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Tugas: Eksplorasi Hyperparameter CNN dan Neural Network

A. CNN

Data : Cifar10
Data Type : Image
Size : 32 x 32 Pixels
Channel size: 3 (RGB)

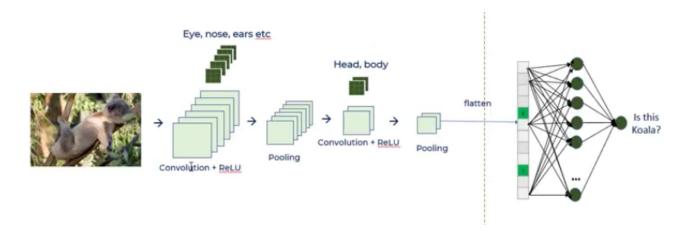
Jumlah data : 50.000 images

Jumlah Class: 10

['pesawat','mobil','burung','kucing','rusa','anjing','katak','kuda','kapal',

'truk']

Skema CNN



Berapa banyaknya convolution layar yang optimal? **Jawab : 2**

Berapa ukuran filter yang optimal untuk setiap convolution layar? $\mathbf{Jawab}: \mathbf{3} \times \mathbf{3}$ Berapa banyaknya filter yang optimal untuk setiap convolution layar? $\mathbf{Jawab}:$

Filter Convolution 1:32 Filter Convolution 2:64

Berapa banyaknya hidden unit yang optimal pada bagian fully connected

network? Jawab: 64

Dari semua pilihan yang disedikan oleh Keras Optimizer, mana yang menghasilkan kinerja paling baik (pada nilai parameter default) ? **Jawab : SGD**

Dari semua pilihan yang disedaikan oleh Keras (Probabilistic) Losses, mana yang menghasilkan kinerja paling baik?

Jawab: sparse_categorical_crossentropy

Eksplorasi 1

Jumlah Convolution layer: 1 Jumlah Maxpooling: 1

ukuran filter: 32 / convolutional layer

filter size : 3 x 3 fungsi aktivasi : Relu

banyaknya hidden unit fully connected network: 64

hasil explorasi 1

eksplorasi 2

Jumlah Convolution layer : 2 Jumlah Maxpooling : 2

ukuran filter: 32 / convolutional layer

filter size : 3 x 3 fungsi aktivasi : relu

banyaknya hidden unit fully connected network: 64

hasil explorasi 2

eksplorasi 3

Jumlah Convolution layer: 2 Jumlah Maxpooling: 2

ukuran filter: 64 / convolutional layer

filter size : 3 x 3 fungsi aktivasi : relu

banyaknya hidden unit fully connected network: 64

eksplorasi 4

Jumlah Convolution layer: 2 Jumlah Maxpooling: 2

ukuran filter 1 : 32 / convolutional layer ukuran filter 2 : 64 / convolutional layer

filter size : 3 x 3 fungsi aktivasi : relu

banyaknya hidden unit fully connected network: 64

eksplorasi 5

Jumlah Convolution layer: 3 Jumlah Maxpooling: 3

ukuran filter 1 : 32 / convolutional layer ukuran filter 2 : 32 / convolutional layer ukuran filter 3 : 64 / convolutional layer

filter size : 3 x 3 fungsi aktivasi : relu

banyaknya hidden unit fully connected network: 64

eksplorasi 6

Jumlah Convolution layer: 2 Jumlah Maxpooling: 2

ukuran filter 1:32 / convolutional layer ukuran filter 2:64 / convolutional layer

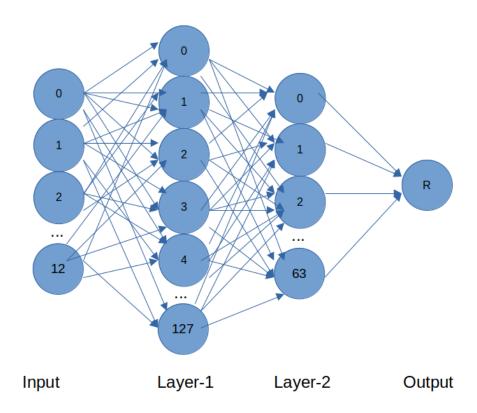
filter size : 3 x 3 fungsi aktivasi : relu Optimizer : SGD

B. Fully Connected NN

Data: Boston Housing Price
Data Type: Tabel numerik (float)

Jumlah Kolom: 13 Jumlah baris: 506

skema NN



Hasil Ekplorasi

Berapa banyaknya hidden layar yang optimal? Jawab: 2

Berapa banykanya hidden unit yang optimal di setiap hidden layar? **Jawab: 128 dan 64** Apa activation function di setiap layaer sehingga hasilnya optimal? **Jawab: Relu** Dari semua pilihan optimer, apa optimizer yang hasilnya optimal? **Jawab: Adam** Dari semua pilihan loss function, apa yang hasilnya optimal?

Jawab = Mean Absolute Error (MAE)

Detail Eksplorasi

Eksplorasi 1

Eksplorasi 2

```
def HousePricePredictionModel():
    model=Sequential()
    model.add(Dense(128,activation='relu',input_shape=(train_x[0].shape)))
    model.add(Dense(64,activation='relu'))
    model.add(Dense(32,activation='relu'))
    model.add(Dense(16,activation='relu'))
    model.add(Dense(8,activation='relu'))
    model.add(Dense(1))
    model.compile(optimizer='Adam',loss='mse', metrics=['mae'])
    return model
```

Eksplorasi 3

```
[142] def HousePricePredictionModel():
    model=Sequential()
    model.add(Dense(128,activation='relu',input_shape=(train_x[0].shape)))
    model.add(Dense(64,activation='relu'))
    model.add(Dense(32,activation='relu'))
    model.add(Dense(16,activation='relu'))
    model.add(Dense(1))
    model.compile(optimizer='Adam',loss='mse', metrics=['mae'])
    return model
```

```
Epoch 95/100
404/404 [===
                        ========] - 2s 4ms/step - loss: 15.5132 - mae: 2.8745 - val loss: 30.8432 - val mae: 3.7887
Epoch 96/100
                       ========] - 1s 4ms/step - loss: 15.4672 - mae: 2.8593 - val loss: 29.3681 - val mae: 3.6561
404/404 [===
Epoch 97/100
404/404 [===
                           ========] - 1s 3ms/step - loss: 13.7625 - mae: 2.7412 - val_loss: 28.2854 - val_mae: 3.5518
Epoch 98/100
404/404 [====
                       ========] - 1s 3ms/step - loss: 15.8283 - mae: 2.9117 - val_loss: 29.8397 - val_mae: 3.6351
Epoch 99/100
                         :=======] - 1s 3ms/step - loss: 16.1588 - mae: 2.8430 - val loss: 26.5212 - val mae: 3.4705
404/404 [===
Epoch 100/100
                       ========] - 1s 3ms/step - loss: 16.0148 - mae: 2.8513 - val_loss: 33.0665 - val_mae: 4.0061
<keras.callbacks.History at 0x7f52f5841310>
```

Eksplorasi 4

```
[145] def HousePricePredictionModel():
    model=Sequential()
    model.add(Dense(128,activation='relu',input_shape=(train_x[0].shape)))
    model.add(Dense(64,activation='relu'))
    model.add(Dense(32,activation='relu'))
    model.add(Dense(1))
    model.compile(optimizer='Adam',loss='mse', metrics=['mae'])
    return model
```

```
Epoch 95/100
                          :=======] - 1s 3ms/step - loss: 14.4224 - mae: 2.7463 - val_loss: 29.2971 - val_mae: 3.6206
404/404 [===
Epoch 96/100
                                ======] - 1s 3ms/step - loss: 15.5237 - mae: 2.8958 - val_loss: 28.7757 - val_mae: 3.5866
404/404 [==:
Epoch 97/100
404/404 [===
                               ======] - 1s 3ms/step - loss: 15.8912 - mae: 2.8684 - val loss: 32.9450 - val mae: 3.8845
Epoch 98/100
                       :=========] - 1s 4ms/step - loss: 17.3785 - mae: 2.8855 - val_loss: 31.2902 - val_mae: 3.7443
404/404 [===:
Epoch 99/100
404/404 [===
                           ========] - 1s 4ms/step - loss: 16.5671 - mae: 2.9186 - val loss: 28.8252 - val mae: 3.6248
Epoch 100/100
404/404 [=
                          :=======] - 1s 4ms/step - loss: 14.8916 - mae: 2.7747 - val loss: 28.6249 - val mae: 3.7039
<keras.callbacks.History at 0x7f52f5597910>
```

Eksplorasi 5

```
[151] def HousePricePredictionModel():
    model=Sequential()
    model.add(Dense(128,activation='relu',input_shape=(train_x[0].shape)))
    model.add(Dense(64,activation='relu'))
    model.add(Dense(1))
    model.compile(optimizer='Adam',loss='mse', metrics=['mae'])
    return model
```

Epoch 95/100	_	_
404/404 [==========] - 1s 3ms/step - loss: 15.1599 - mae:	2.7534 - val_loss: 35.3469	- val_mae: 4.1609
Epoch 96/100		
404/404 [===================================	2.8228 - val_loss: 33.5135	val_mae: 4.0289
Epoch 97/100		
404/404 [===================================	2.8924 - val_loss: 29.2708	val_mae: 3.6973
Epoch 98/100		
404/404 [===================================	2.8844 - val loss: 29.8191	- val mae: 3.8377
Epoch 99/100		
404/404 [=============] - 1s 3ms/step - loss: 15.0095 - mae:	2.7845 - val_loss: 32.4882	- val_mae: 3.8248
Epoch 100/100		
404/404 [===========] - 1s 3ms/step - loss: 15.8758 - mae:	2.8643 - val loss: 27.6126	- val mae: 3.5228
<pre><keras.callbacks.history 0x7f52f5447790="" at=""></keras.callbacks.history></pre>	_	_