

The Regression side of Supervised Learning

LATEST SUBMISSION GRADE
100%

1. What is the hypothesis space of linear regression?

1 / 1 point

☒ The set of flat hyperplanes

✓ **Correct**

Correct, in general linear regression considers all possible flat planes in the appropriate dimensionality.

☒ The set of straight lines.

✓ **Correct**

Correct, by default linear regression in two dimensions considers all possible straight lines.

☐ The best-fit line

☐ The set of curved lines.

☐ All hypothesis that give numbers instead of classes.

3. What's the problem with doing regression to find numeric class labels directly?

1 / 1 point

- ☐ You can't actually convert class labels to numbers.
- ☐ Regression doesn't work for binary values.
- ☐ Classifications are categorical rather than numeric values.
- ☒ It just works better to separate classes.
- ☐ Classification isn't convex.
- ☐ Transfer functions break loss functions.

✓ **Correct**

True! When you care about class labels, trying to fit a line (or hyperplane) to exactly recreate the classes is more difficult than finding a decision boundary.

4. Why might we not want our model to fit perfectly to our training data?

1 / 1 point

- ☐ A function with both local minima and global minima.
- ☒ For any two points on a graph, the line connecting the points is on or above the line of the graph.
- ☐ For any two points on a graph, the line connecting the points is on or below the line of the graph.
- ☐ A function with neither local minima or global minima.

✓ **Correct**

Correct! This is exactly the definition of a convex graph.

6. Why do we need iterative functions other than gradient descent to optimize loss functions?

1 / 1 point

- ☐ We don't need anything but gradient descent to optimize loss functions.
- ☒ Because not all loss functions are differentiable everywhere.
- ☐ Because the L2 loss function can have sharp corners.
- ☒ Randomness in training data

✓ **Correct**

Correct! Each time you sample new data from the same source and run a new analysis, creating a new model. Most phenomena are noisy, so the data will be different each time due to randomness.

☒ Model complexity

✓ **Correct**

Correct! A less complex model means higher bias and lower variance. A more complex model means lower bias and higher variance.

☐ Bad dart throwers

8. L1 and L2 regularizers penalize:

1 / 1 point

- ☐ The magnitude of training data
- ☐ The lambda parameter
- ☐ The distance between the line and the training data.
- ☒ The magnitude of weights in the loss function

✓ **Correct**

Correct! The magnitude of weights is a good proxy to measure complexity. Because we don't want our function to be too complex, we use a regularizer.