

AIDA 2154 – Final Project: Deep Learning-based Monitoring of Solar Panels

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Course: Computer Vision (Fall 2025)
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Project Overview

This notebook implements an **Object Detection** system using **YOLOv11** to identify thermal anomalies in photovoltaic modules. The system is trained on the *Infrared Solar Modules* dataset to detect and localize 12 distinct classes of faults (e.g., Hotspots, Diode failures) and environmental issues (Vegetation, Soiling).

Methodology

- Data Preparation:** Dynamic configuration of the dataset paths.
- Model Training:** Fine-tuning a pre-trained `yolo11n` (Nano) model.
- Evaluation:** Analyzing Mean Average Precision (mAP) and confusion matrices.
- Inference:** Demonstrating the model on unseen test images.

[1]: *# Cell 1: Environment Setup & GPU Check*
Installs dependencies if missing, and imports necessary libraries.

```
import os
import sys
import yaml
import random
import glob
import cv2
import torch
import matplotlib.pyplot as plt

try:
    import ultralytics
    from ultralytics import YOLO
except ImportError:
    print("Installing Ultralytics YOLO...")
    !pip install -q ultralytics
    import ultralytics
    from ultralytics import YOLO

# Verify installation
ultralytics.checks()

# HARDWARE CHECK
if torch.cuda.is_available():
    print(f"✅ GPU Detected: {torch.cuda.get_device_name(0)}")
    BATCH_SIZE = 16 # Standard Production Batch Size
else:
    print("⚠️ GPU NOT Detected. Running on CPU.")
    BATCH_SIZE = 4

print(f"Project Root: {os.getcwd()}")
```

Ultralytics 8.3.235 Python-3.11.14 torch-2.10.0.dev20251205+cu128 CUDA:0 (NVIDIA GeForce RTX 5070 Laptop GPU, 8151MiB)
 Setup complete (20 CPUs, 31.1 GB RAM, 362.6/927.8 GB disk)
 ✅ GPU Detected: NVIDIA GeForce RTX 5070 Laptop GPU
 Project Root: C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project

[2]: *# Cell 2: Dynamic Data Configuration*
This cell automatically locates the dataset and creates the YOLO config file.
This ensures the notebook runs on any machine without manual path editing.

```
# 1. Locate the Dataset
possible_locations = [
    os.path.join("InfraredSolarModules", "ImageSet"), # Expected Location
    "datasets/InfraredSolarModules/ImageSet",         # Alternative
    "ImageSet"                                         # Flat structure
]

dataset_root = None
for path in possible_locations:
    if os.path.exists(os.path.abspath(path)):
        dataset_root = os.path.abspath(path)
```

```

        break

if dataset_root is None:
    raise FileNotFoundError("❌ CRITICAL: Dataset not found. Please ensure 'InfraredSolarModules/ImageSet' exists.")

print(f"✅ Dataset found at: {dataset_root}")

# 2. Create the Configuration Dictionary
config = {
    'path': dataset_root,
    'train': 'train/images',
    'val': 'valid/images',
    'test': 'test/images',
    'names': {
        0: 'Cell',
        1: 'Cell-Multi',
        2: 'Cracking',
        3: 'Diode',
        4: 'Diode-Multi',
        5: 'Hot-Spot',
        6: 'Hot-Spot-Multi',
        7: 'No-Anomaly',
        8: 'Offline-Module',
        9: 'Shadowing',
        10: 'Soiling',
        11: 'Vegetation'
    }
}

# 3. Save as YAML
config_filename = 'final_solar_config.yaml'
with open(config_filename, 'w') as f:
    yaml.dump(config, f)

print(f"✅ Config file '{config_filename}' generated.")

✅ Dataset found at: C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\InfraredSolarModules\ImageSet
✅ Config file 'final_solar_config.yaml' generated.

```

```

[3]: # Cell 3: Model Training (Task 3)
# We use YOLOv11 Nano (Detection) for speed and efficiency.

# Load Model
model = YOLO('yolo11n.pt')

print(f"🔥 Starting Training Pipeline with Batch Size: {BATCH_SIZE}...")

# Train
results = model.train(
    data='final_solar_config.yaml',
    epochs=15,
    imgsz=640,          # Standard detection size
    batch=BATCH_SIZE,   # Set to 16 (Standard)
    project='Solar_Final_Runs',
    name='yolo11_detect',
    exist_ok=True,
    plots=True,

    # --- COMPATIBILITY FLAGS ---
    # These settings prevent kernel crashes on Windows/Nightly builds
    workers=0,          # Forces main-process Loading (Prevents DataLoader crashes)
    amp=False,          # Disables Mixed Precision (Prevents SegFaults on new GPUs)
    cache=False,        # Reads from disk to avoid memory spikes
)

```

🔥 Starting Training Pipeline with Batch Size: 16...

Ultralytics 8.3.235 Python-3.11.14 torch-2.10.0.dev20251205+cu128 CUDA:0 (NVIDIA GeForce RTX 5070 Laptop GPU, 8151MiB)
engine\trainer: agnostic_nms=False, amp=False, augment=False, auto_augment=randaugment, batch=16, bgr=0.0, box=7.5, cache=False, cfg=None, classes=Non
one, close_mosaic=10, cls=0.5, compile=False, conf=None, copy_paste=0.0, copy_paste_mode=flip, cos_lr=False, cutmix=0.0, data=final_solar_config.yam
l, degrees=0.0, deterministic=True, device=None, dfl=1.5, dnn=False, dropout=0.0, dynamic=False, embed=None, epochs=15, erasing=0.4, exist_ok=True, e
fliplr=0.5, flipud=0.0, format=torchscript, fraction=1.0, freeze=None, half=False, hsv_h=0.015, hsv_s=0.7, hsv_v=0.4, imgsz=640, int8=False, iou=0.
7, keras=False, kobj=1.0, line_width=None, lr=0.001, lrf=0.01, mask_ratio=4, max_det=300, mixup=0.0, mode=train, model=yolo11n.pt, momentum=0.937, m
osaic=1.0, multi_scale=False, name=yolo11_detect, nbs=64, nms=False, opset=None, optimize=False, optimizer=auto, overlap_mask=True, patience=100, pe
rspective=0.0, plots=True, pose=12.0, pretrained=True, profile=False, project=Solar_Final_Runs, rect=False, resume=False, retina_masks=False, save=T
rue, save_conf=False, save_crop=False, save_dir=C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect, save_frames=Fals
e, save_json=False, save_period=-1, save_txt=False, scale=0.5, seed=0, shear=0.0, show=False, show_boxes=True, show_conf=True, show_labels=True, sim
ply=True, single_cls=False, source=None, split=val, stream_buffer=False, task=detect, time=None, tracker=botsort.yaml, translate=0.1, val=True, ve
rbose=True, vid_stride=1, visualize=False, warmup_bias_lr=0.1, warmup_epochs=3.0, warmup_momentum=0.8, weight_decay=0.0005, workers=0, workspace=Non
e

Overriding model.yaml nc=80 with nc=12

		from	n	params	module	arguments
0	-1	1	464	ultralytics.nn.modules.conv.Conv		[3, 16, 3, 2]
1	-1	1	4672	ultralytics.nn.modules.conv.Conv		[16, 32, 3, 2]
2	-1	1	6640	ultralytics.nn.modules.block.C3k2		[32, 64, 1, False, 0.25]
3	-1	1	36992	ultralytics.nn.modules.conv.Conv		[64, 64, 3, 2]
4	-1	1	26080	ultralytics.nn.modules.block.C3k2		[64, 128, 1, False, 0.25]
5	-1	1	117712	ultralytics.nn.modules.conv.Conv		[128, 128, 3, 2]

```

5      -1 1 147712 ultralytics.nn.modules.conv.Conv [128, 128, 3, 2]
6      -1 1 87040 ultralytics.nn.modules.block.C3k2 [128, 128, 1, True]
7      -1 1 295424 ultralytics.nn.modules.conv.Conv [128, 256, 3, 2]
8      -1 1 346112 ultralytics.nn.modules.block.C3k2 [256, 256, 1, True]
9      -1 1 164608 ultralytics.nn.modules.block.SPPF [256, 256, 5]
10     -1 1 249728 ultralytics.nn.modules.block.C2PSA [256, 256, 1]
11      0 1 0 torch.nn.modules.upsampling.Upsample [None, 2, 'nearest']
12     [-1, 6] 1 0 ultralytics.nn.modules.conv.Concat [1]
13     -1 1 111296 ultralytics.nn.modules.block.C3k2 [384, 128, 1, False]
14     -1 1 0 torch.nn.modules.upsampling.Upsample [None, 2, 'nearest']
15     [-1, 4] 1 0 ultralytics.nn.modules.conv.Concat [1]
16     -1 1 32096 ultralytics.nn.modules.block.C3k2 [256, 64, 1, False]
17     -1 1 36992 ultralytics.nn.modules.conv.Conv [64, 64, 3, 2]
18     [-1, 13] 1 0 ultralytics.nn.modules.conv.Concat [1]
19     -1 1 86720 ultralytics.nn.modules.block.C3k2 [192, 128, 1, False]
20     -1 1 147712 ultralytics.nn.modules.conv.Conv [128, 128, 3, 2]
21     [-1, 10] 1 0 ultralytics.nn.modules.conv.Concat [1]
22     -1 1 378880 ultralytics.nn.modules.block.C3k2 [384, 256, 1, True]
23     [16, 19, 22] 1 433012 ultralytics.nn.modules.head.Detect [12, [64, 128, 256]]
YOLO11n summary: 181 layers, 2,592,180 parameters, 2,592,164 gradients, 6.5 GFLOPs

```

Transferred 448/499 items from pretrained weights

Freezing layer 'model.23.dfl.conv.weight'

train: Fast image access (ping: 0.20.1 ms, read: 180.3124.6 MB/s, size: 46.0 KB)

train: Scanning C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\InfraredSolarModules\ImageSet\train\labels.cache... 6924 images, 7 backgrounds, 0 corrupt: 100% ————— 6924/6924 6.9Mit/s 0.0s

WARNING Box and segment counts should be equal, but got len(segments) = 189, len(boxes) = 19845. To resolve this only boxes will be used and all segments will be removed. To avoid this please supply either a detect or segment dataset, not a detect-segment mixed dataset.

val: Fast image access (ping: 0.10.1 ms, read: 204.161.7 MB/s, size: 59.9 KB)

val: Scanning C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\InfraredSolarModules\ImageSet\valid\labels.cache... 400 images, 0 backgrounds, 0 corrupt: 100% ————— 400/400 0.0s

Plotting labels to C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect\labels.jpg...

optimizer: 'optimizer=auto' found, ignoring 'lr0=0.01' and 'momentum=0.937' and determining best 'optimizer', 'lr0' and 'momentum' automatically...

optimizer: AdamW(lr=0.000625, momentum=0.9) with parameter groups 81 weight(decay=0.0), 88 weight(decay=0.0005), 87 bias(decay=0.0)

Image sizes 640 train, 640 val

Using 0 dataloader workers

Logging results to C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect

Starting training for 15 epochs...

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
1/15	4.27G	1.597	3.806	1.275	43	640: 100% ————— 433/433 2.0it/s 3:35<0.5s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 1.8it/s 7.2s0.6s
	all	400	1119	0.288	0.278	0.195 0.127
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
2/15	3.89G	1.448	2.535	1.182	82	640: 100% ————— 433/433 2.1it/s 3:29<0.5s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 1.9it/s 6.9s0.6s
	all	400	1119	0.338	0.408	0.328 0.216
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
3/15	4G	1.412	2.198	1.158	45	640: 100% ————— 433/433 2.6it/s 2:44<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 3.1it/s 4.1s0.3s
	all	400	1119	0.322	0.378	0.32 0.208
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
4/15	3.96G	1.366	1.977	1.136	40	640: 100% ————— 433/433 3.1it/s 2:21<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 3.2it/s 4.1s0.3s
	all	400	1119	0.385	0.546	0.395 0.258
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
5/15	3.96G	1.338	1.825	1.117	40	640: 100% ————— 433/433 3.1it/s 2:19<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 3.6it/s 3.6s0.3s
	all	400	1119	0.387	0.488	0.398 0.278
Closing dataloader mosaic						
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
6/15	3.96G	1.304	1.711	1.125	27	640: 100% ————— 433/433 3.4it/s 2:07<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 3.6it/s 3.6s0.3s
	all	400	1119	0.458	0.538	0.51 0.351
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
7/15	3.96G	1.263	1.601	1.101	19	640: 100% ————— 433/433 2.6it/s 2:45<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 2.8it/s 4.6s0.4s
	all	400	1119	0.527	0.562	0.539 0.391
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
8/15	3.96G	1.23	1.51	1.088	50	640: 100% ————— 433/433 3.2it/s 2:17<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 3.7it/s 3.6s0.3s
	all	400	1119	0.582	0.626	0.6 0.439
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
9/15	3.96G	1.212	1.46	1.071	18	640: 100% ————— 433/433 2.6it/s 2:45<0.4s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 2.5it/s 5.2s0.4s
	all	400	1119	0.539	0.667	0.582 0.428
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
10/15	3.96G	1.176	1.392	1.057	34	640: 100% ————— 433/433 2.6it/s 2:47<0.3s
	Class	Images	Instances	Box(P	R	mAP50 mAP50-95): 100% ————— 13/13 2.6it/s 5.0s0.4s
	all	400	1119	0.64	0.612	0.643 0.482
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	640: 100%	433/433	2.8it/s	2:35<0.3s
11/15	3.96G	1.162	1.329	1.053	24	640: 100%	433/433	2.8it/s	2:35<0.3s	
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.9it/s	4.5s0.4s
	all	400	1119	0.604	0.708	0.676	0.508			
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	640: 100%	433/433	2.6it/s	2:46<0.4s
12/15	3.96G	1.139	1.29	1.042	21	640: 100%	433/433	2.6it/s	2:46<0.4s	
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.5it/s	5.1s0.4s
	all	400	1119	0.647	0.638	0.687	0.508			
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	640: 100%	433/433	2.6it/s	2:49<0.4s
13/15	3.96G	1.115	1.241	1.029	39	640: 100%	433/433	2.6it/s	2:49<0.4s	
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.4it/s	5.3s0.4s
	all	400	1119	0.683	0.652	0.691	0.517			
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	640: 100%	433/433	2.5it/s	2:51<0.3s
14/15	3.96G	1.096	1.211	1.025	24	640: 100%	433/433	2.5it/s	2:51<0.3s	
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.5it/s	5.2s0.4s
	all	400	1119	0.667	0.685	0.703	0.537			
Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size	640: 100%	433/433	2.6it/s	2:48<0.3s
15/15	3.96G	1.083	1.173	1.016	39	640: 100%	433/433	2.6it/s	2:48<0.3s	
	Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.5it/s	5.1s0.4s
	all	400	1119	0.646	0.696	0.713	0.552			

15 epochs completed in 0.705 hours.

Optimizer stripped from C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect\weights\last.pt, 5.5MB
Optimizer stripped from C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect\weights\best.pt, 5.5MB

Validating C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect\weights\best.pt...

Ultralytics 8.3.235 Python-3.11.14 torch-2.10.0.dev20251205+cu128 CUDA:0 (NVIDIA GeForce RTX 5070 Laptop GPU, 8151MiB)

YOLO11n summary (fused): 100 layers, 2,584,492 parameters, 0 gradients, 6.3 GFLOPs

Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%	13/13	2.2it/s	6.0s0.4s
all	400	1119	0.647	0.696	0.713	0.552			
Cell	29	34	0.826	0.699	0.838	0.637			
Cell-Multi	39	77	0.461	0.506	0.505	0.383			
Cracking	138	209	0.577	0.752	0.725	0.591			
Diode	142	263	0.795	0.863	0.884	0.728			
Diode-Multi	70	128	0.572	0.609	0.653	0.516			
Hot-Spot	122	323	0.717	0.789	0.787	0.622			
Hot-Spot-Multi	44	69	0.725	0.783	0.824	0.619			
No-Anomaly	14	16	0.499	0.562	0.492	0.317			

Speed: 0.3ms preprocess, 3.6ms inference, 0.0ms loss, 2.0ms postprocess per image

Results saved to C:\Users\markm\Desktop\AIDA2154_Solar_Panel_Project\Solar_Final_Runs\yolo11_detect

```
[2]: import yaml
import torch
from ultralytics.data import build_data_loader
from ultralytics.cfg import get_cfg
from ultralytics.data.utils import check_det_dataset

print("🔍 Starting Data Loader Stress Test...")

try:
    # 1. Load default YOLO configuration (avoiding the 'val' key conflict)
    # We set the data config in overrides, not as the base config
    cfg = get_cfg(cfg=None, overrides={'data': 'final_solar_config.yaml', 'imgsz': 640, 'batch': 16})

    # 2. Parse the dataset file explicitly to get the image path
    data_info = check_det_dataset('final_solar_config.yaml')
    train_image_path = data_info['train']

    print(f"📂 Reading images from: {train_image_path}")

    # 3. Build the data loader (CPU only, no GPU math)
    loader = build_data_loader(cfg, batch=16, img_path=train_image_path, stride=32, rank=-1, mode="train")

    # 4. Stress Test: Read 50 batches from disk into RAM
    print("🔄 Iterating through 50 batches (this tests your CPU, RAM, and Disk)...")
    for i, batch in enumerate(loader):
        if i >= 50: break
        # Print progress every 5 batches to keep output clean
        if i % 5 == 0:
            print(f"✅ Loaded batch {i}/50")

    print("\n🎉 SUCCESS: Data Loading is STABLE.")
    print("🏆 CONCLUSION: Your Data, Disk, and RAM are healthy.")
    print("🚫 The kernel crash is confirmed to be the Python 3.13 <-> PyTorch Nightly incompatibility.")

except Exception as e:
    print(f"❌ FAILURE: Data Loader crashed.")
    print(f"📄 Error details: {e}")
```

🔍 Starting Data Loader Stress Test...

❌ FAILURE: Data Loader crashed.

Error details: argument of type 'NoneType' is not iterable

```
[ ]: # Cell 4: Validation & Metrics (Task 4)
# Evaluate the model on the validation set.
```

```

print("📌 Running Validation...")
metrics = model.val()

print(f"\n📊 Results:")
print(f"    mAP@50:    {metrics.box.map50:.3f}")
print(f"    mAP@50-95: {metrics.box.map:.3f}")
print(f"    Precision: {metrics.box.mean_results()[0]:.3f}")
print(f"    Recall:    {metrics.box.mean_results()[1]:.3f}")

```

```

[ ]: # Cell 5: Inference Demonstration (Task 4 Demo)
      # Visualizing results on the unseen 'Test' set.

test_images_path = os.path.join(dataset_root, 'test', 'images')
test_files = glob.glob(os.path.join(test_images_path, "*.jpg"))

if test_files:
    # Pick 3 random images
    samples = random.sample(test_files, min(len(test_files), 3))

    print(f"🔍 Running inference on {len(samples)} random test images...")

    for img_path in samples:
        # Run Inference
        results = model.predict(img_path, conf=0.25, verbose=False)

        # Plotting
        result_plot = results[0].plot()

        # Convert BGR to RGB for Matplotlib
        result_rgb = cv2.cvtColor(result_plot, cv2.COLOR_BGR2RGB)

        plt.figure(figsize=(10, 8))
        plt.imshow(result_rgb)
        plt.axis('off')
        plt.title(f"Test Image: {os.path.basename(img_path)}")
        plt.show()
    else:
        print("⚠️ No test images found in the dataset folder.")

```