

# Experiment Tracking and Model Management

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- \* Introduction to experiment tracking and model management
- \* MlFlow :- Tracking of experiment  
logging / recording all experiment
- \* Why login is important ?  
- Organization of production pipeline properly
- \* MlFlow interface  
logging of experiments  
- MlFlow interface helping us for
  - 1) What is the accuracy ?
  - 2) What is the score you getting ?
  - 3) What are hyperparameters used ?
 Visualise or see what all experiments conducted.

```

      graph TD
      MlFlow --> Tracking[Tracking of experiment]
      MlFlow --> Model[Model Management]
    
```

Terminology :-

- Experiment Run :- each trial of some experiment
- Metadata for each experiment run :- every information related to an experiment run
  - Hyperparameters used
  - train size and test size
  - Data used
  - Algorithm used
  - Score of model
- Artifacts :- output file from experiment run (e.g. .pkl file, .csv file)

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- \* Why tracking :-
  - Organization and optimization much simple
  - Reproducibility
- \* installation :-
  - pip install mlflow
  - mlflow ui  
↓ go to ↓  
127.0.0.1:5000 (User interface)
  - cmd with file location (.py)  
mlflow ui --backend-store-uri sqlite:///mlflow.db
- \* MlFlow Experiment tracking

```

      graph TD
      MlFlow --> Experiment[Experiment tracking]
      MlFlow --> Model[Model Management]
      Experiment --> Run[Experiment RUN]
      Experiment --> Meta[Meta data]
      Experiment --> Artifact[Artifact]
      Model --> Registry[model registry]
      Registry --> Store[storing the model]
    
```

step 1 :- import mlflow, mlflow

step 2 :- mlflow.set\_tracking\_uri('Database Uri') {tracking of 'sqlite:///mlflow.db' exp name}

mlflow.set\_experiment('Experiment\_Name')

hp = 0.01

model = LogisticRegression(c = hp)

artifact = model.fit(X\_train, y\_train) ← metadata

y\_test\_prediction = model.predict(X\_test)

metadata = acc = metrics.accuracy\_score(y\_test, y\_test\_prediction)

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MlFlow ↔ NEPTUNE ↔ W&B (Weight and Biases)

free      paid      paid

step 3 Start experiment run :-

> with mlflow.start\_run():

track Metadata (set tag) :-

```

mlflow.set_tag("key", "value")
      fun1   key   value
mlflow.set_tag("Algo", "Logit")
      key   value
    
```

mlflow.log\_param("c", hp)

mlflow.log\_param("data\_path", "data/iris.csv")

mlflow.log\_metric("Accuracy", variable)

mlflow.sklearn.log\_model("Variable name of model", Path)

↓

framework ⇒ TensorFlow, keras, pytorch

= Alternate way to log model

> mlflow.log\_artifact("local-path", artifact\_path="model")

### Step 1 - Import MLFlow

```
import mlflow
```

### Step 2 - Set the tracker and experiment

```
mlflow.set_tracking_uri(DATABASE_URI)  
mlflow.set_experiment("EXPERIMENT_NAME")
```

### Step 3 - Start a experiment run

```
with mlflow.start_run():
```

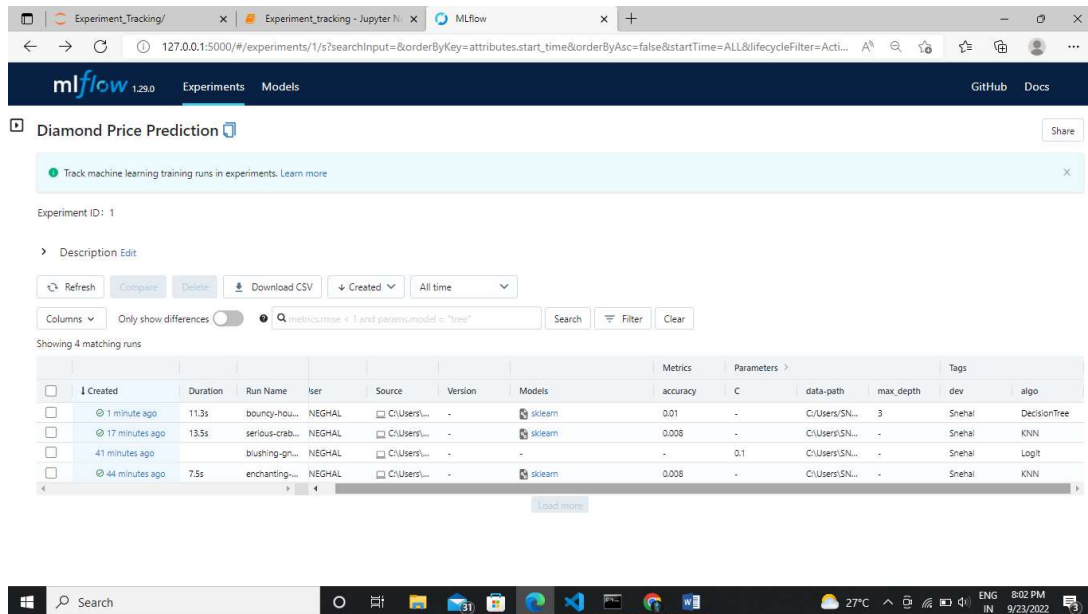
### Step 4 - Logging the metadata

```
mlflow.set_tag(KEY, VALUE)  
mlflow.log_param(KEY, VALUE) mlflow.log_metric(KEY, VALUE)
```

### Step 5 - Logging the model and other files (2 ways)

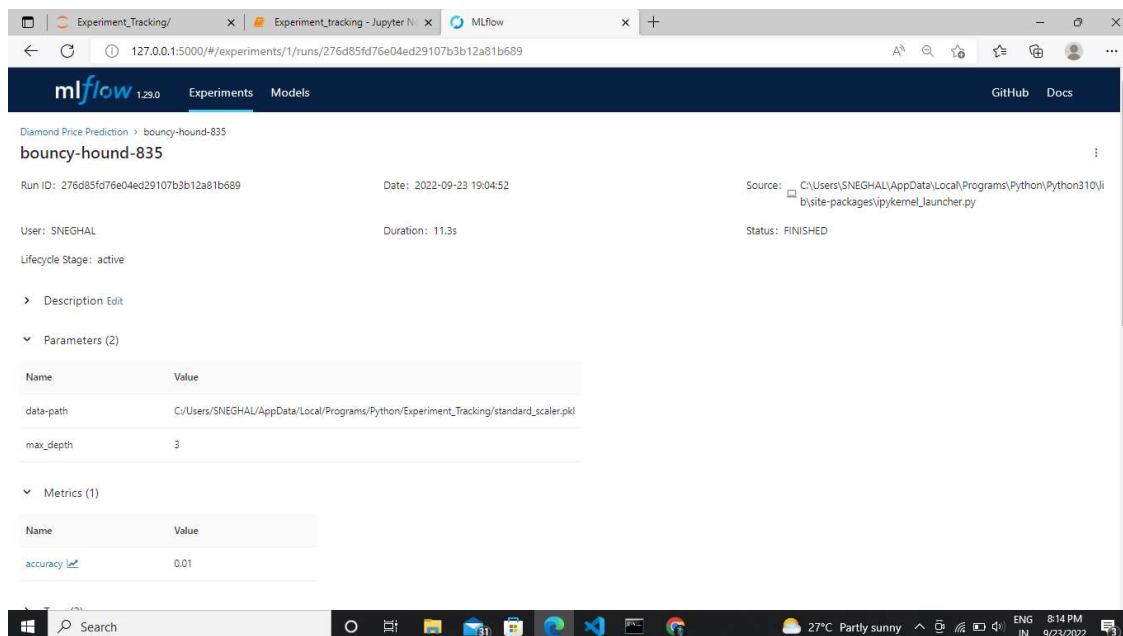
```
Way 1 - mlflow.<FRAMEWORK>.log_model(MODEL_OBJECT, artifact_path="PATH")  
Way 2 - mlflow.log_artifact(LOCAL_PATH, artifact_path="PATH")
```

## MLflow Interface for Tracking Experiments



The screenshot shows the MLflow web interface for the 'Diamond Price Prediction' experiment. The top navigation bar includes 'Experiments' and 'Models' tabs. Below the header, there's a 'Share' button and a message: 'Track machine learning training runs in experiments. Learn more'. The 'Experiment ID: 1' is displayed. A 'Description Edit' section is visible with buttons for 'Refresh', 'Compare', 'Delete', 'Download CSV', and a dropdown for 'Created' (set to 'All time'). A search bar contains the query 'metrics.rmse < 1 and params.model = "tree"'. Below this, a table lists 4 matching runs. The table has columns for 'Created', 'Duration', 'Run Name', 'User', 'Source', 'Version', 'Models', 'Metrics', 'Parameters', and 'Tags'. The runs are sorted by 'Created' time.

Created	Duration	Run Name	User	Source	Version	Models	Metrics	Parameters	Tags
1 minute ago	11.3s	bouncy-hou...	NEGHAL	C:\Users\...	-	sklearn	accuracy: 0.01	-	C:\Users\SN... 3 Snehali DecisionTree
17 minutes ago	13.5s	serious-cra...	NEGHAL	C:\Users\...	-	sklearn	0.008	-	C:\Users\SN... Snehali KNN
41 minutes ago		blushing-gr...	NEGHAL	C:\Users\...	-	-	-	0.1	C:\Users\SN... Snehali Logit
44 minutes ago	7.5s	enchanted-...	NEGHAL	C:\Users\...	-	sklearn	0.008	-	C:\Users\SN... Snehali KNN

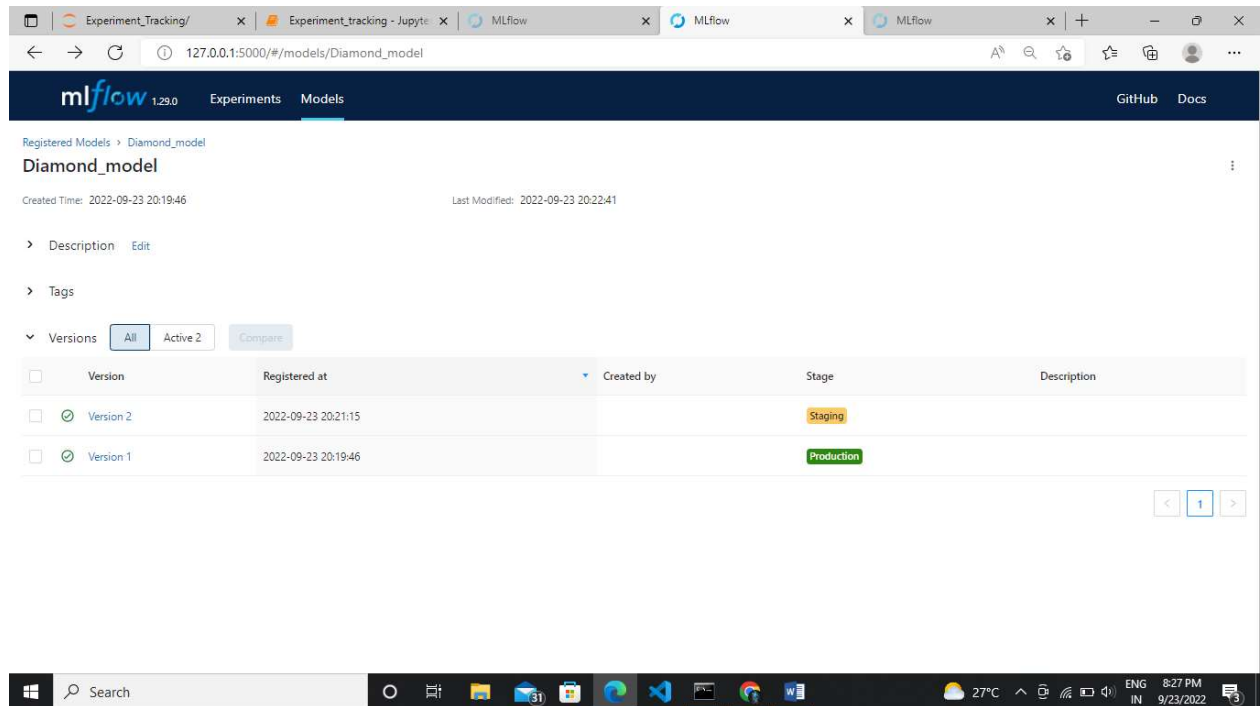


The screenshot shows the MLflow web interface for a specific run named 'bouncy-hound-835'. The top navigation bar includes 'Experiments' and 'Models' tabs. Below the header, the run details are displayed. The 'Run ID' is '276d85fd76e04ed29107b3b12a81b689'. The 'Date' is '2022-09-23 19:04:52'. The 'Source' is 'C:\Users\SNEGHAL\AppData\Local\Programs\Python\Python310\lib\site-packages\ipykernel\_launcher.py'. The 'User' is 'SNEGHAL'. The 'Duration' is '11.3s'. The 'Status' is 'FINISHED'. The 'Lifecycle Stage' is 'active'. A 'Description Edit' section is visible. Below this, the 'Parameters (2)' are listed in a table. The 'Metrics (1)' are also listed in a table.

Name	Value
data-path	C:\Users\SNEGHAL\AppData\Local\Programs\Python\Experiment_Tracking\standard_scaler.pkl
max_depth	3

Name	Value
accuracy	0.01

## MLflow Interface for Model Management



Registered Models > Diamond\_model

### Diamond\_model

Created Time: 2022-09-23 20:19:46 Last Modified: 2022-09-23 20:22:41

> Description [Edit](#)

> Tags

▼ Versions [All](#) [Active 2](#) [Compare](#)

<input type="checkbox"/>	Version	Registered at	Created by	Stage	Description
<input type="checkbox"/>	✓ Version 2	2022-09-23 20:21:15		Staging	
<input type="checkbox"/>	✓ Version 1	2022-09-23 20:19:46		Production	

< 1 >

Windows Search 27°C 8:27 PM 9/23/2022