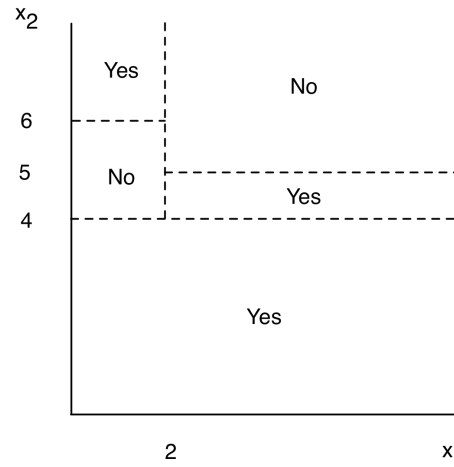


Name: _____

1. **Decision Trees** Consider the following figure, which shows the decision surface learned by a decision tree over two numerical features X_1 and X_2 . The tree predicts a target attribute that takes on the values “Yes” and “No.”



Draw a decision tree that is consistent with the above classification decisions. Try to make your tree as “compact” as possible (i.e., as few decision nodes as possible).

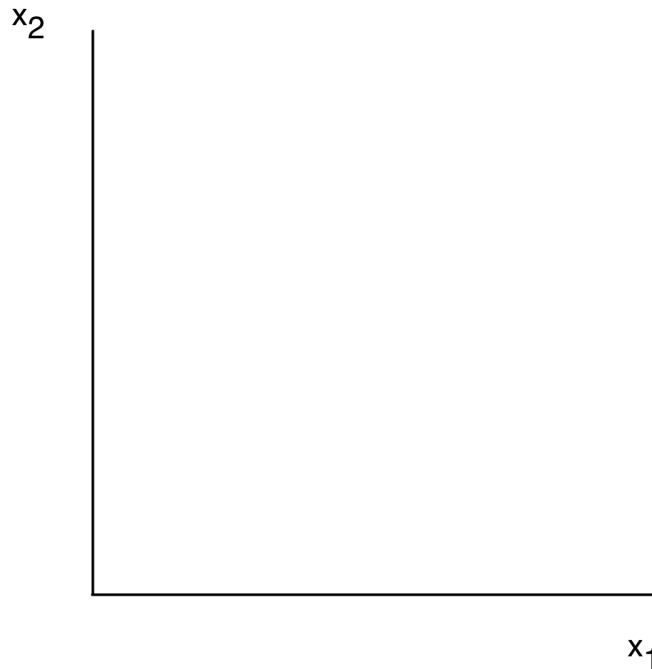
2. **Perceptrons** Suppose you use a perceptron to learn a linear discriminant of a target attribute that takes on two values “Yes” and “No.” You give it labeled data that has two features, X_1 and X_2 and it returns the following set of three weights: $w_0 = 12$, $w_1 = 2$, $w_2 = -4$.

Recall that a perceptron makes a prediction by calculating

$$h(x) = \text{sign} \left(\sum_{i=0}^d w_i x_i \right)$$

Here we'll assume that $h(x) = 1$ will be interpreted as a prediction of “Yes” and $h(x) = -1$ will be interpreted as a prediction of “No.”

- (a) What label would the input $x_1 = 2$, $x_2 = 5$ be assigned?
- (b) Draw the decision boundary in the feature space below. In other words, show which points will be labeled “Yes” and which ones labeled “No.”



3. **Feature Engineering** You are a Hollywood producer. You have a script in your hand and you want to make a movie. Before starting, however, you want to predict if the film you want to make will rake in huge profits, or utterly fail at the box office. You hire two critics A and B to read the script and rate it on a scale of 1 to 5 (assume only integer scores). Each critic reads it independently and announces their verdict. Of course, the critics might be biased and/or not perfect. For instance, for the past five movies you made, these are the critics scores and the performance of the movie:

Movie Name	A	B	Profit?
Mad Max	2	2	No
Harry Potter I	4	5	No
Harry Potter II	3	2	Yes
Mystic Pizza	3	4	Yes
Citizen Kane	2	3	Yes

You decide to use machine learning to learn a classifier that will predict the profitability of the movie based on the reviewers' ratings. Suppose you directly use the scores given above as features. That is, x_A = score given by A and x_B = score given by B.

Would the perceptron converge on this data? (Hint: you might try plotting the data as points in x_A, x_B space.)

4. Your co-executive producer, Mark Wahlberg, thinks your whole approach is ridiculous. He claims that a perceptron isn't "powerful" enough to make predictions because it's too simplistic. He wants some evidence that a perceptron could get the right answer under various scenarios.

For each scenario below, circle it if a perceptron using the features above could indeed perfectly classify the data. You can assume that for each scenario, you will have available a large amount of training data that follows the pattern described in the scenario.

- (a) One or more good reviews leads to success. If the total of their scores is more than 8, then the movie will definitely be a success and otherwise it will fail.
- (b) Mediocre reviews breeds success. Your movie will succeed if and only if each reviewer gives either a score of 2 or a score of 3.
- (c) Your reviewers have complementary tastes. Your movie will succeed if and only if both reviewers agree.

5. At Mark Wahlberg's insistence, you try using a *different* set of features. Consider the following feature space:

$$\begin{aligned}
 x_0 &= 1 && \text{(The bias feature, always equal to 1)} \\
 x_{1A} &= 1 && \text{if score given by A is 1, 0 otherwise} \\
 x_{1B} &= 1 && \text{if score given by B is 1, 0 otherwise} \\
 x_{2A} &= 1 && \text{if score given by A is 2, 0 otherwise} \\
 x_{2B} &= 1 && \text{if score given by B is 2, 0 otherwise} \\
 &\dots && \dots \\
 x_{5A} &= 1 && \text{if score given by A is 5, 0 otherwise} \\
 x_{5B} &= 1 && \text{if score given by B is 5, 0 otherwise}
 \end{aligned}$$

Consider again the three scenarios in part (4). Using a perceptron with the new features, which of the three scenarios can be perfectly classified? If it can be correctly classified, **describe a setting of the weight vectors that would achieve perfect classification.** (There are 11 weights in total, one per feature. The weights are w_0 and then weights w_{iA} and w_{iB} for each $i = 1 \dots 5$.)

- (a) One or more good reviews leads to success. If the total of their scores is more than 8, then the movie will definitely be a success and otherwise it will fail.

 - (b) Mediocre reviews breeds success. Your movie will succeed if and only if each reviewer gives either a score of 2 or a score of 3.

 - (c) Your reviewers have complementary tastes. Your movie will succeed if and only if both reviewers agree.
6. For any above scenario above where the perceptron would fail, suggest a feature or features that could be added to ensure perfect classification.