Congratulations! You passed!

Grade received 100% **To pass** 80% or higher

Go to next item

Quantization and Pruning

Model customization

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To	tal points 7	
1.	True Or False: Today, due to developments in machine learning research, and performance improvements for mobile and edge devices, there exists a wide range of options to deploy a machine learning solution locally.	1 / 1 point
	○ False	
	True	
	Correct That's right! With mobile devices becoming increasingly more powerful and at the same time cheaper, these devices are now able to deploy machine learning solutions at the edge.	
2.	Which of the following benefits does machine learning provide to mobile & IoT businesses that use it? (Select all that apply)	1/1 point
	Automating operational efficiency.	
	Correct That's right! Mobile and IoT deployments can streamline your business and help you make accurate predictions. Also, the automation of some processes can decrease the time of information analysis, and therefore, can be crucial to improve operational efficiency.	
	☐ Eliminating risk.	
	Strengthening security.	
	Correct That's right! With ever increasing number of breaches and confidential data theft, companies want to strengthen their security. Employing ML in mobile and IoT security can help detect intrusions, protect your data, and respond to incidents automatically.	
	Improving user experience with data.	
	Correct That's right! Businesses with a mobile or IoT strategy know how technology can capture and transform data to offer greater access to consumer information and therefore devise better means to enhance their user experiences.	
3.	ML Kit brings Google's machine learning expertise to mobile developers. Which of the following are features of ML Kit? (Select all that apply)	1 / 1 point
	Access to cloud-based web services	
	Correct That's right! With ML, you can upload your models through the Firebase console and let the service take care of hosting and serving them to your app users.	
	On-device model training	

 [-127, 127], in range [-128, 127]. [-128, 127], equal to 0 [-128, 127], equal to 0 ○ Correct That's right! In per-tensor weights, there are two complement values in the range [-127, 127], with zero-point equal to 0 in approximates. 	scanning, and object detection, among others.	\bigcirc
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	ould be dynamic such as activations within	
6. True Or False: One family of optimizations, known as pruning, aims to remove neural network connections, increasing the number of parameters involved in the computation.	s, increasing the number of parameters involved in 1/1 point	
○ True		O Tr
False		● Fa
Correct That's right! The pruning optimization aims to eliminate neural network connections, but instead of increasing the number of parameters, you have to	increasing the number of parameters, you have to	_

7.	Which of the following describe the benefits of applying sparsity with a pruning routine? (Select all that apply)
	Can be used in tandem with quantization to get additional benefits
	Correct That's right! In some experiments, weight pruning is compatible with quantification, resulting in compounding benefits. Therefore, it is possible to further compress the pruned model by applying post-training quantization.
	▼ Better storage and/or transmission
	Correct That's right! An immediate benefit that you can get out of pruning is disk compression of sparse tensors. Thus, you can reduce the model's size for its storage and transmission by applying simple file compression to the pruned checkpoint.
	Method perform well at a large scale

That's right! You can even gain speeds in the CPU and ML throttles that fully exploit integer precision efficiencies in some cases.

Gain speedups in CPU and some ML accelerators

⊘ Correct

1/1 point

 $reduce\ them.\ With\ pruning,\ you\ can iower\ the\ overall\ parameter\ count\ in\ the\ network\ and\ reduce\ their\ storage\ and\ computational\ cost.$