## ▲ Try again once you are ready

Grade received 66.66% To pass 80% or higher

Try again

## **Dimensionality Reduction**

Total points 6

1. Fill in the blanks with the correct answer according to the descriptions in the boxes below:

0 / 1 point

Before... when it was all about \_\_\_\_\_1

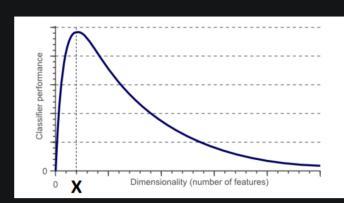
- Domain experts selected features
- Designed feature transforms
   Small number of more
   relevant features were
   enough
- Now... 2 is about integrating everything
  - Data generation and storage is less of a problem
  - Squeeze out the best from data
  - More high-dimensional data having more features
- 1. Data mining. 2. Data Science.
- 1. Dimensionality reduction. 2. Data Science.
- 1. Data mining. 2. Dimensionality reduction.
- 1. Data Science. 2. Data mining.

⊗ Incorrect

Sorry, but you're only half right. Blank 1 refers to the recent past in which data generation and data storage were a lot more costly, so the focus was on data mining. Today, nonetheless, data science involves more data transformation and processing steps such as dimensionality reduction.

2. What does the X value represent?

0 / 1 point



- The worst number of features for making predictions.
- The cursed number of dimensions.
- The optimal number of features.
- The number of features that reaches the maximum classification error.

⊗ Incorrec

Sorry, no. The curse of dimensionality is a broad concept that encompasses the many issues that occur when dealing with high-dimensional data, especially those regarding separability, sparsity, and the amount of data.

3. Which of the following are problems of high dimensionality in model performance? (Select all that apply)

1/1 point

Solutions take longer to reach global optimum

✓) Corn

Right on track! Very often, reaching a global optimum is a more difficult task when dealing with highdimensional problems.

	Smaller hypothesis space.
~	Higher runtimes and system requirements
	<ul> <li>Correct         Correct The more dimensions, the higher the system requirements. Therefore, dimensionality reduction helps optimize the system's performance.     </li> </ul>
~	The possibility of more correlated features is greater.
	<ul> <li>Correct         You've got it! When having more dimensions, it is possible to have more correlated features making the selection of the most relevant features a more difficult task.     </li> </ul>
a Wi	nat does the following line of code refer to? count_params(model_n.trainable_variables)
C	The number of testing parameters for Model n.
С	) The number of dimensions for Model n.
•	) The number of training parameters for Model n.
С	The number of classes for Model n.
	<ul> <li>Correct         That's right! This code line allows to count the number of training parameters for the input model.     </li> </ul>
	e amount of training data available, the complexity of the decision surface, and the classifier type define the mber ofto be used
С	) Spaces
С	) Models
C	) Datasets
•	) Features
	<ul> <li>Correct         That's right! These three aspects define the amount of features that will be used in a machine learning problem.     </li> </ul>
•	ue Or False: Classification subspaces allow to minimize separation among classes, while regression subspaces are ed for maximizing correlation between projected data and response variable.
С	) True
•	) False
	<ul> <li>Correct         That's right! Classification subspaces maximize the separation among classes, while regression intends to maximize the correlation between two variables.     </li> </ul>