

Problem Definition

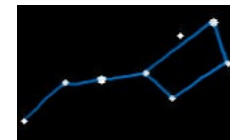
- Apparent **lack of annotated astronomical data.**
- **Applications:**
 1. Satellite positioning
 2. Telescope Detection
 3. Astronomy Research
- Proposing detection method that can be extended beyond just constellations.

Objectives

- Create generative **cGAN** model (**dots2stars**) to make realistic, synthetic astronomy data.
- Create finetuned detection model (**Star Spotter**) to validate synthetic images.
- Comparative constellation detection (procedural vs. generative dataset curation).

Final Design

#1. Constellation Generation



Star Coordinates



dots2stars



Real-Looking Image

#2. Constellation Detection



Real Images



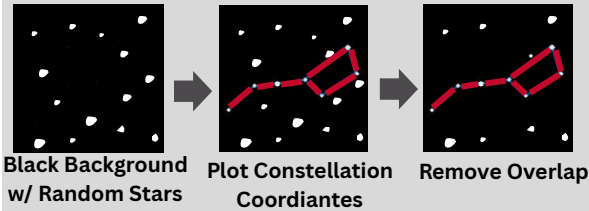
Star Spotter



Constellation Detections

Design Calculations & Analysis

Procedural Constellation Generation



Real-Images Dataset

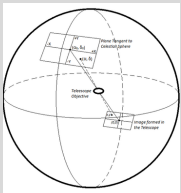
- Online-Sourced
- 1750 Img, 6130 Instances
- 14 Constellation Classes

Balancing cGAN Loss

- Generator **maximizes making realistic images.**
- Discriminator **minimizes mistaking synthetic images for real images.**
- Adjustable Parameter λ

$$\mathcal{L} = \mathcal{L}_{GAN}(G, D) + \lambda \mathcal{L}_{L_1}(G)$$

3D Coordinates to 2D Coordinates

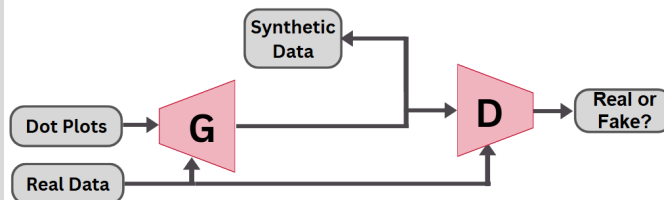


$$x = \cos(\theta) \cos(\phi)$$

$$y = \sin(\theta) \cos(\phi)$$

Conditional GAN

- **G**enerator
- **D**iscriminator
- Compares the synthetic images against real images



Prototype & Test Results

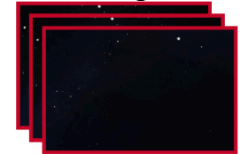
Pre-Trained Models

- YOLO26
- RF-DETR
- D-FINE

Fine-Tune on:

1. Real Data
2. Synthetic Data
3. Real + Synthetic

Evaluate on Real Testing Subset



Model	Nano		Small	
	mAP50	mAP50-95	mAP50	mAP50-95
YOLO26+Real	81.08	48.10	92.29	57.07
YOLO26+Dots	0.04	0.01	1.86	0.93
YOLO26+Real+Dots	11.7	4.46	22.09	10.09
YOLO26+GanGen	0.37	0.06	0.11	0.04
YOLO26+Real+GanGen	6.50	1.91	10.28	4.30