MACHINE LEARNING MODEL DEPLOYMENT WITH IBM CLOUD WATSON STUDIO [PHASE-3]

INTRODUCTION:

IBM Cloud Watson Studio is a comprehensive platform for data science and machine learning that provides various tools and services to streamline the model development process. In this guide, we will create a machine learning model using Watson Studio to address a predictive use case.

1. Define the Predictive Use Case:

The first step is to define the problem you want to solve using machine learning. This could be anything from predicting customer churn to image classification. For this example, let's assume we want to build a model to predict customer churn for a telecommunications company.

2. Select a Relevant Dataset:

Next, you need to find and select a dataset that is relevant to your use case. You can use publicly available datasets or your own data if applicable. In this case, you might use a dataset containing customer information and historical churn data.

3. Import the Dataset:

In Watson Studio, you can import the dataset into your project. Watson Studio provides a user-friendly interface to upload and manage data. The dataset should be in a compatible format (e.g., CSV, Excel).

4. Preprocess the Data:

Data preprocessing is a crucial step. You may need to handle missing values, clean the data, and transform it to make it suitable for modelling. Watson Studio provides tools for data cleaning and transformation.

5. Select Features:

In this step, you choose which features (attributes) from the dataset will be used as input for your machine learning model. Feature selection can have a significant impact on model performance.

6. Train the Machine Learning Model:

Watson Studio offers a variety of machine learning algorithms and libraries. You can build and train your model using these tools. For customer churn prediction, you might use classification algorithms like logistic regression, decision trees, or a neural network.

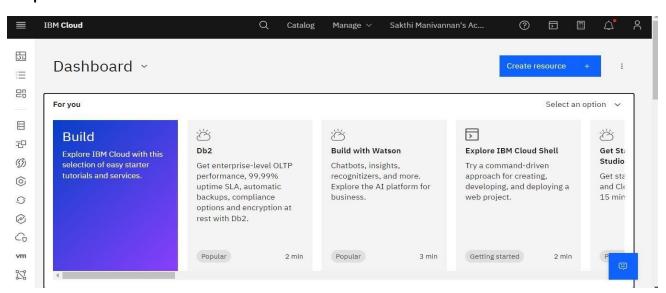
7. Evaluate and Tune the Model:

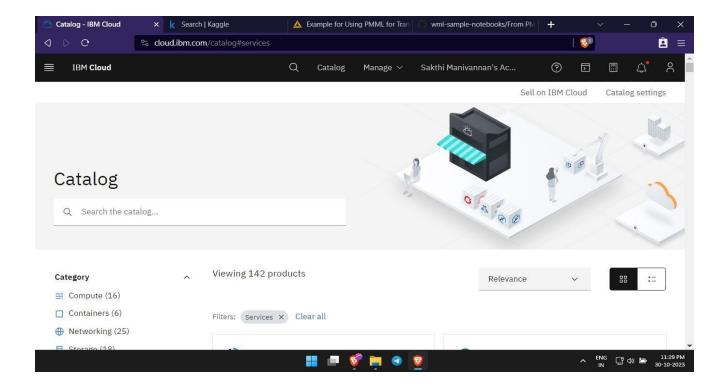
After training, it's essential to evaluate the model's performance using metrics such as accuracy, precision, recall, and F1-score. You can iterate on your model by tuning hyperparameters and selecting the best-performing one.

8. Deploy the Model:

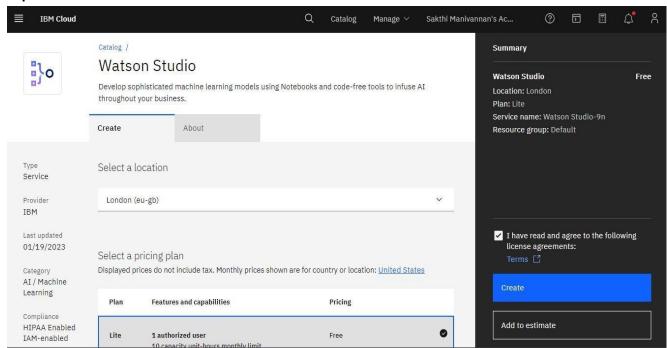
Once you have a satisfactory model, you can deploy it to make predictions on new data. Watson Studio provides deployment options, including APIs for real-time predictions.

Step1:

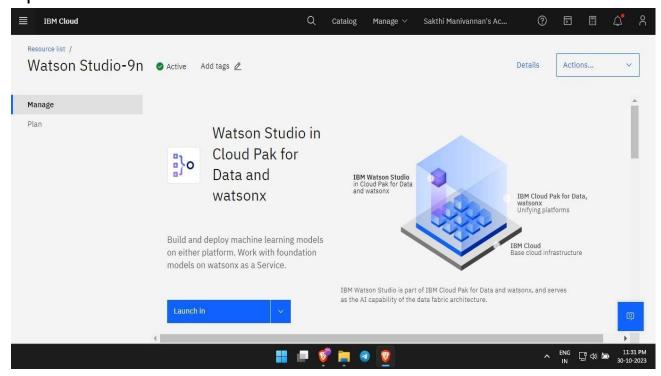


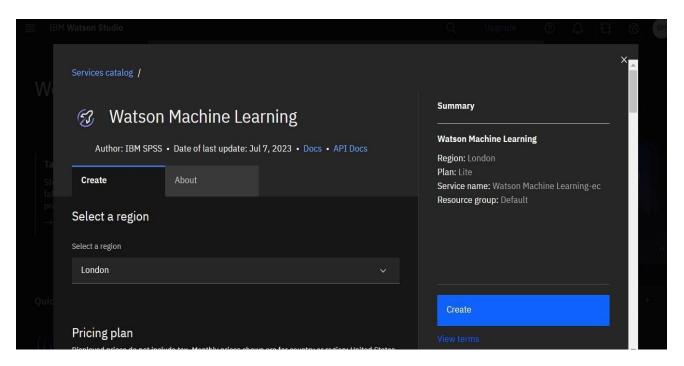


Step2:

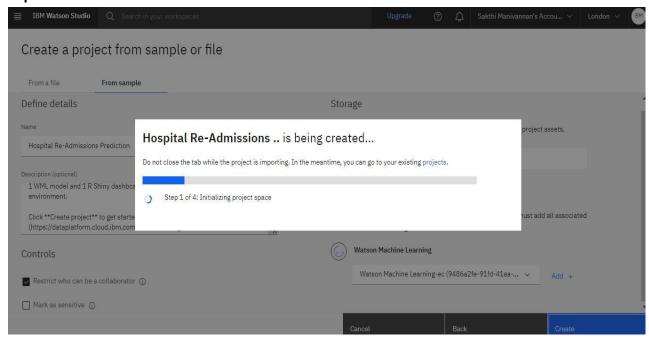


Step3:

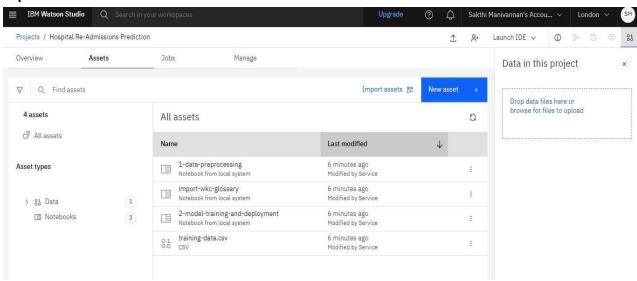


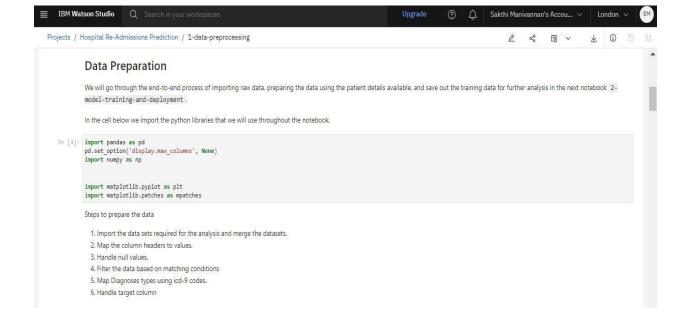


Step4:

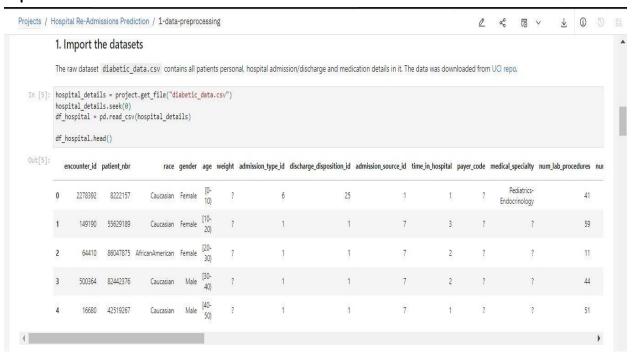


Step5:

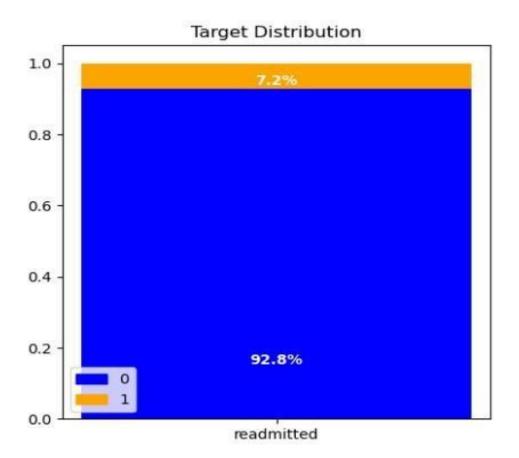




Step6:



Step7:



Step8:

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readmitted

In [16]: df_diag.columns=[x.upper() for x in df_diag.columns.tolist()]
project.save_data('training-data.csv', df_diag.to_csv(index=False), overwrite=True)

Out[16]: {'file_name': 'training-data.csv',
    'message': 'File saved to project storage.',
    'bucket_name': 'hospitalreadmissionspredictionrel-donotdelete-pr-w4kl9yicbyk3wg',
    'asset_id': 'ff@d1894-a19c-41b1-a357-addc5ed9d830'}

The data is prepared and stored in the file training-data.csv , which we can view in the project's data assets. This data set is used in 2-model-training-and-deployments to predict and deploy hospital readmissions using sci-kit and wml libraries. Alternatively, if you want to experiment with the auto Al capability, this data can be used in an auto-ai pipeline to predict hospital readmissions.
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CONCLUSION:

we've outlined the process of building a machine learning model using IBM Cloud Watson Studio. Starting from defining the predictive use case, importing, and preprocessing the data, selecting features, training the model, and deploying it, Watson Studio offers a comprehensive set of tools to streamline the entire process.

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