National College of Ireland

Configuration Manual

MSc Research Project
Cloud Computing

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Project Submission Sheet
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Programme:	Cloud Computing
Year:	2024
Module:	MSc Research Project
Supervisor:	Shivani Jaswal
Submission Due Date:	29/01/2025
Project Title:	Configuration Manual
Word Count:	693
Page Count:	8

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Configuration Manual

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1. Creating Azure Resource group and Workspace

To create the workspace:

- Log in to the Azure Portal.
- Navigate to Azure Machine Learning.
- Create a new Resource Group and Workspace as shown in Figure 1.

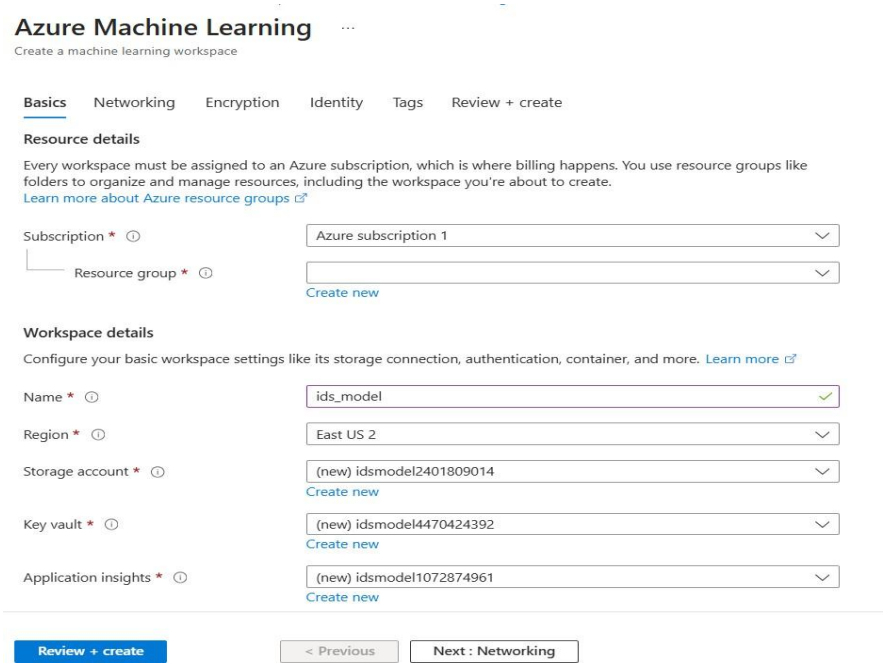

Figure 1: Azure ML Workspace Creation

2. Importing Libraries

Libraries required for data preprocessing, model training, and deployment were imported into Azure Notebooks. This setup is shown in Figure 2.

```python
import numpy as np
import matplotlib.pyplot as plt
from azureml.core import Workspace, Dataset
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, precision_score, recall_score, confusion_matrix
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.naive_bayes import GaussianNB
from sklearn.neighbors import KNeighborsClassifier
from sklearn.preprocessing import LabelEncoder
from azureml.core import Workspace, Model
from azureml.core.environment import Environment
from azureml.core.conda_dependencies import CondaDependencies
from azureml.core.model import InferenceConfig
import requests
import json
```

Figure 2: Library Imports

3. ML Model Registration

The trained Random Forest model (Random_Forest_best_model.pkl) was registered in the Azure ML Workspace. The Python script for model registration is as shown in Figure 3.

```python
# Connect to the Azure ML Workspace
ws = Workspace.from_config()
print(ws)

# Register the model in Azure ML
registered_model = Model.register(workspace=ws,
                        model_name="best_intrusion_detection_model",  # Name of the registered model
                        model_path="Random_Forest_best_model.pkl")    # Path to your saved model

```

```
Workspace.create(name='pred_intrution', subscription_id='ea31d986-323f-42ba-bc31-aef14ad815cc',
resource_group='intrution_detection_ml')
Registering model best_intrusion_detection_model
```

Figure 3: ML Model Registration

4. Deploying Model

The registered model was deployed as a Real-Time Endpoint. Deployment settings and the deployment script are as in Figure 4

```
service = Model.deploy(workspace=ws,
                       name='intrution-prediction-service',
                       models=[registered_model],
                       inference_config=inference_config,
                       deployment_config=aci_config)
service.wait_for_deployment(show_output=True)
```

6] ✓

```
/tmp/ipykernel_3387/656689626.py:1: FutureWarning: azureml.core.model:
To leverage new model deployment capabilities, AzureML recommends using CLI/SDK v2 to deploy models as online endpoint,
please refer to respective documentations
https://docs.microsoft.com/azure/machine-learning/how-to-deploy-managed-online-endpoints /
https://docs.microsoft.com/azure/machine-learning/how-to-attach-kubernetes-anywhere
For more information on migration, see https://aka.ms/acimoemigration
```

Figure 4: Model Deployment

5. Microsoft Forms Setup

Step 1: Form Creation

- Open **Microsoft Forms** and create a new form titled "Intrusion Detection Data Entry".
- Add fields for key attributes:
    - **Duration (ms)** (Number)
    - **Source Bytes** (Number)
    - **Destination Bytes** (Number)
    - **Protocol Type** (Dropdown: TCP, UDP, ICMP)
    - **Service** (Dropdown: HTTP, FTP, SMTP, Other)
    - **Flag** (Dropdown: SF, S0, REJ, etc.)
    - **Wrong Fragment** (Dropdown: 0, 1, 2)

Step 2: Set Required Fields

- Make all fields required to avoid missing data.

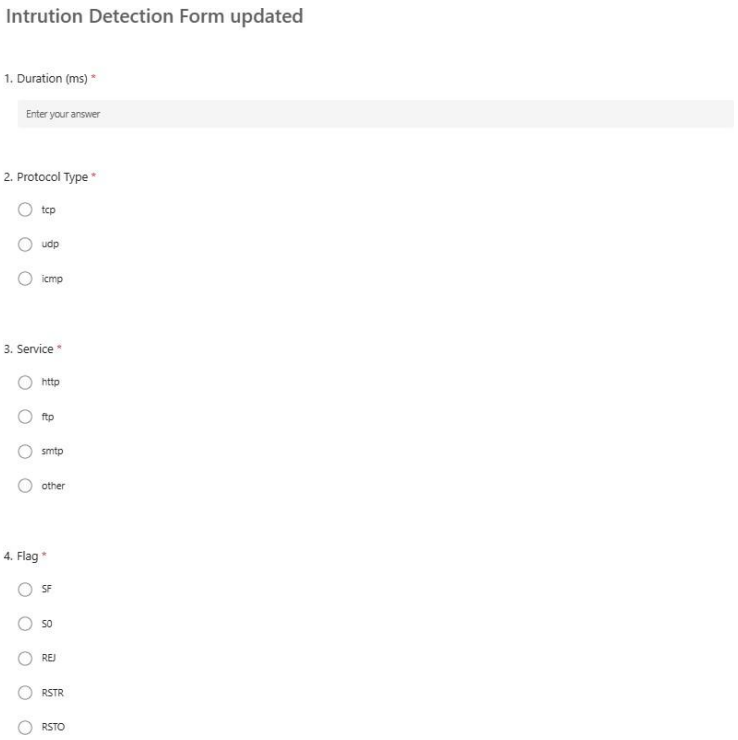

Figure 5: Forms Setup

6. Configuring Power Automate

Step 1: Trigger Setup

- Trigger: **"When a new response is submitted"** (Microsoft Forms).
- Choose the form created in the previous step.

Step 2: Fetch Response Details

- Action: **"Get response details".**
- Link it to the form and fetch all responses dynamically.

Step 3: Data Transformation

- Initialize variables for each field:
    - Example: protocol_type, service, flag as Strings.
    - Use **Switch Actions** to encode categorical values (e.g., TCP = 0, HTTP = 1).

Step 4: Compose Input Data

- Use the **Compose** action to format the data into a list: [

    @{variables('duration')},

    @{variables('protocol_type_encoded')},

    @{variables('service_encoded')},

    @{variables('flag_encoded')},

@{variables('src_bytes')}, @{variables('dst_bytes')},

0, @{variables('wrong_fragment')}, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 5, 3,

0, 0, 0, 0, 1, 0, 0, 100, 100, 0, 0, 0, 0, 0, 0, 0, 0

]

Step 5: HTTP Request to ML Model

- Action: "**HTTP**"
  - **Method:** POST
  - **URI:** Scoring URI from Azure ML deployment.

```json
{
  "Content-Type": "application/json"
}
```

- Body:

```json
{
  "data": @{outputs('Compose')}
}
```

Figure 7: HTTP Request Header and Body

Step 6: Parse JSON

- Parse the HTTP response to extract the ML model's prediction:

```json
{
  "type": "object",
  "properties": {
    "result": {
      "type": "array",
      "items": {
        "type": "integer"
      }
    }
  }
}
```

Figure 8: Json parse for Model prediction

Step 7: Conditional Check

- Action: **Condition**
    - **Condition:** result [0] is equal to 1
        - **If Yes:** Send an alert email (e.g., "Intrusion Detected").
        - **If No:** Log the event as normal traffic.