

# Keras - Model Compilation, Evaluation, Prediction

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# Compilation

- The compilation is the final step in creating a model.
- Once the compilation is done, we can move on to training phase.

# Loss

- In machine learning, Loss function is used to find error or deviation in the learning process. Keras requires loss function during model compilation process.
- Keras provides quite a few loss function in the losses module and they are as follows –
  - mean\_squared\_error
  - mean\_absolute\_error
  - mean\_absolute\_percentage\_error
  - mean\_squared\_logarithmic\_error
  - squared\_hinge
  - hinge
  - categorical\_hinge

# Loss

- logcosh
- huber\_loss
- categorical\_crossentropy
- sparse\_categorical\_crossentropy
- binary\_crossentropy
- kullback\_leibler\_divergence
- poisson
- cosine\_proximity
- is\_categorical\_crossentropy

# Loss

- All above loss function accepts two arguments –
  - `y_true` – true labels as tensors
  - `y_pred` – prediction with same shape as `y_true`

# Optimizers

- In machine learning, Optimization is an important process which optimizes the input weights by comparing the prediction and the loss function.
- Keras provides quite a few optimizers as a module, optimizers and they are as follows:
- SGD – Stochastic gradient descent optimizer.  
`keras.optimizers.SGD(learning_rate = 0.01, momentum = 0.0, nesterov = False)`

# Optimizers

- RMSprop – RMSProp optimizer.
  - `keras.optimizers.RMSprop(learning_rate = 0.001, rho = 0.9)`
- Adagrad – Adagrad optimizer.
  - `keras.optimizers.Adagrad(learning_rate = 0.01)`
- Adadelta – Adadelta optimizer.
  - `keras.optimizers.Adadelta(learning_rate = 1.0, rho = 0.95)`

# Optimizers

- Adam – Adam optimizer.
  - `keras.optimizers.Adam(learning_rate = 0.001, beta_1 = 0.9, beta_2 = 0.999, amsgrad = False)`
- Adamax – Adamax optimizer from Adam.
  - `keras.optimizers.Adamax(learning_rate = 0.002, beta_1 = 0.9, beta_2 = 0.999)`
- Nadam – Nesterov Adam optimizer.
  - `keras.optimizers.Nadam(learning_rate = 0.002, beta_1 = 0.9, beta_2 = 0.999)`



# Metrics

- In machine learning, Metrics is used to evaluate the performance of your model. It is similar to loss function, but not used in training process. Keras provides quite a few metrics as a module, metrics and they are as follows

accuracy

binary\_accuracy

categorical\_accuracy

sparse\_categorical\_accuracy

top\_k\_categorical\_accuracy

sparse\_top\_k\_categorical\_accuracy

cosine\_proximity

clone\_metric

# Metrics

- Similar to loss function, metrics also accepts below two arguments –
  - y\_true – true labels as tensors
  - y\_pred – prediction with same shape as y\_true

# Compile the Model

- Keras model provides a method, `compile()` to compile the model. The argument and default value of the `compile()` method is as follows

```
compile(  
    optimizer,  
    loss = None,  
    metrics = None,  
    loss_weights = None,  
    sample_weight_mode = None,  
    weighted_metrics = None,  
    target_tensors = None  
)
```

# Compile the Model

- The important arguments are as follows –
  - loss function
  - Optimizer
  - metrics
- A sample code to compile the mode is as follows –

```
from keras import losses
from keras import optimizers
from keras import metrics
model.compile(loss = 'mean_squared_error',
              optimizer = 'sgd', metrics = [metrics.categorical_accuracy])
```

# Model Training

- Models are trained by NumPy arrays using fit(). The main purpose of this fit function is used to evaluate your model on training.
- This can be also used for graphing model performance. It has the following syntax –  
`model.fit(X, y, epochs = , batch_size = )`
- Here,
  - X, y – It is a tuple to evaluate your data.
  - epochs – no of times the model is needed to be evaluated during training.
  - batch\_size – training instances.

# Model Evaluation

- Evaluation is a process during development of the model to check whether the model is best fit for the given problem and corresponding data.
- Keras model provides a function, evaluate which does the evaluation of the model.
- It has three main arguments,
  - Test data
  - Test data label
  - verbose - true or false

# Model Evaluation

- The evaluate function is used to evaluate the test data.

```
score = model.evaluate(x_test, y_test, verbose = 0)
```

```
print('Test loss:', score[0])
```

```
print('Test accuracy:', score[1])
```

# Model Prediction

- Prediction is the final step and our expected outcome of the model generation. Keras provides a method, predict to get the prediction of the trained model. The signature of the predict method is as follows,

```
predict(  
    x,  
    batch_size = None,  verbose = 0,  
    steps = None,  callbacks = None,  
    max_queue_size = 10,  workers = 1,  
    use_multiprocessing = False  
)
```



# Example:

- Classification Example

# Thank you

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## Web Resources

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