

Congratulations! You passed!

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| 1. | Which of these best describes unsupervised learning? | 1 / 1 point |
|----|--|-------------|
| | A form of machine learning that finds patterns using labeled data (x, y) | |
| | A form of machine learning that finds patterns using unlabeled data (x). | |
| | A form of machine learning that finds patterns in data using only labels (y) but without any inputs (x). | |
| | A form of machine learning that finds patterns without using a cost function. | |
| | Correct Unsupervised learning uses unlabeled data. The training examples do not have targets or labels "y". Recall the T-shirt example. The data was height and weight but no target size. | |
| | | |
| 2. | | 1/1 point |
| | Which of these statements are true about K-means? Check all that apply. | , . |
| | The number of cluster assignment variables $c^{(i)}$ is equal to the number of training examples. | |
| | | |
| | $c^{(i)}$ describes which centroid example (i) is assigned to. | |
| | $lacksquare$ If you are running K-means with $K=3$ clusters, then each $c^{(i)}$ should be 1, 2, or 3. | |
| | \bigcirc Correct $c^{(i)}$ describes which centroid example(i) is assigned to. If $K=3$, then $c^{(i)}$ would be one of 1,2 or 3 assuming counting starts at 1. | |
| | $lacksquare$ The number of cluster centroids μ_k is equal to the number of examples. | |
| | If each example x is a vector of 5 numbers, then each cluster centroid μ_k is also going to be a vector of 5 numbers. | |
| | ⊘ Correct | |
| | The dimension of μ_k matches the dimension of the examples. | |
| | | |
| 3. | | 1/1 point |
| | You run K-means 100 times with different initializations. How should you pick from the 100 resulting solutions? | |
| | Pick the last one (i.e., the 100th random initialization) because K-means always improves over time | |
| | Average all 100 solutions together. | |
| | lacksquare Pick the one with the lowest cost J | |
| | O Pick randomly that was the point of random initialization. | |
| | ✓ Correct K-means can arrive at different solutions depending on initialization. After running repeated trials, choose the solution with the lowest cost. | |
| | | |
| 4. | You run K-means and compute the value of the cost function $J(c^{(1)},\ldots,c^{(m)},\mu_1,\ldots,\mu_K)$ after each iteration. Which of these statements should be true? | 1/1 point |
| | Because K-means tries to maximize cost, the cost is always greater than or equal to the cost in the previous iteration. | |
| | The cost will either decrease or stay the same after each iteration | |
| | There is no cost function for the K-means algorithm. | |
| | ○ The cost can be greater or smaller than the cost in the previous iteration. but it decreases in the long run. | |

| 5. | In K-means, the elbow method is a method to | |
|----|---|--|
| | Choose the maximum number of examples for each cluster | |
| | Choose the best number of samples in the dataset | |
| | Choose the best random initialization | |
| | Choose the number of clusters K | |
| | Correct The elbow method plots a graph between the number of clusters K and the cost function. The 'bend' in the cost curve can suggest a natural value for | |

⊘ Correct

The cost never increases. K-means always converges.

 $\ensuremath{\mathsf{K}}.$ Note that this feature may not exist or be significant in some data sets.