

Data Science Project Report

Criselyn Santiago MoTIS-9

Walmart: Store Sales Forecasting

The data which formed part of my project was the Walmart dataset obtained from Kaggle. The data contained weekly sales of various departments within different stores over period of time. Most of the work put into the project

I mentioned three steps in my proposal, same as the other projects, then, I can proceed also to those steps. Those steps are data cleaning, predictive modeling and visualization. Let's start.

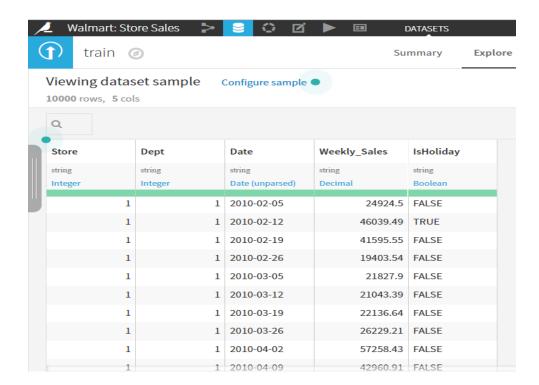
Data Cleaning

Cleaning process of our data and building some features will be done in data cleaning. In this part, I will do some transformation in my dataset using some recipes like prepare, group, join, enrich, sort, etc. in Dataiku DSS. In this step, I need to analyze my dataset in order to clean it properly and to apply some features which is useful for the dataset. In my case, I just used prepare and join recipe.

• Here are my datasets:

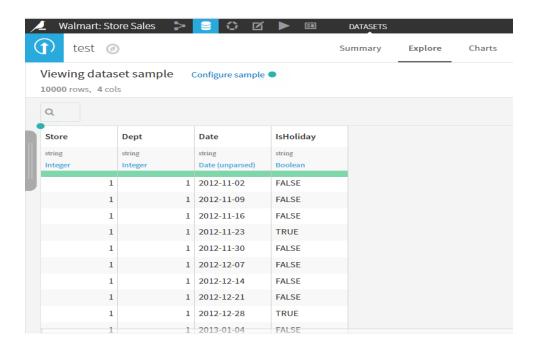
TRAIN DATASET

My train dataset is clean as what we noticed in the picture below. So, I don't need to clean it.



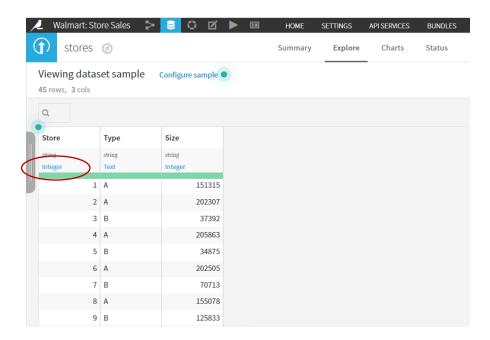
TEST DATASET

Same thing with the train dataset.

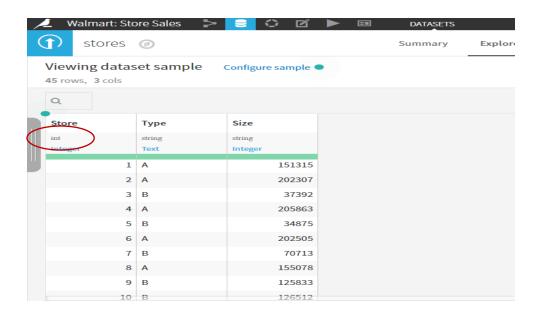


STORE DATASET

In store dataset, I just change the storage type of "Store" from **string** to **int**. The reason why I changed it because later on, I will join it to feature and train dataset and feature and test dataset.

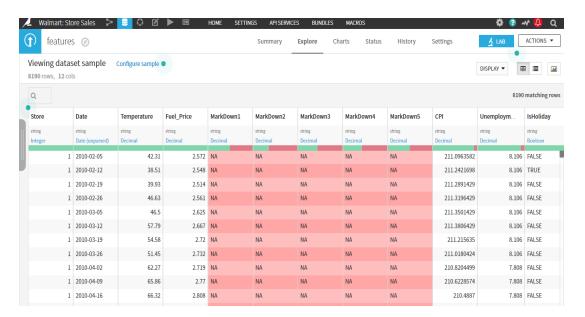






FEATURE DATASET (Using PREPARE RECIPE)

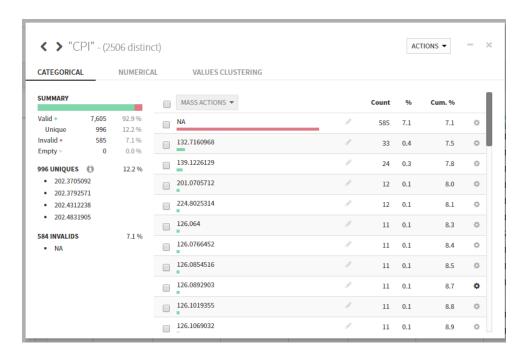
In this dataset, I did a lot of changes but not too much. As we all know, the meanings are automatically detected from the contents of the columns. In Markdown 1-5, CPI, and Unemployment, there should be done some transformation.

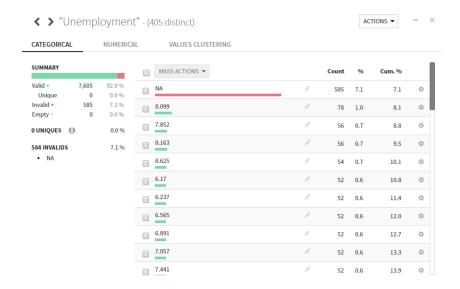


First, I need to change the NA values of Markdown 1-5. I replaced the NA values with another value called -999999. I put something different from zero because I saw that zero is a possible valid value. Look the result below:

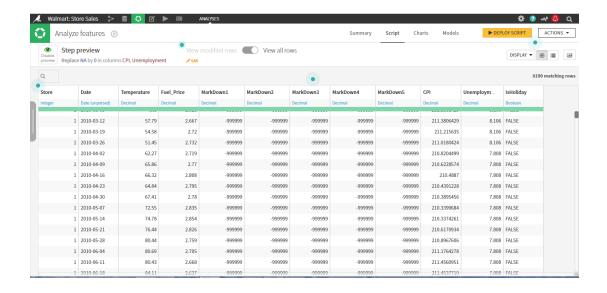
MarkDown1	MarkDown2	MarkDown3	MarkDown4	MarkDown5
Decimal	Decimal	Decimal	Decimal	Decimal
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
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-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999
-999999	-999999	-999999	-999999	-999999

Second, the CPI and Unemployment have also NA values (look the pictures below). And, I don't want to delete those rows because there is a possibility that it will affect the final dataset later on. So, I decided to put 0 as their values.



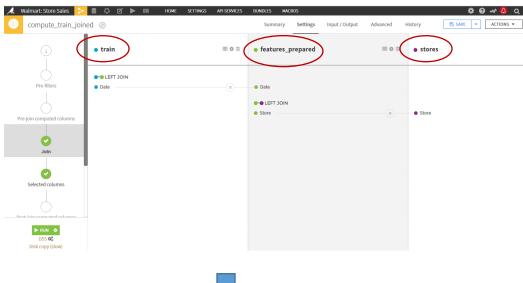


So, here is the cleaned feature dataset.



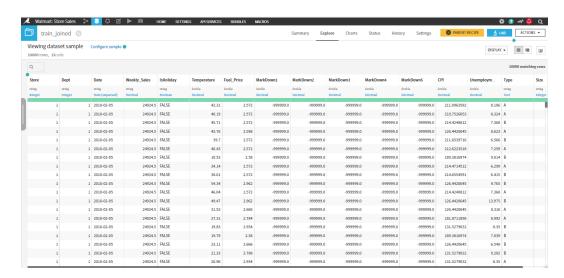
• Let's go to **JOIN RECIPE** which is dedicated to joins between two or more datasets. I merged the three data files (train, features and stores) to see the effect of different variables on sale. I merged also the test, features and stores dataset. Let's have a look.

TRAIN - FEATURES_PREPARED - STORES

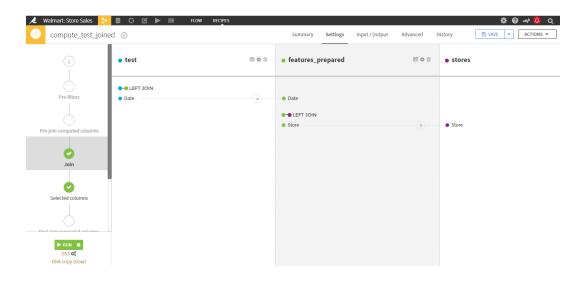




TRAIN_JOINED

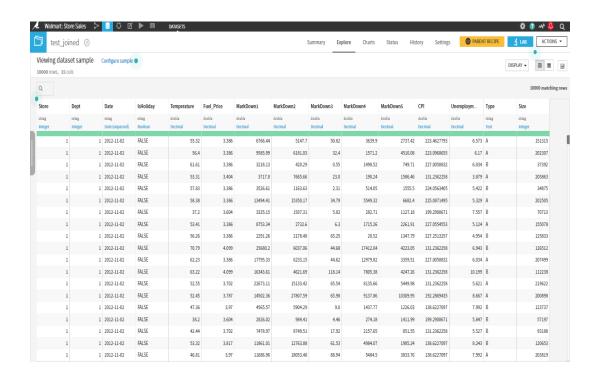


TEST - FEATURES_PREPARED - STORES



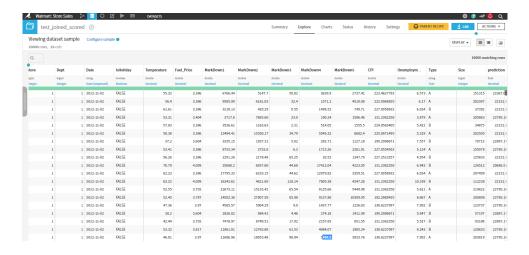


TEST_JOINED



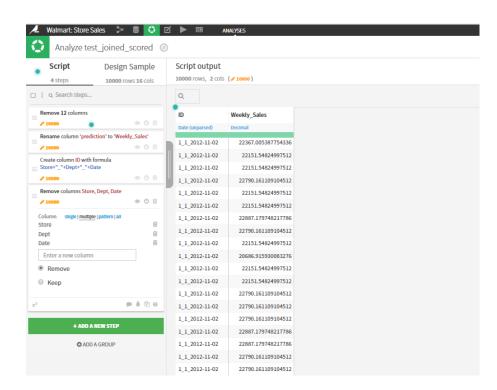
• After I scored the model that I chose, I had a new dataset which I need to make some transformation to the dataset.

Test_joined_scored



Final Column of the Final Dataset

- I removed 12 columns which are IsHoliday, temperature, fuel_price, markdown 1-5, CPI, Unemployment, type and size
- I renamed the prediction column to weekly_sales
- I created a formula to concatenate the three columns

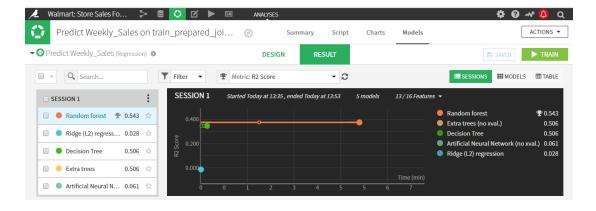


Predictive Modeling

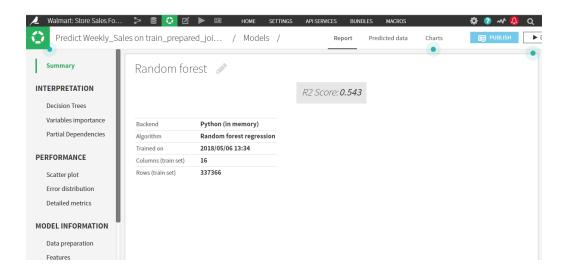
Building and deploying a predictive model. In this part, I can use a lot of model like random forest, decision tree, extra trees, artificial neutral network etc., then, at last I will choose the best model.

Now, I'm done to clean and do some engineering features in my dataset. So, it's time to do a prediction model. I used "weekly_sales" as my target when I did the prediction. I did a lot of models to train my dataset like Random Forest, Ridge (L2) regression, Decision Tree, Extra trees and artificial Neural Network. In my first flow that I had done, the Random Forest was in the top or recommended to use but then, when I did another flow for my project, the recommended one was the Extra trees. At last, I decided to use the Extra trees as my model. The reason why I did two different flow because I'm not satisfied from the result of my first flow. When I did, another session from my first flow, the result are always the same. Until, I decided to make another flow for my project.

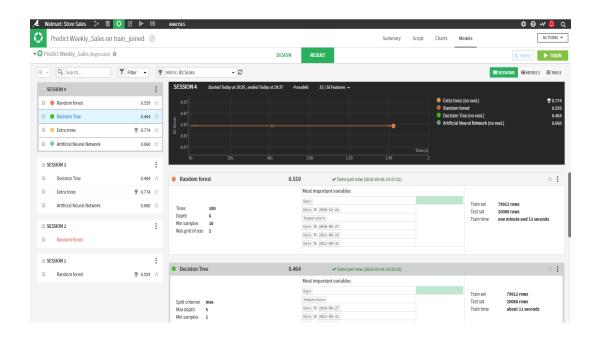
 Below was the result of the first FLOW of my project that I had done and I chose Random Forest as the model.

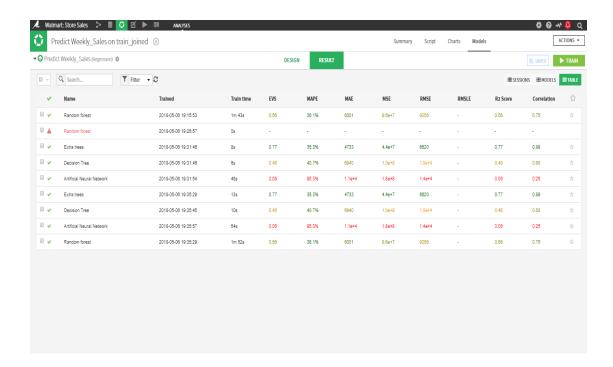


RESULT

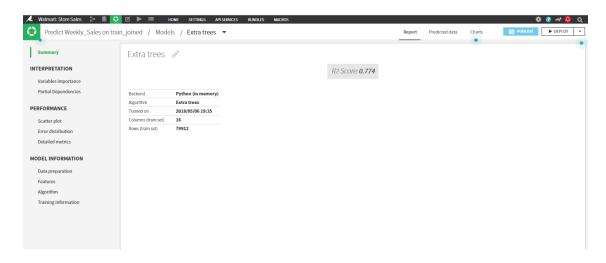


 Below was the result of the second FLOW of my project that I had done and I chose Extra trees as the model.



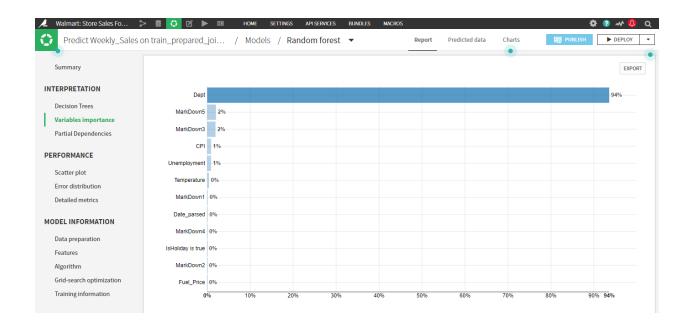


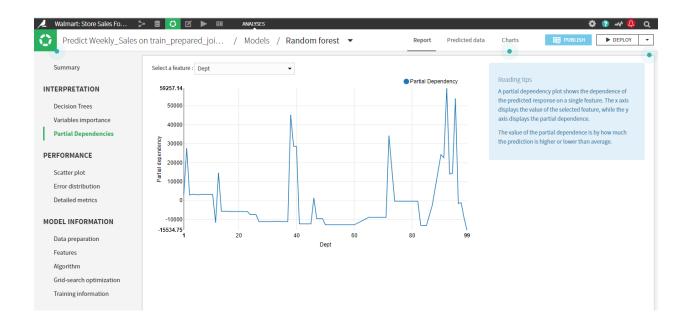
RESULT



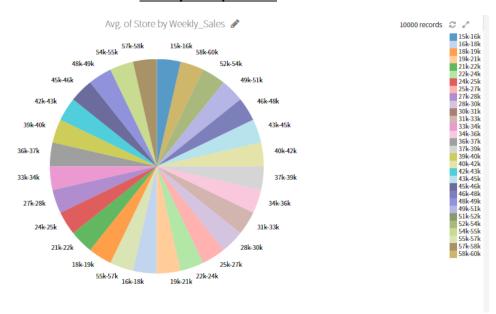
Visualization

Creating a useful visualization of our predicted data and the dataset.



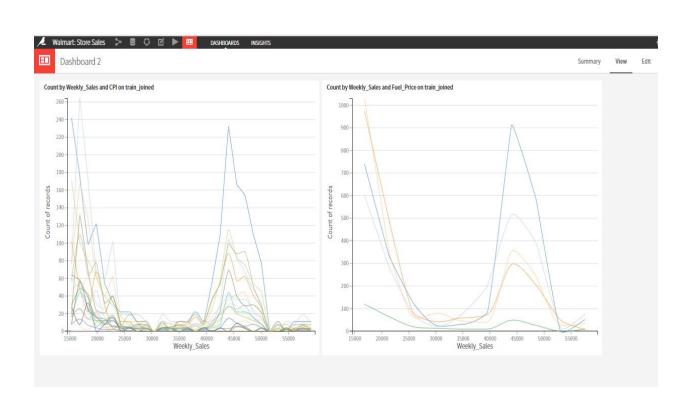


Weekly sales per store



DASHBOARDS

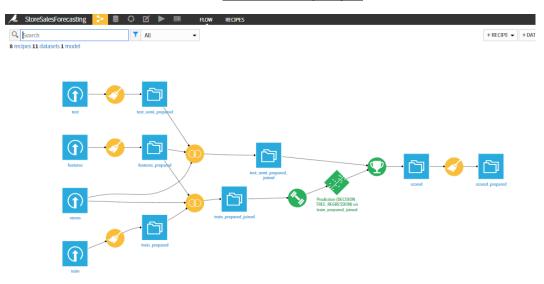




FLOW OF THE PROJECT

Actually, they are almost the same. At the end, I decided not to clean the test and train dataset (look at cleaning section the overview of train and test dataset).

First Flow of my Project



Second Flow of my Project

(This is my final flow)

