

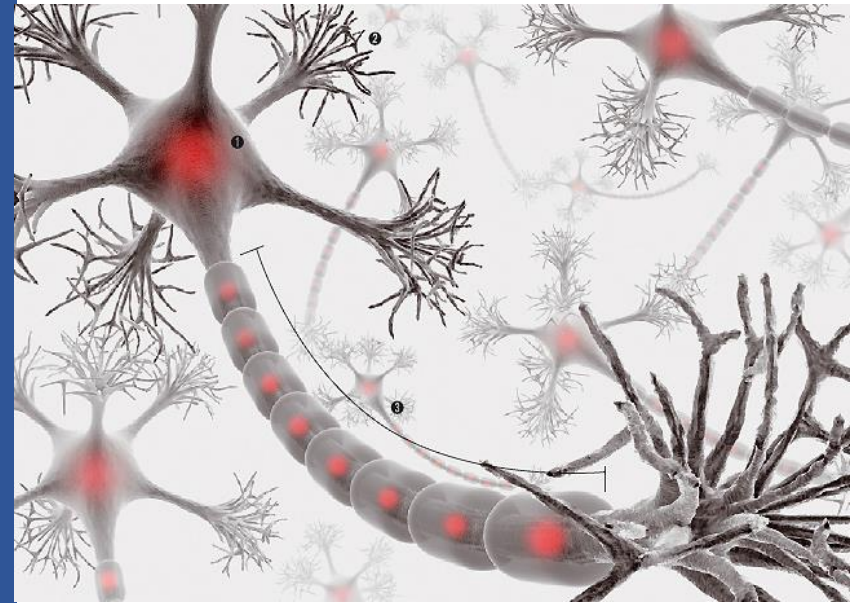
# Keras

## 학습 목표

- Tensorflow의 High-Level Wrapper인 Keras의 사용법을 이해한다.
- .

## 주요 내용

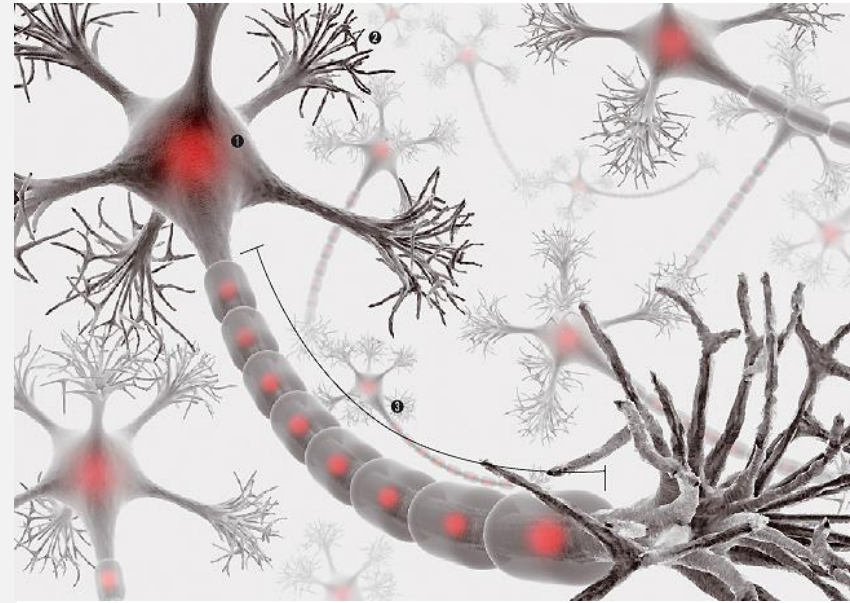
1. 클래스 구조
2. 모델 정의 방법
3. 모델 훈련 방법
4. 활용 클래스



# Keras

- **High-Level 신경망 스펙** (<https://keras.io>) (2015. 03)
- TensorFlow 1.2 : [tf.contrib.keras](#)로 추가됨
- TensorFlow 1.4 : [tf.keras](#)로 한단계 승격 (tf.layers → tf.keras)
- [TensorFlow 2.0 : 1st Class Python API](#)
- tf.layer, tf.contrib.layers(Slim)는 Deprecated 됨
- Keras 2.3.x 가 multi-backend Keras의 마지막 major release  
따라서, tf.keras를 사용하는 것이 좋음

# 1 클래스 계층 구조



# Class Hierarchy

## 변수 컨테이너 (tf.Variable)

`variables()`, `trainable_variables()`

## 계층 정의 (파라미터, Forward Pass)

`__call__()` → `build()` → `add_weights()`  
| → `call()`  
`add_loss()`

## 신경망 계층 통합

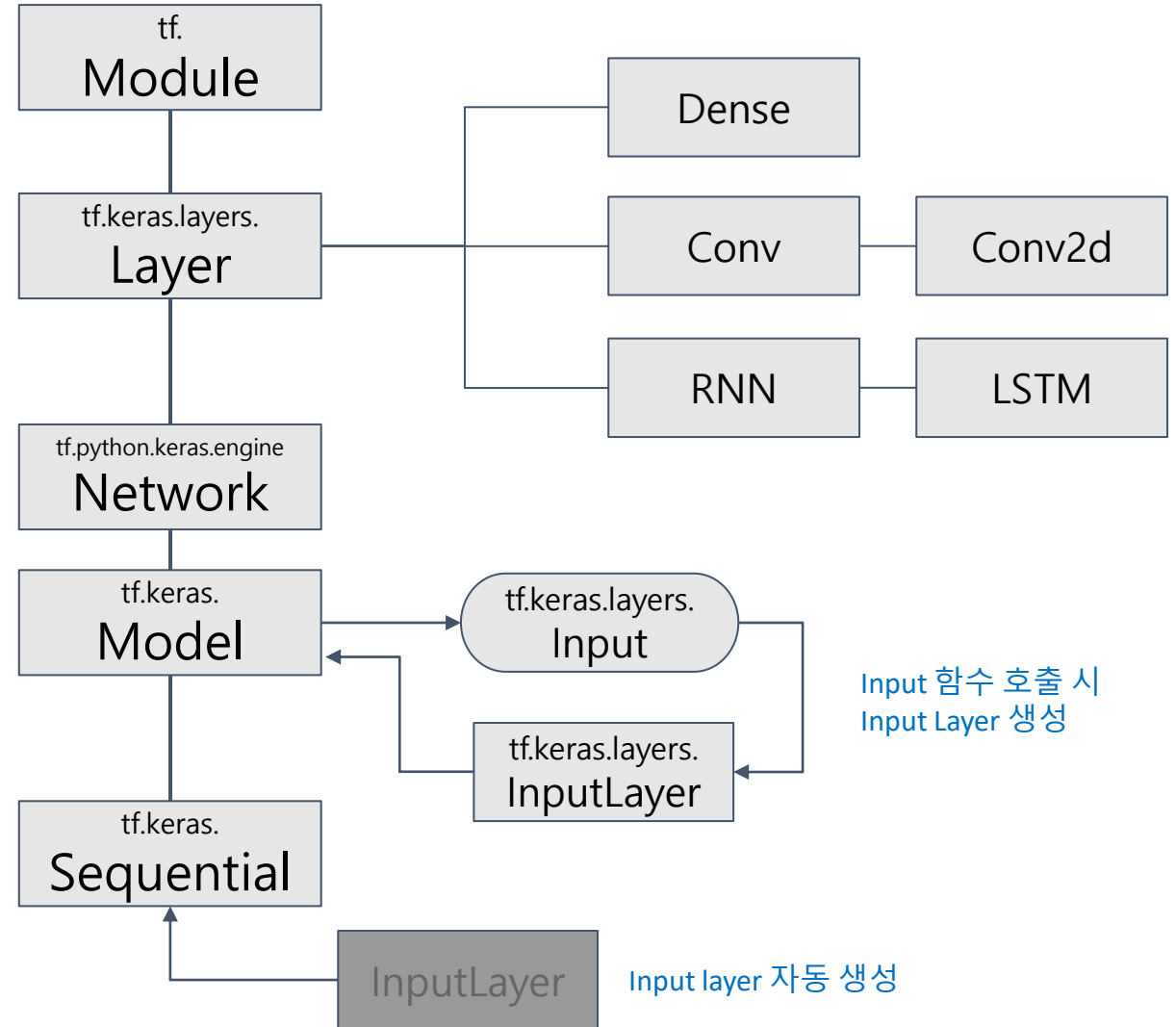
`layers()`, `summary()`, `save()`

## 모델 훈련/검증/테스트

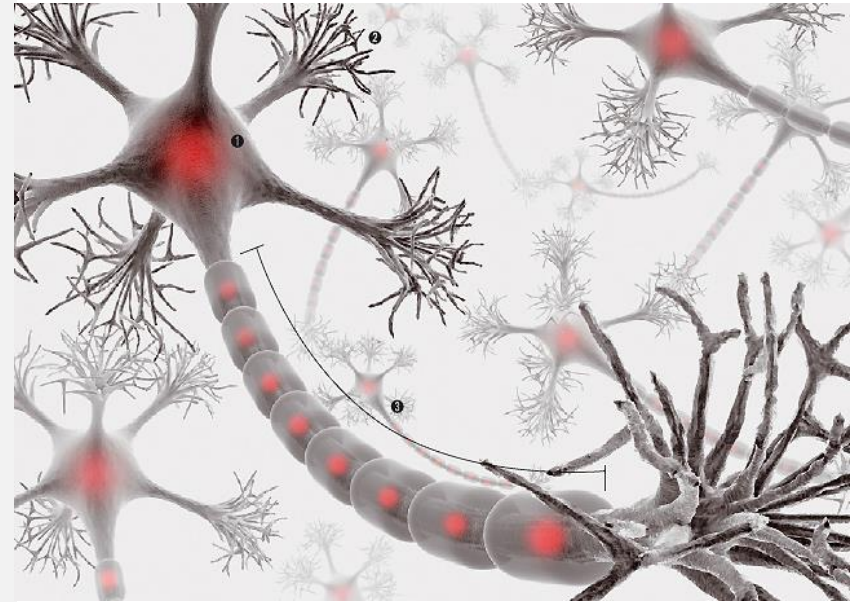
`compile()`, `fit()`, `evaluate()`, `predict()`

## 순차 모델 구성

`add()`



## 2 모델 정의 방법



# Keras 모델 정의

Simple Models



Sequential API

+

Built-in Layers

Functional API

+

Built-in Layers

Functional API

+

Custom layers

Custom metrics

Custom losses

Subclassing

Complex Model

# Sequential API

```
from tensorflow import tf
```

```
model = tf.keras.Sequential()  
model.add(tf.keras.layers.Dense(64, activation='relu'))  
model.add(tf.keras.layers.Dense(64, activation='relu'))  
model.add(tf.keras.layers.Dense(10, activation='softmax'))
```

# 훈련 설정

```
model.compile(optimizer=tf.keras.optimizers.Adam(0.001),  
              loss='categorical_crossentropy',  
              metrics=['accuracy'])
```

# 모델 훈련

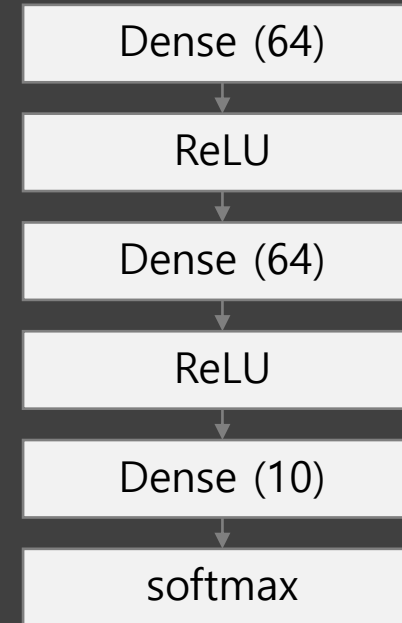
```
model.fit(train_data, labels, epochs=10, batch_size=32)
```

# 모델 평가

```
model.evaluate(test_data, labels)
```

# 샘플 예측

```
model.predict(new_sample)
```



```
model = tf.keras.Sequential([  
    tf.keras.layers.Dense(64),  
    tf.keras.layers.Dense(64),  
    tf.keras.layers.Dense(10),  
])
```

# Functional API

```
from tensorflow import tf

# 입력과 출력을 연결해서 임의의 모델 그래프 생성
input = tf.keras.Input(shape=(784,), name='img') # 입력 플레이스 홀더 반환
h1 = tf.keras.layers.Dense(64, activation='relu')(input) # 각 계층 별로 Tensor를 전달하고 리턴 받음
h2 = tf.keras.layers.Dense(64, activation='relu')(h1)
output = tf.keras.layers.Dense(10, activation='softmax')(h2)

# 모델 생성
model = tf.keras.Model(input, output) # 입력 Tensor와 Output Tensor를 모델에 지정

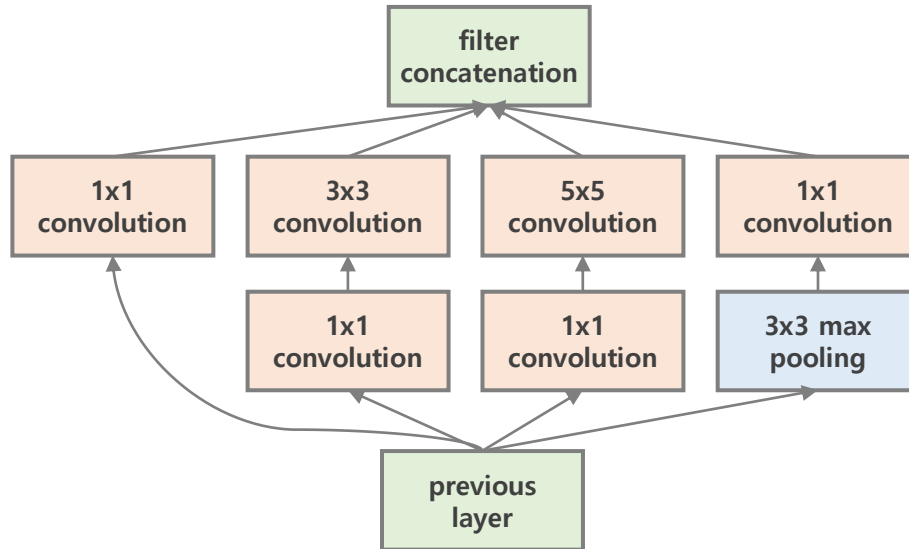
# 훈련 설정
model.compile(optimizer=tf.keras.optimizers.Adam(0.001),
              loss='categorical_crossentropy',
              metrics=['accuracy'])

...
```

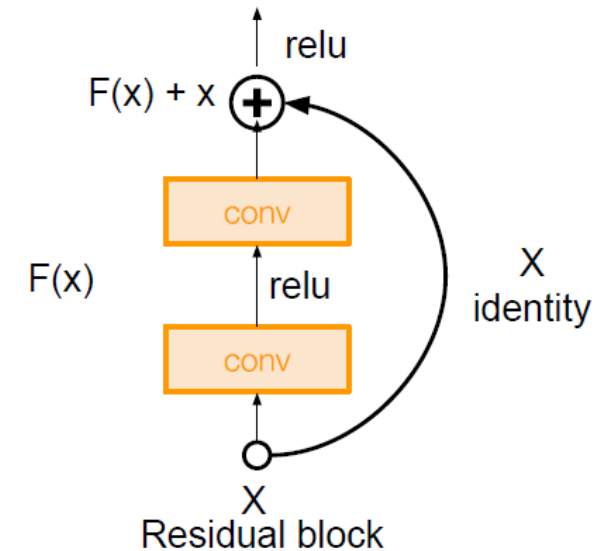


# Functional API

“Inception module”



“Residual block”



- 다중 입력 모델
- 다중 출력 모델
- 층을 공유하는 모델 (동일한 층을 여러 번 호출합니다)
- 데이터 흐름이 차례대로 진행되지 않는 모델 (예를 들면 잔차 연결(residual connections)).

# Custom Layer

```
from tensorflow import tf
class MyLayer(tf.keras.layers.Layer):

    def __init__(self, units, activation=None, **kwargs):
        self.units = units
        self.activation = keras.activations.get(activation)
        super().__init__(**kwargs)

    def build(self, input_shape):
        self.weight = self.add_weight(name='kernel',
                                       shape=(input_shape[1], self.units),
                                       initializer='uniform')
        self.bias = self.add_weight(name='bias',
                                    shape=(self.units,),
                                    initializer='zeros')
        super().build(input_shape)

    def call(self, X):
        z = tf.matmul(X, self.weight) + self.bias
        return self.activation(z)
```

# Custom Model

```
from tensorflow import tf

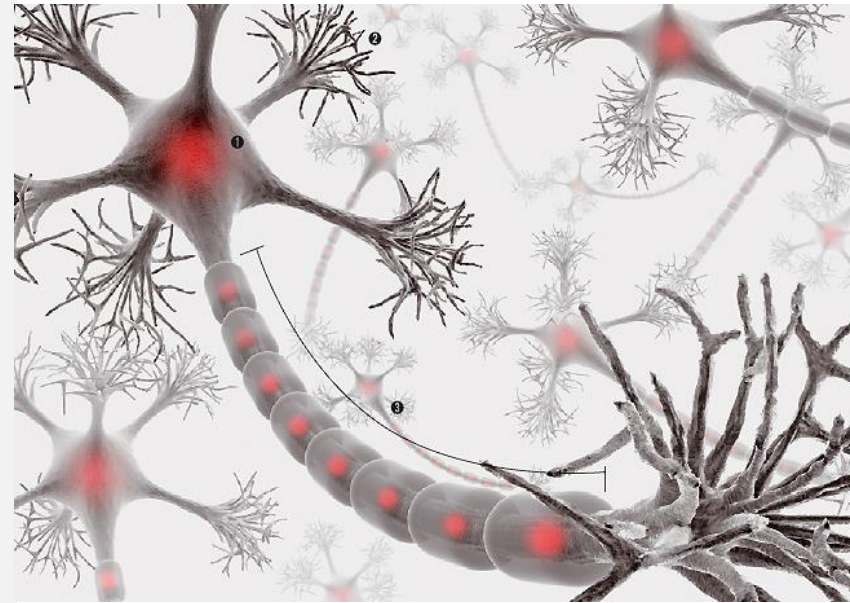
class MyModel(tf.keras.Model):

    def __init__(self, **kwargs):
        self.hidden = MyLayer(10, activation="relu")
        self.output = MyLayer(1)
        super().__init__(**kwargs)

    def call(self, input):
        h = self.hidden(input)
        return self.output(h)

model = MyModel()
```

# 3 모델 훈련 방법



# Training 방식

Quick Experiment



`model.fit()`

`model.fit()`

Iterate on the data

Custom Training Loop

callbacks

- Checkpoint
- Early stopping
- Tensorboard
- Slack notification

`train_on_batch()`

`test_on_batch()`

`predict_on_batch()`

- GAN
- Curriculum Learning

GradientTape

- New optimization algorithm
- Learn to learn (meta learning)

Advanced Training

# model.compile

훈련에 필요한 Optimizer, Loss, Metric을 설정하는 단계

## 회귀 모델 예시

1. 이름으로 지정 (Default 값으로 실행할 때)

```
model.compile(optimizer=tf.keras.optimizers.Adam(0.01),  
              loss='mse',      # 평균 제곱 오차  
              metrics=['mae']) # 평균 절댓값 오차
```

## 분류 모델 예시

2. 객체를 생성해서 전달 (파라미터를 지정할 필요가 있을 때)

```
model.compile(optimizer=tf.keras.optimizers.RMSprop(0.01),  
              loss=tf.keras.losses.CategoricalCrossentropy(),  
              metrics=[tf.keras.metrics.CategoricalAccuracy()])
```

# model.fit

모델을 고정된 epoch 수로 훈련

```
history = model.fit( train_data, train_labels,  
                    epochs=1000, validation_split = 0.2, verbose=0,  
                    callbacks=[Earlystopping(),  
                               Tensorboard(),  
                               ModelCheckpoint()])
```

- **batch\_size**: 배치 크기 (default 32)
- **epochs**: 총 epoch 수 (epoch는 training set을 한번 실행하는 단위)
- **validation\_split**: training set에서 validation set으로 사용할 비율 ( (0,1) 사이의 값)
- **verbose**: 훈련 진행 상황 모드 0 = silent, 1 = progress bar, 2 = one line per epoch
- **callbacks**: 훈련하면서 실행할 콜백 리스트

# tf.GradientTape

```
@tf.function
```

```
def train_step(input, target):
```

```
    with tf.GradientTape() as tape:
```

```
        # forward Pass
```

```
        predictions = model(input)
```

```
        # compute the loss
```

```
        loss = tf.reduce_mean(
            tf.keras.losses.sparse_categorical_crossentropy(
                target, predictions, from_logits=True))
```

```
        # compute gradients
```

```
        grads = tape.gradient(loss, model.trainable_variables)
```

```
        # perform a gradient descent step
```

```
        optimizer.apply_gradients(zip(grads, model.trainable_variables))
```

```
    return loss
```

Forward Pass

Gradient 계산

Parameter Update



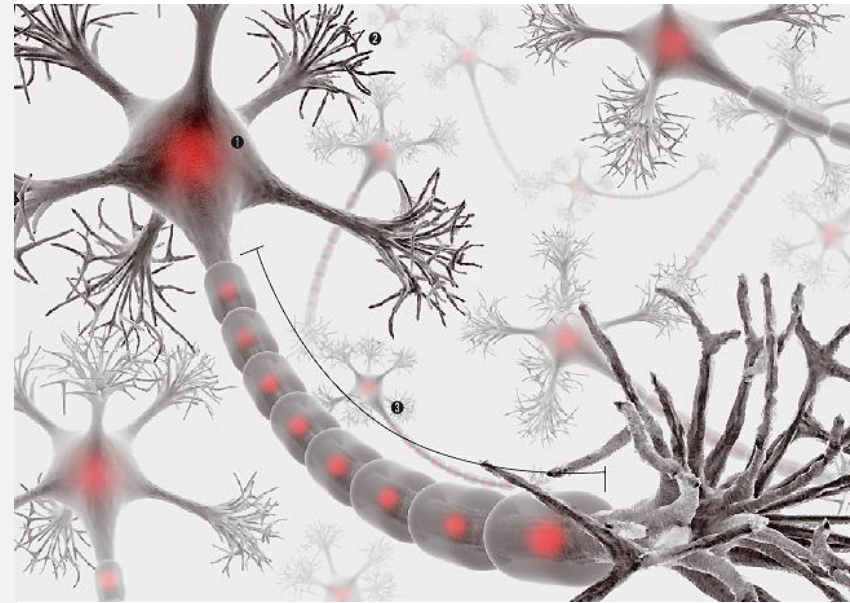
# Keras + eager mode

- Keras는 default로 `@tf.function` 사용

```
class CustomModel(tf.keras.models.Model):  
  
    @tf.function  
    def call(self, input_data):  
        if tf.reduce_mean(input_data) > 0:  
            return input_data  
        else:  
            return input_data // 2
```

- Eager mode를 사용하려면 다음 설정을 해야 함
  - `model = Model(dynamic=True)` or
  - `model.compile(..., run_eagerly=True)`

# 4 활용 클래스



## 참고 tf.keras.layers

- **class Conv2D**: 2D convolution layer (e.g. spatial convolution over images).
- **class Dense**: Just your regular densely-connected NN layer.
- **class Flatten**: Flattens the input. Does not affect the batch size.
- **class Reshape**: Reshapes an output to a certain shape.
- **class InputLayer**: Layer to be used as an entry point into a Network (a graph of layers).
- **class MaxPool2D**: Max pooling operation for spatial data.
- **class AveragePooling2D**: Average pooling operation for spatial data.
- **class GlobalAveragePooling2D**: Global average pooling operation for spatial data.
- **class BatchNormalization**: Normalize and scale inputs or activations. (Ioffe and Szegedy, 2014).
- **class Dropout**: Applies Dropout to the input.
- **class Embedding**: Turns positive integers (indexes) into dense vectors of fixed size.
- **class SimpleRNN**: Fully-connected RNN where the output is to be fed back to input.
- **class LSTM**: Long Short-Term Memory layer - Hochreiter 1997.
- **class GRU**: Gated Recurrent Unit - Cho et al. 2014.

Layer

모양 변경

Pooling

정규화

RNN 계열

[https://www.tensorflow.org/api\\_docs/python/tf/keras/layers](https://www.tensorflow.org/api_docs/python/tf/keras/layers)

## 참고 `tf.keras.layers`

- `class Softmax`: Softmax activation function.
- `class ReLU`: Rectified Linear Unit activation function.
- `class LeakyReLU`: Leaky version of a Rectified Linear Unit.
- `class ELU`: Exponential Linear Unit.

### Activation Function

[https://www.tensorflow.org/api\\_docs/python/tf/keras/layers](https://www.tensorflow.org/api_docs/python/tf/keras/layers)

## 참고 `tf.keras.optimizers`

- `class SGD`: Stochastic gradient descent and momentum optimizer.
- `class Adagrad`: Optimizer that implements the Adagrad algorithm.
- `class RMSprop`: Optimizer that implements the RMSprop algorithm.
- `class Adam`: Optimizer that implements the Adam algorithm.

[https://www.tensorflow.org/api\\_docs/python/tf/keras/optimizers](https://www.tensorflow.org/api_docs/python/tf/keras/optimizers)

## 참고 `tf.keras.losses`

- **class `MeanSquaredError`**: Computes the mean of squares of errors between labels and predictions.
- **class `MeanAbsoluteError`**: Computes the mean of absolute difference between labels and predictions.
- **class `BinaryCrossentropy`**: Computes the cross-entropy loss between true labels and predicted labels.
- **class `CategoricalCrossentropy`**: Computes the crossentropy loss between the labels and predictions.
- **class `SparseCategoricalCrossentropy`**: Computes the crossentropy loss between the labels and predictions.

[https://www.tensorflow.org/api\\_docs/python/tf/keras/losses](https://www.tensorflow.org/api_docs/python/tf/keras/losses)

## 참고 `tf.keras.metrics`

- **class Accuracy**: Calculates how often predictions matches labels.
- **class MeanAbsoluteError**: Computes the mean absolute error between the labels and predictions.
- **class MeanSquaredError**: Computes the mean squared error between `y_true` and `y_pred`.

[https://www.tensorflow.org/api\\_docs/python/tf/keras/metrics](https://www.tensorflow.org/api_docs/python/tf/keras/metrics)

## 참고 `tf.keras.callbacks`

- **class `EarlyStopping`**: Stop training when a monitored quantity has stopped improving.
- **class `ModelCheckpoint`**: Save the model after every epoch.
- **class `TensorBoard`**: Enable visualizations for TensorBoard.
- **class `LearningRateScheduler`**: Learning rate scheduler.

[https://www.tensorflow.org/api\\_docs/python/tf/keras/callbacks](https://www.tensorflow.org/api_docs/python/tf/keras/callbacks)



**Thank you!**

