1.

### Congratulations! You passed!

Grade received 100% Latest Submission Grade 100% To pass 80% or higher

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1/1 point

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## Softmax regression (4 possible outputs)

$$x z_1 = \vec{w}_1 \cdot \vec{x} + b_1$$

$$a_1 = \frac{e^{z_1}}{e^{z_1} + e^{z_2} + e^{z_3} + e^{z_4}}$$

$$= P(y = 1|\vec{x}) \ O.30$$

For a multiclass classification task that has 4 possible outputs, the sum of all the activations adds up to 1. For a multiclass classification task that has 3 possible outputs, the sum of all the activations should add up to ....

- O It will vary, depending on the input x.
- O More than 1
- O Less than 1
- 1

#### Correct Correct

Yes! The sum of all the softmax activations should add up to 1. One way to see this is that if  $e^{z_1}=10$ ,  $e^{z_2}=20$ ,  $e^{z_3}=30$ , then the sum of  $a_1+a_2+a_3$  is equal to  $\frac{e^{z_1}+e^{z_2}+e^{z_3}}{1+e^{z_2}+e^{z_3}}$  which is 1.

2.

# Cost

# Logistic regression

$$z = \overrightarrow{w} \cdot \overrightarrow{x} + b$$

$$a_1 = g(z) = \frac{1}{1 + e^{-z}} = P(y = 1 | \overrightarrow{x})$$

$$a_2 = 1 - a_1 = P(y = 0 | \overrightarrow{x})$$

$$a_3 = 1 - a_1 = P(y = 0 | \overrightarrow{x})$$

$$a_4 = 1 + a_1 = P(y = 0 | \overrightarrow{x})$$

$$a_5 = 1 + a_1 = P(y = 0 | \overrightarrow{x})$$

$$a_7 = 1 + a_1 = P(y = 0 | \overrightarrow{x})$$

$$a_8 = 1 + a_1 = P(y = 0 | \overrightarrow{x})$$

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# Softmax regression

$$a_{1} = \frac{e^{z_{1}}}{e^{z_{1}} + e^{z_{2}} + \dots + e^{z_{N}}} = P(y = 1|\vec{x})$$

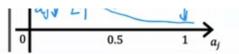
$$\vdots$$

$$a_{N} = \frac{e^{z_{N}}}{e^{z_{1}} + e^{z_{2}} + \dots + e^{z_{N}}} = P(y = N|\vec{x})$$

Crossentropy loss
$$loss(a_1, ..., a_N, y) = \begin{cases} -\log a_1 & \text{if } y = 1 \\ -\log a_2 & \text{if } y = 2 \end{cases}$$

$$\vdots$$

$$-\log a_N & \text{if } y = N \end{cases}$$



For multiclass classification, the cross entropy loss is used for training the model. If there are 4 possible classes for the output, and for a particular training example, the true class of the example is class 3 (y=3), then what does the cross entropy loss simplify to? [Hint: This loss should get smaller when  $a_3$  gets larger.]

- $\bigcirc -log(a_3)$
- O z\_3
- $\bigcirc \frac{-log(a_1) + -log(a_2) + -log(a_3) + -log(a_4)}{4}$
- O z\_3/(z\_1+z\_2+z\_3+z\_4)

3.

Correct. When the true label is 3, then the cross entropy loss for that training example is just the negative of the log of the activation for the third neuron of the softmax. All other terms of the cross entropy loss equation  $(-log(a_1), -log(a_2), and - log(a_4))$  are ignored

1/1 point

```
MNIST (more numerically accurate)
          from tensorflow.keras import Sequential
          from tensorflow.keras.layers import Dense
          model = Sequential([
            Dense (units=25, activation='relu')
           Dense (units=15, activation='relu')
           Dense (units=10, activation='linear') )]
loss
          from tensorflow.keras.losses import
           SparseCategoricalCrossentropy
         model.compile(...,loss=SparseCategoricalCrossentropy(from_logits=True) )
fit
          model.fit(X,Y,epochs=100)
predict
          logits = model(X)
          f_x = tf.nn.softmax(logits)
```

For multiclass classification, the recommended way to implement softmax regression is to set from logits=True in the loss function, and also to define the model's output layer with...

- a 'linear' activation
- a 'softmax' activation

Yes! Set the output as linear, because the loss function handles the calculation of the softmax with a more numerically stable method.