

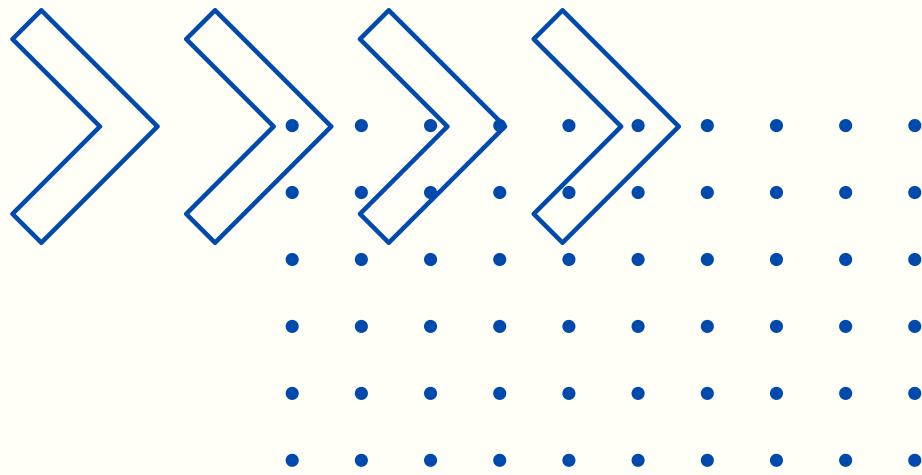


# MILITARY AIRCRAFT DETECTION

Early detection of military aircraft can mean the difference between winning or losing a war.



# TEAM



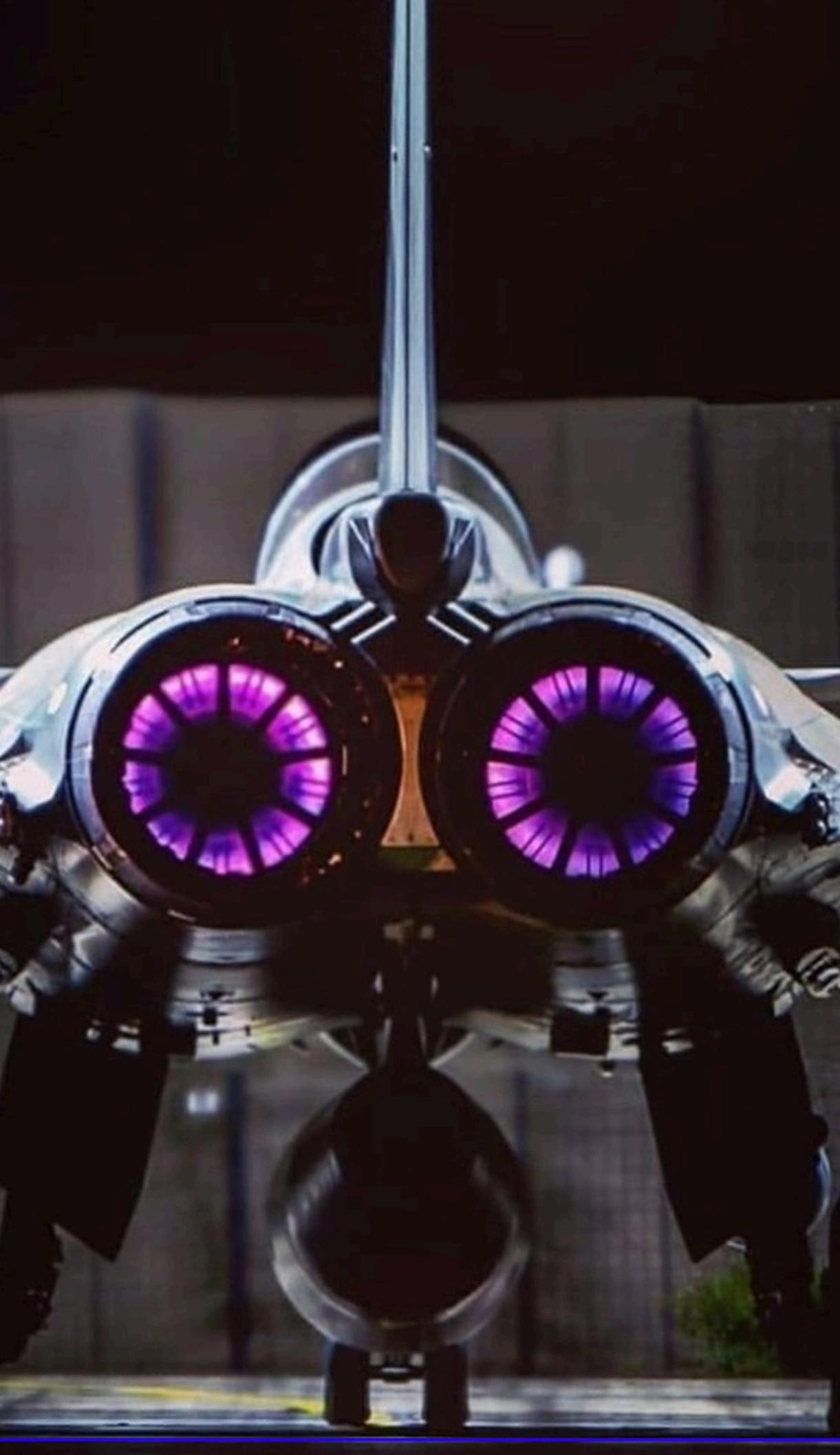
Daniel La Rotta



**KNOW THE SKY AND  
KNOW THE EARTH, AND  
YOUR VICTORY WILL BE  
COMPLETE**

---

*El arte de la guerra*



# JUSTIFICATION

Detecting military aircraft like the F-16, F-18, F-15, F-35, and C-130 is crucial for national security, intelligence, and readiness. It helps identify threats, improves surveillance, and offers insights into foreign military strategies and technologies.





## OBJECTIVES

To detect military aircraft such as the F-16, F-18, F-15, F-35, and C-130 using specialized machine learning models, such as Convolutional Neural Networks (CNNs), to improve the identification and monitoring of these aircraft, thereby strengthening national security and defense capabilities.



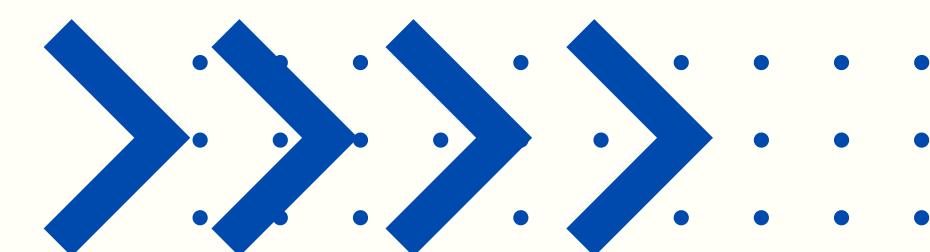
# DATASET

## KAGGEL

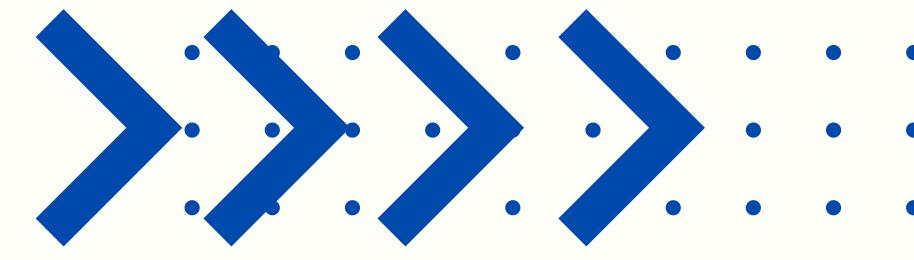
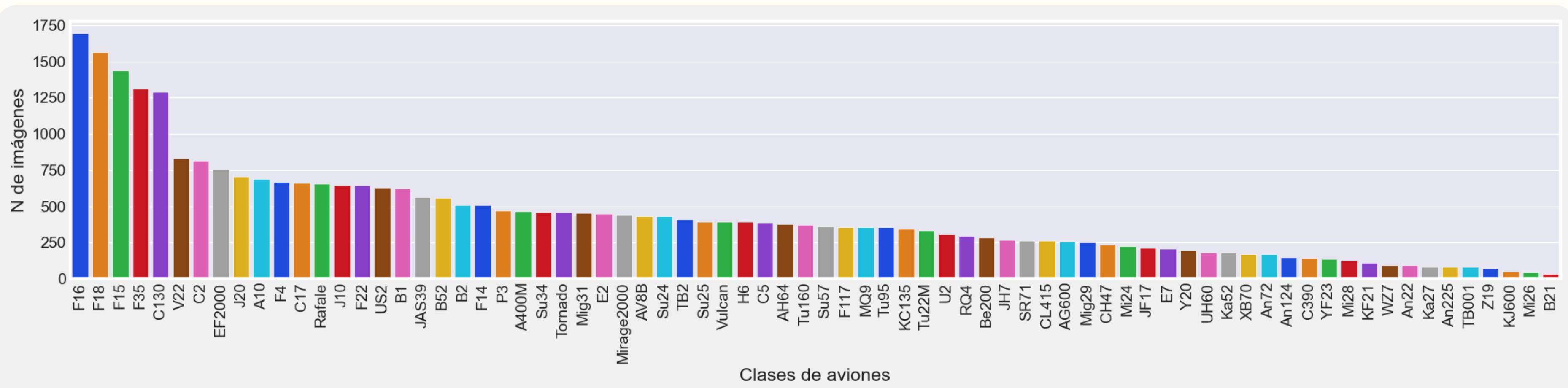
T NAKAMURA - 10GB

### MILITARY AIRCRAFT DETECTION DATASET

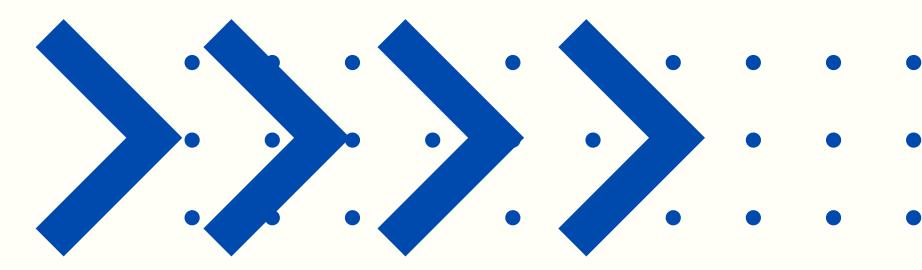
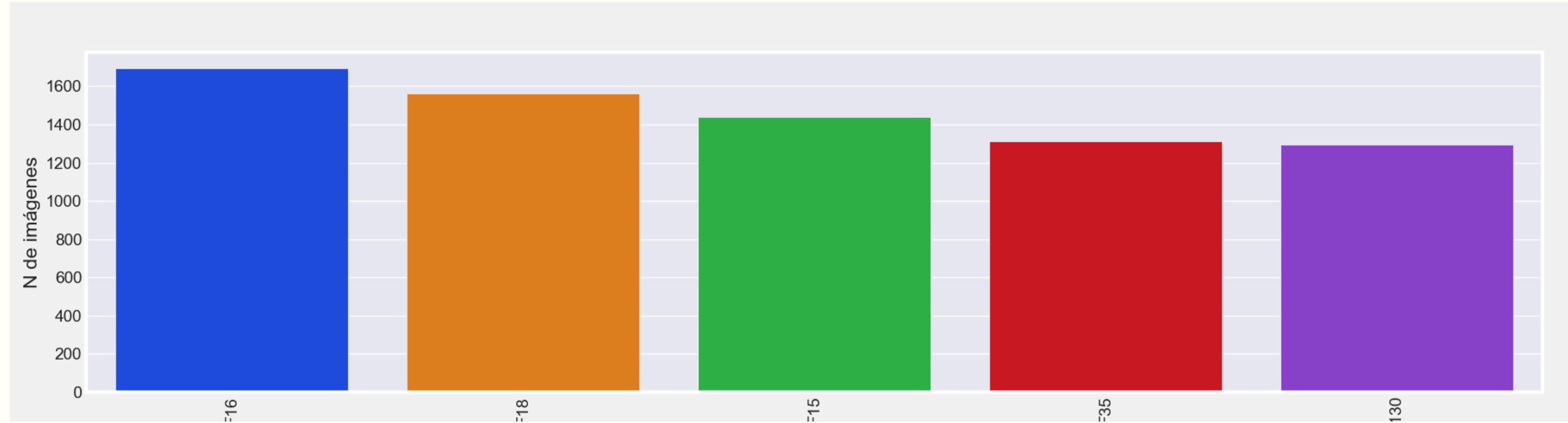
- This dataset is designed for object detection of military aircraft. The dataset encompasses 74 different military aircraft types, with some types merged as one class along with their variants.



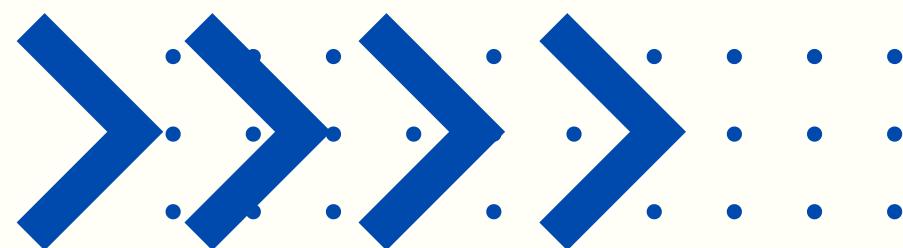
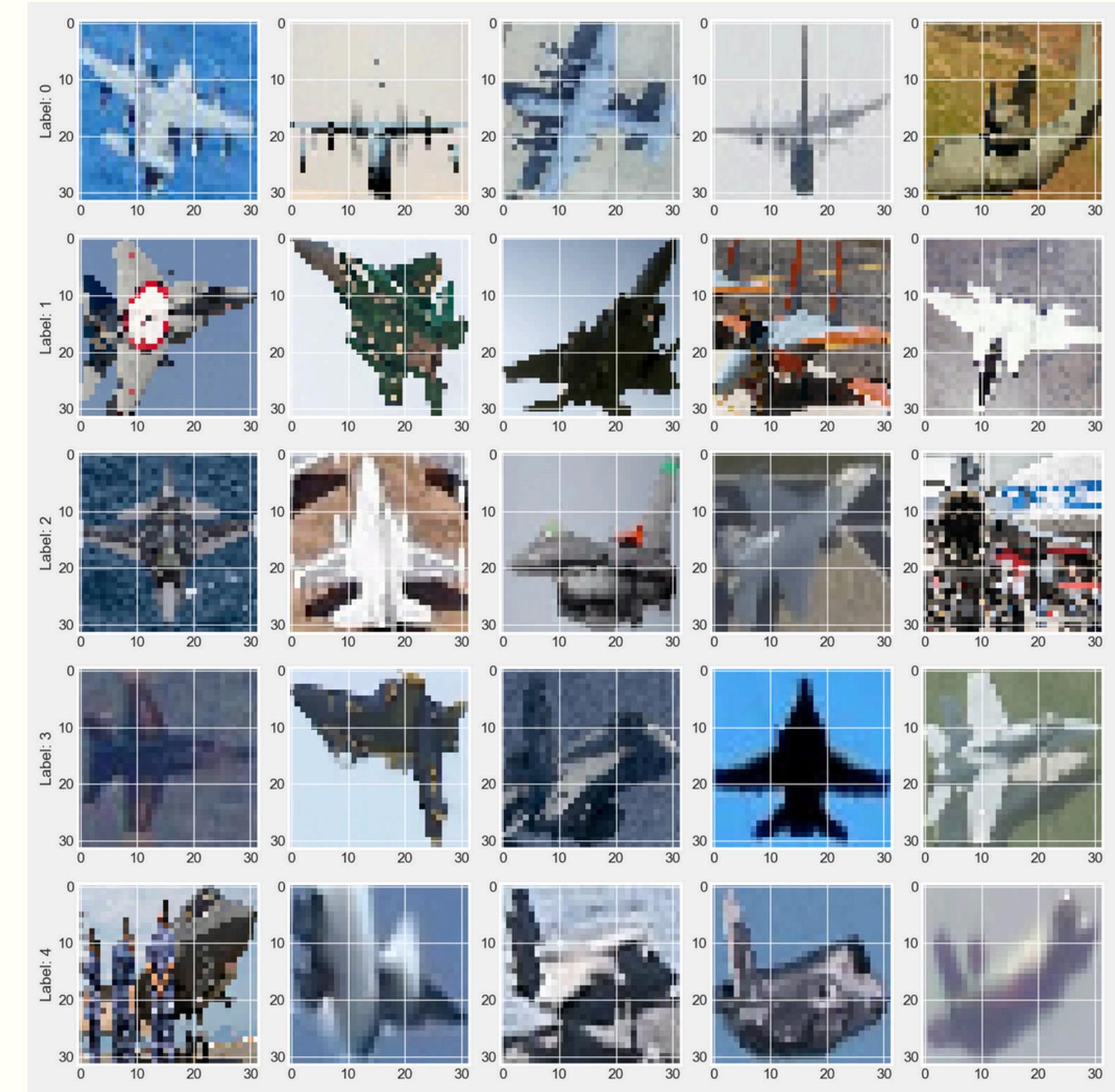
# DATASET



# DATASET FINAL

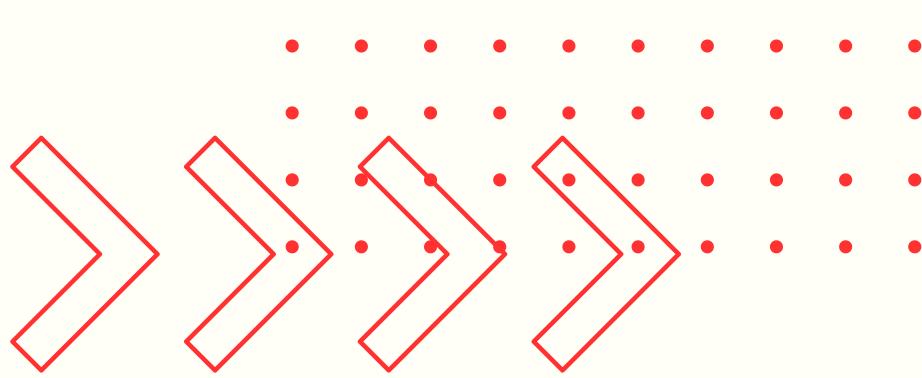


# **DATASET FINAL**



# RESULTS

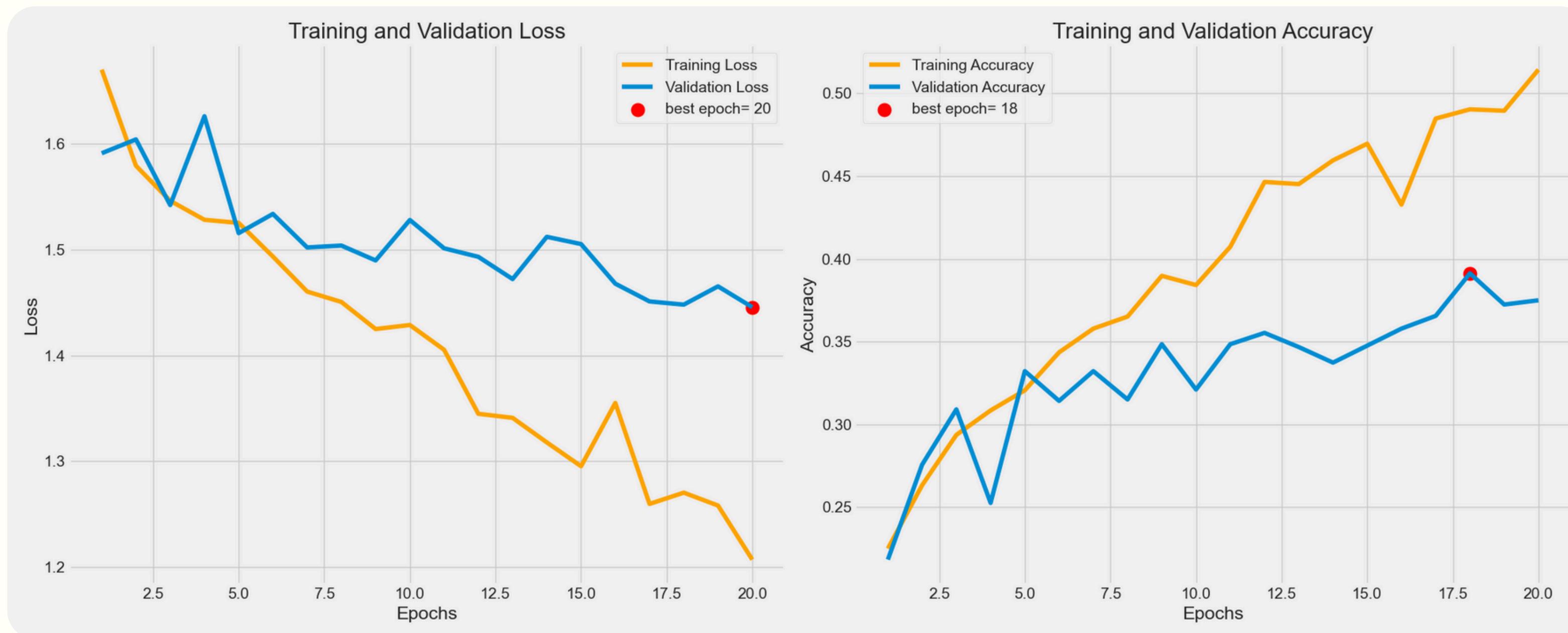
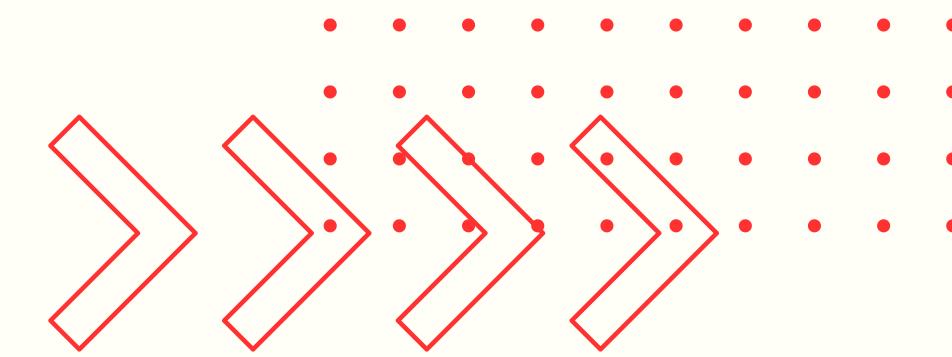




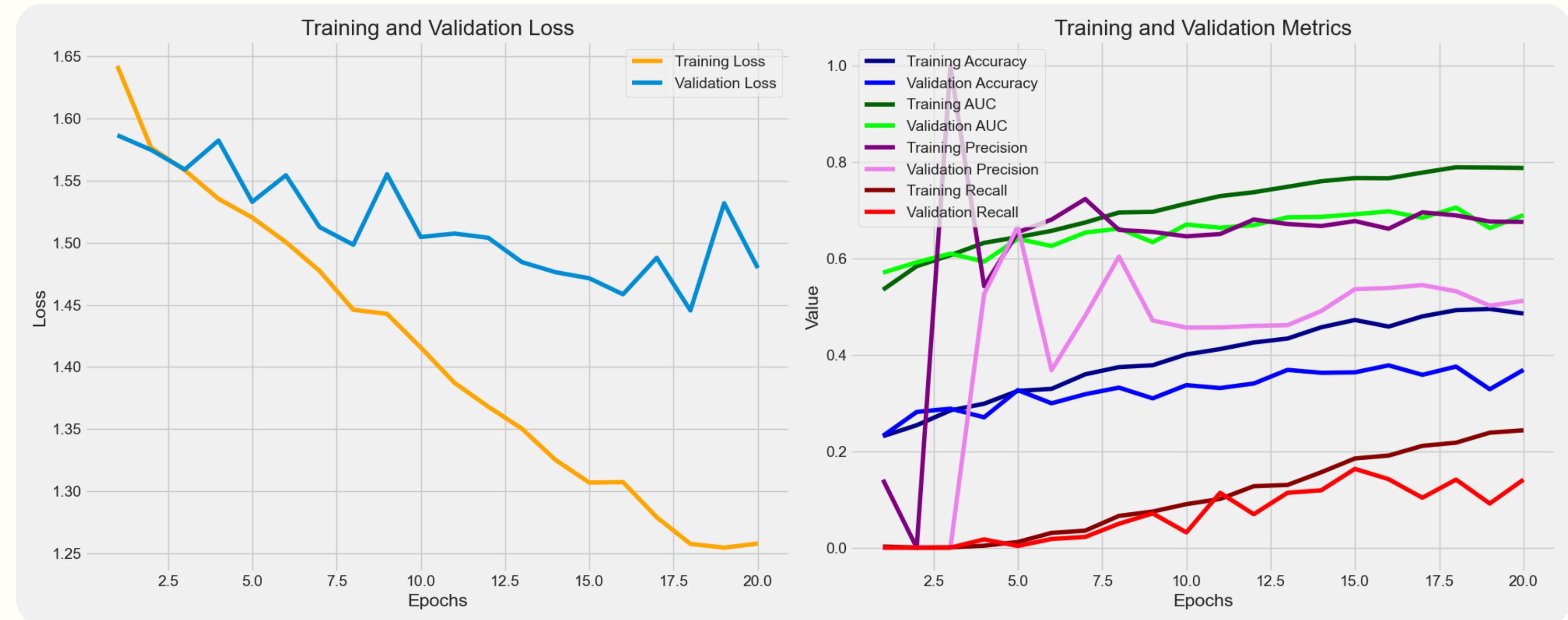
# DNN BASE

Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 3072)	0
dense_3 (Dense)	(None, 128)	393,344
dense_4 (Dense)	(None, 256)	33,024
dense_5 (Dense)	(None, 5)	1,285

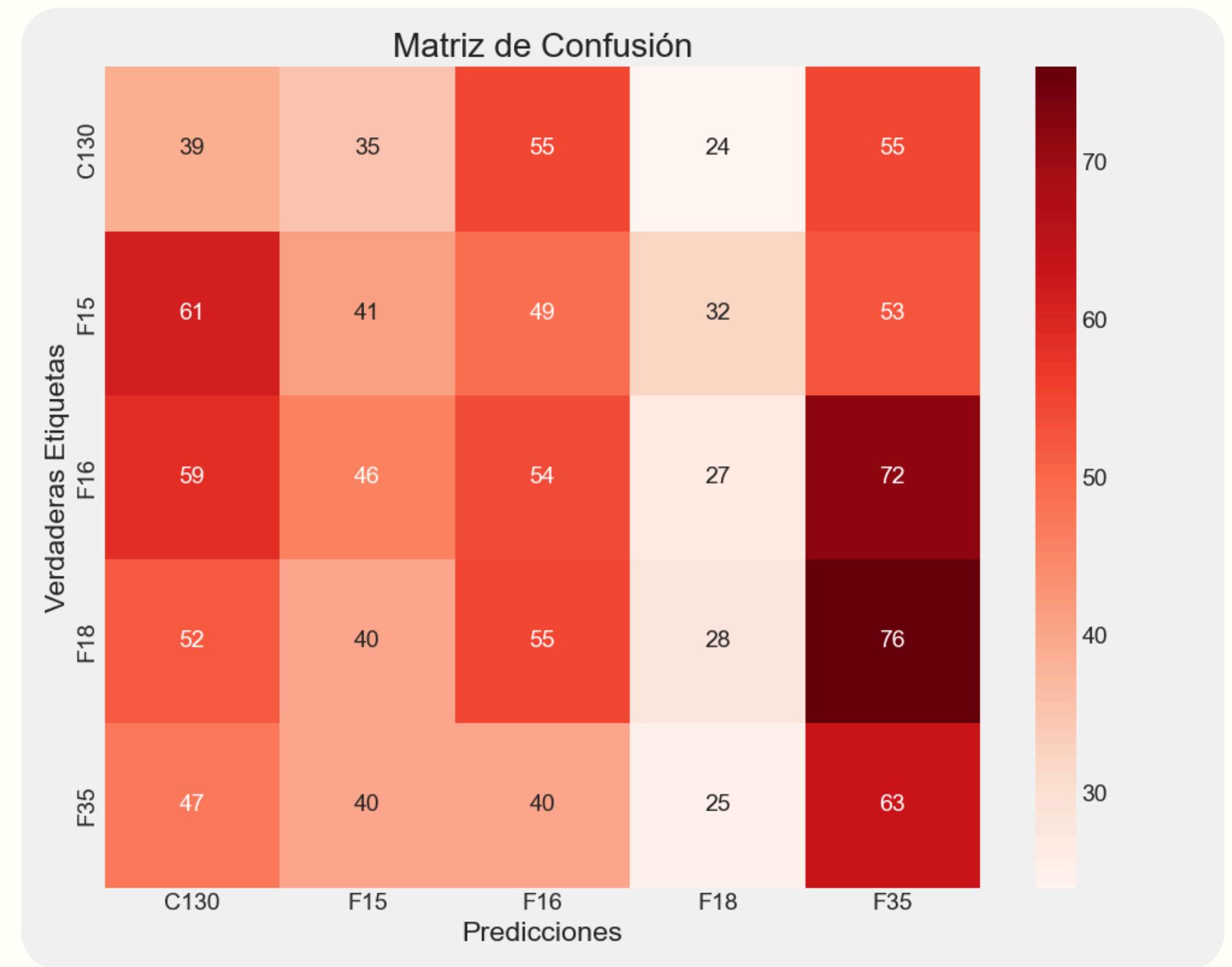
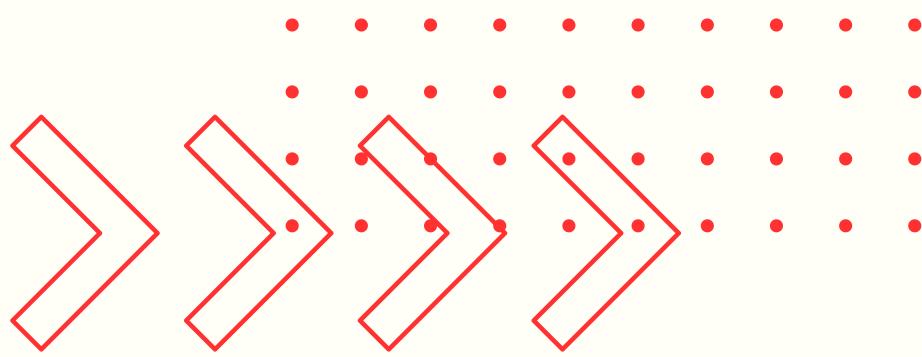
# DNN BASE

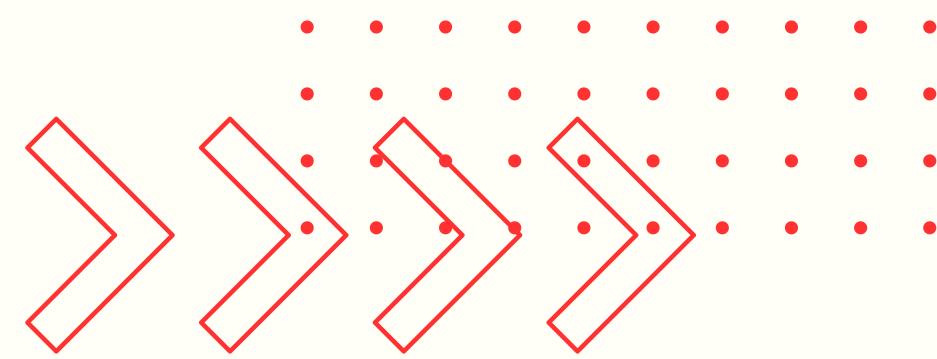


# DNN BASE



# DNN BASE





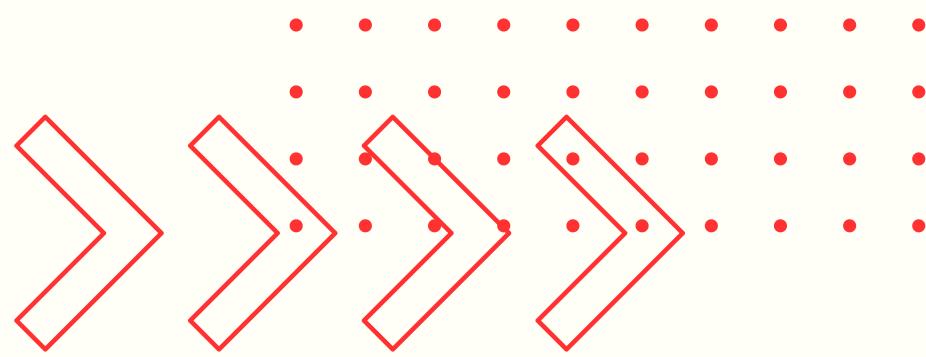
# DNN BASE

## Reporte de Clasificación:

	precision	recall	f1-score	support
C130	0.15	0.19	0.17	208
F15	0.20	0.17	0.19	236
F16	0.21	0.21	0.21	258
F18	0.21	0.11	0.14	251
F35	0.20	0.29	0.24	215
accuracy			0.19	1168
macro avg	0.19	0.20	0.19	1168
weighted avg	0.20	0.19	0.19	1168

Reporte de clasificación guardado en: model1\_DNN\_classification\_report.txt

# DNN VARIANTES



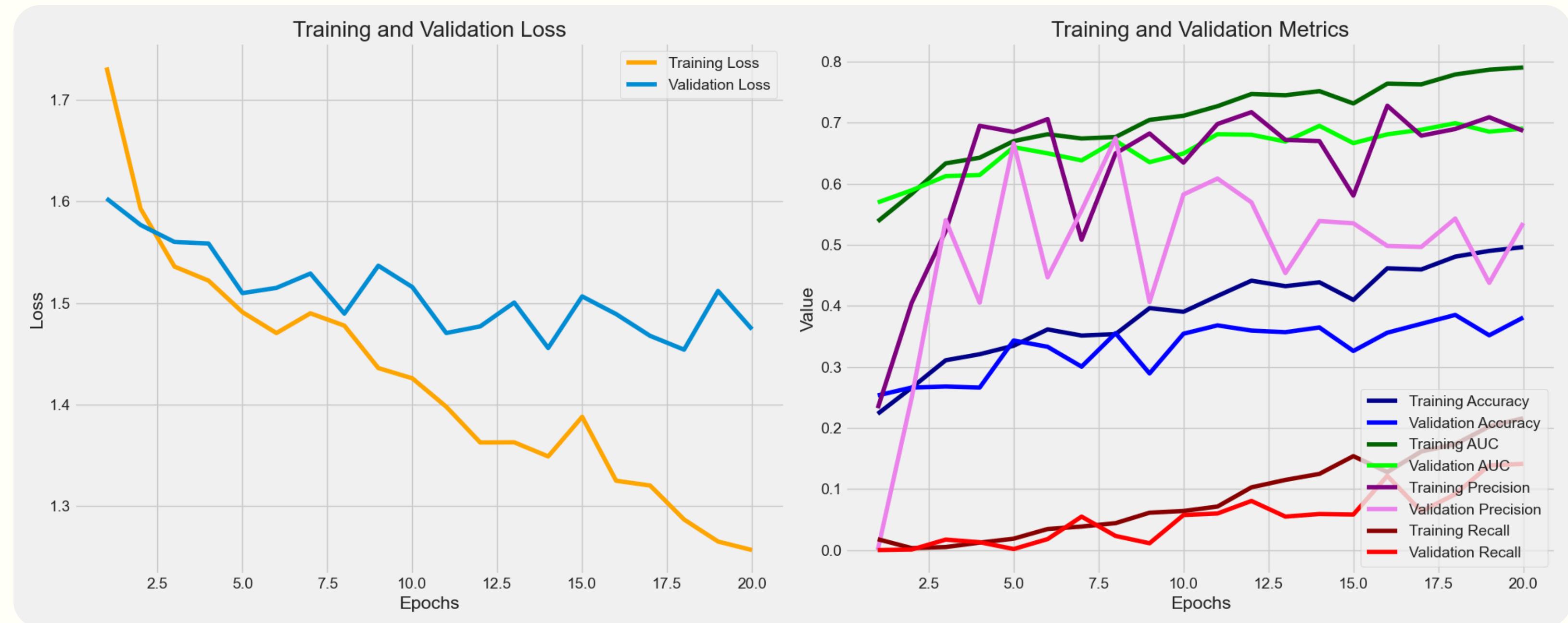
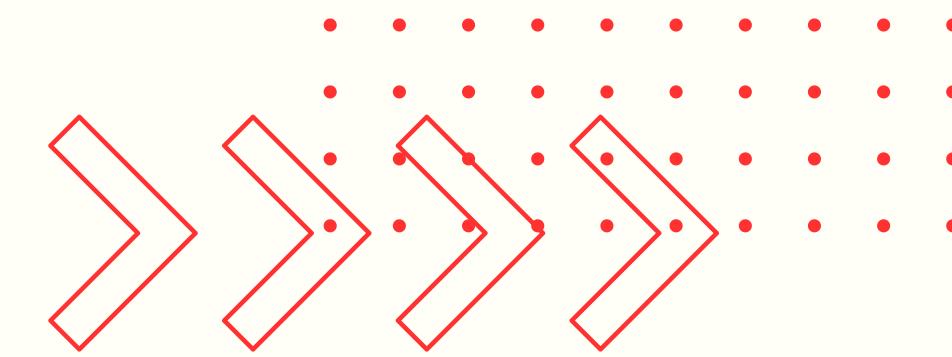
DNN BASE

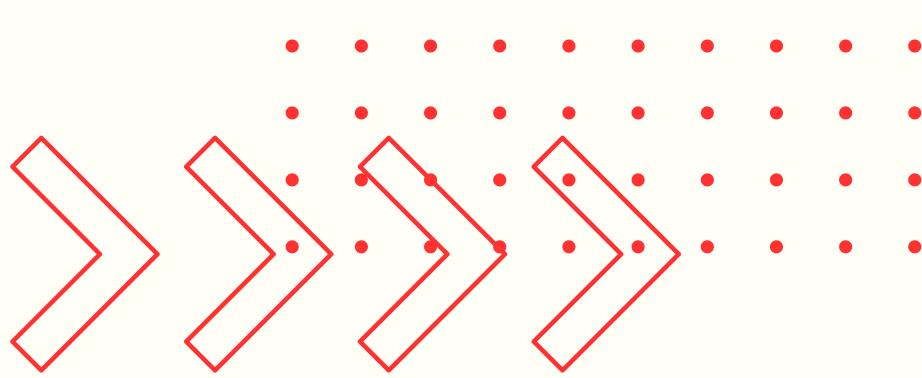
Layer (type)	Output Shape	Param #
flatten_1 (Flatten)	(None, 3072)	0
dense_3 (Dense)	(None, 128)	393,344
dense_4 (Dense)	(None, 256)	33,024
dense_5 (Dense)	(None, 5)	1,285

DNN INVERSA

Layer (type)	Output Shape	Param #
flatten_4 (Flatten)	(None, 3072)	0
dense_12 (Dense)	(None, 256)	786,688
dense_13 (Dense)	(None, 128)	32,896
dense_14 (Dense)	(None, 5)	645

# DNN INVERSA





# DNN VARIANTES

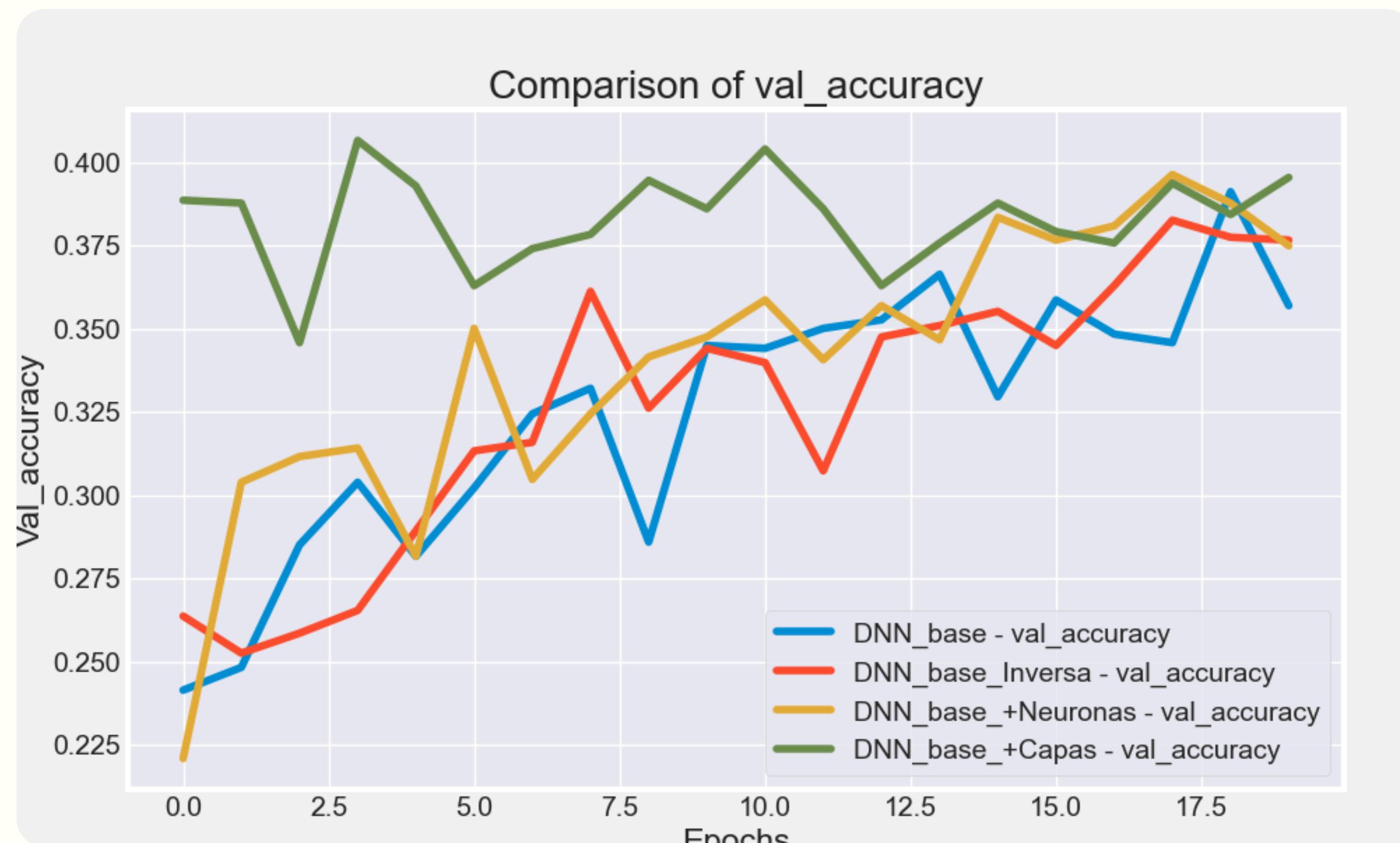
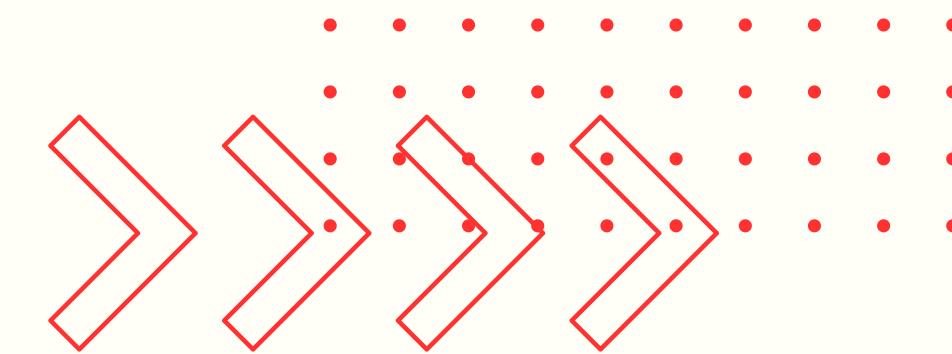
DNN + NEURONAS

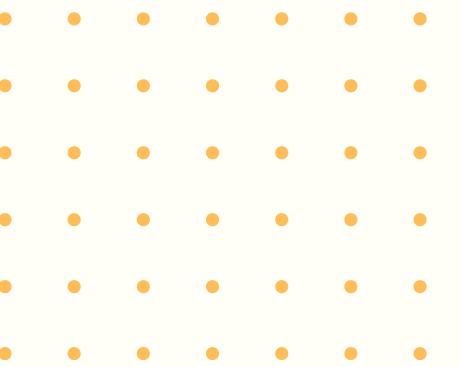
Layer (type)	Output Shape	Param #
flatten_11 (Flatten)	(None, 3072)	0
dense_33 (Dense)	(None, 128)	393,344
dense_34 (Dense)	(None, 256)	33,024
dense_35 (Dense)	(None, 512)	131,584
dense_36 (Dense)	(None, 5)	2,565

DNN + CAPAS

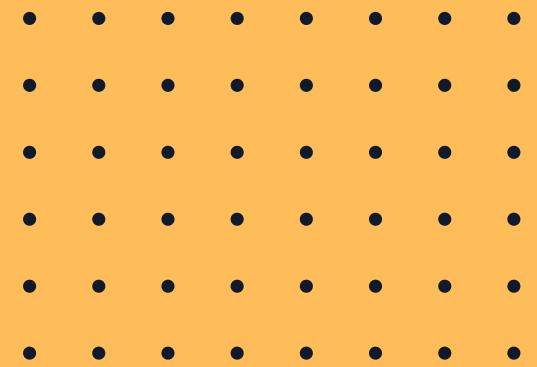
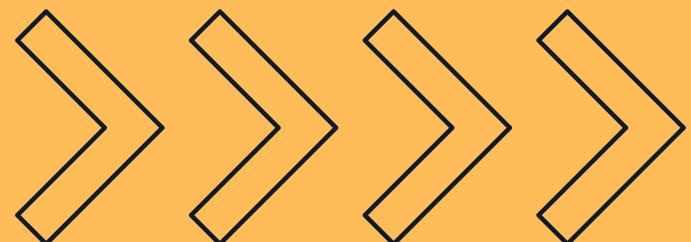
Layer (type)	Output Shape	Param #
flatten_12 (Flatten)	(None, 3072)	0
dense_37 (Dense)	(None, 32)	98,336
dense_38 (Dense)	(None, 72)	2,376
dense_39 (Dense)	(None, 128)	9,344
dense_40 (Dense)	(None, 512)	66,048
dense_41 (Dense)	(None, 5)	2,565

# DNN VARIANTES





# CNN CUSTOM

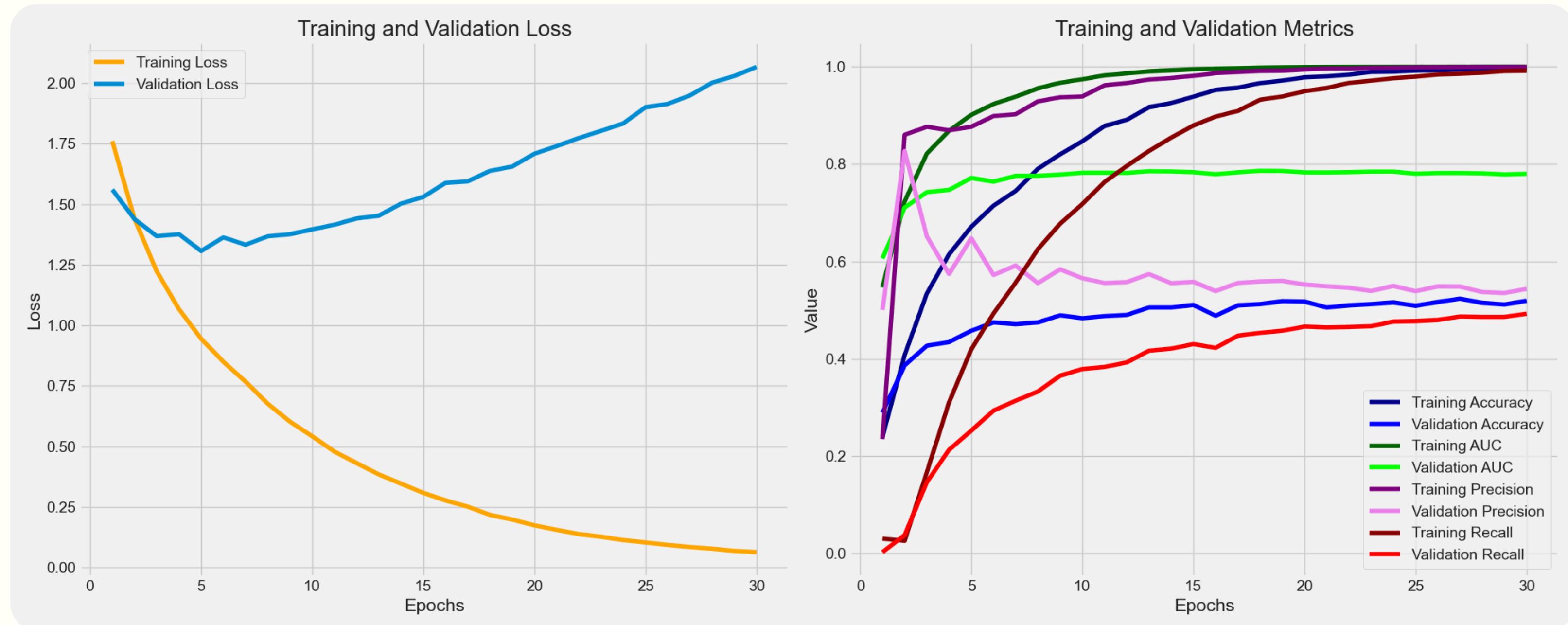




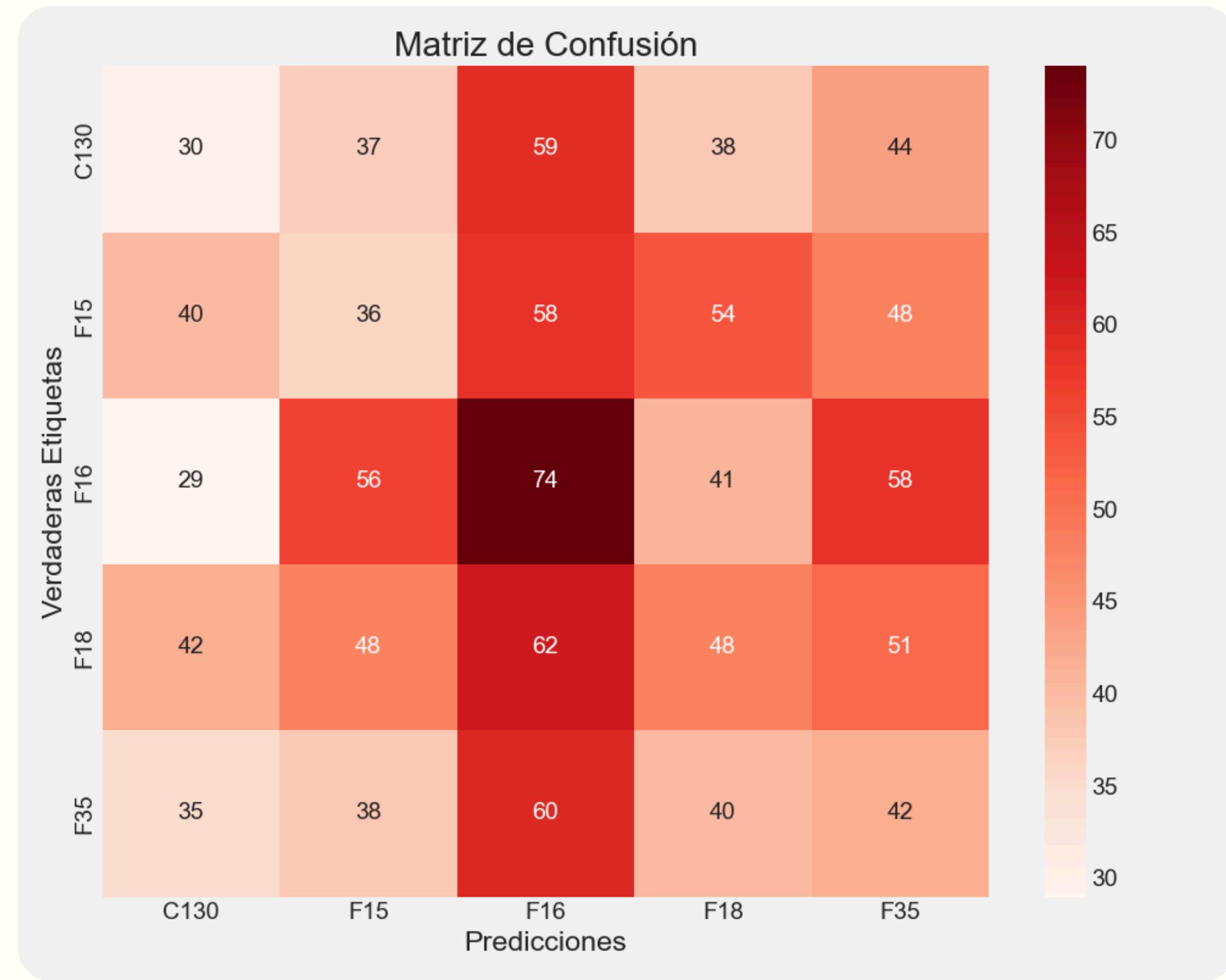
# CNN BASE

Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 30, 30, 75)	2,100
flatten_19 (Flatten)	(None, 67500)	0
dense_59 (Dense)	(None, 5)	337,505

# CNN BASE



# CNN BASE



# CNN VARIANTES



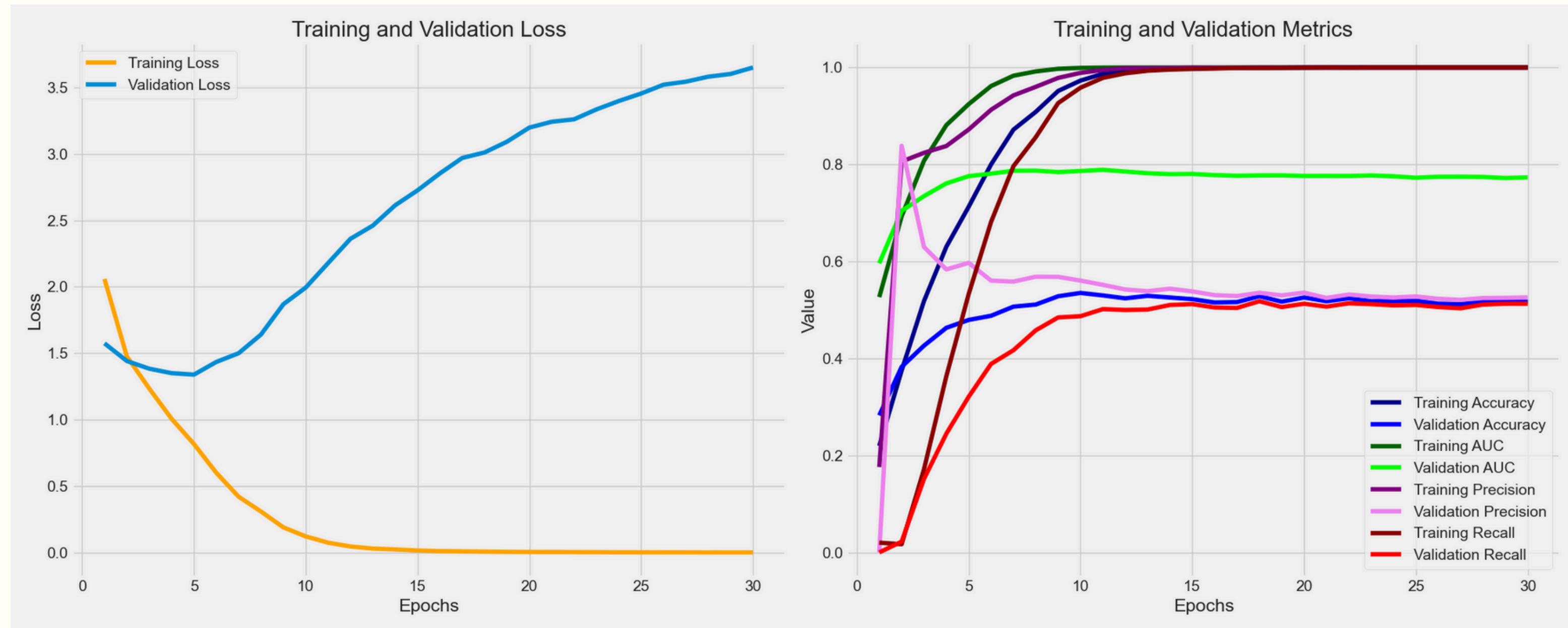
CNN + CAPAS

CNN + CAPAS +  
DROPOUT

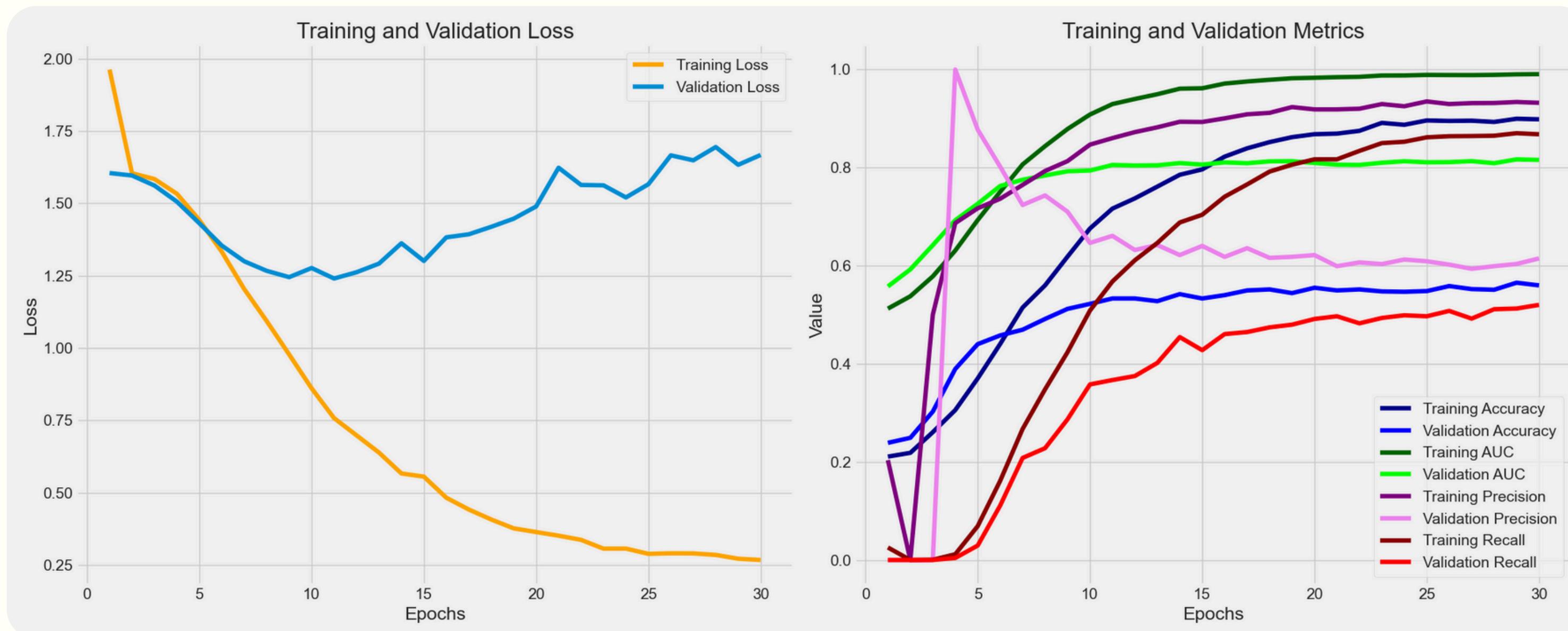
Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 30, 30, 75)	2,100
flatten_5 (Flatten)	(None, 67500)	0
dense_16 (Dense)	(None, 256)	17,280,256
dense_17 (Dense)	(None, 128)	32,896
dense_18 (Dense)	(None, 5)	645

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 30, 30, 75)	2,100
flatten_7 (Flatten)	(None, 67500)	0
dense_22 (Dense)	(None, 256)	17,280,256
dropout (Dropout)	(None, 256)	0
dense_23 (Dense)	(None, 128)	32,896
dropout_1 (Dropout)	(None, 128)	0
dense_24 (Dense)	(None, 5)	645

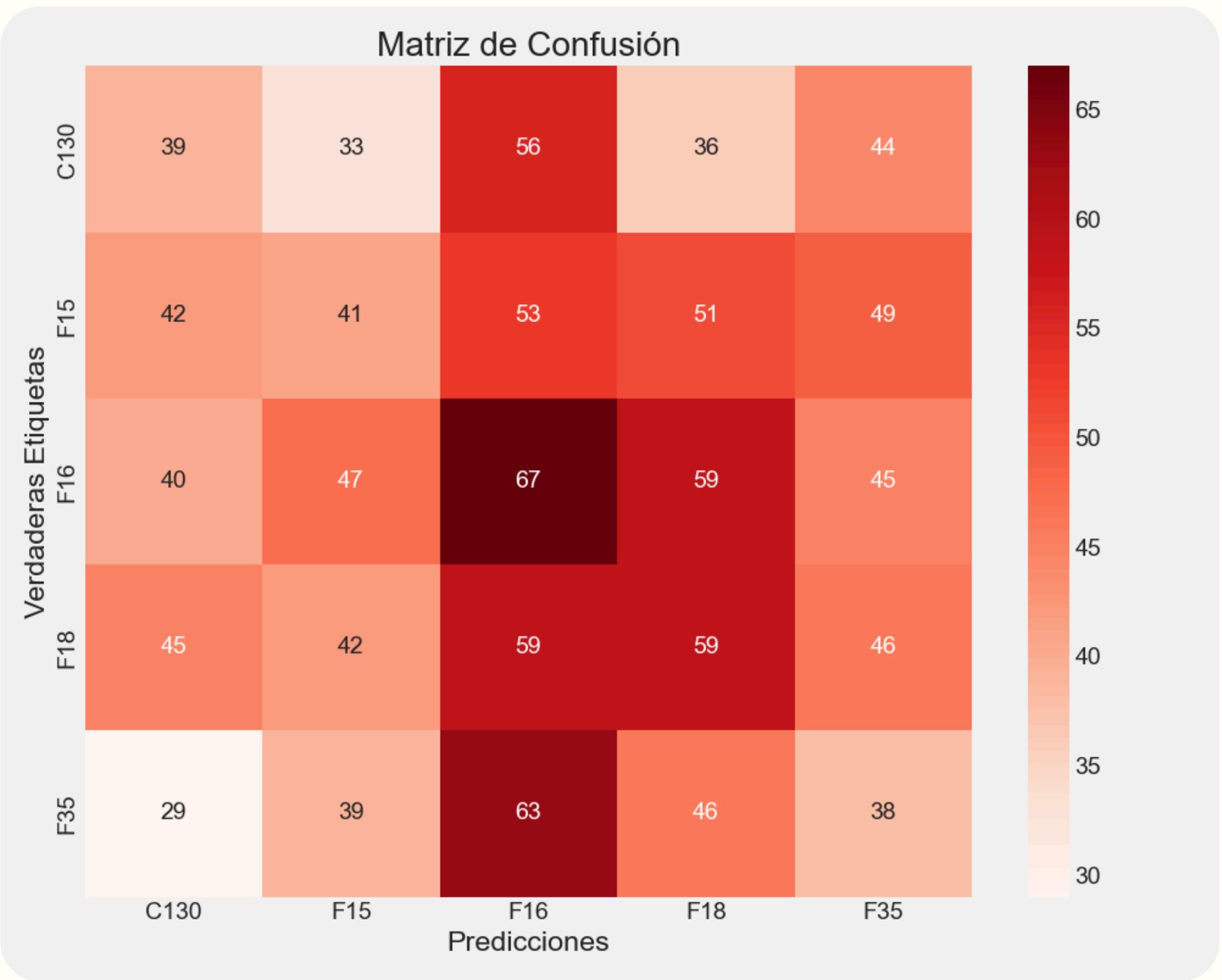
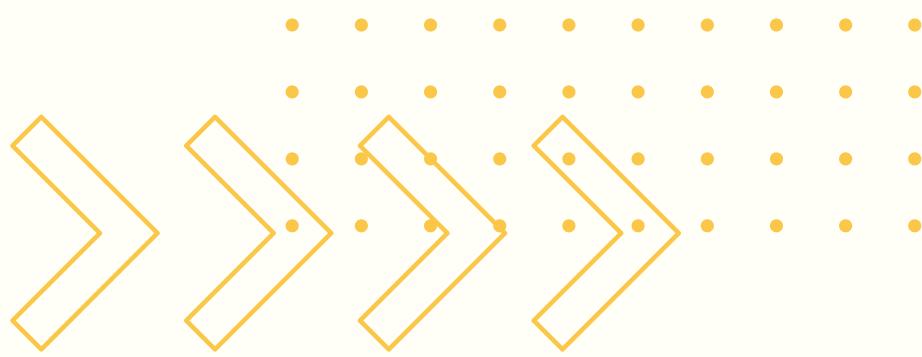
# CNN + CAPAS



# CNN + CAPAS + DROPOUT

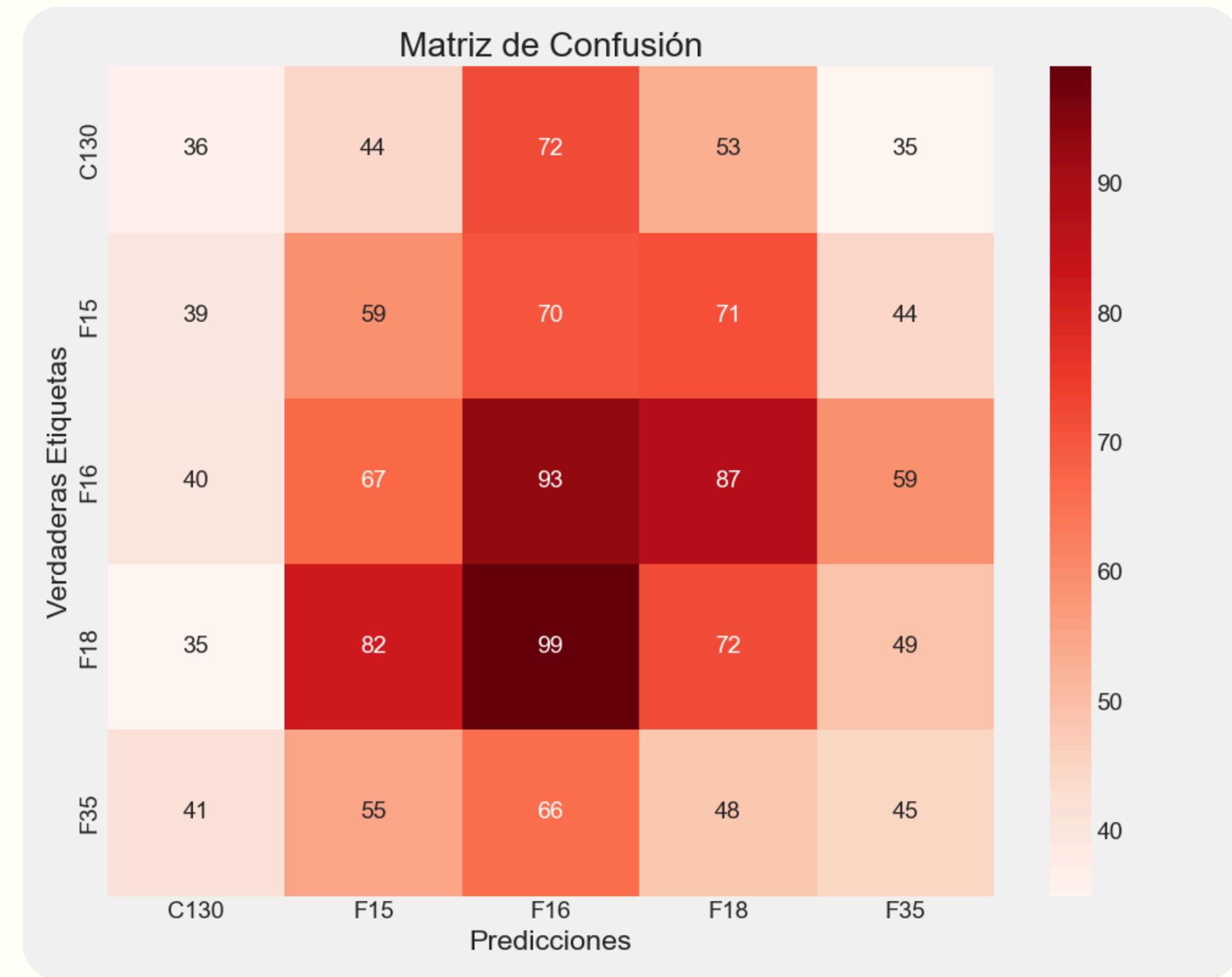


# CNN + CAPAS

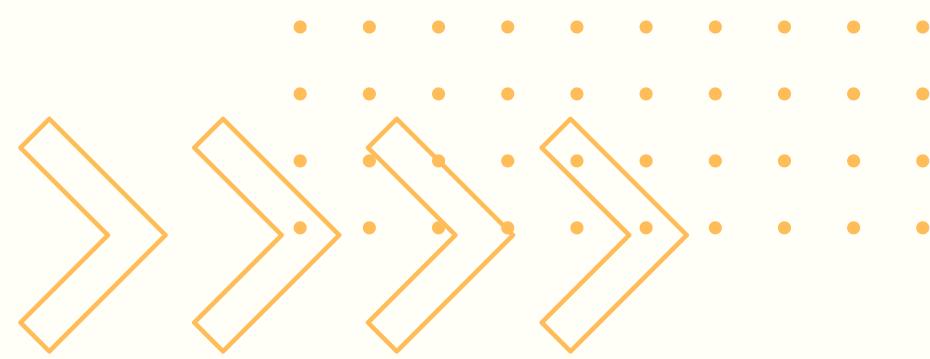




# CNN + CAPAS + DROPOUT



# CNN VARIANTES



CNN + CONV + CAPAS +  
DROPOUT

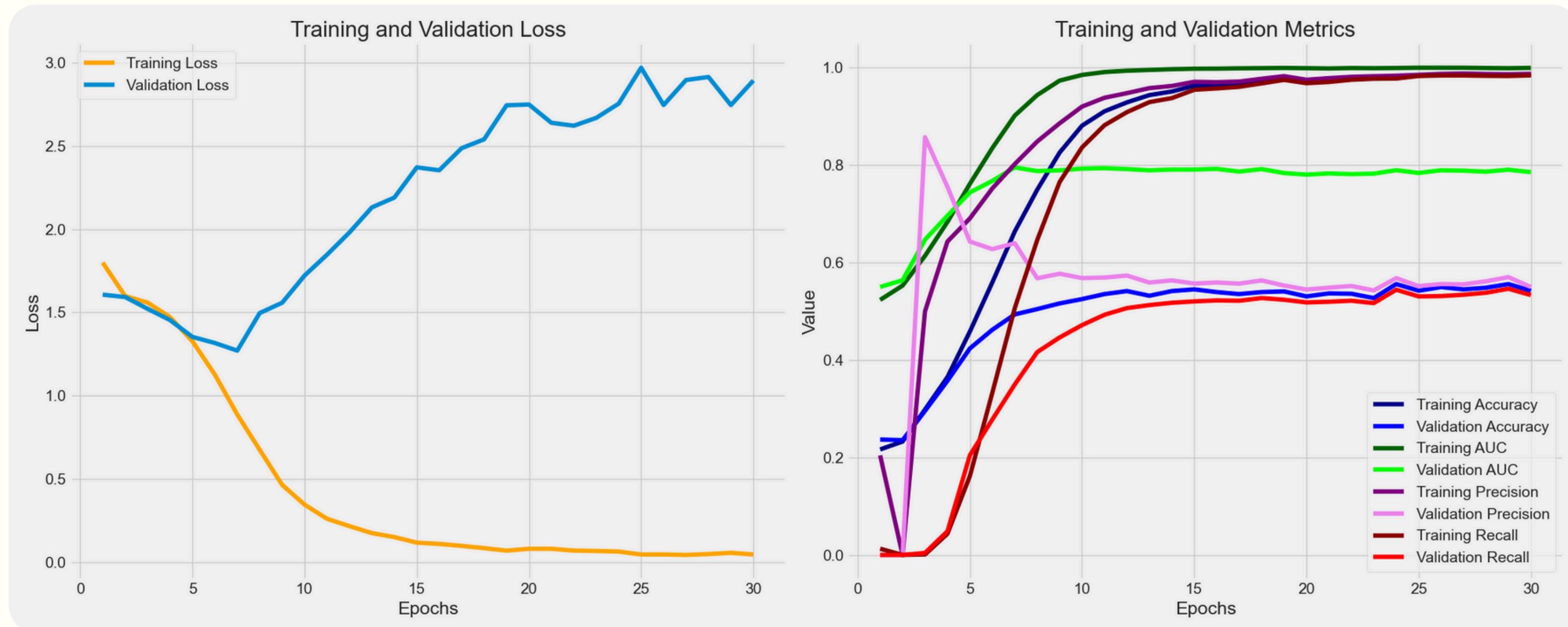
CNN + CONV + CAPAS +  
DROPOUT + MAXPOLING

Layer (type)	Output Shape	Param #
conv2d_12 (Conv2D)	(None, 31, 31, 32)	416
dropout_12 (Dropout)	(None, 31, 31, 32)	0
conv2d_13 (Conv2D)	(None, 30, 30, 64)	8,256
dropout_13 (Dropout)	(None, 30, 30, 64)	0
conv2d_14 (Conv2D)	(None, 29, 29, 128)	32,896
flatten_18 (Flatten)	(None, 107648)	0
dense_56 (Dense)	(None, 256)	27,558,144
dropout_14 (Dropout)	(None, 256)	0
dense_57 (Dense)	(None, 128)	32,896
dropout_15 (Dropout)	(None, 128)	0
dense_58 (Dense)	(None, 5)	645

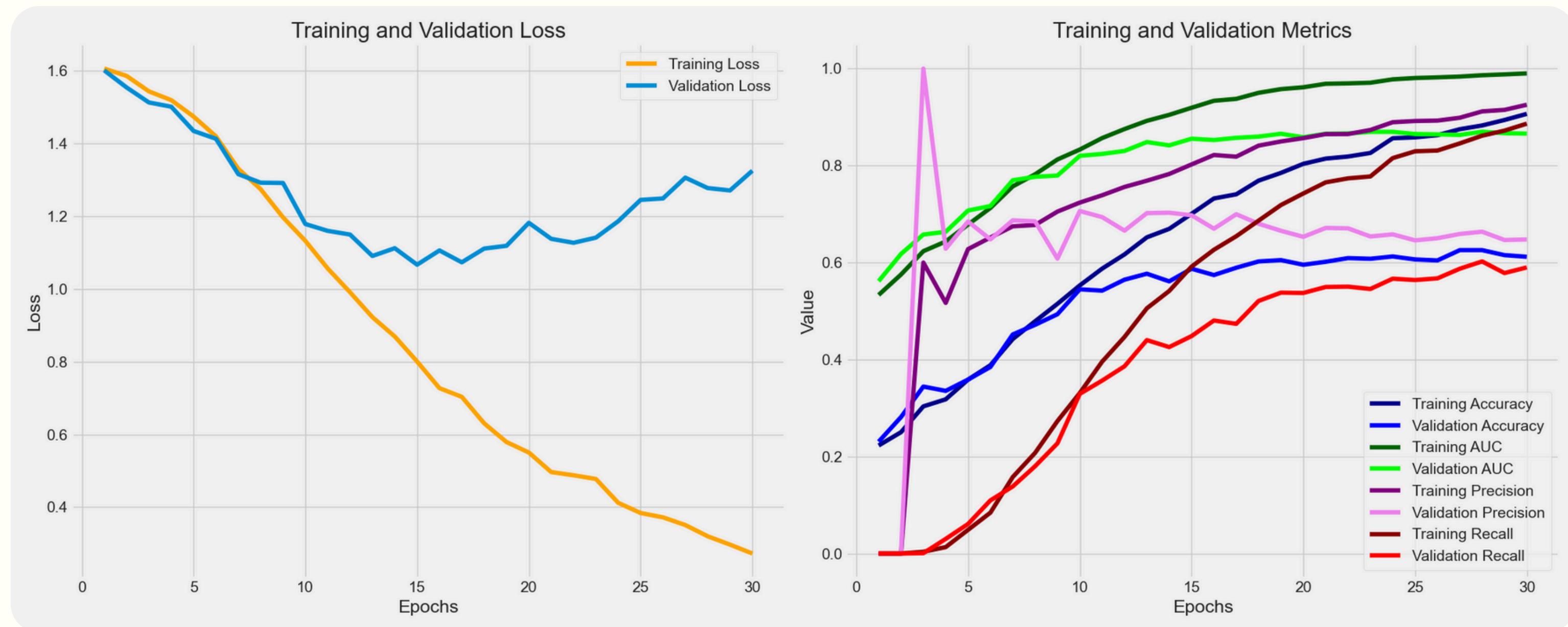
Layer (type)	Output Shape	Param #
conv2d_18 (Conv2D)	(None, 30, 30, 32)	896
max_pooling2d (MaxPooling2D)	(None, 15, 15, 32)	0
dropout_20 (Dropout)	(None, 15, 15, 32)	0
conv2d_19 (Conv2D)	(None, 13, 13, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 6, 6, 64)	0
dropout_21 (Dropout)	(None, 6, 6, 64)	0
conv2d_20 (Conv2D)	(None, 4, 4, 128)	73,856
max_pooling2d_2 (MaxPooling2D)	(None, 2, 2, 128)	0
flatten_20 (Flatten)	(None, 512)	0
dense_62 (Dense)	(None, 256)	131,328
dropout_22 (Dropout)	(None, 256)	0
dense_63 (Dense)	(None, 128)	32,896
dropout_23 (Dropout)	(None, 128)	0
dense_64 (Dense)	(None, 5)	645



# CNN + CONV + CAPAS + DROPOUT

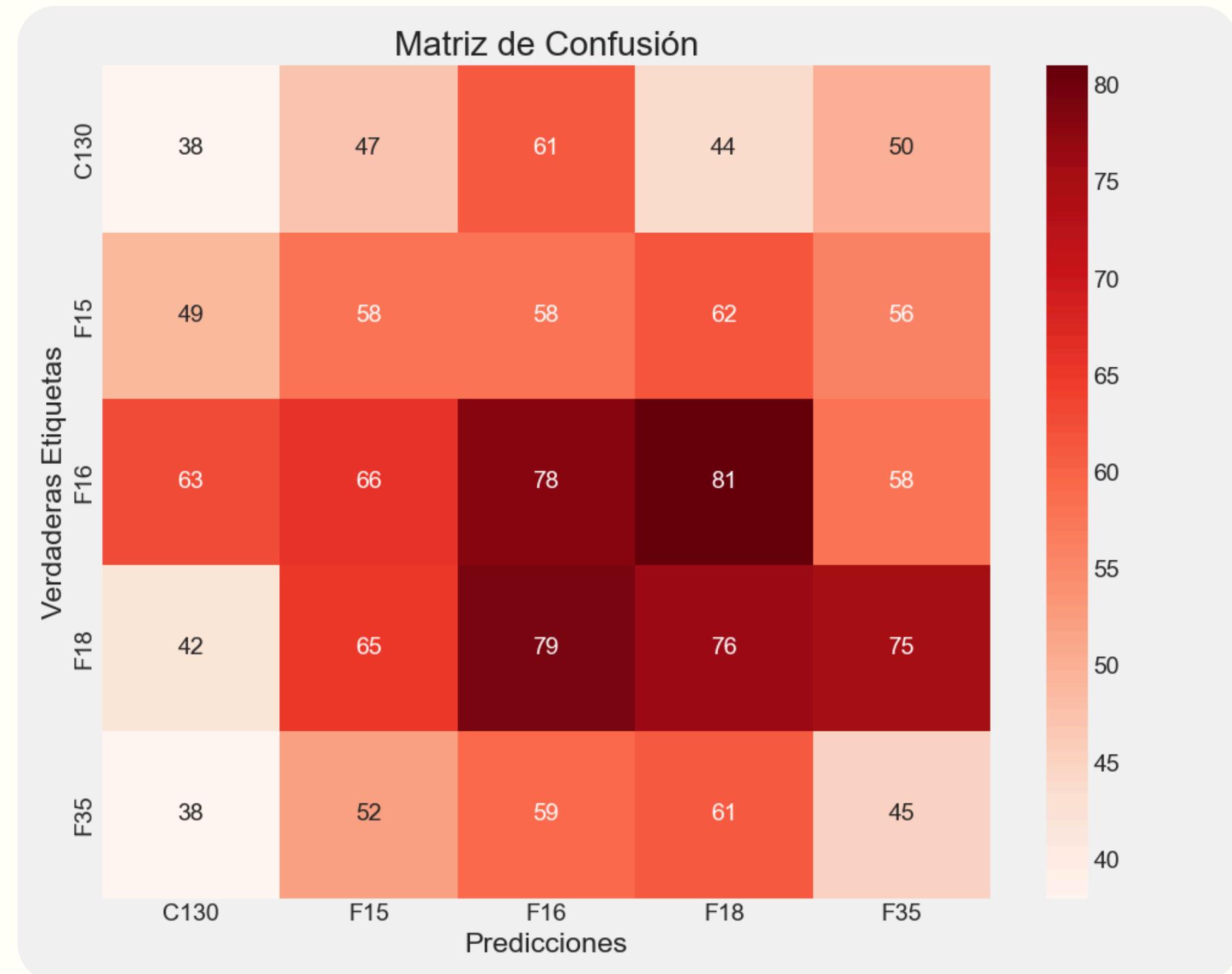


# CNN + CONV + CAPAS + DROPOUT + MAXPOOLING

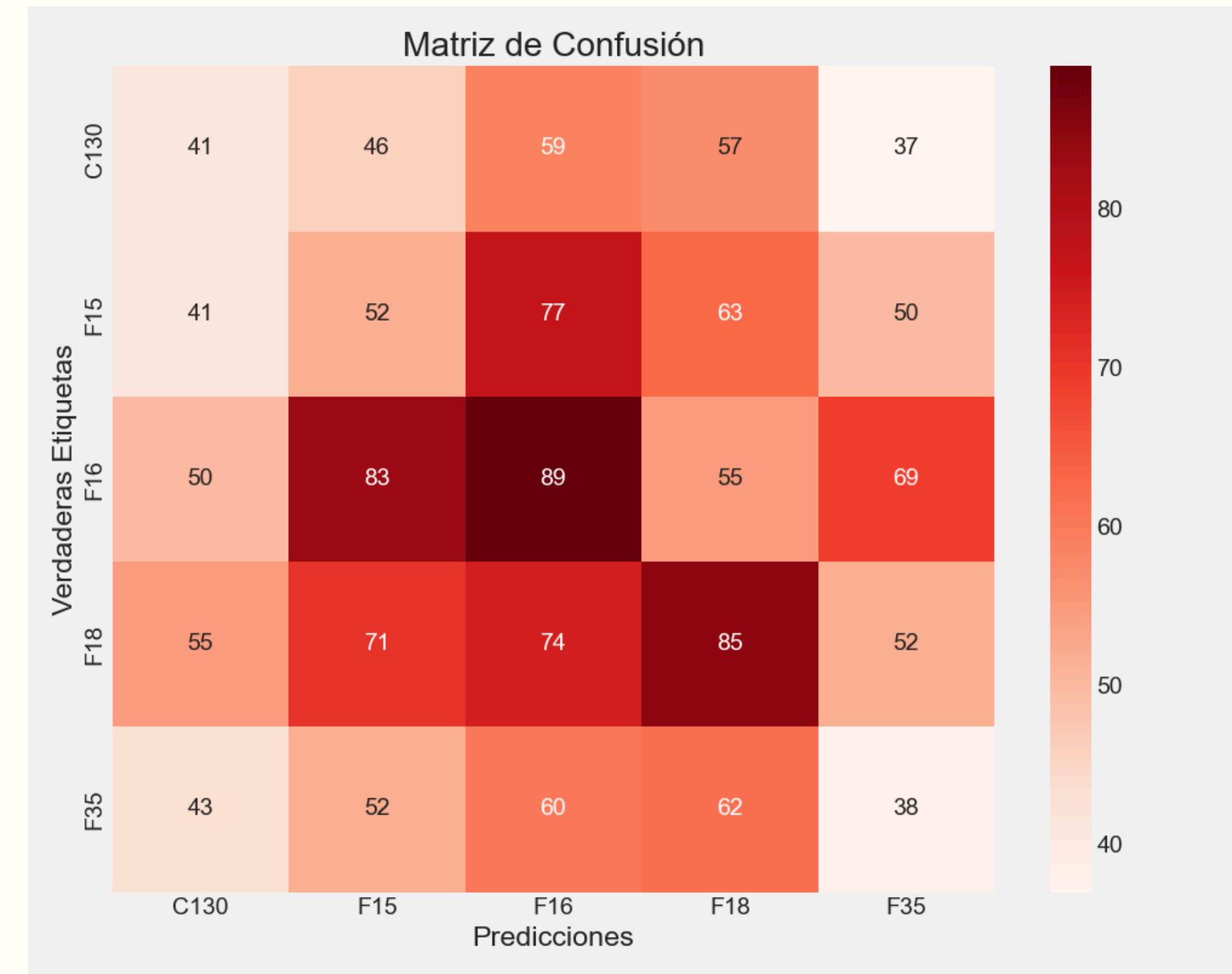


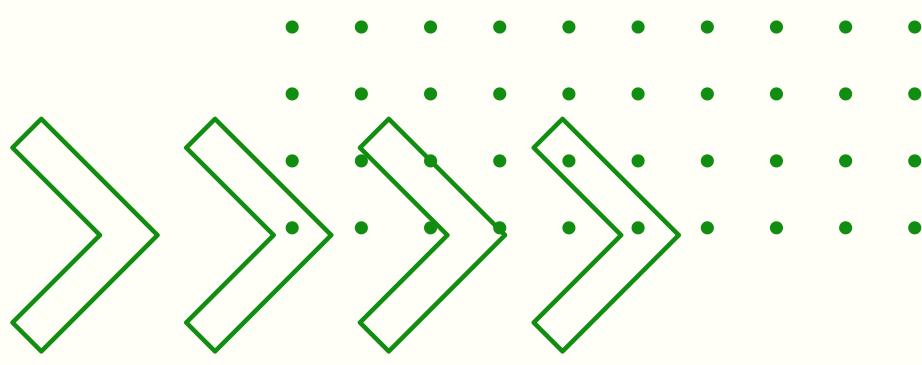


# CNN + CONV + CAPAS + DROPOUT



# CNN + CONV + CAPAS + DROPOUT + MAXPOOLING





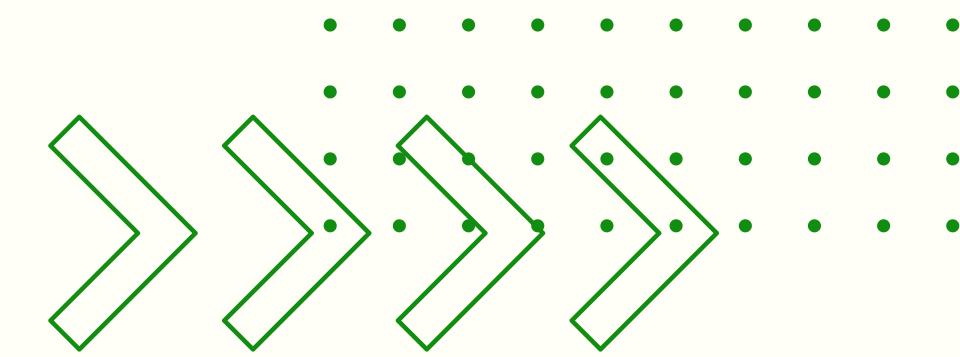
# TRANSFER LEARNING

INCEPTION V3

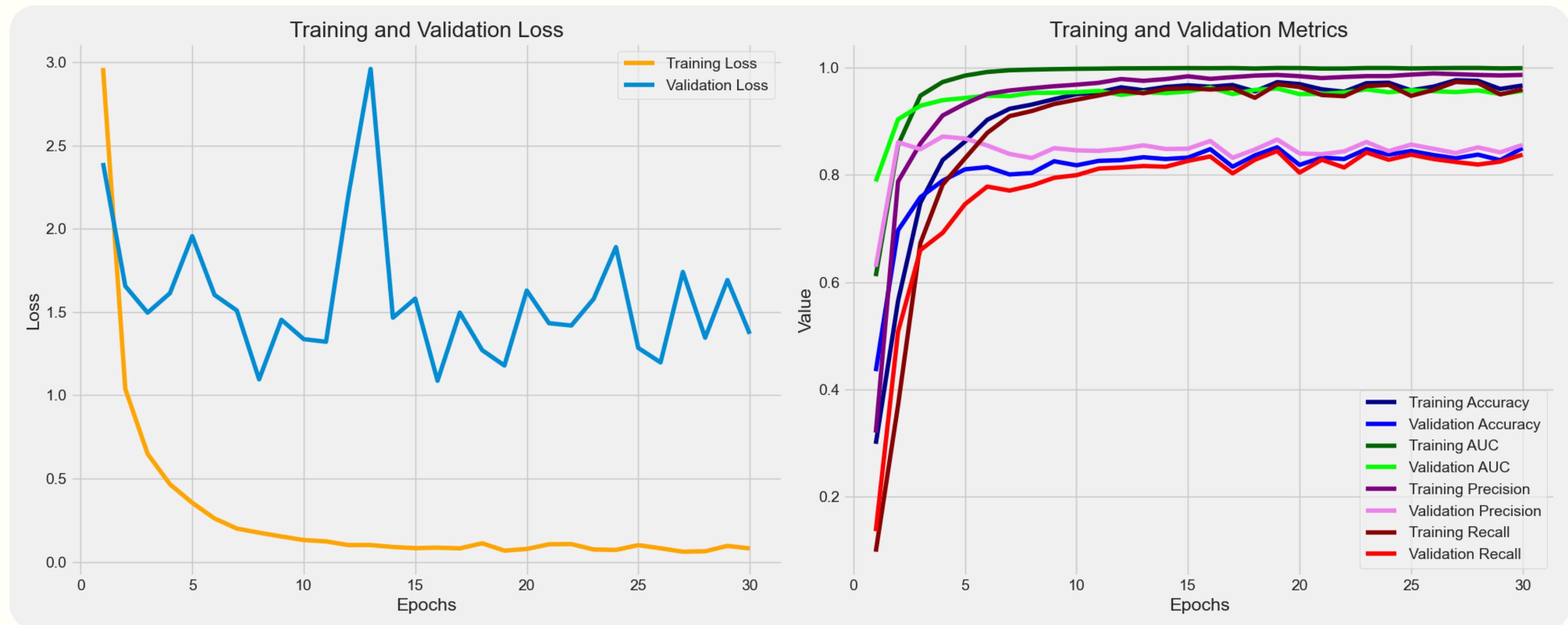
**Total params:** 37,385,893 (142.62 MB)  
**Trainable params:** 28,410,629 (108.38 MB)  
**Non-trainable params:** 8,975,264 (34.24 MB)

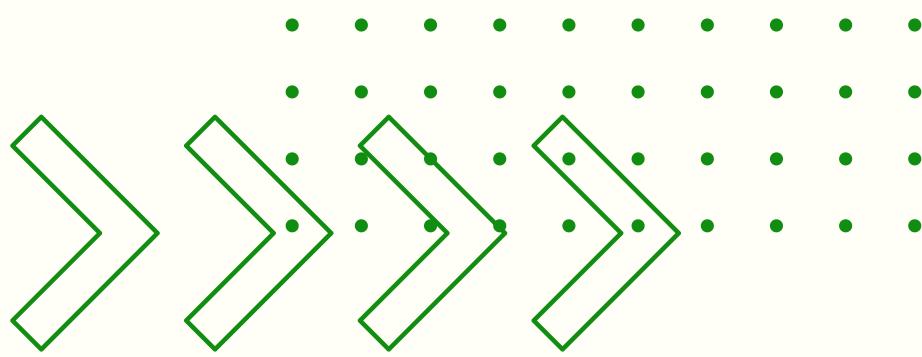
EfficientNetB3

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 30, 30, 75)	2,100
flatten_7 (Flatten)	(None, 67500)	0
dense_22 (Dense)	(None, 256)	17,280,256
dropout (Dropout)	(None, 256)	0
dense_23 (Dense)	(None, 128)	32,896
dropout_1 (Dropout)	(None, 128)	0
dense_24 (Dense)	(None, 5)	645

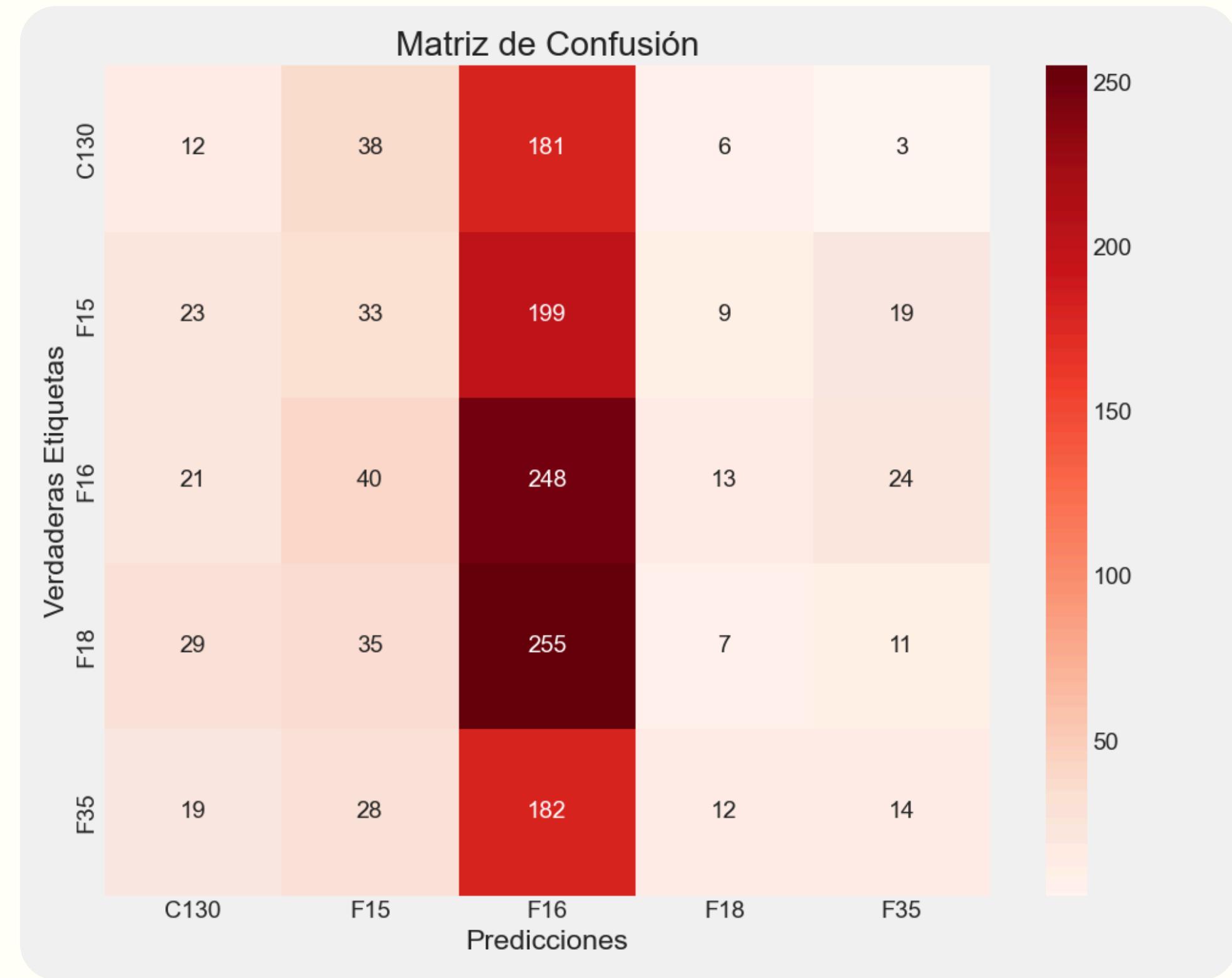


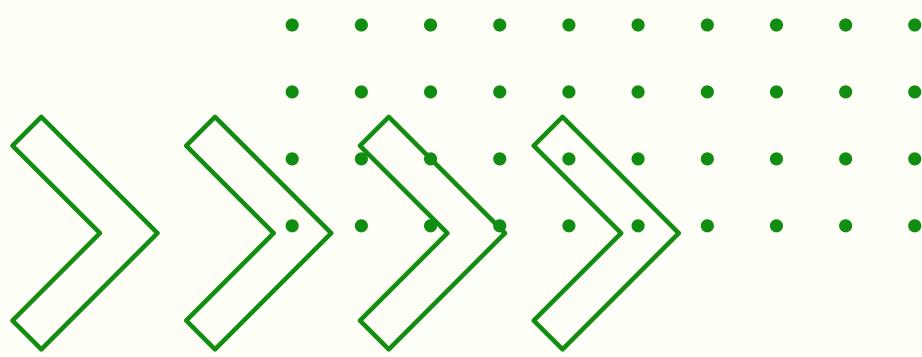
# INCEPTION V3





# INCEPTION V3

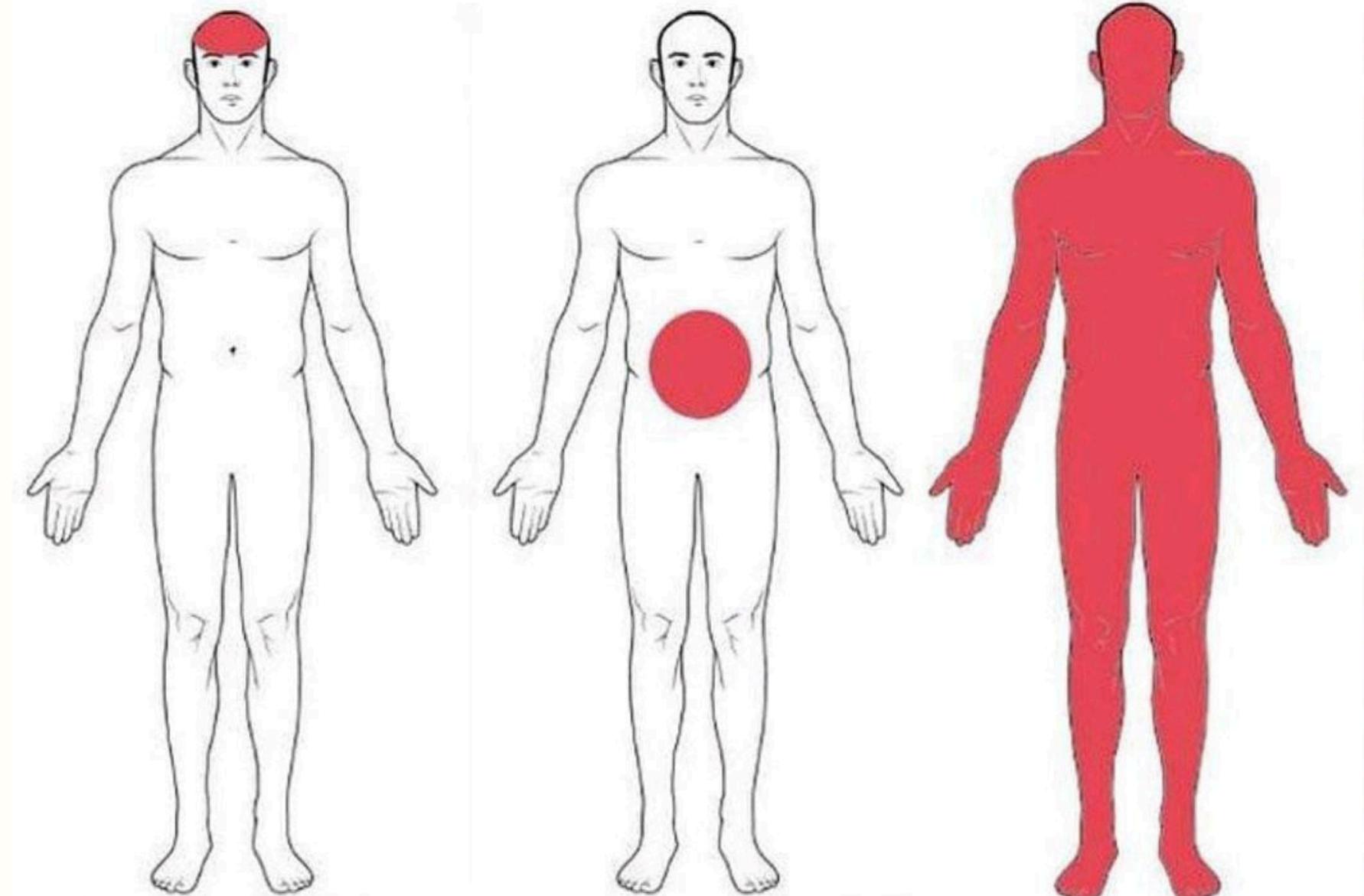
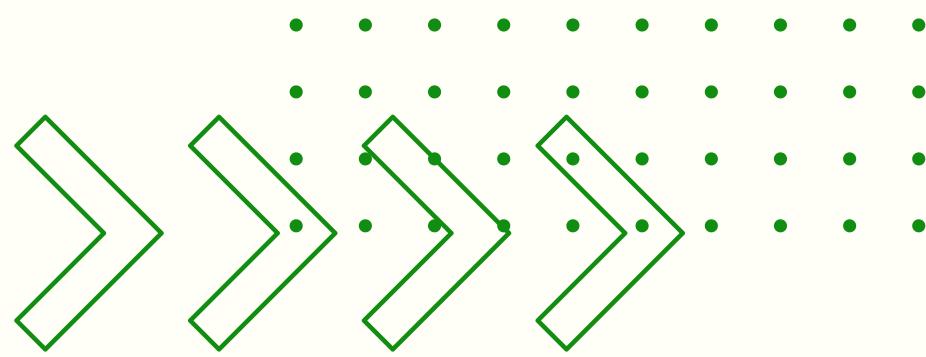




# INCEPTION V3

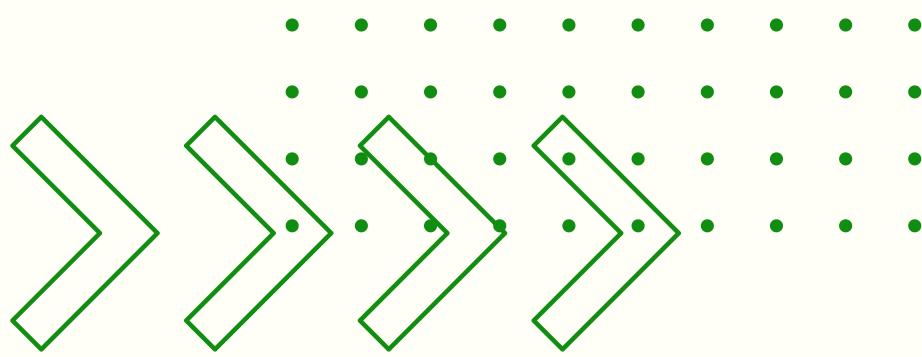
	precision	recall	f1-score	support
C130	0.12	0.05	0.07	240
F15	0.19	0.12	0.14	283
F16	0.23	0.72	0.35	346
F18	0.15	0.02	0.04	337
F35	0.20	0.05	0.09	255
accuracy			0.21	1461
macro avg	0.18	0.19	0.14	1461
weighted avg	0.18	0.21	0.15	1461

# Where does it hurt?

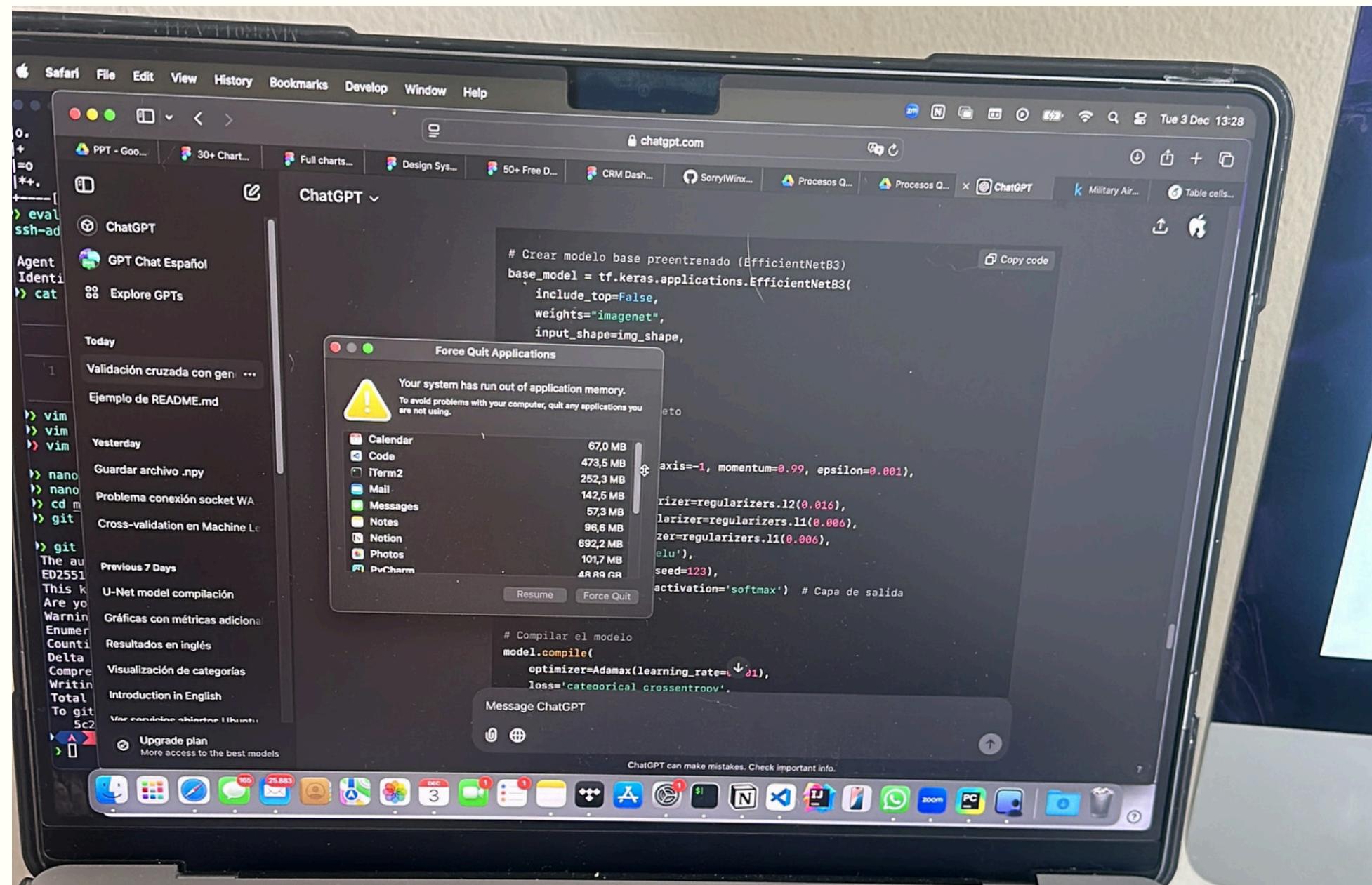


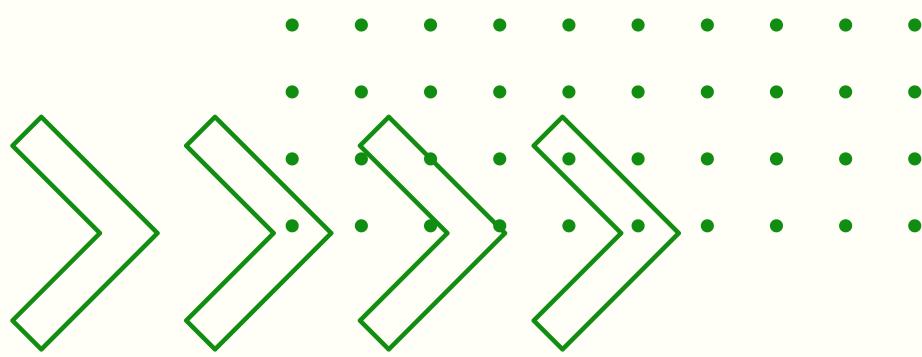
**Head ache**

**Stomach ache** Train acc: 0.98  
Test acc: 0.12



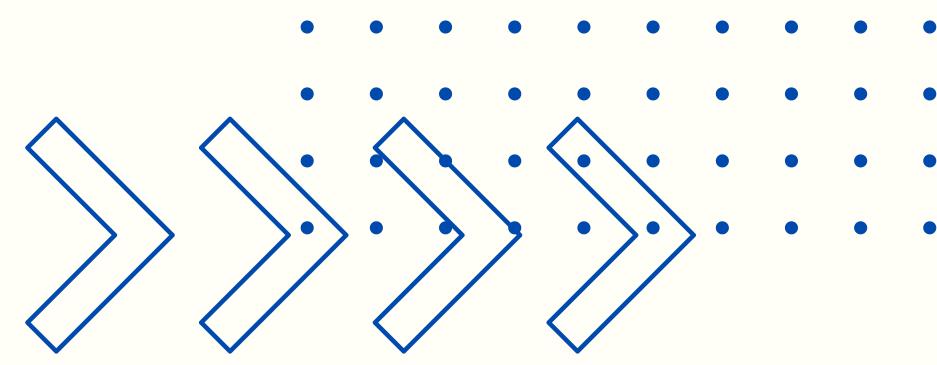
# EFFICIENTNETB3





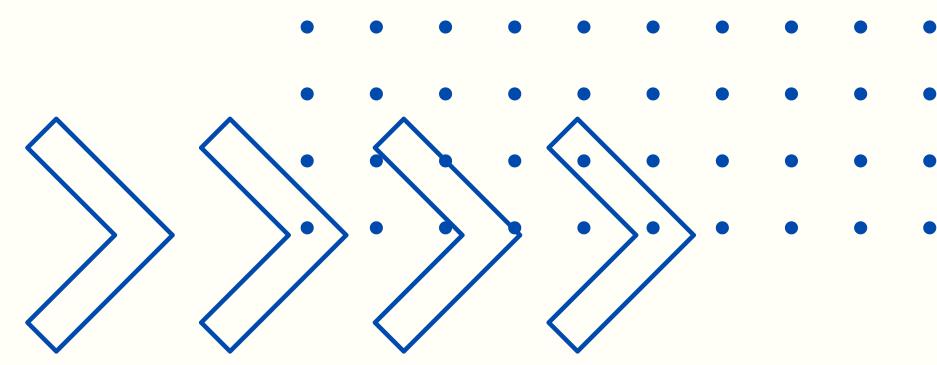
# EFFICIENTNETB3





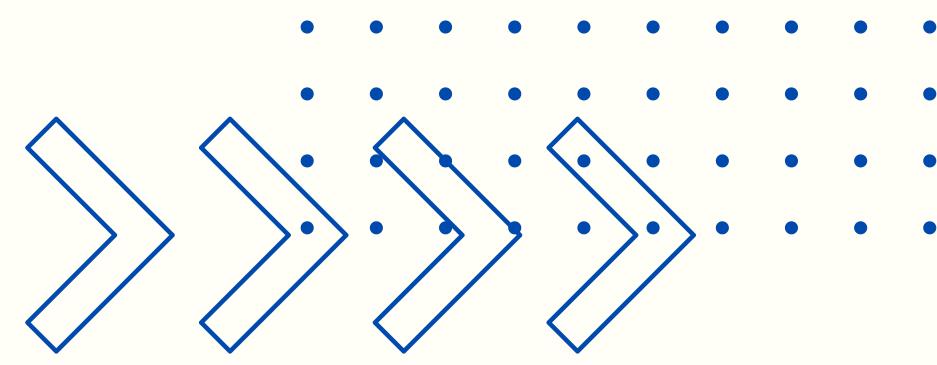
# RESULTADOS

Models \ Metrics	accuracy	auc	precision	recall
DNN base	0.4285	0.7184	0.5593	0.2163
DNN Inversa	0.4073	0.7195	0.5944	0.1896
DNN + Neuronas	0.4079	0.7189	0.5392	0.2307
DNN + Capas	0.4401	0.7403	0.5143	0.3450



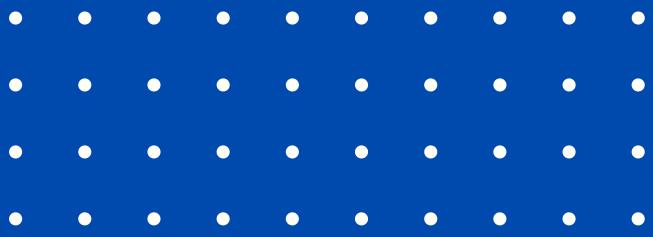
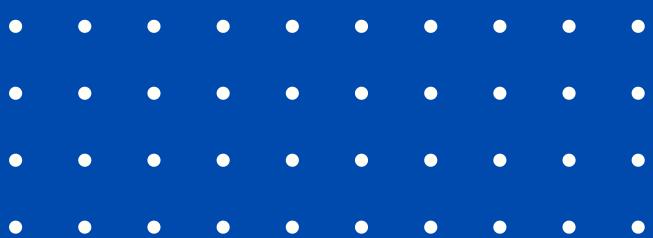
# RESULTADOS

Models \ Metrics	accuracy	auc	precision	recall
CNN base	0.5133	0.7830	0.5353	0.4771
CNN + Capas	0.5270	0.7625	0.5340	0.5209
CNN + Capas + Dropout	0.5633	0.8090	0.6005	0.5277
CNN + Conv + Capas + Dropout + MaxPooling	0.5202	0.7720	0.5659	0.4846
CNN V2	0.6194	0.8633	0.6477	0.5927
	0.6735	0.8922	0.7638	0.5667



# RESULTADOS

Models	Metrics	accuracy	auc	precision	recall
Inception V3		0.8323	0.9511	0.8385	0.8282
EfficientNetB3		Error	Error	Error	Error



# Gachas.