### Watson Studio: Machine Learning with SparkML

### Introduction

In this lab, we will explore machine learning using Spark ML. We will exploit Spark ML's high-level APIs built on top of DataFrames to create and tune machine learning pipelines. We will utilize Spark ML's feature transformers to convert, modify and scale the features that will be used to develop the machine learning model. Finally, we will evaluate and cross validate our model to demonstrate the process of determining a best fit model, load the results in the database, and save the model to the model repository.

We are using machine learning to try to predict records that a human has not seen or vetted before. We will use these predictions to sort the highest priority records for a human to look at. We will use as a training set for the algorithm simulated data that has been vetted by an analyst as high, medium or low.

#### **End-to-End Data Science**

The general flow of the End to End Data Science PoT will be guided by the activities shown in Figure 1- End to End Flow. This lab spans the Prepare Data, Build Model, and Save and Deploy activities.

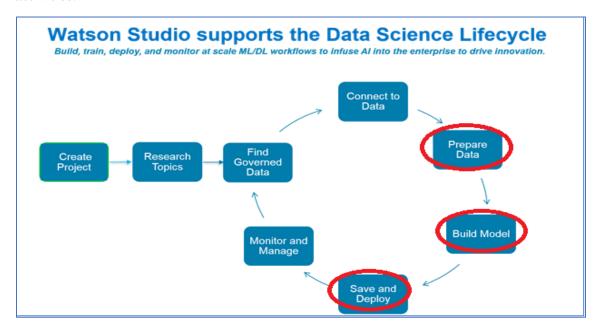


Figure 1- End to End Flow

# **Objectives**

Upon completing the lab, you will know how to:

- Join data from three sources.
- Identify labels and transform data.
- Conduct feature engineering for algorithm data.

- Declare a machine learning model.
- Setup the Pipeline for data transforms and training.
- Train the model.
- Evaluate and show model results
- Automatically tune model
- Score data and load into a new DB2 table.
- Save the model to the model repository.

# **Female Human Trafficking Data**

The data sets used for this lab consist of **simulated** travel itinerary data. The use case corresponds to an analyst reviewing the travel data to assign a risk of trafficking. The risk is recorded as the VETTING\_LEVEL column in the dataset. Some of the records have already been analyzed and have a VETTING\_LEVEL of low (value is 30), medium (value is 20), or high risk (value is 10). Others have not yet been vetted (value is 100). We will use the data that has been vetted to train a model to predict the risk for the unvetted records. This can be used to automate the process and augment the analyst. For example, one option would be to send the predicted high-risk persons to the analyst for further investigation.

The OCCUPATION data included in the travel data is very granular. For modeling purposes, it was decided to categorize the OCCUPATION data. Two additional datasets are used for this purpose. The occupation.csv dataset maps the granular occupation data to a category code. The categories dataset maps a category code to a category description. These datasets will be joined to the main dataset to prepare the data for modeling.

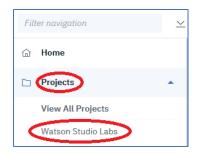
Other columns in the dataset are similarly very granular and could also be categorized for modeling purposes. This lab does not include steps to accomplish this, but it would be similar to what was done for the occupation column.

# Lab Steps

# Step 1 - Create a Jupyter Notebook

1. Click on the hamburger icon , then click on **Projects**, and then **Watson Studio Labs** (or whatever you named the project)

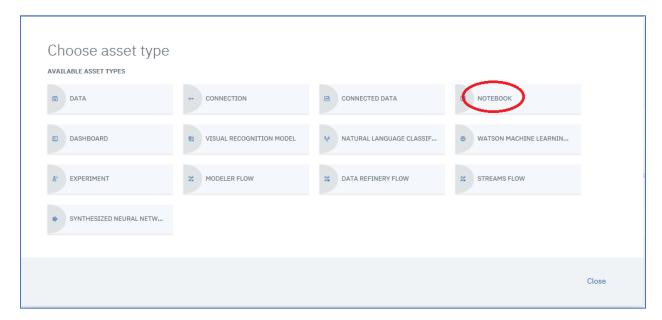




2. We are now going to create a notebook in our project. This notebook will be created from a url that points to the Machine Learning with SparkML notebook in the github repository. Click the **Add to project** link.



#### 3. Click on NOTEBOOK



4. Click on **From URL** under New Notebook, enter **Machine Learning with SparkML** for the **Name**, and optionally enter a **Description**.

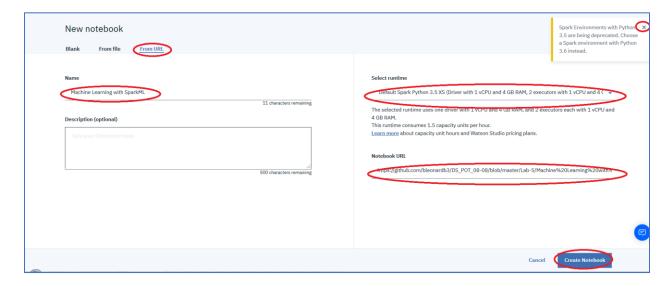
Select the Runtime. You will need to change the default selection. MAKE SURE TO SELECT Default Spark Python 3.5 XS (Driver with 1vCPU ...

Close the deprecation pop-up.

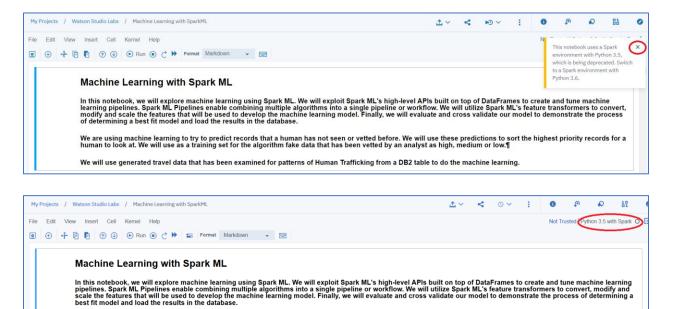
### Cut and paste the following url into the **Notebook URL** field.

https://github.com/bleonardb3/DS\_POT\_08-08/blob/master/Lab-5/Machine%20Learning%20with%20SparkML.ipynb

#### Click Create Notebook.



5. Close the deprecation pop-up message. Please make sure the notebook has Python 3.5 with Spark in the top right corner.



6. A Jupyter notebook consists of a series of cells. These cells are of 2 types (1) documentation cells containing markdown, and (2) code cells (denoted by a bracket on the left of the cell) where you write Python code, R, or Scala code depending on the type

We are using machine learning to try to predict records that a human has not seen or vetted before. We will use these predictions to sort the highest priority records for a human to look at. We will use as a training set for the algorithm fake data that has been vetted by an analyst as high, medium or low.

We will use generated travel data that has been examined for patterns of Human Trafficking from a DB2 table to do the machine learning.

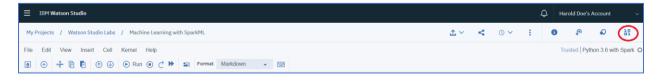
of notebook. Code cells can be run by putting the cursor in the code cell and pressing **<Shift><Enter>** on the keyboard. Alternatively, you can execute the cells by clicking on **Run icon** on the menu bar that will run the current cell (where the cursor is located) and then select the cell below. In this way, repeatedly clicking on **Run** executes all the cells in the notebook. When a code cell is executed the brackets on the left change to an asterisk '\*' to indicate the code cell is executing. When completed, a sequence number appears.

### **Step 2: Insert Generated Code**

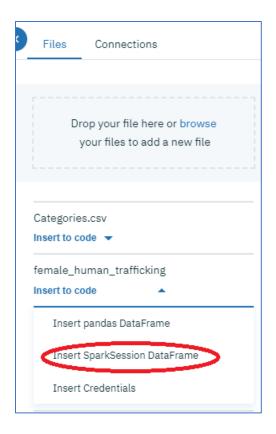
1. Before executing the cells in the notebook, we are going to use the IBM value-add code generator to insert code in 4 code cells. Scroll down in the notebook until you see **Read Data Asset- female\_human\_trafficking – See Lab Instructions** and put the cursor in the code cell underneath the comment lines. (Comments begin with the # sign).



2. Click on the 1/0 icon. at the top right.



3. Click on the insert to code down arrow trafficking and click on insertSparkSession DataFrame.



4. Locate the variable df\_data\_n (n is a number). This is a generated variable. We need to change this variable name to **trafficking\_df**.

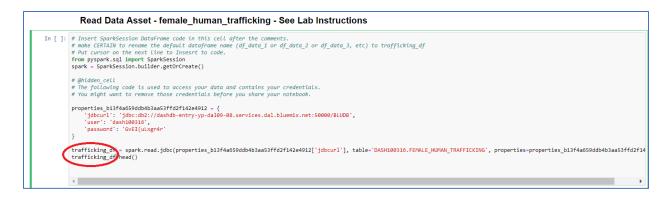
```
Read Data Asset - female_human_trafficking - See Lab Instructions

In []: # Insert SparkSession DataFrame code in this cell after the comments.
# make CERTAIN to rename the default dataFrame name (df_data_1 or df_data_2 or df_data_3, etc) to trafficking_df
# Put curson on the next line to Insert to code.
from pyspark.sql import SparkSession. builder.getOrCreate()
# @hidden_cell
# The following code is used to access your data and contains your credentials.
# You might wont to remove those credentials before you share your notebook.

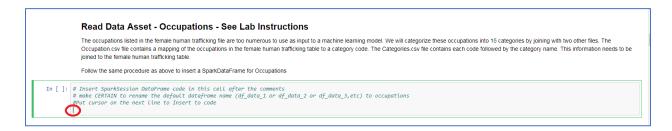
properties_b13f4a669ddbb3aa53ffd2f142e4912 = {
    'jdbcurl': 'jdbc:db2://dashdb-entry-yp-dal09-08.services.dal.bluemix.net:50000/BLUDB',
    'user': 'dashl00316',
    'password': 'GVEI{ulxgr4r'}
}

data_df_1_spark.read.jdbc(properties_b13f4a659ddb4b3aa53ffd2f142e4912['jdbcurl'], table='DASH100316.FEMALE_HUMAN_TRAFFICKING', properties_b13f4a659ddb4b3aa53ffd2f142e4912
| data_df_1_bead()
```

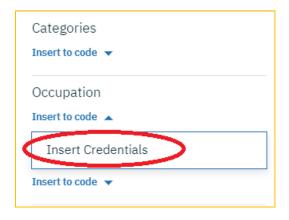
Change to.



5. Scroll down to **Read Data Asset – Occupations – See Lab Instructions.** Click cursor underneath the commented lines in the code cell.



6. Click on Click on Credentials.



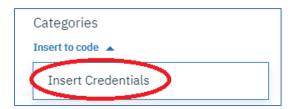
7. Locate the variable **credentials**. Make sure the variable does not have a number appended (e.g. **credentials\_1** or **credentials\_2** or **credentials\_3**, etc). If it does, change the variable to be **credentials** (without a number).

#### Change to:

8. Scroll down to **Read Data Asset – Categories – See Lab Instructions.** Click cursor underneath the commented lines in the code cell.

```
In []: # Insert Categories credentials in this cell after the comments
# make CERTAIN that if credentials gets generated as credentials_1 or credentials_2 or credentials_n where n is a number to rename as credentials
#Put cursor on the next line to Insert to code
```

9. Click on Credentials. Click on Insert Credentials.



10. Locate the variable **credentials**. Make sure the variable does not have a number appended (e.g. **credentials\_1** or **credentials\_2** or **credentials\_3**, etc). If it does, change the variable to be **credentials** (without a number).

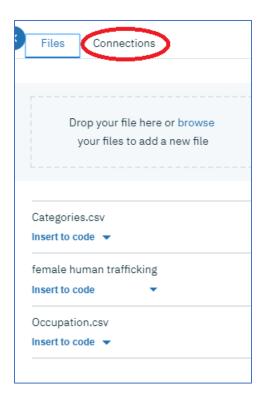
#### Change to:

11. Scroll down further towards the middle of the notebook until you see **Insert the** database credentials – see Lab Instructions. Click cursor underneath the commented lines in the code cell.

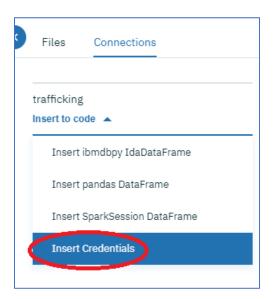
```
Insert the database credentials - see Lab Instructions

In []: # Insert database connection credentials below Make sure the name that is used is credentials. If credentials_1 is shown, please change to credentials.
```

12. Click on Connections.



13. Click on Credentials. Click on Insert to code down arrow underneath trafficking. Click on Insert



14. Locate the variable **credentials**. Make sure the variable does not have a number appended (e.g. **credentials\_1** or **credentials\_2** or **credentials\_3**, etc). If it does, change the variable to be **credentials** (without a number).

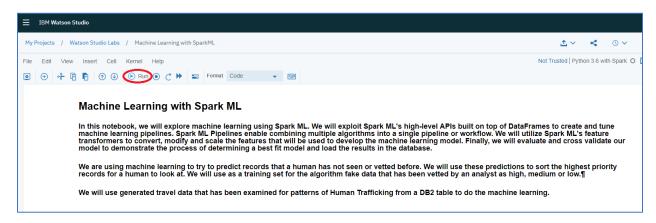
```
In [ ]: # Insert database connection credentials below
# Make sure the name that is used is credentials. If credentials_1 is shown, please change to credentials.
# @hidden_cell
# The following code contains the credentials for a connection in your Project.
# You might want to remove those credentials before you share your notebook.

credentials_2 {
    "username': 'dash100316',
        'password': """GVEI{uLxgr4r""",
        'sg_service_url': 'https://sgmanager.ng.bluemix.net',
        'database': 'BLUDB',
        'host': 'dashdb-entry-yp-dal09-08.services.dal.bluemix.net',
        'port': '50000',
        'url': 'https://undefined'
}
```

### Change to:

### Step 3: Execute the code cells in the notebook

1. Scroll back to the top of the notebook. Execute each of the code cells in order by clicking into each code cell starting at the top and pressing the **Shift>Enter>** keys or by clicking into the first code cell and using the Run icon in the menu bar at the top. Read the documentation to gain an understanding of the code that is executing.



- ✓ Joined data from three sources.
- ✓ Identified labels and transformed data.
- ✓ Conducted feature engineering for algorithm data.
- ✓ Declared a machine learning model.
- ✓ Created the Pipeline for data transforms and training.
- ✓ Trained the model.
- ✓ Evaluated and showed model results.
- ✓ Automatically tuned model.
- ✓ Scored data and loaded into a new DB2 table.
- ✓ Saved the model to the model repository.