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Iniciar

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You can describe the topic of the section here

You can describe the topic of the section here









INTRODUCCION



Desarrollo de un modelo ML supervisado para Clasificación de imágenes



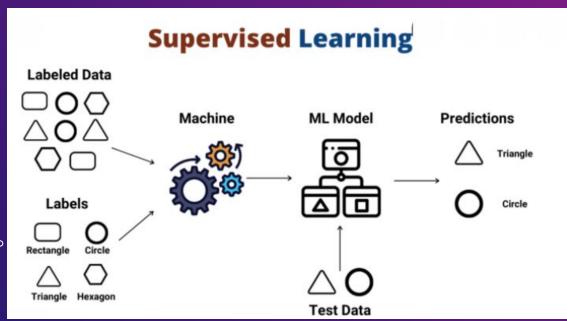






PROBLEMA VS SOLUCIÓN

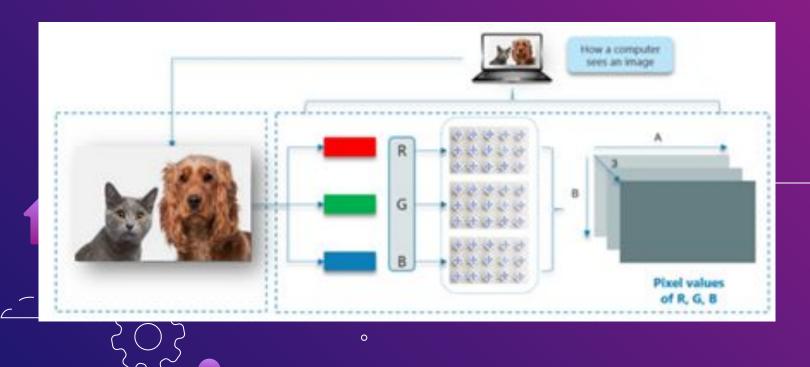




















SETPARA ENTRENAR EL MODELO







CIFAR-10

- Instituto Canadiense de Investigación Avanzada
- 60000 imágenes en color de 32x32
- 10 clases
- Target : 0-9







CIFAR-10

Normalizar la data

```
# Parse numbers as floats
train_x=train_X.astype('float32')
test_X=test_X.astype('float32')
# Normalize data
train_X=train_X/255.0
test_X=test_X/255.0
```

```
print("Shape of x train: {}".format(train X.shape))
print("Shape of y train: {}".format(train Y.shape))
print()
print("Shape of x test: {}".format(test X.shape))
print("Shape of y test: {}".format(test Y.shape))
Shape of x train: (50000, 32, 32, 3)
Shape of y train: (50000, 1)
Shape of x test: (10000, 32, 32, 3)
Shape of y test: (10000, 1)
 train X = \text{train } X.\text{reshape}(50000,32*32*3)
 test X = \text{test } X.\text{reshape}(10000, 32*32*3)
  (50000, 3072)
  (10000, 3072)
```





##





PARA TRAINNING









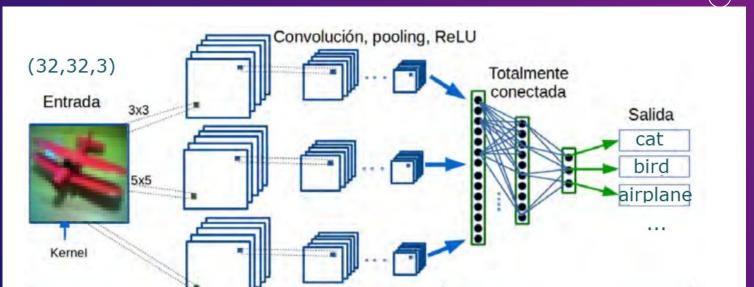






REDES CONVOLUCIONALES - CNN

Extracción de características



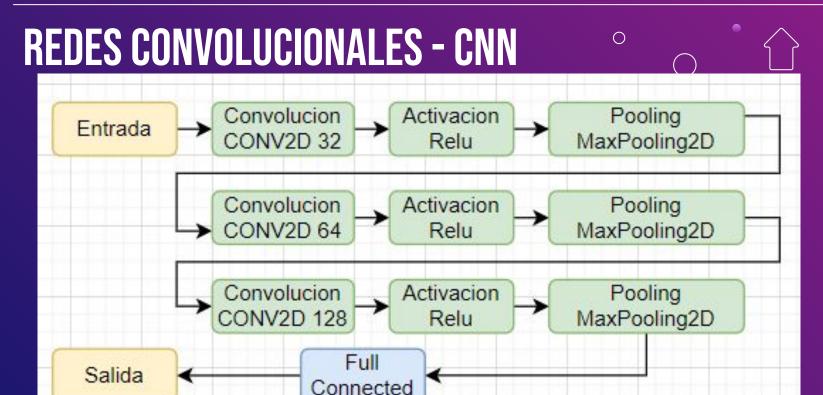
Clasificación



- CONV2
- RELU
- POOLING











MENU

REDES CONVOLUCIONALES - CNN







```
model = Sequential()
model.add(Conv2D(32, kernel size=(3, 3), activation='relu', input shape=(32,32,3)))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Conv2D(64, kernel size=(3, 3), activation='relu'))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Conv2D(128, kernel size=(3, 3), activation='relu'))
model.add(MaxPooling2D(pool size=(2, 2)))
model.add(Flatten())
model.add(Dense(256, activation='relu'))
model.add(Dense(128, activation='relu'))
model.add(Dense(num classes, activation='softmax'))
```



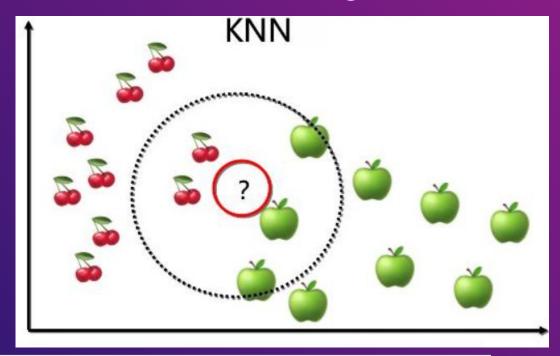


KNN - VECINOS MÁS CERCANOS





- Calcular la distancia entre el item a clasificar y el resto de items
- Seleccionar los "k" elementos más cercanos
- Realizar una "votación de mayoría" entre los k puntos







LIBRERÍAS USADAS

from sklearn import metrics

from sklearn.neighbors import KNeighborsClassifier

import dependencies
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import Dropout
from keras.layers import Flatten
from keras.constraints import maxnorm
from keras.layers.convolutional import Conv2D
from keras.layers.convolutional import MaxPooling2D
from keras.datasets import cifar10















##



















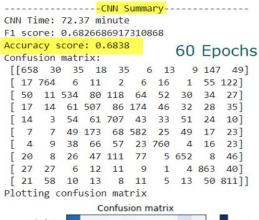


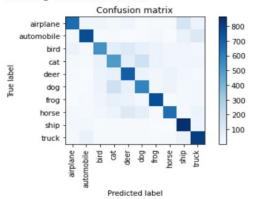
EVALUACION

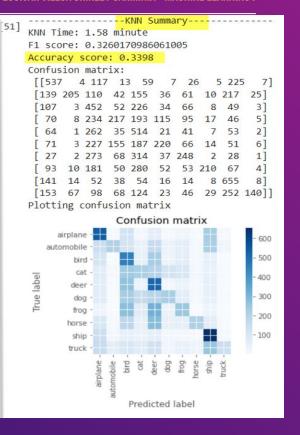


- F1 Score
- Accuracy
- Confusion Matrix



















& RECOMENDACIONES •







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CONCLUSIONES

REDES CONVOLUCIONALES 68%

KNN 33%

KMEANS 8%

RECOMENDACIONES

- Trabajo futuro , mejoras
- Estudiar el algoritmo elegido
- Estudiar procesamiento









• GRACIAS!

























