# Implementasi Klasifikasi Gambar menggunakan CNN's

Langkah-langkah Data Pre-processing:

```
In [1]: # Mengimport Library yang dibutuhkan
    from __future__ import absolute_import, division, print_function, unicode_literals
    # TensorFLow dan tf.keras
    import tensorflow as tf
    from tensorflow import keras

# Library pembantu
    import numpy as np
    import matplotlib.pyplot as plt

In [2]: # Import Dataset
    cifar10 = keras.datasets.cifar10
```

```
# Proses Normalisasi gambar
           train images = train images / 255.0
           test images = test images / 255.0
          Jumlah Gambar Pelatihan :(50000, 32, 32, 3)
          Jumlah Label Pelatihan :(50000, 1)
          Jumlah Gambar Uji :(10000, 32, 32, 3)
          Jumlah Label Uji :(10000, 1)
          Kumpulan Label Pelatihan :[[6]
           [9]
           [9]
           [9]
           [1]
           [1]]
          Kumpulan Label Uji :[[3]
           [8]
           [8]
           . . .
           [5]
           [1]
           [7]]
 In [3]:
           # Bentuk ulang data input dari (32, 32) menjadi (32, 32, 1)
           w, h = 32, 32
           train images = train images.reshape(train images.shape[0], w, h, 3)
           test images = test images.reshape(test images.shape[0], w, h, 3)
In [28]:
           ###### from tensorflow.keras import backend as K
           import tensorflow as tf
           K.clear session()
           model = tf.keras.Sequential()
           # Menentukan bentuk input di lapisan pertama jaringan saraf
           model.add(tf.keras.layers.Conv2D(filters=64, kernel size=4, padding='same', activation='relu', input shape=(32,32,3)))
           model.add(tf.keras.layers.MaxPooling2D(pool size=2))
           # model.add(tf.keras.layers.Dropout(0.3))
           # Lapisan Konvolusi Kedua
           model.add(tf.keras.layers.Conv2D(filters=32, kernel size=4, padding='same', activation='relu'))
           model.add(tf.keras.layers.MaxPooling2D(pool size=2))
           model.add(tf.keras.layers.Dropout(0.3))
           # Lapisan Fully Connected
           model.add(tf.keras.layers.Flatten())
           model.add(tf.keras.layers.Dense(256, activation='relu'))
           model.add(tf.keras.layers.Dropout(0.5))
```

```
model.add(tf.keras.layers.Dense(10, activation='softmax'))
# Melihat ringkasan model
model.summary()
```

Model: "sequential"

Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	32, 32, 64)	3136
max_pooling2d (MaxPooling2D)	(None,	16, 16, 64)	0
conv2d_1 (Conv2D)	(None,	16, 16, 32)	32800
max_pooling2d_1 (MaxPooling2	(None,	8, 8, 32)	0
dropout (Dropout)	(None,	8, 8, 32)	0
flatten (Flatten)	(None,	2048)	0
dense (Dense)	(None,	256)	524544
dropout_1 (Dropout)	(None,	256)	0
dense_1 (Dense)	(None,	10)	2570
Total params: 563,050 Trainable params: 563,050	=====		======

Total params: 563,050 Trainable params: 563,050 Non-trainable params: 0

## Kompilasi Model

## **Melatih Model**

```
Epoch 2/10
   Epoch 3/10
   Epoch 4/10
   Epoch 5/10
   Epoch 6/10
   Epoch 7/10
   Epoch 8/10
   Epoch 9/10
   Epoch 10/10
   Out[30]: <tensorflow.python.keras.callbacks.History at 0x7f53ec5a9a60>
In [31]:
    # Mengevaluasi Model menggunakan Test Set
    test_loss, test_acc = model.evaluate(test_images, test_labels, verbose=2)
    print(f'\nTest accuracy: {test acc:0.2f}')
    print(f'\ntest loss: {test loss:0.2f}')
   313/313 - 0s - loss: 0.8497 - accuracy: 0.7051
   Test accuracy: 0.71
   test loss: 0.85
```

#### Membuat Prediksi:

```
In [32]: # Membuat prediksi dari data test_images
    predictions = model.predict(test_images)

In [33]: # Mengubah input data dari (28, 28) menjadi (28, 28, 1)
    w, h = 32, 32
    train_images = train_images.reshape(train_images.shape[0], w, h, 3)
    test_images = test_images.reshape(test_images.shape[0], w, h, 3)
```

# Fungsi bantuan untuk Visualisasi gambar

def plot image(i, predictions array, true label, img):

```
predictions array,true label, img = predictions array, true label[i], img[i]
             plt.grid(False)
             plt.xticks([])
             plt.yticks([])
             plt.imshow(img, cmap=plt.cm.binary)
             predicted label = np.argmax(predictions array)
             if predicted label == true label:
               color = 'blue'
             else:
               color = 'red'
             plt.xlabel("{} {:2.0f}% ({})".format(class_names[predicted_label],
                                           100*np.max(predictions array),
                                           class names[true label]),
                                           color=color)
           def plot value array(i, predictions array, true label):
             predictions array, true label = predictions array, true label[i]
             plt.grid(False)
             plt.xticks(range(10))
             plt.yticks([])
             thisplot = plt.bar(range(10), predictions array, color="#777777")
             plt.ylim([0, 1])
             predicted label = np.argmax(predictions array)
             thisplot[predicted label].set color('red')
             thisplot[true label].set color('blue')
In [34]:
           import numpy as np
           test labels=np.ravel(test labels)
           test labels
Out[34]: array([3, 8, 8, ..., 5, 1, 7], dtype=uint8)
In [35]:
           # Plot X gambar pertama pada test images, label hasil prediksi , dan label yang sebenarnya.
           # Warnai prediksi yang benar dengan warna biru dan prediksi yang salah dengan warna merah.
```

```
num\_rows = 5
  num cols = 3
  num images = num rows*num cols
  plt.figure(figsize=(3*2*num_cols, 2*num_rows))
 for i in range(num_images):
    plt.subplot(num_rows, 2*num_cols, 2*i+1)
    plot image(i, predictions[i], test labels, test images)
    plt.subplot(num_rows, 2*num_cols, 2*i+2)
    plot value array(i, predictions[i], test labels)
  plt.tight_layout()
  plt.show()
                                                                                                                     0 1 2 3 4 5 6 7 8 9
                     0 1 2 3 4 5 6 7 8 9
                                                                     0 1 2 3 4 5 6 7 8 9
    cat 44% (cat)
                                                                                                   ship 40% (ship)
                                                                                                                     0 1 2 3 4 5 6 7 8 9
 airplane 76% (airplane)
                     0 1 2 3 4 5 6 7 8 9
                                                   frog 69% (frog)
                                                                     0 1 2 3 4 5 6 7 8 9
                                                                                                   frog 97% (frog)
                     0 1 2 3 4 5 6 7 8 9
                                                                     0 1 2 3 4 5 6 7 8 9
                                                                                                                     0 1 2 3 4 5 6 7 8 9
utomobile 96% (automobile)
                                                   frog 75% (frog)
                                                                                                    cat 57% (cat)
                                                                     0 1 2 3 4 5 6 7 8 9
                                                                                                                     0 1 2 3 4 5 6 7 8 9
                     0 1 2 3 4 5 6 7 8 9
                                                 airplane 63% (airplane)
                                                                                                  truck 100% (truck)
utomobile 83% (automobile)
```

#### Kesimpulan:

frog 25% (dog)

0 1 2 3 4 5 6 7 8 9

horse 99% (horse)

0 1 2 3 4 5 6 7 8 9

truck 100% (truck)

0 1 2 3 4 5 6 7 8 9

Menjalankan kedua model selama 5 Epochs, berikut adalah tabel perbandingannya:

	Model	Train Accuracy	Train Loss	Test Accuracy	Test Loss
	Fully connected Neural Networks -After 10 Epochs	0.8497	0.7051	0.71	0.85
In [ ]:					