

SPARK ML

1.Terraform Apply:

```
MINGW64:/c/Users/YuriiiHordiichuk/Desktop/m08_sparkml_python_azure-mas...
azurerm_databricks_workspace.bdcc: Still creating... [01m50s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m00s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m10s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m20s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m30s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m40s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [02m50s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [03m00s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [03m10s elapsed]
azurerm_databricks_workspace.bdcc: Still creating... [03m20s elapsed]
azurerm_databricks_workspace.bdcc: Creation complete after 3m29s [id=/subscriptions/316ab800-22b5-40ab-be41-8595de4fb2d4/resourceGroups/rg-dev-westeurope/providers/Microsoft.Databricks/workspaces/dbw-dev-westeurope-j2ta]
Releasing state lock. This may take a few moments...

Apply complete! Resources: 5 added, 0 changed, 0 destroyed.

Outputs:
resource_group_name = "rg-dev-westeurope-j2ta"

AzureAD+YuriiiHordiichuk@EPPLWROW0218 MINGW64 ~/Desktop/m08_sparkml_python/master/terraform
```

2.Baseline model result:

```
▶    ✓ 12:53 PM (2s)

model = mlflow.pyfunc.load_model(f"models:/{{model_name}}@production")

# Sanity-check: This should match the AUC logged by MLflow
print(f'AUC: {roc_auc_score(y_test, model.predict(X_test))}')

▶ (4) Spark Jobs

Downloading artifacts: 100% 9/9 [00:00<00:00, 21.4
AUC: 0.8540300975814177
```

Features importance:

```
feature_importances = pd.DataFrame(model.feature_importances_, index=X_train.columns.tolist(), columns=['importance'])
feature_importances.sort_values('importance', ascending=False)
```

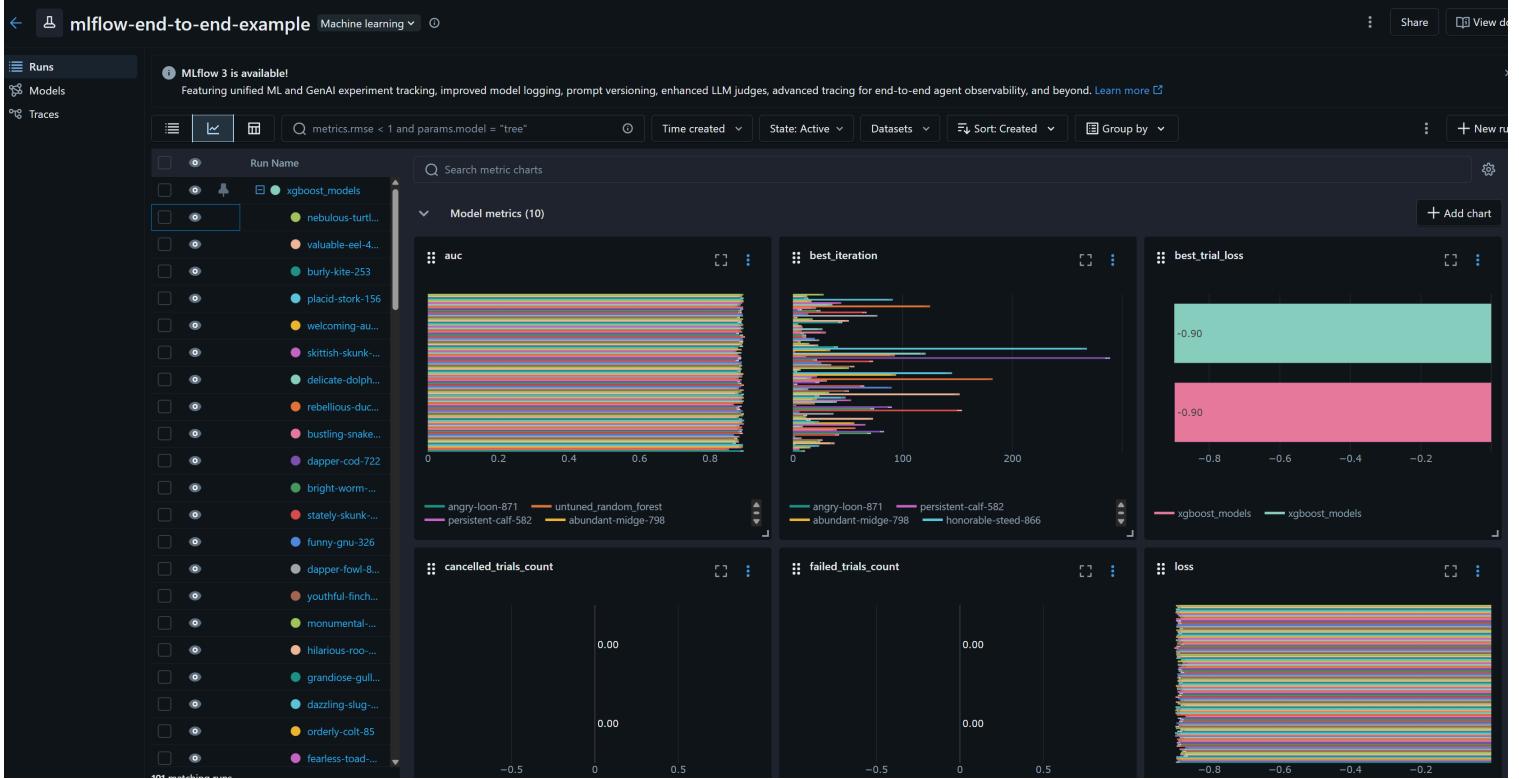
	importance
alcohol	0.160192
density	0.117415
volatile_acidity	0.093136
chlorides	0.086618
residual_sugar	0.082544
free_sulfur_dioxide	0.080473
pH	0.080212
total_sulfur_dioxide	0.077798
sulphates	0.075780
citric_acid	0.071857
fixed_acidity	0.071841
is_red	0.002134

Model available in models tab:

The screenshot shows the Databricks Catalog Explorer interface. On the left, the Catalog sidebar lists databases like 'Serverless Starter Warehouse', 'Serverless', and 'dbw_dev_westeuropa.j2ta'. Under 'dbw_dev_westeuropa.j2ta', the 'default' database is selected. The main area shows the 'Overview' tab for the 'default' schema, which is described as a 'Default schema (auto-created)'. Below this, there are tabs for 'Tables 0', 'Volumes 0', 'Models 2', and 'Functions 0'. The 'Models 2' tab is active, displaying a table with two rows. The first row is 'myregisteredmodel1' (Owner: ffwsw@gmail.com, Created at: Dec 26, 2025, 01:49 PM). The second row is 'wine_quality' (Owner: ffwsw@gmail.com, Created at: Dec 26, 2025, 12:53 PM). The 'wine_quality' row is highlighted with a red box.

Name	Owner	Created at
myregisteredmodel1	ffwsw@gmail.com	Dec 26, 2025, 01:49 PM
wine_quality	ffwsw@gmail.com	Dec 26, 2025, 12:53 PM

3. New model xgboost_models experiments:



Version 2 with better results added:

The screenshot shows the Databricks catalog interface. The top navigation bar includes 'Catalog Explorer', 'databricks-dev-west-europe-j20', and 'default'. Below the navigation, there's a card for the 'wine_quality' dataset, which has a star icon and a 'spark' icon.

The main content area has tabs for 'Overview', 'Details', and 'Permissions'. The 'Overview' tab is active. It contains a 'Description' section with an 'AI generate' button and an 'Add' button. Below this is a 'Versions' section with a table:

Status	Version	Tags	Aliases	Deployment job s...	Active endpoints	Comment
✓	Version 2		@ production		model-test	
✓	Version 1		@ archived			

Below the table is an 'Activity log' section. It says 'Activity logs capture updates from deployment jobs. Learn more.' followed by a table:

Time	Version	Log
2025-12-26 13:00:19	Version 2	Model version 2 registered by user ffwsdw@gmail.com.
2025-12-26 12:53:55	Version 1	Model version 1 registered by user ffwsdw@gmail.com.

4. Server thee model. I rewrote this cell to successfully log the model:

▶ ✓ 01:49 PM (9s) 53

```

import mlflow
from mlflow.models.signature import infer_signature

model_name = "wine_quality"
X_sample = X_train.iloc[:10]

# Load the model to infer signature
model = mlflow.pyfunc.load_model(f"models:/'{model_name}@production")
y_pred = model.predict(X_sample)
signature = infer_signature(X_sample, y_pred)

with mlflow.start_run():
    mlflow.sklearn.log_model(
        sk_model=model._model_impl, # Access the underlying sklearn model
        artifact_path="my_model",
        registered_model_name="MyRegisteredModel1",
        signature=signature,
        input_example=X_sample
    )

```

▶ (8) Spark Jobs

▼ (1) MLflow run
Logged 1 run to an experiment in MLflow. Learn more

Downloading artifacts: 100% [██████████] 9/9 [00:00<00:00, 30.95it/s]

/databricks/python/lib/python3.11/site-packages/mlflow/types/utils.py:435: UserWarning: Hint: Inferred schema contains integer column(s). Integer columns in Python cannot represent missing values. If your input data contains missing values at inference time, it will be encoded as floats and will cause a schema enforcement error. The best way to avoid this problem is to infer the model schema based on a realistic data sample (training dataset) that includes missing values. Alternatively, you can declare integer columns as doubles (float64) whenever these columns may have missing values. See 'Handling Integers With Missing Values' <<https://www.mlflow.org/docs/latest/models.html#handling-integers-with-missing-values>> for more details.

warnings.warn(

Uploading artifacts: 100% [██████████] 11/11 [00:00<00:00, 14.40it/s]

Successfully registered model 'dbw dev westeurope j2ta.default.myregisteredmodel1'.

I created a served endpoint. Access token and model URL obtained:

Serving endpoints >

test-model-name

Ready https://adb-7405616284574485.5.azure.databricks.net/serving-endpoints/test-model-name/invocations

Edit AI Gateway

Gateway

Usage monitoring: system.serving.endpoint_usage

Dimension table: system.serving.served_entities

Inference tables: Not enabled

Rate limits: Not enabled

About this endpoint

Created by Yuri Hordiichuk

Route Not enabled

optimization

Tags

Add

Serverless usage policy

Active configuration

Entity	Version	Name	State	Compute	Traffic (%)	Description
dbw.dev.westeurope.j2ta.default.myregisteredmodel1	Version 1	myregisteredmodel1-1	Ready	CPU, Custom 0-4 concurrency (0-4 DBU)	100	No description

5. The model prediction from the endpoint agree with the result from local model:

The model predictions from the endpoint should agree with the results from locally evaluating the model.

```
▶ ✓ Just now (<1s) 60
# Model serving is designed for low-latency predictions on smaller batches of data
num_predictions = 5
served_predictions = score_model(X_test[:num_predictions])
model_evaluations = model.predict(X_test[:num_predictions])

# Compare the results from the deployed model and the trained model
pd.DataFrame({
    "Model Prediction": model_evaluations,
    "Served Model Prediction": np.array(served_predictions['predictions'], dtype=np.float32),
})
}
```

Model Prediction	Served Model Prediction
0	0.000269
1	0.003019
2	0.020137
3	0.044904
4	0.019414

Request appeared on the chart:

