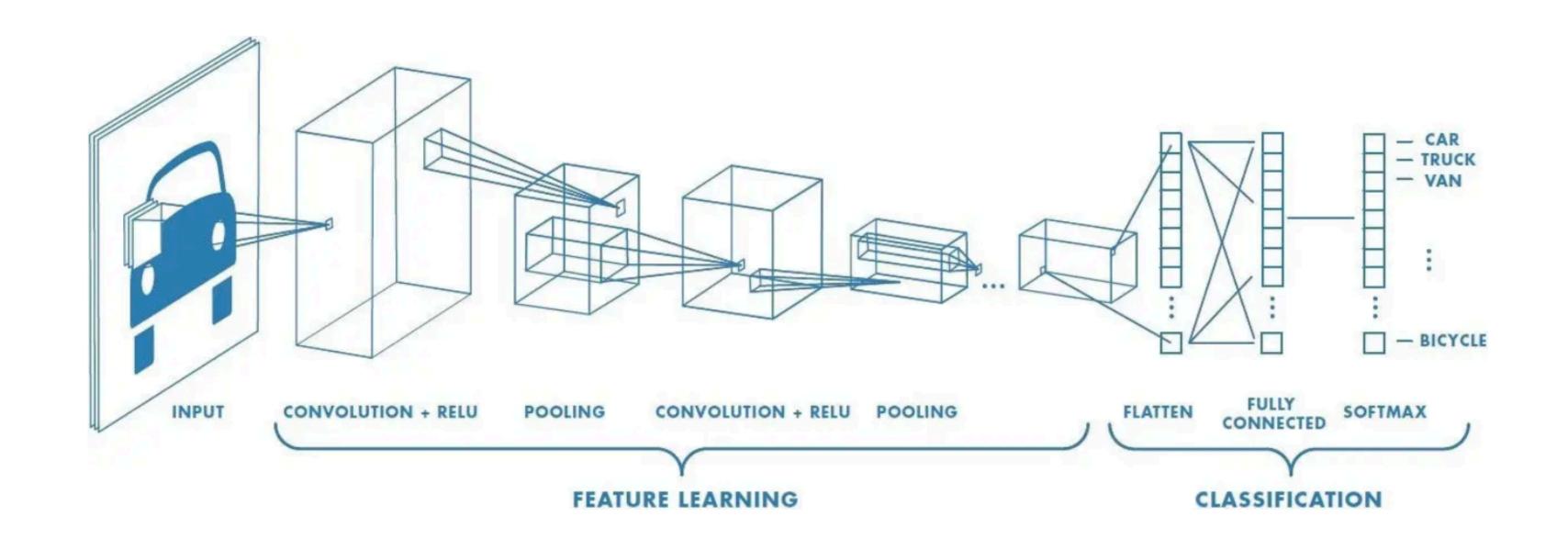
# CIFAR-10 IMAGE CLASSIFICATION

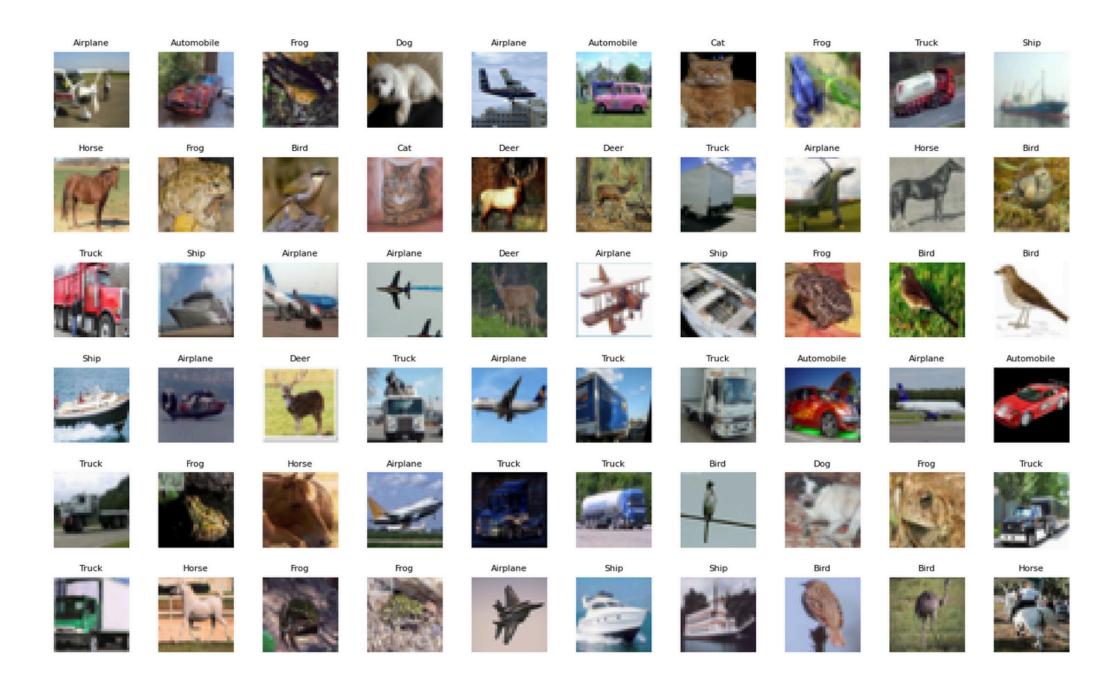
Adonis Kingsley Granita

# **OBJECTIVE**

Build a Convolutional Neural Network (CNN) to classify images into 10 distinct classes from the CIFAR-10 dataset.



# CIFAR-10



- **O** AIRPLANE
- **1** AUTOMOBILE
- 2 BIRD
- 3 CAT
- 4 DEER
- 5 DOG
- 6 FROG
- 7 HORSE
- 8 SHIP
- 9 TRUCK

# DATA PREPROCESSING

# **NORMALIZATION**

The pixel values where divided by 255, so that the scale ranges from 0 to 1

# **HOT-ONE-ENCODING**

I applied one-hot encoding to the labels converting them into binary matrix.

# **DATA SPLIT**

# TRAINING SET

42500 images

# **VALIDATION SET**

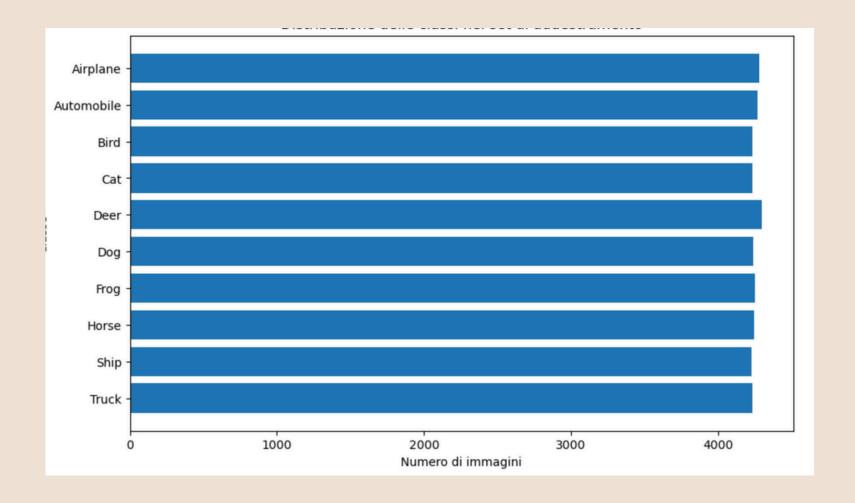
7500 images, 15% of the training set

# **TEST SET**

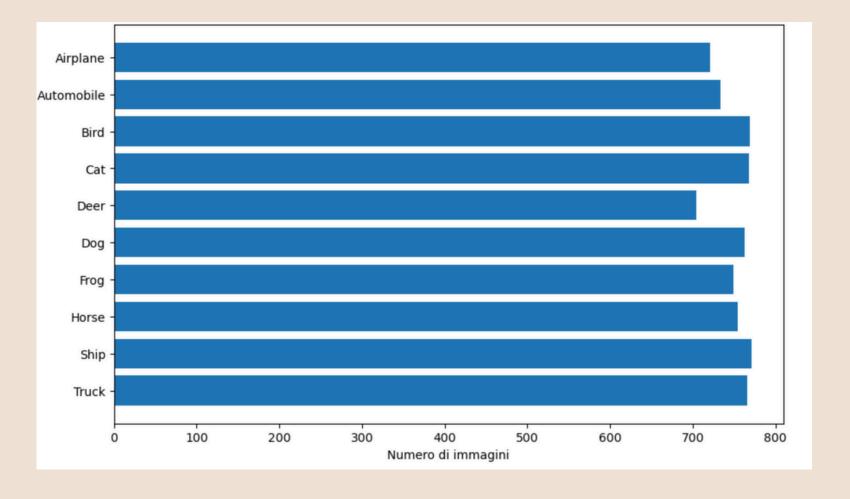
10 000 images

# CLASS DISTRIBUTION

### **TRAINING SET**



## **VALIDATION SET**



# EVALUATION METRICS

# **ACCURACY**

 $Accuracy = \frac{Number of Correct Predictions}{Total Number of Predictions}$ 

percentage of correct predictions made by the model out of all predictions.

# CATEGORICAL CROSSENTROPY LOSS

$$L_i = -\sum_{j=1}^C y_{ij} \log(\hat{y}_{ij})$$

how well the predicted probability distribution aligns with the true class, focusing on maximizing the probability of the correct class.

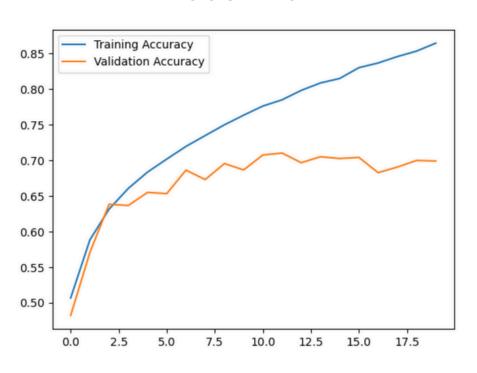
# MODEL 1

# **PROBLEM**

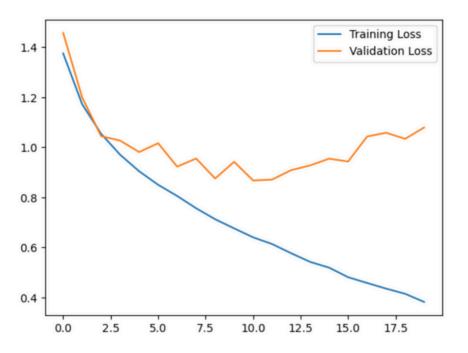
# **OVERFITTING**

Layer (type)	Output Shape
conv2d_6 (Conv2D)	(None, 30, 30, 32)
max_pooling2d_4 (MaxPooling2D)	(None, 15, 15, 32)
conv2d_7 (Conv2D)	(None, 13, 13, 64)
max_pooling2d_5 (MaxPooling2D)	(None, 6, 6, 64)
conv2d_8 (Conv2D)	(None, 4, 4, 64)
flatten_2 (Flatten)	(None, 1024)
dense_4 (Dense)	(None, 64)
dense_5 (Dense)	(None, 10)

### **ACCURACY**



### **CATEGORICAL CROSSENTROPY LOSS**



**TOTAL PARAMS: 122,570** 

TEST ACCURACY: 0.6877

**TEST LOSS: 1.1005** 

# MODEL 2

### **SOLUTIONS**

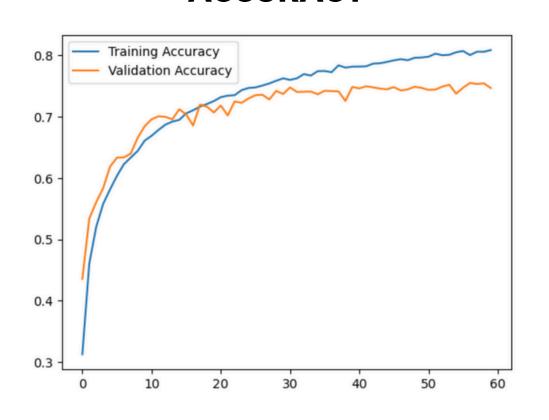
- DROPOUT
- DATA AUGMENTATION

## **PROBLEM**

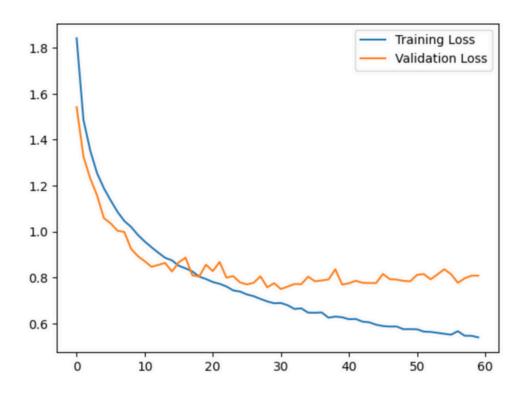
OVERFITTING

Layer (type)		Output Shape		
random_flip (RandomFlip)		(None, 32, 32, 3)		
conv2d_3 (Conv2D)	(None, 30, 30, 32)			
max_pooling2d_2 (MaxPool	(None, 15, 15, 32)			
conv2d_4 (Conv2D)	(None, 13, 13, 64)			
max_pooling2d_3 (MaxPooling2D)		(None, 6, 6, 64)		
conv2d_5 (Conv2D)		(None, 4, 4, 64)		
flatten_1 (Flatten)		(None, 1024)		
dense_2 (Dense)		(None, 64)		
dropout (Dropout)		(None, 64)		
dense_3 (Dense)		(None, 10)		

### **ACCURACY**



### **CATEGORICAL CROSSENTROPY LOSS**



**TOTAL PARAMS: 367,712** 

TEST ACCURACY: 0.7556 TEST LOSS: 0.8183

# MODEL 3

# **SOLUTIONS**

**PROBLEM** 

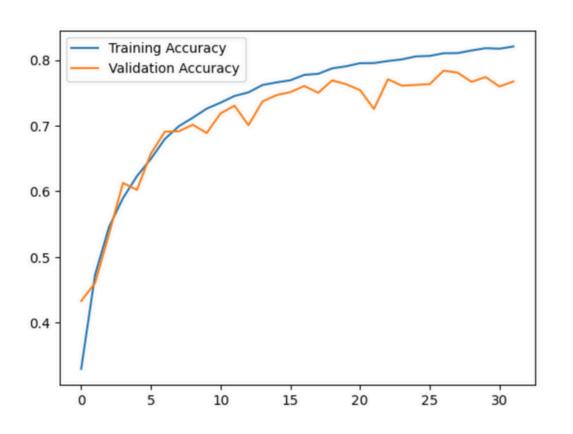
- L2 REGULARIZATION
- MORE COMPLEX MODEL
- PERFORMANCE COULD BE BETTER

Layer (type)	Output Shape
random_flip_15 (RandomFlip)	(None, 32, 32, 3)
conv2d_59 (Conv2D)	(None, 32, 32, 64)
max_pooling2d_39 (MaxPooling2D)	(None, 16, 16, 64)
conv2d_60 (Conv2D)	(None, 16, 16, 128)
max_pooling2d_40 (MaxPooling2D)	(None, 8, 8, 128)
conv2d_61 (Conv2D)	(None, 8, 8, 256)
max_pooling2d_41 (MaxPooling2D)	(None, 4, 4, 256)
conv2d_62 (Conv2D)	(None, 4, 4, 256)
flatten_16 (Flatten)	(None, 4096)
dense_37 (Dense)	(None, 128)
dense_38 (Dense)	(None, 64)
dense_39 (Dense)	(None, 10)

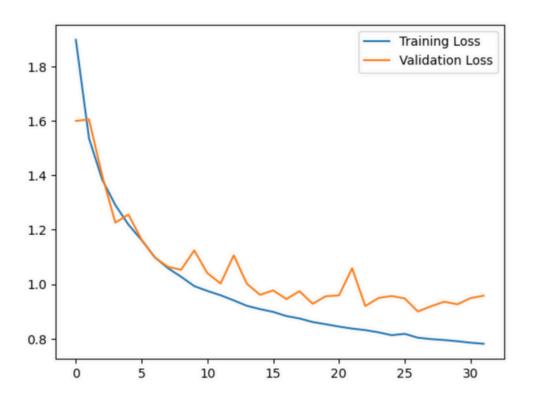
kernel\_regularizer=regularizers.l2(0.001)

**TOTAL PARAMS: 1,494,218** 

### **ACCURACY**



### **CATEGORICAL CROSSENTROPY LOSS**



TEST ACCURACY: 0.8169

**TEST LOSS: 0.8241** 

# FINAL MODEL

- INCREASE DEPTH AND FILTERS
  - STARTS AT 64, DOUBLES EACH BLOCK, UP TO 512
- DROPOUT
  - 0.2 ON THE FIRT BLOCKS
  - 0.5 ON THE LAST
- L2 REGULARIZERS
  - APPLIED TO ALL CONV LAYERS
- BATCH NORMALIZATION
  - AFTER ALL CONV LAYERS
- TRAINING CALLBACKS
  - EARLY STOPPING
  - LEARNING RATE SCHEDULING

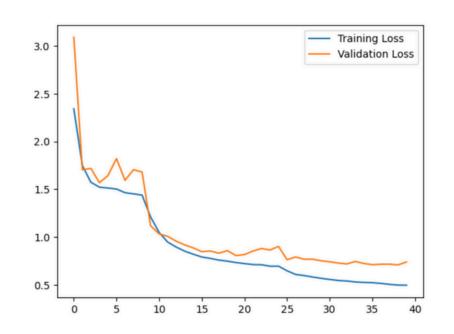
Layer (type)	Output Shape
random_flip_4 (RandomFlip)	(None, 32, 32, 3)
conv2d_15 (Conv2D)	(None, 32, 32, 64)
batch_normalization (BatchNormalization)	(None, 32, 32, 64)
max_pooling2d_10 (MaxPooling2D)	(None, 16, 16, 64)
dropout_3 (Dropout)	(None, 16, 16, 64)
conv2d_16 (Conv2D)	(None, 16, 16, 128)
batch_normalization_1 (BatchNormalization)	(None, 16, 16, 128)
max_pooling2d_11 (MaxPooling2D)	(None, 8, 8, 128)
dropout_4 (Dropout)	(None, 8, 8, 128)
conv2d_17 (Conv2D)	(None, 8, 8, 256)
batch_normalization_2 (BatchNormalization)	(None, 8, 8, 256)
max_pooling2d_12 (MaxPooling2D)	(None, 4, 4, 256)
dropout_5 (Dropout)	(None, 4, 4, 256)
conv2d_18 (Conv2D)	(None, 4, 4, 512)
batch_normalization_3 (BatchNormalization)	(None, 4, 4, 512)
max_pooling2d_13 (MaxPooling2D)	(None, 2, 2, 512)
dropout_6 (Dropout)	(None, 2, 2, 512)
flatten_5 (Flatten)	(None, 2048)
dense_10 (Dense)	(None, 128)
dense_11 (Dense)	(None, 64)
dense_12 (Dense)	(None, 10)

# FINAL MODEL EVALUATION

### **ACCURACY**

# 0.9 - Training Accuracy Validation Accuracy 0.8 - 0.7 - 0.6 - 0.5 - 0.4 - 0.3 - 0 - 5 - 10 - 15 - 20 - 25 - 30 - 35 - 40

### **CATEGORICAL CROSSENTROPY LOSS**



TEST ACCURACY: 0.8637 TEST LOSS:0.7141

### **CLASSIFICATION REPORT**

	precision	recall	f1-score	support
0	0.87	0.87	0.87	1000
1	0.94	0.91	0.93	1000
2	0.85	0.81	0.83	1000
3	0.79	0.69	0.74	1000
4	0.83	0.87	0.85	1000
5	0.83	0.79	0.81	1000
6	0.84	0.94	0.88	1000
7	0.90	0.90	0.90	1000
8	0.86	0.94	0.90	1000
9	0.92	0.90	0.91	1000
accuracy			0.86	10000
macro avg	0.86	0.86	0.86	10000
weighted avg	0.86	0.86	0.86	10000

01

# **PETRAINED MODELS**

Use transfer learning could had lead to a better performing model

FUTURE DEVELOPMENTS 02

# **TRAINING TIME**

increasing the number of epoch of the final model could have increase its performance

03

# **MODEL SEARCH**

using for instance Neural Architecture Search - NAS to find the right complexity of the model

# THANKS

**LINK TO THE MODELS**