

Menjadi Powerful Dengan Tensor(flow)

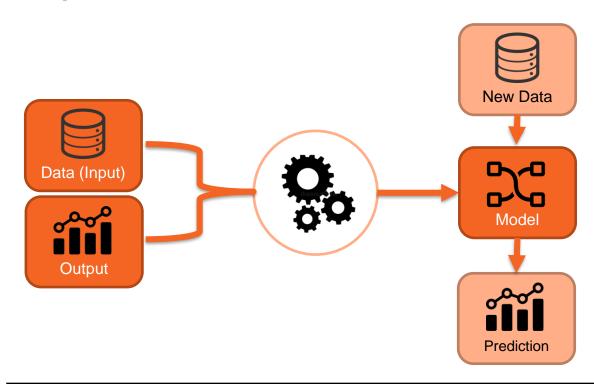
Part 2



Aditya Firman Ihsan • Tensorflow Developer Certified KK Data Science Telkom University

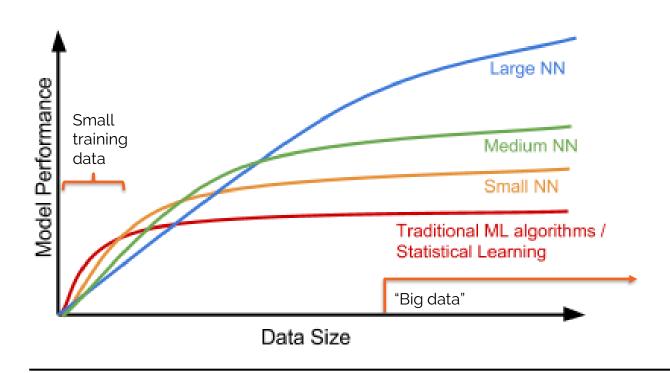
Review dulu

Bagaimana mesin belajar?

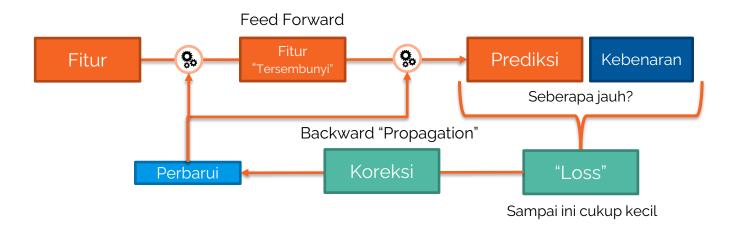


Bagaimana mesin belajar? Supervis<u>ed</u> Unsupervised Reinforcement Classification Regression Clustering Control **GLM** Monte Carlo Neural Neural Neural Neural Decision Trees Networks Networks Networks Networks

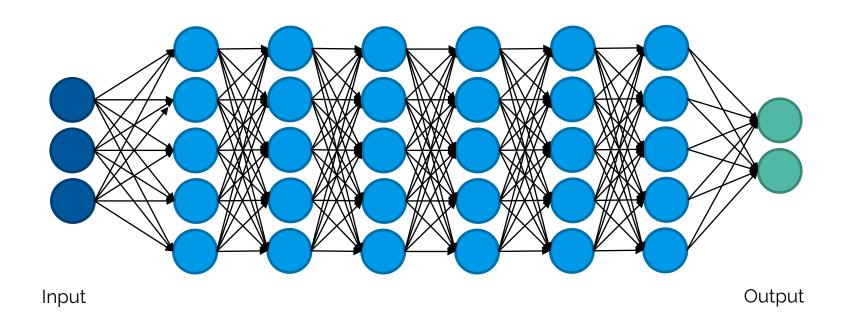
Kenapa Neural Network (NN)?



Apa yang dilakukan NN?



Apa yang dilakukan NN?



Kenapa harus tensor?

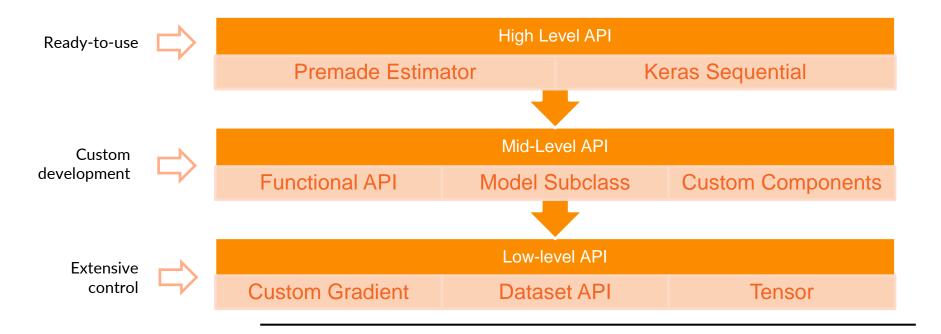
Tensor merupakan array yang menerapkan paradigma differentiable programming

Sederhananya, konsep program dimana perhitungan numerik dapat dihitung turunannya melalui graf komputasi yang dibangun

Tensorflow

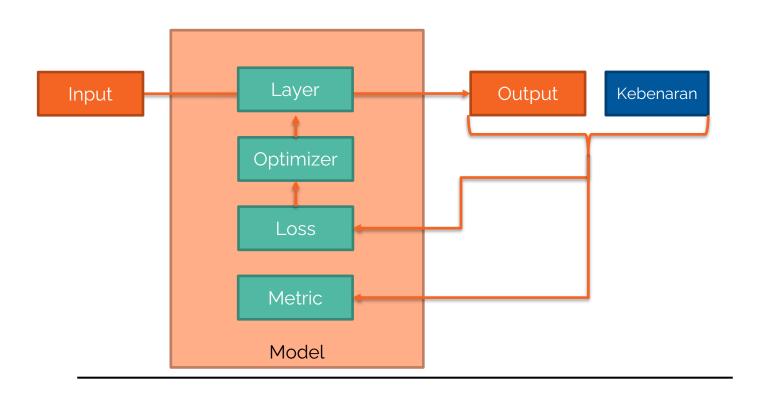
sebuah framework *Deep Learning*, yang memanfaatkan konsep tensor untuk efektivitas komputasi

Fleksibilitas TF

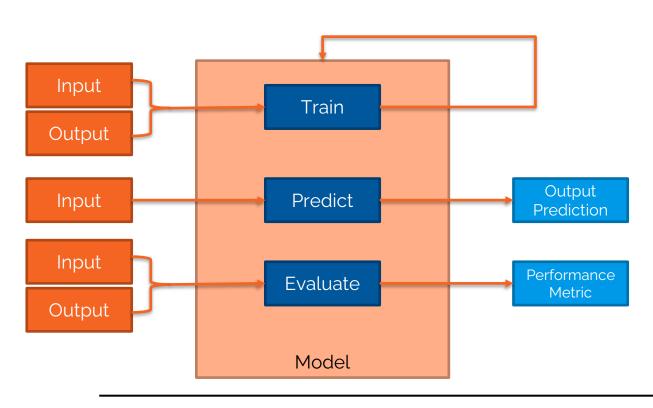


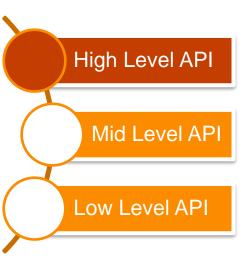
Bagaimana Memakainya?

Apa yang mendefinisikan suatu model NN?



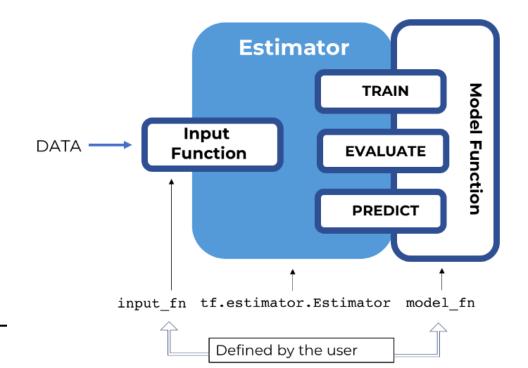
Apa yang mendefinisikan suatu model NN?

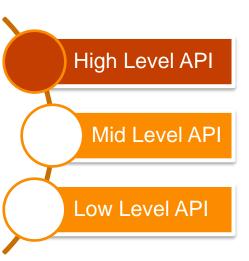




1. Premade Estimator

Estimator adalah kelas di TF yang membungkus suatu model siap pakai.



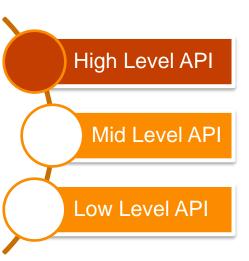


1. Premade Estimator

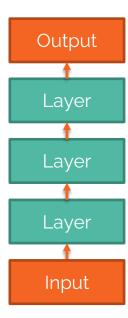
Beberapa Estimator yang disiapkan TF:

BaselineRegressor, BoostedTreesClassifier, BoostedTreesRegressor, DNNClassifier, DNNLinearCombinedClassifier, DNNLinearCombinedRegressor, DNNRegressor, LinearClassifier, LinearRegressor, etc

```
model = tf.estimator.LinearClassifier(feature_columns=feature_columns)
model.train(train_input_fn)
result = model.evaluate(eval_input_fn)
```



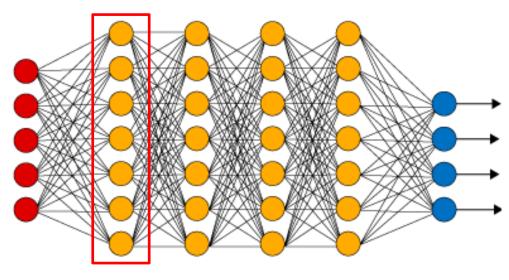
Model sekuensial menyediakan konstruksi model yang sederhana dengan cukup menumpuk layer dalam list



Mid Level API

Low Level API

2. Keras Sequential API

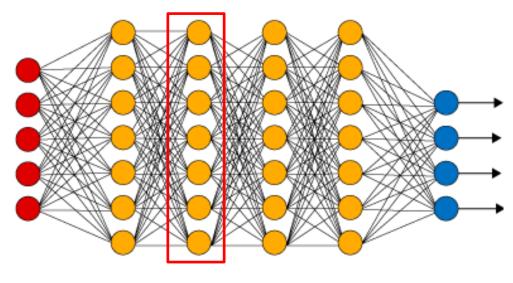




Mid Level API

Low Level API

2. Keras Sequential API



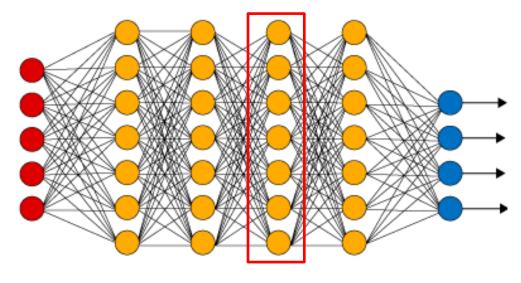
Mid Level API

Low Level API

2. Keras Sequential API

```
from tf.keras.layers import Dense as D

tf.keras.Sequential([
    D(7, activation='relu', input_shape=(5,))
    D(7, activation='relu'),
    D(7, activation='relu'),
    D(7, activation='relu'),
    D(4, activation='softmax')
])
```

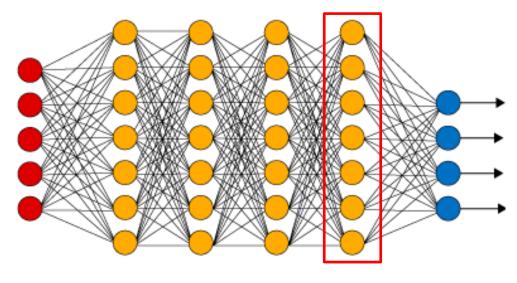


Mid Level API

Low Level API

2. Keras Sequential API

```
from tf.keras.layers import Dense as D
tf.keras.Sequential([
    D(7, activation='relu', input_shape=(5,))
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    D(7, activation='relu'),
    D(4, activation='softmax')
])
```

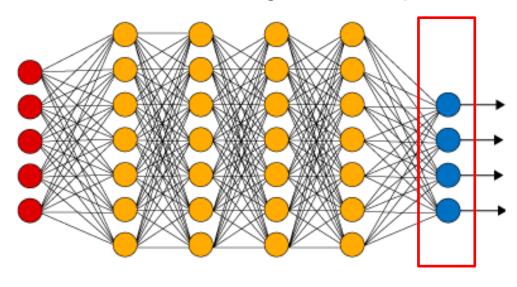


Mid Level API

Low Level API

2. Keras Sequential API

```
from tf.keras.layers import Dense as D
tf.keras.Sequential([
    D(7, activation='relu', input_shape=(5,))
    D(7, activation='relu'),
    D(7, activation='relu'),
    D(7, activation='relu'),
    D(4, activation='softmax')
])
```



High Level API Mid Level API berikut Low Level API Conv 1 Conv 2 Convolution Convolution (5 x 5) kernel (5 x 5) kernel Max-Pooling Max-Pooling valid padding valid padding (2×2) (2×2)

n1 channels

 $(24 \times 24 \times n1)$

INPUT

 $(28 \times 28 \times 1)$

2. Keras Sequential API

fc_3
Fully-Connected

Neural Network

ReLU activation

n2 channels

 $(4 \times 4 \times n2)$

n3 units

n2 channels

 $(8 \times 8 \times n2)$

n1 channels

 $(12 \times 12 \times n1)$

Ataupun Convolutional Neural Network (CNN) sederhana seperti berikut

fc 4

Fully-Connected

Neural Network

(with

dropout)

OUTPUT

High Level API Mid Level API berikut Low Level API **Fully-Connected** Neural Network Conv 1 Conv 2 ReLU activation Convolution Convolution (5 x 5) kernel (5 x 5) kernel Max-Pooling Max-Pooling valid padding valid padding (2×2) (2×2)

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Ataupun Convolutional Neural Network (CNN) sederhana seperti

fc 4

Fully-Connected

Neural Network

(with

dropout)

OUTPUT

```
from tf.keras import layers as L
tf.keras.Sequential([
  L.Conv2D(filters=32,
           kernel_size=(5,5),
           input shape=(28, 28, 1)),
  L.MaxPool2D((2,2)),
  L.Conv2D(filters=32,
           kernel size=(5,5)),
  L.MaxPool2D((2,2)),
  L.Flatten(),
  L.Dense(256, activation='relu'),
  L.Dense(10, activation='softmax')
```

High Level API Mid Level API berikut Low Level API **Fully-Connected** Neural Network Conv 1 Conv 2 ReLU activation Convolution Convolution (5 x 5) kernel (5 x 5) kernel Max-Pooling Max-Pooling valid padding valid padding (2×2) (2×2)

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INPUT

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2. Keras Sequential API

fc_3

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Ataupun Convolutional Neural Network (CNN) sederhana seperti

fc 4

Fully-Connected

Neural Network

(with

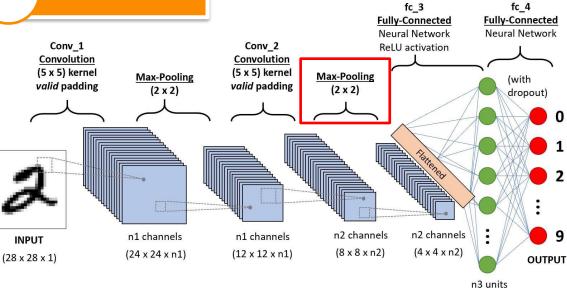
dropout)

OUTPUT

```
from tf.keras import layers as L
tf.keras.Sequential([
  L.Conv2D(filters=32,
           kernel size=(5,5),
           input shape=(28, 28, 1)),
  L.MaxPool2D((2,2)),
  L.Conv2D(filters=32,
           kernel size=(5,5)),
  L.MaxPool2D((2,2)),
  L.Flatten(),
  L.Dense(256, activation='relu'),
  L.Dense(10, activation='softmax')
```

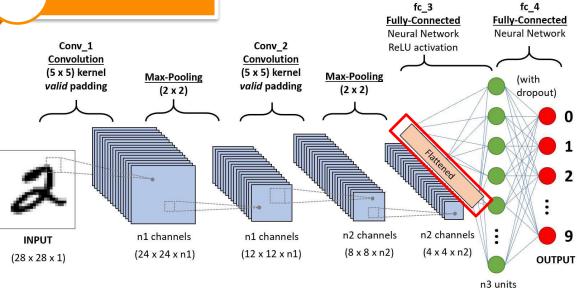
High Level API Mid Level API Low Level API Conv 1 Convolution (5 x 5) kernel Max-Pooling valid padding (2×2)

2. Keras Sequential API



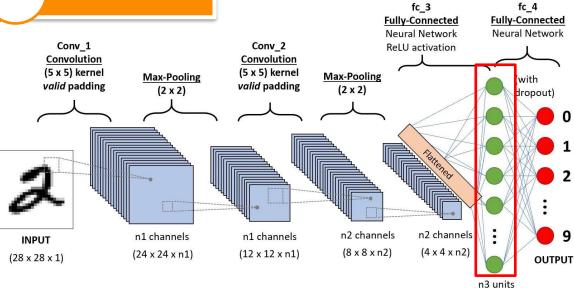
High Level API Mid Level API Low Level API Conv 1 Convolution (5 x 5) kernel Max-Pooling valid padding

2. Keras Sequential API



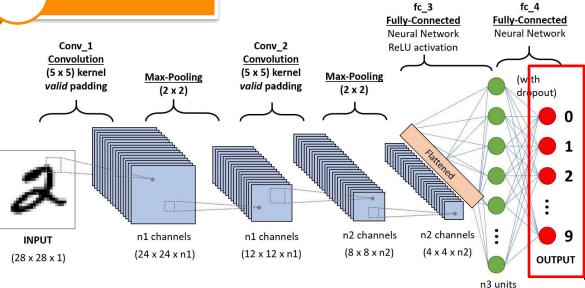
High Level API Mid Level API Low Level API Conv 1 Convolution (5 x 5) kernel Max-Pooling valid padding

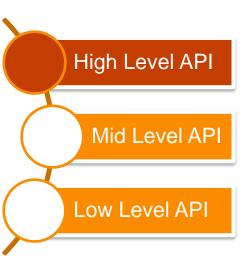
2. Keras Sequential API



High Level API Mid Level API Low Level API Conv 1 Convolution (5 x 5) kernel Max-Pooling

2. Keras Sequential API





Beberapa layer yang tersedia:

- Dense
- Batch Normalization
- Dropout

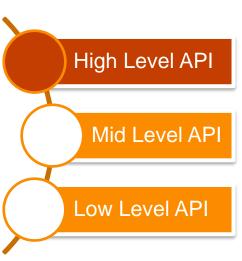
(Khusus CNN)

- Convolutional (1D/2D/3D, Transpose/Separable)
- Pooling (Max/Avg, Global, 1D/2D/3D)
- Cropping (1D/2D/3D)

(Khusus RNN)

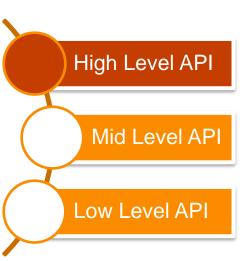
- Embedding
- SimpleRNN
- GRU
- LSTM
- Bidirectional

dll



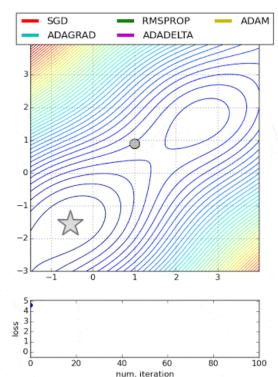
Proses kompilasi dan melatih model dapat dilakukan dengan sederhana

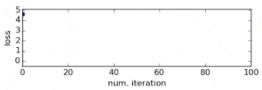
```
model.compile(optimizer='adam', loss='mae')
model.fit(x, y, batch_size=32, epochs=50)
```

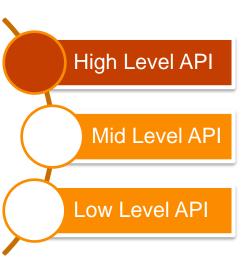


Beberapa optimizer yang tersedia:

- SGD (Stochastic Gradient Descent)
- Adam (Adaptive Moment Estimation)
- RMSProp (Root Mean Squared Propagation)
- AdaGrad (Adaptive Gradient Descent)
- AdaDelta (AdaGrad with Delta)





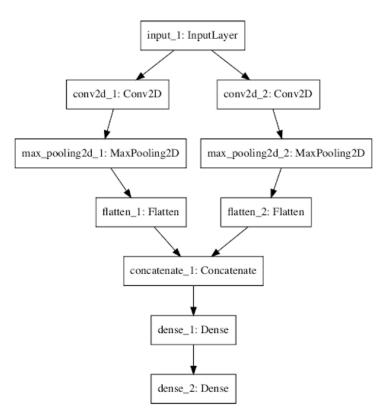


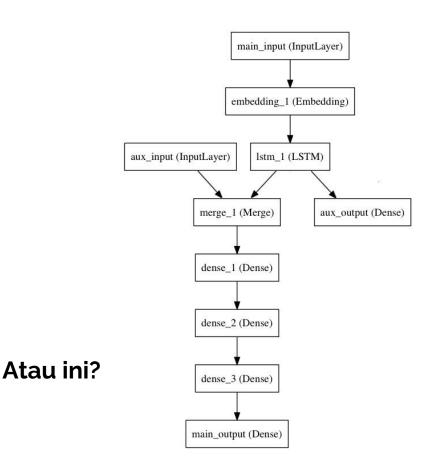
Beberapa fungsi loss yang tersedia:

- Binary & Categorical Crossentropy
- MSE
- RMSE
- MAE
- Huber
- Cosine Similarity

—

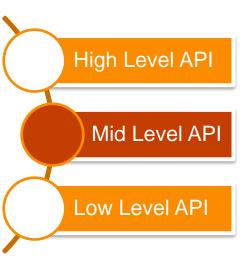
Bagaimana kalau kita punya model seperti ini





Model DNN "lurus" begitu saja tidak akan banyak bermanfaat.

Selain model sekuensial, TF memiliki dua cara lain untuk membangun model: *functional API* dan *model subclassing*



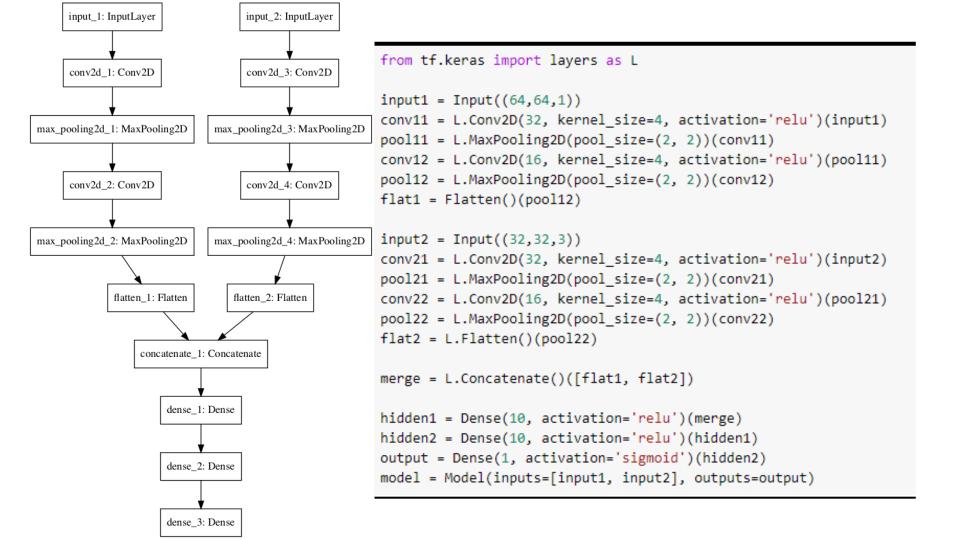
1. Keras Functional API

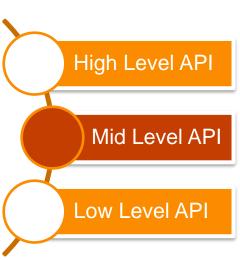
Membangun model dengan memerlakukan setiap layer seperti fungsi, yang secara fleksibel bisa ditentukan inputnya dari layer mana ke layer mana.

Model cukup didefinisikan dengan menetapkan input dan outputnya

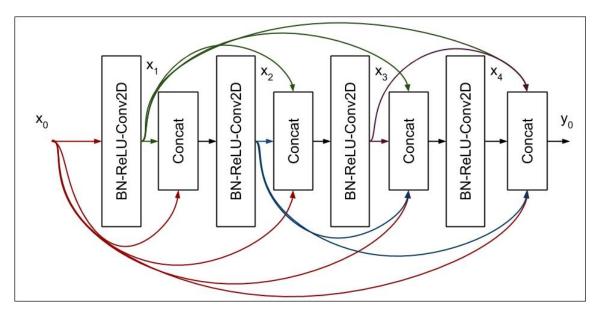
```
input = tf.keras.Input((10,))
dense1 = tf.keras.layers.Dense(32, activation='relu')(input)
dense2 = tf.keras.layers.Dense(32, activation='relu')(dense1)
output = tf.keras.layers.Dense(1)(dense2)

model = tf.keras.Model(inputs=input, output=output)
```

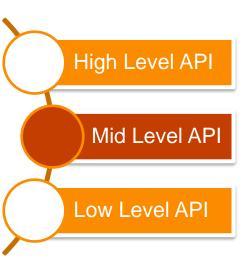




Bagaimana kalau modelnya seperti ini?



Tentu dimungkinkan dengan Functional API, tapi kodenya akan sangat kompleks dan panjang

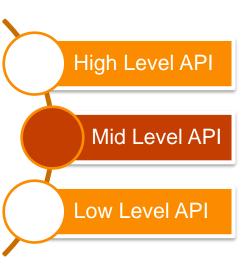


Membangun model dengan membuat kelas turunan Keras Model secara manual.

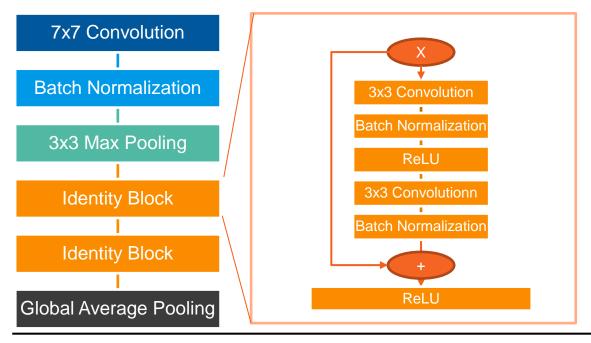
Bermanfaat terutama apabila terdapat banyak "blok" di dalam modelnya yang bisa dianggap sebagai submodel

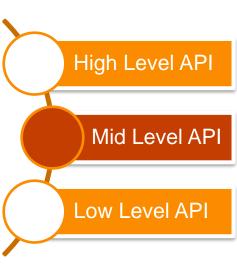
```
class CustomModel(tf.keras.Model):
    def __init__(self, *args, **kwargs):
        super(CustomModel, self).__init__(name='')
        # di sini kita inisialisasi semua layer yang dibutuhkan

def call(self, input):
    # di sini ditentukan apa yang akan dilakukan
    # bila model dipanggil dengan suatu input tensor
    output = None
    return output
```



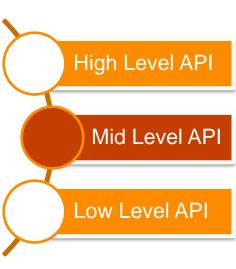
Misal kita ingin bangun model seperti ini





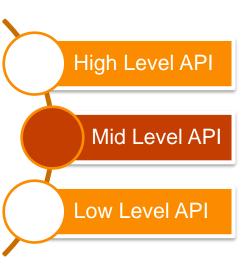
Mula-mula, definisikan Identity Block sebagai sebuah model sendiri

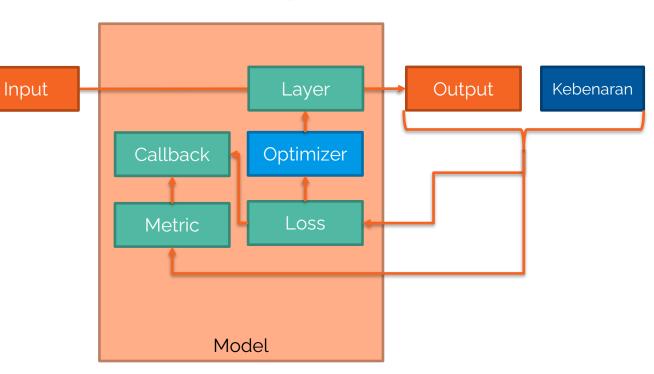
```
class IdentityBlock(tf.keras.Model):
   def init (self, filters, kernel size):
        super(IdentityBlock, self). init (name='')
        self.conv1 = tf.keras.layers.Conv2D(filters, kernel size, padding='same')
        self.bn1 = tf.keras.layers.BatchNormalization()
        self.conv2 = tf.keras.layers.Conv2D(filters, kernel size, padding='same')
        self.bn2 = tf.keras.layers.BatchNormalization()
        self.act = tf.keras.layers.Activation('relu')
        self.add = tf.keras.layers.Add()
                                                                        3x3 Convolution
   def call(self, input tensor):
       x = self.conv1(input tensor)
       x = self.bn1(x)
       x = self.act(x)
       x = self.conv2(x)
       x = self.bn2(x)
       x = self.add([x, input tensor])
       x = self.act(x)
       return x
                                                                            ReLU
```

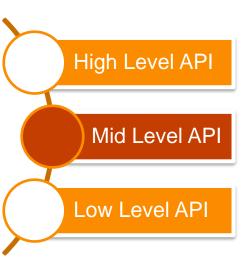


Kemudian gunakan blok tersebut di dalam model yang sesungguhnya

```
class MiniResNet(tf.keras.Model):
   def init (self, num classes):
       super(MiniResNet, self). init ()
       self.conv = tf.keras.layers.Conv2D(64, 7, padding='same')
                                                                                    7x7 Convolution
       self.bn = tf.keras.layers.BatchNormalization()
       self.act = tf.keras.layers.Activation('relu')
                                                                                 Batch Normalization
       self.max pool = tf.keras.layers.MaxPool2D((3, 3))
       self.id1a = IdentityBlock(64, 3)
       self.id1b = IdentityBlock(64, 3)
                                                                                   3x3 Max Pooling
       self.global pool = tf.keras.layers.GlobalAveragePooling2D()
       self.classifier = tf.keras.layers.Dense(num classes, activation='softmax')
                                                                                     Identity Block
   def call(self, inputs):
       x = self.conv(inputs)
       x = self.bn(x)
                                                                                     Identity Block
       x = self.act(x)
       x = self.max pool(x)
                                                                                    Global Average
       x = self.id1a(x)
                                                                                        Pooling
       x = self.id1b(x)
       x = self.global pool(x)
       return self.classifier(x)
```



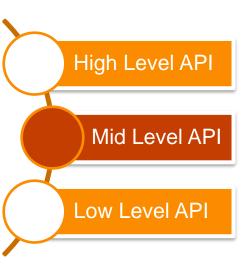




Semua komponen yang dibutuhkan ketika melatih model DNN:

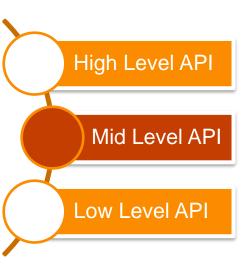
Layer, Loss, Metric, dan juga Callback

bisa dibuat manual sesuai kebutuhan. Khusus optimizer, override-nya harus dilakukan di low level API



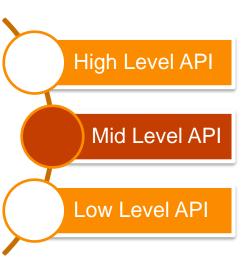
Format dasar custom layer:

```
class CustomLayer(tf.keras.layers.Layer):
    def __init__(self, *args, **kwargs):
        # inisiasi atribut
    def build(self, input_shape):
        # membangun parameter yang dipakai
    def call(self, inputs):
        # mengolah input jadi output dari layer ini
        return outputs
```



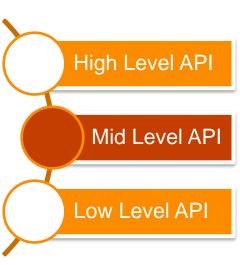
Format dasar custom loss:

```
class CustomLoss(tf.keras.losses.Loss):
    def __init__(self, *args, **kwargs):
        # inisiasi atribut
    def call(self, y_true, y_pred):
        # menghitung loss dari label prediksi
        # dan label yang benar
        return loss
```



Format dasar custom metric:

```
class CustomMetric(tf.keras.metrics.Metric):
    def __init__(self, *args, **kwargs):
        # inisiasi atribut
    def update_state(self, *args, **kwargs):
        # menghitung dan menambahkan nilai metrik
        # sesuai input yang diberikan
    def result(self):
        # mengembalikan akumulasi dari metrik yang
        # sudah dihitung
```



Format dasar custom callback:

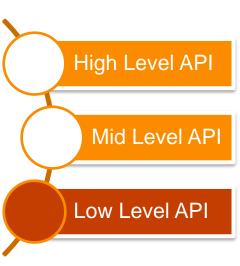
```
class CustomCallback(tf.keras.callbacks.Callback):
    def __init__(self, *args, **kwargs):
        # inisiasi atribut

def on_epoch_begin(self, epoch, logs):
        # callback ketika epoch dimulai

def on_epoch_end(self, epoch, logs):
        # callback ketika epoch berakhir

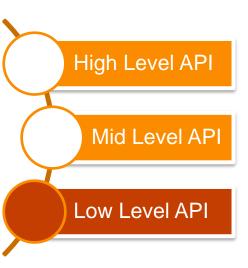
def on_batch_begin(self, batch, logs):
        # callback ketika batch dimulai

def on_batch_end(self, batch, logs):
    # callback ketika batch berakhir
```



1. Gradient Tape

Gradient Tape merekam semua operasi yang dilakukan dan secara otomatis menghitung turunannya. Sangat memudahkan proses backpropagation

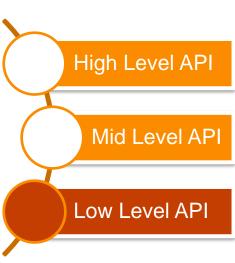


2. Dataset API

Tensorflow punya API khusus untuk mengelola data pipeline.

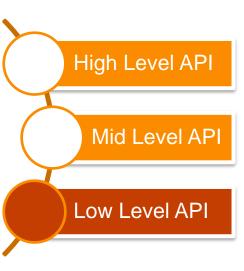
Misal, kita punya data time series, yang mau diolah menjadi input suatu model

```
def windowed_dataset(series, window_size, batch_size, shuffle_buffer):
    dataset = tf.data.Dataset.from_tensor_slices(series)
    dataset = dataset.window(window_size + 1, shift=1, drop_remainder=True)
    dataset = dataset.flat_map(lambda window: window.batch(window_size + 1))
    dataset = dataset.shuffle(shuffle_buffer).map(lambda window: (window[:-1], window[-1]))
    dataset = dataset.batch(batch_size).prefetch(1)
    return dataset
```



2. Dataset API

```
array = np.arange(0, 20)
    print(array)
    dataset = windowed_dataset(array, 4, 3, 1000)
    print("processed:")
    for x, y in dataset.take(1):
     for i in range(len(x)):
        print(x.numpy()[i], "->", y.numpy()[i])
      2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19]
processed:
[1 2 3 4] -> 5
[3 4 5 6] -> 7
[15 16 17 18] -> 19
```



Dan masih banyak lagi

Low level API berarti semuanya serba custom dan self-build, memanfaatkan seluruh fitur inti dari Tensorflow. Mau pakai yang mana? Apapun modelnya, frameworknya Tensorflow

Yuk, coba dulu



bit.ly/hands-on-tf

Any Question?

