

FIT 3181/5215 Deep Learning

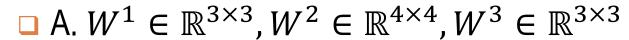
# Quiz for: Feed-forward Neural Nets with TensorFlow

**Teaching Team** 

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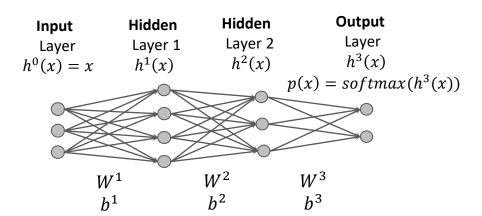
□ Given the following feed-forward neural network. What are the shapes of weight matrices if we follow the convention in the lecture (not TF implementation)?



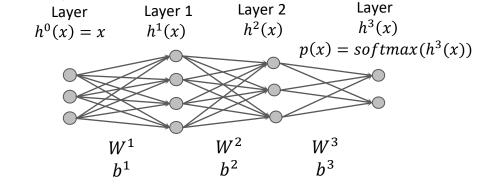
$$\square$$
 B.  $W^1 \in \mathbb{R}^{4 \times 3}$ ,  $W^2 \in \mathbb{R}^{3 \times 4}$ ,  $W^3 \in \mathbb{R}^{2 \times 3}$ 

$$\square$$
 C.  $W^1 \in \mathbb{R}^{3\times4}$ ,  $W^2 \in \mathbb{R}^{4\times3}$ ,  $W^3 \in \mathbb{R}^{3\times2}$ 

$$\square$$
 D.  $W^1 \in \mathbb{R}^{4 \times 4}$ ,  $W^2 \in \mathbb{R}^{3 \times 3}$ ,  $W^3 \in \mathbb{R}^{2 \times 2}$ 



□ Given the following feed-forward neural network. What are the shapes of weight matrices if we follow the convention in the lecture (not TF implementation)?



Hidden

Hidden

Input

Output

$$\square$$
 A.  $W^1 \in \mathbb{R}^{3 \times 3}$ ,  $W^2 \in \mathbb{R}^{4 \times 4}$ ,  $W^3 \in \mathbb{R}^{3 \times 3}$ 

$$\square$$
 B.  $W^1 \in \mathbb{R}^{4 \times 3}$ ,  $W^2 \in \mathbb{R}^{3 \times 4}$ ,  $W^3 \in \mathbb{R}^{2 \times 3}$  [x]

$$\square$$
 C.  $W^1 \in \mathbb{R}^{3\times 4}$ ,  $W^2 \in \mathbb{R}^{4\times 3}$ ,  $W^3 \in \mathbb{R}^{3\times 2}$ 

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 D.  $W^1 \in \mathbb{R}^{4 \times 4}$ ,  $W^2 \in \mathbb{R}^{3 \times 3}$ ,  $W^3 \in \mathbb{R}^{2 \times 2}$ 

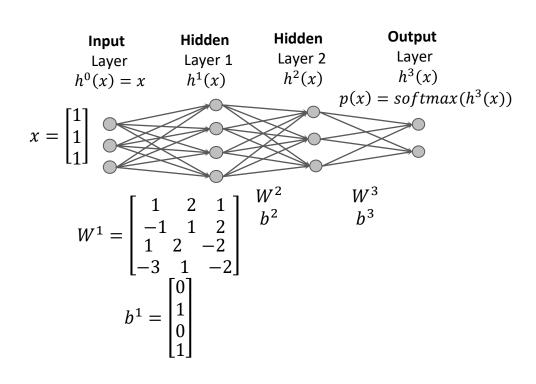
□ Given the following feed-forward neural network. Assume that we input to the network feature vector  $x = [1 \ 1 \ 1]^T$ . What is the values of pre-activations  $\overline{h}^1$ ?

$$\blacksquare$$
 A.  $\bar{h}^1 = [4\ 2\ 1 - 4]$ 

$$\blacksquare$$
 B.  $\bar{h}^1 = [4 \ 3 \ 1 \ -3]$ 

$$\Box$$
 C.  $\bar{h}^1 = [4 \ 3 \ 1 \ -3]^T$ 

$$\Box$$
 D. $\bar{h}^1 = [4 \ 2 \ 1 - 4]^T$ 



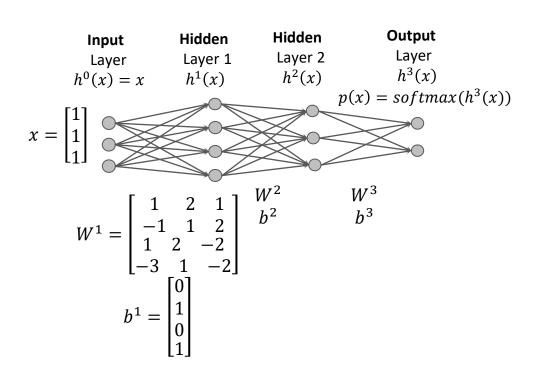
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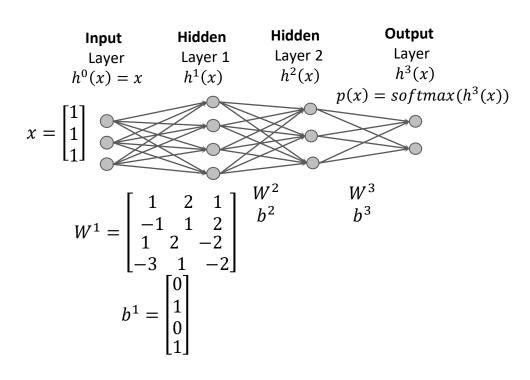
□ Given the following feed-forward neural network. Assume that we input to the network feature vector  $x = [1 \ 1 \ 1]^T$ . What is the hidden values  $h^1$  if we use ReLU activation function?

$$\square$$
 A.  $h^1 = [0 \ 0 \ 0 - 4]^T$ 

$$\blacksquare$$
 B. h<sup>1</sup> =  $[4 \ 3 \ 1 \ 0]^T$ 

$$\Box$$
 C. h<sup>1</sup> =  $[0\ 0\ 0\ -3]^T$ 

$$\Box$$
 D. h<sup>1</sup> = [4 2 1 - 4]<sup>T</sup>



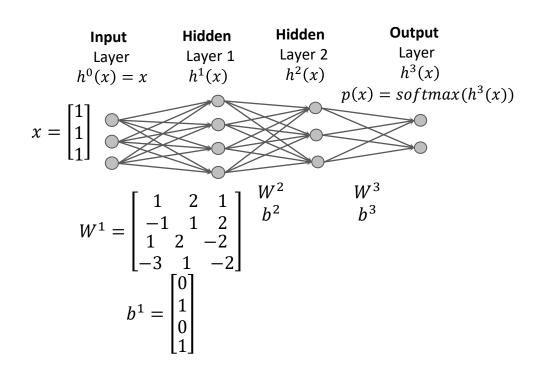
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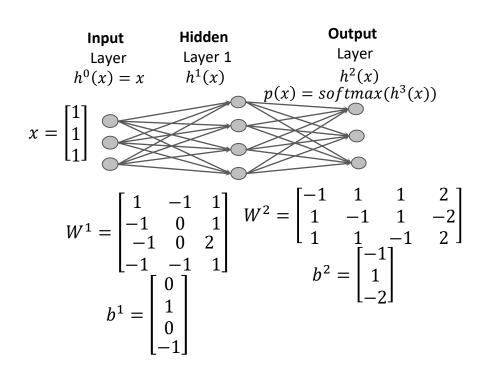
$$\Box$$
 D. h<sup>1</sup> = [4 2 1 - 4]<sup>T</sup>



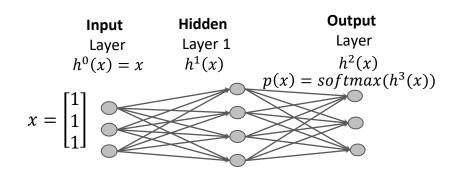
□ Given the following feed-forward neural network. Assume that we input to the network feature vector  $x = [1 \ 1 \ 1]^T$ . What are the correct statements if we use ReLU activation function? (MC)

$$\square$$
 A. h<sup>1</sup> = [1 1 1 0]<sup>T</sup>

- $\square$  B. h<sup>1</sup> = [1 1 1 1]<sup>T</sup>
- $\Box$  C. Logit  $h^2 = [0 \ 2 \ -1]^T$
- $\Box$  D. Logit  $h^2 = [0 \ 2 \ 0]^T$



□ Given the following feed-forward neural network. Assume that we input to the network feature vector  $x = [1 \ 1 \ 1]^T$ . What are the correct statements if we use ReLU activation function? (MC)



$$\square$$
 A.  $h^1 = [1 \ 1 \ 1 \ 0]^T$  [x]

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 B. h<sup>1</sup> = [1 1 1 1]<sup>T</sup>

$$\Box$$
 C. Logit  $h^2 = [0 \ 2 \ -1]^T$  [x]

$$\Box$$
 D. Logit  $h^2 = [0 \ 2 \ 0]^T$ 

$$W^{1} = \begin{bmatrix} 1 & -1 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 2 \\ -1 & -1 & 1 \end{bmatrix} \times \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \qquad W^{2} = \begin{bmatrix} -1 & 1 & 1 & 2 \\ 1 & -1 & 1 & -2 \\ 1 & 1 & -1 & 2 \end{bmatrix} \times \begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

$$b^{1} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \end{bmatrix} \qquad b^{2} = \begin{bmatrix} -1 \\ 1 \\ -2 \end{bmatrix}$$

Given an implementation as below. Which of following architecture is correct (SC).

```
dnn_model = Sequential()
dnn_model.add(Dense(units=32, input_shape=(784,), activation='relu'))
dnn_model.add(Dense(units=64, activation='relu'))
dnn_model.add(Dense(units=64, activation='relu'))
dnn_model.add(Dense(units=32, activation='relu'))
dnn_model.add(Dense(units=10, activation='softmax'))
```

- $\triangle$  A. 784 $\rightarrow$ 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(ReLU)
- □ B.  $784 \rightarrow 32(ReLU) \rightarrow 64(ReLU) \rightarrow 64(ReLU) \rightarrow 32(ReLU) \rightarrow 10(Softmax)$
- $\square$  C. 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(ReLU)
- □ D. 784(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(Softmax)

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- $\triangle$  A. 784 $\rightarrow$ 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(ReLU)
- □ B.  $784 \rightarrow 32(ReLU) \rightarrow 64(ReLU) \rightarrow 64(ReLU) \rightarrow 32(ReLU) \rightarrow 10(Softmax) [x]$
- $\square$  C. 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(ReLU)
- □ D. 784(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 64(ReLU) $\rightarrow$ 32(ReLU) $\rightarrow$ 10(Softmax)

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```

- A. The model has 5 Fully Connected layers
- B. The batch size is 784
- C. The model can work with an arbitrary batch size
- □ D. The model's output is a logit value and in range [-inf, +inf]

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dnn_model.add(Dense(units=32, activation='relu'))
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```

- □ A. The model has 5 Fully Connected layers [x]
- B. The batch size is 784
- C. The model can work with an arbitrary batch size [x]
- □ D. The model's output is a logit value and in range [-inf, +inf]

Given an implementation as below. What are outputs of the two print functions (SC).

```
[5]: dnn_model = Sequential()
    dnn_model.add(Dense(units=32, input_shape=(784,), activation='relu'))
    dnn_model.add(Dense(units=64, activation='relu'))
    dnn_model.add(Dense(units=32, activation='relu'))
    dnn_model.add(Dense(units=32, activation='relu'))
    dnn_model.add(Dense(units=10, activation='softmax'))

[8]: hidden1 = dnn_model.layers[0]
    weights, biases = hidden1.get_weights()
    print('shape W=',weights.shape)
    print('shape b=',biases.shape)
```

- □ A. (32, 32), (32,)
- □ B. (32, 784), (784,)
- □ C. (784, 32), (32,)
- □ D. (784, 32), (784,)

Given an implementation as below. What are outputs of the two print functions (SC).

```
[5]: dnn_model = Sequential()
  dnn_model.add(Dense(units=32, input_shape=(784,), activation='relu'))
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  print('shape W=',weights.shape)
  print('shape b=',biases.shape)
```

- □ A. (32, 32), (32,)
- □ B. (32, 784), (784,)
- □ C. (784, 32), (32,) [x]
- □ D. (784, 32), (784,)

Given an implementation as below. What is the total parameters of the model (SC).

```
dnn_model = Sequential()
dnn_model.add(Dense(units=20, input_shape=(10,), activation='relu'))
dnn_model.add(Dense(units=20, activation='relu'))
dnn_model.add(Dense(units=10, activation='softmax'))
```

- □ A. 800
- □ B. 830
- □ C. 840
- D. 850

Given an implementation as below. What is the total parameters of the model (SC).

```
dnn_model = Sequential()
dnn_model.add(Dense(units=20, input_shape=(10,), activation='relu'))
dnn_model.add(Dense(units=20, activation='relu'))
dnn_model.add(Dense(units=10, activation='softmax'))
```

- □ A. 800
- □ B. 830
- **C.** 840
- □ D. 850 [x]

```
Architecture 10 \rightarrow 20 (ReLU) \rightarrow 20 (ReLU) \rightarrow 10 (softmax) (20 \times 10 + 20) + (20 \times 20 + 20) + (10 \times 20 + 10) = 850
```