

# Sequential models with Keras

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- ▶ Will be introduced later on

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- ▶ Can get weights as list of numpy arrays:  
`layer.get_weights()`
- ▶ Set layer weights with `layer.set_weights(weights)`
- ▶ Each layer has a defining configuration, `layer.get_config()`

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- ▶ Evaluate model, persist or deploy model, start new experiment etc.

# Compiling models I: loss functions

```
# Option 1: Importing from loss module (preferred)  
from keras.losses import mean_squared_error  
model.compile(loss=mean_squared_error, optimizer=...)  
  
# Option 2: Using strings  
model.compile(loss='mean_squared_error', optimizer=...)
```

# Compiling models I: optimizers

*# Option 1: Load optimizers from module (preferred)*

```
from keras.optimizers import SGD
```

```
sgd = SGD(lr=0.01,          # learning rate >= 0
          decay=1e-6,       # lr decay after updates
          momentum=0.9)     # Momentum parameter used for SGD
model.compile(loss=..., optimizer=sgd)
```

*# Option 2: pass string (default parameters will be used)*

```
model.compile(loss=..., optimizer='sgd')
```

# Fit, evaluate and predict

```
# Fit a model to train data
```

```
model.fit(x_train, y_train,  
          batch_size=32,  
          epochs=10,  
          validation_data=(x_val, y_val))
```

```
# Evaluate on test data
```

```
evaluate(x_test, y_test, batch_size=32)
```

```
# Predict labels
```

```
predict(x_test, batch_size=32)
```