

✔ Congratulations! You passed!

Grade
received 100%

Latest Submission
Grade 100%

To pass 80% or
higher

Go to next item

Retake the assignment in **7h 57m**

1. What are the three main components of MLflow?

1 / 1 point

- ☐ 1. Deployment
2. Models
3. Packaging
- ☐ 1. Experiments
2. Packaging
3. Deployment
- ☒ 1. Tracking
2. Projects
3. Models

✔ **Correct**
Correct!

2. What are three useful items that MLflow can track with the Tracking UI?

1 / 1 point

- ☒ 1. Parameters
2. Metrics
3. Artifacts

- ☐ 1. Performance
- 2. Accuracy
- 3. Parameters

- ☐ 1. Parameters
- 2. Performance
- 3. Artifacts

✓ **Correct**

Correct!

3. Where do MLflow runs get recorded?

1 / 1 point

- ☒ They get recorded either locally in files or remotely to a tracking server.
- ☐ They get recorded remotely to a tracking server which could also be Azure ML Studio.
- ☐ They get recorded locally in files in an **mlruns** directory wherever you ran your program.

✓ **Correct**

Correct!

4. What are three properties that are essential in MLflow projects?

1 / 1 point

- ☐ 1. Git Project
- 2. Dependencies
- 3. Entry points

- ☐ 1. Name
- 2. Parameters
- 3. Entry points

- ☒ 1. Name
- 2. Dependencies
- 3. Entry points

✓ **Correct**

Correct!

5. What are three main ways to interact with MLflow, including runs?

1 / 1 point

- ☐ 1. Command-line interface
- 2. AWS Sagemaker
- 3. Azure ML Studio
- ☐ 1. Scikit-learn
- 2. Projects
- 3. Tracking UI
- ☒ 1. Command-line interface
- 2. Python API
- 3. REST API

☒ **Correct**
Correct!

6. What is one benefit of using MLflow projects?

1 / 1 point

- ☐ It is a solid foundation for running distributed ML tasks
- ☐ That it provides a way to deploy a run on AWS Sagemaker
- ☒ That it provides a repeatable way to install and run ML tasks

☒ **Correct**
Correct!

7. What are three common components of an MLflow project?

1 / 1 point

- ☒ 1. **MLproject** file
- 2. A dependencies or env YAML file
- 3. One or more scripts to run tasks

- ☐ 1. A `pipeline.py` file
 - 2. A notebook (`ipynb`) file
 - 3. One or more scripts to run tasks
-
- ☐ 1. An **MLflow** file
 - 2. A `pipeline.py` file
 - 3. A dependencies or env YAML file

☒ **Correct**
Correct!

8. What is the right Python method to use to register an existing model with **mlflow** 1 / 1 point
? The choices below use an **onnx-gpt2** example model

☐ `from mlflow import create_registered_model`

`create_registered_model("onnx-gpt2")`

☐ `from mlflow import MlflowRegisterModel`

`client = MlflowClient()`
`MlflowRegisterModel(client, "onnx-gpt2")`

☒ `from mlflow import MlflowClient`

`client = MlflowClient()`
`client.create_registered_model("onnx-gpt2")`

☒ **Correct**
Correct!

9. What are the three stages supported for models in MLflow? 1 / 1 point

- ☒ 1. Staging
- 2. Production
- 3. Archived

- ☐ 1. Testing
2. Production
3. Deleted

- ☐ 1. Testing
2. Production
3. Archived

☒ **Correct**
Correct!

10. What happens with the MLflow Python library if you try to update a model version using `client.update_model_version()` if a previous version wasn't created?

1 / 1 point

- ☒ It fails with an exception because it is a requirement to have a previous version created
- ☐ The API calls succeeds but no actual version is updated. Internally MLflow sets the version to 0
- ☐ It creates a version 1 if no version exists. It isn't a requirements to have a previous version created.

☒ **Correct**
Correct!