



Università degli Studi di Milano - Bicocca
Università della Svizzera Italiana

DOUBLE DEGREE
MASTER OF SCIENCE IN COMPUTER SCIENCE



Interpretation of Neural Networks and Advanced Image Augmentation for Visual Control of Drones in Human Proximity

Supervisor: Dr. Alessandro Giusti
Co-Supervisor: PhD Stud. Dario Mantegazza

Master Thesis by
MARCO FERRI
ID 807130

PROBLEM INTRODUCTION



[Mantegazza et al. 2019]

HUMAN POSE ESTIMATION
FOR FLYING A DRONE USING
IMAGES FROM ITS CAMERA



ONLY WORKS IN
THE DRONE ARENA



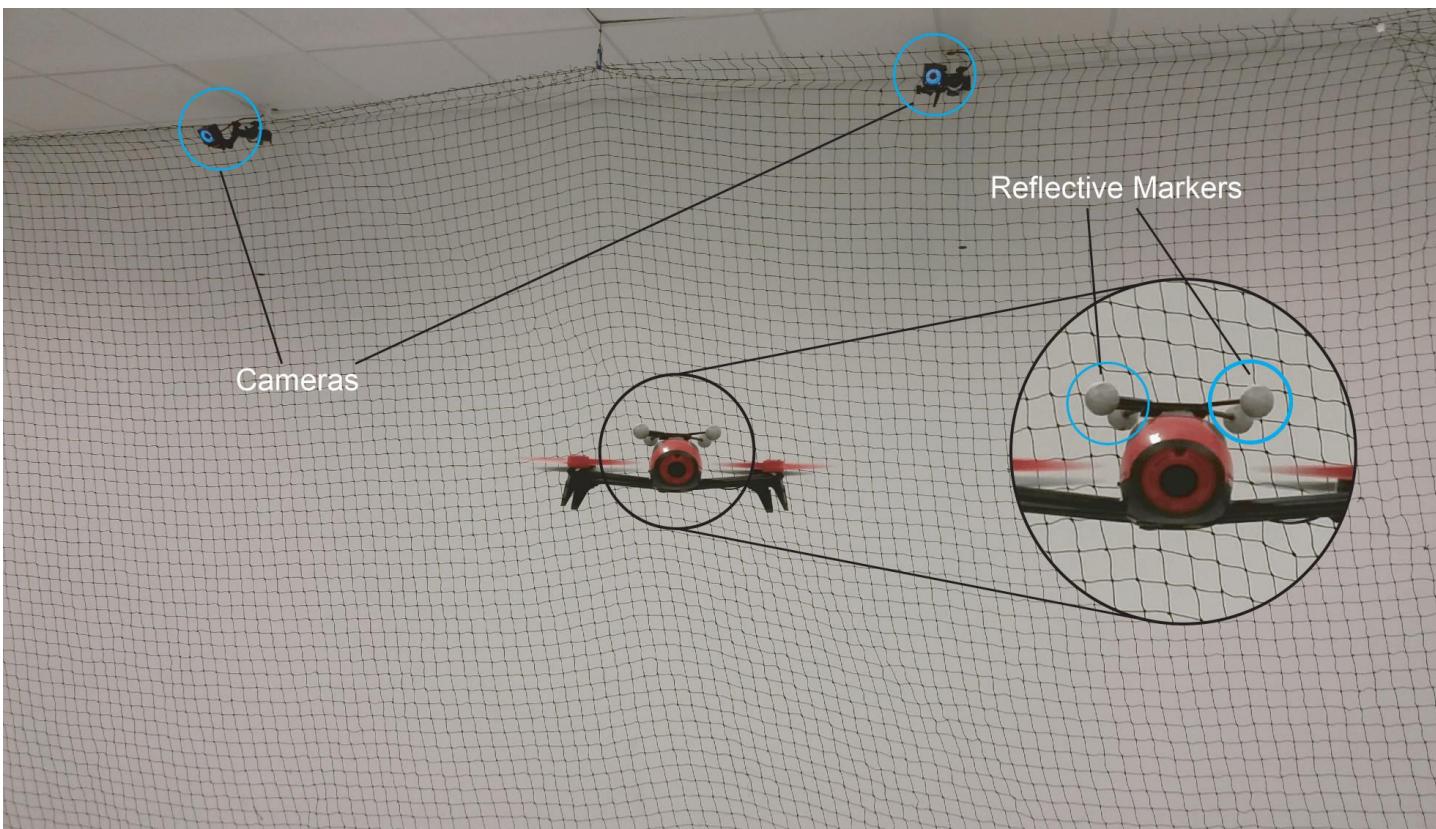
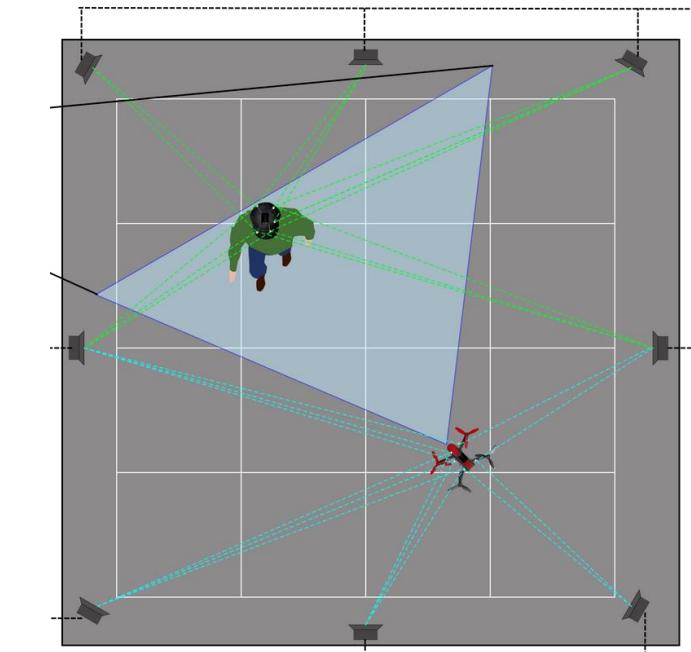
ENHANCE MODEL'S
GENERALIZATION

HOW THE SYSTEM WORKS

- Supervised Learning
- Data collected in a dedicated drone arena
- Motion Capture to acquire the GT



USER'S 3D POSE WITH
RESPECT TO THE DRONE



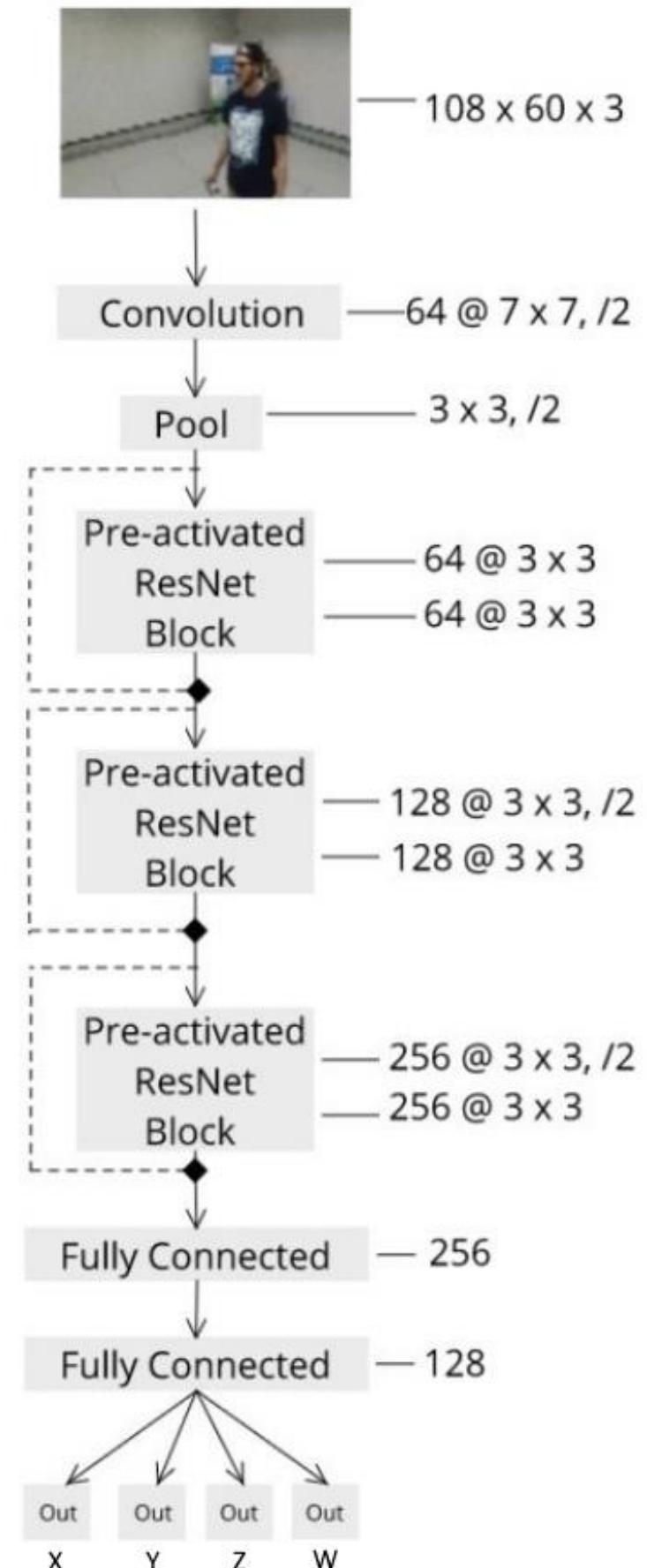
HOW THE SYSTEM WORKS

REGRESSION ON THE USER'S 3D COORDINATES

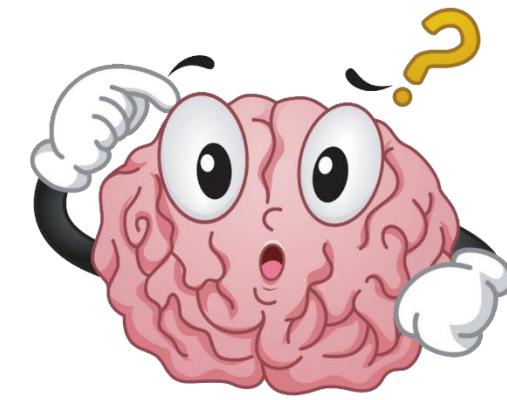
- X Distance from the camera
- Y Horizontal alignment
- Z Vertical alignment
- W Head's orientation (yaw)

RESNET | RESIDUAL NEURAL NETWORK

- Input camera images (108 x 60 RGB)
- Output 4 regression variables



OUR CONTRIBUTION



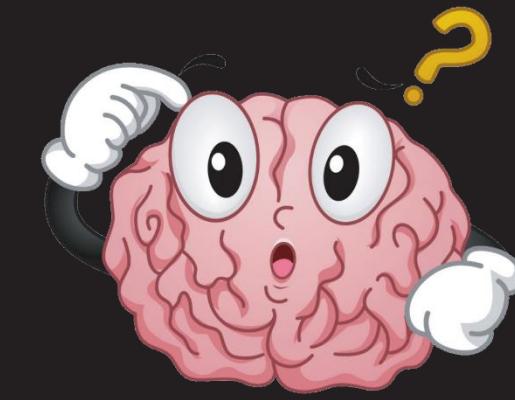
MODEL
INTERPRETATION
TO FIND THE PROBLEM



AUGMENTATION
STRATEGY
TO GENERALIZE THE TASK



EXPERIMENTAL
EVALUATIONS
TO VALIDATE THE SOLUTION



MODEL INTERPRETATION

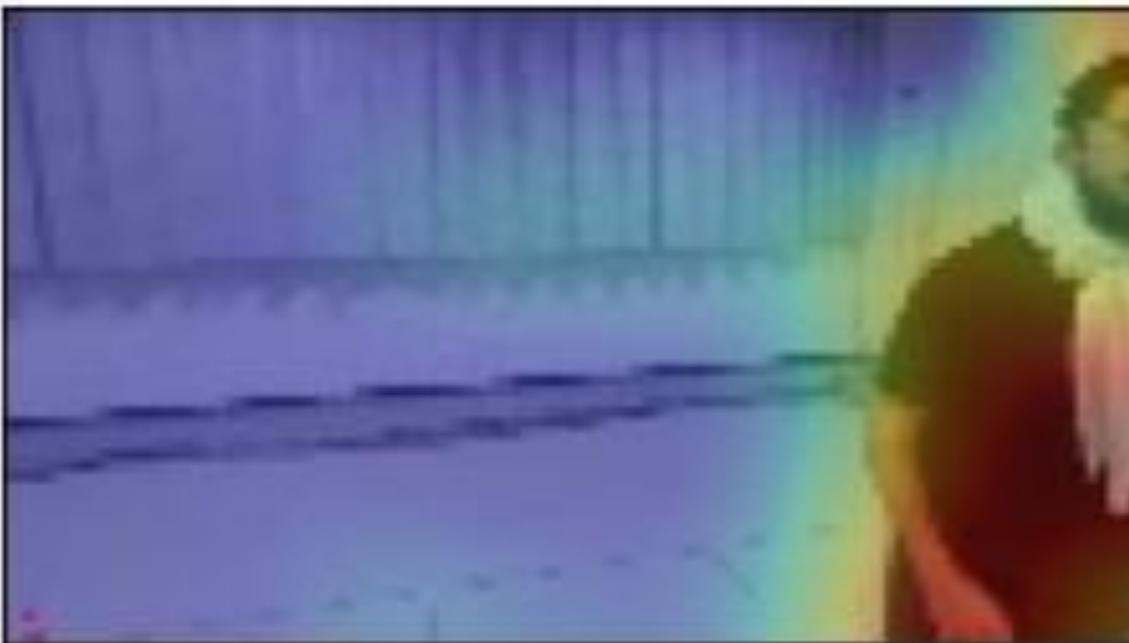
TO FIND THE PROBLEM

GRAD-CAM

HIGHLIGHTS THE REGIONS OF THE INPUT IMAGES
MOST RESPONSIBLE FOR THE NEURAL NETWORK'S OUTPUT



GRAD-CAM EXPECTED OBSERVATIONS



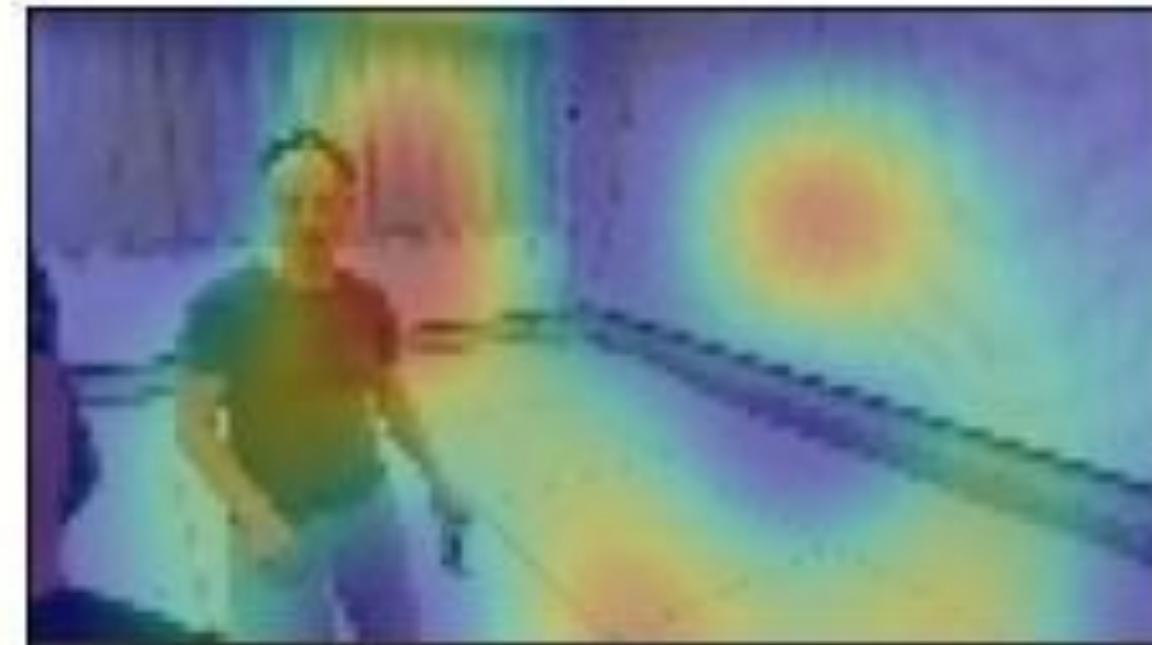
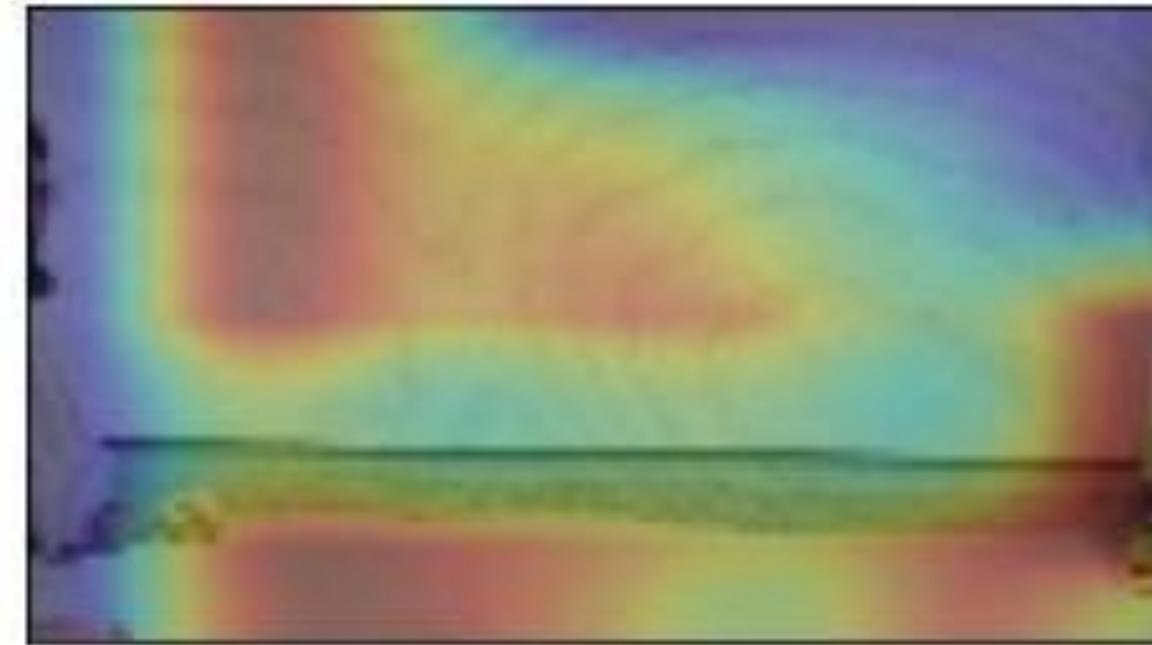
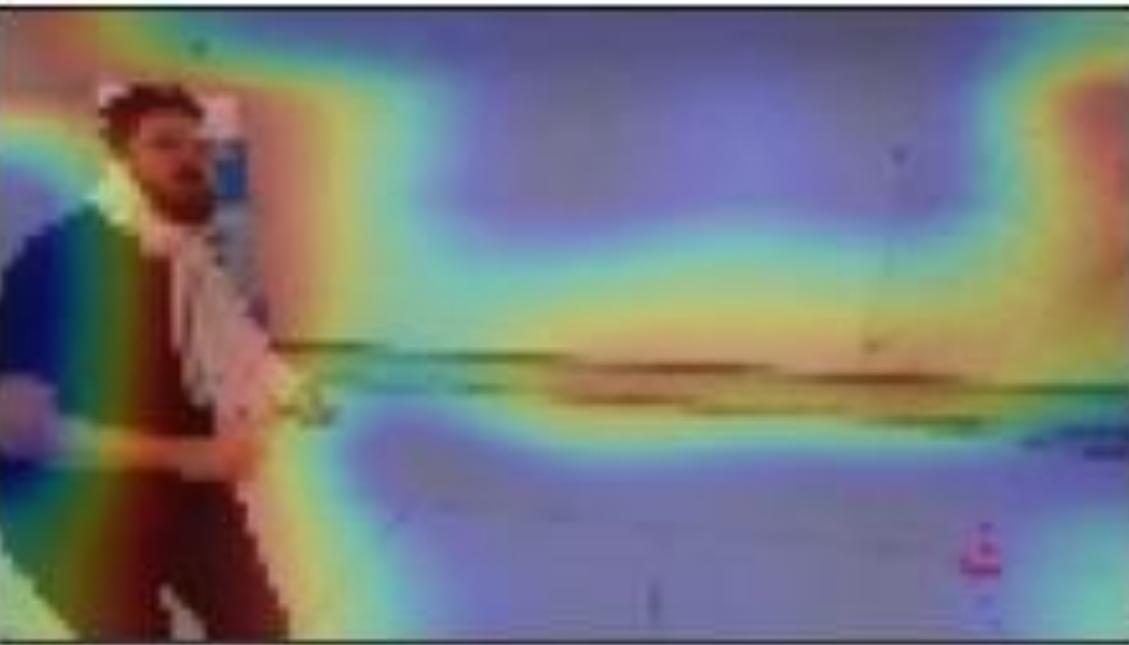
GRAD-CAM REVEALS: BACKGROUND OBJECTS



GRAD-CAM REVEALS: CURTAINS



GRAD-CAM REVEALS: WALLS



GRAD-CAM REVEALS: MULTIPLE PEOPLE



GRAD-CAM SUMMARY



THE MODEL IS DEPENDANT ON THE DRONE ARENA
IN WHICH THE DATASET HAS BEEN COLLECTED



WE WANT TO ELIMINATE THE
OVERFITTING ON THE TRAINING SET

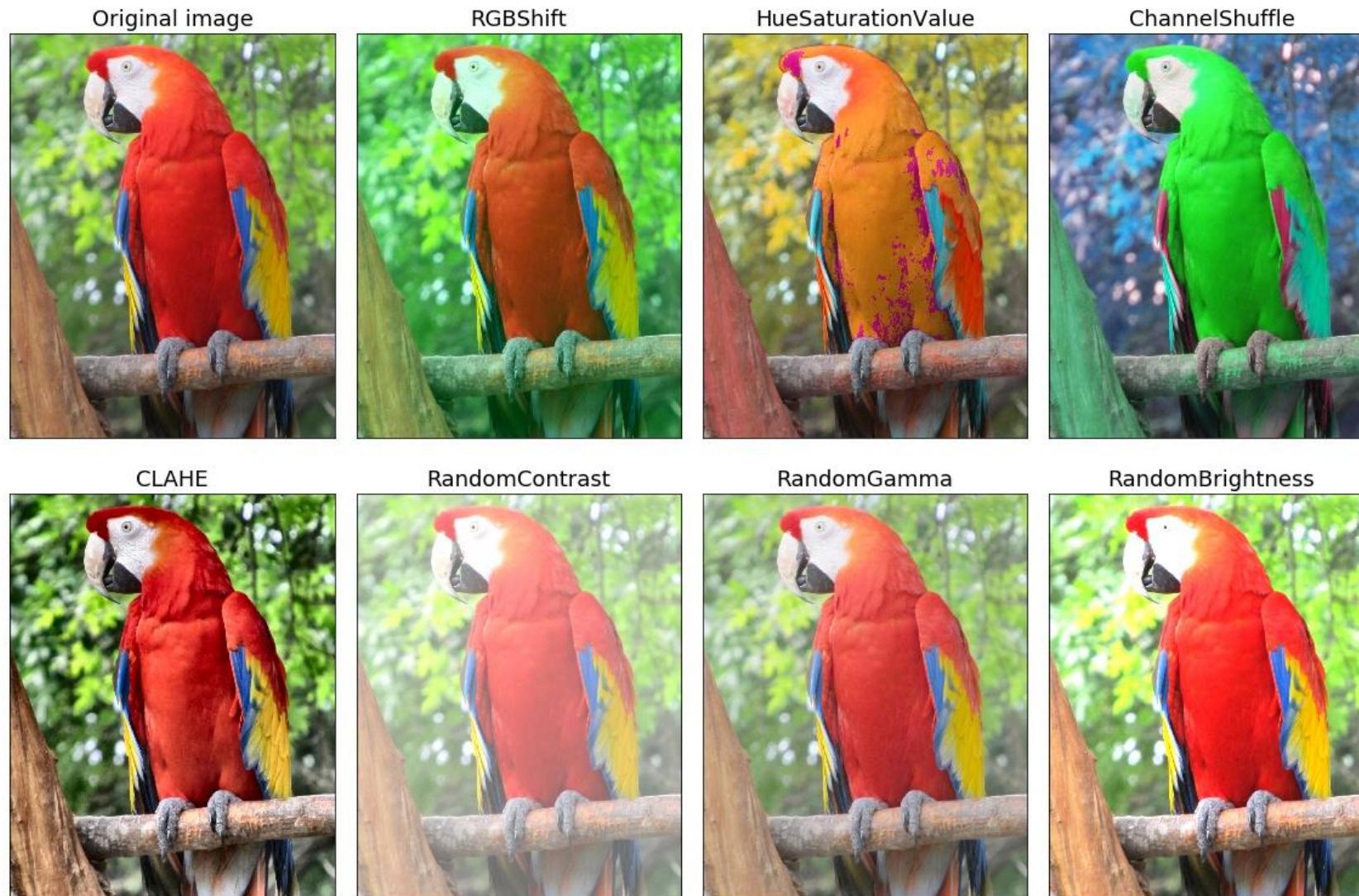


AUGMENTATION STRATEGY TO GENERALIZE THE TASK

WHAT IS DATA AUGMENTATION

WIDELY USED IN MACHINE LEARNING TO GENERATE NEW TRAINING SAMPLES FOR REDUCING OVERFITTING

CLASSIC IMAGE AUGMENTATION



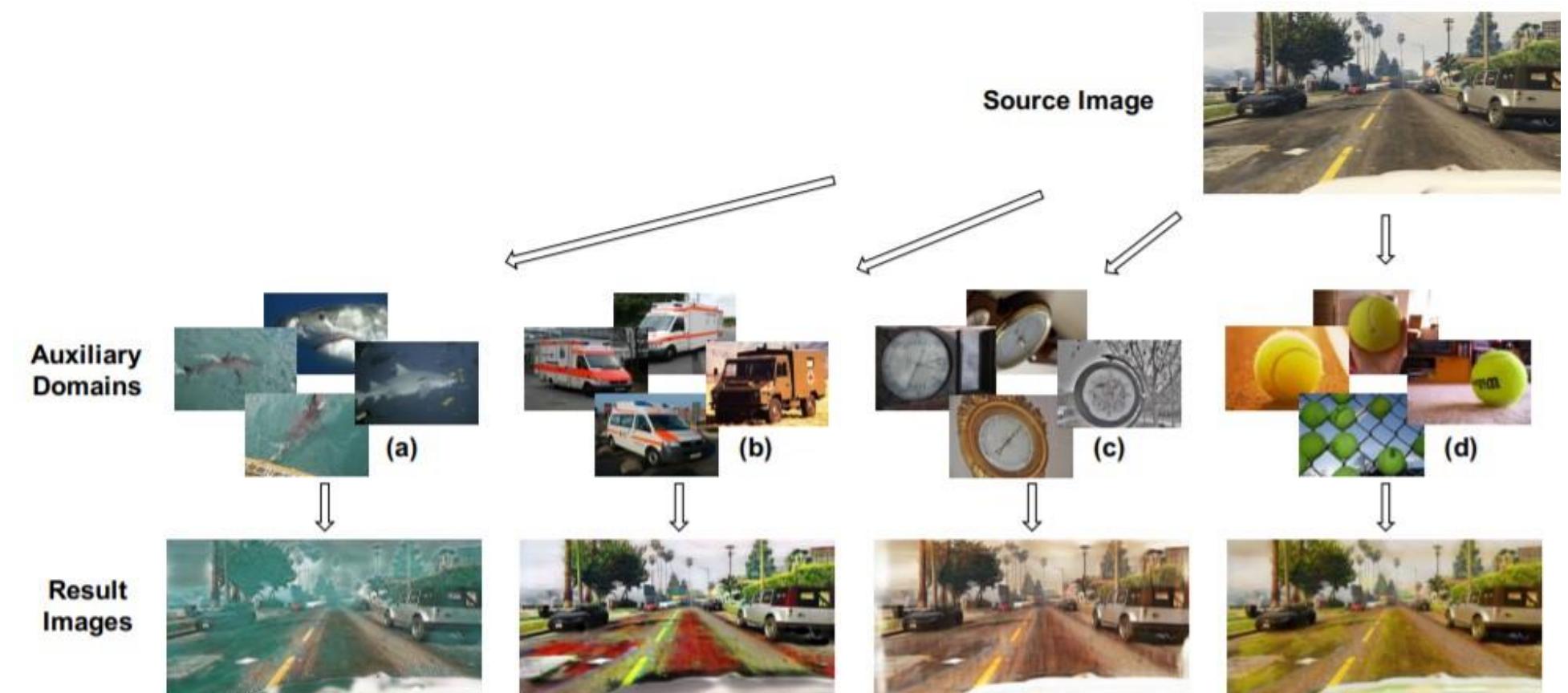
WHAT IS DATA AUGMENTATION

WIDELY USED IN MACHINE LEARNING TO GENERATE NEW TRAINING SAMPLES FOR REDUCING OVERFITTING

DOMAIN RANDOMIZATION

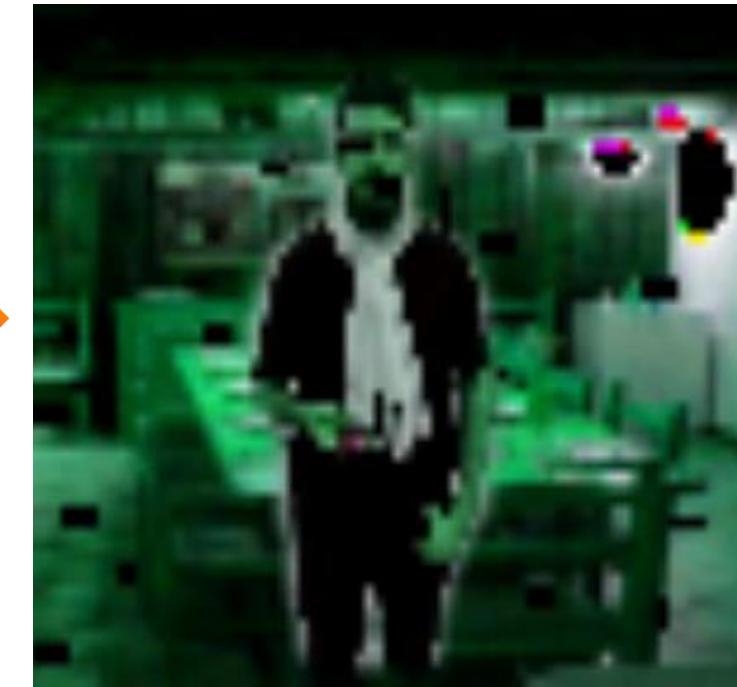
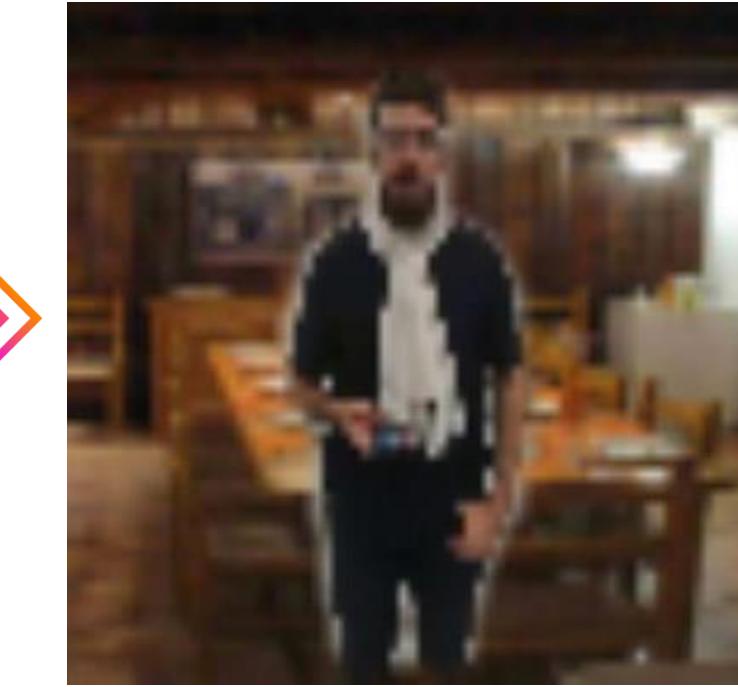
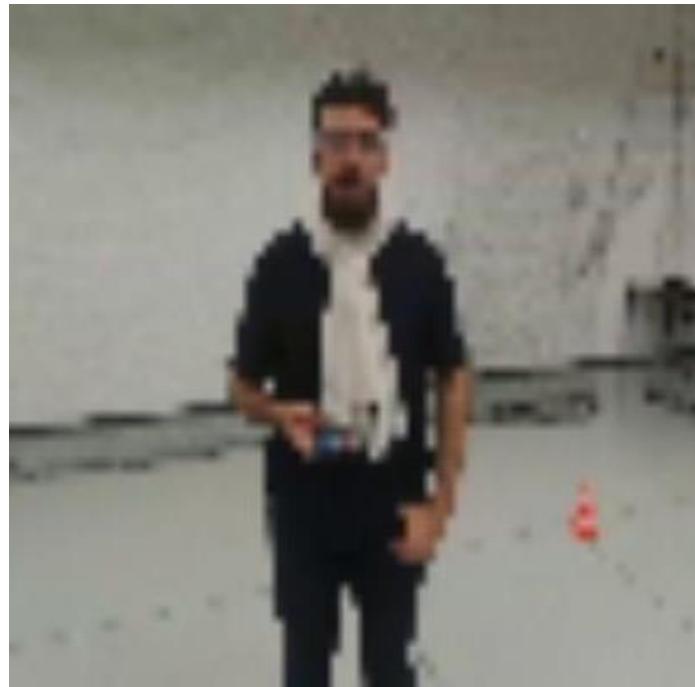


[Tobin et al. 2017]



[Yue et al. 2019]

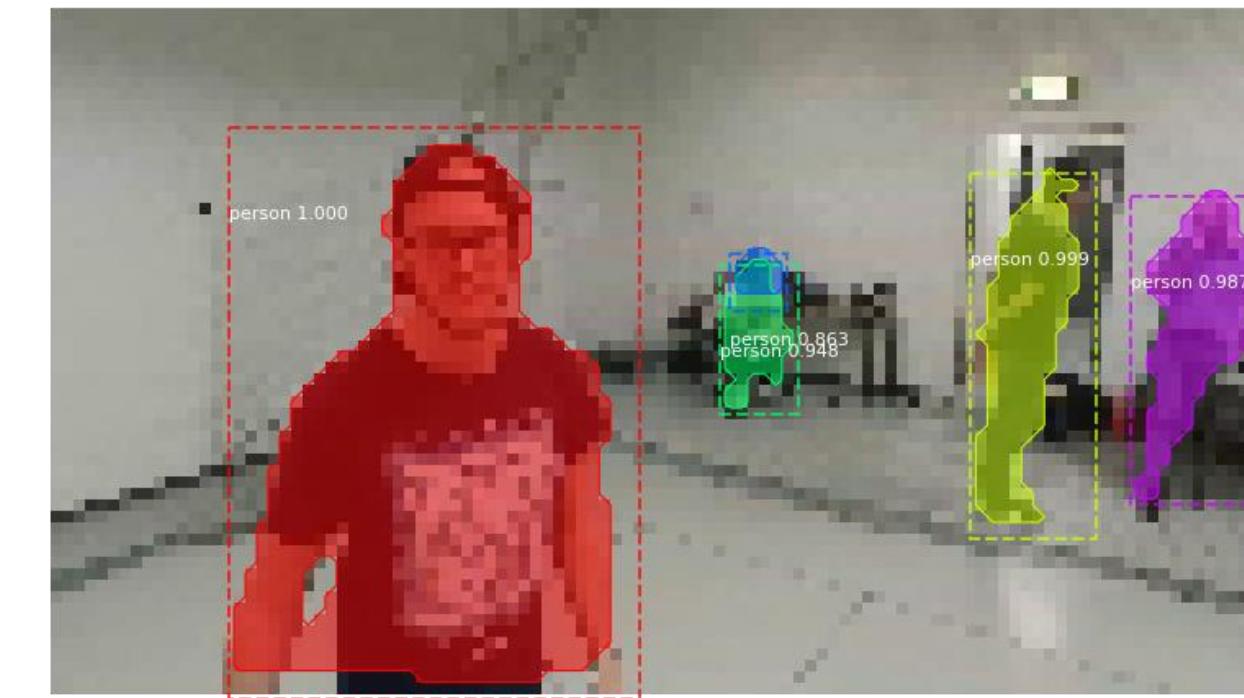
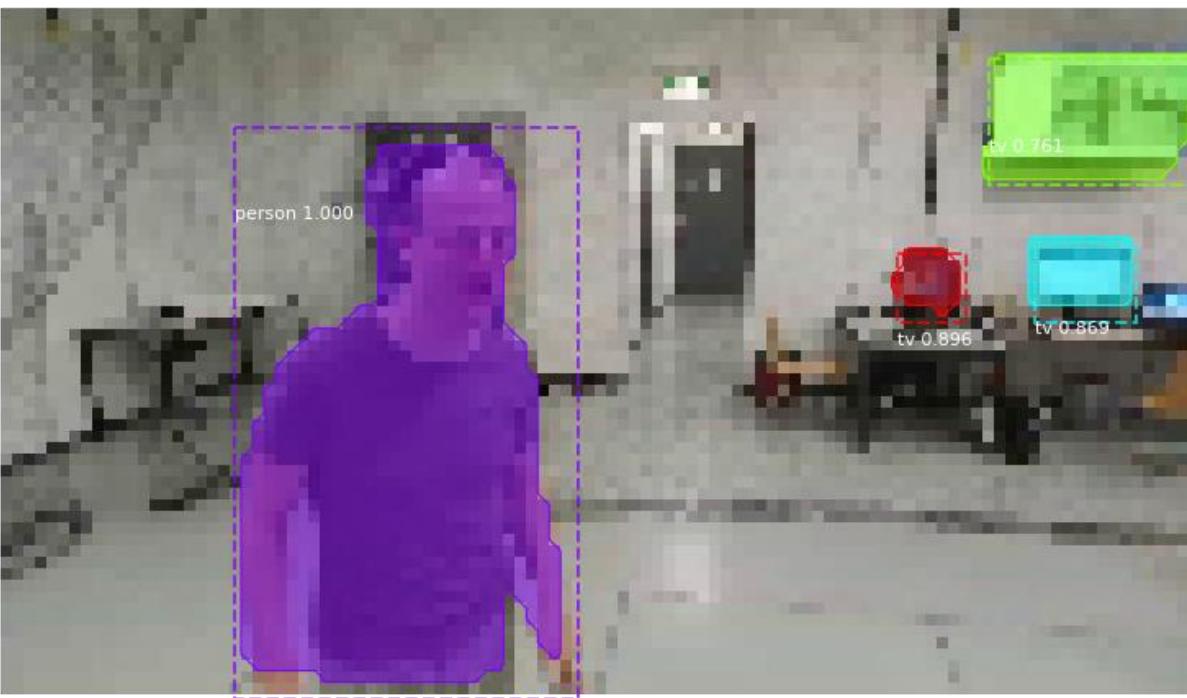
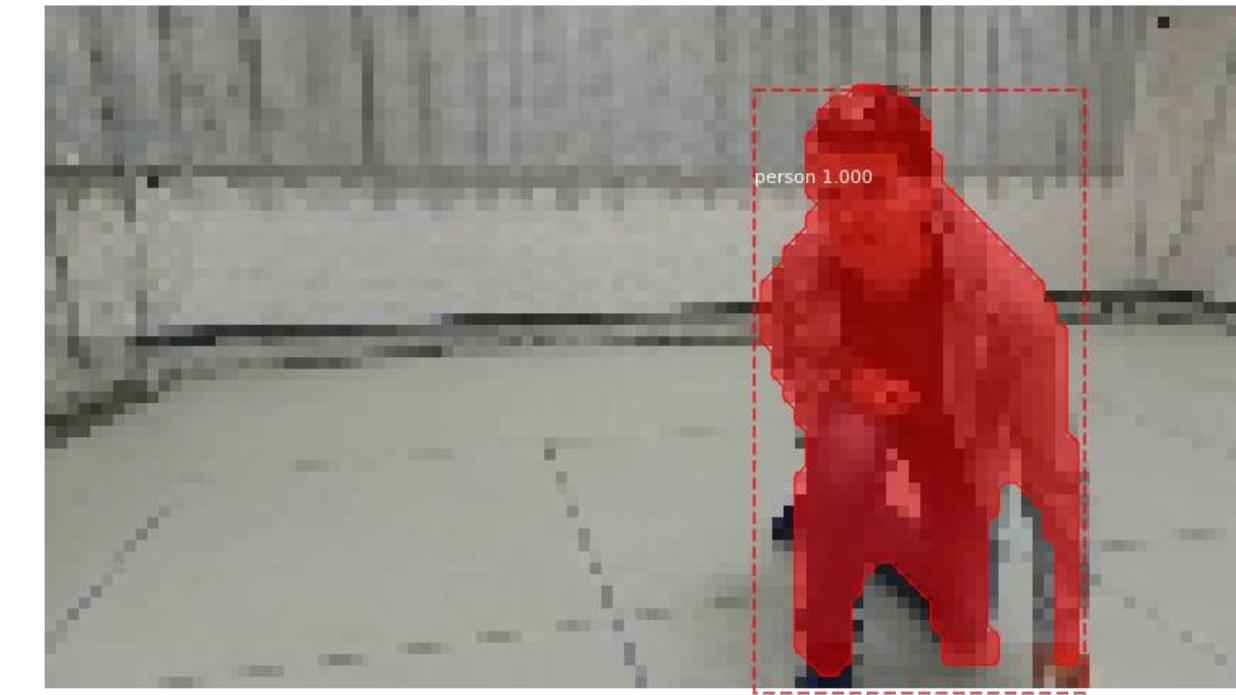
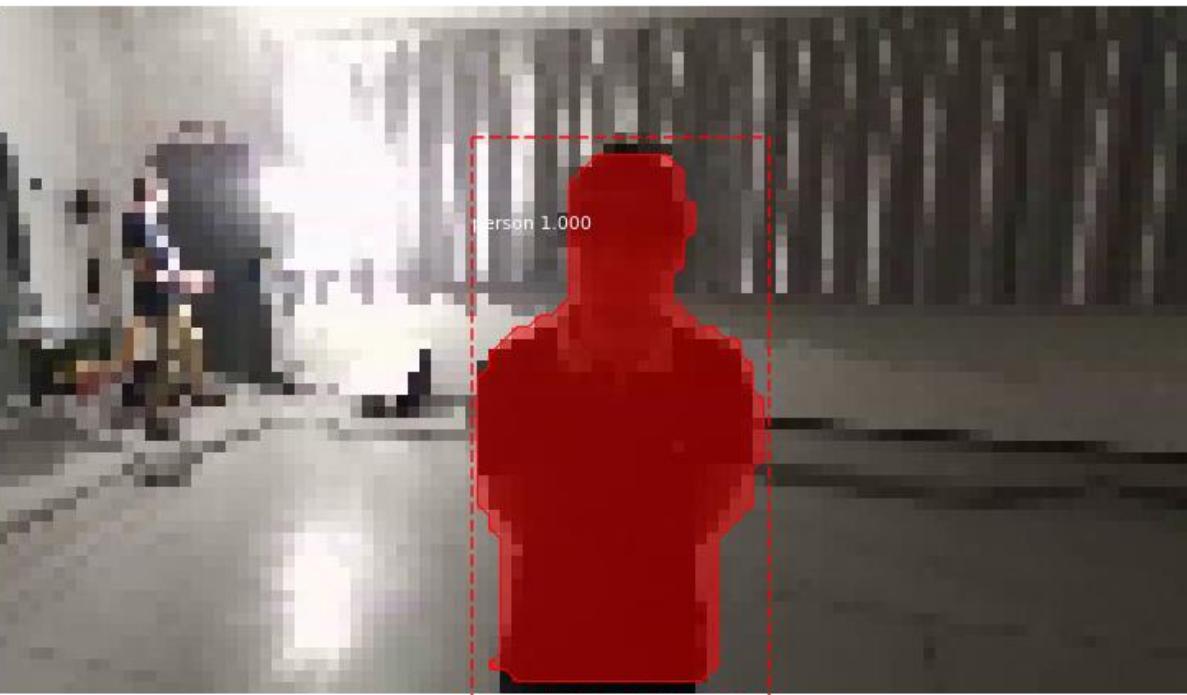
OUR AUGMENTATION STRATEGY



- 1** MASKING THE USERS IN THE IMAGES
- 2** REPLACING WITH NEW BACKGROUNDS
- 3** APPLYING CLASSIC IMAGE AUGMENTATIONS
- 4** TRAINING THE MODEL ON AUGMENTED DATASET

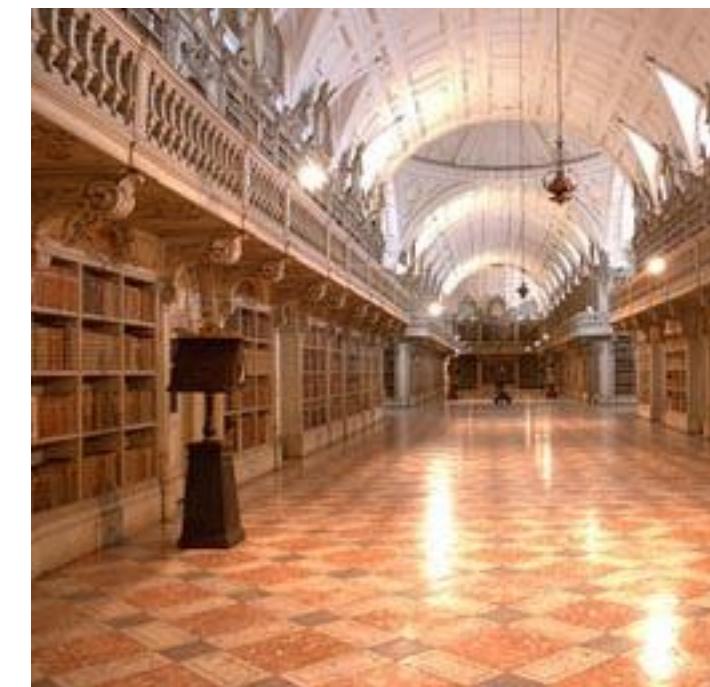
PERSON MASKING WITH MASK R-CNN

1



BACKGROUNDS DATASET

2



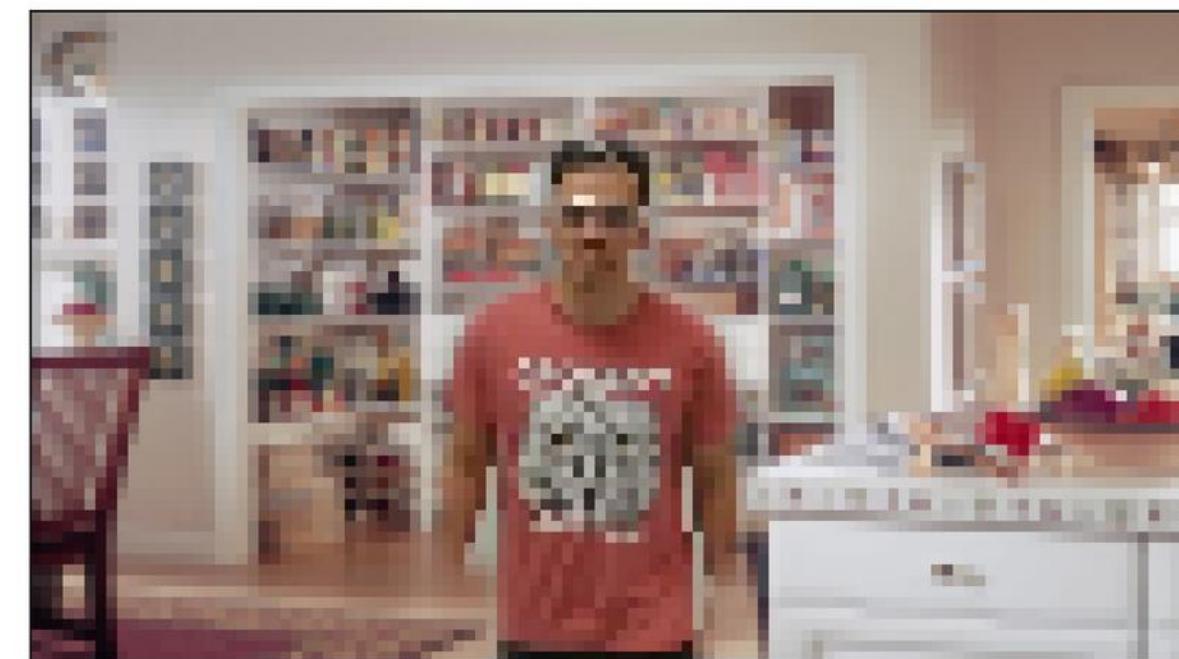
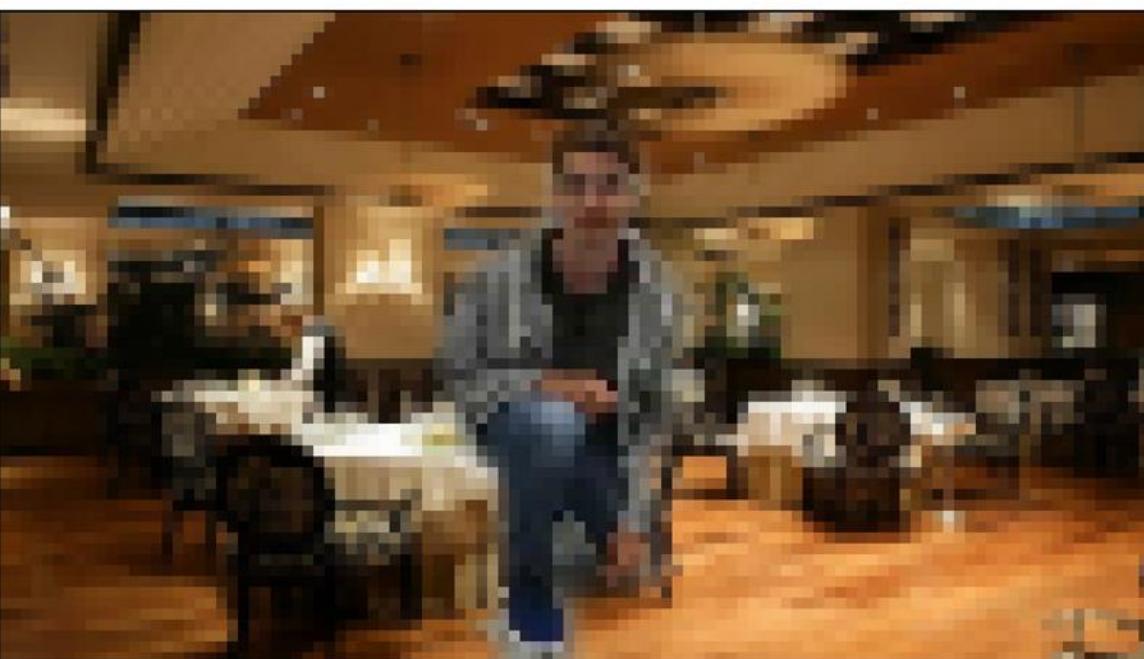
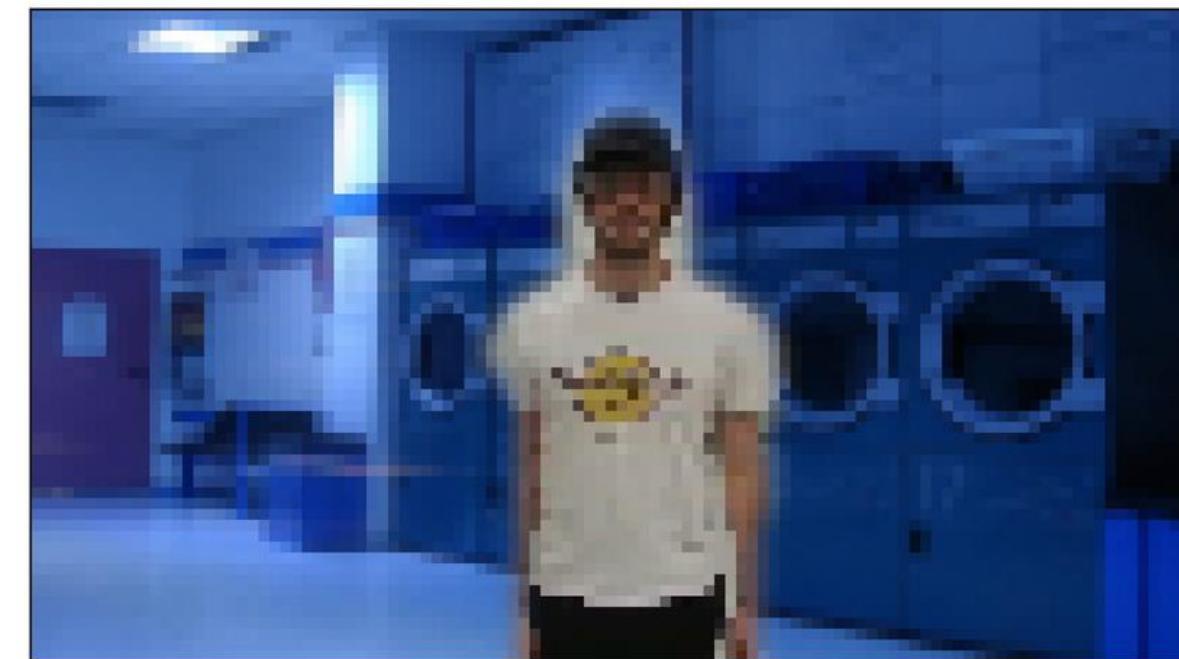
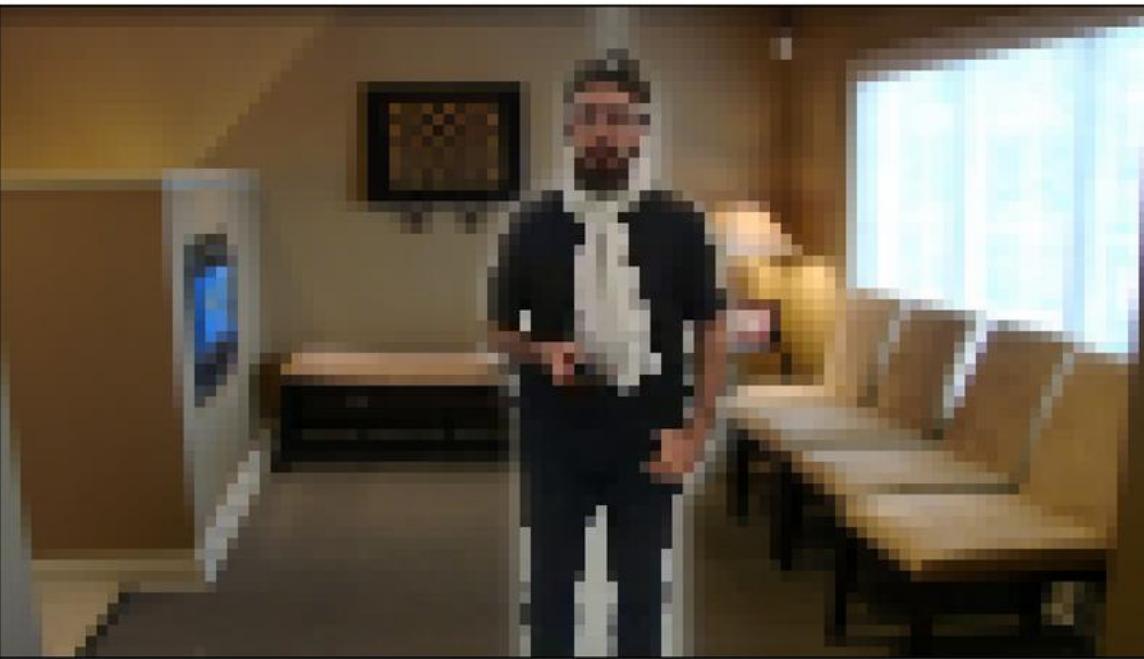
INDOOR SCENE RECOGNITION - CVPR 2009

[\[Quattoni et al. 2009\]](#)

15'620 IMAGES
67 CATEGORIES

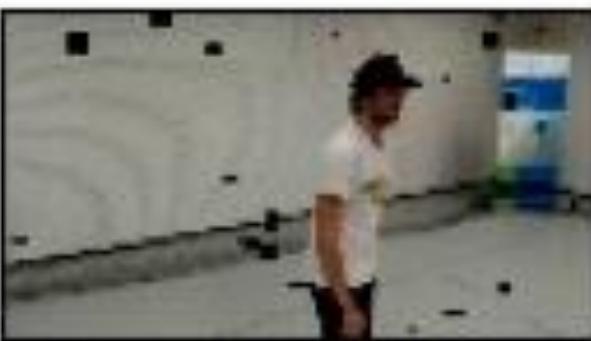
BACKGROUND REPLACEMENT

2



CLASSIC IMAGE AUGMENTATION

3



MODELS DEFINITION

4

A

TRAINED ON
ORIGINAL
DATASET

ARENA MODEL

B

TRAINED ON
BACKGROUND REPLACED
DATASET

CVPR MODEL

C

TRAINED ON
BG REPLACED & IMG
AUGMENTED DATASET

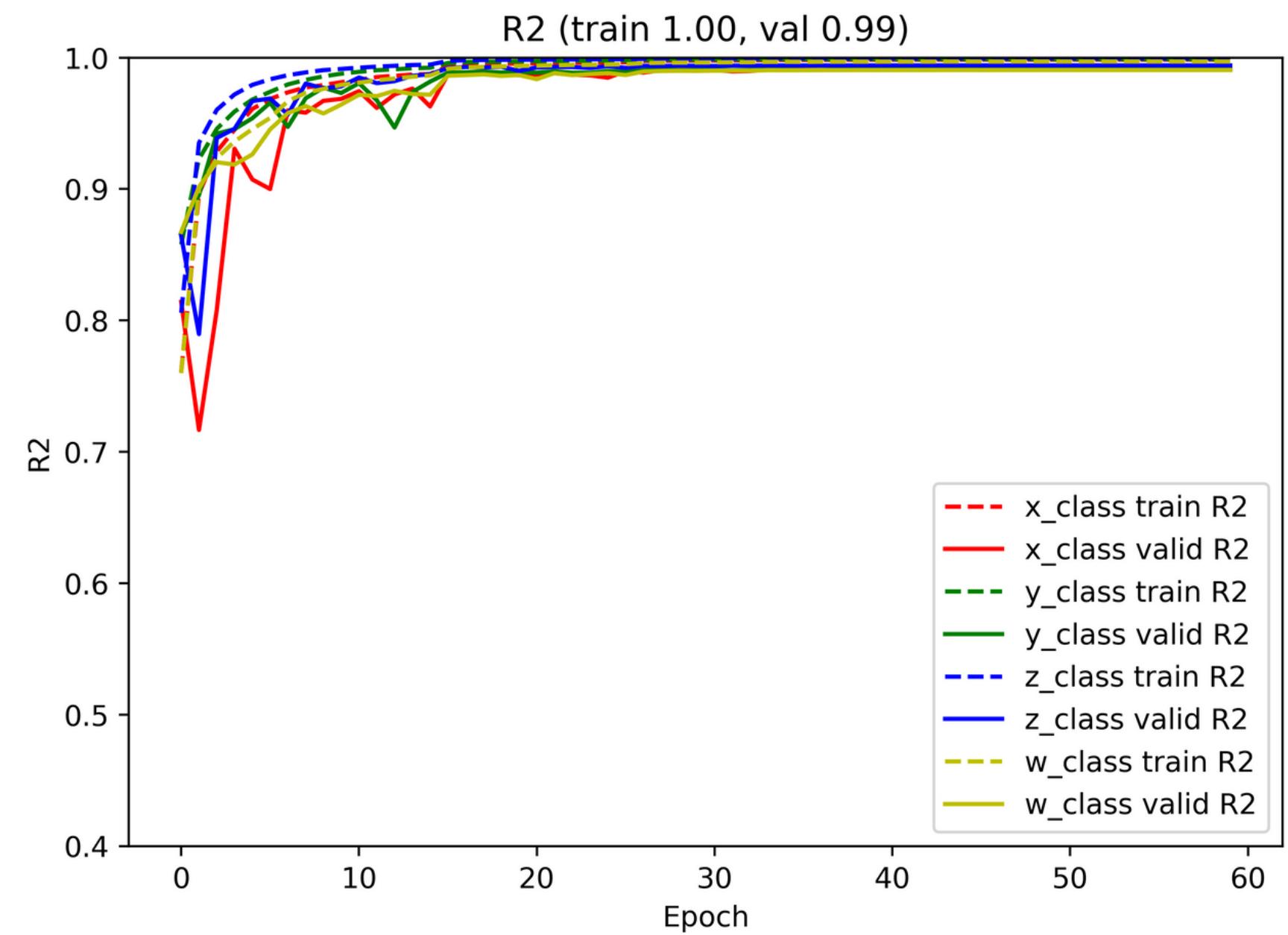
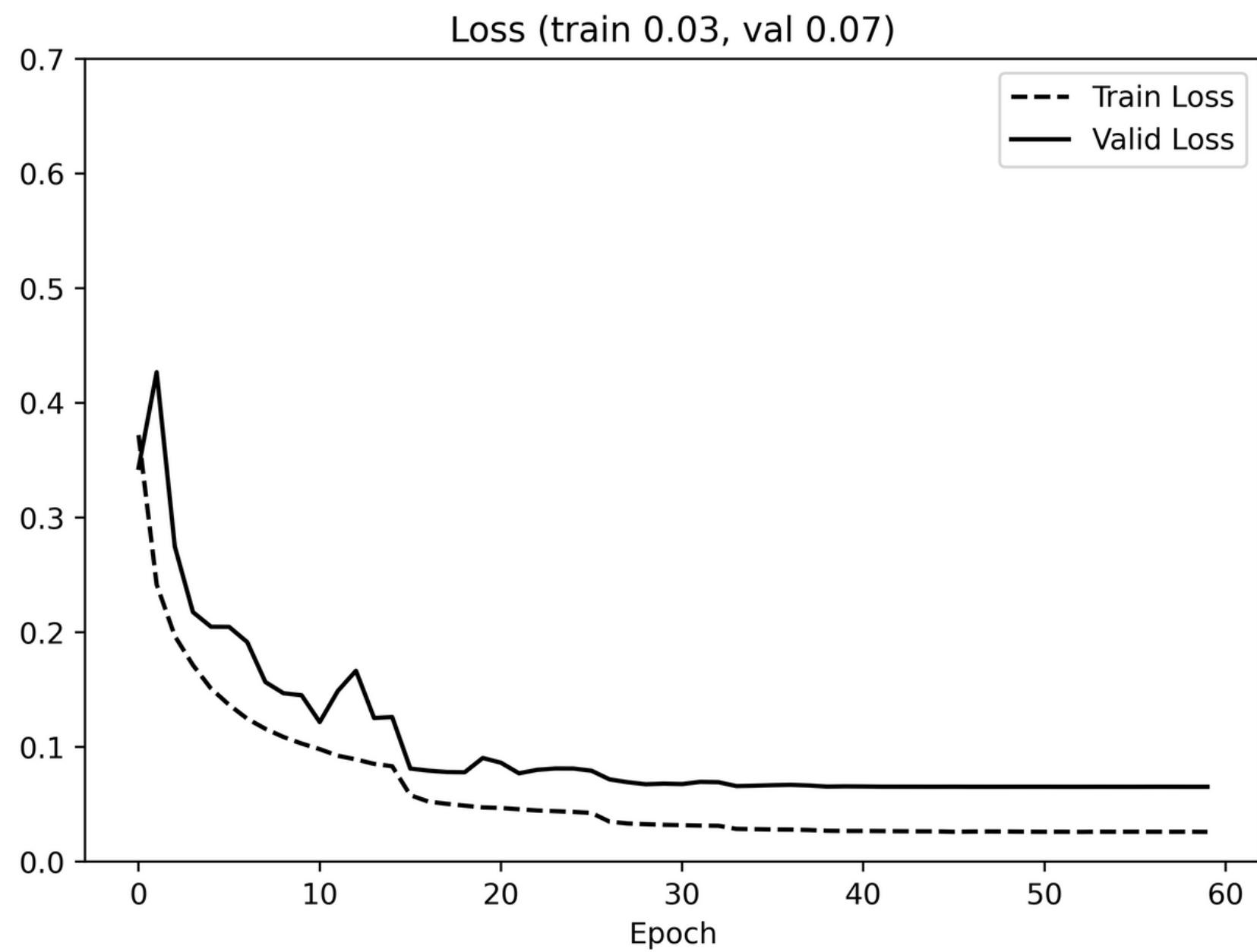
CVPR AUG MODEL

TRAINING

A

4

ARENA MODEL

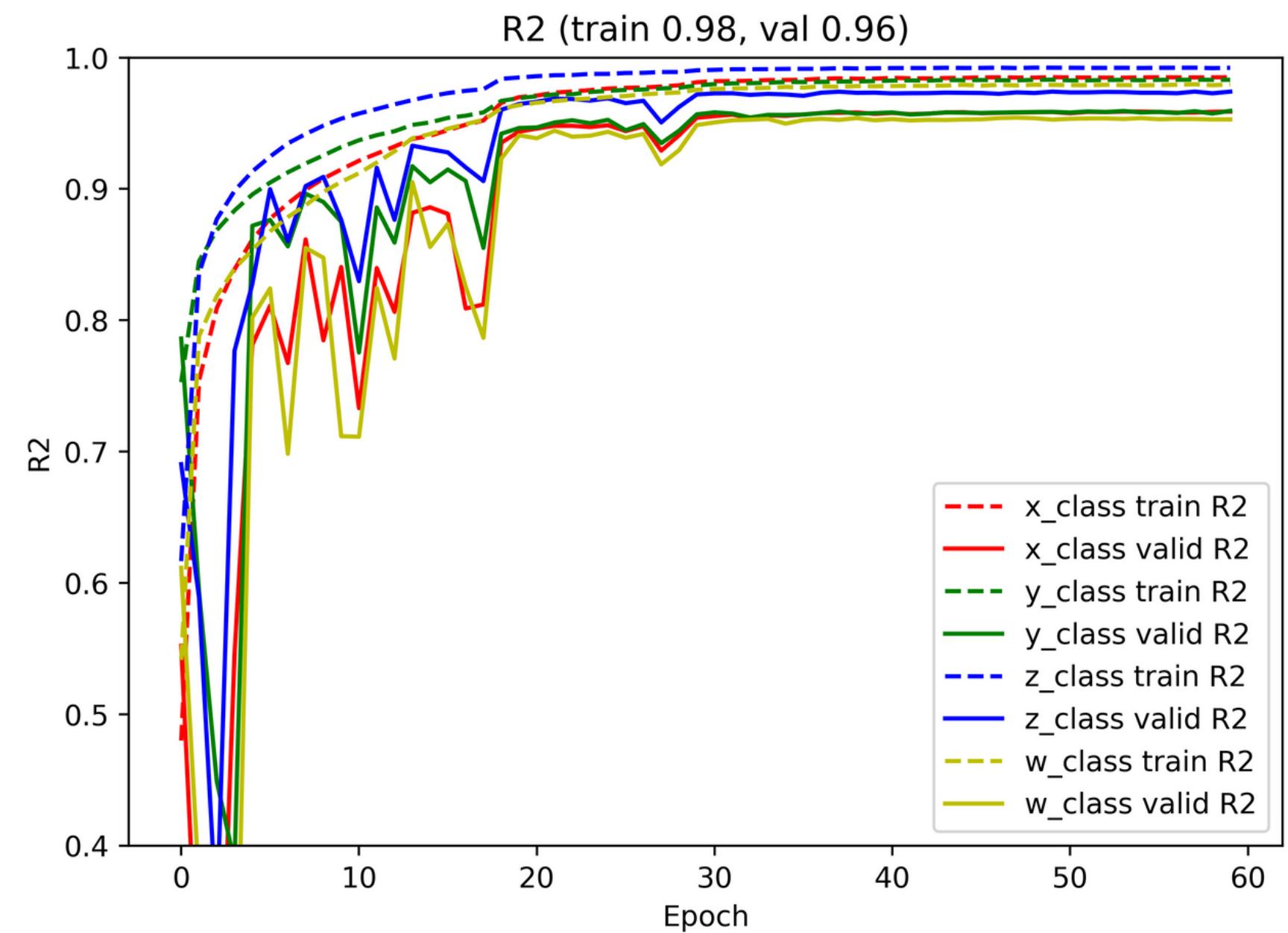
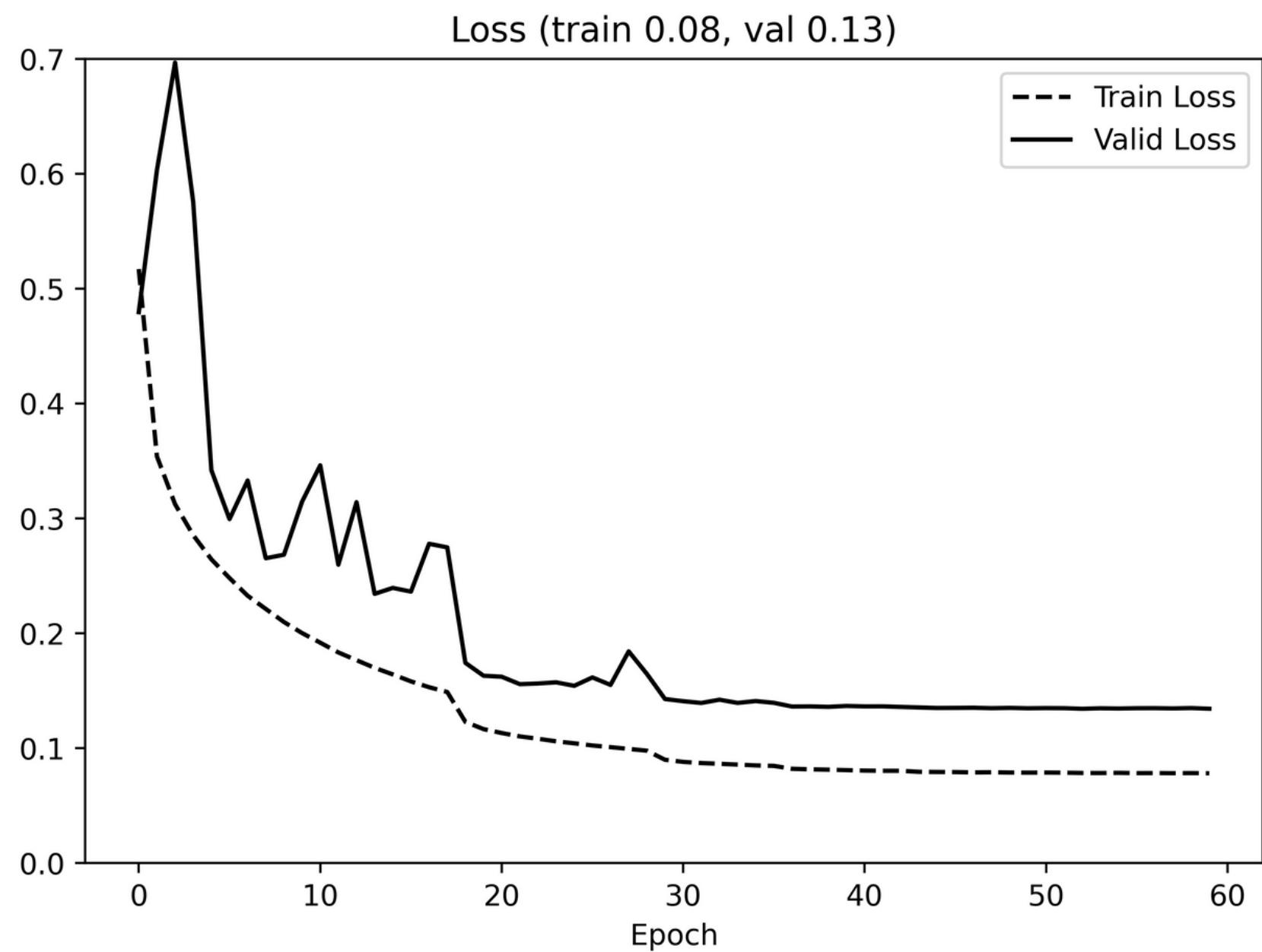


TRAINING

B

4

CVPR MODEL

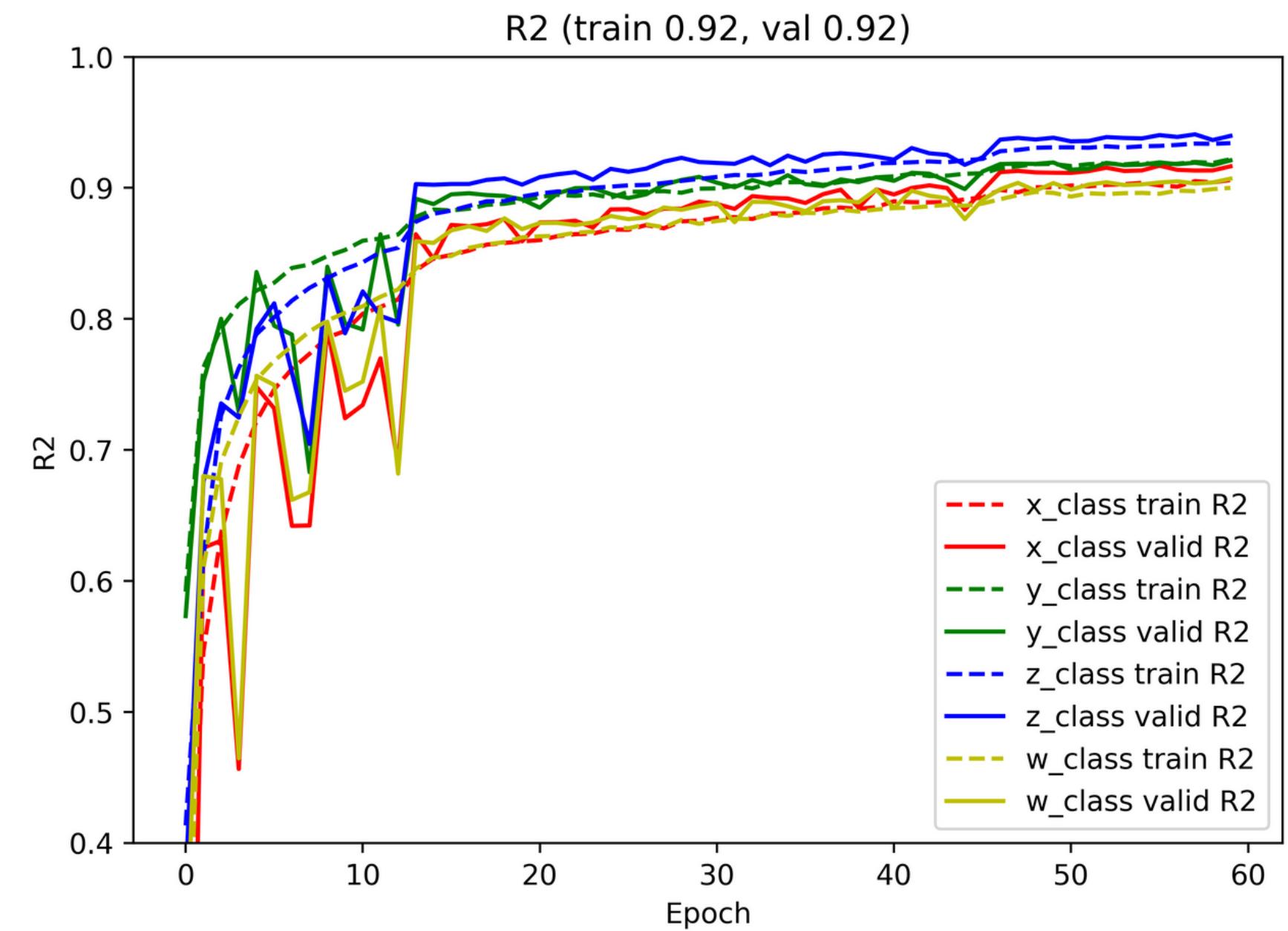
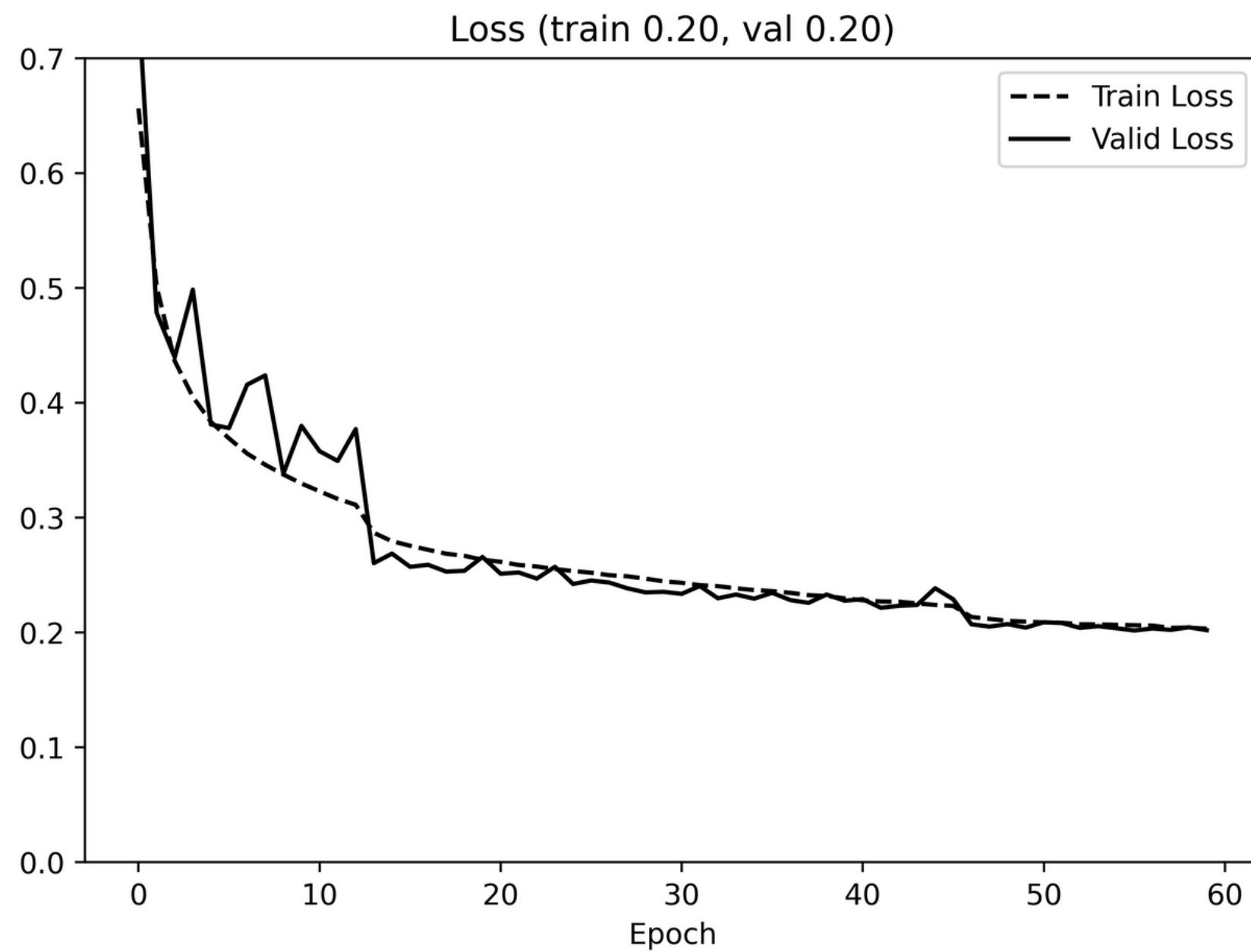


TRAINING

C

4

CVPR AUG MODEL





EXPERIMENTAL EVALUATIONS TO VALIDATE THE SOLUTION

TEST SETS DEFINITION



Original



through background replacement



through background replacement

**ARENA
TEST SET**

**INDOOR 1
TEST SET**

**INDOOR 2
TEST SET**

QUANTITATIVE EVALUATION: ARENA TEST SET

A

B

C

		Arena	CVPR	CVPR Aug
TEST LOSS (MAE)		0,41	0,42	0,36
R2	x	0,81	0,69	0,83
	y	0,86	0,82	0,87
	z	0,79	0,79	0,86
	w	0,74	0,74	0,78

QUANTITATIVE EVALUATION: INDOOR 1 TEST SET

A

B

C

		Arena	CVPR	CVPR Aug
TEST LOSS (MAE)		0,86	0,44	0,37
R2	x	0,16	0,58	0,83
	y	0,30	0,81	0,83
	z	0,21	0,78	0,85
	w	0,08	0,70	0,76

QUANTITATIVE EVALUATION: INDOOR 2 TEST SET

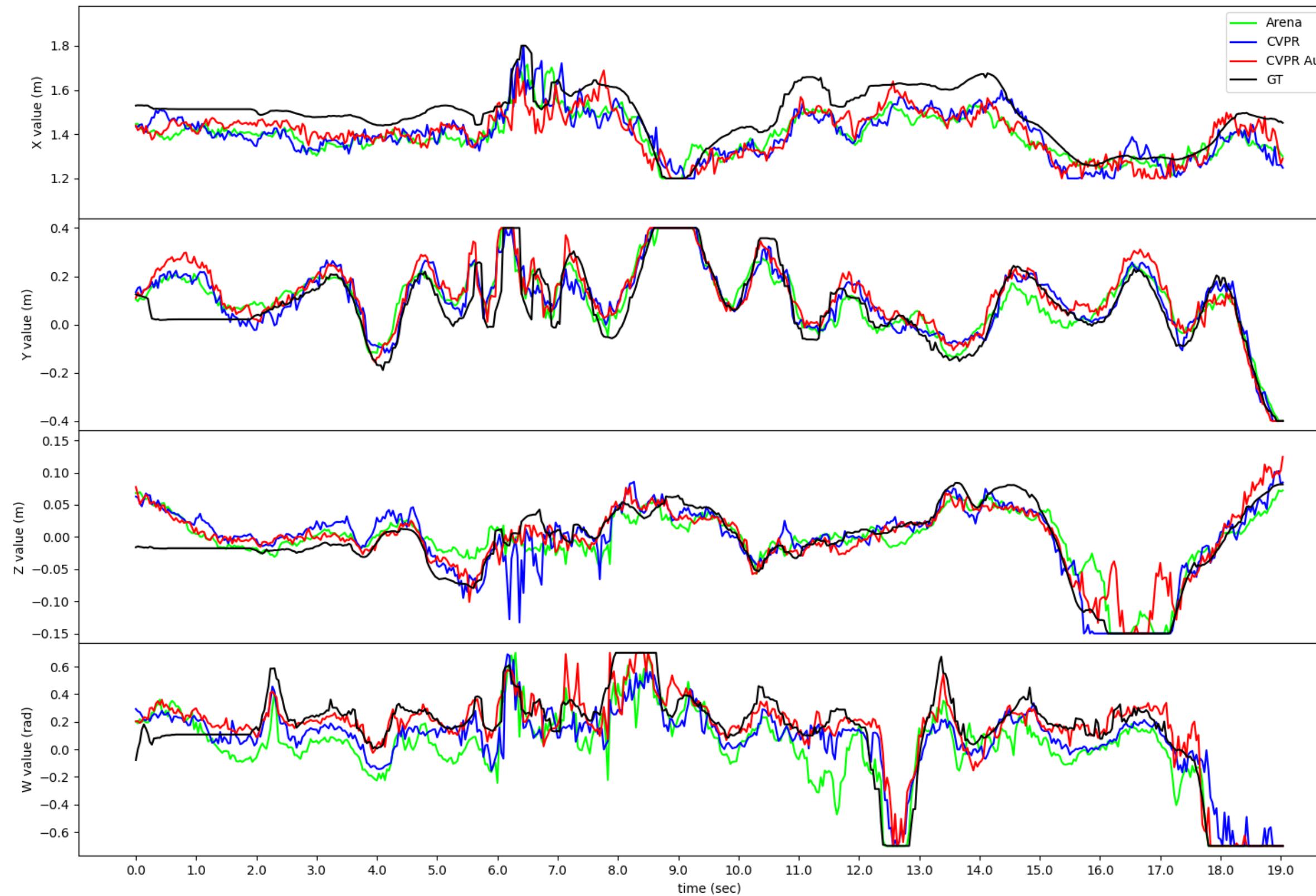
A

B

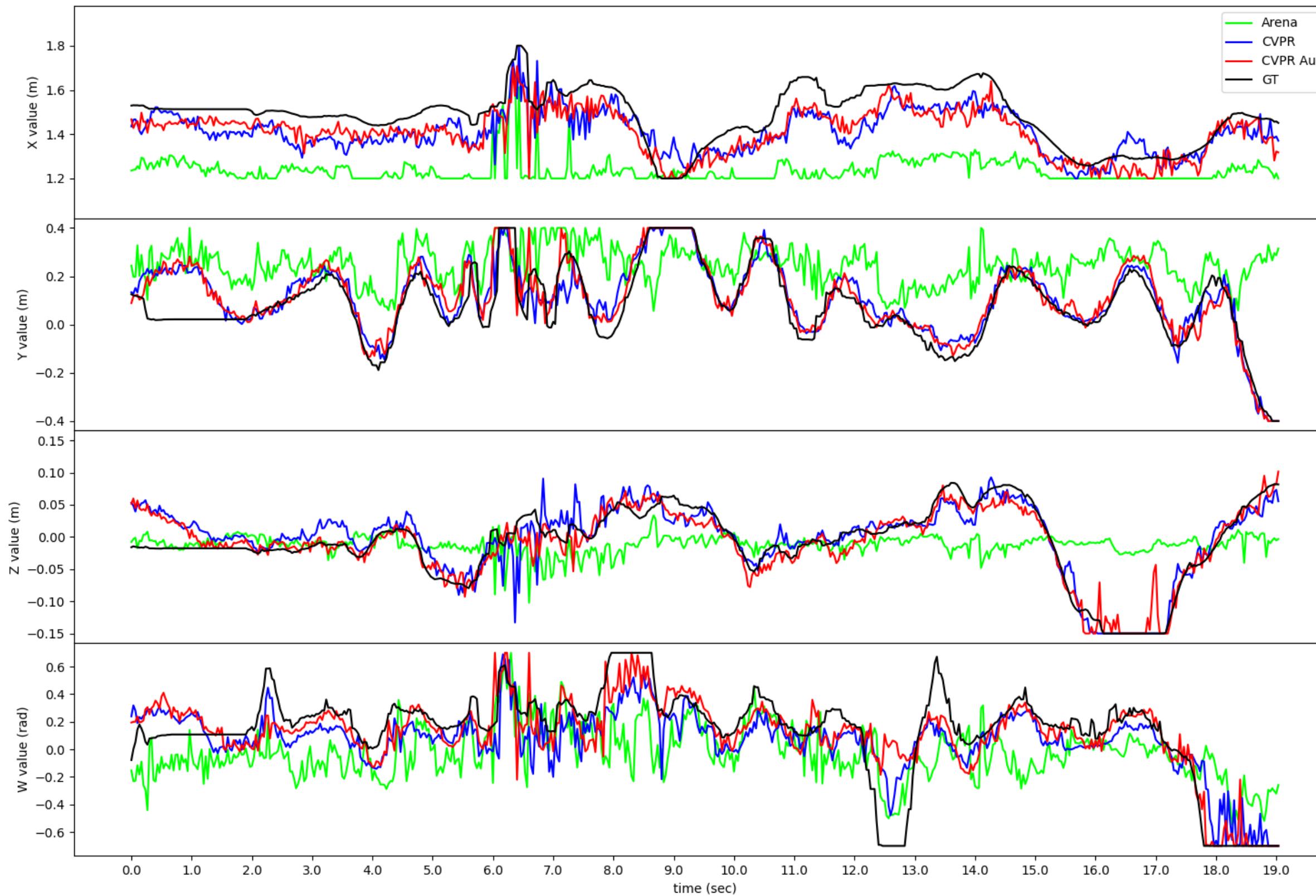
C

		Arena	CVPR	CVPR Aug
TEST LOSS (MAE)		1,00	0,45	0,37
R2	x	-0,86	0,64	0,81
	y	0,51	0,82	0,84
	z	-2,78	0,77	0,86
	w	0,24	0,67	0,75

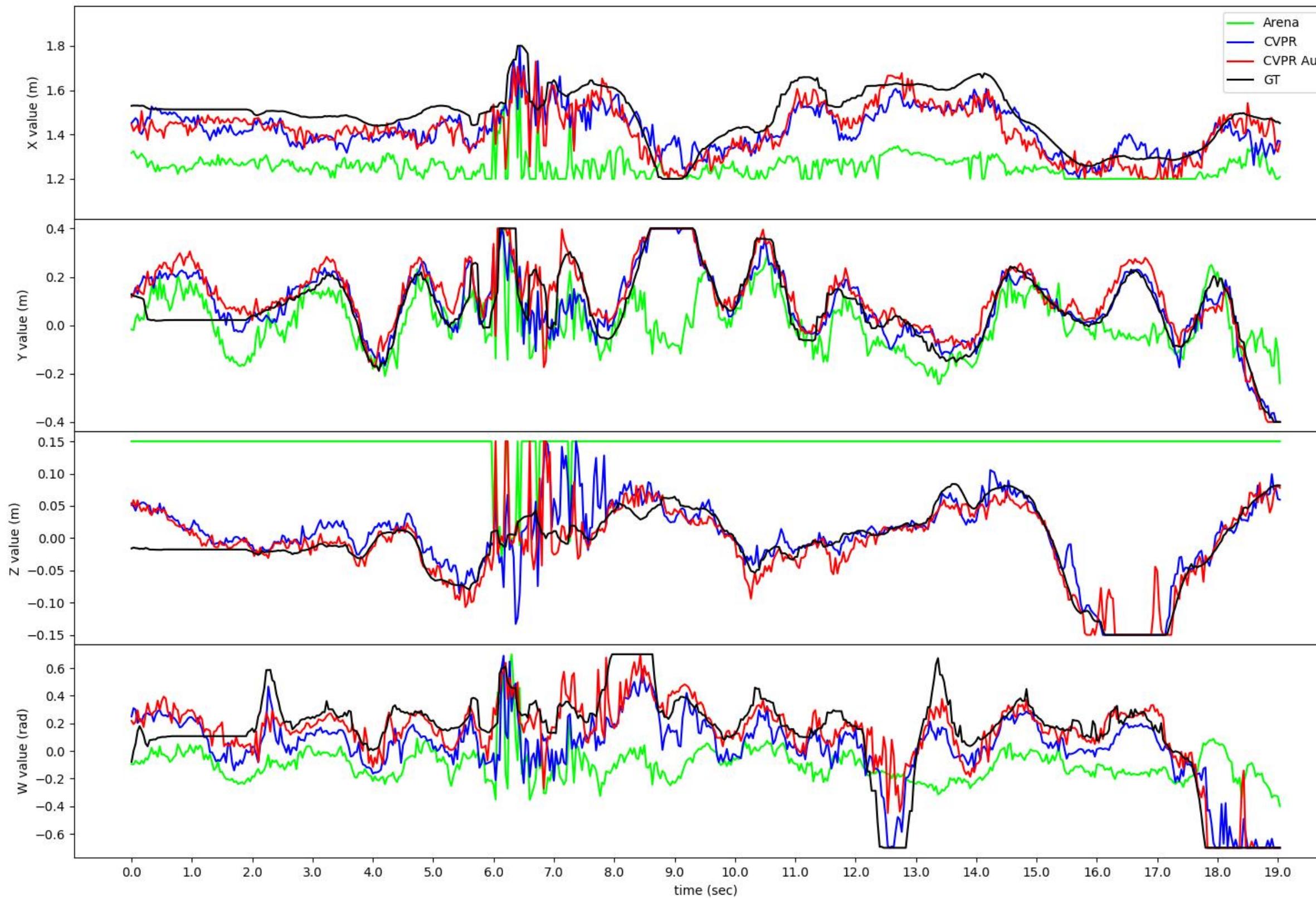
TIMELINE EVALUATION: ARENA TEST SET



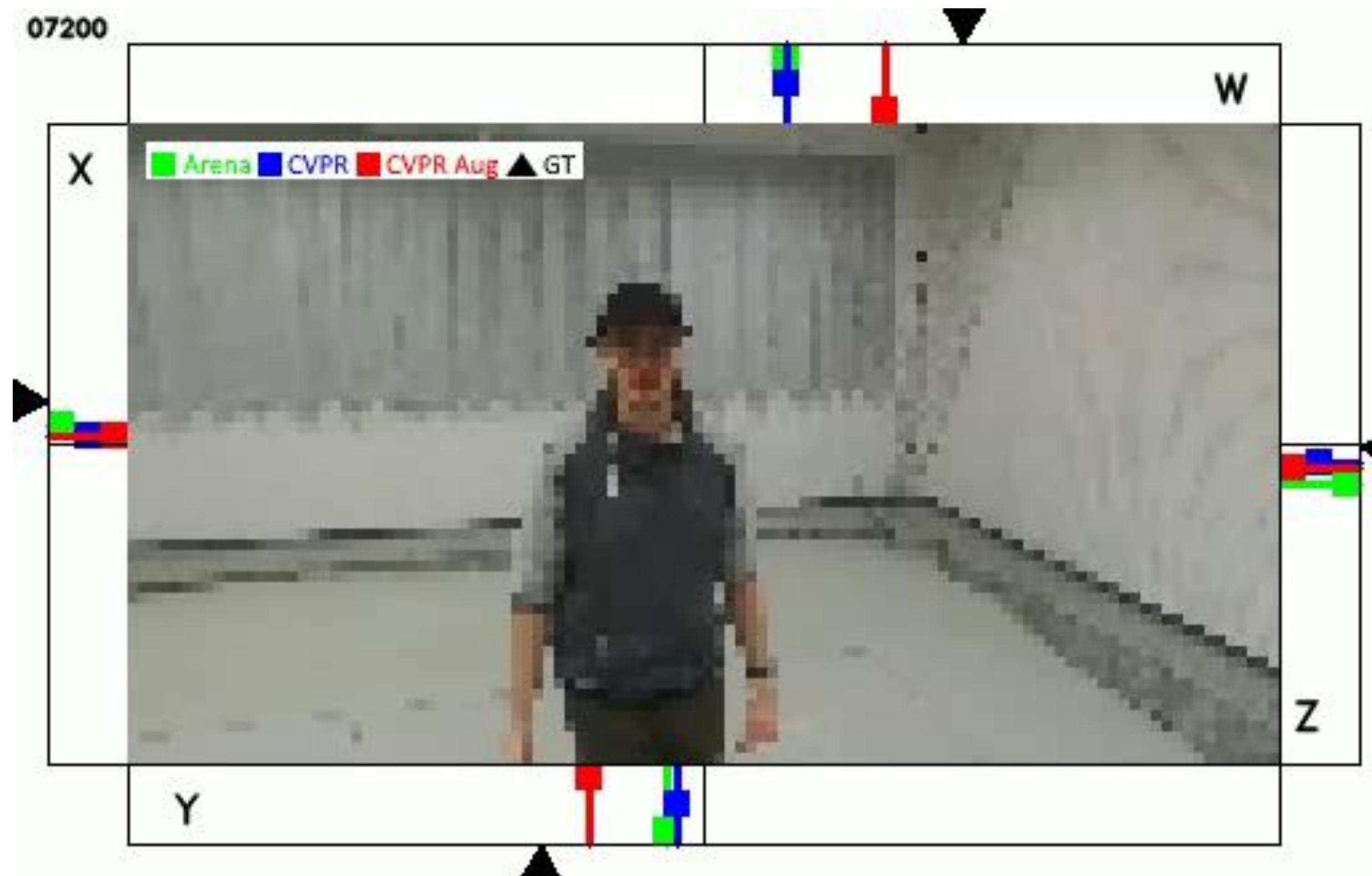
TIMELINE EVALUATION: INDOOR 1 TEST SET



TIMELINE EVALUATION: INDOOR 2 TEST SET



PER-FRAME EVALUATION: ARENA TEST SET



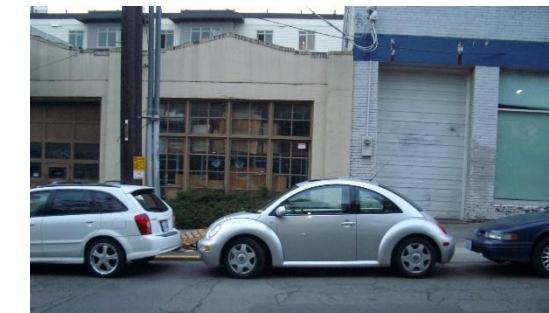
PER-FRAME EVALUATION: INDOOR 1 TEST SET



PER-FRAME EVALUATION: INDOOR 2 TEST SET



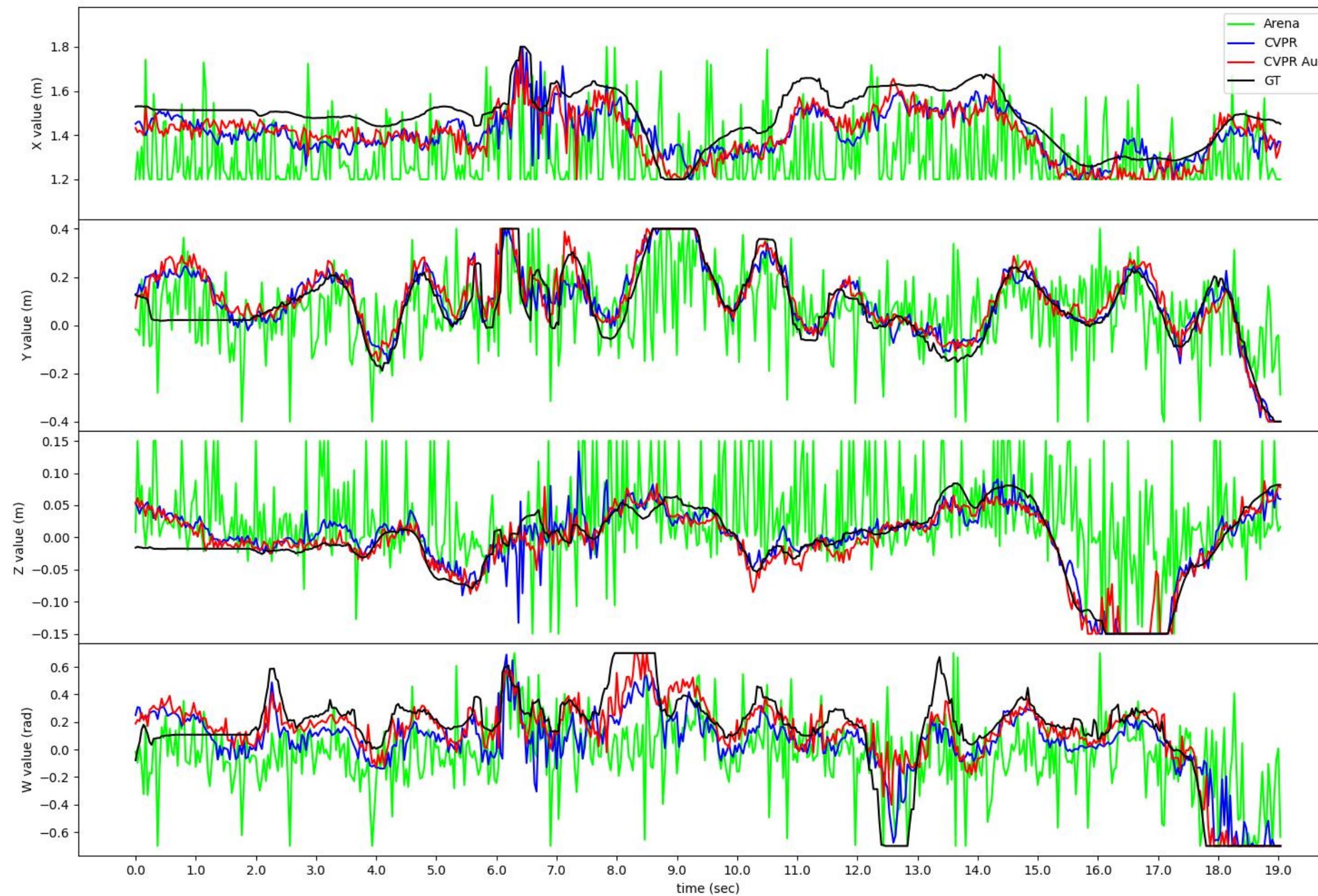
ADDITIONAL TEST SET



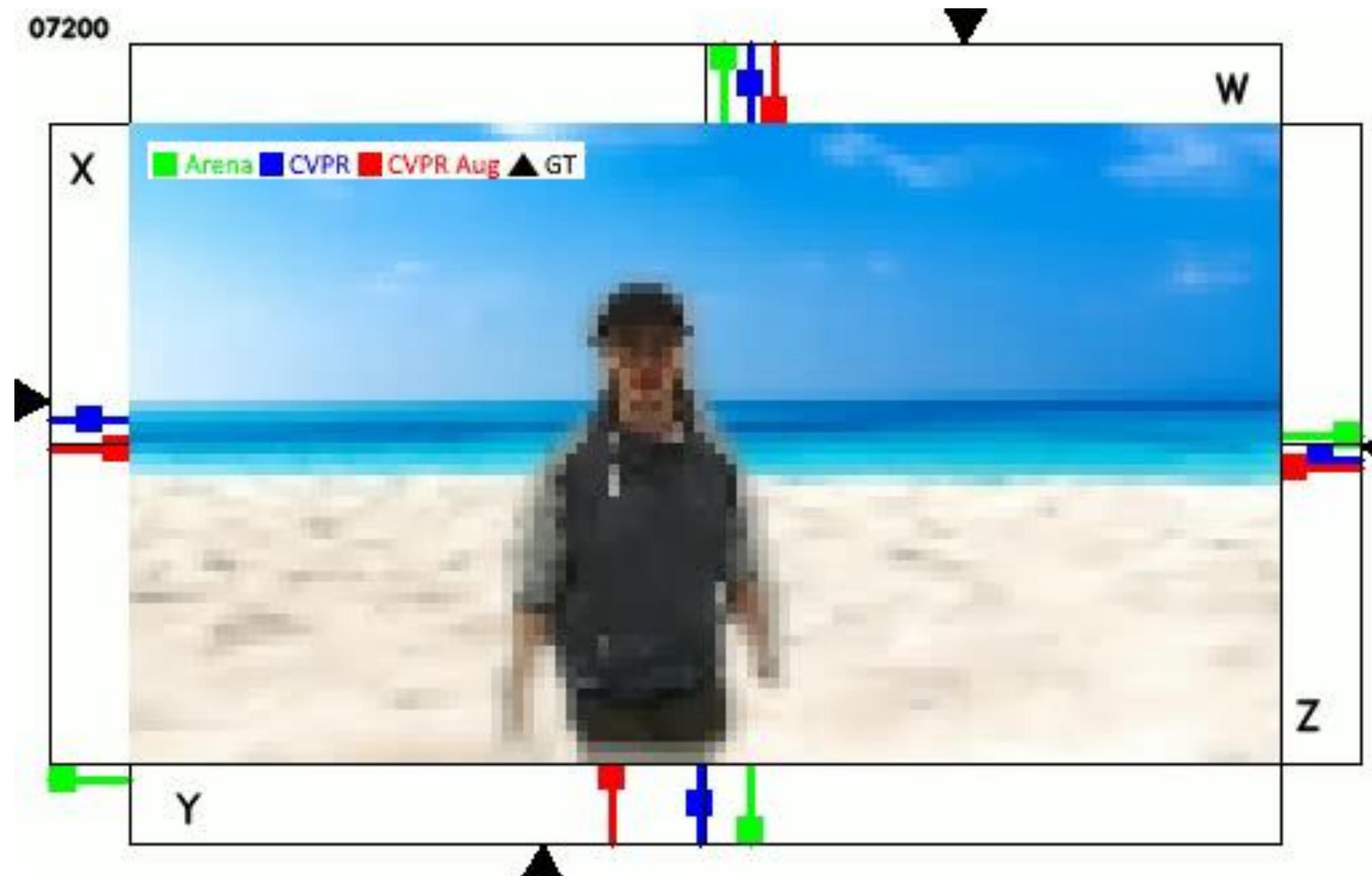
many more ...

MIXED TEST SET

TIMELINE EVALUATION: MIXED TEST SET



PER-FRAME EVALUATION: MIXED TEST SET



SMARTPHONE EVALUATION: OVERVIEW



SMARTPHONE EVALUATION: VERTICAL ALIGNMENT



SMARTPHONE EVALUATION: USER'S DISTANCE



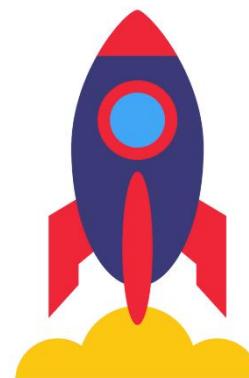
CONCLUSIONS



QUANTITATIVE AND QUALITATIVE EVALUATION PROVED THAT THE AUGMENTATION STRATEGY IS ABLE TO ENHANCE MODEL'S GENERALIZATION



BOTH BACKGROUND REPLACEMENT AND CLASSIC IMAGE AUGMENTATION ARE REQUIRED TO OBTAIN ROBUST RESULTS



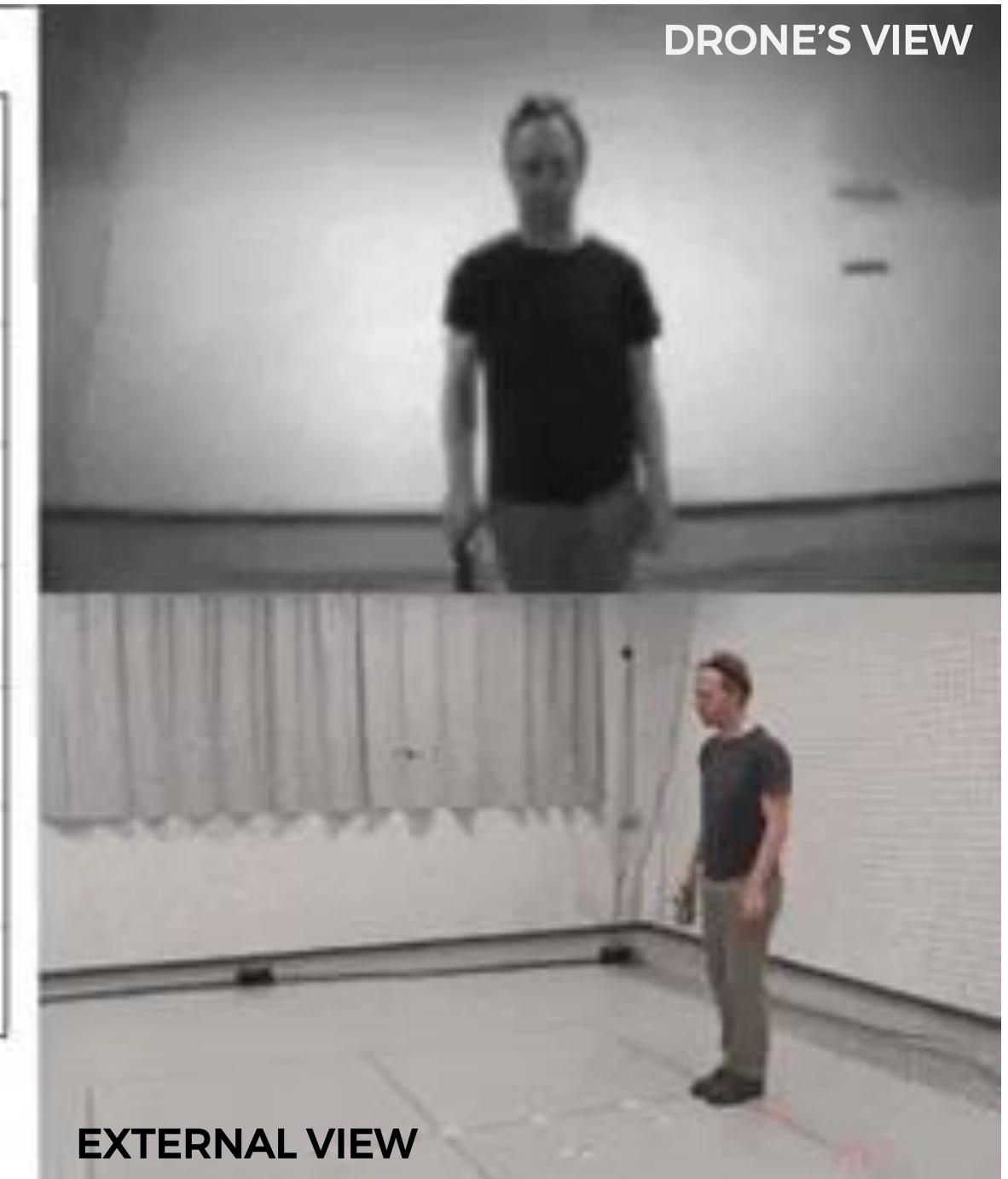
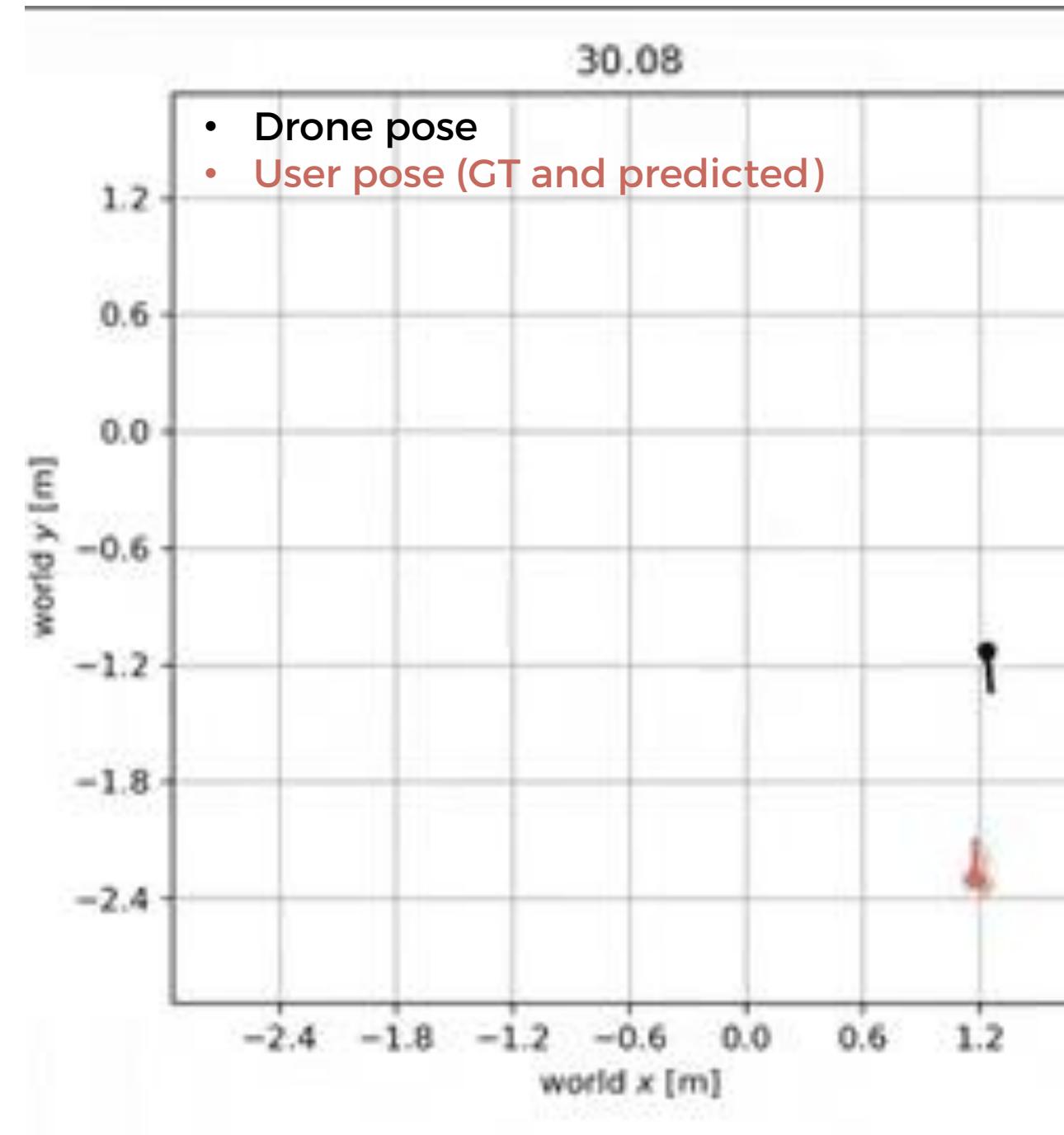
OUR APPROACH IS EXTENDABLE TO OTHER TASKS WHICH REQUIRES OBJECT OR HUMAN RECOGNITION

FUTURE WORKS

- TEST THE MODEL ON THE REAL DRONE (Parrot Bebop 2)
- FURTHER INTERPRETATION ON THE NEW MODEL
- TEST THE AUGMENTATION STRATEGY ON DIFFERENT DRONES



FUTURE WORKS: CRAZYFLIE





THANKS FOR YOUR KIND ATTENTION

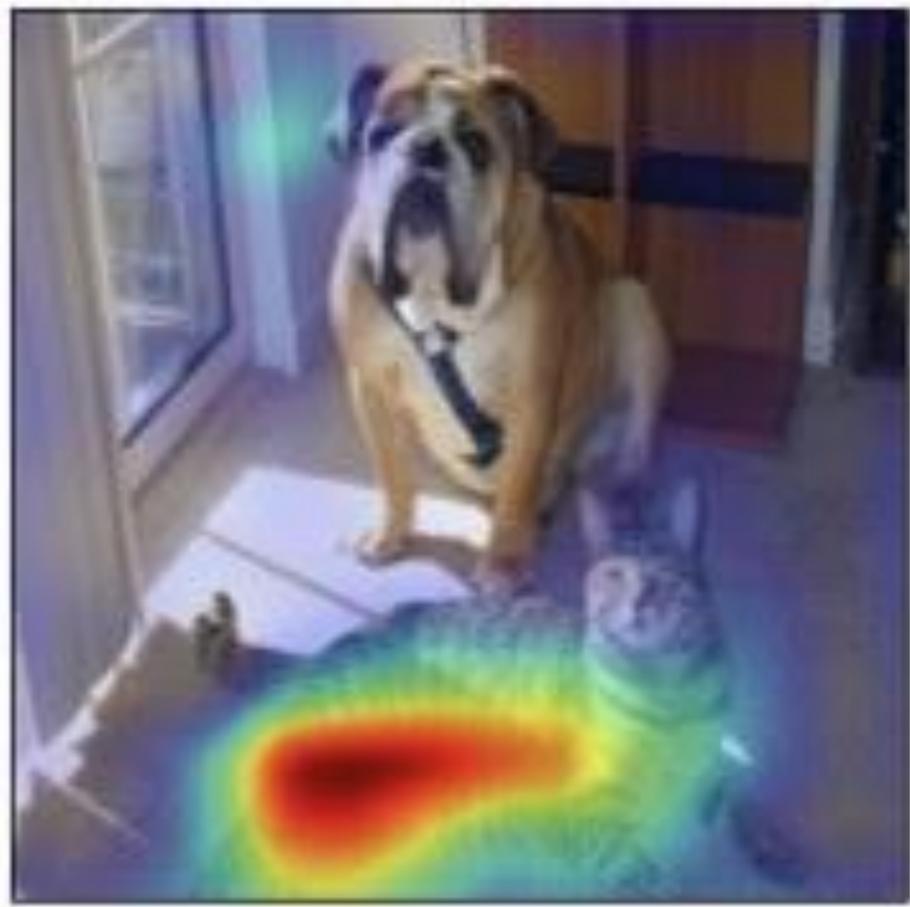
Supervisor: Dr. Alessandro Giusti
Co-Supervisor: PhD Stud. Dario Mantegazza

Master Thesis by
MARCO FERRI
ID 80730

ADDITIONAL SLIDES

HOW GRAD-CAM WORKS

Grad-CAM for "Cat"

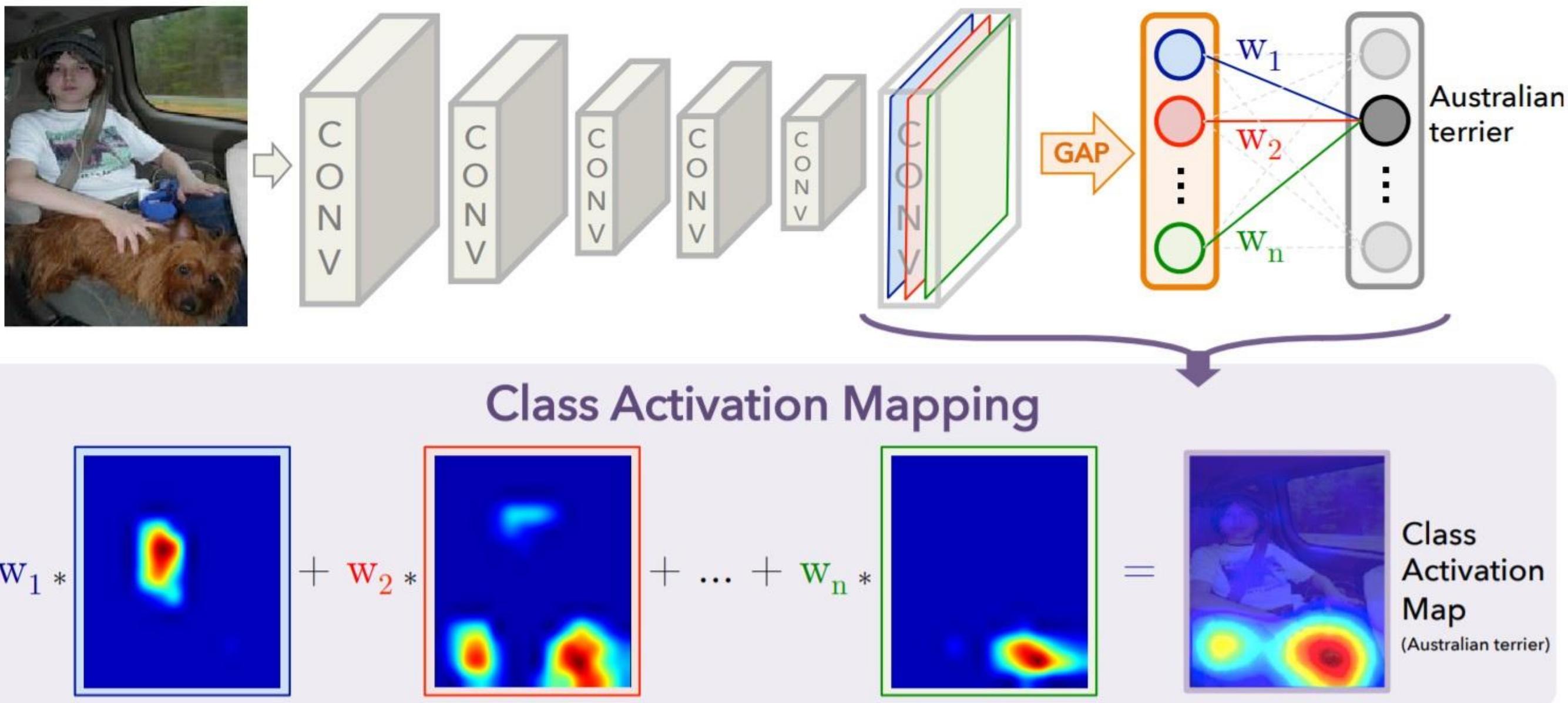


[Selvaraju et al.
2019]

Grad-CAM for "Dog"

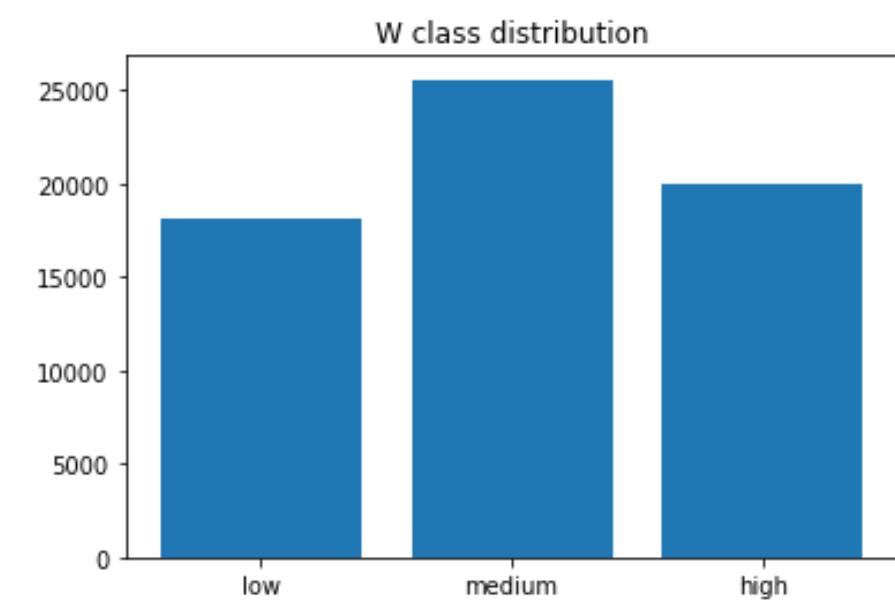
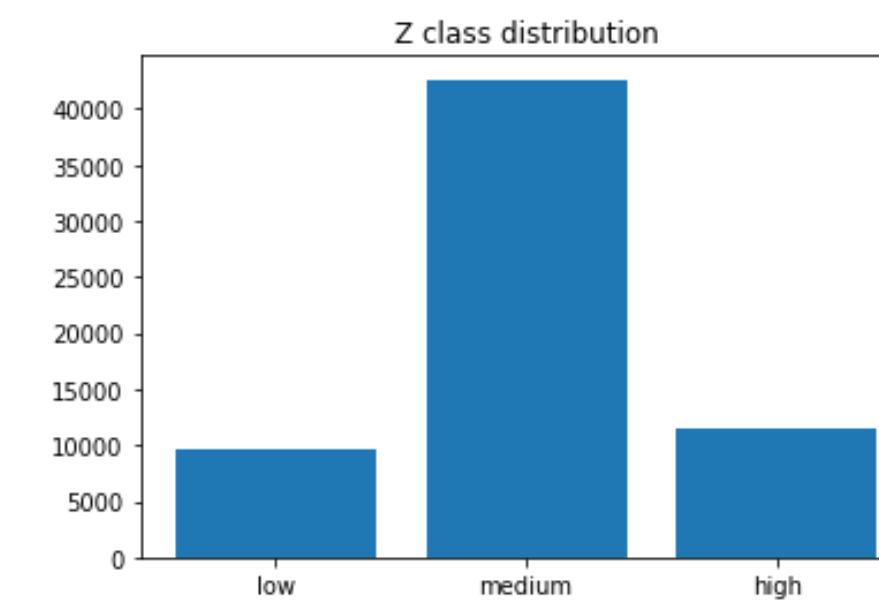
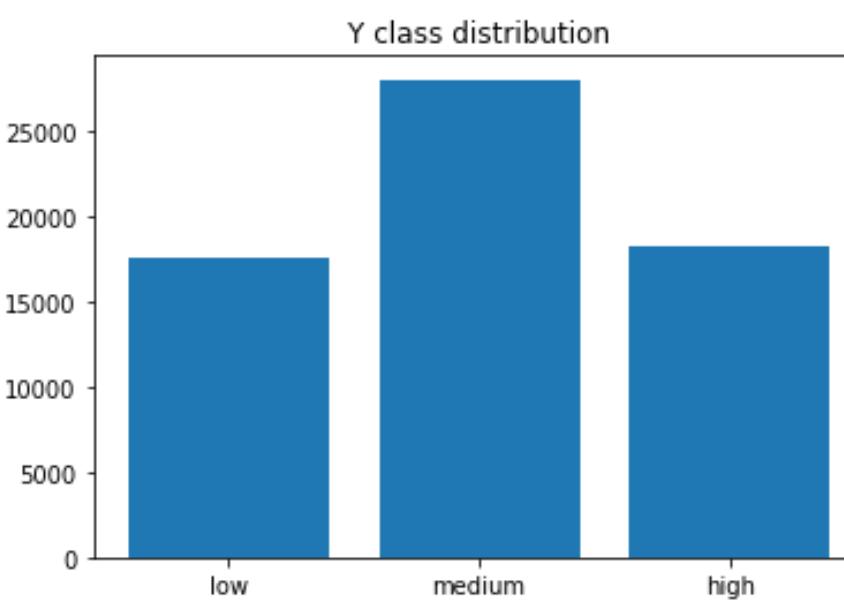
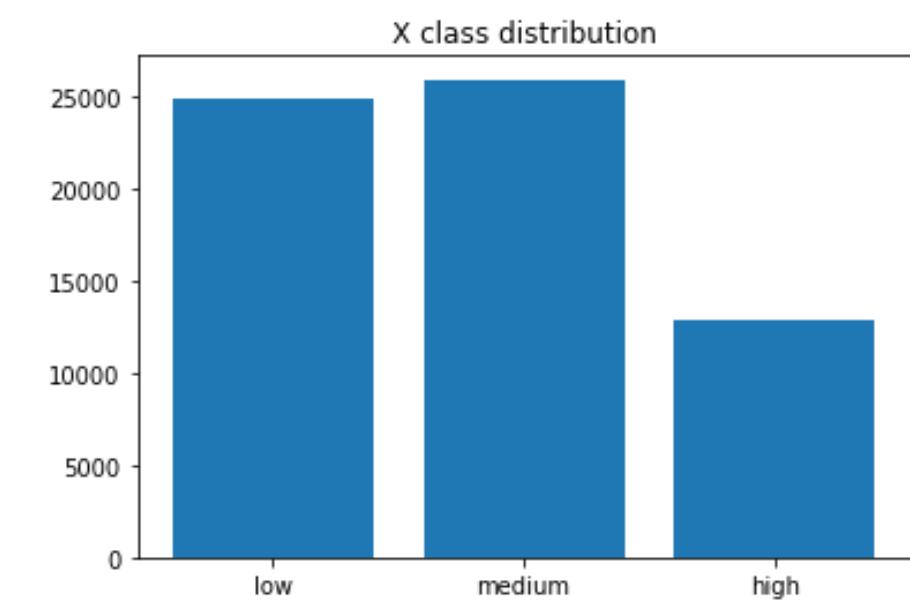
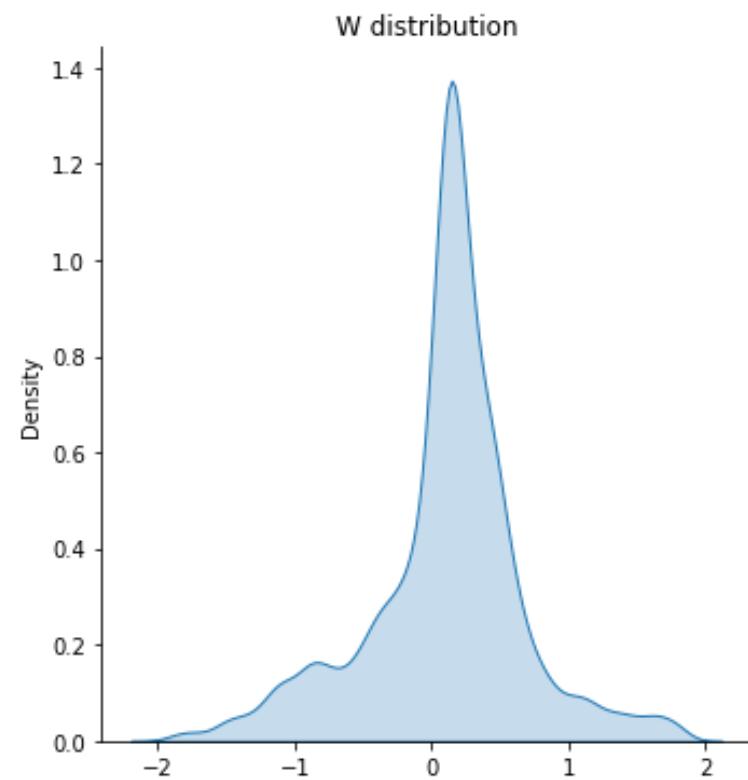
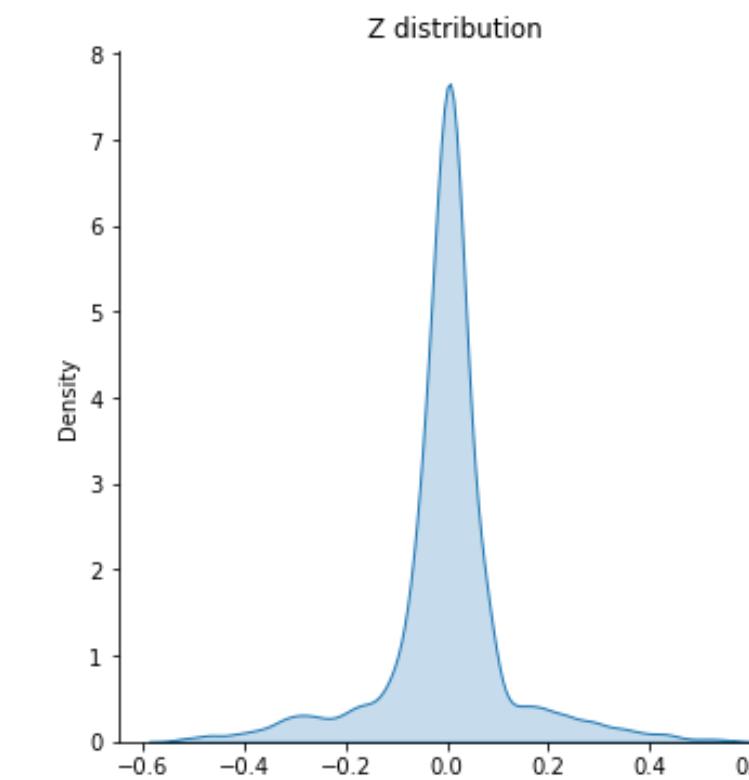
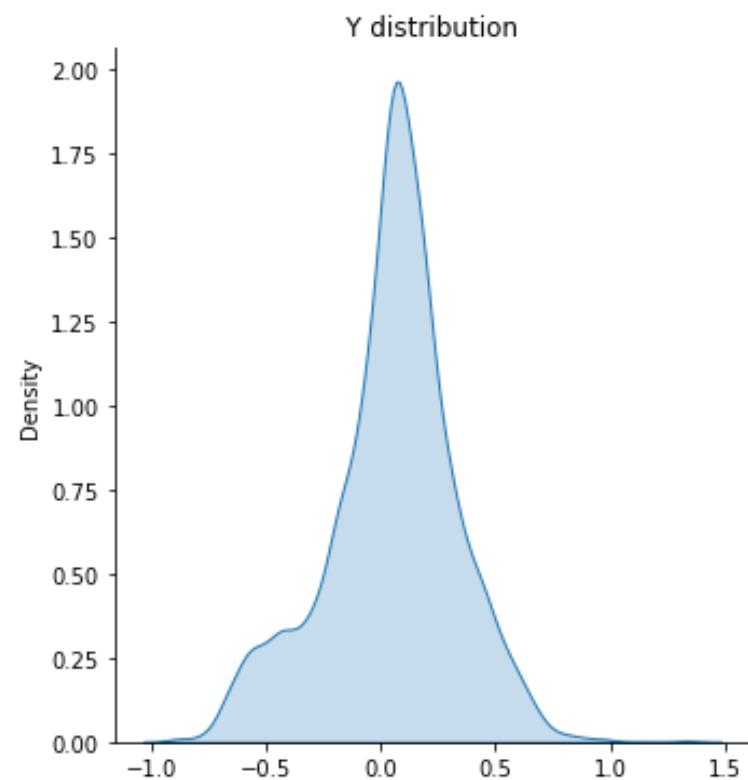
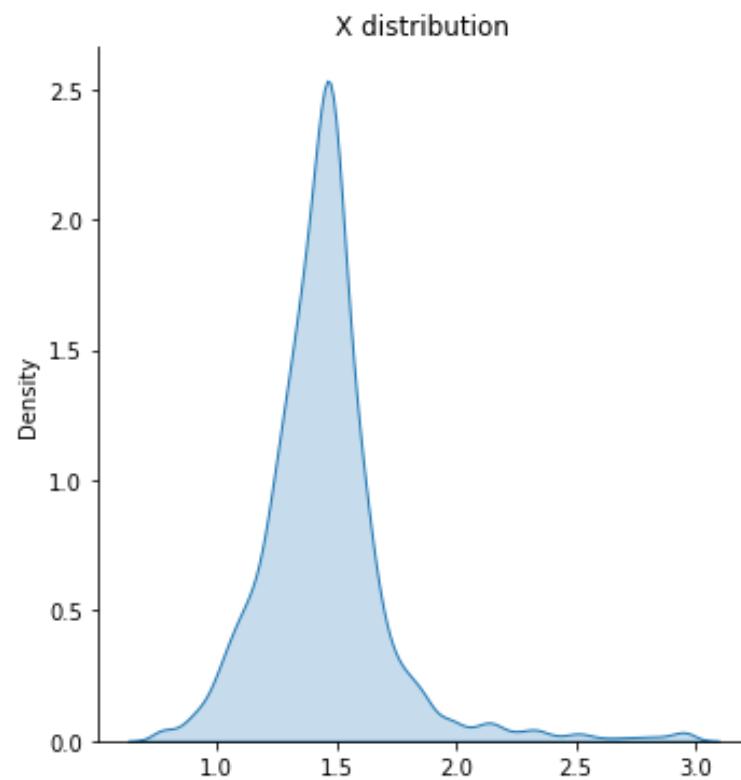


HOW GRAD-CAM WORKS



[Selvaraju et al.
2019]

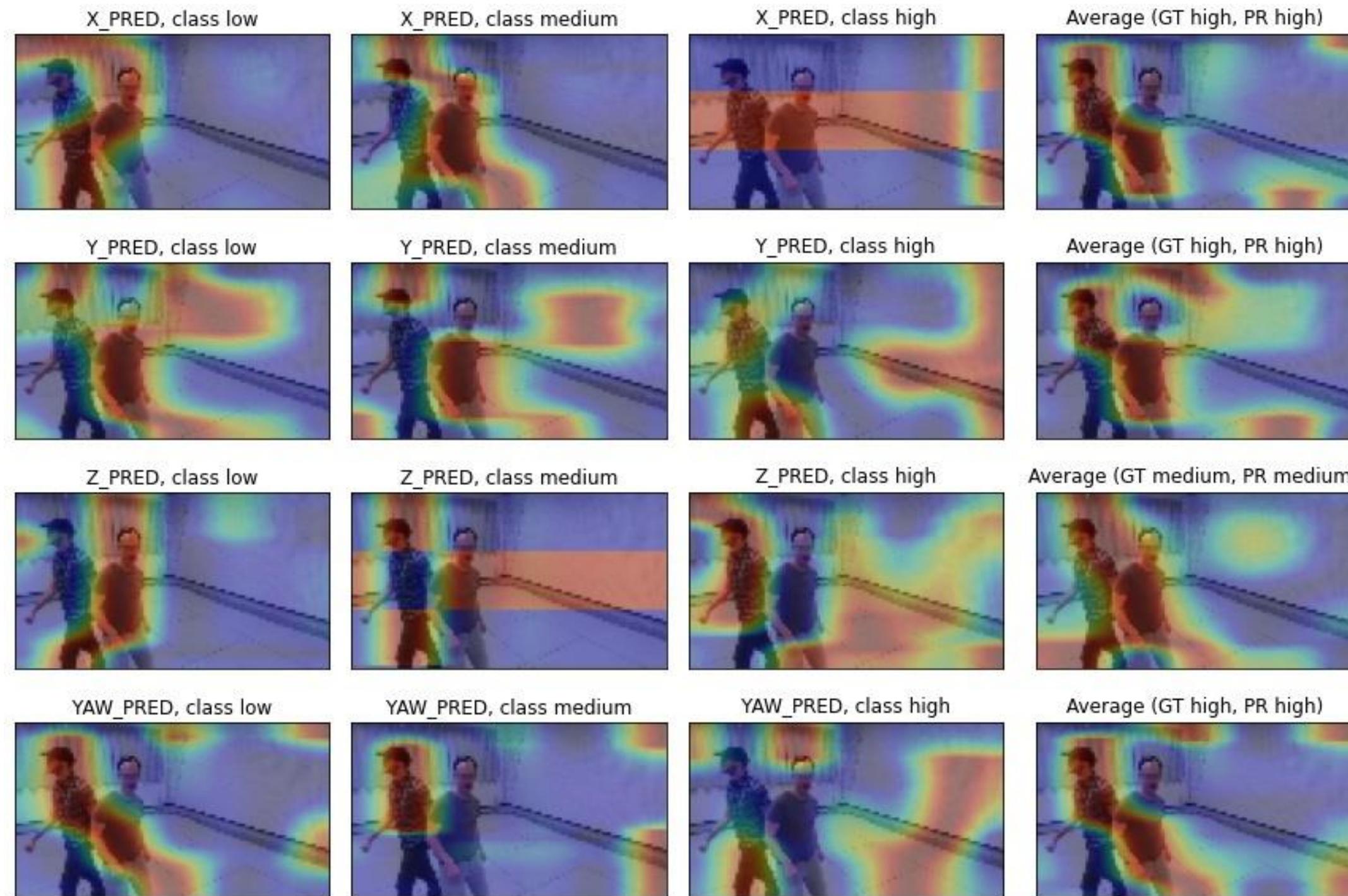
GRAD-CAM: FROM REGRESSION TO CLASSIFICATION



GRAD-CAM PER CLASS



GRAD-CAM PER CLASS



SMARTPHONE EVALUATION: MULTIPLE PEOPLE



SMARTPHONE EVALUATION: FAST MOVEMENTS



SMARTPHONE EVALUATION: FOREGROUND PERSON

