

MLOps Exam 2025 - EN

Summary	The exam assignment consists of two major parts. The first part focus heavily on the Azure Machine Learning story. The goal is to build a pipeline and automate data preprocessing and AI training. In the second part, you will work with HuggingFace to deploy your AI model.
Deadline	You submit no later than 12:30 pm.
Submit	Place the answers to the questions in a .docx or Notion document. You do not need to pay attention to formatting. Take screenshots when prompted. Feel free to add additional notes and screenshots. All your code and this document will be placed in a ZIP file and uploaded via Leho in a timely manner.

Objectives

- 1D.2.5 The student responsibly selects and implements the appropriate data service at the level of input, coding, output or maintenance
- 1D.3.1 The student integrates data tailored to the application
- 1D.3.4 The student implements the most efficient algorithm to build an AI application
- 1E.3.1 The student develops an interactive application using a framework or combination of frameworks.
- 2.2.2 The student handles version control in a simple context

Agreements

Accepted

- Internet access, lab solutions, GitHub Copilot, ChatGPT, ... and if available your brains.

Not allowed

- Communication with others in any way.



Before you start, read the entire document carefully!

General tips

- If you get stuck, try to rejoin at a later point in the assignment. That way, you will leave fewer points unresolved.
- Ask a question when you are completely stuck. That way you won't lose unnecessary time. We may give you an interim solution if necessary. This may cost one or more points, but you can continue working after that.
- Add one screenshot too many rather than too few. Make sure all relevant information is visible so we can correctly assess your work.
- Use GitHub to keep your code safe during the exam.

Assignment

You will have about 3 hours for this assignment, if you don't finish in time, submit as far as you got, you will also earn a large portion of points for how you tackle your project.

Game of Throne House Predictor

The AI model in this exam is not of great importance, the process is mainly about AI training and the components in a pipeline. Therefore, it is important that you focus on that.

Submit

Create a `.docx` file or a Notion document (export as PDF) with the requested screenshots and a brief explanation of your approach. Note which parts succeeded and which did not.

You should then submit the following via **GitHub Classroom** and **Leho**:

- All the source code
- The `.docx` file or the PDF export from Notion with all screenshots and explanations
- The prompts used (see instructions below)

Submit prompts

We want to track your usage of AI, so we can also give you points on how you got to a certain solution if the answer isn't 100% correct. To track everything correctly, you'll need to submit your prompts and chat history. Below are some instructions for popular tools that we have used. Perhaps your tool works a little different

ChatGPT / Gemini

1. Close the sidebar.
2. Press `Ctrl` or `Cmd` + `S` to save the entire conversation as `.html`.
3. Place this file in a `prompts` folder within your project.

Claude

1. Copy the full conversation.
2. Paste it into Word.
3. Save as a `.pdf`.
4. Place this file in the `prompts` folder.

Copilot Chat (in VSCode)

1. Open the Control Panel of VSCode.
2. Choose **Export Chat**.
3. This generates a `.json` file.
4. Place this file in the `prompts` folder.

Cursor / Antigravity / Perplexity

- Search how to do it yourself, in a way that we can find most of your history.
-

Azure

Step 0 - Setup of your Azure Machine Learning Workspace.

- For this assignment, you will use your Azure Machine Learning workspace as you used for your final assignment. Either your own environment or use MCT's existing MLOps resource group on Azure.

Request access to this environment if needed!

- Start up a Compute machine while you take a moment to review the rest of the project. Take a machine that you will later run your Data Processing pipeline on. This one will suffice: **Standard_DS11_v2**

Step 1 - Data Uploading

Points: 3

The data you have been given comes from an AI-generated dataset where there is an attempt to classify which `House` a Game of Thrones character belongs to based on some features.

The task is for you to pour the data preprocessing, as done in some of the cells below, into an Azure pipeline. You've been given some code blocks below that can help with the processing.

Download the data here!!

[got_persona_dataset_100.xlsx](#)

Input = `got_persona_dataset_100.xlsx`

Code:

- Make sure the Excel can be loaded. Use the `openpyxl` package:

```
!/anaconda/envs/azureml_py310_sdkv2/bin/python -m pip install openpyxl
```

- Make sure the dataset is uploaded on Azure as a **Tabular** dataset.
- For this, use code to convert the Excel to the correct format.

You may create this via SDK, CLI+YAML or via the GUI Editor.



Screenshot (3 points)

Data Asset with the Tabular data

The screenshot shows the Azure Machine Learning portal interface. The left sidebar contains navigation options like 'All workspaces', 'Home', 'Model catalog', 'Authoring', 'Assets', 'Data', 'Jobs', 'Components', 'Pipelines', 'Environments', 'Models', 'Endpoints', 'Manage', 'Compute', 'Monitoring', 'Data Labeling', and 'Connections'. The main area displays the details for a dataset named 'persona_dataset' (Version: 1 (latest)). The 'Attributes' section shows: Type: Table (mtable), Named asset URI: azureml:persona_dataset:1, Created by: Danylo Bordunov, Current version: 1, Latest version: 1, Created time: Dec 17, 2025 9:58 AM, and Modified time: Dec 17, 2025 9:58 AM. The 'Tags' section shows 'No tags'. The 'Description' section shows 'exam'. The 'Data sources' section shows 'Datastore: workspaceblobstore' with a relative path 'UI/2025-12-17_085324_UTG/get_mtable/'. Below this, there are links for 'View in datastores browse', 'View in Azure Portal', and 'Show MLTable file preview'. At the bottom, the 'Datastore URI' and 'Storage URI' are listed.

Step 1 - AI Data Preparation

Points: 5

You can, using the code below, transform your Data Asset from the raw data, to a trainable version in Training and Testing Features and Targets.

Use a component and a Pipeline to do this.

`conda.yaml` environment file.

```
name: got-prepare-env
channels:
  - conda-forge
dependencies:
  - python=3.10
  - pip
  - pip:
    - pandas
    - scikit-learn
```

`prepare_component.py` for the splitting into Train and Testing data.

```
import argparse
import json
import os
from glob import glob

import pandas as pd
from sklearn.model_selection import train_test_split

def write_mltable(folder: str, filename: str):
    # MLTable that points to one file
    with open(os.path.join(folder, "MLTable"), "w", encoding="utf-8") as f:
        f.write(
f"""paths:
  - file: ./{filename}
transformations:
  - read_delimited:
      delimiter: ","
      encoding: "utf8"
      header: all_files_same_headers
```

```

"""
    )

def main():
    parser = argparse.ArgumentParser()
    parser.add_argument("--input_folder", type=str, required=True)

    parser.add_argument("--target_col", type=str, default="house_affiliation")

    parser.add_argument("--test_size", type=float, default=0.2)
    parser.add_argument("--seed", type=int, default=42)

    parser.add_argument("--stratify", type=int, default=1) # 1=true, 0=false

    parser.add_argument("--out_train", type=str, required=True)
    parser.add_argument("--out_test", type=str, required=True)

    args = parser.parse_args()

    # locate csv
    csv_path = os.path.join(args.input_folder, "data.csv")
    if not os.path.exists(csv_path):
        csvs = glob(os.path.join(args.input_folder, "*.csv"))
        if not csvs:
            raise FileNotFoundError(f"No CSV found in input folder: {args.input_folder}")
        csv_path = csvs[0]

    df = pd.read_csv(csv_path)
    if args.target_col not in df.columns:
        raise ValueError(f"Target column '{args.target_col}' not found. Columns: {list(df.columns)}")

    # ✅ Drop columns that should never be features (prevent leakage)
    # - character_name is basically an ID / label proxy

```

```

# - character_id is also an ID
leakage_cols = ["character_id", "character_name"]
drop_cols = [c for c in leakage_cols if c in df.columns]

y = df[args.target_col].astype(str)
X = df.drop(columns=[args.target_col] + drop_cols, errors="ignore")

strat = None
if args.stratify == 1:
    vc = y.value_counts()
    too_small = vc[vc < 2]
    if len(too_small) == 0:
        strat = y
        print("✅ Using stratified split.")
    else:
        strat = None
        print("⚠️ Stratify disabled: at least one class has < 2 samples.")
        print("Classes with too few samples:", too_small.to_dict())

X_train_raw, X_test_raw, y_train, y_test = train_test_split(
    X,
    y,
    test_size=args.test_size,
    random_state=args.seed,
    stratify=strat
)

# One-hot encode categoricals on TRAIN, then align TEST
X_train = pd.get_dummies(X_train_raw, dummy_na=True)
X_test = pd.get_dummies(X_test_raw, dummy_na=True)

# Align columns so train/test match exactly
X_train, X_test = X_train.align(X_test, join="left", axis=1, fill_value=0)

# Save outputs
os.makedirs(args.out_train, exist_ok=True)

```



```

os.makedirs(args.out_test, exist_ok=True)

X_train_path = os.path.join(args.out_train, "X_train.csv")
y_train_path = os.path.join(args.out_train, "y_train.csv")
X_test_path = os.path.join(args.out_test, "X_test.csv")
y_test_path = os.path.join(args.out_test, "y_test.csv")

X_train.to_csv(X_train_path, index=False, encoding="utf-8")
y_train.to_frame(name=args.target_col).to_csv(y_train_path, index=False, encoding="utf-8")

X_test.to_csv(X_test_path, index=False, encoding="utf-8")
y_test.to_frame(name=args.target_col).to_csv(y_test_path, index=False, encoding="utf-8")

# Save metadata for reproducibility & inference compatibility
with open(os.path.join(args.out_train, "feature_columns.json"), "w", encoding="utf-8") as f:
    json.dump({"columns": list(X_train.columns)}, f, ensure_ascii=False, indent=2)

classes = sorted(y.unique().tolist())
with open(os.path.join(args.out_train, "label_classes.json"), "w", encoding="utf-8") as f:
    json.dump({"classes": classes}, f, ensure_ascii=False, indent=2)

# Add MLTable definitions (point to the main files)
write_mltable(args.out_train, "X_train.csv")
write_mltable(args.out_test, "X_test.csv")

print("✅ Done")
print("Target:", args.target_col)
print("Train shape:", X_train.shape, "Test shape:", X_test.shape)
print("Num classes:", len(classes))

```

```
if __name__ == "__main__":  
    main()
```

You may create this via SDK, CLI+YAML or via the GUI Editor.



Screenshot (2 points)

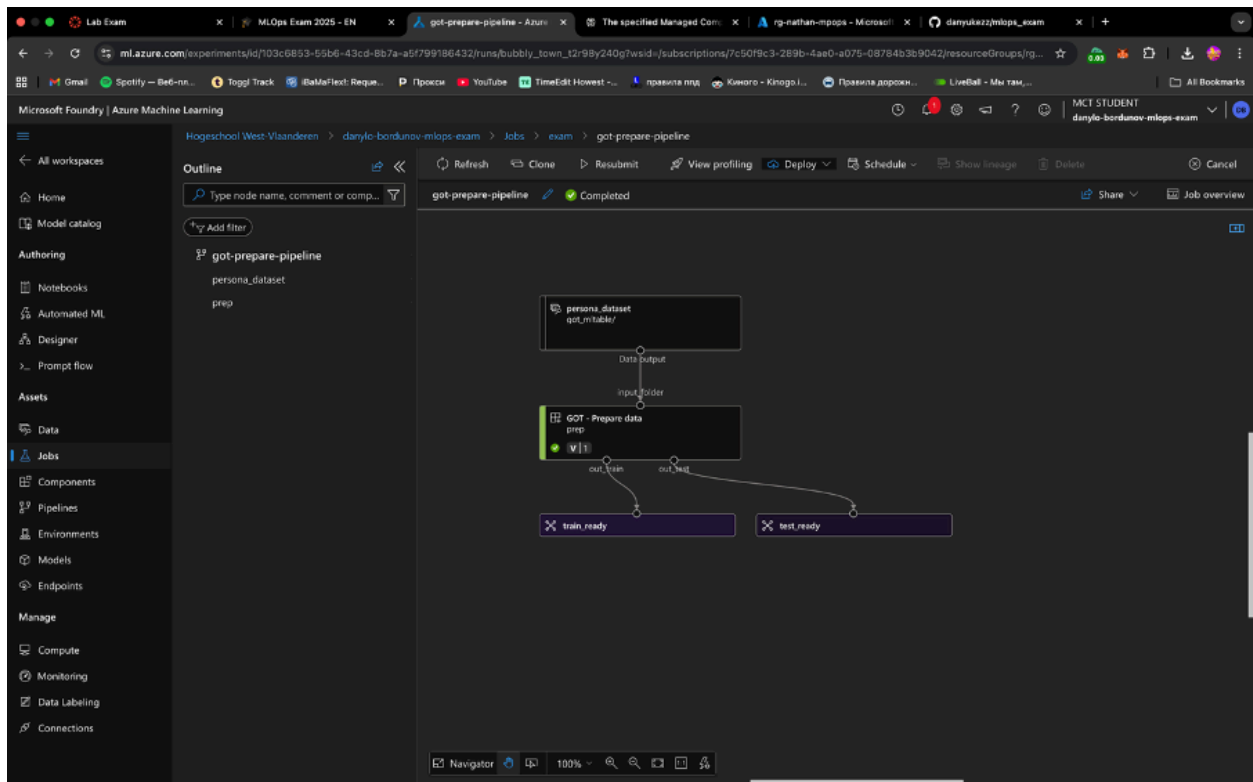
Your component for AI Data Preparation in the Component tab of Azure.

The screenshot shows the Azure Machine Learning interface. The left sidebar contains navigation options: All workspaces, Home, Model catalog, Authoring, Notebooks, Automated ML, Designer, Prompt flow, Assets, Data, Jobs, Components (selected), Pipelines, Environments, Models, Endpoints, Manage, Compute, Monitoring, Data Labeling, and Connections. The main panel displays the 'GOT - Prepare data' component details. The 'Details' tab is active, showing attributes: Name (got_prepare_data), Component ID (a09195fd-b6ea-46d4-b777-6108f25ccc3c), Created by (Danylo Bordinov), Version (1), Type (command), Created on (Dec 17, 2025 10:52 AM), and Last updated on (Dec 17, 2025 10:52 AM). The 'Environment' section shows the environment name and ID. The 'Tags' section indicates 'No properties'. The 'Component preview' section shows a visual representation of the component as a box with input and output ports.



Screenshot (2 points)

Your **successful** pipeline with **Input** and **Output** visible.



Screenshot (1 point)

Your stored Data Assets for **Train** and **Testing** data.

The screenshot displays the Azure Machine Learning portal interface. The left sidebar contains navigation options: All workspaces, Home, Model catalog, Authoring (Notebooks, Automated ML, Designer, Prompt flow), Assets (Data, Jobs, Components, Pipelines, Environments, Models, Endpoints), and Manage (Compute, Monitoring, Data Labeling, Connections). The main content area shows the 'got-train-ready' dataset details for Version: 1 (latest). The 'Attributes' section lists: Type: Table (mitable), Named asset URI: azureml:got-train-ready:1, Created by: Danylo Bordunov, Current version: 1, Latest version: 1, Created time: Dec 17, 2025 11:08 AM, Modified time: Dec 17, 2025 11:08 AM, and Created by job: 64ebb93f-0306-495e-bbc4-cee211c38a63. The 'Tags' section shows 'No tags'. The 'Description' section has a placeholder 'Click edit icon to add a description'. The 'Data sources' section lists the Datastore as 'workspaceblobstore' and provides the Relative path: 'azureml/64ebb93f-0306-495e-bbc4-cee211c38a63/out_train/'. It also includes links for 'View in datastores browse', 'View in Azure Portal', and 'Show MLTable file preview'. The 'Datastore URI' is 'azureml/subscriptions/7c50f9c3-289b-4ae0-a075-08784b3b9042/resourceGroups/rg-nathan-mpops/providers/Microsoft.MachineLearningServices/...', and the 'Storage URI' is 'https://danylobordunov6803185837.blob.core.windows.net/azureml-blobstore-69b9c7e8-780a-44...'. The 'URI' section shows the Location as 'http://uri'.

The screenshot displays the Azure Machine Learning portal interface for the 'got-test-ready' dataset. The left sidebar is identical to the previous screenshot. The main content area shows the 'got-test-ready' dataset details for Version: 1 (latest). The 'Attributes' section lists: Type: Table (mitable), Named asset URI: azureml:got-test-ready:1, Created by: Danylo Bordunov, Current version: 1, Latest version: 1, Created time: Dec 17, 2025 11:08 AM, Modified time: Dec 17, 2025 11:08 AM, and Created by job: 64ebb93f-0306-495e-bbc4-cee211c38a63. The 'Tags' section shows 'No tags'. The 'Description' section has a placeholder 'Click edit icon to add a description'. The 'Data sources' section lists the Datastore as 'workspaceblobstore' and provides the Relative path: 'azureml/64ebb93f-0306-495e-bbc4-cee211c38a63/out_test/'. It also includes links for 'View in datastores browse', 'View in Azure Portal', and 'Show MLTable file preview'. The 'Datastore URI' is 'azureml/subscriptions/7c50f9c3-289b-4ae0-a075-08784b3b9042/resourceGroups/rg-nathan-mpops/providers/Microsoft.MachineLearningServices/...', and the 'Storage URI' is 'https://danylobordunov6803185837.blob.core.windows.net/azureml-blobstore-69b9c7e8-780a-44...'. The 'URI' section shows the Location as 'http://uri'.

Step 2 - AI Training - Decision Tree Classifier

Points: 7

For the AI model, we work with a Decision Tree Classifier. The Environment for this is also given, as well as the code to initiate the training itself.

```
name: got-train-dt-env
channels:
  - conda-forge
dependencies:
  - python=3.10
  - pip
  - pip:
    - pandas
    - scikit-learn
    - joblib
```

`training.py` - Code has been given, not yet in the right structure to do everything in a component manner though

```
# Load data
X_train_path = os.path.join(args.train_ready, "X_train.csv")
y_train_path = os.path.join(args.train_ready, "y_train.csv")
X_test_path = os.path.join(args.test_ready, "X_test.csv")
y_test_path = os.path.join(args.test_ready, "y_test.csv")

for p in [X_train_path, y_train_path, X_test_path, y_test_path]:
    if not os.path.exists(p):
        raise FileNotFoundError(f"Missing required file: {p}")

X_train = pd.read_csv(X_train_path)
y_train = pd.read_csv(y_train_path)[args.target_col].astype(str)

X_test = pd.read_csv(X_test_path)
y_test = pd.read_csv(y_test_path)[args.target_col].astype(str)

# Train model
```

```
clf = DecisionTreeClassifier(  
    max_depth=args.max_depth,  
    min_samples_split=args.min_samples_split,  
    min_samples_leaf=args.min_samples_leaf,  
    random_state=args.random_state,  
)  
clf.fit(X_train, y_train)  
  
# Evaluate  
y_pred = clf.predict(X_test)  
acc = float(accuracy_score(y_test, y_pred))  
  
report = classification_report(y_test, y_pred, output_dict=True, zero_division=  
0)
```

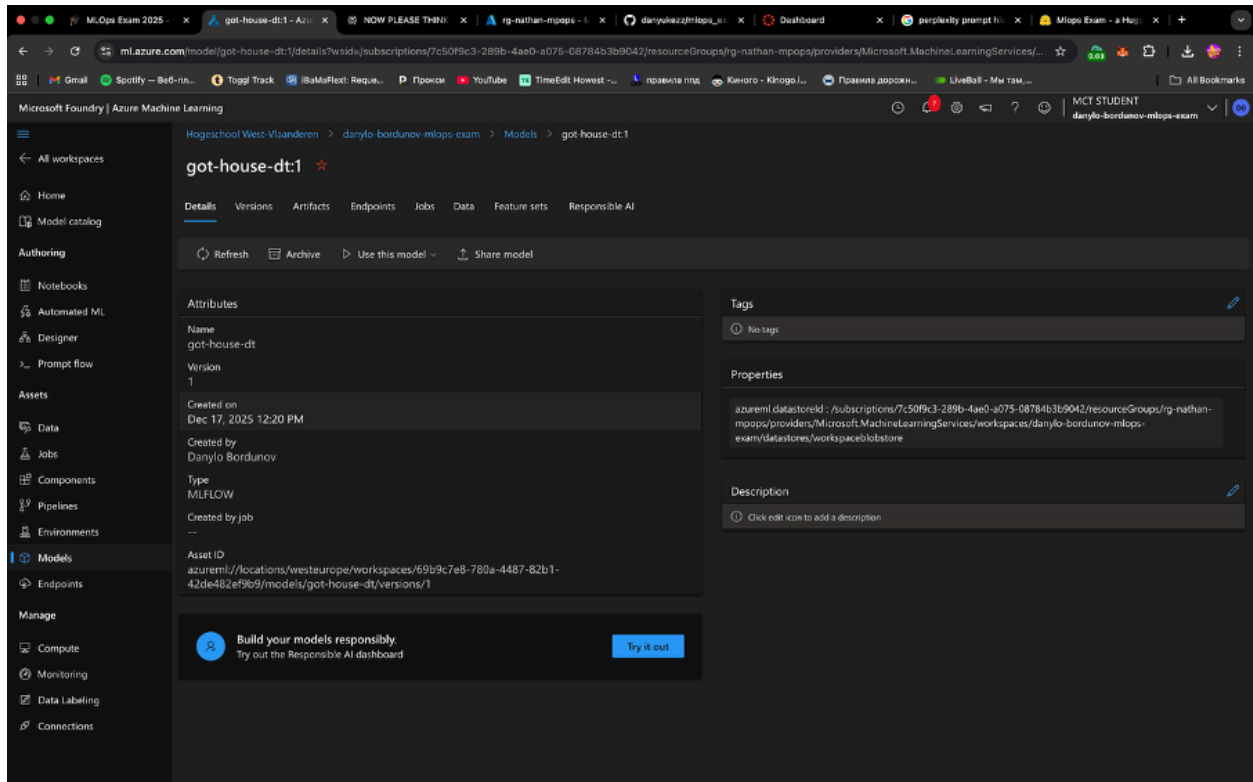
Checklist

- ✓ Log the metrics from the Classification Report into the Azure ML Experiments using **MLFlow Logging**
- ✓ Make sure your AI model gets stored and registered in your Azure ML Pipeline
 - ✓ This is done in a separate step with this component:
azureml://registries/azureml/components/register_model/versions/0.0.21



Screenshot (1 point)

Your registered AI model in the **Model** view.



Screenshot (2 points)

Your Classification Report metrics from MLFlow

The screenshot shows the Azure Machine Learning portal interface. The main window displays the 'GOT - Train Decision Tree' experiment. The 'Outputs' tab is active, showing a list of outputs on the left sidebar and a detailed view of the 'classification_report.json' output on the right. The JSON output contains the following metrics:

```

{
  "accuracy": 0.7,
  "report": {
    "Arqyn": {
      "precision": 1,
      "recall": 1,
      "f1-score": 1,
      "support": 1
    },
    "Barathene": {
      "precision": 0,
      "recall": 0,
      "f1-score": 0,
      "support": 1
    },
    "Free Folk": {
      "precision": 1,
      "recall": 1,
      "f1-score": 1,
      "support": 1
    },
    "Greyjoy": {
      "precision": 1,
      "recall": 0.5,
      "f1-score": 0.6666666666666666,
      "support": 2
    },
    "Lannister": {
      "precision": 0.8,
      "recall": 0.8,
      "f1-score": 0.8,
      "support": 5
    },
    "Mormont": {

```



Screenshot (2 points)

Your component for AI Training.

Screenshot of the Microsoft Foundry | Azure Machine Learning interface showing the 'GOT - Train Decision Tree' component details.

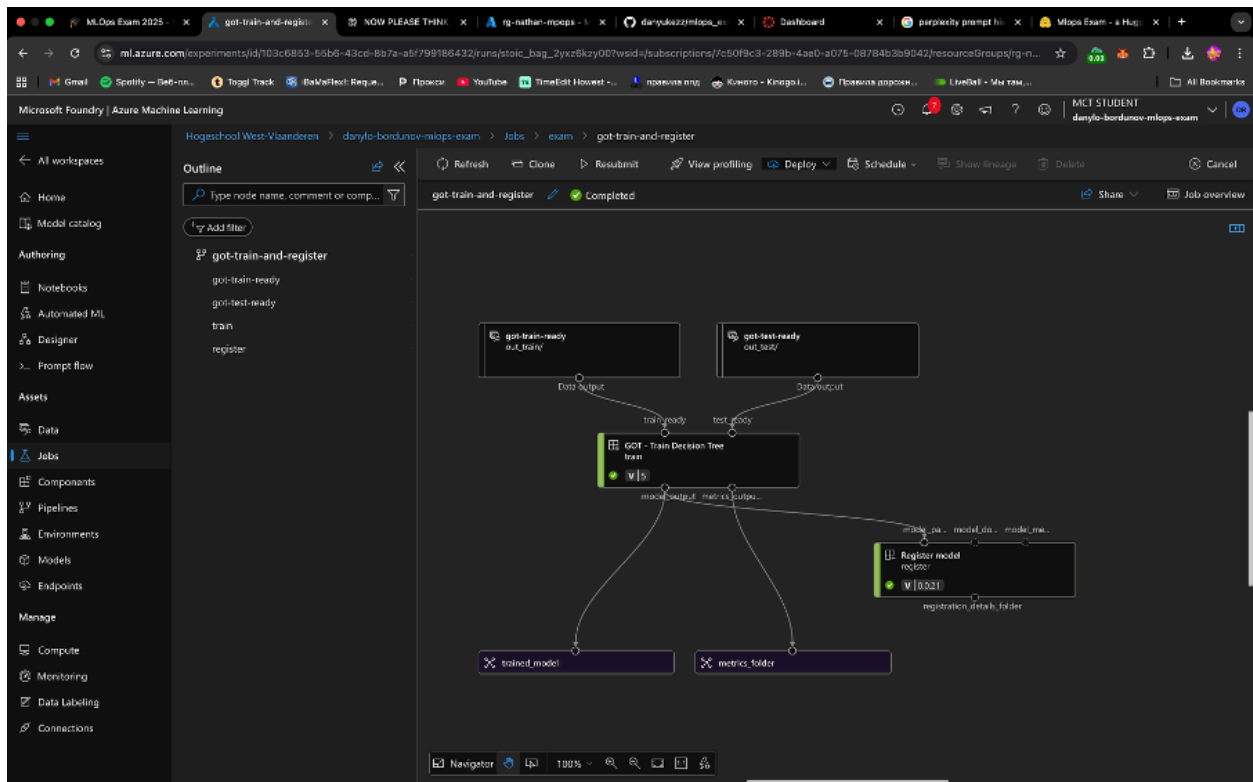
The interface displays the component's attributes, environment, and tags. The component is named 'got_train_dt' and is a command type. It is created by Danylo Bordunov and is currently in version 2. The environment is 'CIIV2AnonymousEnvironment:a0d82c843698535aa7f21eb30fe115dc48a456424a02117c4fe576991f3569ec'. The component is currently in a 'No properties' state.

The component preview shows a diagram of the 'GOT - Train Decision Tree' component, which is a rectangular box with four connection points (two on the left, two on the right).



Screenshot (2 points)

Your **working** pipeline with **Input** and **Output**.



Step 3 - Deployment

Points: 5



You can perform this step on your local PC, without using Azure Machine Learning.

I do think it would be nice if there is an API in FastAPI or Gradio where the AI model can be applied.

You will need to send a request to an API endpoint (`/predict`) with the JSON payload below.

I expect a response where you show which **House** a character belongs to based on the JSON input entered.

Also post this on HuggingFace The necessary inputs can then be entered as follows:

```
{
  "region": "The North",
  "primary_role": "Commander",
  "alignment": "Lawful Good",
  "status": "Alive",
  "species": "Human",
  "honour_1to5": 4,
  "ruthlessness_1to5": 2,
  "intelligence_1to5": 3,
  "combat_skill_1to5": 4,
  "diplomacy_1to5": 3,
  "leadership_1to5": 4,
  "trait_loyal": true,
  "trait_scheming": false
}
```

The screenshot shows a VS Code editor with a project named 'mllops_exam'. The Explorer sidebar on the left shows the file structure, including 'app.py', 'schemas.py', 'dockerignore', 'Dockerfile', 'requirements.txt', 'got_mitable', '.gitignore', 'mllops_exam_report.docx', and 'README.md'. The main editor window displays the 'app.py' file, which is a FastAPI application. The code imports 'PredictRequest' and 'PredictResponse' from 'app.schemas', loads an MLflow model from 'model_dir', and defines a health endpoint and a prediction endpoint. The prediction endpoint uses 'mlflow.pyfunc.load_model' to load the model and 'model.predict' to make predictions. The status bar at the bottom indicates the file is at line 28, column 1, with 4 spaces, UTF-8 encoding, and LF line endings. The status bar also shows 'Python 3.12.10' and 'Continue (NE)' as the active interpreter.

```
fastapi > app > app.py > ...
5 from app.schemas import PredictRequest, PredictResponse
6
7 MODEL_DIR = "model" # folder containing ML model
8 model = mlflow.pyfunc.load_model(MODEL_DIR)
9
10 app = FastAPI(title="GOT House Predictor", version="1.0")
11
12 @app.get("/health")
13 def health():
14     return {"status": "ok"}
15
16 @app.post("/predict", response_model=PredictResponse)
17 def predict(payload: PredictRequest):
18     X = pd.DataFrame(payload.model_dump())
19     y = model.predict(X)
20
21     # MLflow pyfunc often returns numpy array / pandas series / dataframe depending on flavor
22     if hasattr(y, "iloc"):
23         pred = y.iloc[0]
24     else:
25         pred = y[0]
26
27     return {"house": str(pred)}
28
```



Screenshot (3 points)

Your `/docs` endpoint showing a successful execution of the input data with output results. If you chose Gradio, you may also show a screenshot of your frontend, preferably with input and output visible.



Screenshot (2 points)

Show your HuggingFace Space where this is running.

The screenshot shows a web browser displaying the HuggingFace Space page for 'danyukezz/mlops_exam'. The page has a dark theme and includes the following sections:

- Get started with your Gradio Space!**: A message stating 'Your new space has been created, follow these steps to get started (or read the full [documentation](#))'.
- Start by cloning this repo by using:**: Instructions for cloning the repository using HTTPS, SSH, or curl. The SSH command is: `git clone https://huggingface.co/spaces/danyukezz/mlops_exam`. The curl command is: `curl -Lsf https://hf.co/cli/install.sh | bash`.
- Create your Gradio app.py file:**: A code block showing the following Python code:

```
import gradio as gr

def greet(name):
    return "Hello " + name + "!"

demo = gr.Interface(fn=greet, inputs="text", outputs="text")
demo.launch()
```
- Then commit and push:**: A code block showing the following shell commands:

```
git add app.py
git commit -m "Add application file"
git push
```

Submit

Submission instructions are at the top of this document. Will you check the following?

- ☒ Have you posted the necessary screenshots everywhere?
- ☒ Did you turn off your Azure services at the end of the exam?

✓ Did you download all your code on your PC, and push it on GitHub?

✓ Did you compress the total project into a `.zip` folder?

If you did everything, export this document as a `.pdf` and upload it along with your `.zip` folder on Leho.

Additional

You may also provide your feedback on this course in the document 😊.

Questions?

You can send me a message directly on Teams, or raise your hand in the classroom.

Technical problems with your computer or the network, you should deal with them through the Helpdesk. They can solve those problems.

Good luck!