

! Try again once you are ready

TO PASS 80% or higher



GRADE 79.16%

Distributed Strategy

Tatest submission grade 79.16%

1.	Whi	ich (of the following is <i>true</i> about training your model using data parallelism technique? Check all that are true.	0.75 / 1 point
		The	ne same model architectures are used on different machines, and each machine processes the entire data set.	
	✓		of the data is on 1 master machine, and copies of the data are then distributed to machines having different odel architectures based on their capacity of processing the data.	
		!	This should not be selected Incorrect! Data parallelism is meant to improve efficiency by not having to store or process all of the data on the same machine.	
	~	The	ne full data set is split up and subsets of the data are stored across multiple machines	
	,	~	Correct Correct! Data parallelism is meant to improve efficiency by not having to store or process all of the data on the same machine.	
	~	We	eights from different machines are aggregated and updated into a single model.	
	,	/	Correct Correct! All the learnings from training on multiple machines should be used to update a single model.	
2.	In T		sorFlow version 2, tf.distribute.Strategy class supports Check all that apply.	1/1 point
	•	/	Correct!	
	~	Gra	raph Mode	
	,	~	Correct!	
3.			of the following are <i>true</i> of both MirroredStrategy and TPU Strategy? Check all that are <i>true</i> . The same model is replicated on each core.	1 / 1 point
	,	~	Correct Correct! Both of these strategies use multiple cores on the same machine (either GPU for Mirrored Strategy or TPU for TPU strategy)	
	~	Vai	uriables are synchronized (mirrored) across each replica of the model	
	•	/	Correct Correct! Variables are mirrored across the copies of the model.	
		Us	ses multiple machines	

uses a single machine / Correct Correct! Both of these strategies use a single machine. 4. To modify training code to work with Mirrored Strategy, which of the following should we do? Choose all that apply. ✓ Put code that creates the model object inside the scope of "with strategy.scope(J". ✓ Correct Correct: the model creation code should be written within the scope of the strategy. Put the code that creates, compiles and fits the model inside the scope of "with strategy.scope()". Increase the batch size as long as the number is 2ⁿ (e.g. 64, 128, 256 etc). Adjust the batch size to equal the batch size per replica times the number of replicas ✓ Correct Correct! The batch size that the model can handle is now the number of examples that can be processed across all replicas of the model. 5. To modify training code to work with distributed data, which of the following should we do? Choose all that apply. Replace the code that updates the model weights (calculating loss, calculating gradients, and applying the gradients) so that each training step handles all replicas at once. ✓ Use *strategy.reduce* to aggregate the losses across the replicas. ✓ Correct Correct! After the replicas all train, update their weights, and return their losses, their losses are aggregated using strategy.reduce Use *strategy.run* to run the code that updates the model weights (calculating loss, calculating the gradients, and applying the gradients). / Correct Correct! Use strategy.run and pass in a function that contains the code which updates the model weights and returns the calculated loss. Use strategy.experimental_distribute_dataset to convert training and test sets into distributed datasets. / Correct Correct! 6. To use the TPU strategy, there are some steps that you'll take before running the training code. Please think about which line of code implements each step and choose the set of code that performs these steps in this order. 1 Get the TPU address 2 Find the TPU cluster 3 Connect to the TPU cluster 4 Initialize the TPU cluster 5 Create your TPU strategy • tpu_address = 'grpc://' + os.environ['COLAB_TPU_ADDR'] tf.distribute.cluster_resolver.TPUClusterResolver(tpu_address) tf.config.experimental_connect_to_cluster(tpu) tf.tpu.experimental.initialize_tpu_system(tpu) strategy = tf.distribute.experimental.MirroredStrategy(tpu)

```
1 strategy = tf.distribute.experimental.TPUStrategy(tpu)
2 tpu_address = 'grpc://' + os.environ['COLAB_TPU_ADDR']
3 tf.distribute.cluster_resolver.TPUClusterResolver(tpu_address)
4 tf.config.experimental_connect_to_cluster(tpu)
5 tf.tpu.experimental.initialize_tpu_system(tpu)
```

```
tpu_address = 'grpc://' + os.environ['COLAB_TPU_ADDR']
tf.config.experimental_connect_to_cluster(tpu)
tf.distribute.cluster_resolver.TPUClusterResolver(tpu_address)
tf.tpu.experimental.initialize_tpu_system(tpu)
strategy = tf.distribute.experimental.TPUStrategy(tpu)
```

tpu_address = 'grpc://' + os.environ['COLAB_TPU_ADDR']
tf.distribute.cluster_resolver.TPUClusterResolver(tpu_address)

tf.config.experimental_connect_to_cluster(tpu)

tf.tpu.experimental.initialize_tpu_system(tpu)

strategy = tf.distribute.experimental.TPUStrategy(tpu)

Incorrect

 $Incorrect!\ To\ create\ a\ TPU\ strategy,\ we'll\ use\ \textit{tf.distribute.experimental.TPUStrategy()}\ and\ not\ \textit{tf.distribute.experimental.MirroredStrategy()}$