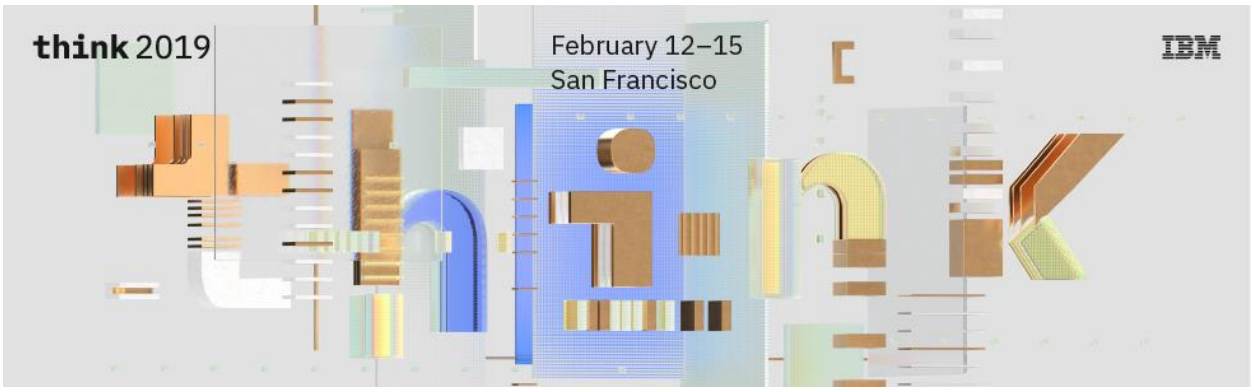

Lab Center – Hands-on Lab

Session #3259

Session Title: Introduction to Data Science using Watson Studio

Lab-2: Deploy a Continuous Machine Learning Model using Watson Studio

Bernard Beekman, IBM, beekmanb@us.ibm.com

Michael Cronk, IBM, michael.cronk@ibm.com

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Introduction

In this lab, you will be using 2017 Chicago building data to make Chicago a safer place by building a model to predict when buildings are likely to fail inspection. You can then use the model to find which buildings are most dangerous and attend to those first. Continuous learning will be set up to monitor the model performance and adjust if necessary.

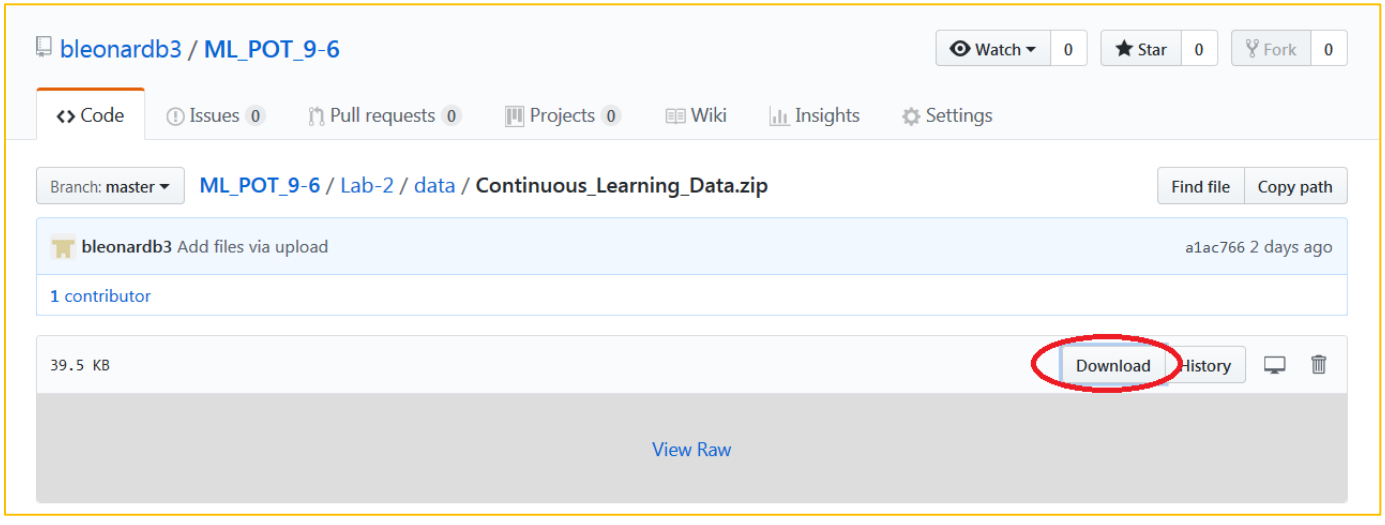
Goals:

- Upload data to your Watson Studio project.
- Train, compare, and select a machine learning model.
- Set up continuous learning capabilities.
- Deploy a machine learning model.

Step 1 Upload a New Data Asset.

Before we build our models, we need to load data into our project.

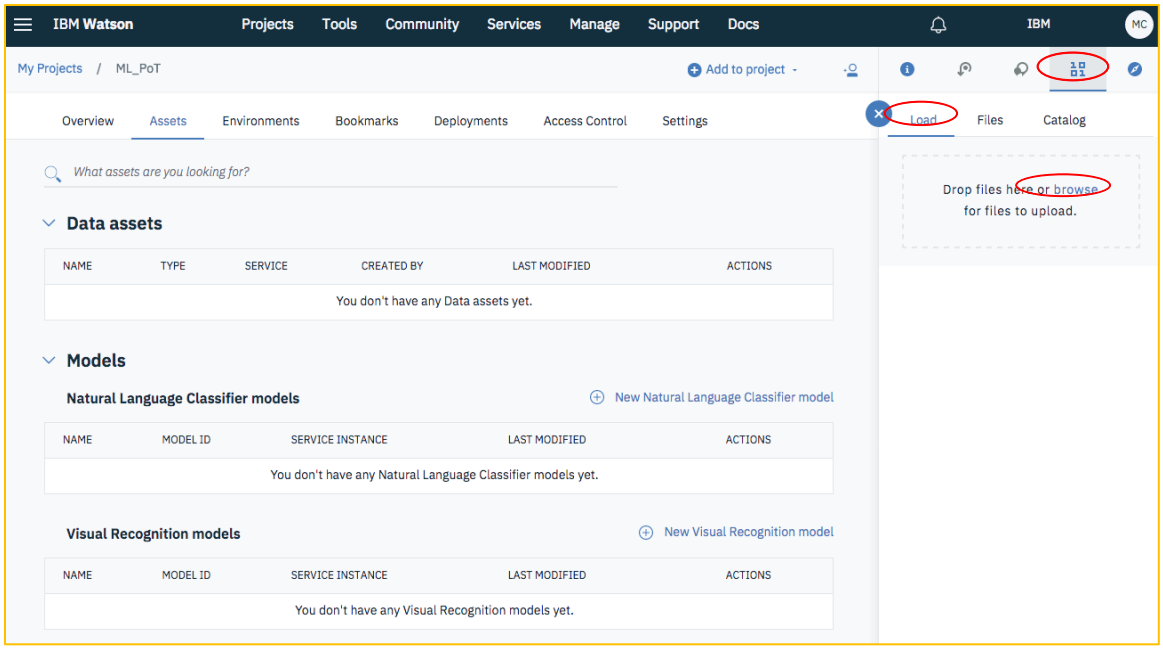
1. Click on Building Data
2. Click on Download.



3. Save the zip file to your desktop and unzip the contents. This zip file contains data for building inspections for each month. We will begin by training our model for the month of September and then add October and November data to show how we can continuously adjust over time.
4. Return to your assets page. From Watson Studio, you can click on the Projects

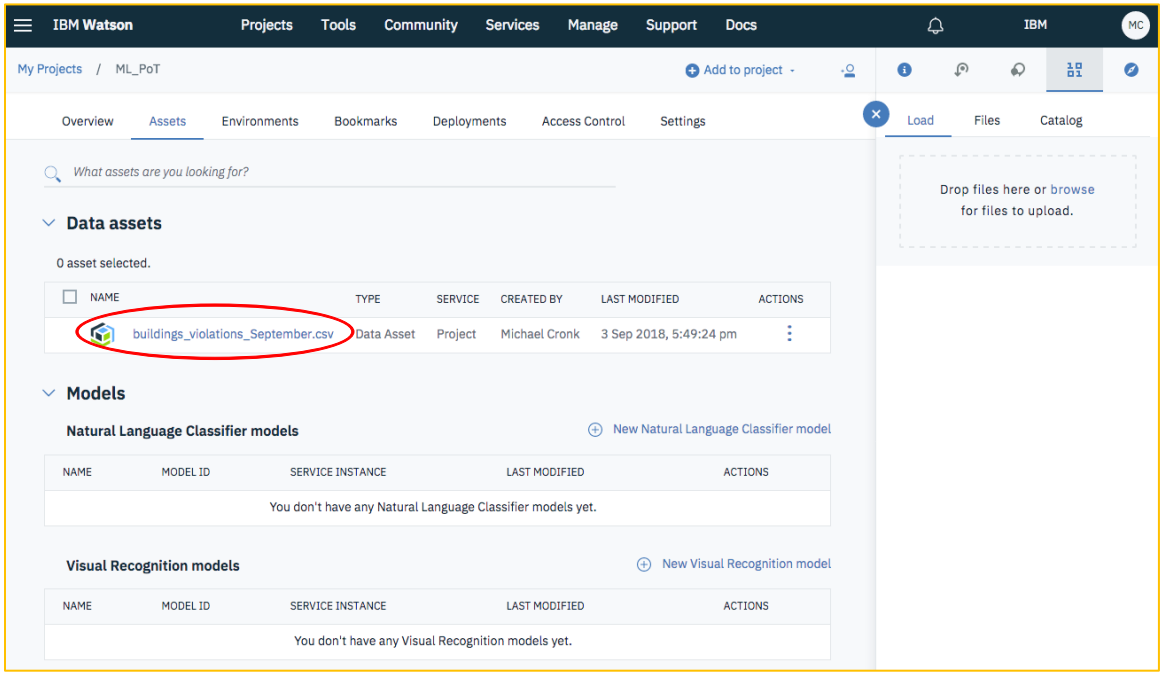
dropdown on the top navbar and then click on the name of your project.

5. In the top right of the screen, click the blue icon with 1's and 0's. This will open the data pallet.
6. Click on Load, browse, select the building_violations_September.csv file and click open.

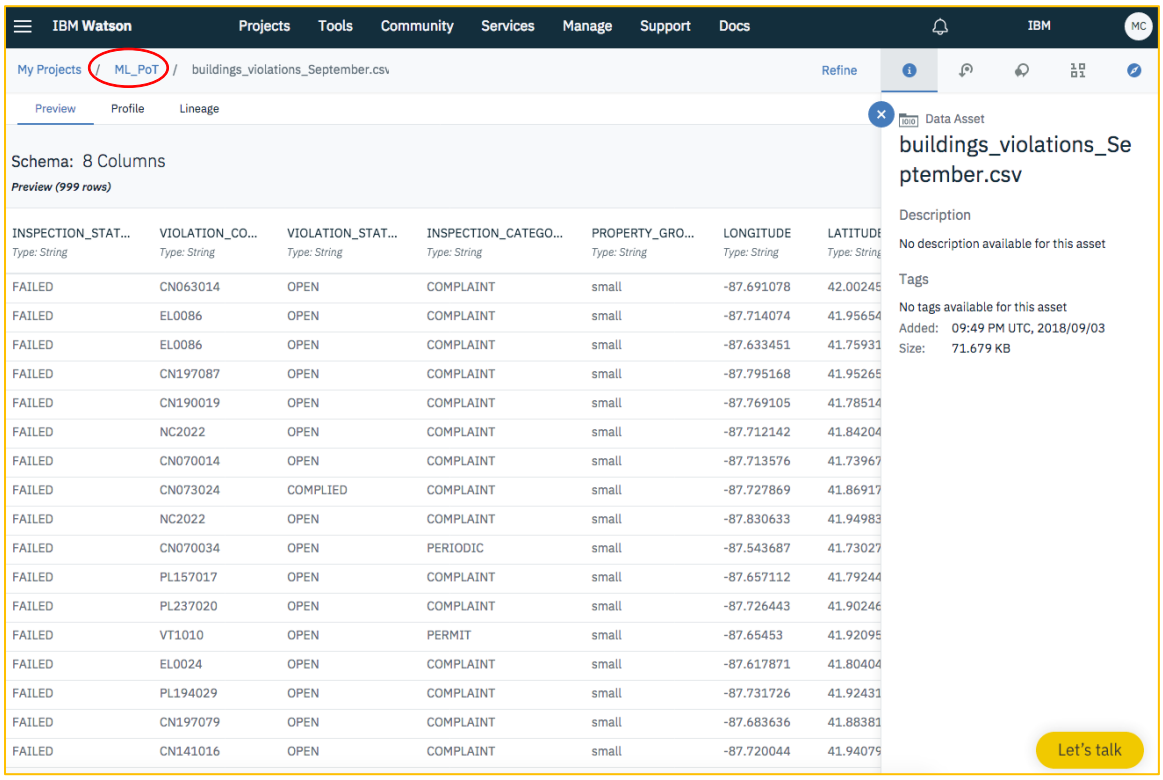


The CSV file should now be listed under "Data assets." It is now accessible by the Watson Studio modeling tools and applications.

7. Click on building_violations_September.csv under Data assets.

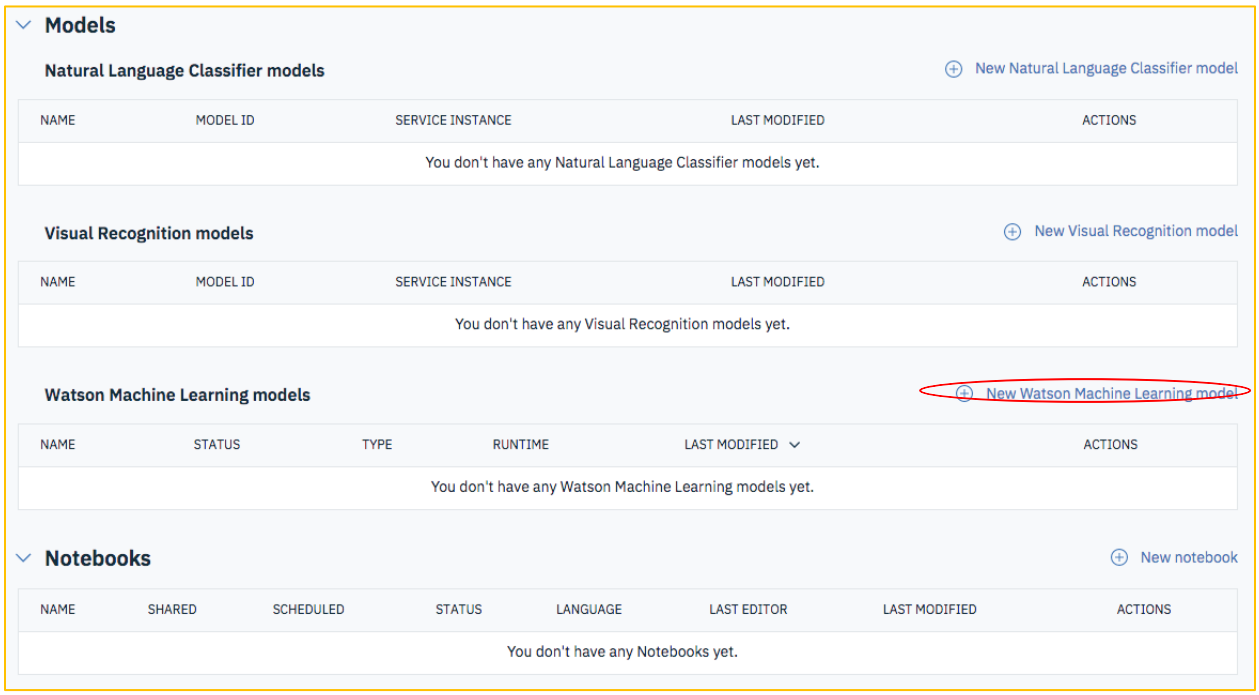


After clicking on building_violations_September.csv, you are brought to a data view. Once you are satisfied that this is the data you want, click on your project name to return to your project's assets.

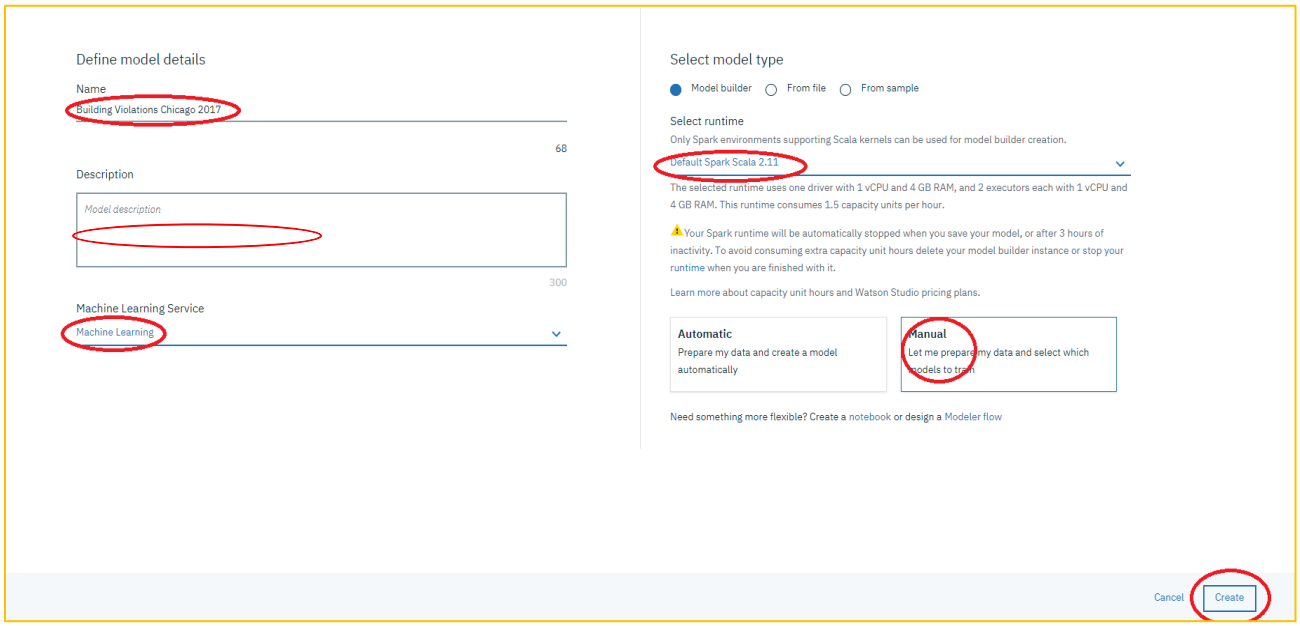


Step 2 Train, Compare, and Select a Machine Learning Model

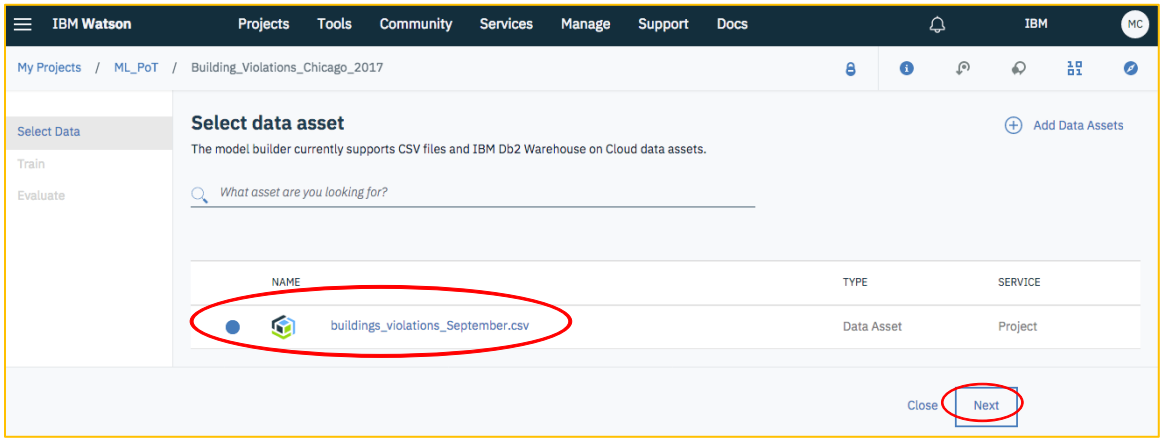
1. Scroll down and click New Watson Machine Learning model.



2. Enter a Name and a Description. Note that the Machine Learning Service should already be filled in from the prerequisites. Select the Default Spark Scala environment, Select Manual, and click Create. We are creating a model using "Model builder" but can also create models in notebooks as well as with SPSS or Deep Learning flow modelers.

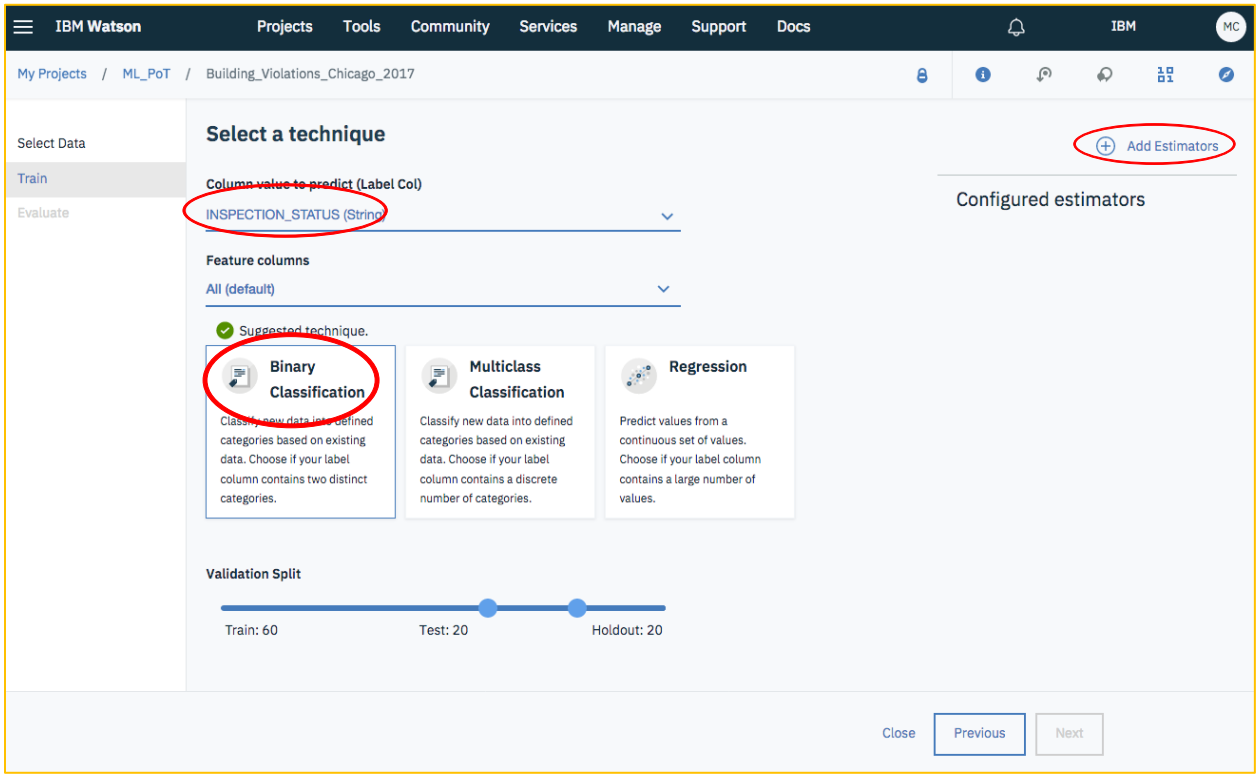


3. You will be brought to a "Select data asset" page. Select building_violations_September.csv as our data asset and click Next.



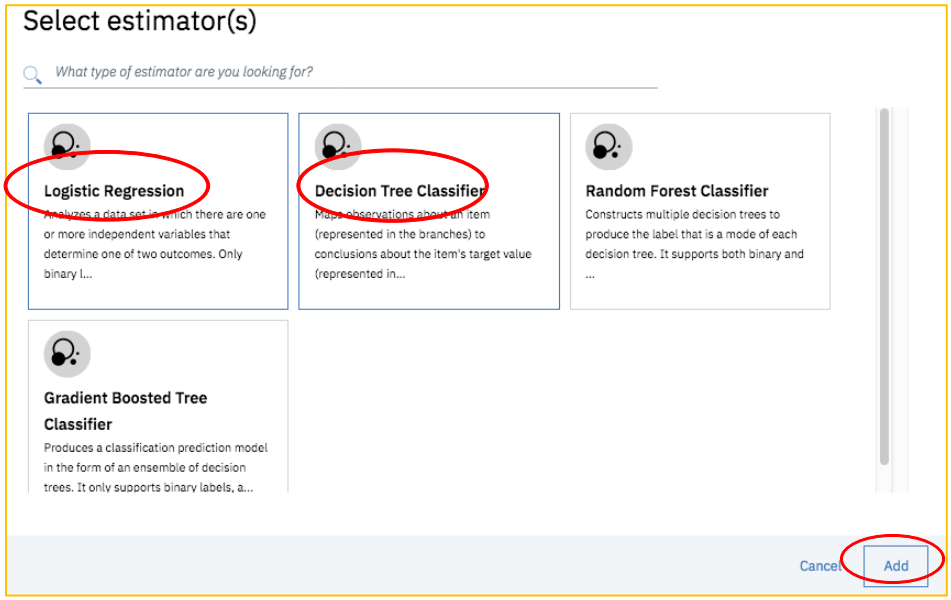
We will now choose which techniques may work well given our data and given our goal to predict which buildings are most likely to not pass inspection.

4. On the “Select a technique” page, click on the Select Label Col dropdown and select INSPECTION_STATUS(String).
5. Select Binary Classification. Note that this is often automatically suggested by Watson Studio.
6. Click Add Estimators in the upper right corner of the page.



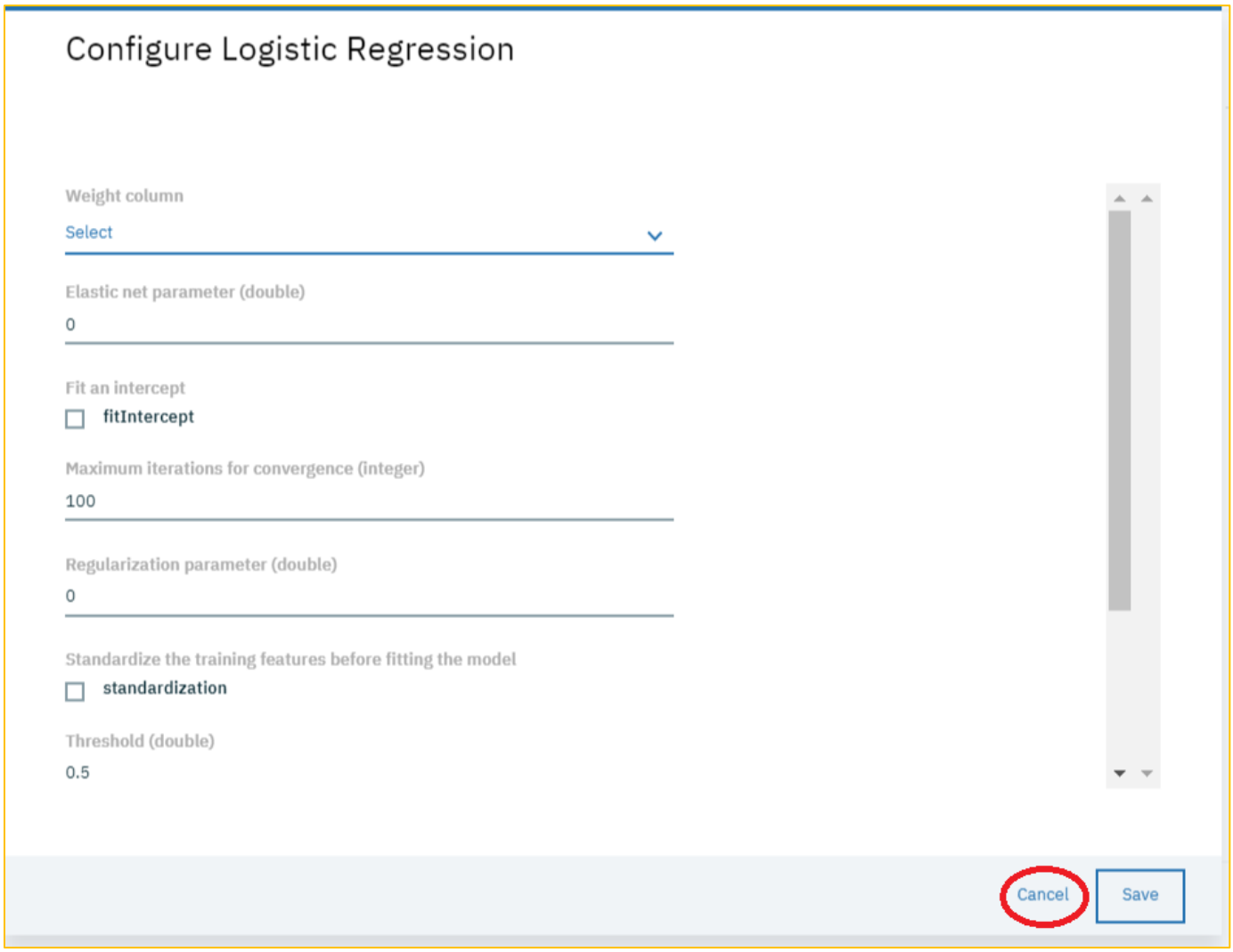
Watson Studio only presents the estimators that would work with our selections on the previous page.

7. Select Logistic Regression and Decision Tree Classifier.
8. Click Add.

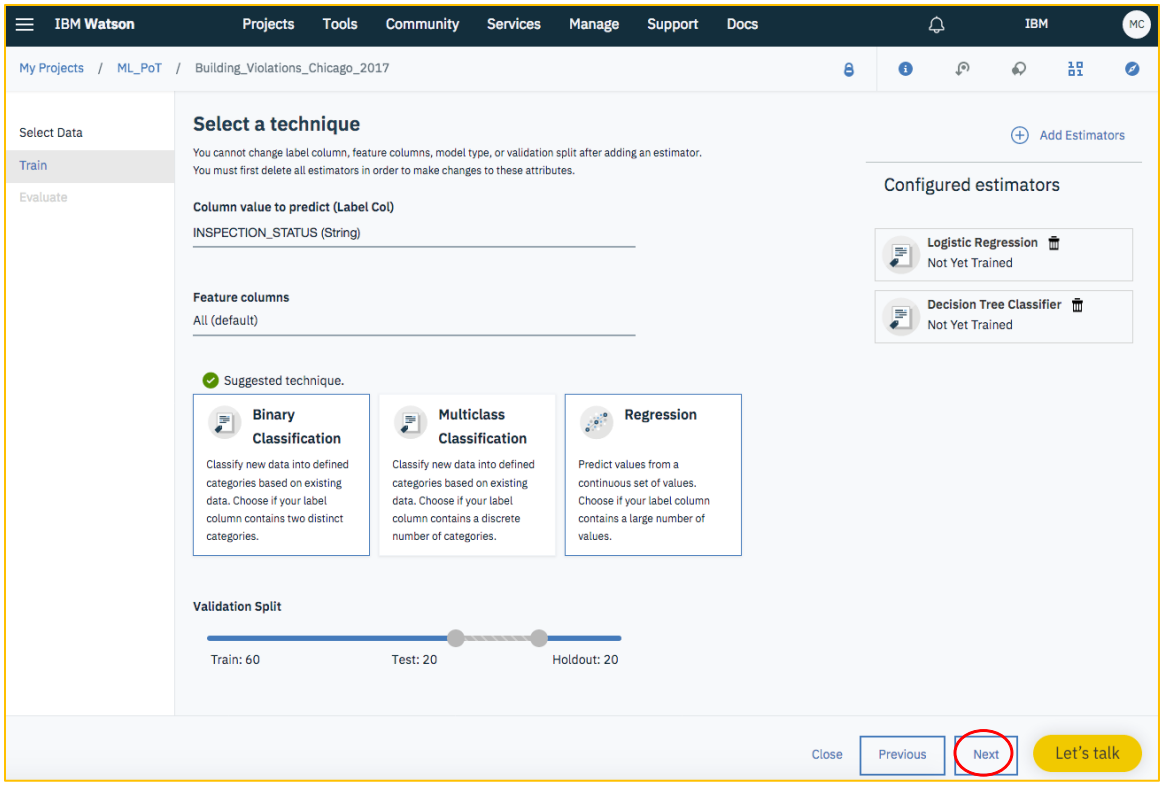


9. Note that you can click on the configured estimators to tune the hyperparameters associated with the algorithm. Click on Logistic Regression.

10. We will not change any hyperparameters, so click on Cancel.

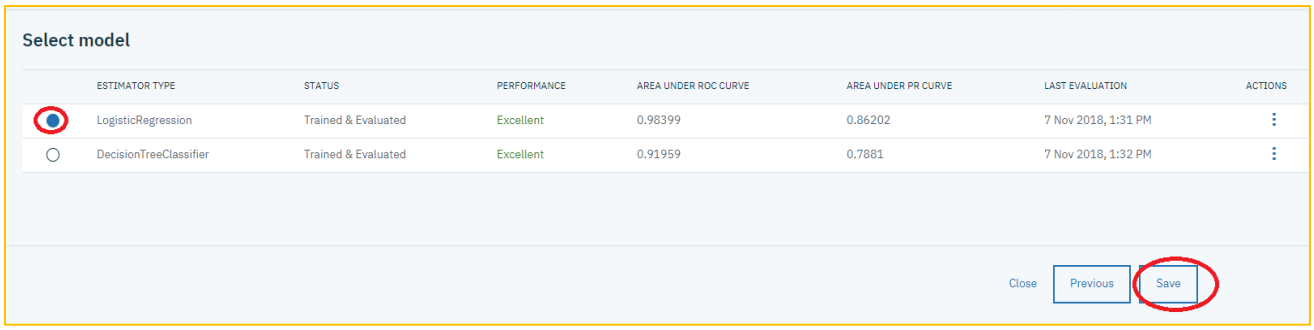


11. Back on the “Select a technique” page, click Next.

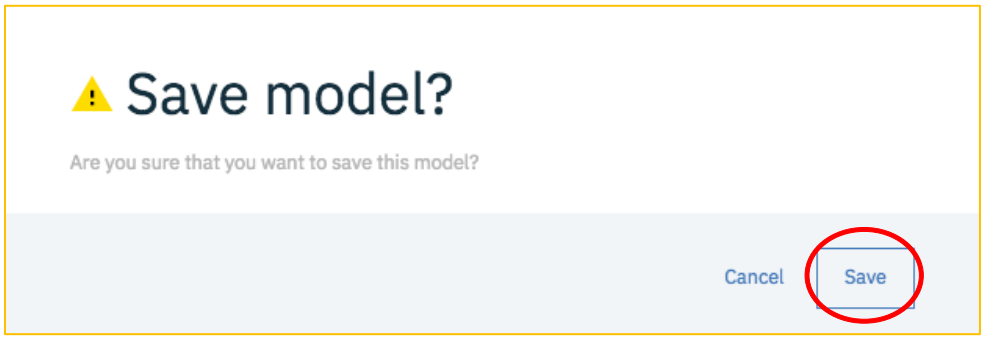


The Select model page will allow us to compare the results of different estimator types.

12. Both models did well. The Logistic Regression model performed slightly better. Select Logistic Regression.
13. Click Save.

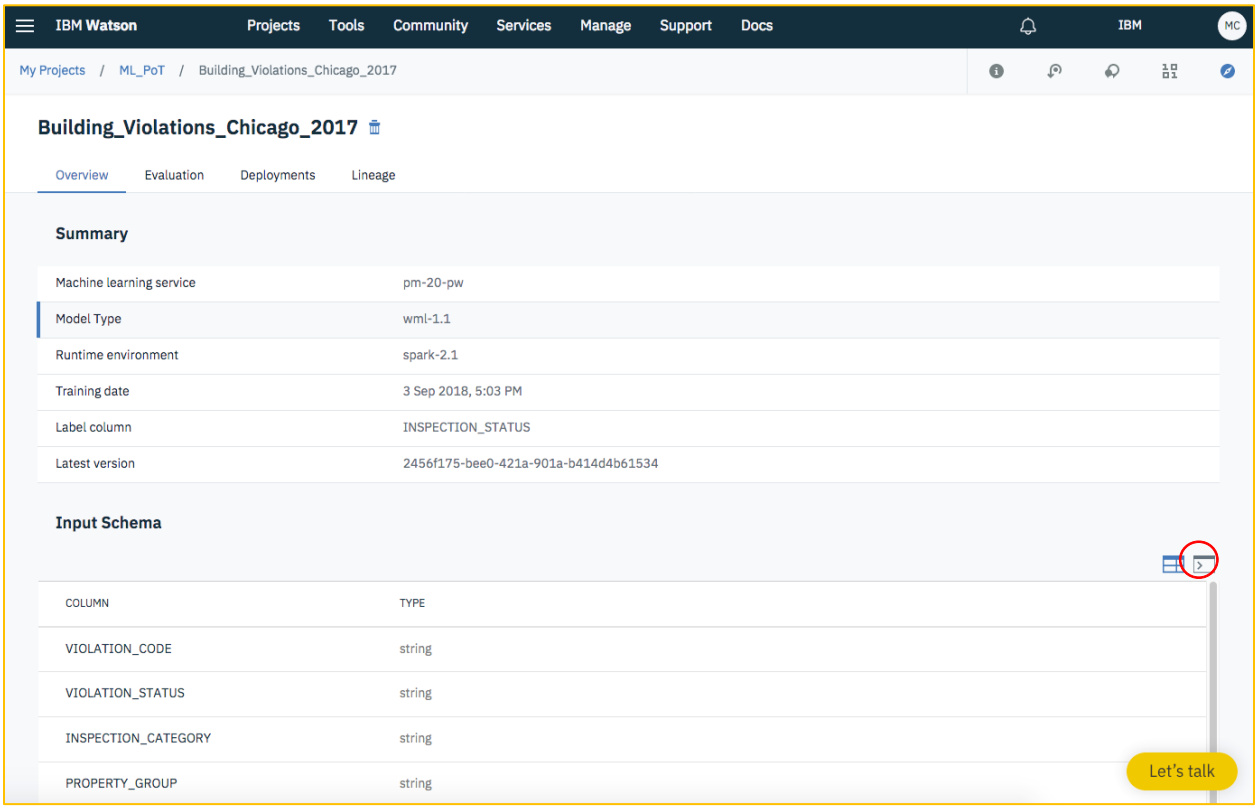


14. When the “Save model” window appears, click Save.

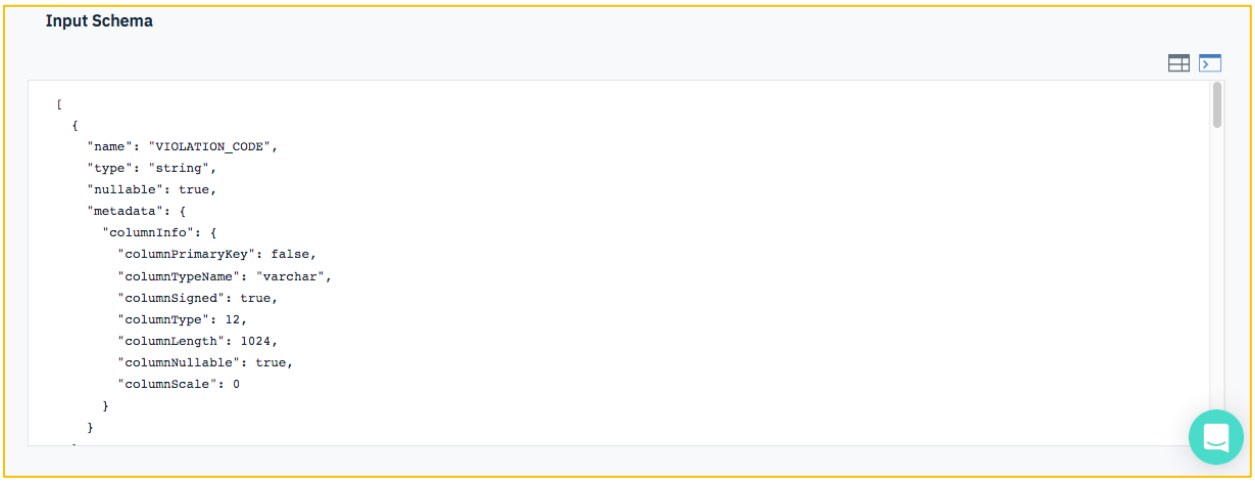


We have just saved the model in our Watson Machine Learning service and can now view information about the model specifications, details of construction, and input schema. The model can also be exposed as an API and used by data pipelines, applications, or other external systems.

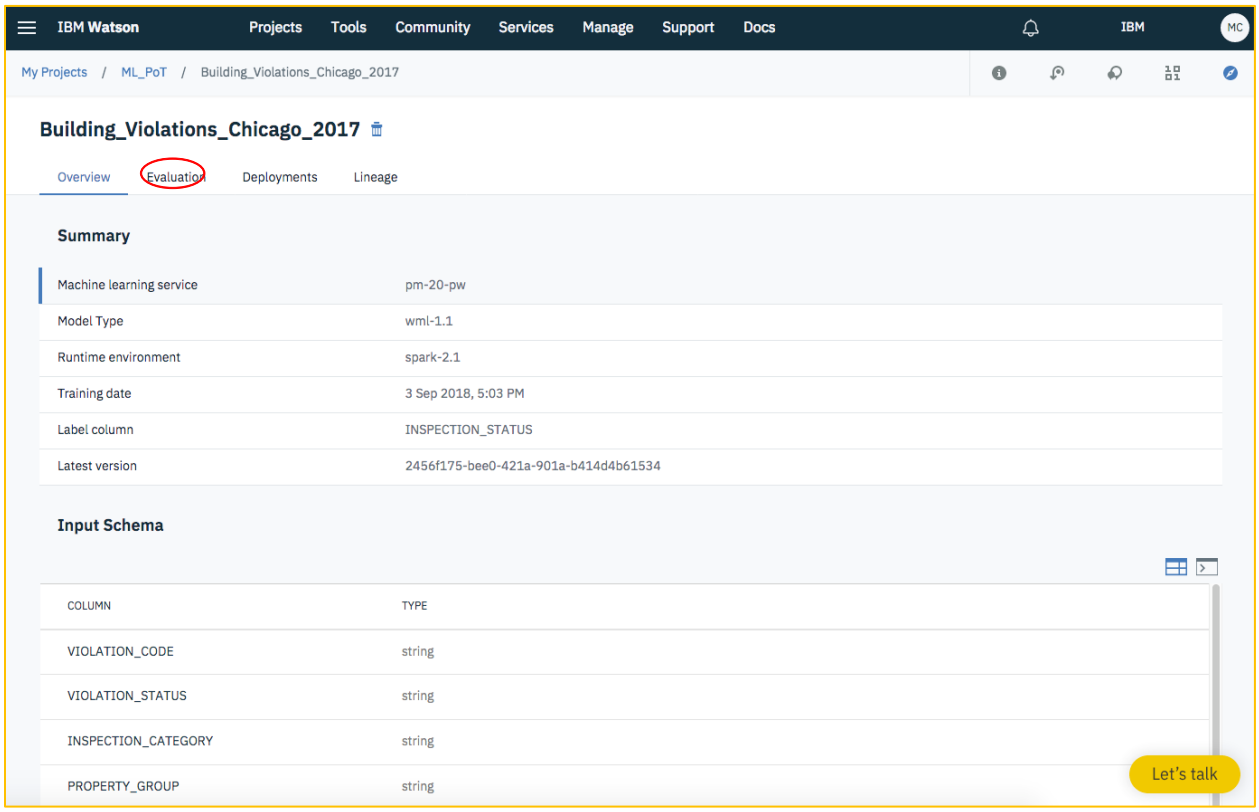
15. Click on the console button with an “angle bracket” to view the schema in JSON format.



This schema can be copied and used elsewhere to help existing systems easily interact with our model and the Watson Machine Learning service.



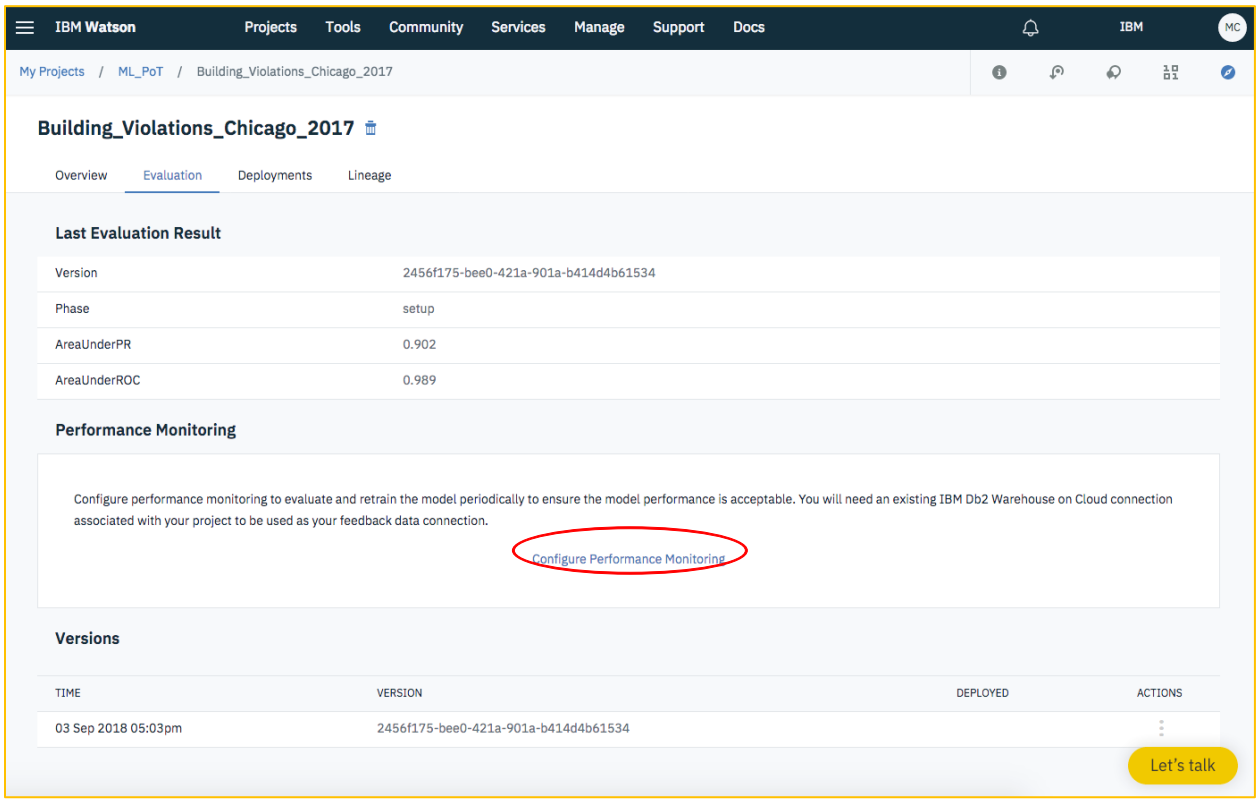
16. Scroll up and click on the Evaluation tab.



Step 3 Set up Continuous Learning Model Capabilities

Model performance is critical for solving data science problems. This page provides information and functionality to aid in continuous training, tuning, and redeployment. We will now set up parameters to automatically retrain our model when performance falls below a certain threshold.

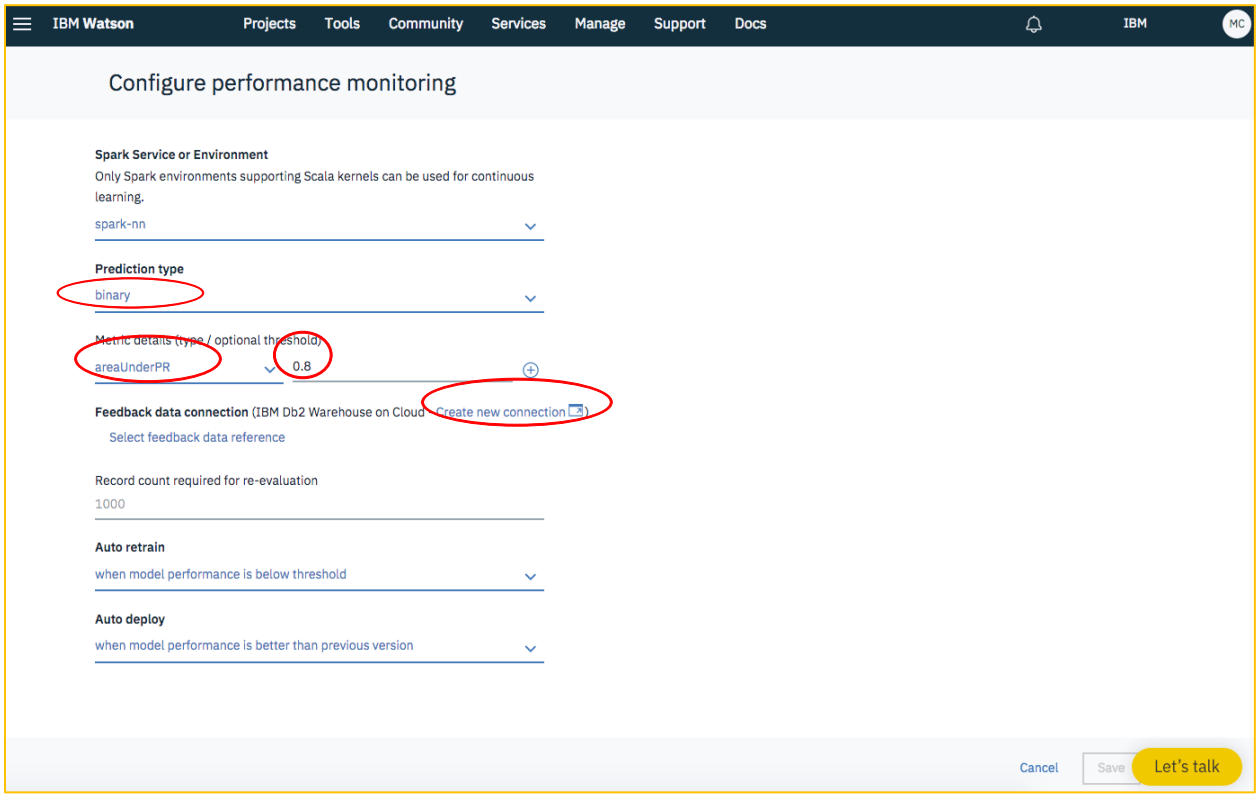
1. Click on Configure Performance Monitoring.



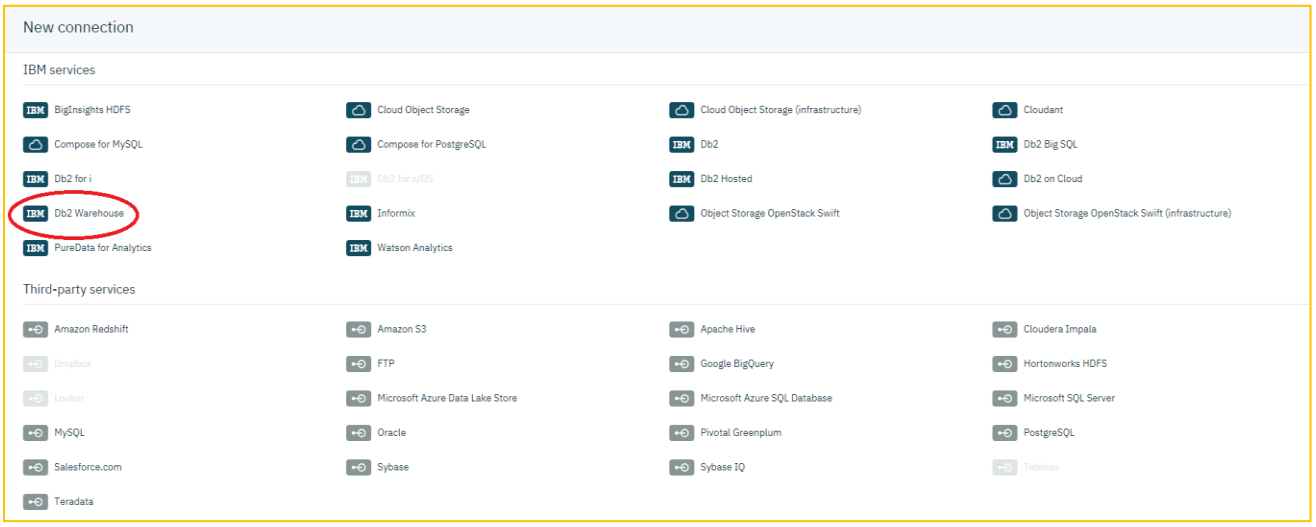
2. Under Prediction type, select binary.
3. Under Metric details, select areaUnderPR and enter 0.8.

This relies on a database table of new data. We will use DB2 Warehouse on Cloud, a data store optimized for analytic data sets.

4. Click on Create a new connection.



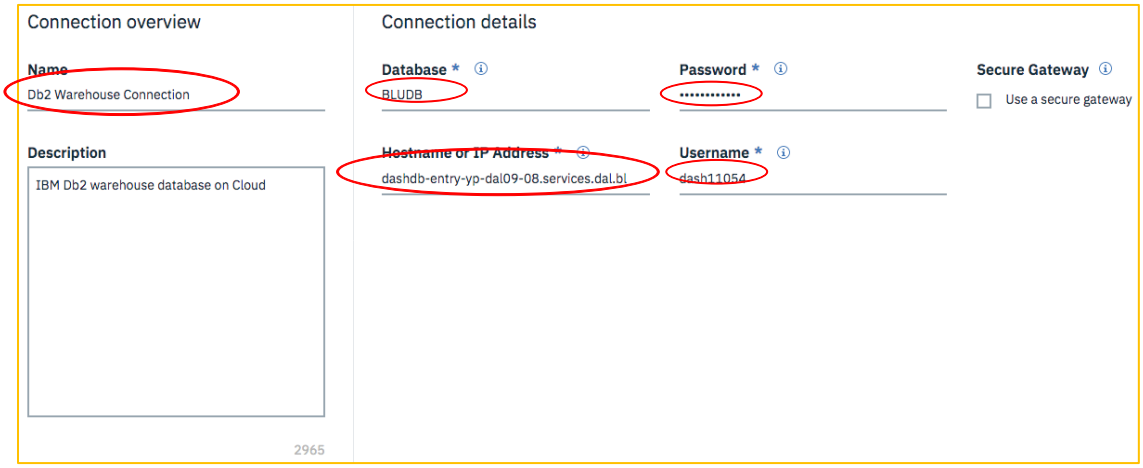
5. On the “New connection” page, click DB2 Warehouse.



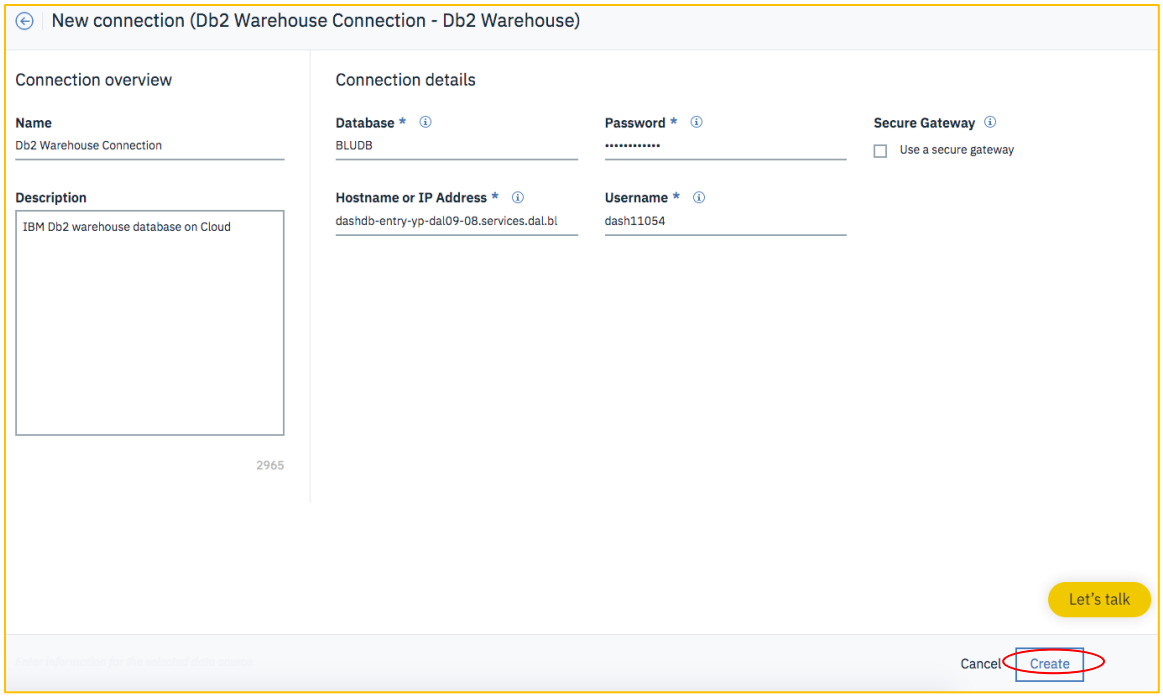
6. Access [DB2 Credentials]

Ask your instructor for your assigned database.

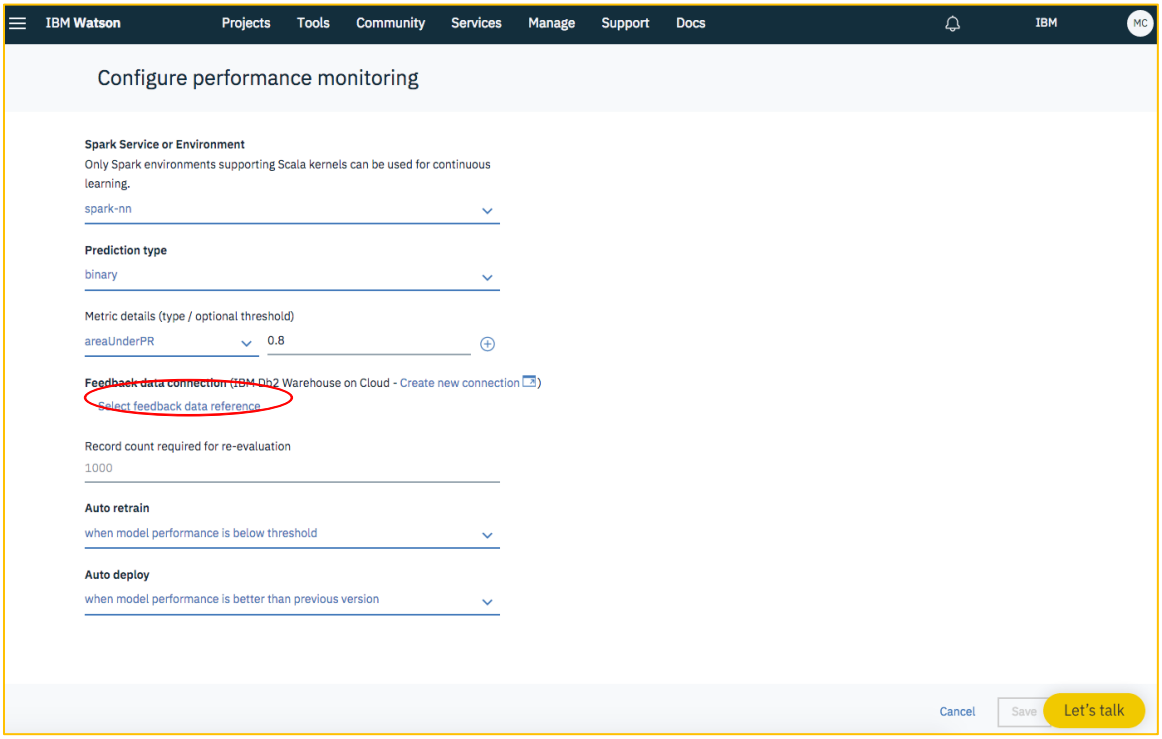
7. Enter a Name for your connection (e.g. DB2 Warehouse Connection).
8. Using the credentials cut and paste the fields under Hostname, Database, Username, and Password.



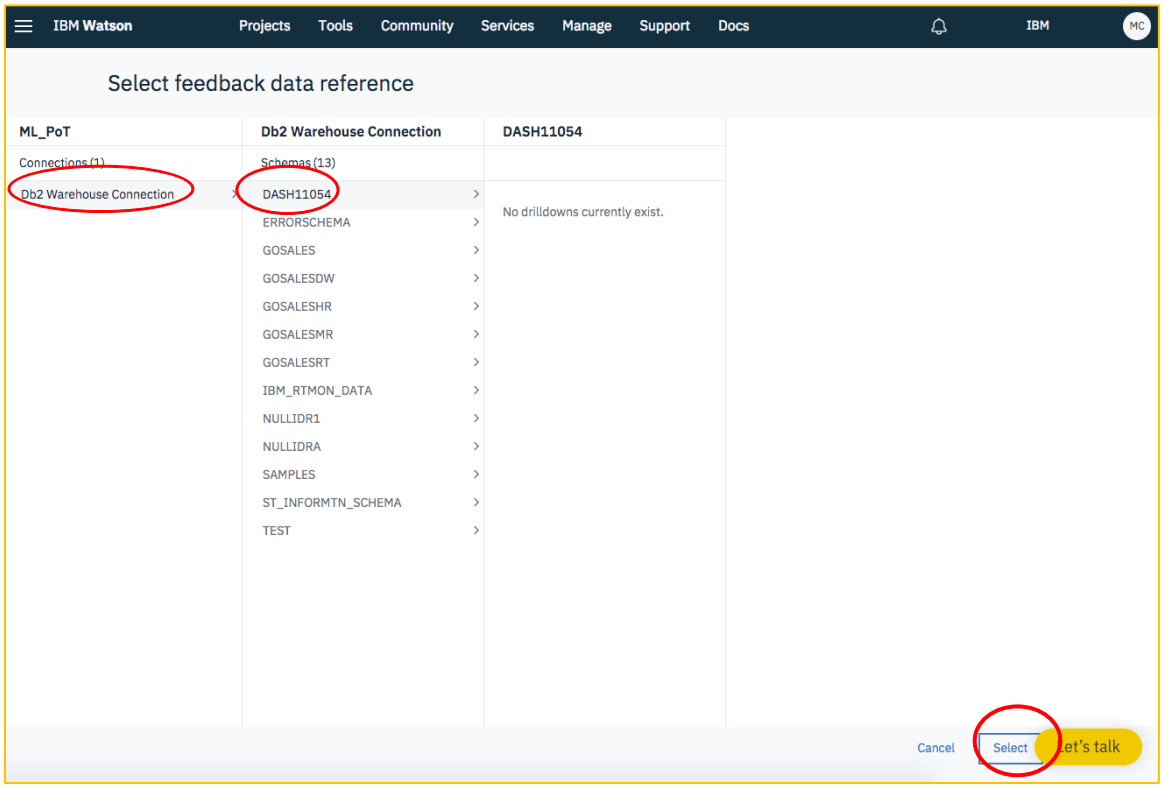
9. Click Create.


New connection (Db2 Warehouse Connection - Db2 Warehouse)
Connection overview
Name
Db2 Warehouse Connection
Description
IBM Db2 warehouse database on Cloud
2965
Connection details
Database *
BLUDB
Password *
Secure Gateway
Use a secure gateway
Hostname or IP Address *
dashdb-entry-yp-dal09-08.services.dal.bl
Username *
dash11054
Let's talk
Cancel
Create

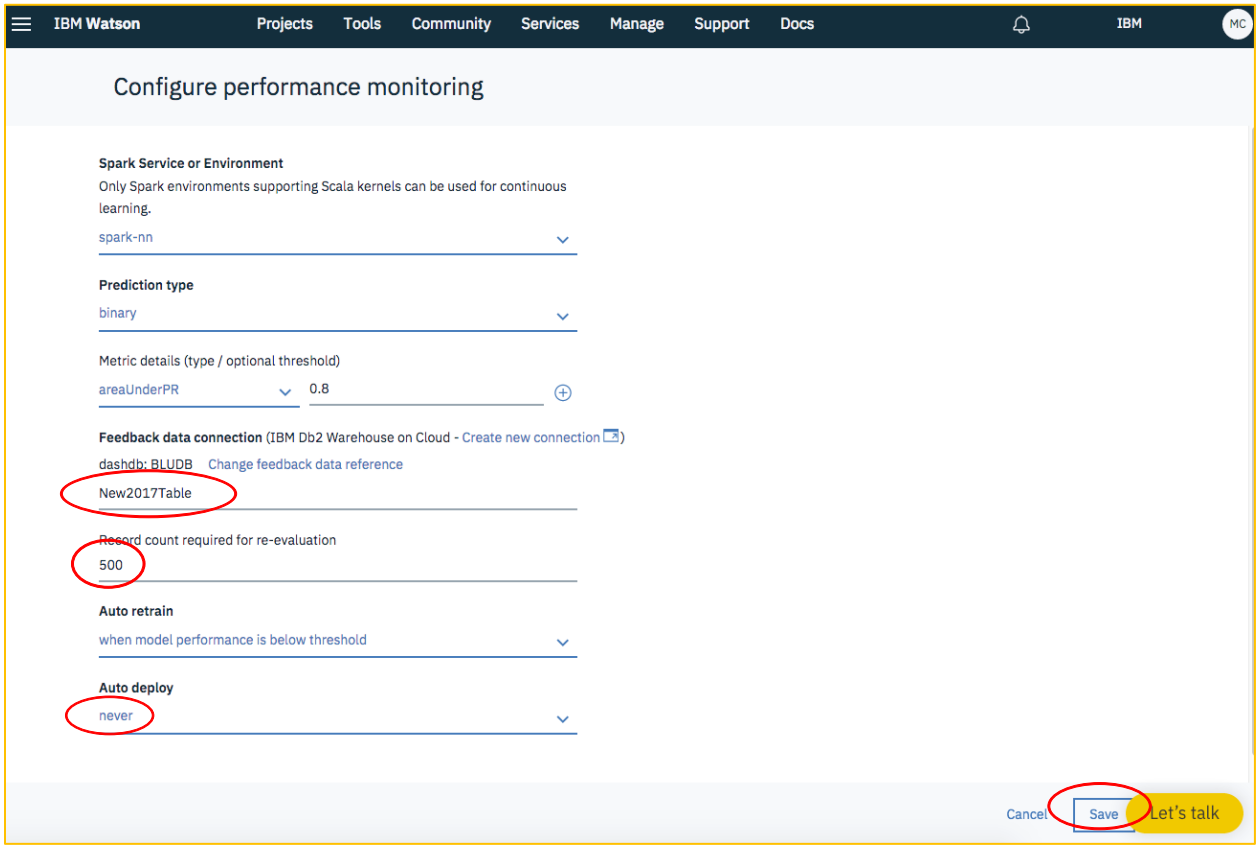
10. Return to the “Configure performance monitoring page” and click Select feedback data reference.



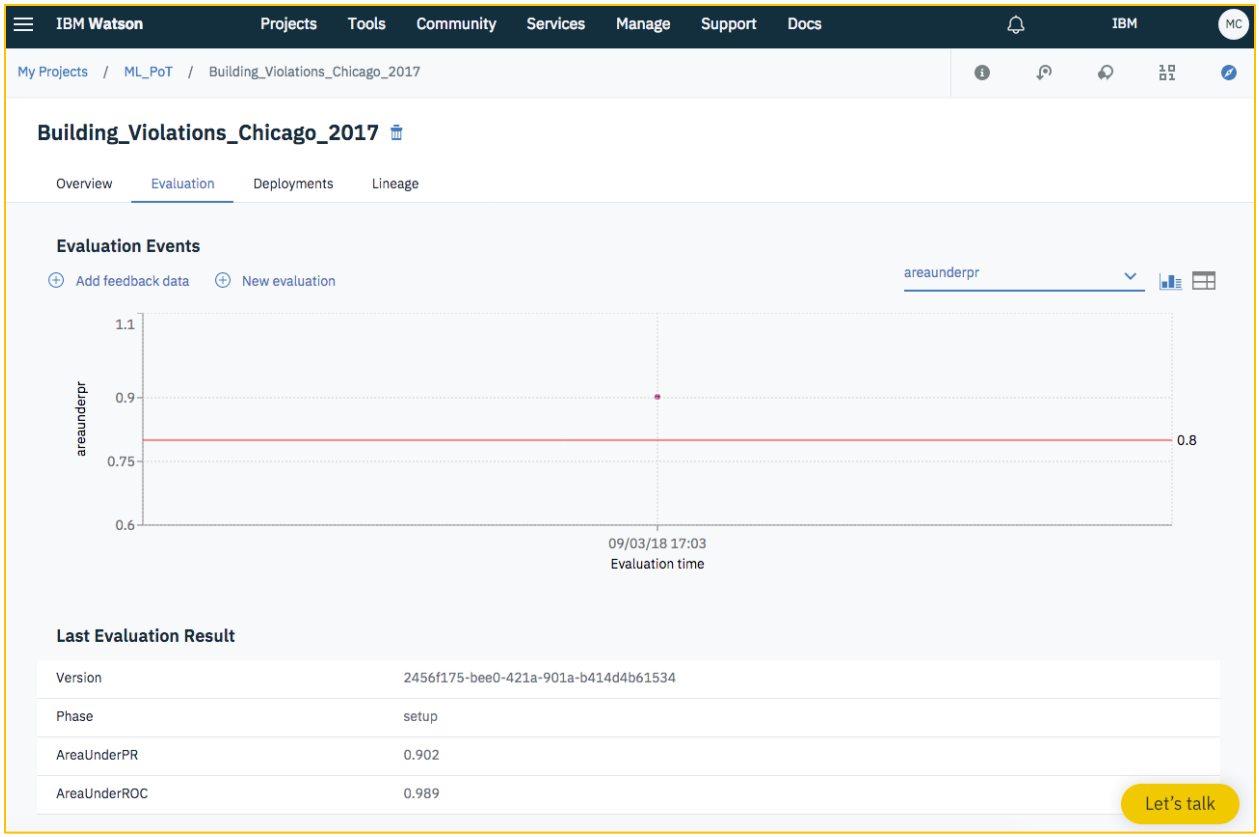
11. Click on your DB2 Warehouse-xx name.
12. Select the schema that matches DASH####.
13. Click select.



14. After returning to the “Configure performance monitoring” page, enter a unique table name (attendees are sharing the DB2 Warehouse service so make your name unique by appending your initials to the table name).
15. Enter 500 as the Record count.
16. Leave the default for Auto retrain. Under Auto deploy, select never.
17. Click Save.

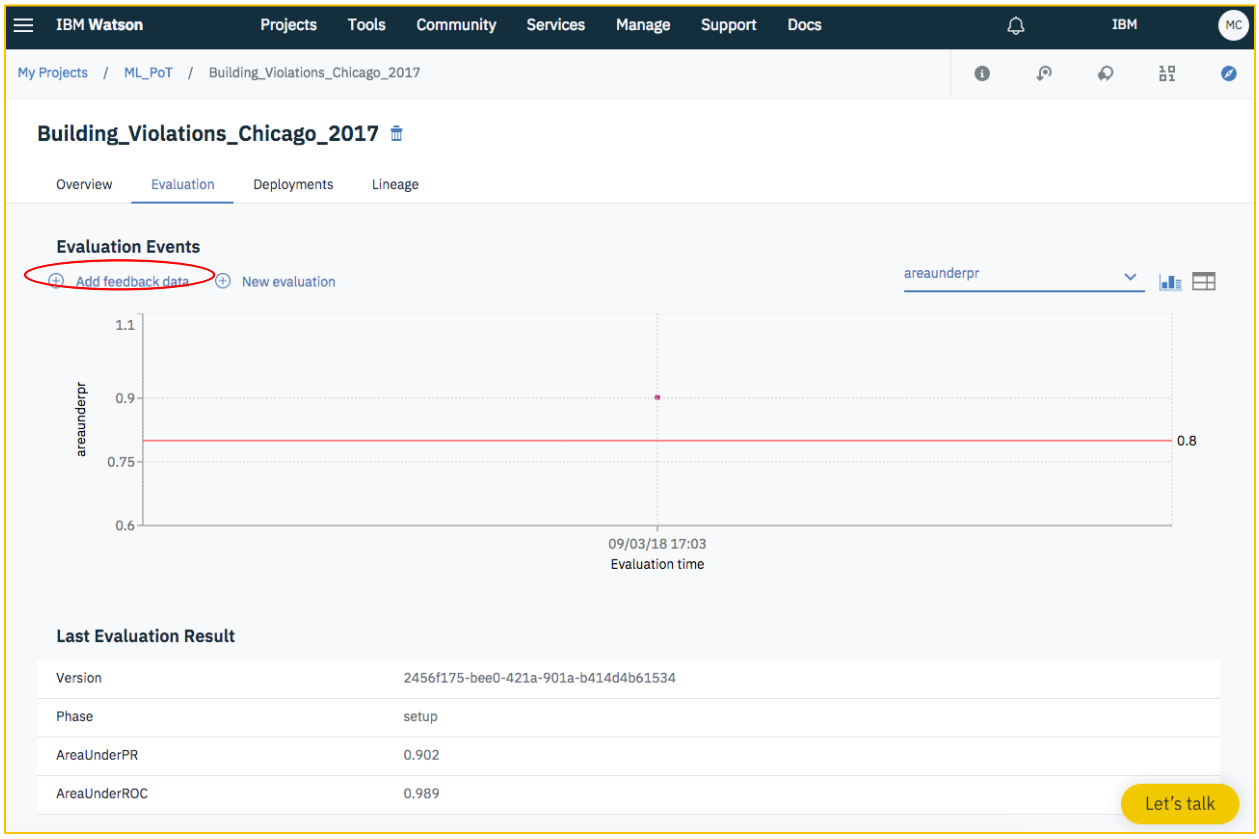


We have just set our model to retrain whenever its performance falls below 0.8. The evaluation screen has changed to reflect the decisions we have just made. Notice the graph lists our current areaunderPR and the threshold we have set. Clicking on Add feedback data will upload a new dataset. Clicking on new evaluation will then evaluate the model and check to see how the model compares to our 0.8 threshold.



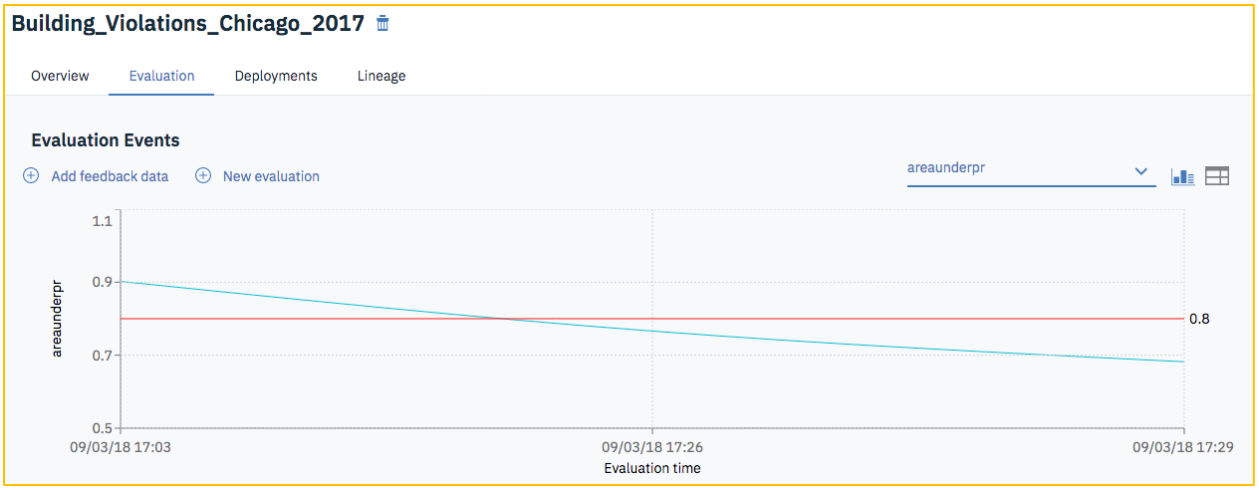
Suppose September has passed, and we now have October data. We can trigger a new evaluation by adding more feedback data.

18. Click Add feedback data.
19. Select building_violations_October.csv.

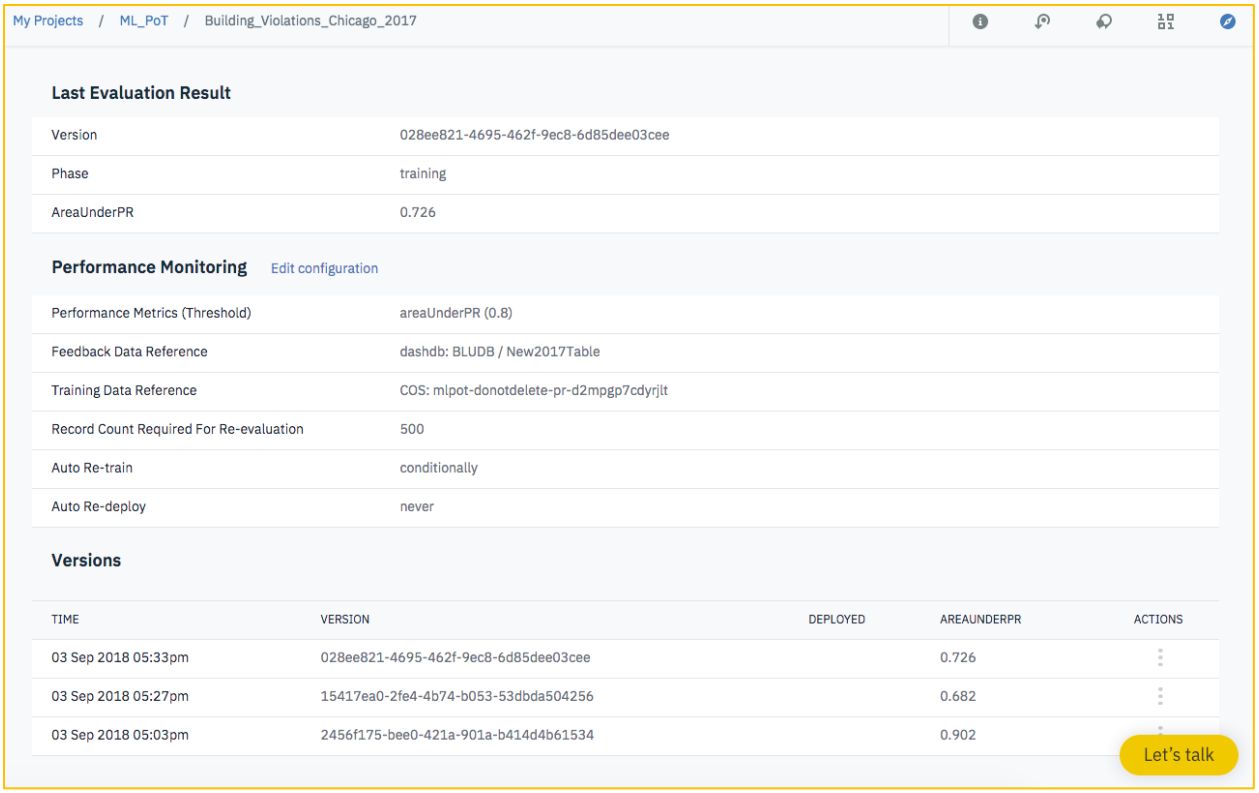


20. When the “New evaluation” window appears, click New Evaluation. Re-evaluating the model may take a few minutes. This is a great time to grab coffee.

Notice that the October data has caused our model to fall below the 0.8 threshold and a new version has been trained. Although we run through these steps graphically, you can also configure a Watson Machine Learning continuous learning model entirely through APIs.



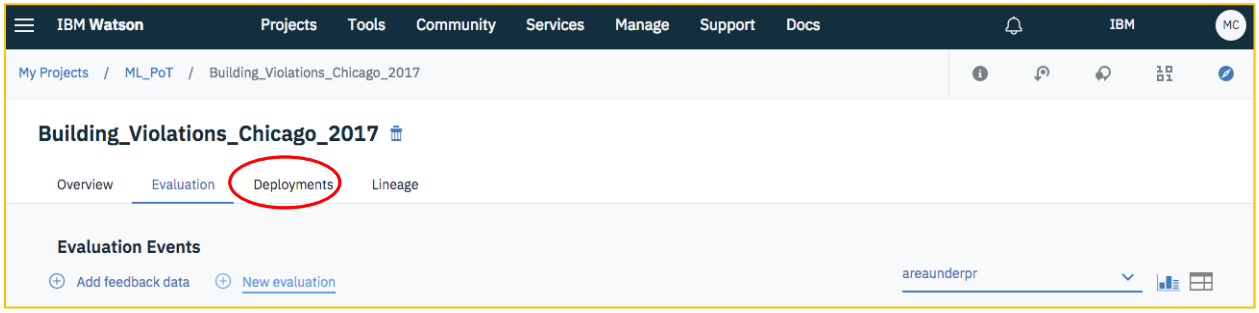
You can continue by adding the November data by the same process of hitting **Add feedback data** and adding the November file. If you scroll down you will be able to see all model versions listed. We have three versions, but no deployments.



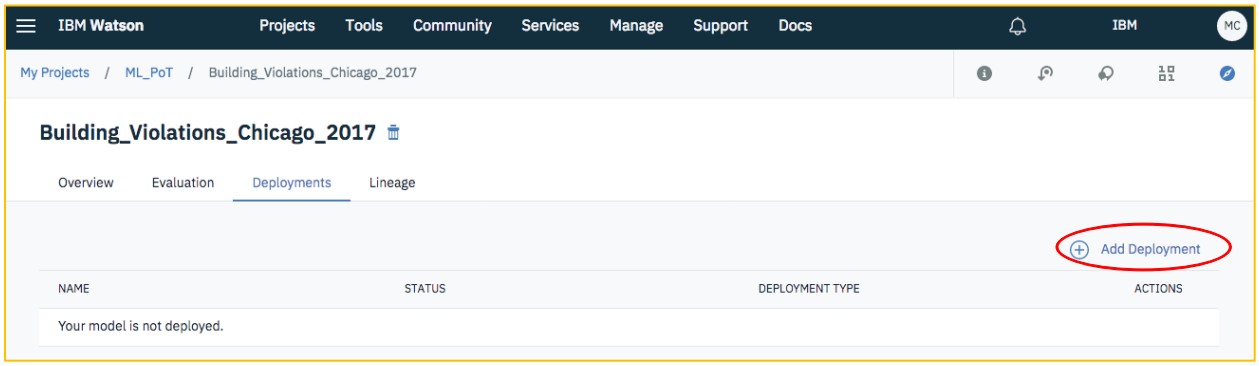
Step 4 Deploy a Machine Learning Model.

Now we will focus on deployment.

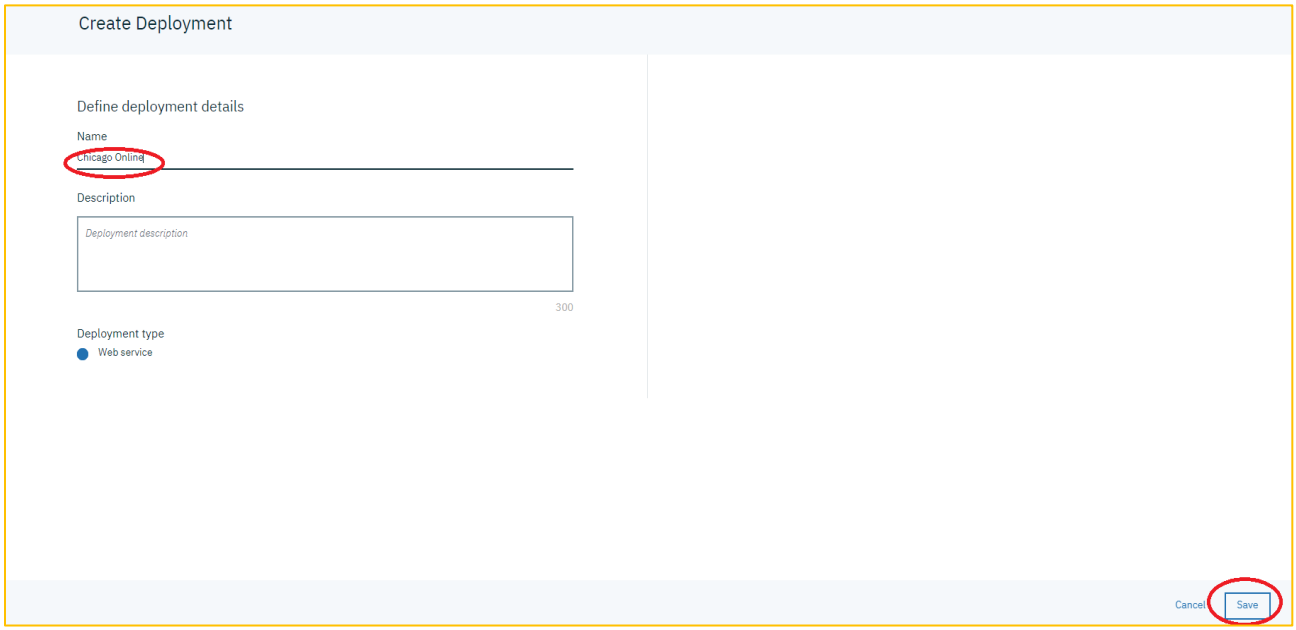
1. Click the Deployment tab.



2. Add Deployment.

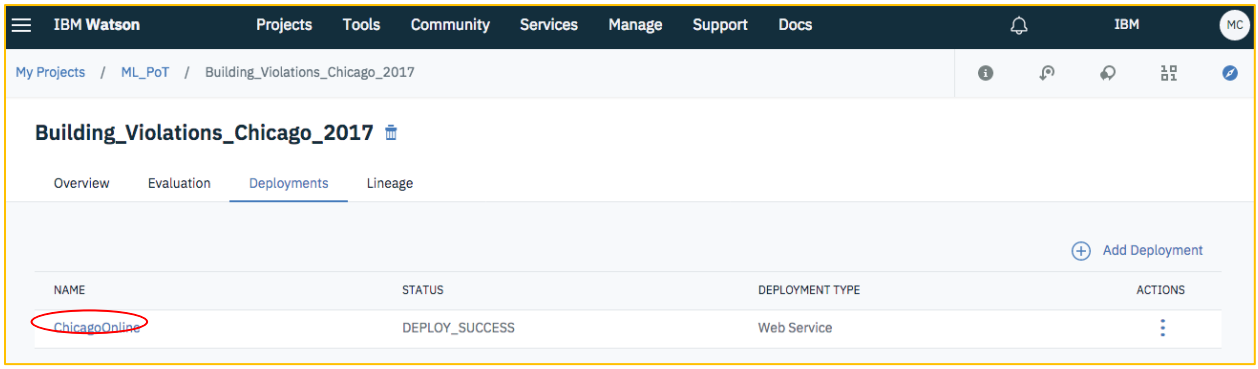


3. Enter a Name (e.g. ChicagoOnline).
4. Enter a Description.
5. Click Save.



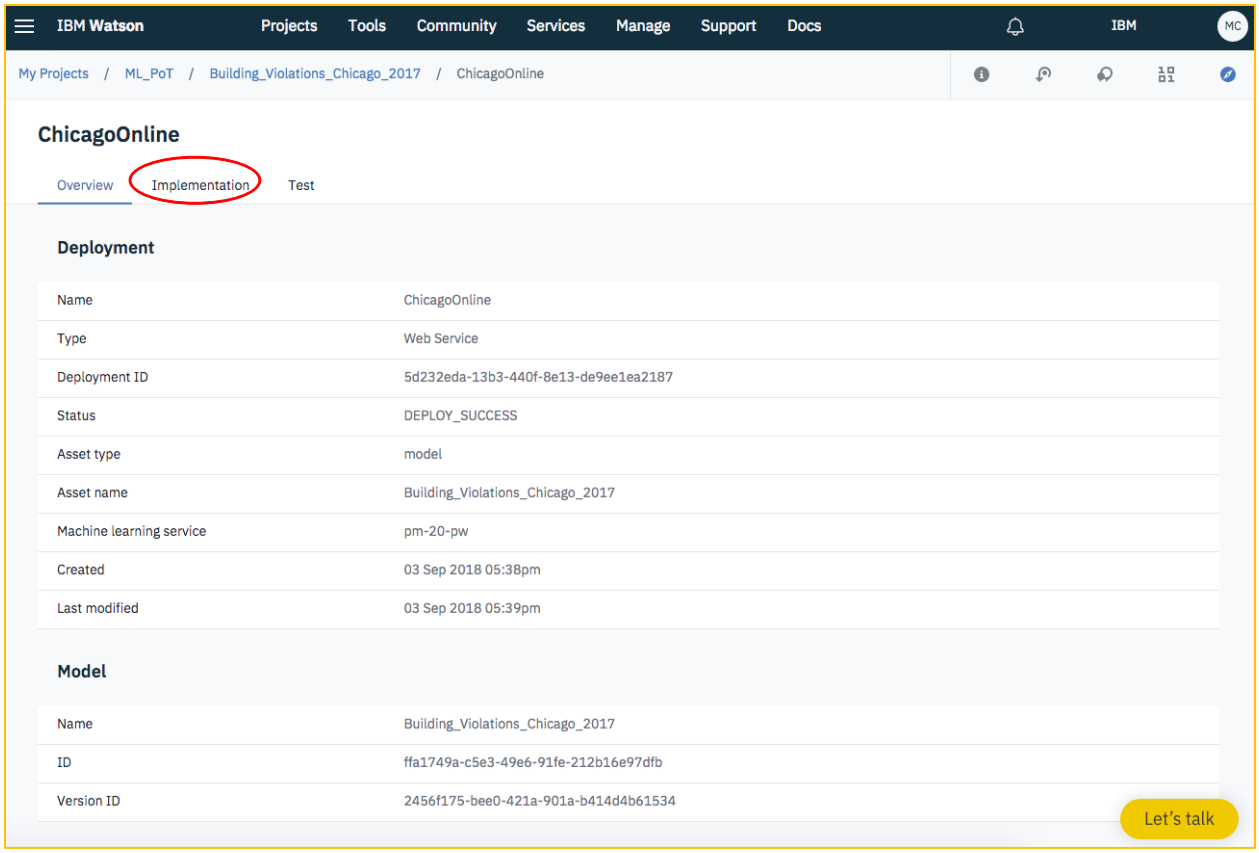
The model is now accessible by external systems.

6. Click on your deployment (e.g. ChicagoOnline).



You are automatically brought to the Overview tab lists information related to the model deployment including services used and version number.

7. Click the Implementation tab.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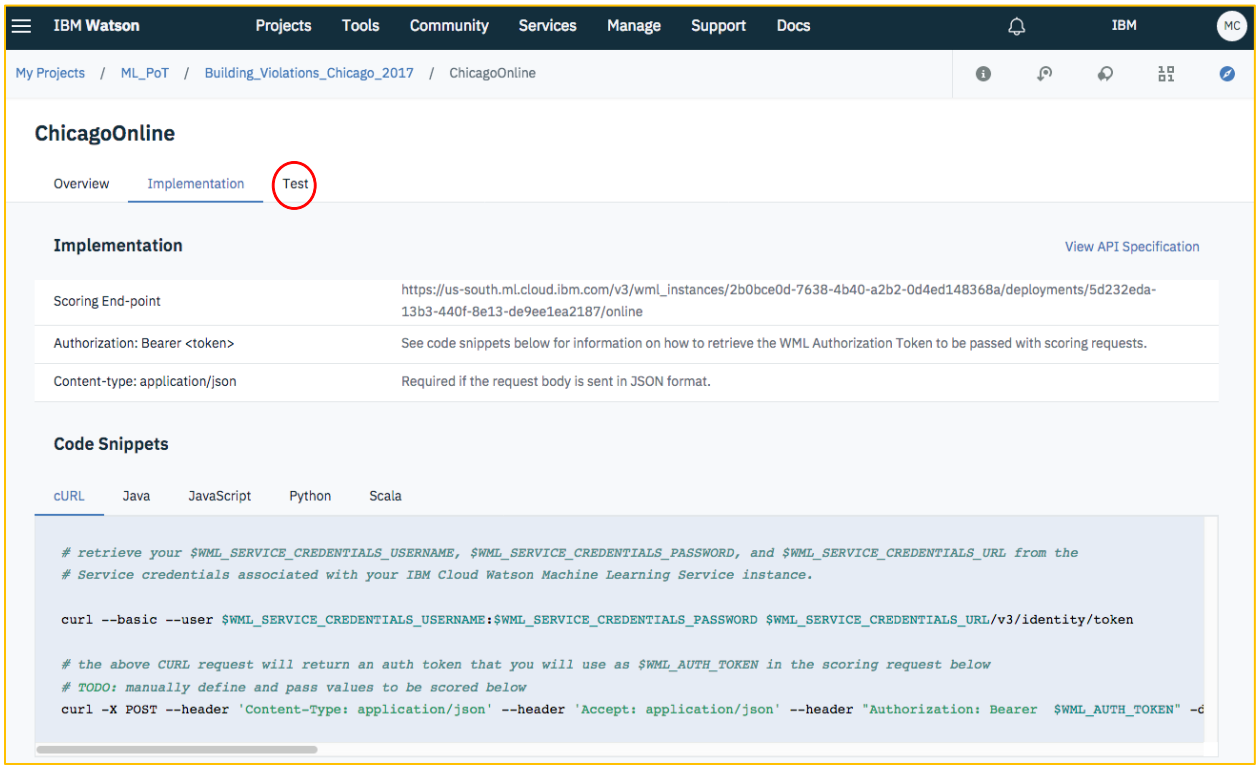
IBM
MC
My Projects / ML_PoT / Building_Violations_Chicago_2017 / ChicagoOnline
ChicagoOnline
Overview
Implementation
Test
Deployment
Name ChicagoOnline
Type Web Service
Deployment ID 5d232eda-13b3-440f-8e13-de9ee1ea2187
Status DEPLOY_SUCCESS
Asset type model
Asset name Building_Violations_Chicago_2017
Machine learning service pm-20-pw
Created 03 Sep 2018 05:38pm
Last modified 03 Sep 2018 05:39pm
Model
Name Building_Violations_Chicago_2017
ID ffa1749a-c5e3-49e6-91fe-212b16e97dfb
Version ID 2456f175-bee0-421a-901a-b414d4b61534
Let's talk

The implementation tab provides developers information to help minimize the time it takes to develop models and place them in a production environment.

8. Click on the Test tab.



The Test tab allows manual testing of the deployed model and viewing of results.

9. Enter:

   VIOLATION_CODE: CN063014

   VIOLATION _STATUS: OPEN

   INSPECTION_CATEGORY: COMPLAINT

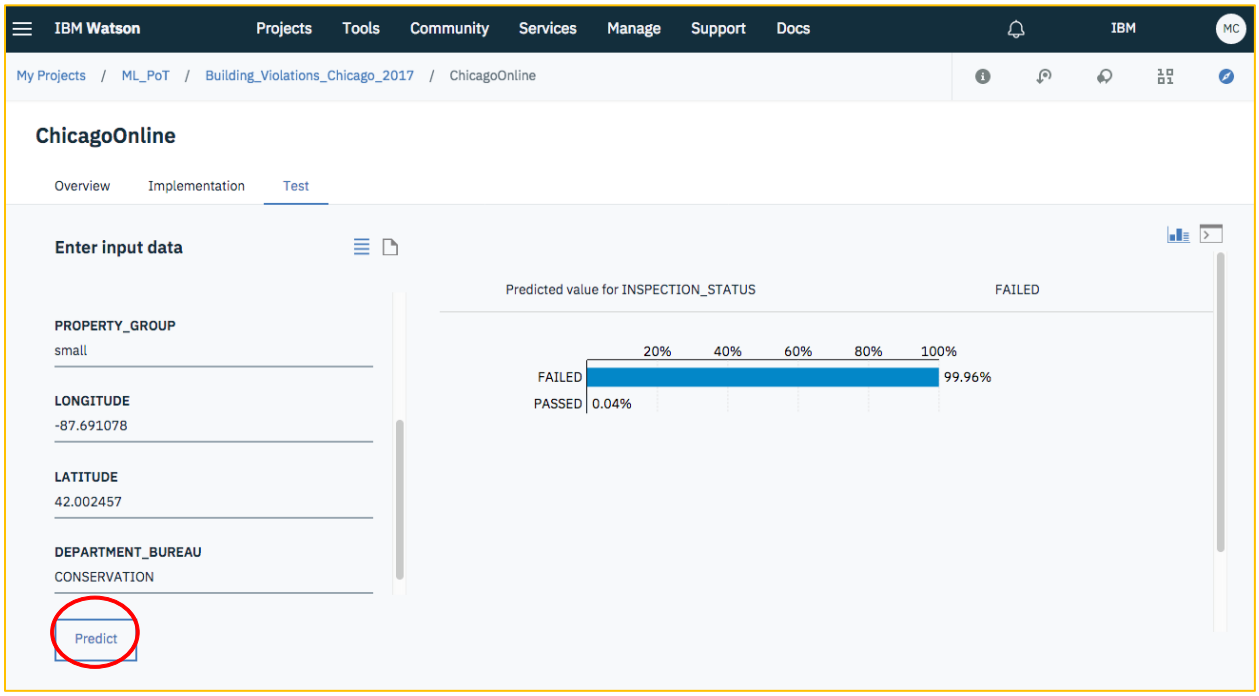
   PROPERTY_GROUP: small

   LONGITUDE: -87.691078

   LATITUDE: 42.002457

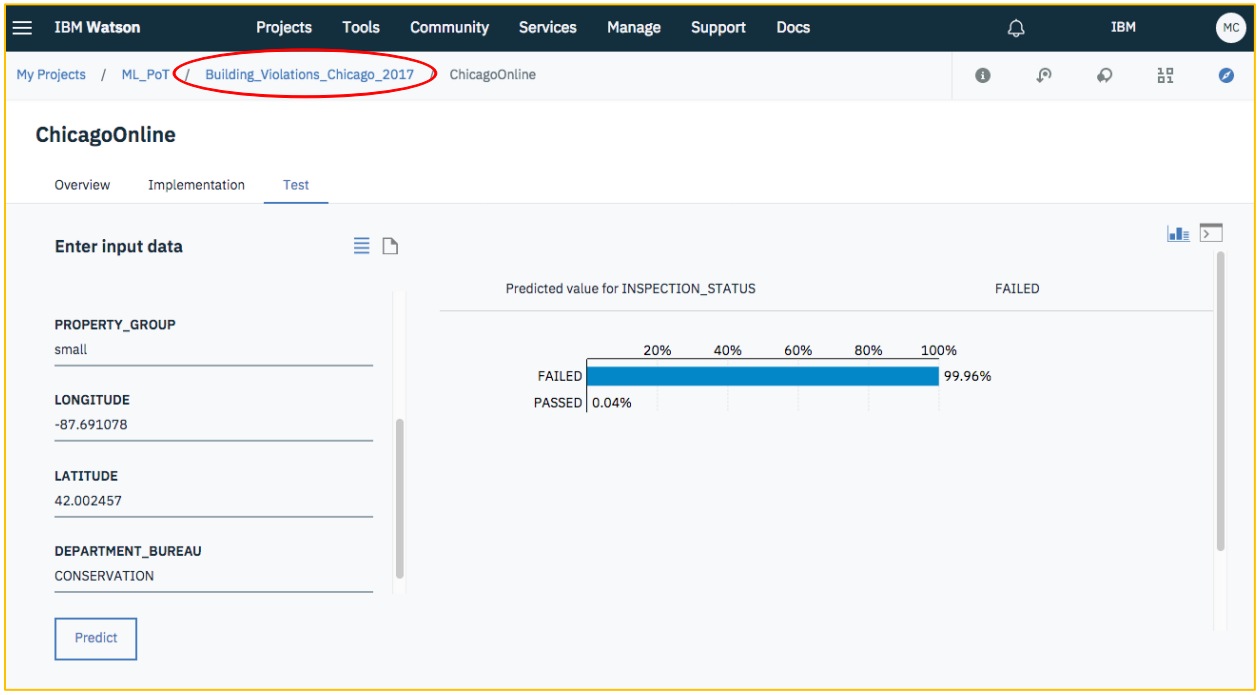
   DEPARTMENT_BUREAU: CONSERVATION

10. Click Predict.

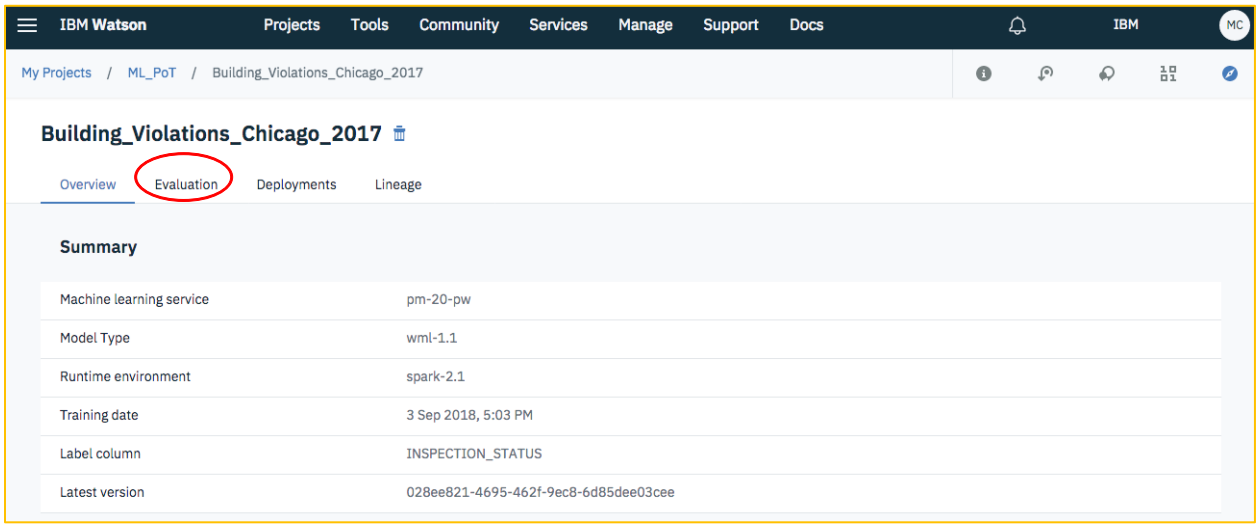


According to our model, a building with our input data has a 0.04% chance of passing inspection (your numbers may differ slightly)

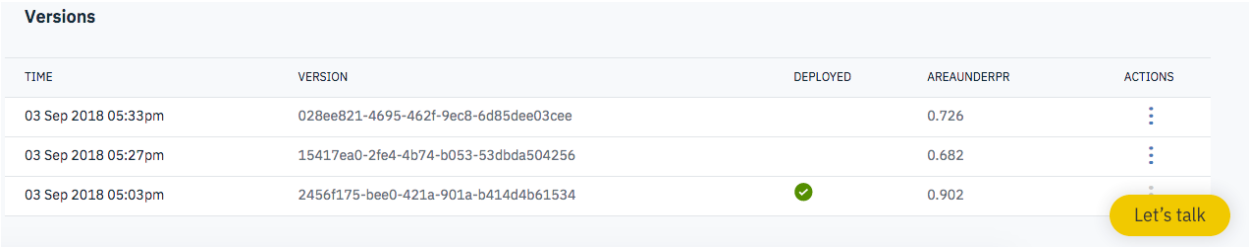
11. Feel free to change the input data and run new predictions. When ready, click on you model name (e.g. Building_Violations_Chicago_2017)



12. Click on the Evaluation tab.



If you scroll to the bottom you will be able to see the different model versions that have been created and which model is currently deployed. You can select the vertical ellipse under ACTIONS to change the deployed version.



During this lab we have very quickly compared various machine learning models and chosen the best one tuned to our dataset and objectives. We then created a continuous machine learning model that automatically monitors and retrains allowing Watson Machine Learning to keep applications, data pipelines, or external systems relying on the machine learning model as up to date as possible.

You have completed Lab 02

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