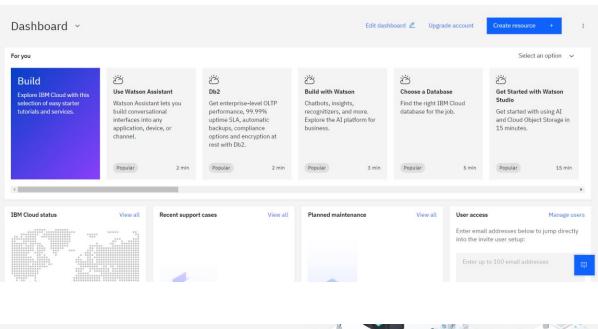
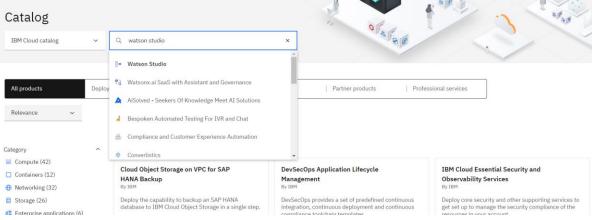
LAB 2- Advance Data Analytics and Data visualization using pandas/Watson Studio

Step 1: Access Watson Studio

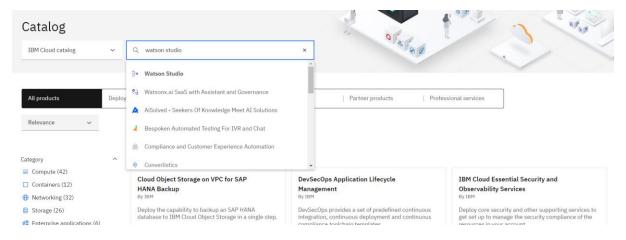
- 1. Log in to IBM Cloud: Go to IBM Cloud and sign in to your account.
- 2. **Search for Watson Studio**: In the IBM Cloud catalog, type "Watson Studio" in the search bar.





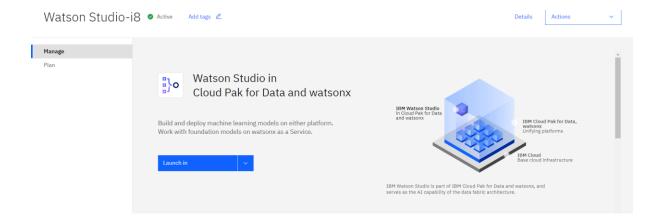
Step 2: Create a Service in Watson Studio

1. **Create Service**: Click on Watson Studio and follow the prompts to create a new Watson Studio service. Select the appropriate plan based on your needs.



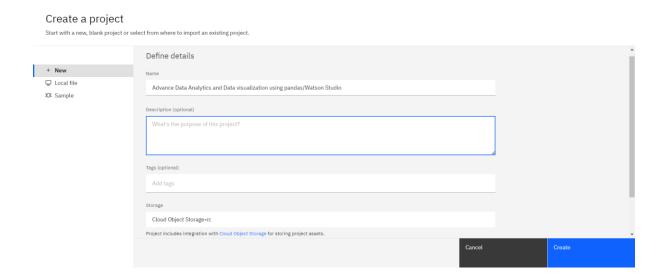
Step 3: Launch IBM Cloud Pak for Data

- 1. **Access Cloud Pak**: Navigate to IBM Cloud Pak for Data if it's part of your environment, or go directly to Watson Studio.
- 2. Open Watson Studio: Click on the Watson Studio service to launch it.



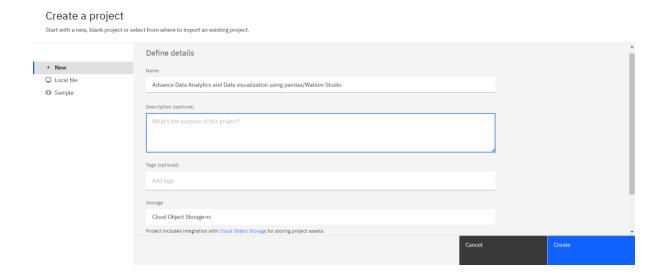
Step 4: Create a New Project

- 1. Create Project: In Watson Studio, click on "New Project."
- 2. Select Project Type: Choose "Default" for a standard project.
- 3. **Provide Project Details**: Enter a name and description for your project and click "Create."



Step 5: Create a New Asset

- 1. Navigate to Assets: Within your project, click on the "Assets" tab.
- 2. Create Asset: Click on "Add Asset" or "Create New Asset."

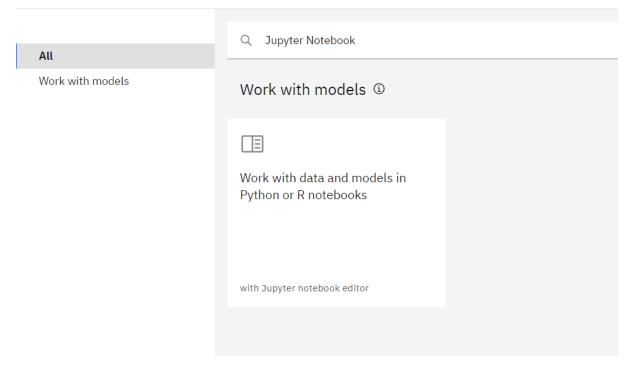


Step 6: Search for Jupyter Notebook

- 1. Find Notebook: In the asset creation menu, search for "Jupyter Notebook."
- 2. **Select Jupyter Notebook**: Choose the option to create a new Jupyter Notebook.

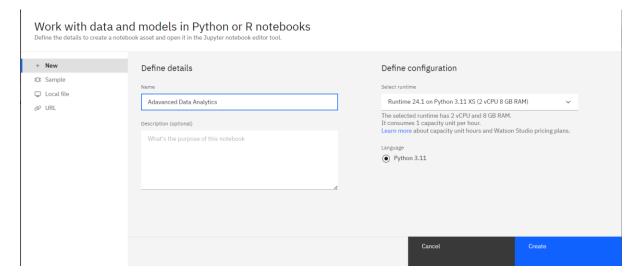
What do you want to do?

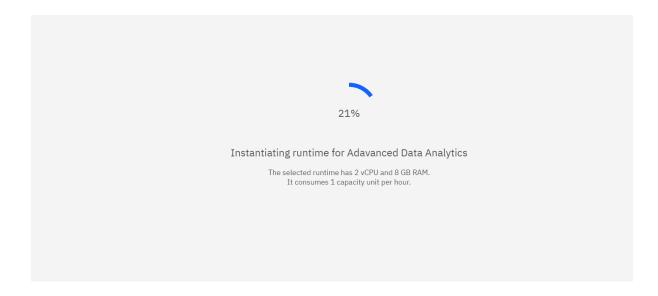
Select a task based on your goal. You'll use a tool to create an asset for that goal.



Step 7: Provide Notebook Details

- 1. **Name Your Notebook**: Enter a name for your Jupyter Notebook (e.g., "Iris Data Analysis").
- 2. **Select Environment**: Choose the appropriate Python environment if prompted.





Step 8: Install Required Libraries

1. **Install Libraries**: In the first cell of your notebook, run the following code to install necessary libraries:

```
Requirement already satisfied: pandas in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (2.1.4)
Requirement already satisfied: matplotlib in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (3.8.0)
Requirement already satisfied: seaborn in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (0.12.2)
Requirement already satisfied: numpyc2,>=1.23.2 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from pandas) (1.26.4)
Requirement already satisfied: python-dateutil>=2.8.2 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from pandas) (2.8.2)
Requirement already satisfied: python-dateutil>=2.8.2 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from pandas) (2.8.2)
Requirement already satisfied: tzdata>=2022.1 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from pandas) (2023.3)
Requirement already satisfied: contourpy>=1.0.1 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (1.2.0)
Requirement already satisfied: cycler>=0.10 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (4.51.0)
Requirement already satisfied: kiwisolver>=1.0.1 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (4.51.0)
Requirement already satisfied: packaging>=20.0 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (10.3.0)
Requirement already satisfied: packaging>=20.0 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (10.3.0)
Requirement already satisfied: packaging>=23.1 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (3.0.9)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-packages (from matplotlib) (3.0.9)
Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-RT24.1/lib/python3.11/site-pac
```

Step 9: Import Libraries

1. **Import Libraries**: In the next cell, import the libraries:

```
import seaborn as sns
import pandas as pd
```

Step 10: Load the Iris Dataset

1. **Fetch Dataset**: Load the Iris dataset using Seaborn:

```
# Load the Iris dataset
iris = sns.load_dataset('iris')
# Display the first few rows of the dataset
print(iris.head())
  sepal_length sepal_width petal_length petal_width species
               3.5
                        1.4 0.2 setosa
0
         5.1
         4.9
                   3.0
                               1.4
                                          0.2 setosa
1
                               1.3
                                          0.2 setosa
                   3.2
2
         4.7
                               1.5
3
         4.6
                    3.1
                                          0.2 setosa
4
          5.0
                    3.6
                               1.4
                                          0.2 setosa
```

Step 11: Explore the Data

1. Basic Information: Check the dataset's basic information and summary statistics:

```
# Display basic information about the dataset
print(iris.info())
# Describe the numerical features
print(iris.describe())
# Check for missing values
print(iris.isnull().sum())
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 # Column Non-Null Count Dtype
0 sepal_length 150 non-null float64
1 sepal_width 150 non-null float64
 2 petal_length 150 non-null float64
 3 petal_width 150 non-null float64
4 species 150 non-null object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
None
```

```
sepal_length sepal_width
                                  petal_length
                                                 petal_width
         150.000000
                     150.000000
                                    150.000000
                                                  150.000000
count
           5.843333
                        3.057333
                                      3.758000
                                                    1.199333
mean
           0.828066
                        0.435866
                                      1.765298
                                                    0.762238
std
min
           4.300000
                        2.000000
                                      1.000000
                                                    0.100000
25%
                                                    0.300000
           5.100000
                        2.800000
                                      1.600000
50%
           5.800000
                        3.000000
                                      4.350000
                                                    1.300000
75%
           6.400000
                        3.300000
                                      5.100000
                                                    1.800000
           7.900000
                        4.400000
                                      6.900000
                                                    2.500000
max
sepal_length
                0
sepal_width
petal_length
                0
petal_width
                0
species
                0
dtype: int64
```

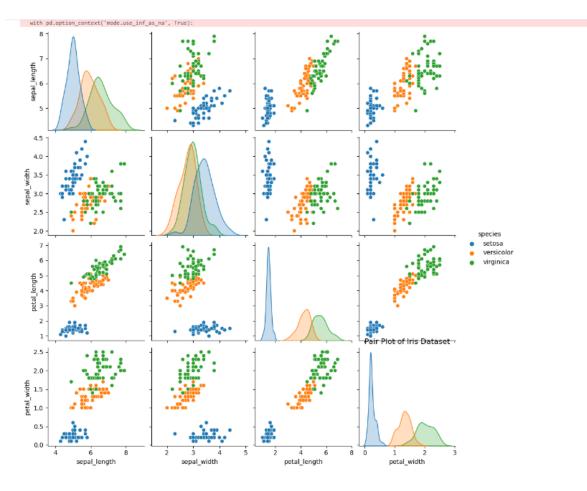
Step 12: Visualize the Data

1. Create Visualizations

Pair plot

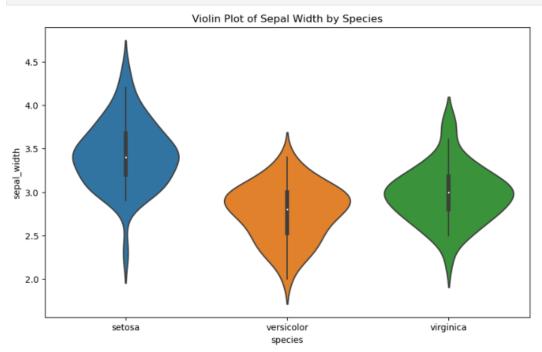
```
import matplotlib.pyplot as plt

# Pair plot to visualize relationships between features
sns.pairplot(iris, hue='species')
plt.title('Pair Plot of Iris Dataset')
plt.show()
```



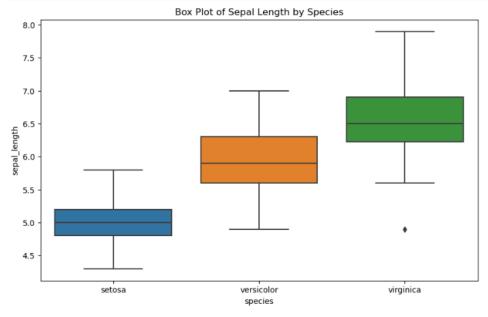
• Violin plot:

```
# Violin plot for Sepal Width by Species
plt.figure(figsize=(10, 6))
sns.violinplot(x='species', y='sepal_width', data=iris)
plt.title('Violin Plot of Sepal Width by Species')
plt.show()
```



• Boxplot:

```
# Box plot for Sepal Length by Species
plt.figure(figsize=(10, 6))
sns.boxplot(x='species', y='sepal_length', data=iris)
plt.title('Box Plot of Sepal Length by Species')
plt.show()
```



Step 13: Advanced Data Analysis

1. Group Analysis: Conduct group analysis to find averages by species

```
: # Group by species and calculate mean of each feature
  grouped_data = iris.groupby('species').mean().reset_index()
  print(grouped_data)
       species sepal_length sepal_width petal_length petal_width
  0
        setosa 5.006
                            3.428 1.462
                                                          0.246
    versicolor
                      5.936
                                  2.770
                                              4.260
                                                          1.326
     virginica
                      6.588
                                  2.974
                                              5.552
                                                          2.026
```