

**Lab Center – Hands-on Lab**

**Session 4874A**

#### **Visualizing Deployed Models with R Shiny and Watson Machine Learning**

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Table of Contents

[Disclaimer 3](#_Toc509159642)

[Sign up for IBM Watson Studio using your IBM Cloud Account 5](#_Toc509159643)

[Working with Watson Studio 6](#_Toc509159644)

[Create a project 6](#_Toc509159645)

[Add data assets to project 6](#_Toc509159646)

[Building a flow in Modeler 7](#_Toc509159647)

[Adding data using the flow 7](#_Toc509159648)

[Preparing data for modeling 7](#_Toc509159649)

[Training the model 8](#_Toc509159650)

[Saving the model to the project 10](#_Toc509159651)

[Creating a deployment for the model 11](#_Toc509159652)

[Building your Shiny Dashboard 11](#_Toc509159653)

[We Value Your Feedback! 13](#_Toc509159654)

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# Sign up for IBM Watson Studio using your IBM Cloud Account

If you have not yet tried IBM Watson Studio using your IBM Cloud Account then follow the steps below. Otherwise jump to the next section.

1. Go to datascience.ibm.com
2. Click Sign Up

If you already have an IBM Cloud Account then click on ‘Use it to sign up for IBM Watson’. Then click on the ‘IBM Watson Studio’ logo in the top left navigation bar, which will take you to the ‘Welcome’ page.

If you do not already have an IBM Cloud Account then do so through the ‘Register for IBM Cloud’ option.

1. Register using your email address.
2. Create your free IBM Cloud account and verify your email.
3. Once you log in to your IBM Cloud account after verification you will be taken to Watson Studio.

# Working with Watson Studio

The purpose of this lab session is to learn about the machine learning lifecycle and and gain skills in deploying and visualizing predictive models. We will start the lab by adding data from the Watson Studio community into a new project then quickly building a model using SPSS Modeler flows. Once we train a model, we can save it in the Watson Machine Learning (WML) repository where it can easily be deployed. Once we have the deployment we will create a dashboard using RStudio and Shiny.

### Create a project

All analytical assets within Watson Studio are organized into ‘Projects’. Projects are containers which can include data assets, Notebooks, Flows and models among other items. Projects can be personal or shared with a team of collaborators. It is a best practice to create a new project when you start a new analytics task.

1. From the ‘Welcome’ page of Watson Studio click on ‘New project’.
2. Enter a project name, and select your target IBM Cloud Object Storage Instance.
3. If you do not have a Cloud Object Storage service defined then click on ‘Add’ and select the ‘Lite’ plan, click ‘Create’ and then ‘Confirm’, and click ‘Refresh’ to detect this as the defined compute engine, then click ‘Create’.

### Add data assets to project

Data assets are added to a project to make them available for any of the tools included in Watson Studio. Data assets added to a project are also available for any of the collaborators sharing that project to use. Examples of data assets are files, such as .csv or .json or database tables added through a data connection.

1. Click on the ‘Community’ tab in the top bar. From here you can see content in the Watson Studio Community, which includes data sets.
2. Navigate to the data set ‘UCI: Adult - Predict income’, then click on the add (+) icon on the community card for this data set, select your project and click ‘Add’. Once this completes it will display ‘Added’. This may take up to a minute.
3. Click on the ‘View Project’ link.
4. Within the project’s ‘Assets’ tab you will see ‘Data assets’ shown including the one added above. Click on the ‘UCI: Adult - Predict income’ name to preview the data.
5. Return to the project by clicking on the project name in the navigation breadcrumb.

# Building a flow in Modeler

1. From the navigation bar select ‘Add to project’ then ‘Modeler flow’.
2. Provide a name and use the default runtime ‘IBM SPSS Modeler’, and click ‘Create’.

### Adding data using the flow

1. On the right side of the screen there is the list of the data assets which we added previously. Drag and drop the ‘UCI: Adult - Predict income’ data asset onto the canvas in the middle of the screen.

This dataset contains demographic information for the for adults including whether their salary is above or below $50k annually. The original data source can be found here: <https://archive.ics.uci.edu/ml/datasets/Adult>.

1. Right click on the ‘UCI: Adult - Predict income.csv’ node, or click on the three dots within the node, and select Preview to examine this data set.

As you can see this dataset contains a number of columns; including information about adults in the study like their age, education, marital status, etc. You can click the **Profile** and **Visualizations** tabs to explore the data more. Since this lab is more about understanding the full machine learning lifecycle we won’t focus much on exploratory data analysis, but rather move forward to model building. Click the name of your flow in the breadcrumb to return to the flow editor.

### Preparing data for modeling

1. Add a Filter node from the node palette on the left hand side. The Filter node is a ‘Field Operations’ node (or you can use the search bar).
2. Connect this node to the data set node then double click to open the properties on the right side.
3. Clicking the add columns button allows you to choose which fields you want to include for your model. Use the filtering available to select the following fields:

* Age
* Education-Num
* Marital\_status
* Occupation
* Hours\_per\_week
* Prediction

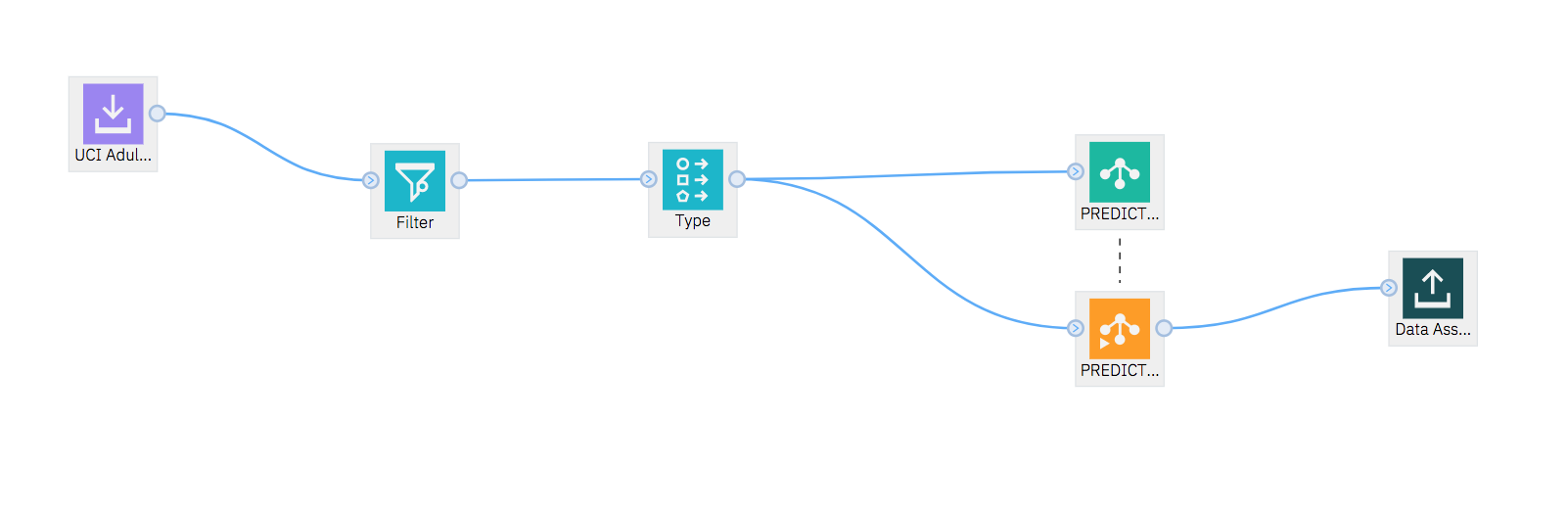
1. Click **OK** then change the Mode to “Retain the selected fields…” then click **Save**
2. Now add a Type node from the node palette on the left hand side. The Type node is a ‘Field Operations’ node (or you can use the search bar).
3. Double click the Type node and click on ‘Configure Types’ then click ‘Read Values’.
4. Change the *Role* for Prediction to be **Target**, then click **OK** then **Save**
5. This completes the minimal pre-processing we need to do before building a model. Since our target is discrete (either over 50k salary, or below) we need to choose an algorithm for classifying data.

### Training the model

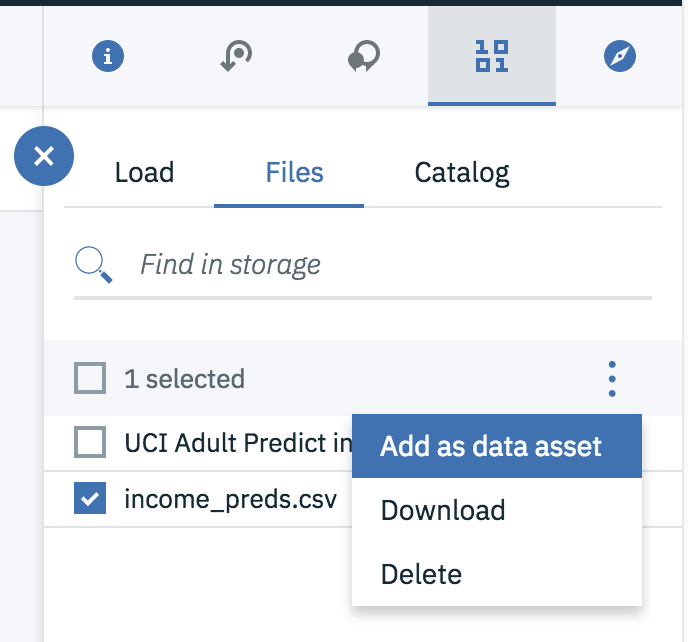
1. Go back to the palette and select a **Neural Network** model to add to the flow.
2. Right click this node and click **Run** to begin model training
3. To investigate this model, right the the model object (orange node) and click **View Model**

#### Now we have a trained model that can be deployed to Watson Machine Learning.

1. From the Export section of the Palette, add a Data Asset Export node to the flow and connect it to the Neural Network model. Double click this and add a name for the file, then run the flow to train the model and export the scored data to Cloud Object Storage

*Your flow should look something like this*

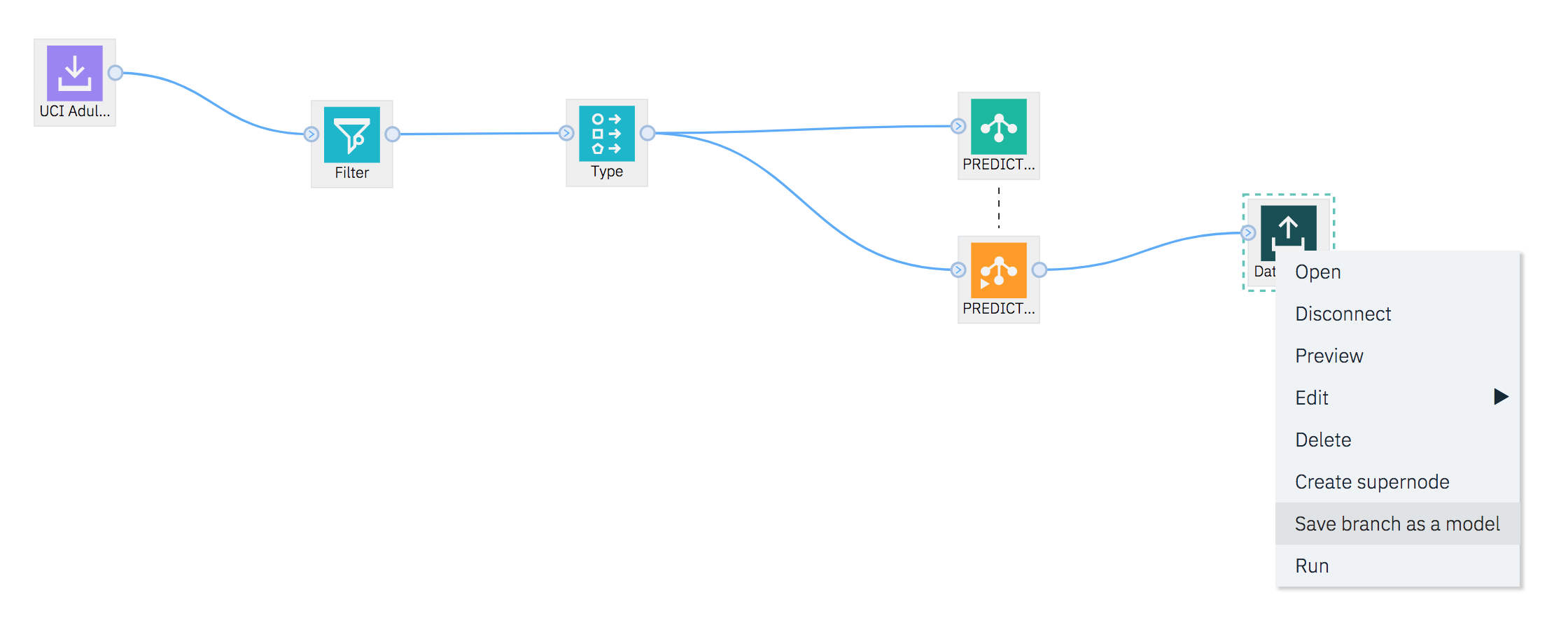
1. You can find this data asset in the right side panel of your project, where it can be added as a project asset if needed.



# Saving the model to the project

To use the model for scoring we need to deploy it in the Watson Studio project. The first step in this process is defining the branch in the flow that will be scored with data.

1. To deploy this flow, right-click on the final node (“Data Asset Export”) and click “Save branch as a model”.



At this point you will be required to create a new Watson Machine Learning service instance to save your model to.

1. In the ‘Required Service Missing’ dialog click on the ‘Create a new Watson Machine Learning service instance’ link.
2. Select the ‘Lite’ plan then click ‘Create’.
3. In the ‘Confirm Creation’ dialog click ‘Confirm’.

You will be returned to the flow and from this point you can use the context menu on the table node click ‘Save as a model’ to use this newly defined Watson Machine Learning service instance.

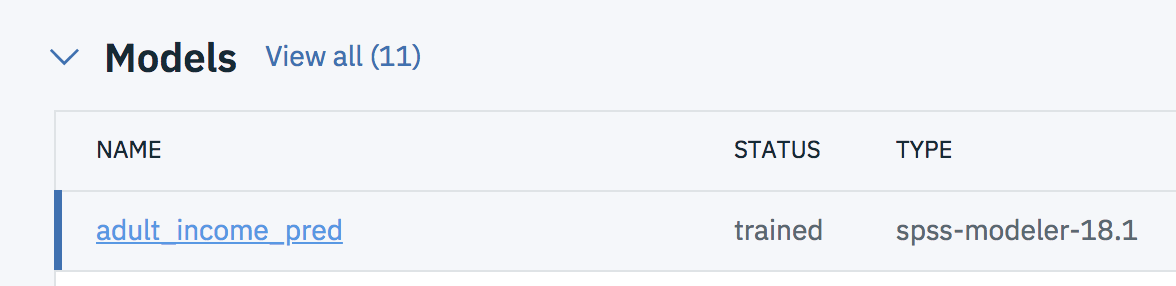
1. On the ‘Save Model’ page give the model a suitable name (such as ‘Income Model’) and make sure the Watson Machine Learning instance you created earlier is selected.
2. Then click ‘Save’.

The model has now been saved as an analytics asset in your Watson Studio project making it available for deployment for scoring.

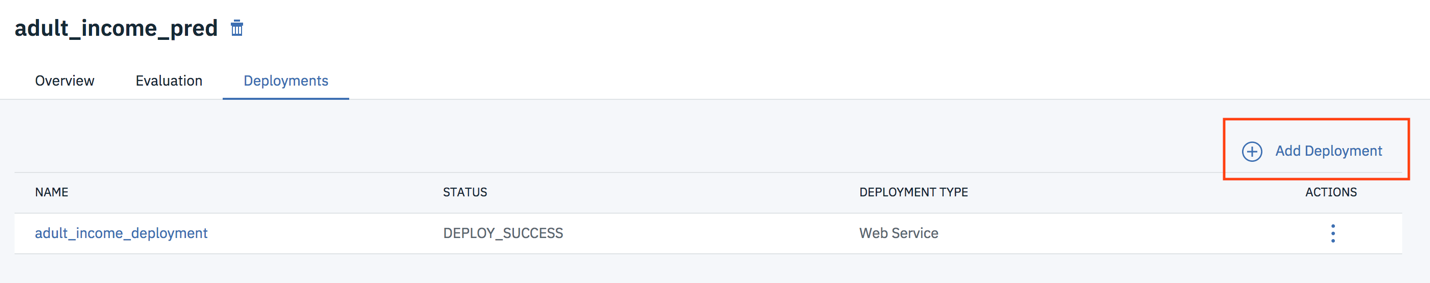
# Creating a deployment for the model

In order to use this model for our Shiny dashboard, we need to create a deployment so we can score new data with the model vis REST API.

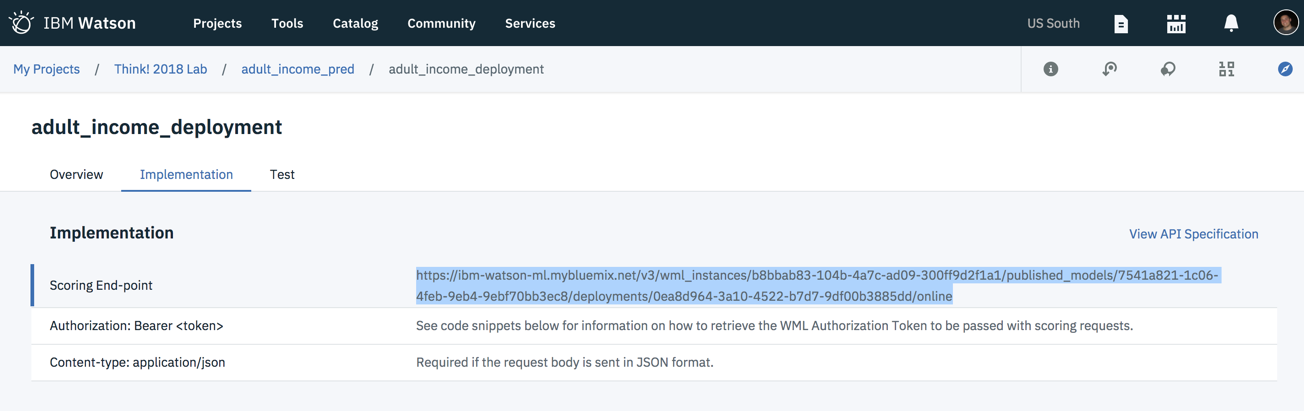
1. Go to the Models section of your project and click the name of the saved Model to see it’s details:



1. After drilling into the model page, go to the Deployments tab and add a deployment:



1. After creating the deployment, you can view more information clicking on the name of the deployment, or clicking on the **Deployments** tab of the project to view all deployments
2. The important information we need from the deployment page is the scoring end-point – this is the url we will use for the dashboard code:



# Building your Shiny Dashboard

To complete this lab, please navigate to <https://github.com/gfilla/PredictingIncomeDashboard>

This repository contains the following steps:

* Finding your Watson Machine Learning credentials
* Understanding how R Shiny and Flexdashboards work
* Understanding how to score deployed models via REST API
* Composing a dashboard that scores your deployed model

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