## Congratulations! You passed!

Grade received 100%

Latest Submission Grade 100% To pass 80% or higher

Go to next item

- 1. Which is an example of a classification task?
  - Based on the size of each tumor, determine if each tumor is malignant (cancerous) or not.
  - O Based on a patient's blood pressure, determine how much blood pressure medication (a dosage measured in milligrams) the patient should be prescribed.
  - O Based on a patient's age and blood pressure, determine how much blood pressure medication (measured in milligrams) the patient should be prescribed.
  - **⊘** Correct

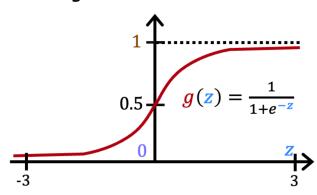
This task predicts one of two classes, malignant or not malignant.

**2.** Recall the sigmoid function is  $g(z)=rac{1}{1+e^{-z}}$ 

1/1 point

1/1 point

## sigmoid function

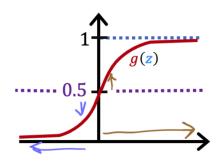


If z is a large positive number, then:

- $\bigcirc \ g(z)$  will be near zero (0)
- $\bigcirc \ g(z)$  is near negative one (-1)
- igodeauticle g(z) is near one (1)
- $\bigcirc \ g(z)$  will be near 0.5

 $\odot$  correct Say z = +100. So  $e^{-z}$  is then  $e^{-100}$ , a really small positive number. So,  $g(z)=rac{1}{1+{
m a \ small \ positive \ number}}$  which is close to 1

3.



1/1 point

A cat photo classification model predicts 1 if it's a cat, and 0 if it's not a cat. For a particular photograph, the logistic regression model outputs g(z) (a number between 0 and 1). Which of these would be a reasonable criteria to decide whether to predict if it's a cat?

- $\bigcirc$  Predict it is a cat if g(z) < 0.5
- O Predict it is a cat if g(z) = 0.5
- $\bigcirc$  Predict it is a cat if g(z) < 0.7
- Predict it is a cat if g(z) >= 0.5

## ✓ Correct

Think of g(z) as the probability that the photo is of a cat. When this number is at or above the threshold of 0.5, predict that it is a cat.

4.		1 / 1 poi
	True/False? No matter what features you use (including if you use polynomial features), the decision boundary learned by logistic regression will be a linear decision boundary.	
	○ True	
	False	
	<ul> <li>Correct         The decision boundary can also be non-linear, as described in the lectures.     </li> </ul>	