#model.load_weights('rasrmdl')
```

CNN model setup

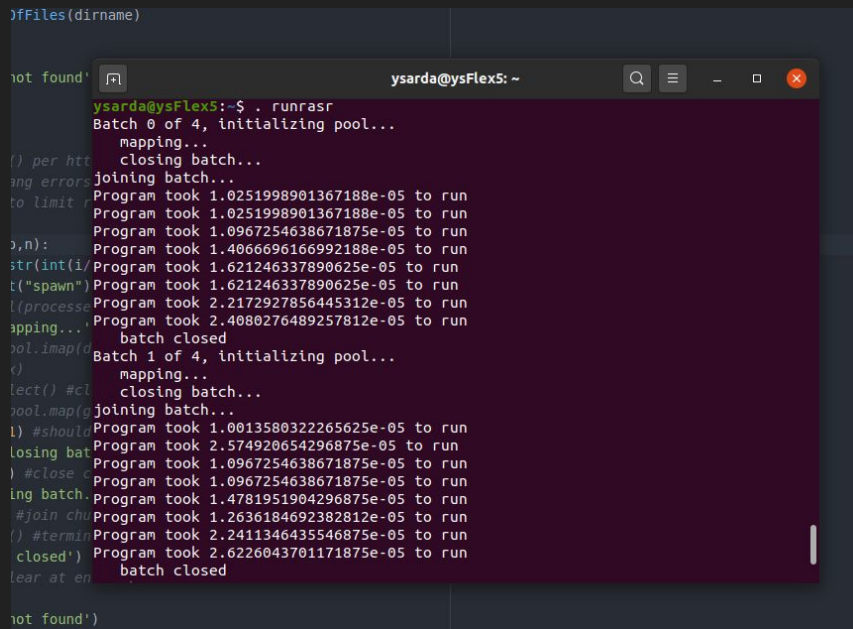
Testing

- 1 sweep is 2500x2500x4 RGBY image
- Batch testing locally: Slow
 - Laptop is 16 GB RAM, 8 core
 - 800% core usage for serial use
- Developed algorithm to split images into overlapping pieces, retaining edge information
- Based on CPU available, image can split into n^2 pieces
- Unfortunately, has to be re-trained for each case



Testing

- CNN is run through sigmoid function to normalize and discretize output
- Sifts site by site, sweep by sweep
- Training is simple, image matched with either 1 or 0 from ARES database, and vectorized
- Parallel training on TACC

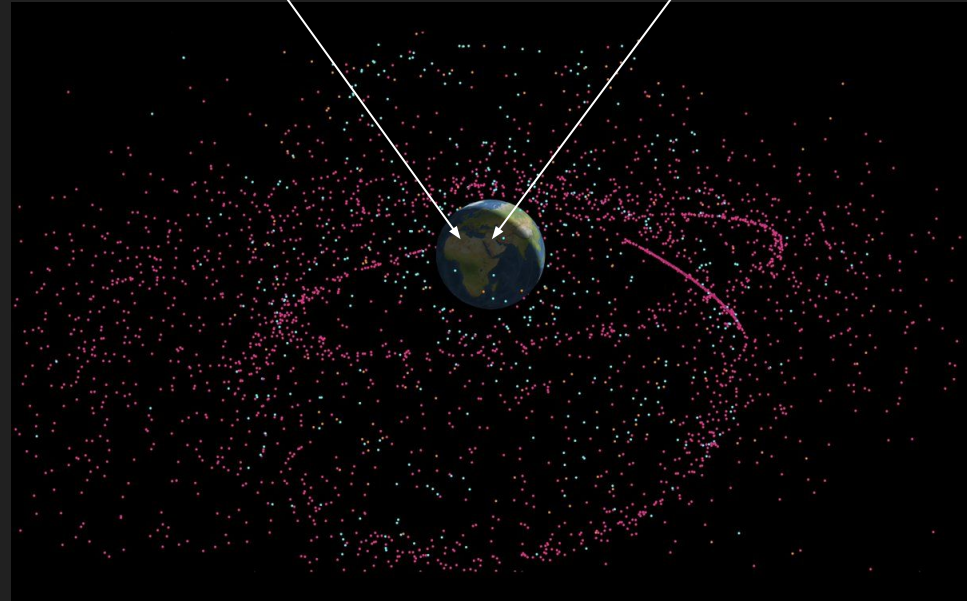
A screenshot of a terminal window titled 'ysarda@ysFlex5: ~'. The terminal shows the execution of a script named 'runrasr'. The output indicates that it is running 'Batch 0 of 4, initializing pool...' and 'mapping...'. It then shows 'closing batch...' and 'joining batch...'. A list of 16 programs is shown, each with a timestamp indicating when it started running. The batch is then closed, and 'Batch 1 of 4, initializing pool...' begins. The terminal window has a dark background with light-colored text. The window title bar shows standard Linux window controls (minimize, maximize, close) and the user's name and host.

```
ysarda@ysFlex5: ~  
ysarda@ysFlex5:~$ . runrasr  
Batch 0 of 4, initializing pool...  
mapping...  
closing batch...  
joining batch...  
Program took 1.0251998901367188e-05 to run  
Program took 1.0251998901367188e-05 to run  
Program took 1.0967254638671875e-05 to run  
Program took 1.4066696166992188e-05 to run  
Program took 1.621246337890625e-05 to run  
Program took 1.621246337890625e-05 to run  
Program took 2.2172927856445312e-05 to run  
Program took 2.4080276489257812e-05 to run  
batch closed  
Batch 1 of 4, initializing pool...  
mapping...  
closing batch...  
joining batch...  
Program took 1.0013580322265625e-05 to run  
Program took 2.574920654296875e-05 to run  
Program took 1.0967254638671875e-05 to run  
Program took 1.0967254638671875e-05 to run  
Program took 1.4781951904296875e-05 to run  
Program took 1.2636184692382812e-05 to run  
Program took 2.2411346435546875e-05 to run  
Program took 2.6226043701171875e-05 to run  
batch closed  
not found'
```

Batch testing screenshot

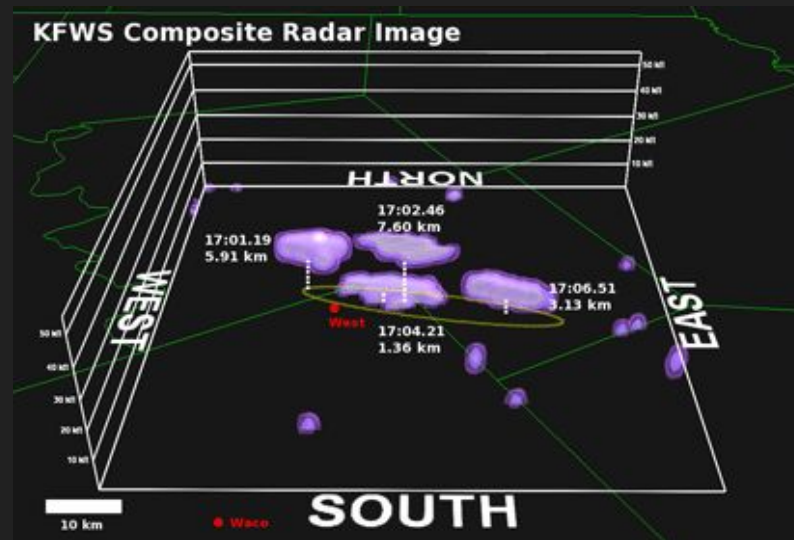
Output

- After detection, relevant data is packaged into a json file
- To be displayed on [ASTRIAGraph](#)
- All code, data, and algorithms are publicly available



Next Evolution

- Moving from 2D sweeps to 3D datacubes, for fidelity and use in kinematics
- This means input: 3D to 4D
- Currently, CNN outputs binary
- OD software being written to integrate into CNN
- Will require more extensive labeling, but allow for centroids



Example concept datacube from Fries [1]

References

Thanks to Dr. Moriba Jah, head of ASTRIA and PI of this project, as well as Ben Miller and Robby Keh.

Shoutout to the Tensorflow tutorials

[1] FRIES, M. and FRIES, J. (2010), Doppler weather radar as a meteorite recovery tool. *Meteoritics & Planetary Science*, 45: 1476-1487.
doi:10.1111/j.1945-5100.2010.01115.x