# COMSATS UNIVERSITY ISLAMABAD, WAH CAMPUS



# **Group-Members:**

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FA21-BCS-080 Ali Shan

**Section:** 

BCS-8A

# **Project-Title:**

# Real-time Weapon Detection

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## **Learning Outcomes:**

**HUSSAIN ALI** 

(FA21-BCS-066):

**Model Training** 

**Model Validation** 

Flow-Diagrams

**ALI SHAN** 

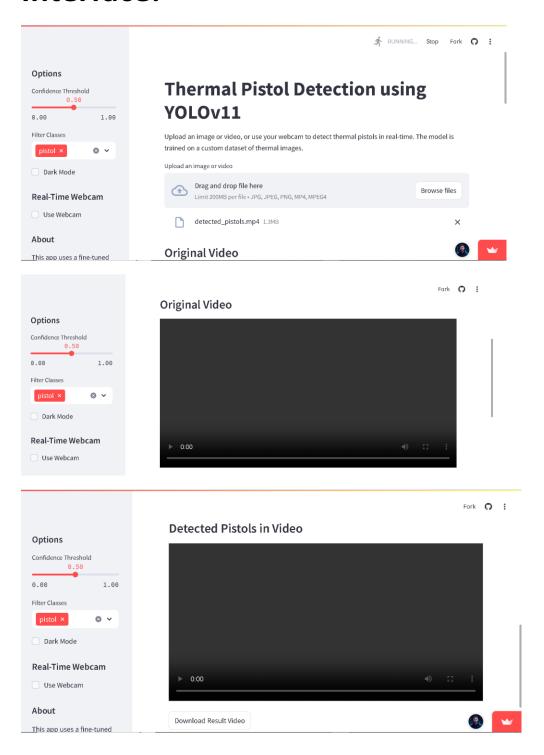
(FA21-BCS-080):

Interface

**Model Training** 

**Model Validation** 

### Interface:

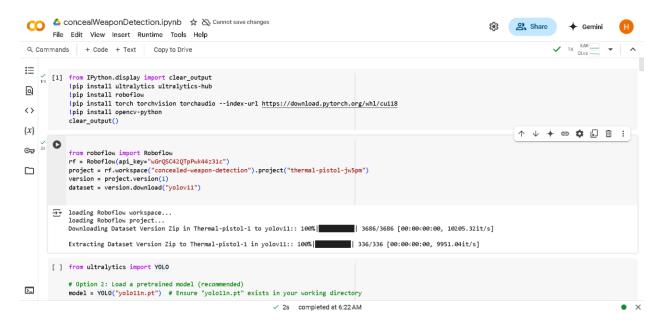


#### Video:



detected\_pistols (2).mp4

## **Model Training**







```
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from ultralytics import YOLO
    # Load the pretrained model
    model = YOLO("yolo11n.pt") # Replace with the path to your pretrained model
    # Fine-tune the model on your custom dataset
    results = model.train(
       data="/content/Thermal-pistol-1/data.yaml", # Path to your dataset configuration file
        epochs=100,
                                  # Number of training epochs
       imgsz=640,
                                  # Image size
       batch=16,
                                  # Batch size
       device=0,
                                  # Use GPU (set device=0 for GPU, device='cpu' for CPU)
       workers=2,
                                  # Number of data loading workers
       lr0=0.01,
                                  # Initial learning rate
       weight_decay=0.0005,
                                  # Weight decay
       optimizer="SGD",
                                  # Optimizer (SGD, Adam, etc.)
        name="yolo11n_finetuned"  # Name of the training run
    # Validate the model on the validation set
    metrics = model.val() # Validate the model
    print(metrics.box.map) # Print mAP (mean Average Precision)
```

#### **Model Validation**

```
# Validate the model on the validation set
                                                                                                                                                                                                                                                        ↑ ↓ + © ‡ 🗓 🗓 :
∷
                      metrics = model.val() # Validate the model
print(metrics.box.map) # Print mAP (mean Average Precision)
Q

→ Ultralytics 8.3.95 

Ø Python-3.11.11 torch-2.6.0+cu124 CUDA:0 (Tesla T4, 15095MiB)

                     Downloading https://ultralytics.com/assets/Arial.ttf to '/root/.config/Ultralytics/Arial.ttf'...

| Pownloading https://ultralytics.com/assets/Arial.ttf to '/root/.config/Ultralytics/Arial.ttf'...

| 755k/755k [00:00:00:00:2.2.4HB/s]
| 755k/755k [00:00:00:00:2.2.4HB/s]
<>
{x}
                                                                                                                                                                          arguments
[3, 16, 3, 2]
[16, 32, 3, 2]
[16, 32, 3, 2]
[32, 64, 1, False, 0.25]
[64, 128, 1, False, 0.25]
[128, 128, 3, 2]
[128, 128, 1, True]
[128, 256, 3, 2]
[256, 256, 1, True]
[256, 256, 1, True]
[256, 256, 1]
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[1]
                                                                            464 ultralytics.nn.modules.conv.Conv
4672 ultralytics.nn.modules.conv.Conv
6640 ultralytics.nn.modules.block.C3k2
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36992
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ultralytics.nn.modules.block.SPPF
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                                                                           346112
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249728
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                                                            -1
                                                                            0 torch.nn.modules.upsampling.Upsample
0 ultralytics.nn.modules.conv.Concat
111296 ultralytics.nn.modules.block.C3k2
                                                                                                                                                                            [1]
[384, 128, 1, False]
[None, 2, 'nearest']
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13
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32096 ultralytics.nn.modules.block.C3k2
36992 ultralytics.nn.modules.conv.Conv
                                                                                                                                                                            [1]
[256, 64, 1, False]
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                                                  [-1, 4]
                                                                            36992 ultralytics.nn.modules.conv.Conv
0 ultralytics.nn.modules.conv.Concat
>_
                                                 [-1, 13] 1
                                                                                                                                                                           [1]
```

```
# Validate the model on the validation set
                                                                                                                                                                                                                                       ↑ ↓ + © ‡ 🗓 🗓 :
                    metrics = model.val() # Validate the model
print(metrics.box.map) # Print mAP (mean Average Precision)
∷
Q
                                                                                                                                                                [128, 128, 3, 2]
             ₹
                    21 [-1, 10] 1 0 ultralytics.nn.modules.comv.Concat

22 -1 1 37880 ultralytics.nn.modules.block.G3k2

23 [16, 19, 22] 1 430867 ultralytics.nn.modules.head.Detect

YOLO11n summary: 181 layers, 2,590,035 parameters, 2,590,019 gradients, 6.4 GFLOPs
                                                                                                                                                                [1]
<>
                                                                                                                                                                [384. 256, 1, True]
                                                                                                                                                                [1, [64, 128, 256]]
{x}
                    Transferred 448/499 items from pretrained weights

TensorBoard: Start with 'tensorboard --logdir runs/detect/yolo11n_finetuned', view at <a href="http://localhost:6006/">http://localhost:6006/</a>
Freezing layer 'model.23.dfl.conv.weight'
⊙ಫ
                    AMP: running Automatic Mixed Precision (AMP) checks...

AMP: checks passed 
train: Scanning /content/Thermal-pistol-1/train/labels... 114 images, 1 backgrounds, 0 corrupt: 100% 114/114 [00:00<00:00, 2260.26it/s]train: New
albumentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01, blur_limit=(3, 7)), ToGray(p=0.01, num_output_channels=3, method='weighted_average'), CL# val: Scanning /content/Thermal-pistol-1/valid/labels... 32 images, 0 backgrounds, 0 corrupt: 100%| 32/32 [00:00<00:00, 1492.00it/s]val: New cache
                    Plotting labels to runs/detect/yolo11n_finetuned/labels.jpg..
                    riorting labels to runs/detect/yololin_rinetuned/labels.jpg...

optimizer: SGO(lr=0.01, momentum=0.937) with parameter groups 81 weight(decay=0.0), 88 weight(decay=0.0055), 87 bias(decay=0.0)

TensorBoard: model graph visualization added 

Image sizes 640 train, 640 val

Using 2 dataloader workers
Logging results to runs/detect/yololin_finetuned

Starting training for 100 epochs...
                                             GPU_mem box_loss cls_loss dfl_loss Instances
>_
                                                                                                                                                                            8/8 [00:04<00:00, 1.96it/s]
                                                               2.234
                                                                                       5.03
                                                                                                    1.969
                                                                                                                                                640: 100%
                                                                                                                    ✓ 13s completed at 6:24AM
```

0	metrics = mod print(metrics				rage Precis	ion)		↑ ↓ <b>†</b> ⊕ <b>‡</b>	
	pr inc (me cr ice	. oox . map)		•					
₹	Epoch	GPU_mem	box_loss			Instances	Size		
	1/100	2.29G Class	2.234 Images	5.03 Instances	1.969 Box(P	7 R		100%  8/8 [00:04<00:00, 1.96it/s] mAP50-95): 100%  1/1 [00:01<00:00, 1.27s/it]	al
	Epoch	GPU_mem	box_loss			Instances	Size	<u> </u>	
	2/100	2.69G	1.93	4.985	1.807	3		100%  8/8 [00:01<00:00, 4.08it/s]	
		Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00, 2.59it/s]	al
	Epoch	GPU mem	box loss	cls loss	dfl loss	Instances	Size		
	3/100	2.7G	1.643	4.638	1.619	1		100% 8/8 [00:02<00:00, 3.32it/s]	
	-,	Class	Images	Instances	Box(P	R		mAP50-95): 100%    1/1 [00:00<00:00, 3.67it/s]	al
	Epoch	GPU mem	box loss	cls loss	dfl loss	Instances	Size		
	4/100	2.71G	1.697	3.972	1.487	3		100% 8/8 [00:01<00:00, 4.53it/s]	
	., ===	Class		Instances	Box(P	R		mAP50-95): 100%  1/1 [00:00<00:00, 3.56it/s]	a!
	Epoch	GPU mem	box loss	cls_loss	dfl loss	Instances	Size		
	5/100	2.72G	1.4	3.828	1.433	2	640:	100% 8/8 [00:01<00:00, 4.84it/s]	
		Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%   1/1 [00:00<00:00, 2.67it/s]	al
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size	10001	
	7/400				1	✓ 13s com			

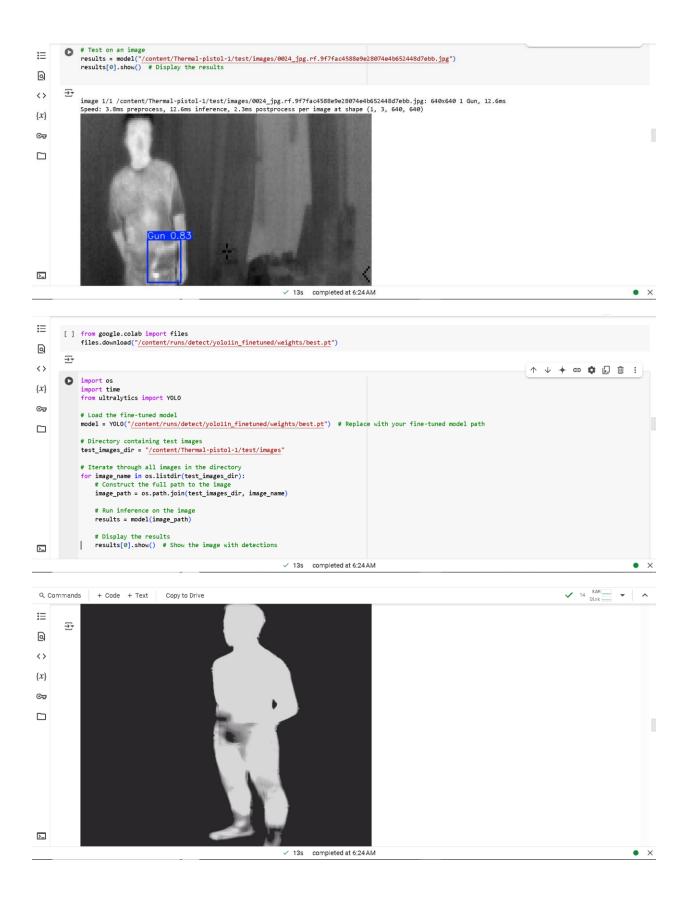
<b>:</b> ≡	0	# Validate th							↑ ↓ <b>+</b> ⊕ <b>\$</b> ᡚ	<u> </u>
Q		print(metrics	.box.map)	# Print m/	AP (mean Ave		ion)		1 0/0 [00/02/00/00](02/0]	
	₹		Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00, 2.09it/s]	all 🗖
<>		Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size		
{ <i>x</i> }		8/100	2.75G Class	1.252 Images	2.787 Instances	1.349 Box(P	3 R		100%  8/8 [00:02<00:00, 3.32it/s] mAP50-95): 100%  1/1 [00:00<00:00, 3.33it/s]	all
© <del></del>										
		Epoch 9/100	GPU_mem 2.77G Class	box_loss 1.37 Images	cls_loss 2.616 Instances	dfl_loss 1.396 Box(P	Instances 2 R		100% 8/8 [00:01<00:00, 4.57it/s] mAP50-95): 100% 1/1 [00:00<00:00, 3.47it/s]	all
		Epoch 10/100	GPU_mem 2.78G Class	box_loss 1.211 Images	cls_loss 2.354 Instances	dfl_loss 1.238 Box(P	Instances 5 R		8/8 [00:01<00:00, 4.75it/s]   mAP50-95): 100%    1/1 [00:00<00:00, 3.83it/s]	all
		Epoch 11/100	GPU_mem 2.79G Class	box_loss 1.202 Images	cls_loss 2.293 Instances	dfl_loss 1.251 Box(P	Instances 2 R		100%  8/8 [00:01<00:00, 4.72it/s] mAP50-95): 100%  1/1 [00:00<00:00, 3.11it/s]	all
>_		Epoch 12/100	GPU_mem 2.79G Class	box_loss 1.364 Images	cls_loss 2.15 Instances	dfl_loss 1.306 Box(P	Instances 5 R		100%  8/8 [00:01<00:00, 4.57it/s] mAP50-95): 100%  1/1 [00:00<00:00, 2.65it/s]	all
							✓ 13s co	mpleted at 6:2	4AM	• ×

<b>≡</b>	0	# Validate the metrics = mod print(metrics	del.val()	# Validate	the model	rage Precis	ion)			↑ ↓ <b>+</b> ∈	
Q	<del>∑</del>	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size			
<>		14/100	2.81G Class	1.478 Images	2.03 Instances	1.442 Box(P	2 R		100%  8/8 [00:01<00:00, 4.60it/s] mAP50-95): 100%  1/1 [00:00<00:00,		all
{x}		Epoch	GPU mem	box_loss	cls loss	d61 1000	Instances	Size			
		15/100	2.82G	1.223	1.856	1.215	instances 2		100%  8/8 [00:01<00:00, 4.68it/s]		
© <del></del>			Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00,	3.85it/s]	all
		Epoch	GPU_mem	box_loss	cls_loss	dfl loss	Instances	Size			
		16/100	2.83G	1.26	1.866	1.28	5		100%  8/8 [00:01<00:00, 4.98it/s]		
			Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00,	3.02it/s]	al1
		Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size			
		17/100	2.84G	1.209	1.937	1.255	4		100% 8/8 [00:01<00:00, 4.61it/s]		
			Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00,	3.37it/s]	a11
		Epoch	GPU_mem	box_loss	cls_loss		Instances	Size			
		18/100	2.85G	1.209	1.852	1.206	2		100% 8/8 [00:02<00:00, 2.92it/s]		
			Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00,	4.031t/s]	all
>_		Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size	1000/1		
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∷	0	metrics = mod		the valida # Validate					↑ ↓ + © <b>‡</b> 🖟 🗓	
		print(metrics	.box.map)	# Print mA	AP (mean Ave	rage Precis	ion)			
Q	⊋₹	21/100	2.88G	1.12	1.581	1.204	instances 2		100%  8/8 [00:01<00:00, 4.14it/s]	4
>			Class	Images	Instances	Box(P	R	mAP50	mAP50-95): 100%  1/1 [00:00<00:00, 1.63it/s]	all
<b>x</b> }		Epoch	GPU_mem	box_loss			Instances	Size	1.000 [	
7		22/100	2.89G Class	1.061 Images	1.673 Instances	1.117 Box(P	2 R		100%   8/8 [00:02:00:00, 3.78it/s] mAP50-95): 100%   1/1 [00:00<00:00, 2.37it/s]	all
5		Epoch	GPU_mem	box_loss			Instances	Size		
		23/100	2.9G Class	1.156 Images	1.527 Instances	1.154 Box(P	5 R		100%   8/8 [00:02:00:00, 3.57it/s] mAP50-95): 100%   1/1 [00:00:00:00, 3.53it/s]	all
		Epoch	GPU_mem	box_loss			Instances	Size		
		24/100	2.91G Class	1.123 Images	1.572 Instances	1.17 Box(P	2 R		100%  8/8 [00:01<00:00, 4.37it/s] mAP50-95): 100%    1/1 [00:00<00:00, 4.29it/s]	a11
		Epoch	GPU_mem	box_loss			Instances	Size		
		25/100	2.92G Class	1.088 Images	1.606 Instances	1.17 Box(P	2 R		100%  8/8 [00:01<00:00, 4.55it/s] mAP50-95): 100%  1/1 [00:00<00:00, 4.49it/s]	all
_		Epoch	GPU_mem	box_loss			Instances	Size		
		26/100	2.93G	1.259	1.689	1.269	2		100%  8/8 [00:01<00:00, 4.52it/s]	
							✓ 13s com	pleted at 6:2	4AM	•

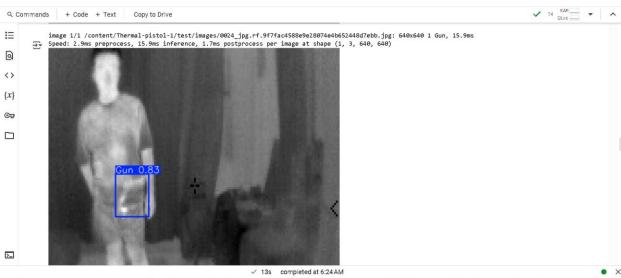












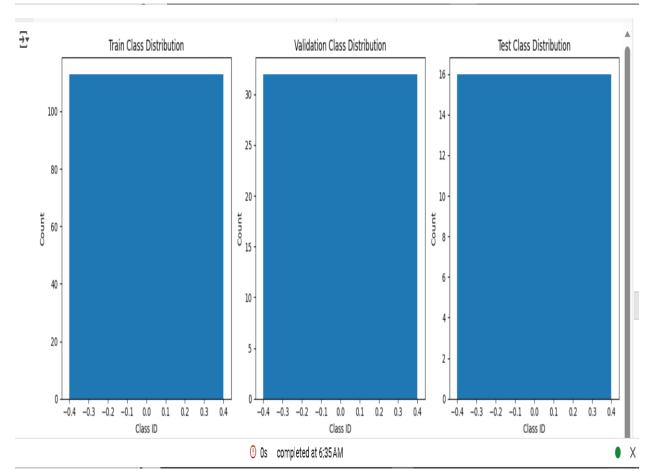
```
|↑ ↓ + ⊖ ‡ 🗓 🗓 : ||
from ultralytics import YOLO
 import matplotlib.pyplot as plt
# Load trained model
model = Y0L0('/content/runs/detect/yolo11n_finetuned/weights/best.pt')
# Validate model
metrics = model.val(
   data='/content/Thermal-pistol-1/data.yaml',
   split='test', # Use test split for final evaluation
      plots=True
# Print key metrics
print(f"mAP50-95: {metrics.box.map:.4f}")
print(f"Precision: {metrics.box.mp:.4f}")
print(f"Recall: {metrics.box.mr:.4f}")
print(f"F1 Score: {2 * (metrics.box.mp * metrics.box.mr) / (metrics.box.mp + metrics.box.mr):.4f}")
# Plot confusion matrix
confusion_matrix = plt.imread('/content/runs/detect/val/confusion_matrix.png')
plt.figure(figsize=(10, 8))
plt.imshow(confusion_matrix)
plt.axis('off')
plt.title('Confusion Matrix')
plt.show()
                                                                                      ① 0s completed at 6:35 AM
                                                                                                                                                                                                                                          • ×
        import yaml
import cv2
import numpy as np
from PIL import Image
                                                                                                                                                                                               ↑ ↓ + ⊕ ‡ ᡚ 🗓 :
   0
          from ultralytics.utils.plotting import plot_images
         # Load dataset config
         with open('/content/Thermal-pistol-1/data.yaml') as f:
   data = yaml.safe_load(f)
         # Class distribution visualization
         v Class_counts = []
for split in ['train', 'valid', 'test']:
    label_dir = f'/content/Thermal-pistol-1/{split}/labels'
    counts = np.zeros(data['nc'])
                for label_file in os.listdir(label_dir):
                    with open(os.path.join(label_dir, label_file)) as f:
    for line in f:
        class_id = int(line.split()[0])
        counts[class_id] += 1
               class_counts.append(counts)
         plt.figure(figsize=(15, 5))
for i, split in enumerate(['Train', 'Validation', 'Test']):
    plt.subplot(1, 3, i+1)
    plt.bar(range(data['nc']), class_counts[i])
    plt.title(f'\split\ class Distribution')
    plt.xlabel('Class ID')
```

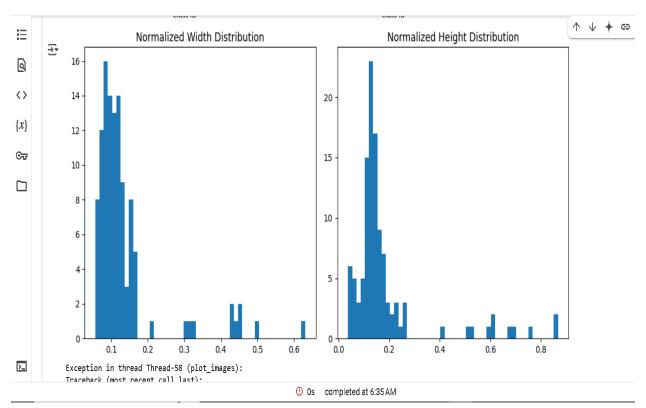
① 0s completed at 6:35 AM

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```
plt.title(f'{split} Class Distribution')
                                                                                                                                     \wedge \downarrow \uparrow
    plt.xlabel('Class ID')
    plt.ylabel('Count')
plt.tight_layout()
plt.show()
# Bounding box size distribution
box_sizes = []
for split in ['train']:
    label_dir = f'/content/Thermal-pistol-1/{split}/labels'
    for label_file in os.listdir(label_dir):
        with open(os.path.join(label_dir, label_file)) as f:
            for line in f:
                _, x, y, w, h = map(float, line.split())
box_sizes.append((w, h))
w, h = zip(*box_sizes)
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.hist(w, bins=50)
plt.title('Normalized Width Distribution')
plt.subplot(1, 2, 2)
plt.hist(h, bins=50)
plt.title('Normalized Height Distribution')
plt.tight_layout()
plt.show()
# Sample training images with labels
```



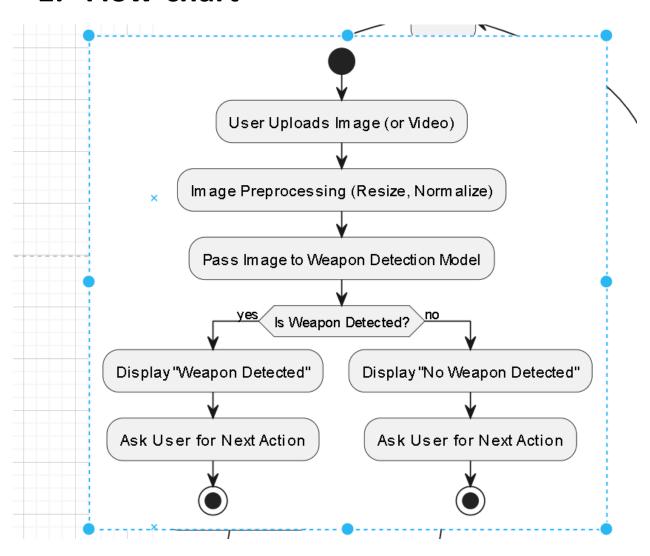




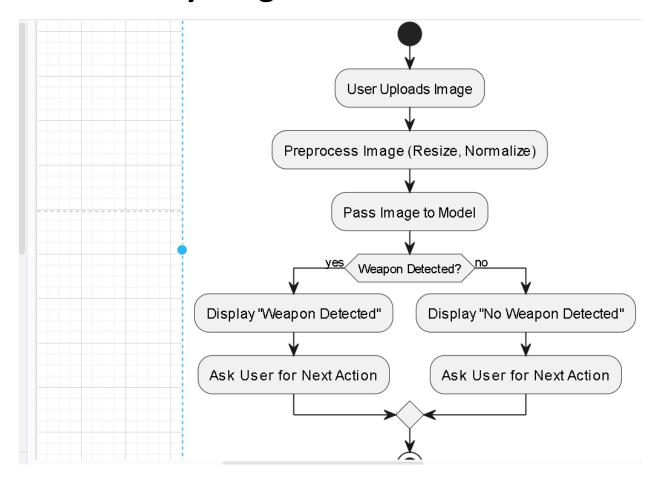
```
# Training metrics visualization
training metrics = plt.imread('/content/runs/detect/yolo11n_finetuned/results.png')
                                                                                                                                                           ↑ ↓ + 🖨 📮 🗓 :
plt.figure(figsize=(12, 6))
plt.imshow(training_metrics)
plt.axis('off')
plt.title('Training Metrics')
plt.show()
fig, axes = plt.subplots(2, 3, figsize=(20, 12))
for ax, img_path in zip(axes.flat, test_images):
    results = model(img_path)
    ax.imshow(results[0].plot())
     ax.axis('off')
plt.tight_layout()
plt.show()
# Precision-Recall curve
pr_curve = plt.imread('/content/runs/detect/val/R_curve.png')
plt.figure(figsize=(10, 8))
plt.imshow(pr_curve)
plt.axis('off')
plt.title('Precision-Recall Curve')
plt.show()
                                                                      ① 0s completed at 6:35 AM
```

## **Flow-Diagrams**

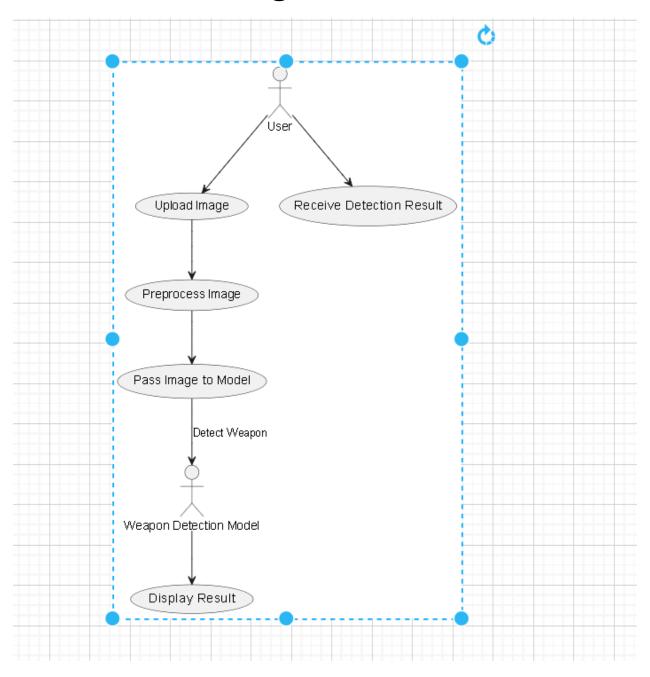
#### 1. Flow-chart



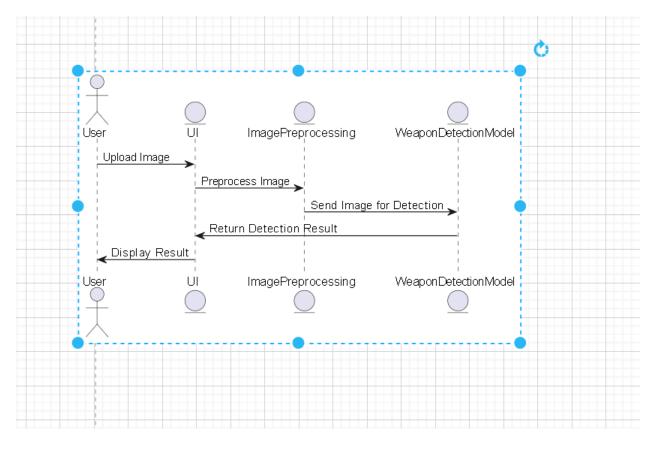
## 2. Activity-Diagram



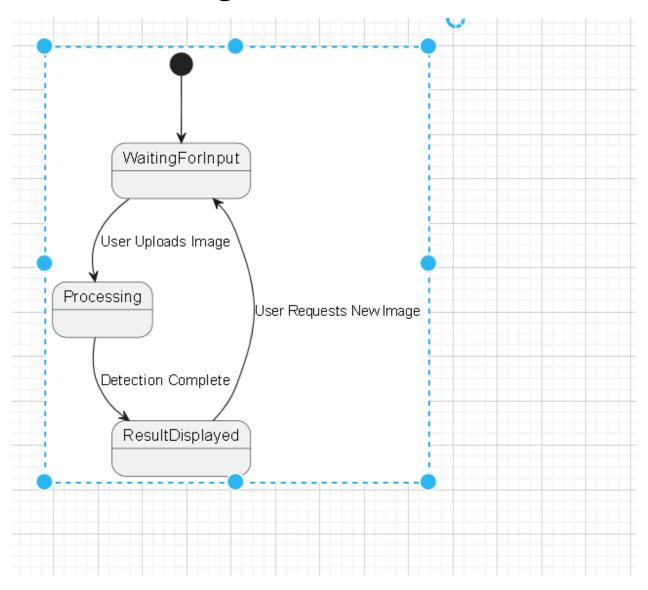
## 3. Use-Case Diagram



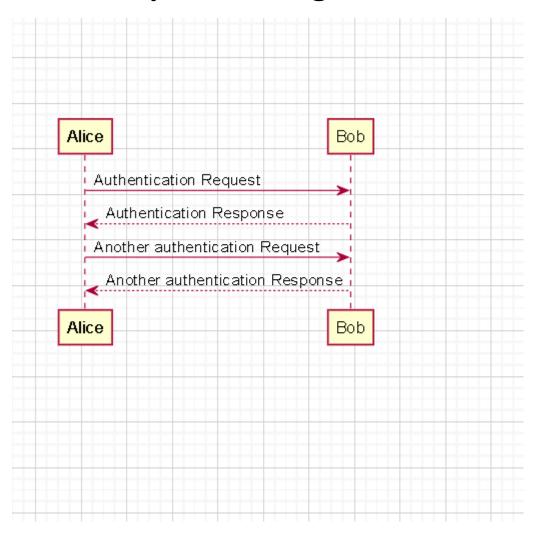
# 4. Sequence-Diagram



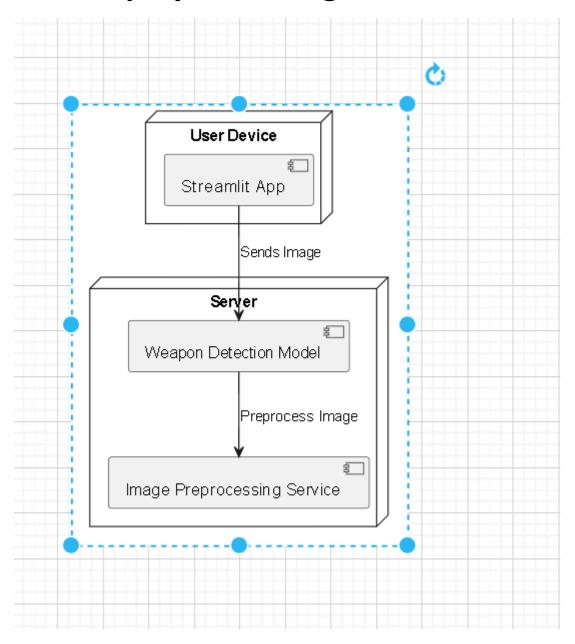
# 5. State-Diagram



# 6. Component-Diagram



## 7. Deployment-Diagram



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