



Nawatech

Crafting Experience Beyond Engineering

nawatech.co



Day 1

Prepared for: HAGI



Explore the Azure Machine
Learning workspace
resources and assets





Introducing Azure Machine Learning

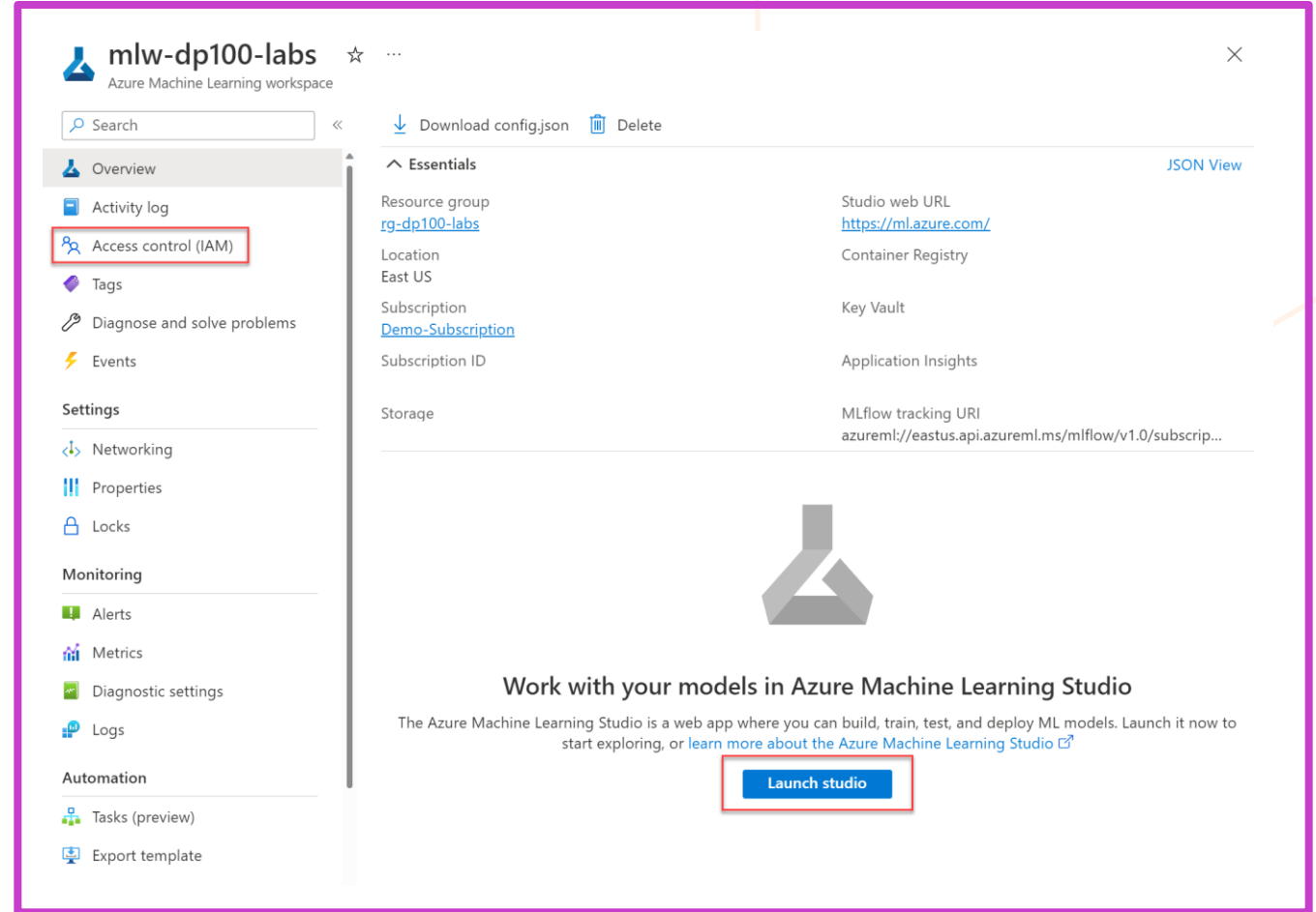
Azure Machine Learning provides a platform for data scientists to train, deploy, and manage their machine learning models on the Microsoft Azure platform.

It provides a comprehensive set of resources and assets to train and deploy effective machine learning models.



Explore the workspace in the Azure portal

- Give others access to the Azure Machine Learning workspace, using the **Access control**.
- Launch the **Azure Machine Learning studio**, an easy-to-use interface to create, manage, and use resources and assets in the workspace.





Identify Azure Machine Learning resources

1

- The workspace – The top-level resource for Azure Machine Learning. The workspace keeps an overview of all logs, metrics, outputs, models, and snapshots of your code.

2

- Compute resources – There are five types of compute in the Azure Machine Learning workspace: compute instances, compute clusters, Kubernetes clusters, attached computes, and serverless compute.

3

- Datastores – All data is stored in datastores, which are references to Azure data services. Four datastores will exist by default.



Identify Azure Machine Learning assets

1

- Models – Save trained models in the workspace. A common way to store such models is to package the model as a Python pickle file (.pkl extension).

2

- Environments – Specify software packages, environment variables, and software settings to run scripts. An environment is stored as an image in the Azure Container Registry created with the workspace when it's used for the first time.

3

- Data – You can use data assets to easily access data every time, without having to provide authentication every time you want to access it.

4

- Components – Make it easier to share code with component in a workspace.



Explore algorithms and hyperparameter values with Automated Machine Learning

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Create a new Automated ML job

✓

Select data asset

✓

Configure job

●

Select task and settings

○

Hyperparameter configuration
(Computer Vision only)

○

Validate and test

📊

Classification

To predict one of several categories in the target column. yes/no, blue, red, green.

📈

Regression

To predict continuous numeric values.

🕒

Time series forecasting

To predict values based on time.

✓

The time series forecasting method requires some additional information.

Time column * ⓘ
WeekStarting (Date) ▾

Time series identifier(s) ⓘ
Autodetect

)

Frequency * ⓘ
Autodetect

Forecast horizon * ⓘ
Autodetect

☐ Enable deep learning ⓘ

Back

Next

Cancel



Run a notebook

The screenshot displays the Azure Machine Learning Studio interface. On the left, the 'Notebooks' sidebar shows a file explorer with a tree structure: Users > madiepev > azure-ml-labs > Instructions > Labs > 01 > src. The file 'Run training script.ipynb' is selected. The main area shows the notebook content, which includes a title 'Run a training script with the Python SDK', a description, a 'Before you start' section, a code cell, and a 'Connect to your workspace' section.

Notebooks

Files Samples

Users

madiepev

azure-ml-labs

Instructions

Labs

01

src

Run training script.ipynb

02

03

04

05

06

07

08

09

10

Run training script.ipynb X

Edit in VS Code (pr...) Compute instance: ci11... No ...

Viewing Last saved a few seconds ago

Run a training script with the Python SDK

You can use the Python SDK for Azure Machine Learning to submit scripts as jobs. By using jobs, you can easily keep track of the input parameters and outputs when training a machine learning model.

Before you start

You'll need the latest version of the **azureml-ai-ml** package to run the code in this notebook. Run the cell below to verify that it is installed.

Note: If the **azure-ai-ml** package is not installed, run `pip install azure-ai-ml` to install it.

+ Code + Markdown

```
1 pip show azure-ai-ml
```

Press shift + enter to run

Connect to your workspace

With the required SDK packages installed, now you're ready to connect to your workspace.

To connect to a workspace, we need identifier parameters - a subscription ID, resource group name, and workspace name. The resource group name and workspace name are already filled in for you. You only need the subscription ID to complete the command.

To find the necessary parameters, click on the subscription and workspace name at the top right of the Studio. A pane will open on the right.



Run a script as a job

- When you submit a job to the workspace, all inputs and outputs will be stored in the workspace.
- There are different types of jobs:
 - **Command:** Execute a single script.
 - **Sweep:** Perform hyperparameter tuning when executing a single script.
 - **Pipeline:** Run a pipeline consisting of multiple scripts or components.

The screenshot displays the Azure ML Studio interface for a job named 'diabetes-train-mlflow'. The breadcrumb navigation at the top shows the path: Microsoft Non-Production > mlw-dp100-labs > Jobs > diabetes-training > diabetes-train-mlflow. The job status is 'Completed' with a green checkmark icon. Below the job name, there are tabs for Overview, Metrics, Images, Child jobs, Outputs + logs, and Code. The Overview tab is active, showing a toolbar with icons for Refresh, Connect to compute, Edit and submit, Register model, Cancel, and Delete. The main content area is divided into three panels: Properties, Inputs, and Outputs. The Properties panel on the left lists job details: Status (Completed), Created on (Nov 4, 2022), Start time (Nov 4, 2022), Duration (1m 36.99s), Compute duration (1m 36.99s), Name (yellow_head), Command (python train.py --training_data \${inputs.diabetes_data} --reg_rate \${inputs.reg_rate}), Created by, Job type (Command), and Experiment (diabetes-training). The Inputs panel on the right shows reg_rate (0.01) and Input name (diabetes_data) with Data (diabetes-data:1). The Outputs panel shows Output name (mlflow_log_model) and Model (azureml_yellow_head_1729586584:1). The Tags panel shows estimator_class (sklearn.linear_model_logistic.LogisticRegression) and estimator_name (LogisticRegression) with model_type (LogisticRegression). The Params panel shows various hyperparameters: C (100.0), class_weight (None), dual (False), fit_intercept (True), intercept_scaling (1), l1_ratio (None), max_iter (100), multi_class (auto), n_jobs (None), penalty (l2), random_state (None), Regularization rate (0.01), solver (liblinear), and tol (0.0001).

Make data available in Azure Machine Learning

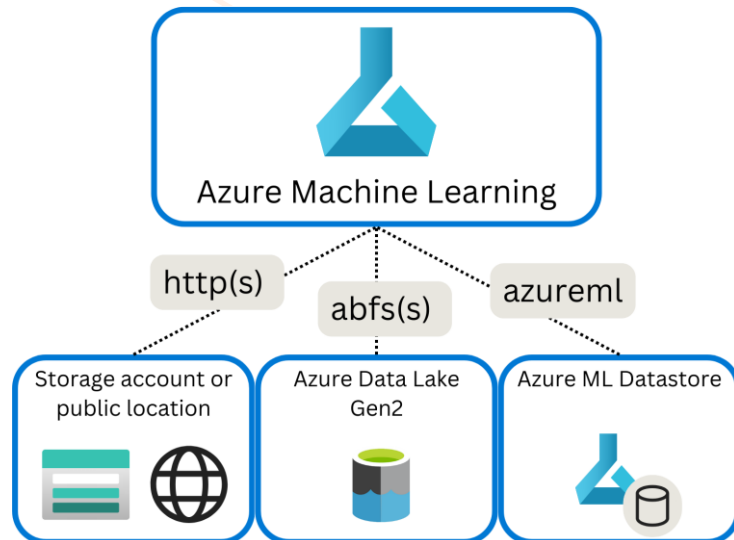




Understand URIs

A URI references the **location of your data**.

For Azure Machine Learning to connect to your data directly, you need to prefix the URI with the appropriate protocol.





Understand datastores

Datastores are **abstractions** for cloud data sources, storing the **connection information**.

The benefits of datastores:

- Provide easy-to-use URLs to your data storage.
- Facilitates data discovery within Azure Machine Learning.
- Securely stores connection information, without exposing secrets and keys to data scientists.



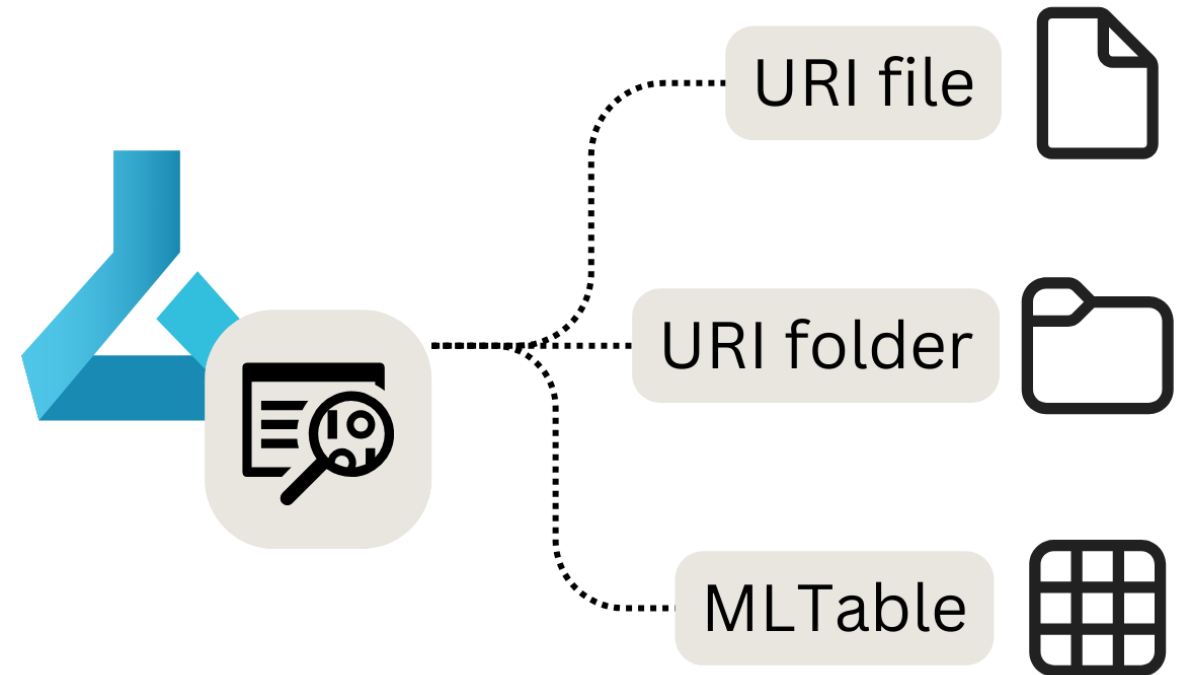


Understand data assets

Data assets are **references** to where the data is stored, how to get access, and any other relevant metadata.

The benefits of data assets:

- Share and reuse data with other members.
- Seamlessly access data during model training (on any supported compute type) without worrying about connection strings or data paths.
- Version the metadata of the data asset.

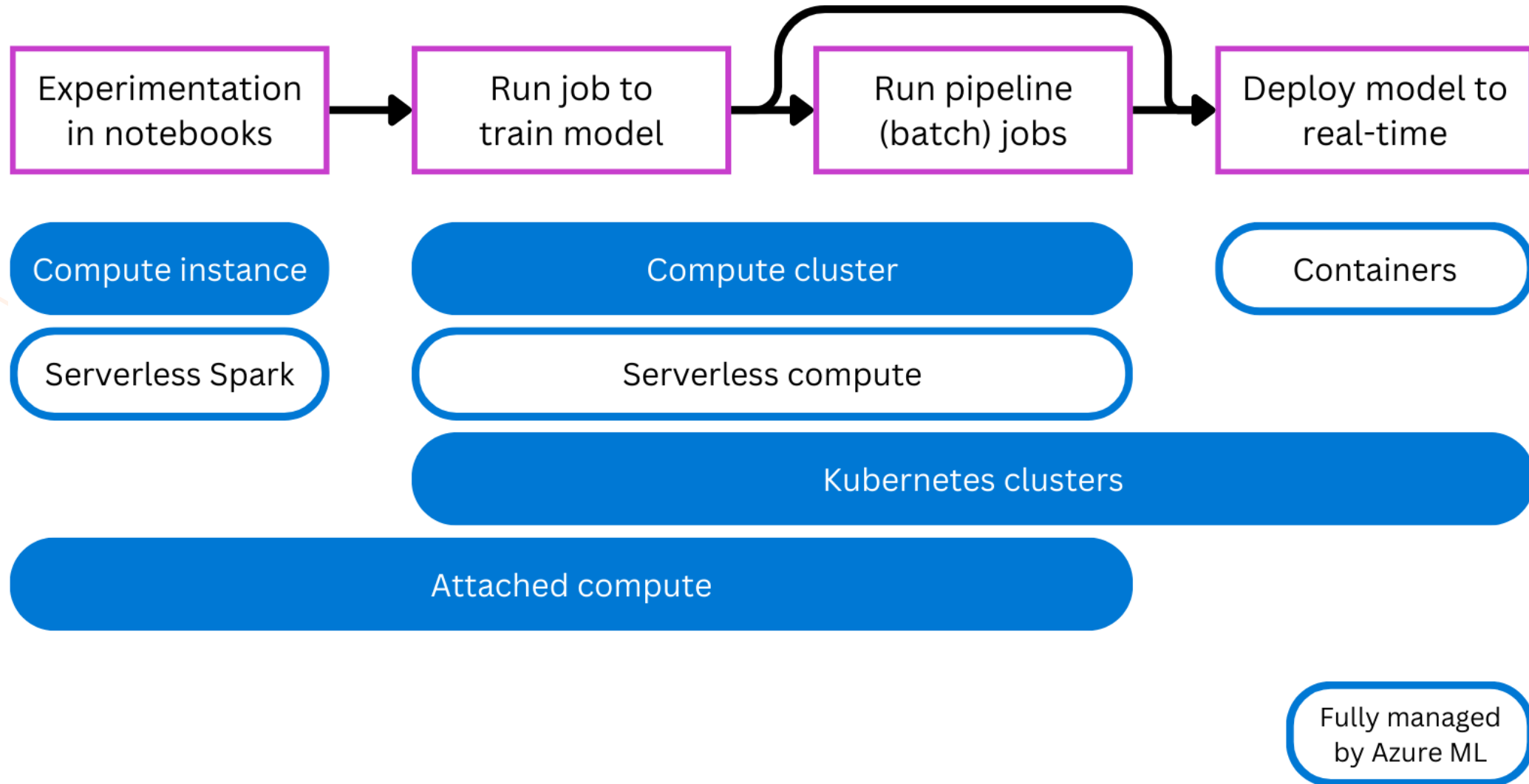


Work with compute resources in Azure Machine Learning





Choose the appropriate compute target



Automated Machine Learning





Explore algorithms and hyperparameter values with Automated Machine Learning

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Create a new Automated ML job

✓ Select data asset

✓ Configure job

● Select task and settings

○ Hyperparameter configuration (Computer Vision only)

○ Validate and test

📊 Classification
To predict one of several categories in the target column. yes/no, blue, red, green.

📈 Regression
To predict continuous numeric values.

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To predict values based on time. ✓

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Frequency * ⓘ
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Forecast horizon * ⓘ
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☐ Enable deep learning ⓘ

Back

Next

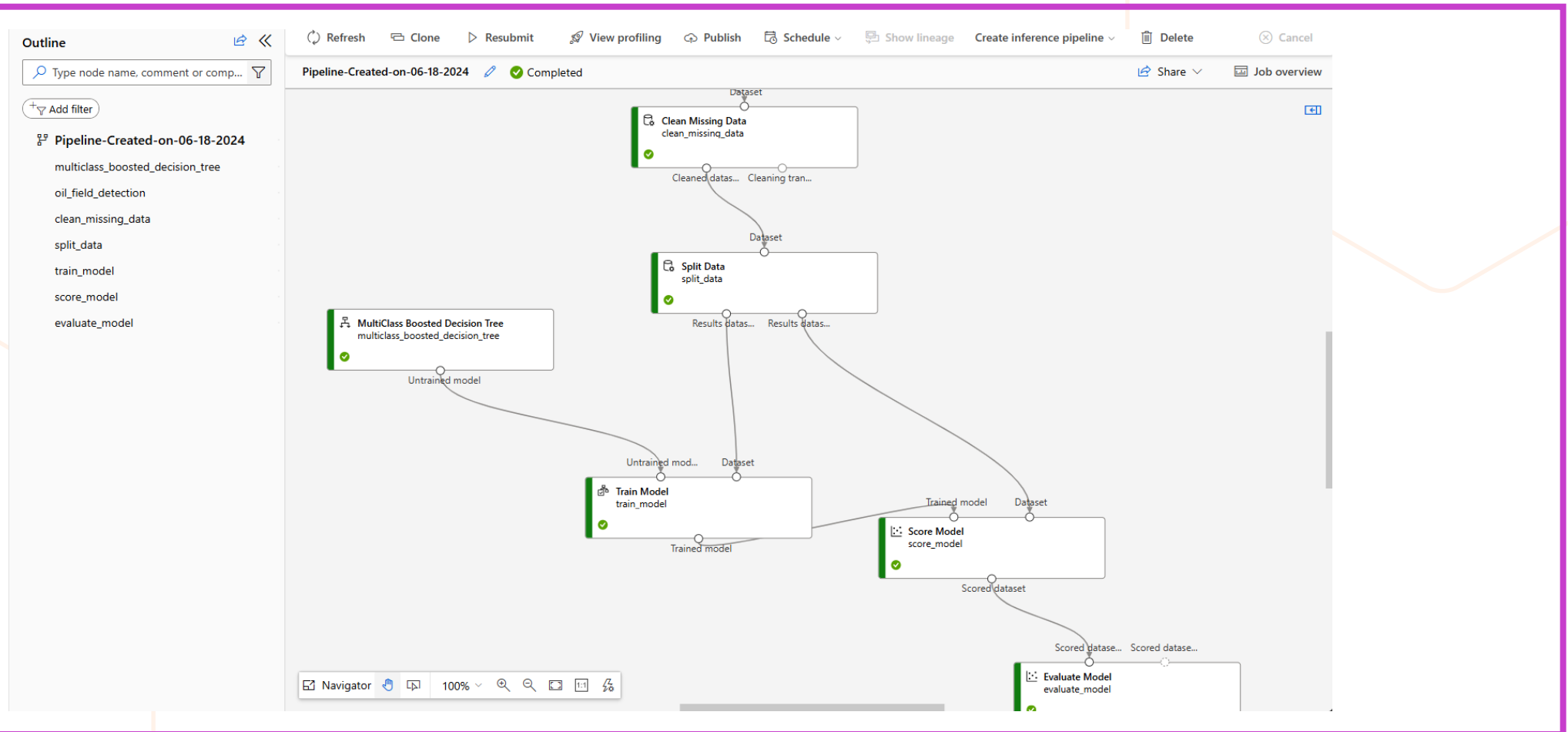
Cancel

Azure Machine Learning Designer





Create pipeline designer



Azure Machine Learning Notebook





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Environment Setup: Azure Machine Learning





Environment Setup



#1 Compute

Prerequisites: Make sure your account already has access and can sign in to the Azure Machine Learning workspace.

Navigate to Compute Instance and Compute Cluster.

Create Compute Instance/Cluster.

Set configuration compute.

#2 Kernel, ML version

Prerequisites: Make sure to having compute instance and some notebook.

Start compute and open a notebook.

Choose kernel based on use case.

Exercise: Azure Machine Learning Hands On





USECASE

Geothermal Gradient Prediction



- Objectives :

This Project is aimed at predicting geothermal characteristics for Colombia, specifically the geothermal gradient, based on available geological and geophysical data. We employ machine learning techniques to make these predictions. This project focuses on predicting the Apparent Geothermal Gradient ($^{\circ}\text{C}/\text{Km}$) as an essential factor in geothermal exploration. The code and the results are in Model.ipynb.

- Methodology :

The project utilizes geospatial data, geophysical information, and geothermal measurements. These datasets are located in the data folder of this repository. The data includes information on well depths, temperatures, geological features, and proximity to volcanic structures.













USECASE

Geothermal Gradient Prediction

- Machine Learning Canvas

THE MACHINE LEARNING CANVAS (V1.1) Designed for: _____ Designed by: _____ Date: _____ Iteration: _____

PREDICTION TASK  <small>Entity on which predictions are made? Possible outcomes? Wait time before observation?</small> Predict the geothermal characteristics especially in gradient based on geophysical and geological data.	DECISIONS  <small>How are predictions turned into proposed value for the end-user? Mention parameters of the process / application that does that.</small> Geothermal is one of essential factors in geothermal exploration.	VALUE PROPOSITION  <small>Who is the end-user? What are their objectives? How will they benefit from the ML system? Mention workflow/interfaces.</small> The end user is the geothermal industry actors (geologist, geophysics, etc.)	DATA COLLECTION  <small>Strategy for initial train set & continuous update. Mention collection rate, holdout on production entities, cost/constraints to observe outcomes.</small> The data is included information of geological, geophysical (well depths, temperatures, proximity to volcanic structures)	DATA SOURCES  <small>Where can we get (raw) information on entities and observed outcomes? Mention database tables, API methods, websites to scrape, etc.</small> Dataset can be obtained from Geothermal Power Database (International Geothermal Association)
IMPACT SIMULATION  <small>Can models be deployed? Which test data to assess performance? Cost/gain values for (in)correct predictions? <u>Fairness constraint</u>?</small> Model can be deployed as restapi, azure ml ops. So it can be used for further application (web/mobile app)	MAKING PREDICTIONS  <small>When do we make real-time / batch pred.? Time available for this + featurization + post-processing? Compute target?</small> We make the batch prediction since it's not time series data. And the compute target can be compute instance, compute cluster or kubernetes	BUILDING MODELS  <small>How many prod models are needed? When would we update? Time available for this (including featurization and analysis)?</small> we can use any regression models as long as the model give an accurate result.		
MONITORING  <small>Metrics to quantify value creation and measure the ML system's impact in production (on end-users and business)?</small> Models can be monitored using mlflow.		FEATURES  <small>Input representations available at prediction time, extracted from raw data sources.</small>		

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USECASE

Geothermal Gradient Prediction using Notebook

Method :

- Open Azure ML Notebooks
- Connect to active compute instance and kernel
- Install all library required
- Connect your notebooks to data assets

The screenshot displays the Azure ML Notebook environment. On the left, a sidebar contains navigation options: All workspaces, Home, Model catalog, Authoring (Notebooks, Automated ML, Designer, Prompt flow, Tracing), Assets (Data, Jobs, Components, Pipelines, Environments, Models, Endpoints), Manage (Compute, Monitoring, Data Labeling, Linked Services, Connections), and Data Store. The main area is divided into a 'Files' pane on the left showing a directory structure for 'Geothermal-Gradient-Ma' (including folders like '_pycache_', 'DICOM-Visualization', 'Feature-Positioning-Cont', 'data', 'deploy-ami', and files like '.gitignore', 'data_pre_norm.csv', 'data_weights.csv', 'full_prediction_points.c', 'Model_V4.ipynb', 'normalized_data_minrr', 'predicted_dataset.csv', 'predicted_new.csv', 'README.md', 'requirements.txt', 'trained_model.pkl', 'lightning_logs', 'SeismicHoloviz-copy', and 'cekipynb.ipynb'), and a central code editor. The code editor shows a Jupyter Notebook cell with the title 'Import Necessary Libraries' containing 11 lines of Python code to import various libraries. Below the code, a terminal window displays a FutureWarning message about glibc version and a note about XGBoost features. The right side of the interface shows the 'Compute Instance' dropdown set to 'Python 3.8 - AzureML' and a 'Kernel' dropdown set to 'Python 3.8 - AzureML'.

```
1 import pandas as pd
2 from sklearn.model_selection import train_test_split
3 import xgboost as xgb
4 import numpy as np
5 import geopandas as gpd
6 from tqdm import tqdm
7 import rasterio
8 import matplotlib.pyplot as plt
9 import seaborn as sns
10 from shapely.geometry import Point
11 from shapely.ops import nearest_points
```

anaconda/envs/azureml_py38/lib/python3.8/site-packages/xgboost/core.py:265: FutureWarning: Your system has an old version of glibc (< 2.28). We will stop supporting Linux distros with glibc older than 2.28 after **May 31, 2025**. Please upgrade to a recent Linux distro (with glibc 2.28+) to use future versions of XGBoost.

Note: You have installed the 'manylinux2014' variant of XGBoost. Certain features such as GPU algorithms or federated learning are not available. To use these features, please upgrade to a recent Linux distro with glibc 2.28+, and install the 'manylinux_2_28' variant.

warnings.warn()

Read the normalized dataset

```
1 #Mounting from data assets Azure ML
2 from azure.ai.ml import MLClient
3 from azure.identity import DefaultAzureCredential
4
5 ml_client = MLClient.from_config(credential=DefaultAzureCredential())
```



USECASE

Geothermal Gradient Prediction using Auto ML

Method :

- Open Azure Automated ML
- Create new project or pick existing project and pick ML type (classification, regression, etc)
- Connect to your dataset.
- Train your model.

Authoring

1



Notebooks



Automated ML



Designer



Prompt flow



Tracing

PREVIEW

Automated ML

Let Automated ML train and find the best model based on your data without writing a single line of code. [Learn more about Automated ML](#)

+ New Automated ML job Refresh

Recent Automated ML jobs

[View all experiments](#)

Display name	Experiment	Status	Created on	Duration	Created by	Compute targ...	Tags
preprocessed-text	first-experiment	Completed	Aug 15, 2024 1:16 PM	6h 12m 26s	Muhammad F...	Serverless	dynamic_allowlist: ...
raw-sample-	first-experiment	Failed	Aug 15, 2024 11:43 AM	12s	Muhammad F...	Serverless	_azureml.Compute: ...
kaifinance-usecase1-demo-tdpclassific	first-experiment	Canceled	Aug 15, 2024 10:03 AM	3h 13m 37s	Muhammad F...	Serverless	dynamic_allowlist: ...

2

PT. Nawa Darsana Teknologi > aml-development > Training job

Submit an Automated ML job

Let's start with some basic information about your training job.

Basic settings

Job name *

Experiment name *

Select existing ☒ Create new ☐

Existing experiment *

Description

Tags

Name	Value	Add
------	-------	-----

Back Stop Cancel

3

PT. Nawa Darsana Teknologi > aml-development > Training job

Submit an Automated ML job

Training method

Basic settings

Task type & data

Choose the type of task that you would like your model to

Select task type *

Classification

To predict one of several categories yes/no, blue, red, green.

Regression

To predict continuous numeric values.

Time series forecasting

To predict values based on time.

Natural language processing

To predict based on text-only data types using multi-class or multi-label classification or named entity recognition.

Computer vision

To predict using multi-class or multi-label image classification, object detection, and instance segmentation.

4





USECASE

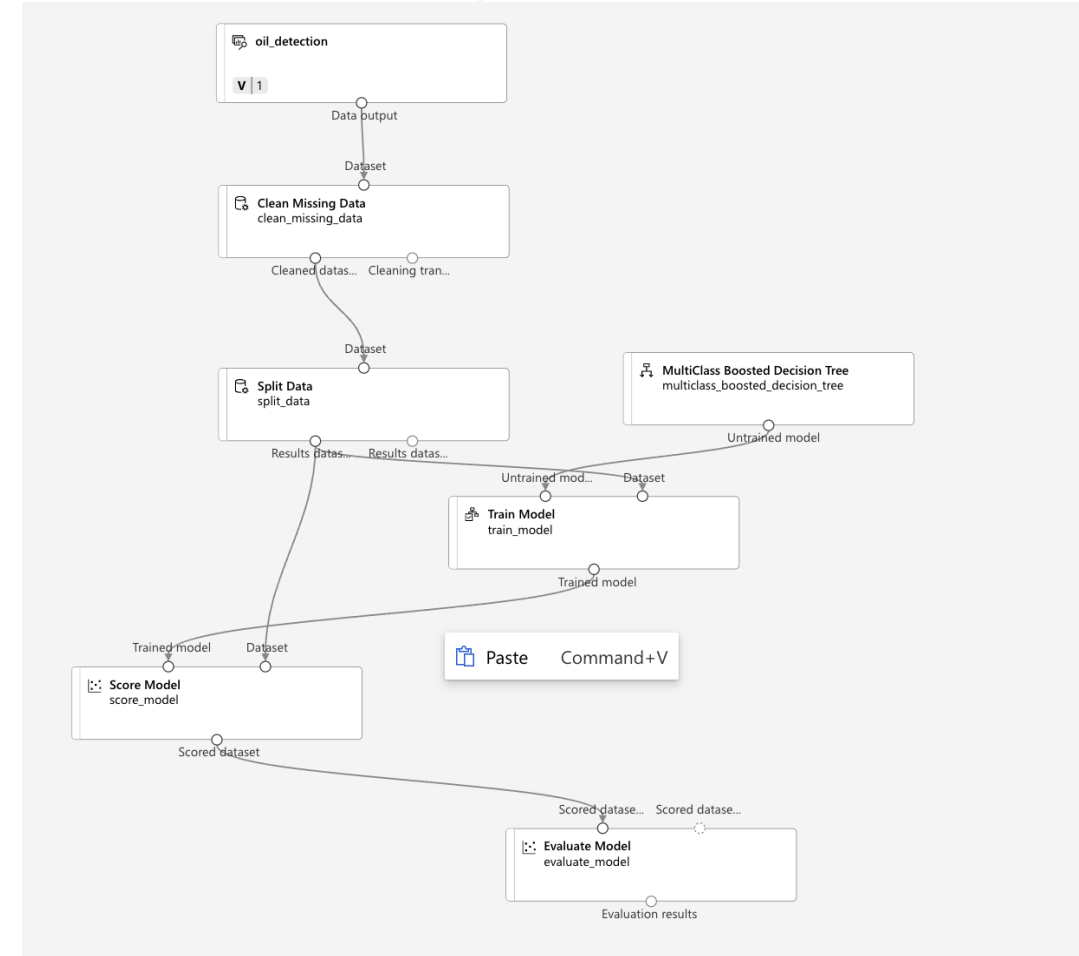
Geothermal Gradient Prediction using AML Designer



Technically the steps we do in Azure Machine Learning Designer resembles the steps on Azure Machine Learning Notebooks / python file. But we transform it to be low code mode, using some flow boxes.

Methods :

- Select AML Designer menu.
- Create new pipeline/designer projects
- Follow the steps as shown on figure





Thank you

Let's discuss our collaborations

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Phone : +62 812 8033 2457