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Exam Professional Machine Learning Engineer All Questions

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EXAM PROFESSIONAL MACHINE LEARNING ENGINEER TOPIC 1 QUESTION 37 DISCUSSIO..

Actual exam question from Google's Professional Machine Learning Engineer

Question #: 37

Topic #: 1

[All Professional Machine Learning Engineer Questions]

You are developing models to classify customer support emails. You created models with TensorFlow Estimators using small datasets on your on-premises system, but you now need to train the models using large datasets to ensure high performance. You will port your models to Google Cloud and want to minimize code refactoring and infrastructure overhead for easier migration from on-prem to cloud. What should you do?

- A. Use AI Platform for distributed training.
- B. Create a cluster on Dataproc for training.
- C. Create a Managed Instance Group with autoscaling.
- D. Use Kubeflow Pipelines to train on a Google Kubernetes Engine cluster.

Show Suggested Answer

by A salsabilsf at June 9, 2021, 11:49 a.m.

Comments

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	 ■ maartenalexander Highly Voted 3 years, 4 months ago A. Al platform provides lower infrastructure overhead and allows you to not have to refactor your code too much (no
	containerization and such, like in KubeFlow).
	A PhilipKoku Most Recent ② 5 months ago
	Selected Answer: A
	A) Al Platform
	upvoted 1 times
	dirgu 5 months, 1 week ago The most suitable option for minimizing code refactoring and infrastructure overhead while enabling large-scale training on
	Google Cloud is:
	A. Use AI Platform for distributed training. * **Simplified Workflow:** AI Platform offers a managed service for training machine learning models. You can train your existing TensorFlow Estimator code with minimal changes, reducing the need for extensive code refactoring. * **Distributed Training:** AI Platform automatically handles distributing your training job across multiple machines, allowing you to leverage the power of Google's cloud infrastructure to train on large datasets efficiently. * **Reduced Infrastructure Overhead:** You don't need to manage the underlying infrastructure (e.g., setting up and maintaining a cluster) yourself. AI Platform takes care of all the infrastructure provisioning and management, minimizing the workload on your team.
	♣ fragkris 11 months ago
	Selected Answer: A
	I chose A. Even though D is a working option, it requires us to create a GKE cluster, which requires more work.
	Sum_Sum 11 months, 3 weeks ago
	Selected Answer: A
	A - because it has native support for TF upvoted 1 times
	♣ harithacML 1 year, 3 months ago
	Selected Answer: A
	A. Use AI Platform for distributed training. : Managed , low infra change migration: yes , although need code refactoring to
	bigquery sql B. Create a cluster on Dataproc for training. : only cluster ? what about training?
	C. Create a Managed Instance Group with autoscaling. : Same Q?
	D. Use Kubeflow Pipelines to train on a Google Kubernetes Engine cluster : only training? upvoted 2 times
U	M25 1 year, 6 months ago Selected Answer: A
	Went with A
	- Fatiy 1 year, 8 months ago
	Selected Answer: A
	Option A is the best choice as AI Platform provides a distributed training framework, enabling you to train large-scale models
	faster and with less effort upvoted 1 times
	Mohamed_Mossad 2 years, 4 months ago
	Selected Answer: A
	using options eliminations answer between A,D will vote for A as it is easier
	i
	— monamos_moosas = years, e menane age
	- using options eliminations answer between A,D will vote for A as it is easier
	upvoted 1 times
	▲ David_ml 2 years, 6 months ago
	Selected Answer: A The answer is A. Al platform also contains kubeflow pipelines. you don't need to set up infrastructure to use it. For D you
	need to set up a kubernetes cluster engine. The question asks us to minimize infrastructure overheard.

upvoted 2 times

🗖 🏜 mmona19 2 years, 6 months ago

Selected Answer: D

D- Kubeflow pipelines with Vertex ai provides you ability to reuse existing code using a TF conatiner in a pipeline. it helps automate the process, there is a gwiklab walking through this.

A-incorrect, question is asking resuse existing code with minimum changes. distributed deployment does not address that.

upvoted 1 times

■ Lavid_ml 2 years, 6 months ago

The answer is A. AI platform also contains kubeflow pipelines. you don't need to set up infrastructure to use it. For D you need to set up a kubernetes cluster engine. The question asks us to minimize infrastructure overheard.

upvoted 2 times

□ ♣ A4M 2 years, 9 months ago

A - better to go with managed service and distributed

upvoted 2 times

DHEEPAK 2 years, 9 months ago

I am 100% sure that the answer is D. Kubeflow pipelines were designed keeping:

A) Portability.

- B) Composability.
- C) Flexibility in mind.

This is the pain point that the kubeflow pipelines address

upvoted 1 times

David_ml 2 years, 6 months ago

The answer is A. AI platform also contains kubeflow pipelines. you don't need to set up infrastructure to use it. For D you need to set up a kubernetes cluster engine. The question asks us to minimize infrastructure overheard.

upvoted 2 times

😑 📤 NamitSehgal 2 years, 10 months ago

Selected Answer: A

TensorFlow Estimators means require distributed and that is key feature for AI platform or later Vertex AI.

upvoted 3 times

■ JobQ 2 years, 10 months ago

I think is A

upvoted 1 times

🗖 🏜 q4exam 3 years, 1 month ago

I think the answer is either A or B, but personally think it is likely B because dataproc is a common tool box on GCP used for ML while AI platform might require refactoring. However, I dont really know A or B

upvoted 3 times

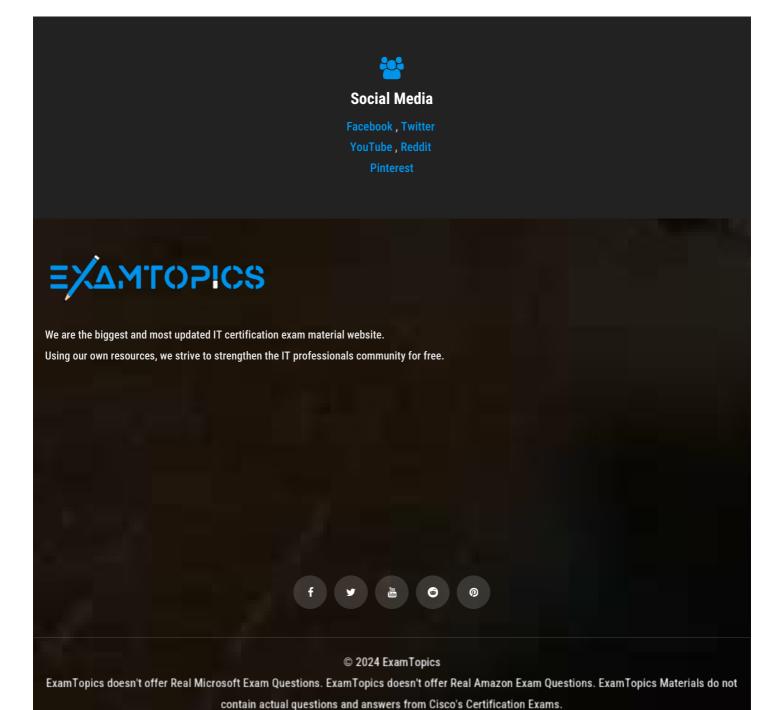
□ ♣ qeorqe_oqnyanov 3 years ago

Another vote for answer A. AI Platform distributed training here.

However, I wanted to share my logic why its not B as well. Dataproc is a managed Hadoop and as such needs a processing engine for ML tasks. Most likely Spark and SparkML. Now Spark code is quite different than pure Python and SparkML is even more different than TFcode. I imagine there might me a way to convert TF code to run on SparkML, but this seems a lot of work. And besides the question specifically wants us to minimize refactoring, so there you have it, we can eliminate option B 100%.

upvoted 5 times

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