

RECOMMENDATION ENGINE

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Data Management and Business Process Modeling

Recommendation Engine using Oracle Cloud

Problem Statement: After developing our Entity Relationship Diagrams, the required tables in an Oracle database, and a few report queries using Oracle's SQL Developer, we wanted to create for our client more value from our Data Management Solution. We realized that building a recommendation engine would help the client's sales team improve its cross-sell probability by helping the team make data-informed recommendations to their customers.

I researched some possible solutions, and realized that I could implement a recommendation engine using Oracle Cloud. Here is a quick run through the steps involved.

Part I: Setup Oracle Cloud and create OML user

- 1. Create an Oracle Cloud Free Tier Account (https://www.oracle.com/cloud/free/)
 - Need a credit card and a phone number to register
- 2. Sign in to Oracle Cloud console using tenant id and credentials generated from step 1 (https://www.oracle.com/cloud/sign-in.html)

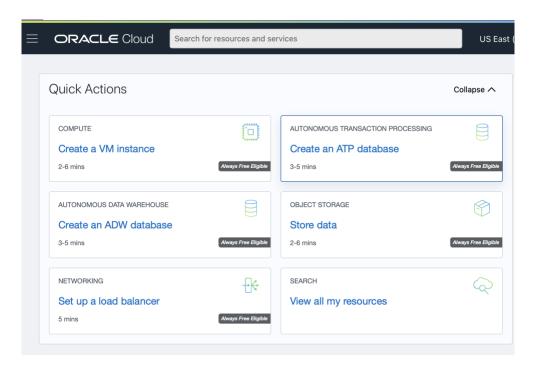


3. Click on "Create an ATP database" to create an Autonomous Transaction Processing Database (Always Free).

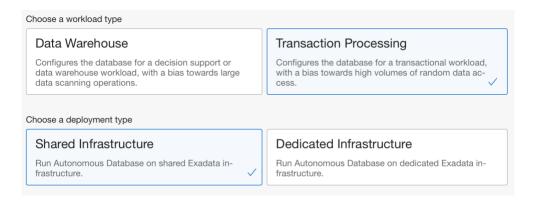
Provide Admin Password (field can be found on the ATP form). Review the default values in the other fields on the form. You may keep the other details the same as suggested by Oracle.

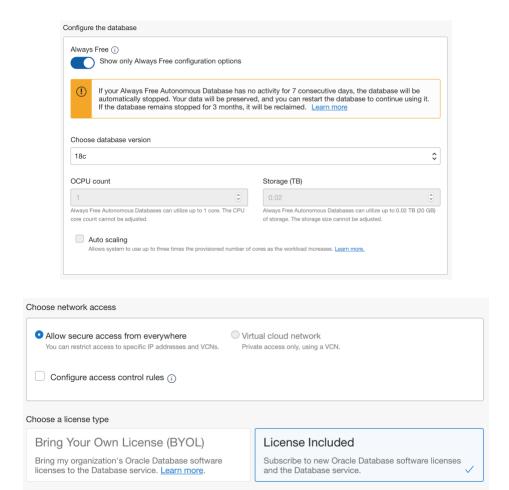
Click "Create Autonomous Database"

Please refer to the following screenshots of the form.

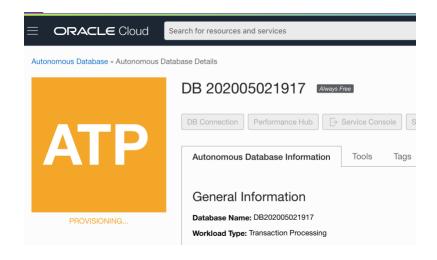


Screenshots of the form:

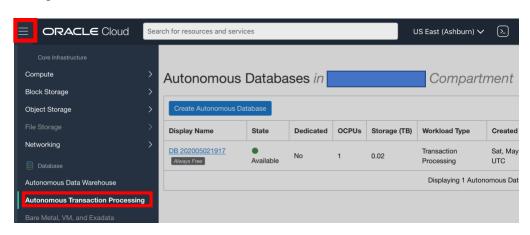


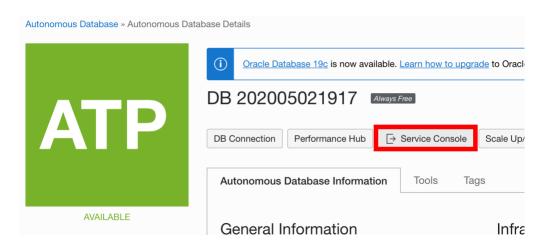


4. The ATP status would change from Provisioning to Available:

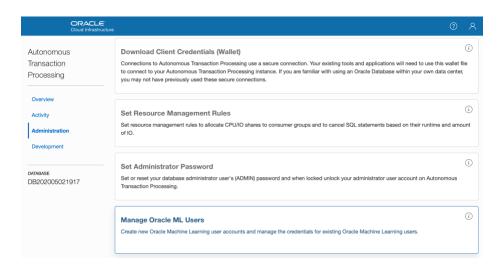


- 5. Create OML user: Now that we have our Autonomous Database available, the next step is to create a new Oracle Machine Learning (OML) user, since the ADMIN a/c is restricted from creating notebooks and jobs. The OML user can create notebooks and other jobs.
 - a. Navigate to the new ATP and click on Service Console

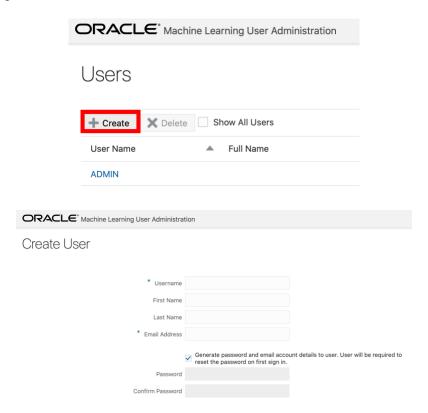




b. On the Service Console screen, click on Administration and then select Manage Oracle ML Users



c. Click on "+ Create" and create the required User by filling in details on the following screen.



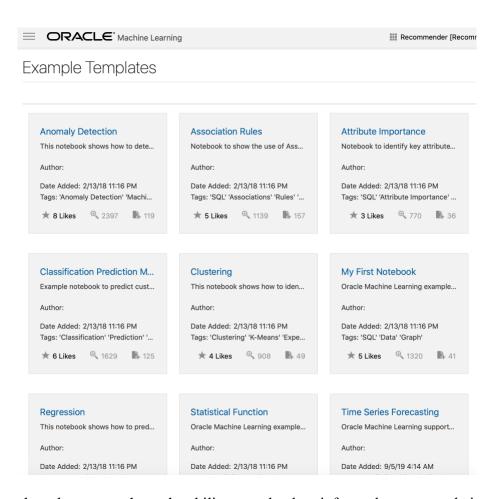
d. Once we have the OML user working, we are all set to import data and use services such as SQL Developer Web and OML Notebooks that help implement machine learning algorithms through analytical PL/SQL functions.

Part II: Import data and implement ML algorithms/Recommendation Engine

- 1. Login using the URL and the OML user credentials provided by the OML ADMIN. The user will get an email containing the login URL and the password reset link.
- 2. Create Project and Workspaces
 - a. The OML service would create a default workspace and project, but you may create a new workspace and project.
 For more details, refer to:
 https://www.oracle.com/webfolder/technetwork/tutorials/obe/cloud/oml/OMLPW-create-project-workspace/html/index.html
- 3. Build tables and populate data using SQL Developer Web. You can also use OML notebooks for this step.
- 4. OML offers notebooks which we will use to build our recommendations. Notebooks are a fast and easy way to share analytical approaches and results of our machine learning pilot with our Data Management and Business Process Modeling audience
 - a. Create a notebook
 - b. Open the notebook in notebook editor and build your model
 For more, refer to:
 https://www.oracle.com/webfolder/technetwork/tutorials/obe/cloud/oml/OMLCR-create-run-notebooks/html/index.html



5. Oracle Machine Learning notebooks service provides example templates, under the "Examples" section, that may be used to implement some popular algorithms and machine learning techniques.



6. We want the sales team to have the ability to make data-informed recommendations to clients and increase cross-sell probability. To demo a pilot, we decided to use the Association Rules machine learning technique and build association rules models using the Apriori algorithm for the recommendations.

The algorithm requires data containing Case ID/Value pairs. A Case ID column such as

The algorithm requires data containing Case ID/Value pairs. A Case ID column such as customer id and a value column such as Product ID or Product name can be used. For more details on the algorithm, refer to:

https://docs.oracle.com/en/database/oracle/oracle-database/12.2/dmcon/apriori.html

7. Join our TRANSACTIONS and PRODUCT table data and load into the SALES_TRANS_CUST table. The model uses this table as the single source of data from where it can build the Association Rules (AR) model.

```
CREATE OR REPLACE VIEW SALES_TRANS_CUST AS

SELECT DISTINCT CUSTOMER_ID, PRODUCT_NAME, PRODUCT_LINE

FROM (SELECT A.CUSTOMER_ID, B.PRODUCT_NAME,

B.PRODUCT_LINE

FROM transactions A, product B

WHERE A.PRODUCT_ID = B.PRODUCT_ID

);
```

8. Drop the current AR_SH_SAMPLE_SETTINGS table which contains the AR Model settings

```
-- Drop table AR_SH_SAMPLE_SETTINGS

BEGIN

EXECUTE IMMEDIATE 'DROP Table AR_SH_SAMPLE_SETTINGS';

EXCEPTION

WHEN OTHERS THEN NULL;

END;
```

9. Build a new AR_SH_SAMPLE_SETTINGS table which will contain our AR model settings

10. Insert Association Rules model settings data such as minimum support, minimum confidence, etc., into AR_SH_SAMPLE_SETTINGS. These parameters can be changed as required to limit build time or Association Rules

```
-- Insert data into AR_SH_SAMPLE_SETTINGS

BEGIN

INSERT INTO AR_SH_SAMPLE_SETTINGS VALUES

(DBMS_DATA_MINING.ASSO_MIN_SUPPORT, 0.04);

INSERT INTO AR_SH_SAMPLE_SETTINGS VALUES

(DBMS_DATA_MINING.ASSO_MIN_CONFIDENCE, 0.1);

INSERT INTO AR_SH_SAMPLE_SETTINGS VALUES

(DBMS_DATA_MINING.ASSO_MAX_RULE_LENGTH, 2);

INSERT INTO AR_SH_SAMPLE_SETTINGS VALUES

(DBMS_DATA_MINING.ODMS_ITEM_ID_COLUMN_NAME,

'PRODUCT_NAME');

COMMIT;

END;
```

11. Now we can create our AR model using the Association mining function supported by Oracle Data Mining. We need to provide the model name, mining function, data table name containing the sales transactions, case id column name, and table name containing model settings for the AR model.

```
-- Create model for Market Basket Analysis using Sales
history transactional data

BEGIN

DBMS_DATA_MINING.CREATE_MODEL(
MODEL_NAME => 'AR_SH_SAMPLE',
MINING_FUNCTION => DBMS_DATA_MINING.ASSOCIATION,
DATA_TABLE_NAME => 'SALES_TRANS_CUST',
CASE_ID_COLUMN_NAME => 'CUSTOMER_ID',
SETTINGS_TABLE_NAME => 'AR_SH_SAMPLE_SETTINGS'
);

END;
```

12. Now that we have our model, we can use it to see what items to recommend when customer is interested in a particular product, say "Math Mayhem - Fun with Fractions" from our sales transaction data. In terms of the Apriori, the antecedent is "Fun with Fractions" and we want the consequent and the associated Lift. The AR Example provided by OML has a template which we will update with our antecedent.

```
-- What items should we recommend when customer is
interested in "Math Mayhem - Fun with Fractions"?
-- Assume the antecedent item is "Math Mayhem - Fun with
Fractions".
-- Since the number of items in antecedent is 1, the
number of items is 2.
-- choose top 5, ordered by rule lift to see top
recommendations by lift
-- Change sort to RULE SUPPORT to see top recommendations
by support
SELECT ROWNUM RANK,
     CONSEQUENT NAME RECOMMENDATION,
     NUMBER OF ITEMS NUM,
     ROUND (RULE SUPPORT, 3) SUPPORT,
     ROUND (RULE CONFIDENCE, 3) CONFIDENCE,
     ROUND (RULE LIFT, 3) LIFT,
     ROUND (RULE REVCONFIDENCE, 3) REVERSE CONFIDENCE
FROM (SELECT * FROM DM$VRAR SH SAMPLE
     WHERE NUMBER OF ITEMS = 2
```

```
AND EXTRACT(antecedent, '//item[item_name="Math Mayhem - Fun with Fractions"]') IS NOT NULL ORDER BY RULE_LIFT DESC, NUMBER_OF_ITEMS)
WHERE ROWNUM <= 5;
```

Output:

```
RANK RECOMMENDATION NUM SUPPORT CONFIDENCE
LIFT REVERSE_CONFIDENCE
1 Chemistry Creations - Dinosaur Wonders 2 0.125 1
8 1
```

The Transactions Data:

A few things to note:

- a. An important constraint to note is that Oracle Data Mining supports association rules that have one or more items in the antecedent and a single item in the consequent.
- b. We could add more values to the antecedent to do a more complex market basket analysis, and we could mine more association rules. However, as the number of unique items increases to x, total number of itemsets increases to 2^x, and the algorithm may require significantly more build time. In such a scenario, we could reduce the itemsets considered by increasing the minimum support and minimum confidence and decreasing the max rule length parameters in the model settings table as well as by decreasing the number of items in our recommendation query to 2.
- c. Although, Oracle Cloud's ATP services are always free, some OML services are currently limited to first 30 days.
- d. For more details on the Apriori algorithm and the associated terms, please refer to: https://docs.oracle.com/en/database/oracle/oracle-database/12.2/dmcon/apriori.html

References:

- 1. https://www.oracle.com/webfolder/technetwork/tutorials/obe/cloud/oml/OMLPW-create-project-workspace/html/index.html
- 2. https://www.oracle.com/webfolder/technetwork/tutorials/obe/cloud/oml/OMLCR-create-run-notebooks/html/index.html
- 3. https://docs.oracle.com/en/database/oracle/oracle-database/12.2/dmcon/apriori.html
- 4. https://docs.oracle.com/en/cloud/paas/autonomous-data-warehouse-cloud/omlug/using-oracle-machine-learning.pdf
- https://docs.oracle.com/en/database/oracle/oracledatabase/12.2/arpls/DBMS_DATA_MINING.html