

Tensorflow (Draft)

Deep Learning Framework

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Ph.D. in Computer Science

Outline

- **Introduction to Tensorflow**
- **Tensorflow/Keras**
- **Model Construction – Linear Regression**
- **Model Construction – Logistic Regression**
- **Model Construction – Softmax Regression**
- **Model Training**
- **Model Saving and Loading**

Introduction

<https://en.wikipedia.org/wiki/TensorFlow>



TensorFlow

Developer(s) Google Brain Team^[1]

Initial release November 9, 2015; 4 years ago

Stable release 2.2.0^[2] / May 6, 2020; 1 month ago

Repository github.com/tensorflow/tensorflow

Written in Python, C++, CUDA

Platform Linux, macOS, Windows, Android, JavaScript^[3]

Type Machine learning library

License Apache License 2.0

Website www.tensorflow.org

Installation

```
pip install tensorflow
```

Package Declaration

```
import tensorflow as tf
```

Example 1

```
1 import tensorflow as tf
2
3 print("TensorFlow version: ", tf.__version__)
4 print("Keras version: ", tf.keras.__version__)
```

```
TensorFlow version: 2.0.0
Keras version: 2.2.4-tf
```

Example 2

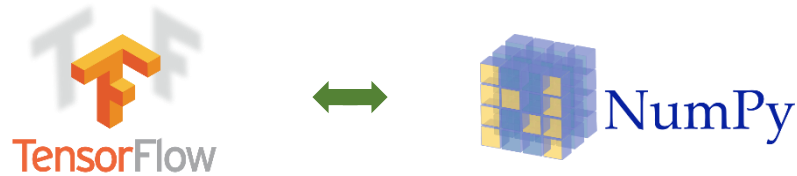
```
1 import tensorflow as tf
2
3 if tf.test.is_gpu_available():
4     print('Running on GPU')
5 else:
6     print('Running on CPU')
```

Running on CPU

Introduction

❖ Tensor

- ❖ ~ ndarray in Numpy
- ❖ Run on both CPU and GPU
- ❖ All tensors are immutable



```
1 import tensorflow as tf
2 import numpy as np
3
4 # create a ndarray
5 an_array = np.array([1,3,5,7,11])
6 print(type(an_array), an_array)
7
8 # convert from ndarray to tensor
9 a_tensor = tf.convert_to_tensor(an_array)
10 print(a_tensor)
11
12 # convert from tensor to ndarray
13 array_2 = a_tensor.numpy()
14 print(type(array_2), array_2)
```

```
<class 'numpy.ndarray'> [ 1  3  5  7 11]
tf.Tensor([ 1  3  5  7 11], shape=(5,), dtype=int32)
<class 'numpy.ndarray'> [ 1  3  5  7 11]
```

Important attributes

```
1 import tensorflow as tf
2 import numpy as np
3
4 # create a 2x3 array
5 an_array = np.array([[1,2],
6                      [3,4],
7                      [5,6]])
8
9 # convert to tensor
10 a_tensor = tf.convert_to_tensor(an_array)
11
12 # no. of elements on each axis
13 print(a_tensor.shape)
14 # data type
15 print(a_tensor.dtype)
16
17 # no. of axes
18 print(a_tensor.ndim)
```

```
(3, 2)
<dtype: 'int32'>
2
```

Introduction

❖ Tensor

❖ Broadcasting

```
1 import tensorflow as tf
2 import numpy as np
3
4 # create two tensors
5 tensor_1 = tf.convert_to_tensor([[1,2],
6                                   [3,4]])
7 tensor_2 = tf.convert_to_tensor([1])
8
9 # add two tensors
10 tensor_3 = tensor_1 + tensor_2
11
12 print('tensor_1: \n', tensor_1)
13 print('tensor_2: \n', tensor_2)
14 print('tensor_3: \n', tensor_3)
```

```
tensor_1:
  tf.Tensor(
[[1 2]
 [3 4]], shape=(2, 2), dtype=int32)
tensor_2:
  tf.Tensor([1], shape=(1,), dtype=int32)
tensor_3:
  tf.Tensor(
[[2 3]
 [4 5]], shape=(2, 2), dtype=int32)
```

```
1 import tensorflow as tf
2 import numpy as np
3
4 # create 2x2 tensor
5 tensor_1 = tf.convert_to_tensor([[1,2],
6                                   [3,4]])
7
8 # add a number to a tensor
9 tensor_2 = tensor_1 + 2
10
11 print('tensor_1: \n', tensor_1)
12 print('tensor_2: \n', tensor_2)
```

```
tensor_1:
  tf.Tensor(
[[1 2]
 [3 4]], shape=(2, 2), dtype=int32)
tensor_2:
  tf.Tensor(
[[3 4]
 [5 6]], shape=(2, 2), dtype=int32)
```

Introduction

❖ Tensor

❖ Important functions

Squared Difference

$$sd = (x - y)^2$$

```
1 import tensorflow as tf
2 import numpy as np
3
4 # create an ndarray
5 x = np.array([1,2, 3, 4])
6 y = 5
7
8 # compute squared difference
9 sd = tf.math.squared_difference(x,y)
10 print(sd)
```

tf.Tensor([16 9 4 1], shape=(4,), dtype=int32)

```
1 import tensorflow as tf
2
3 # create a list
4 x = [1,2, 3, 4]
5 y = 5
6
7 # compute squared difference
8 sd = tf.math.squared_difference(x,y)
9 print(sd)
```

tf.Tensor([16 9 4 1], shape=(4,), dtype=int32)

```
1 import tensorflow as tf
2
3 # create a tensor
4 x = tf.convert_to_tensor([1,2, 3, 4])
5 y = 5
6
7 # compute squared difference
8 sd = tf.math.squared_difference(x,y)
9 print(sd)
```

tf.Tensor([16 9 4 1], shape=(4,), dtype=int32)

Introduction

❖ Tensor

❖ Important functions

random.normal()

```
1 import tensorflow as tf
2
3 rand = tf.random.normal(shape = (3,2), mean=0, stddev=1)
4 print(rand)
```

```
tf.Tensor(
[[ 0.47103247 -0.12765862]
 [-0.26556632 -0.05912822]
 [ 1.0851953  -0.55289406]], shape=(3, 2), dtype=float32)
```

random.uniform()

```
1 import tensorflow as tf
2
3 rand = tf.random.uniform(shape=(3,2), minval=0,
4                           maxval=9, dtype=tf.int32)
5 print(rand)
```

```
tf.Tensor(
[[4 0]
 [1 8]
 [0 2]], shape=(3, 2), dtype=int32)
```

concat()

```
1 import tensorflow as tf
2
3 tensor_1 = tf.random.normal(shape=(2,2), mean=0, stddev=1)
4 tensor_2 = tf.random.normal(shape=(2,2), mean=0, stddev=1)
5
6 # concat two tensors along axis_0
7 tensor_3 = tf.concat([tensor_1, tensor_1], axis=0)
8
9 # concat two tensors along axis_1
10 tensor_4 = tf.concat([tensor_1, tensor_1], axis=1)
11
12 print(tensor_1.shape)
13 print(tensor_2.shape)
14 print(tensor_3.shape)
15 print(tensor_4.shape)
```

```
(2, 2)
(2, 2)
(4, 2)
(2, 4)
```

Introduction

❖ Images in files

❖ Important functions

argmax()

```
1 import tensorflow as tf
2
3 # create a tensor
4 tensor = tf.random.uniform(shape=(3,6), minval=0,
5                             maxval=20, dtype=tf.int32)
6
7 # find the index of the max value
8 max_position_1 = tf.argmax(tensor, axis=0)
9 max_position_2 = tf.argmax(tensor, axis=1)
10
11 print(tensor)
12 print('max_position_1: ', max_position_1)
13 print('max_position_2: ', max_position_2)
```

tf.Tensor(
[[8 10 13 1 12 1]
 [8 19 5 3 16 16]
 [2 16 7 1 4 17]], shape=(3, 6), dtype=int32)
max_position_1: tf.Tensor([0 1 0 1 1 2], shape=(6,), dtype=int64)
max_position_2: tf.Tensor([2 1 5], shape=(3,), dtype=int64)

argmin()

```
1 import tensorflow as tf
2
3 # create a tensor
4 tensor = tf.random.uniform(shape=(3,6), minval=0,
5                             maxval=20, dtype=tf.int32)
6
7 # find the index of the min value
8 min_position_1 = tf.argmin(tensor, axis=0)
9 min_position_2 = tf.argmin(tensor, axis=1)
10
11 print(tensor)
12 print('min_position_1: ', min_position_1)
13 print('min_position_2: ', min_position_2)
```

tf.Tensor(
[[5 0 13 11 4 10]
 [19 6 7 5 17 6]
 [18 9 9 2 1 11]], shape=(3, 6), dtype=int32)
min_position_1: tf.Tensor([0 0 1 2 2 1], shape=(6,), dtype=int64)
min_position_2: tf.Tensor([1 3 4], shape=(3,), dtype=int64)

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- **Model Saving and Loading**

Keras

- ❖ Run on top of TensorFlow
- ❖ Integrated into Tensorflow

Package Declaration

```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 print(keras.__version__)
```

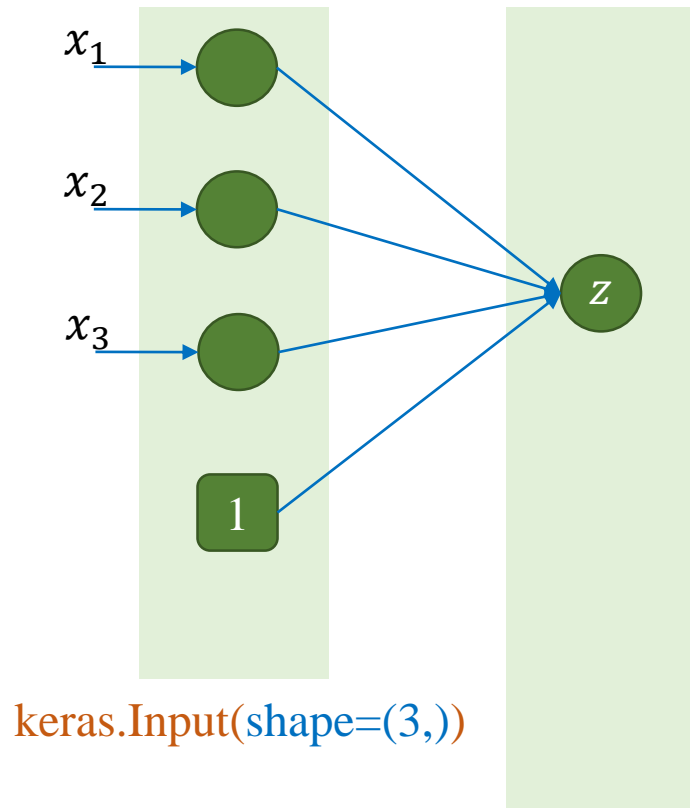
2.2.4-tf



Original author(s)	François Chollet [fr]
Developer(s)	various
Initial release	27 March 2015; 5 years ago
Stable release	2.3.1 ^[1] / 7 October 2019; 8 months ago
Repository	github.com/keras-team/keras 
Written in	Python
Platform	Cross-platform
Type	Neural networks
License	MIT
Website	keras.io 

Keras

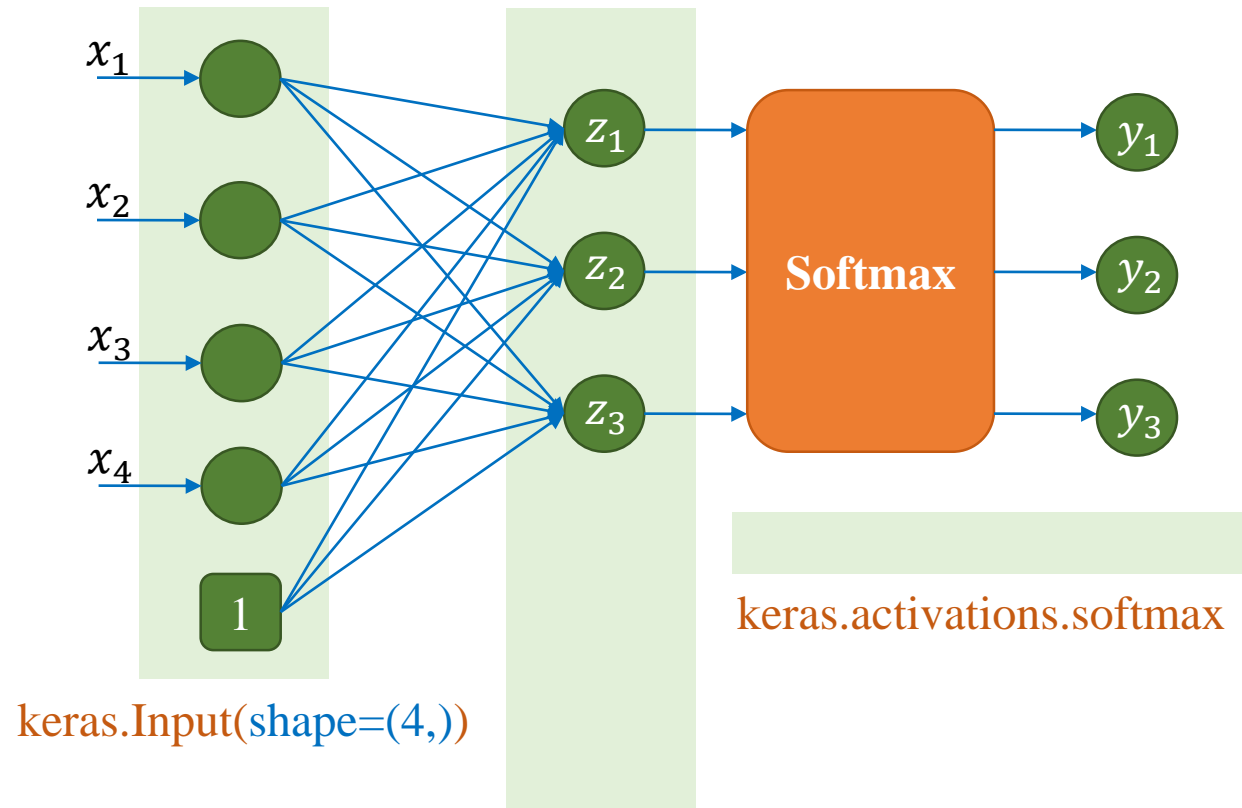
Model



`keras.Input(shape=(3,))`

`keras.layers.Dense(units=1)`

Model



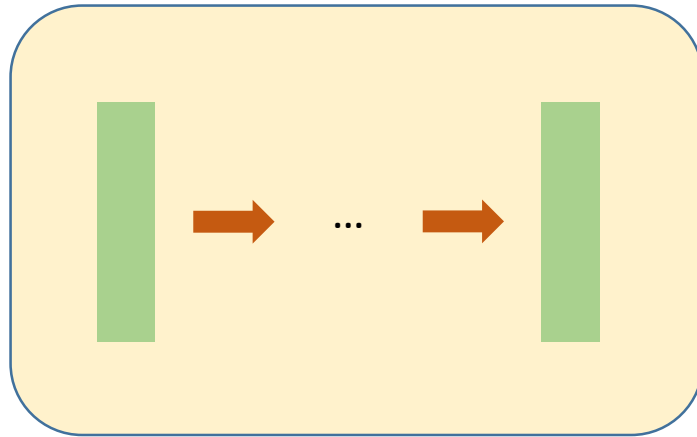
`keras.Input(shape=(4,))`

`keras.layers.Dense(units=3)`

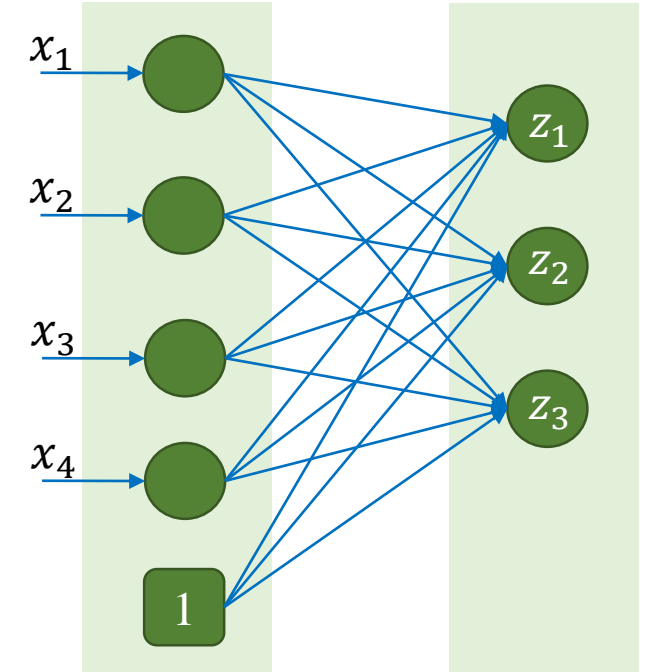
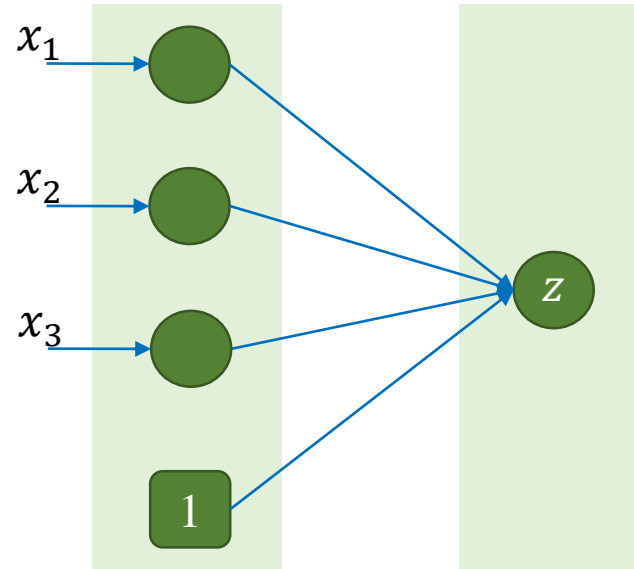
`keras.activations.softmax`

Keras

Model



`keras.Sequential()`

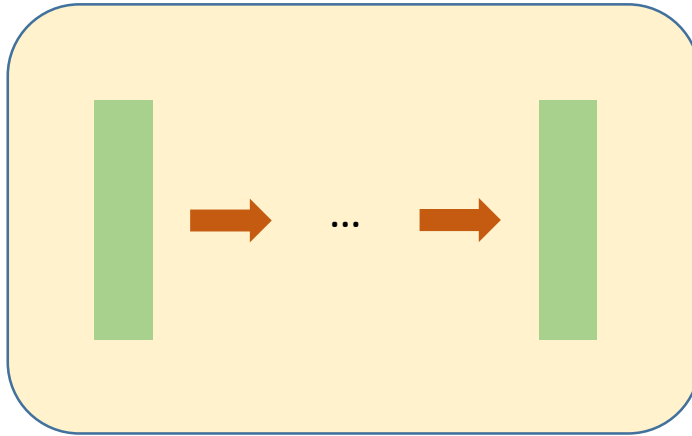


```
1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 # create model
6 model = keras.Sequential()
7 model.add(keras.Input(shape=(3,)))
8 model.add(keras.layers.Dense(1))
```

```
1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 # create model
6 model = keras.Sequential()
7 model.add(keras.Input(shape=(4,)))
8 model.add(keras.layers.Dense(3))
```

Keras

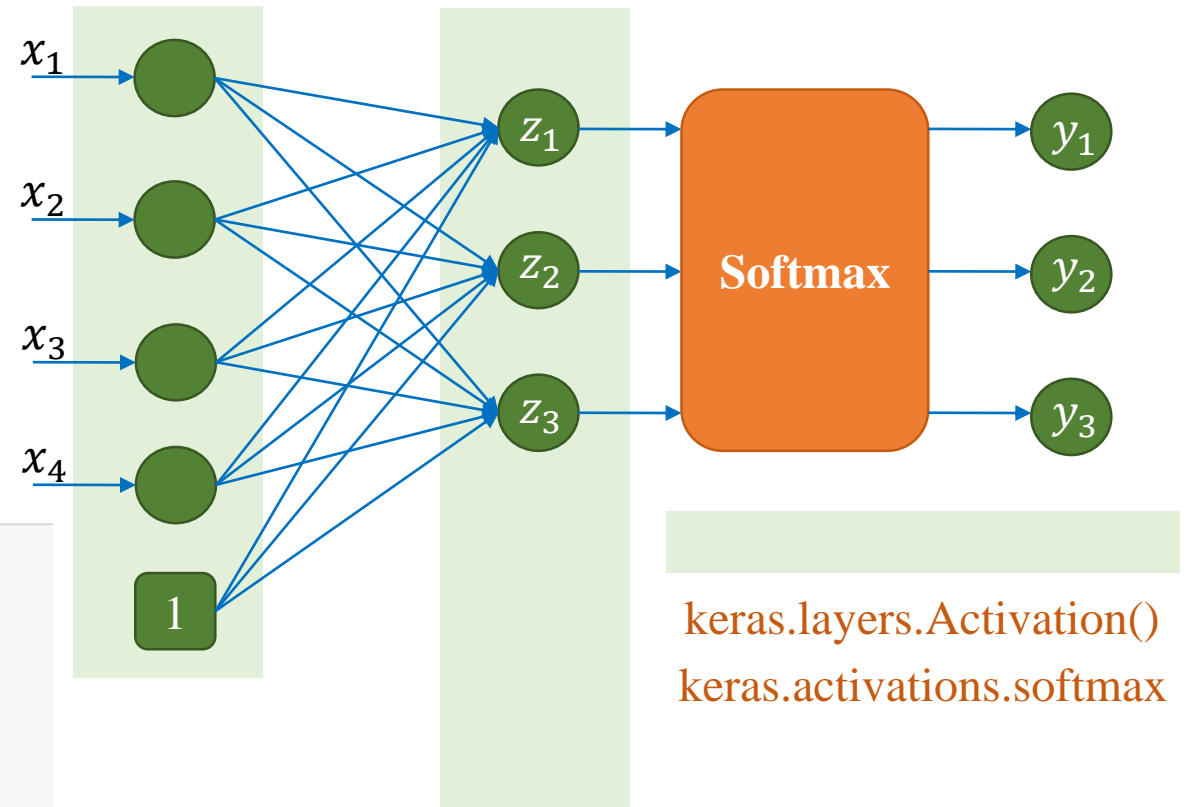
Model



`keras.Sequential()`

```
1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 # create model
6 model = keras.Sequential()
7 model.add(keras.Input(shape=(4,)))
8 model.add(keras.layers.Dense(3))
9 model.add(keras.layers.Activation(keras.activations.softmax))
```

`keras.Input(shape=(4,))`



`keras.layers.Activation()`
`keras.activations.softmax`

`keras.layers.Dense(units=3)`

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Model Construction

❖ Linear regression

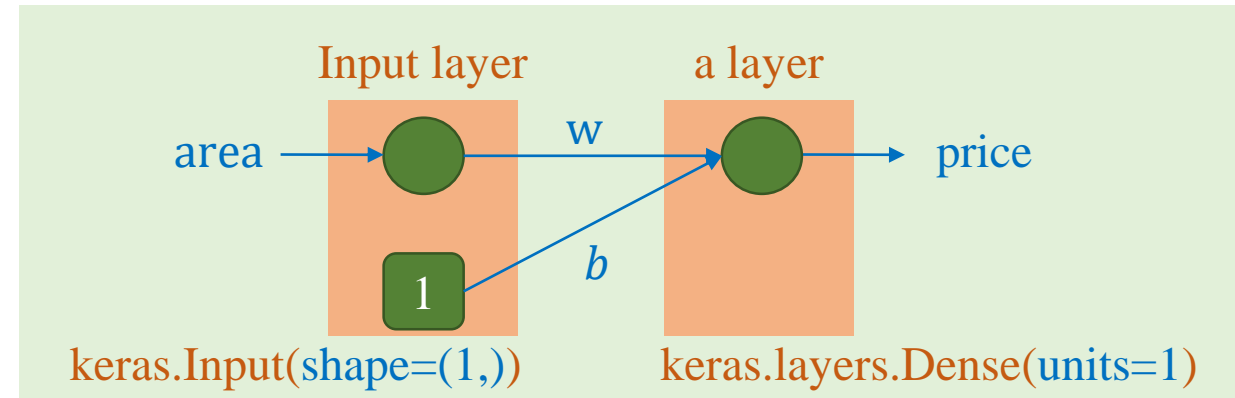
Feature	Label	
area	price	
6.7	9.1	
4.6	5.9	
3.5	4.6	
5.5	6.7	

House price data

Model

$$\text{price} = w * \text{area} + b$$

$$y = wx + b$$



```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(1,)))
7 model.add(keras.layers.Dense(1))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 1)	2

Total params: 2
Trainable params: 2
Non-trainable params: 0

Model Construction

❖ Linear regression

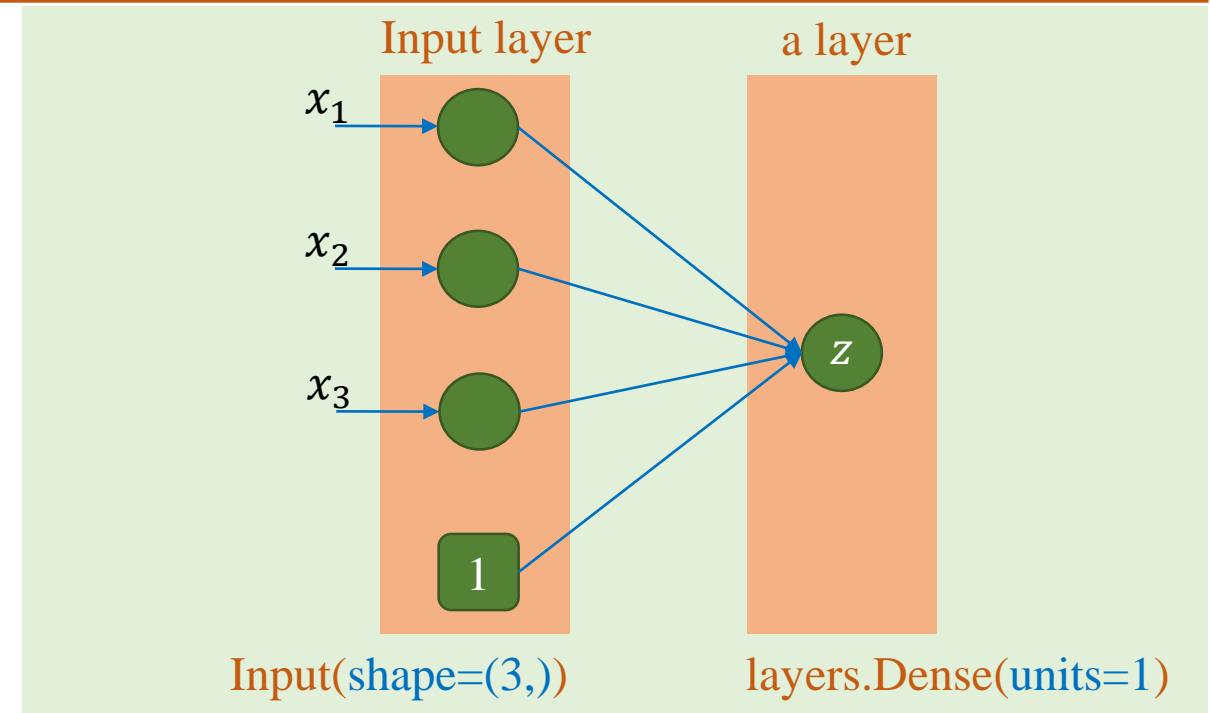
Features			Label
TV	↕ Radio	↕ Newspaper	↕ Sales
230.1	37.8	69.2	22.1
44.5	39.3	45.1	10.4
17.2	45.9	69.3	12
151.5	41.3	58.5	16.5
180.8	10.8	58.4	17.9

Advertising-based sale data

Model

$$\text{Sale} = w_1 * TV + w_2 * Radio + w_3 * Newspaper + b$$

$$y = w_1x_1 + w_2x_2 + w_3x_3 + b$$



```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(3,)))
7 model.add(keras.layers.Dense(1))
8
9 model.summary()
```


Model Construction

❖ Linear regression

Boston House
Price Data

Features														Label
crim	zn	indus	chas	nox	rm	age	dis	rad	tax	ptratio	black	lstat	medv	
0.00632	18	2.31	0	0.538	6.575	65.2	4.09	1	296	15.3	396.9	4.98	24	
0.02731	0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	396.9	9.14	21.6	
0.03237	0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	394.63	2.94	33.4	
0.06905	0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	396.9	5.33	36.2	
0.08829	12.5	7.87	0	0.524	6.012	66.6	5.5605	5	311	15.2	395.6	12.43	22.9	

Model

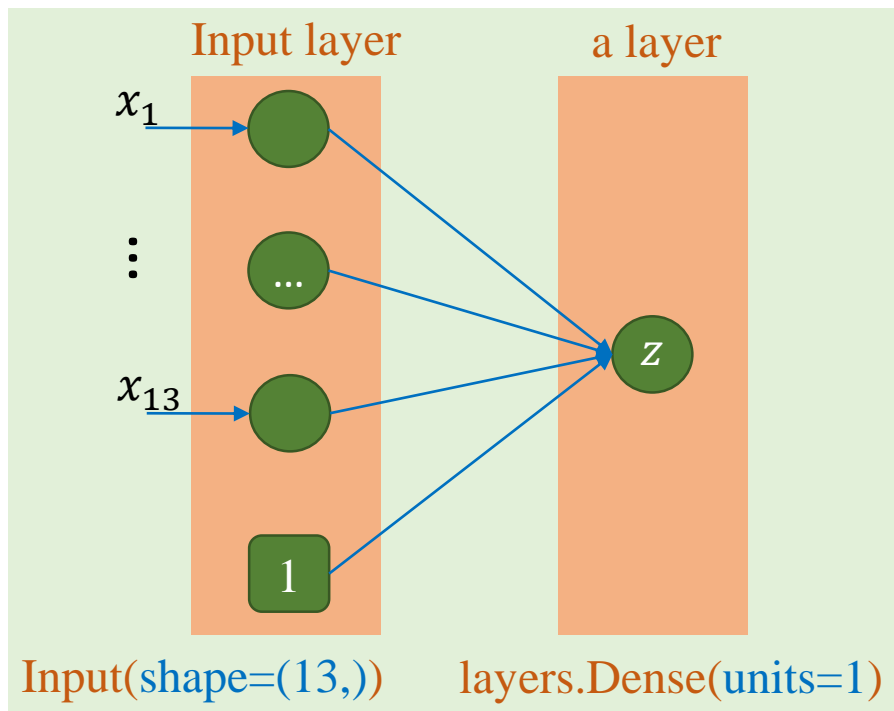
$$\text{medv} = w_1 * x_1 + \dots + w_{13} * x_{13} + b$$

Model Construction

❖ Linear regression

Model

$$\text{medv} = w_1 * x_1 + \dots + w_{13} * x_{13} + b$$



```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(13,)))
7 model.add(keras.layers.Dense(1))
8
9 model.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, 1)	14

Total params: 14

Trainable params: 14

Non-trainable params: 0

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Model Construction

Feature Label

Petal_Length	Category
1.4	0
1	0
1.5	0
3	1
3.8	1
4.1	1

Model

$$z = wx + b$$

$$\hat{y} = \frac{1}{1 + e^{-z}}$$

❖ Logistic regression

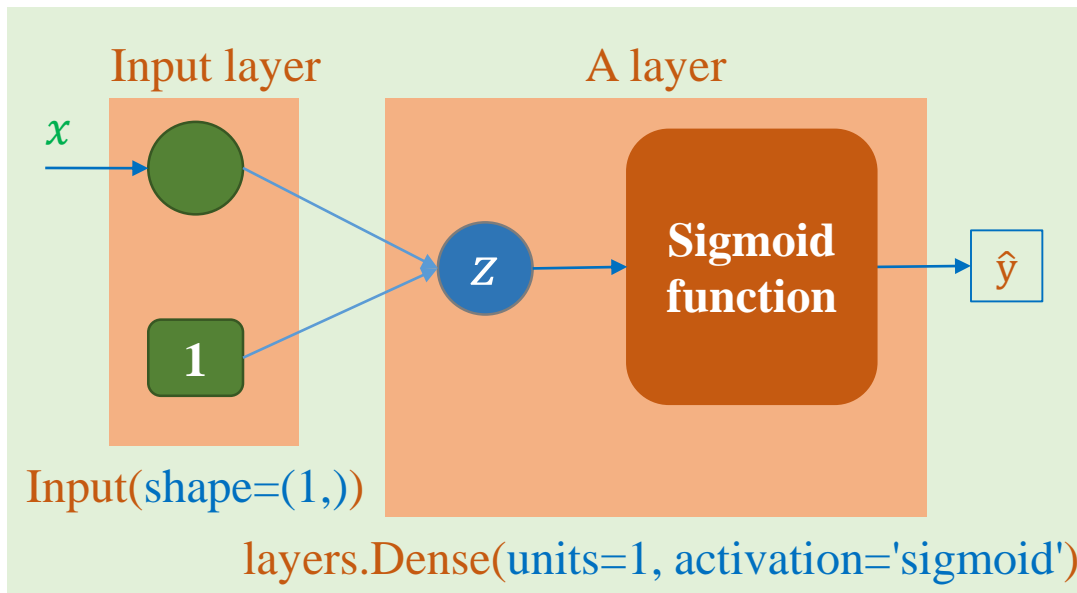
```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(1,)))
7 model.add(keras.layers.Dense(1, activation='sigmoid'))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 1)	2
Total params: 2		
Trainable params: 2		
Non-trainable params: 0		



Model Construction

❖ Logistic regression

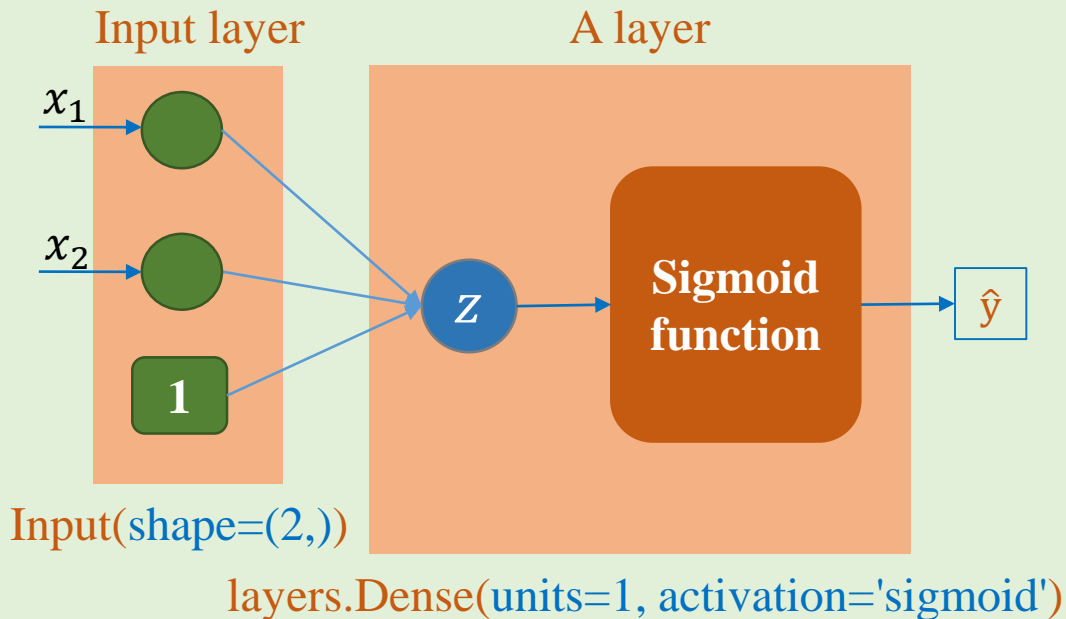
Feature Label

Petal_Length	Petal_Width	Label
1.5	0.2	0
1.4	0.2	0
1.6	0.2	0
4.7	1.6	1
3.3	1.1	1
4.6	1.3	1

Model

$$z = \theta^T x$$

$$\hat{y} = \frac{1}{1 + e^{-z}}$$



```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(2,)))
7 model.add(keras.layers.Dense(1, activation='sigmoid'))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 1)	3
Total params: 3		
Trainable params: 3		
Non-trainable params: 0		

Model Construction

Feature

Label

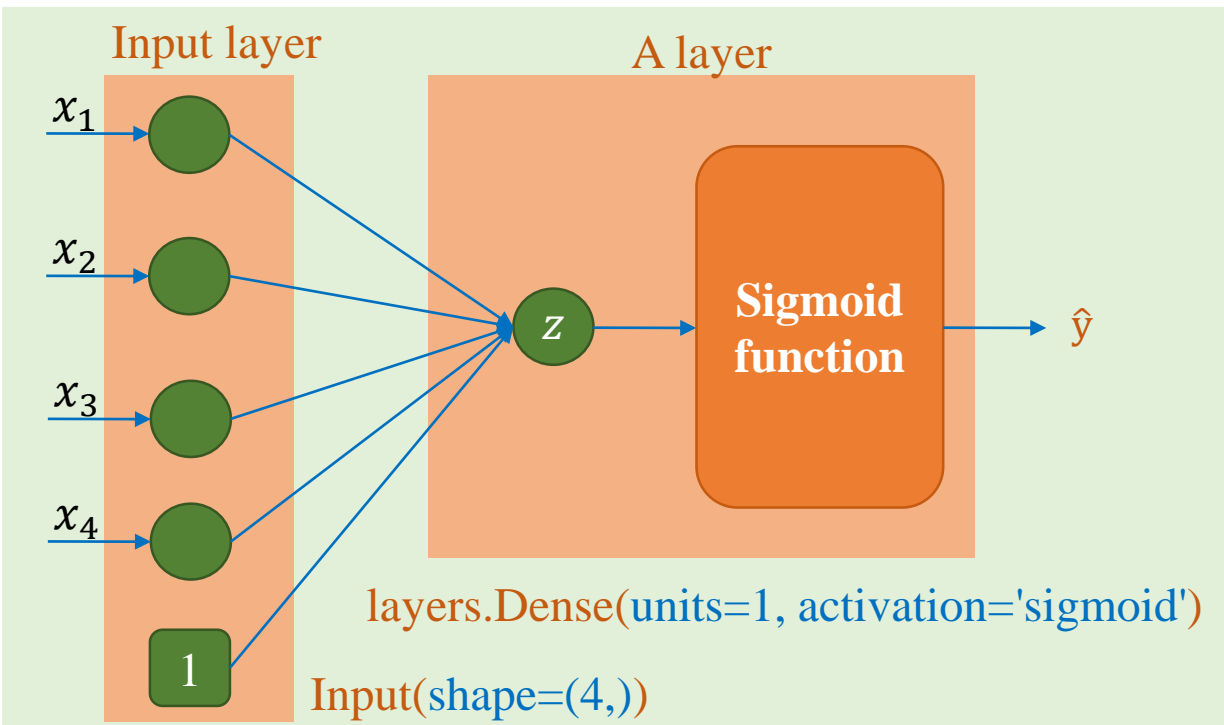
Sepal_Length	Sepal_Width	Petal_Length	Petal_Width	Label
5.2	3.5	1.5	0.2	0
5.2	3.4	1.4	0.2	0
4.7	3.2	1.6	0.2	0
6.3	3.3	4.7	1.6	1
4.9	2.4	3.3	1.1	1
6.6	2.9	4.6	1.3	1

Model

$$z = \theta^T x$$

$$\hat{y} = \frac{1}{1 + e^{-z}}$$

❖ Logistic regression



```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(4,)))
7 model.add(keras.layers.Dense(1, activation='sigmoid'))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 1)	5
Total params: 5		
Trainable params: 5		
Non-trainable params: 0		

Outline

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Model Construction

Feature Label

Petal_Length	Label
1.4	1
1.3	1
1.5	1
4.5	2
4.1	2
4.6	2

Iris Classification Data

$$z_1 = xw_1 + b_1$$

$$z_2 = xw_2 + b_2$$

$$\hat{y}_1 = \frac{e^{z_1}}{\sum_{j=1}^2 e^{z_j}}$$

$$\hat{y}_2 = \frac{e^{z_2}}{\sum_{j=1}^2 e^{z_j}}$$

❖ Softmax regression

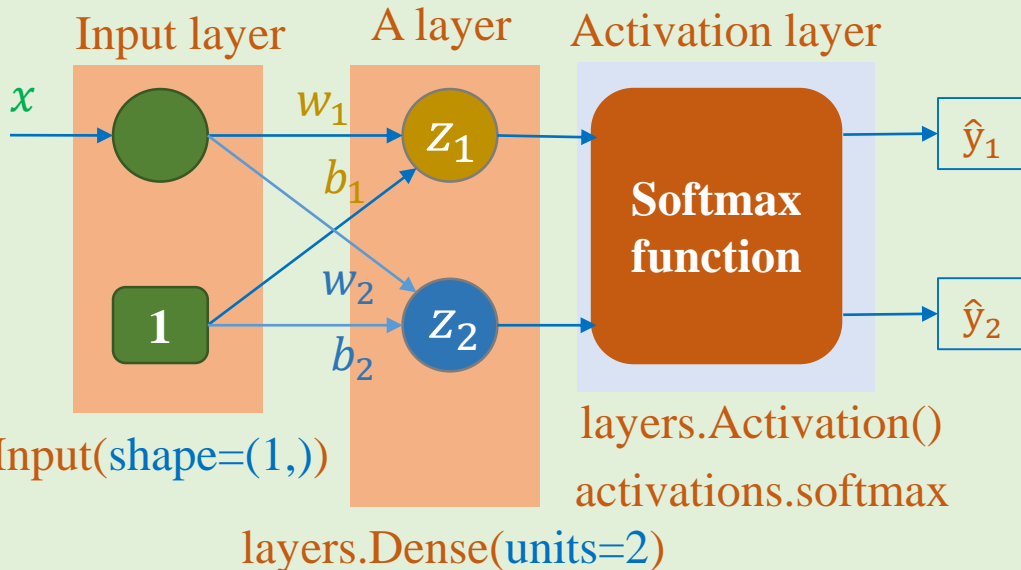
```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(1,)))
7 model.add(keras.layers.Dense(2))
8 model.add(keras.layers.Activation(keras.activations.softmax))
9
10 model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
dense (Dense)	(None, 2)	4

activation (Activation)	(None, 2)	0
=====		

Total params: 4
Trainable params: 4
Non-trainable params: 0



Model Construction

Feature Label

Petal_Length	Label
1.4	1
1.3	1
1.5	1
4.5	2
4.1	2
4.6	2

Iris Classification Data

$$z_1 = xw_1 + b_1$$

$$z_2 = xw_2 + b_2$$

$$\hat{y}_1 = \frac{e^{z_1}}{\sum_{j=1}^2 e^{z_j}}$$

$$\hat{y}_2 = \frac{e^{z_2}}{\sum_{j=1}^2 e^{z_j}}$$

❖ Softmax regression

```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(1,)))
7 model.add(keras.layers.Dense(2, activation='softmax'))
8
9 model.summary()
```

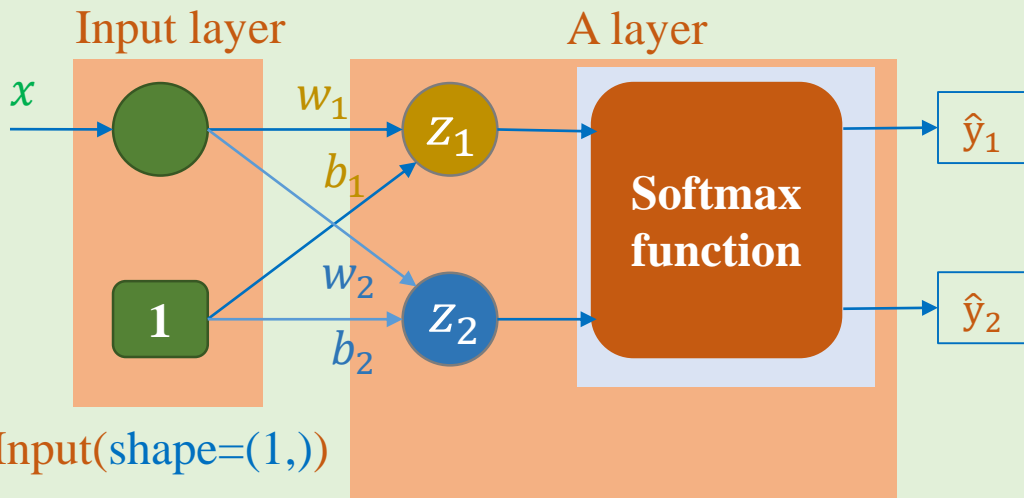
Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 2)	4

Total params: 4

Trainable params: 4

Non-trainable params: 0



Input(shape=(1,))

layers.Dense(units=2, activation='softmax')

Model Construction

❖ Softmax regression

Petal_Length	Label
1.4	1
1.3	1
1.5	1
4.5	2
4.1	2
4.6	2
5.2	3
5.6	3
5.9	3

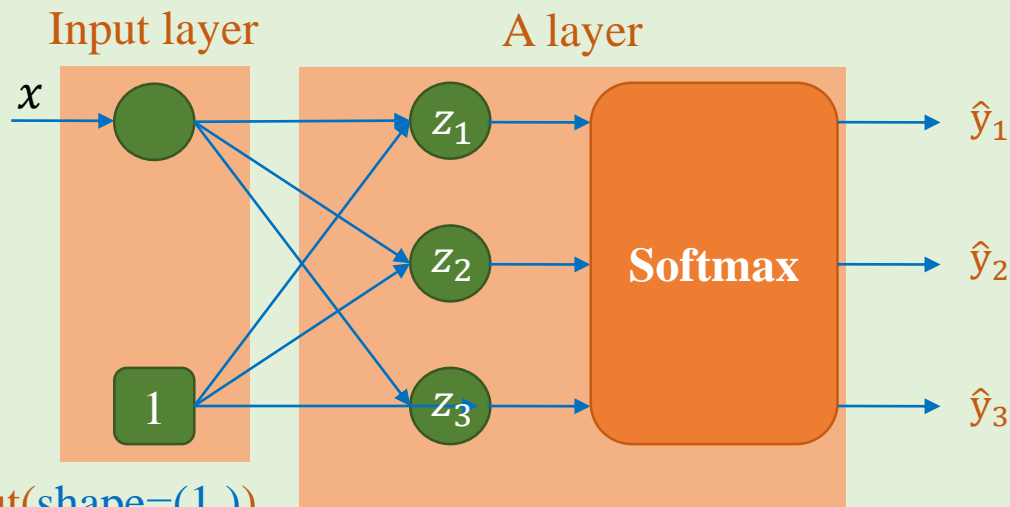
```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(1,)))
7 model.add(keras.layers.Dense(3, activation='softmax'))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 3)	6
Total params: 6		
Trainable params: 6		
Non-trainable params: 0		



layers.Dense(units=3, activation='softmax')

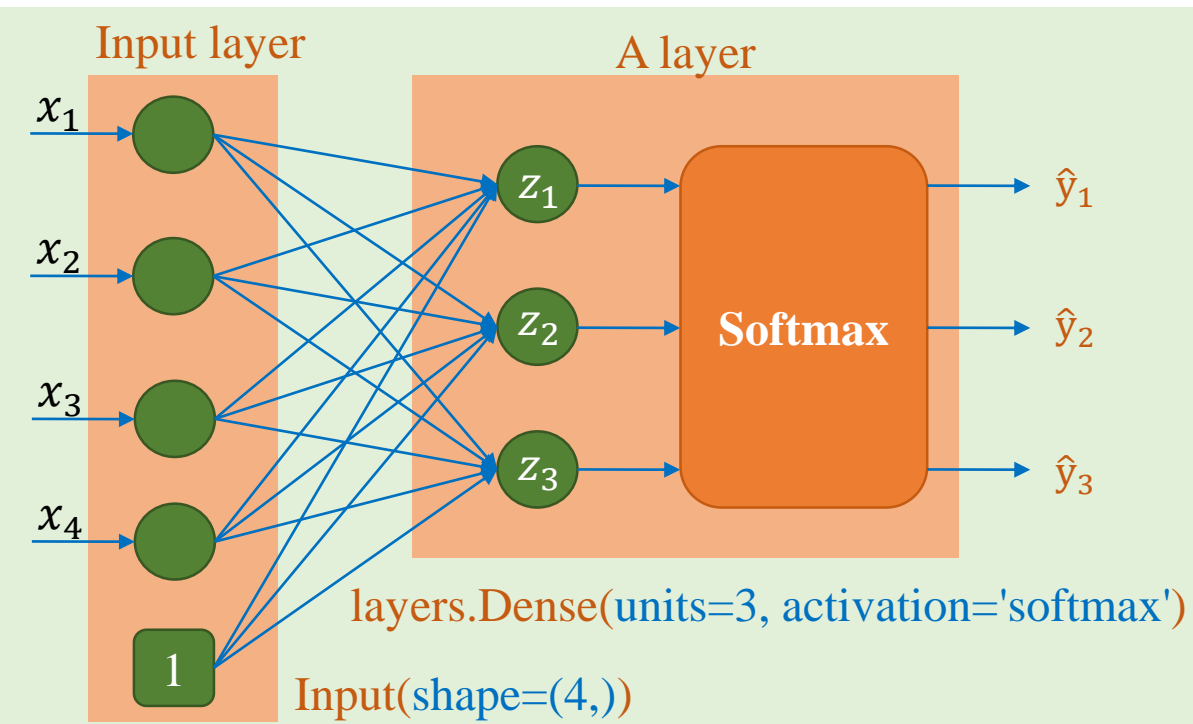
Model Construction

Sepal_Length	Sepal_Width	Petal_Length	Petal_Width	Label
5.2	3.5	1.5	0.2	1
5.2	3.4	1.4	0.2	1
4.7	3.2	1.6	0.2	1
6.3	3.3	4.7	1.6	2
4.9	2.4	3.3	1.1	2
6.6	2.9	4.6	1.3	2
6.4	2.8	5.6	2.2	3
6.3	2.8	5.1	1.5	3
6.1	2.6	5.6	1.4	3

❖ Softmax regression

Forward computation

$$z = \theta^T x \quad \hat{y} = \frac{e^z}{\sum_{i=1}^k e^{z_i}}$$



```

1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = keras.Sequential()
6 model.add(keras.Input(shape=(4,)))
7 model.add(keras.layers.Dense(3, activation='softmax'))
8
9 model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 3)	15
Total params: 15		
Trainable params: 15		
Non-trainable params: 0		

Outline

- **Introduction to Tensorflow**
- **Tensorflow/Keras**
- **Model Construction – Linear Regression**
- **Model Construction – Logistic Regression**
- **Model Construction – Softmax Regression**
- **Model Training**
- **Model Saving and Loading**

Training

❖ Logistic regression

→ Tính output \hat{y}

$$z = \boldsymbol{\theta}^T \mathbf{x}$$
$$\hat{y} = \sigma(z) = \frac{1}{1 + e^{-z}}$$

→ Tính loss (binary cross-entropy)

$$L(\boldsymbol{\theta}) = (-y^T \log \hat{y} - (1-y)^T \log(1-\hat{y}))$$

→ Tính đạo hàm

$$L'_{\boldsymbol{\theta}} = \mathbf{x}^T (\hat{y} - y)$$

→ Cập nhật tham số (Stochastic gradient descent)

$$\boldsymbol{\theta} = \boldsymbol{\theta} - \eta L'_{\boldsymbol{\theta}}$$

Computed automatically

Declare optimizer and loss function

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

Start training

```
model.fit(x-data, y-data, batch-size, epochs)
```

If batch-size=1 → Stochastic training

If batch-size=N → Batch training

If 1<batch-size<N → Mini-batch training

Training

❖ Softmax regression

→ Tính output \hat{y}

$$z = \theta^T x \quad \hat{y} = \frac{e^z}{\sum_{i=1}^k e^{z_i}}$$

→ Tính loss (cross-entropy)

$$L(\theta) = - \sum_{i=1}^k \delta(i, y) \log \hat{y}_i$$

→ Tính đạo hàm

$$\frac{\partial L}{\partial \theta_i} = x(\hat{y}_i - \delta(i, y))$$

→ Cập nhật tham số (Stochastic gradient descent)

$$\theta = \theta - \eta L'_\theta$$

Computed automatically

Declare optimizer and loss function

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

Start training

```
model.fit(x-data, y-data, batch-size, epochs)
```

If $\text{batch-size}=1 \rightarrow$ Stochastic training

If $\text{batch-size}=m \rightarrow$ Batch training

If $1 < \text{batch-size} < m \rightarrow$ Mini-batch training

Training

❖ Linear regression

Feature Label

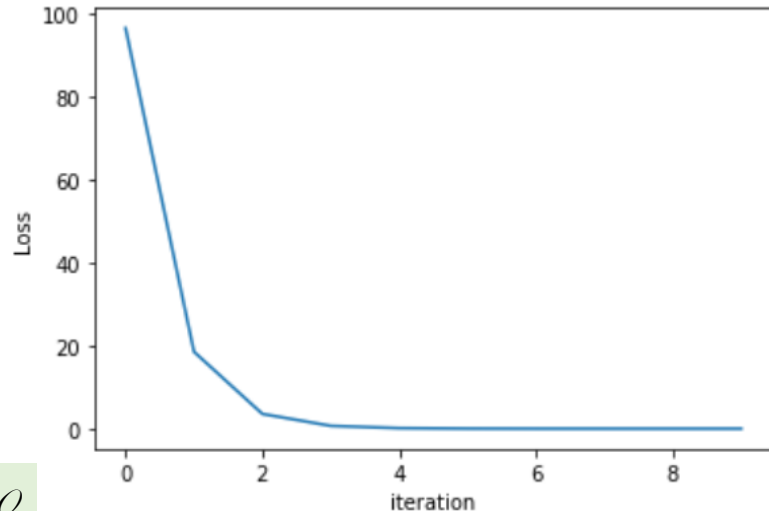
area	price
6.7	9.1
4.6	5.9
3.5	4.6
5.5	6.7

Model

$$\text{price} = w * \text{area} + b$$

$$y = wx + b$$

House price data



```

1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 batch_size = 4
6 epochs = 10
7
8 # Data Preparation
9 data = np.genfromtxt('data.csv', delimiter=',')
10 X = data[:,0:1]
11 y = data[:,1:]
12
13 # create model
14 model = tf.keras.Sequential(
15     [tf.keras.layers.Dense(units=1, input_shape=[1])]
16 )
17 # declare optimization method and loss function
18 opt = keras.optimizers.SGD(learning_rate=0.01)
19 model.compile(optimizer=opt, loss='mse')
20
21 # training
22 history = model.fit(X, y, batch_size, epochs)

```

Train on 4 samples

Epoch 1/10

4/4 [=====] - 0s 28ms/sample - loss: 96.6016

Epoch 2/10

4/4 [=====] - 0s 247us/sample - loss: 18.6414

Epoch 3/10

4/4 [=====] - 0s 499us/sample - loss: 3.6692

Epoch 4/10

4/4 [=====] - 0s 245us/sample - loss: 0.7937

Epoch 5/10

4/4 [=====] - 0s 499us/sample - loss: 0.2415

Epoch 6/10

4/4 [=====] - 0s 499us/sample - loss: 0.1353

Training

❖ Logistic regression

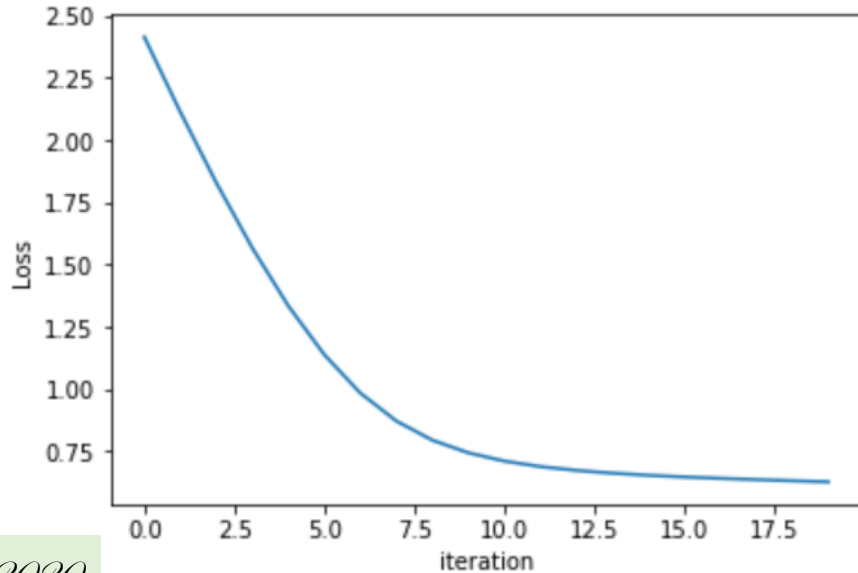
Feature Label

Petal_Length	Category
1.4	0
1	0
1.5	0
3	1
3.8	1
4.1	1

Model

$$z = wx + b$$

$$\hat{y} = \frac{1}{1 + e^{-z}}$$



```

1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 batch_size = 6
6 epochs = 20
7
8 # Data Preparation
9 data = np.genfromtxt('iris_1D.csv', delimiter=',', skip_header=1)
10 X = data[:,0:1]
11 y = data[:,1]
12
13 # create model
14 model = tf.keras.Sequential(
15     [tf.keras.layers.Dense(units=1, activation='sigmoid', input_shape=[1])])
16
17 # declare optimization method and loss function
18 opt = keras.optimizers.SGD(learning_rate=0.1)
19 model.compile(optimizer=opt, loss='binary_crossentropy')
20
21 # training
22 history = model.fit(X, y, batch_size, epochs)

```

Train on 6 samples

Epoch 1/20

6/6 [=====] - 0s 20ms/sample - loss: 2.4125

Epoch 2/20

6/6 [=====] - 0s 339us/sample - loss: 2.1118

Epoch 3/20

6/6 [=====] - 0s 326us/sample - loss: 1.8278

Epoch 4/20

6/6 [=====] - 0s 161us/sample - loss: 1.5661

Epoch 5/20

6/6 [=====] - 0s 333us/sample - loss: 1.3337

Epoch 6/20

6/6 [=====] - 0s 166us/sample - loss: 1.1381

Epoch 7/20

6/6 [=====] - 0s 160us/sample - loss: 0.9842

Training

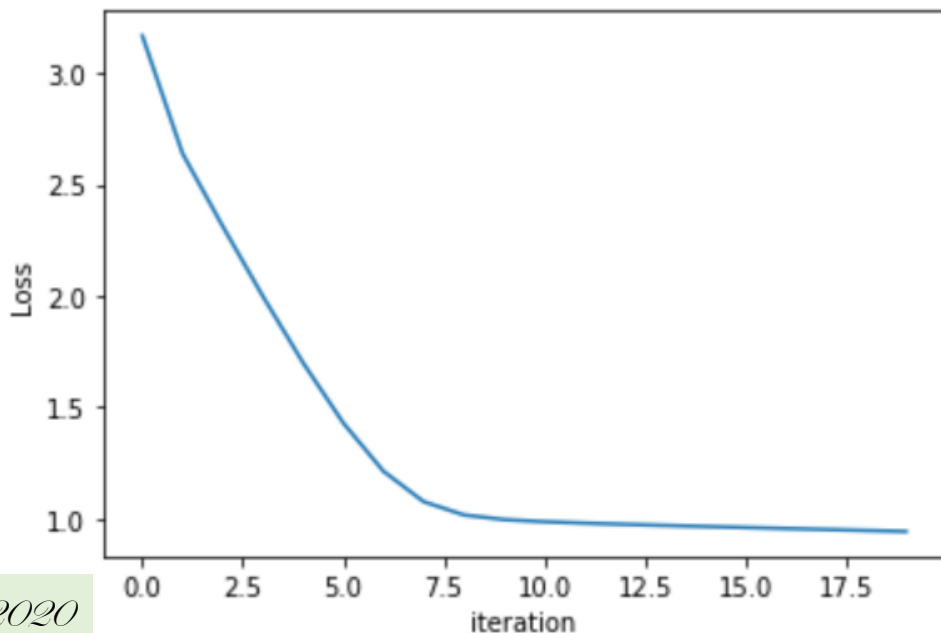
❖ Softmax regression

Model

$$z = \theta^T x$$

$$\hat{y} = \frac{e^z}{\sum_{i=1}^k e^{z_i}}$$

Petal_Length	Label
1.4	1
1.3	1
1.5	1
4.5	2
4.1	2
4.6	2
5.2	3
5.6	3
5.9	3



```

1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 batch_size = 9
6 epochs = 20
7
8 # Data Preparation
9 data = np.genfromtxt('iris_1D_3c.csv', delimiter=',', skip_header=1)
10 X = data[:,0:1]
11 y = data[:,1]
12
13 # create model
14 model = tf.keras.Sequential(
15     [tf.keras.layers.Dense(units=3, activation='softmax', input_shape=[1])])
16
17 # declare optimization method and loss function
18 opt = keras.optimizers.SGD(learning_rate=0.1)
19 model.compile(optimizer=opt, loss='sparse_categorical_crossentropy')
20
21 # training
22 history = model.fit(X, y, batch_size, epochs)

```

Train on 9 samples

```

Epoch 1/20
9/9 [=====] - 0s 19ms/sample - loss: 3.1676
Epoch 2/20
9/9 [=====] - 0s 221us/sample - loss: 2.6389
Epoch 3/20
9/9 [=====] - 0s 110us/sample - loss: 2.3121
Epoch 4/20
9/9 [=====] - 0s 221us/sample - loss: 1.9978
Epoch 5/20
9/9 [=====] - 0s 111us/sample - loss: 1.6993
Epoch 6/20
9/9 [=====] - 0s 220us/sample - loss: 1.4296
Epoch 7/20
9/9 [=====] - 0s 222us/sample - loss: 1.2127
Epoch 8/20
9/9 [=====] - 0s 221us/sample - loss: 1.0761
Epoch 9/20
9/9 [=====] - 0s 222us/sample - loss: 1.0165
Epoch 10/20
9/9 [=====] - 0s 222us/sample - loss: 0.9956

```

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Model Saving and Loading

Model Saving

```
1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 batch_size = 4
6 epochs = 10
7
8 # Data Preparation
9 data = np.genfromtxt('data.csv', delimiter=',')
10 X = data[:,0:1]
11 y = data[:,1:]
12
13 # create model
14 model = tf.keras.Sequential(
15     [tf.keras.layers.Dense(units=1, input_shape=[1])])
16
17 # declare optimization method and loss function
18 opt = keras.optimizers.SGD(learning_rate=0.01)
19 model.compile(optimizer=opt, loss='mse')
20
21 # training
22 history = model.fit(X, y, batch_size, epochs)
23
24 # save model
25 checkpoint_path = "my_model/model.ckpt"
26 model.save_weights(checkpoint_path)
```

Testing

```
1 # testing
2 X_testing = [[5.0]]
3 y_hat = model.predict(X_testing)
4 print(y_hat)
```

[[6.51236]]

Model Loading

```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # create model
5 model = tf.keras.Sequential(
6     [tf.keras.layers.Dense(units=1, input_shape=[1])])
7
8 # load model
9 model.load_weights('my_model/model.ckpt')
10
11 X_testing = [[5.0]]
12 y_hat = model.predict(X_testing)
13 print(y_hat)
```

[[6.5058403]]

Model Saving and Loading

Model Saving

```
1 import numpy as np
2 import tensorflow as tf
3 import tensorflow.keras as keras
4
5 batch_size = 4
6 epochs = 10
7
8 # Data Preparation
9 data = np.genfromtxt('data.csv', delimiter=',')
10 X = data[:,0:1]
11 y = data[:,1:]
12
13 # create model
14 model = tf.keras.Sequential(
15     [tf.keras.layers.Dense(units=1, input_shape=[1])])
16
17 # declare optimization method and loss function
18 opt = keras.optimizers.SGD(learning_rate=0.01)
19 model.compile(optimizer=opt, loss='mse')
20
21 # training
22 history = model.fit(X, y, batch_size, epochs)
23
24 # save entire model
25 model.save('my_model/model.h5')
```

Model Loading

```
1 import tensorflow as tf
2 import tensorflow.keras as keras
3
4 # load model
5 model = tf.keras.models.load_model('my_model/model.h5')
6
7 # testing
8 X_testing = [[5.0]]
9 y_hat = model.predict(X_testing)
10 print(y_hat)
```

```
[[6.5115185]]
```

Tensorflow

❖ Demo

```
Python 3.7.3 (default, Apr 24 2019, 15:29:51) [MSC v.1915 64 bit (AMD64)] ::  
Type "help", "copyright", "credits" or "license" for more information.  
>>>  
>>>  
>>>  
>>>  
>>>  
>>>  
>>>  
>>>  
>>> for epoch in range(n_epochs):  
...     sum_of_losses = 0  
...     gradients = np.zeros((2,1))  
...  
...     for index in range(4):  
...         xi = X_b[index:index+1]  
...         yi = y[index:index+1]
```

Reference

Tensor

<https://www.tensorflow.org/guide/tensor>

TensorFlow 2 quickstart for beginners

<https://www.tensorflow.org/tutorials/quickstart/beginner>

Save and load models

https://www.tensorflow.org/tutorials/keras/save_and_load

