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YOUR EVALUATIONS**

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Convention Center



# Modern Machine Learning

## With Oracle Analytics Cloud and Oracle Autonomous Data Warehouse Cloud

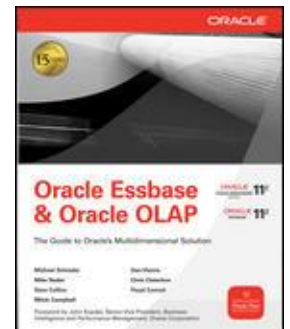
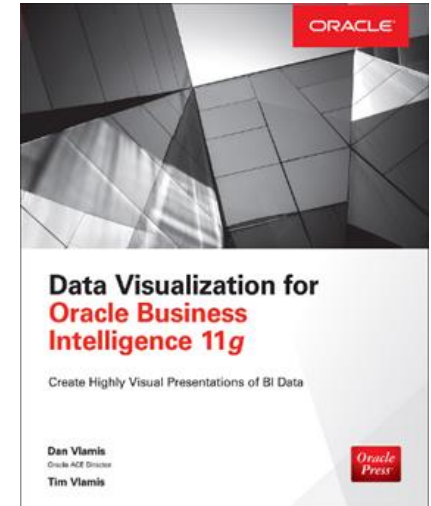
Tim Vlamis

June 26, 2019

Vlamis Software Solutions


# VlamiS Software Solutions

- VlamiS Software founded in 1992 in Kansas City, Missouri
- Developed 200+ Oracle BI and analytics systems
- Specializes in Oracle-based:
  - Enterprise Business Intelligence & Analytics
  - Analytic Warehousing
  - Data Mining and Predictive Analytics
  - Data Visualization
- Multiple Oracle ACEs, consultants average 15+ years
- [www.vlamiS.com](http://www.vlamiS.com) (blog, papers, newsletters, services)
- Co-authors of book “Data Visualization for OBI”
- Co-author of book “Oracle Essbase & Oracle OLAP”
- Oracle University Reseller
- Oracle Gold Partner





## Vice President & Analytics Strategist

- 30+ years in business modeling and valuation, forecasting, and scenario analyses
- Oracle ACE  ORACLE ACE
- Instructor for Oracle University's Predictive Analytics, Data Mining Techniques and Oracle R Enterprise Essentials Courses
- Professional Certified Marketer (PCM) from AMA
- MBA Kellogg School of Management (Northwestern University)
- BA Economics Yale University



# What is Machine Learning?

- The application of advanced analytic algorithms which automatically update their predictions over time.





# Many Words Used for Similar Concepts

Predictive Analytics

Regression Data Mining

SQL Analytics Anomaly Detection

Python

Adaptive Intelligence

Data Science

Diagnostic Analytics

Classification

AI

Advanced Analytics

Algorithm Descriptive Analytics

SQL

R

Clustering

Artificial Intelligence

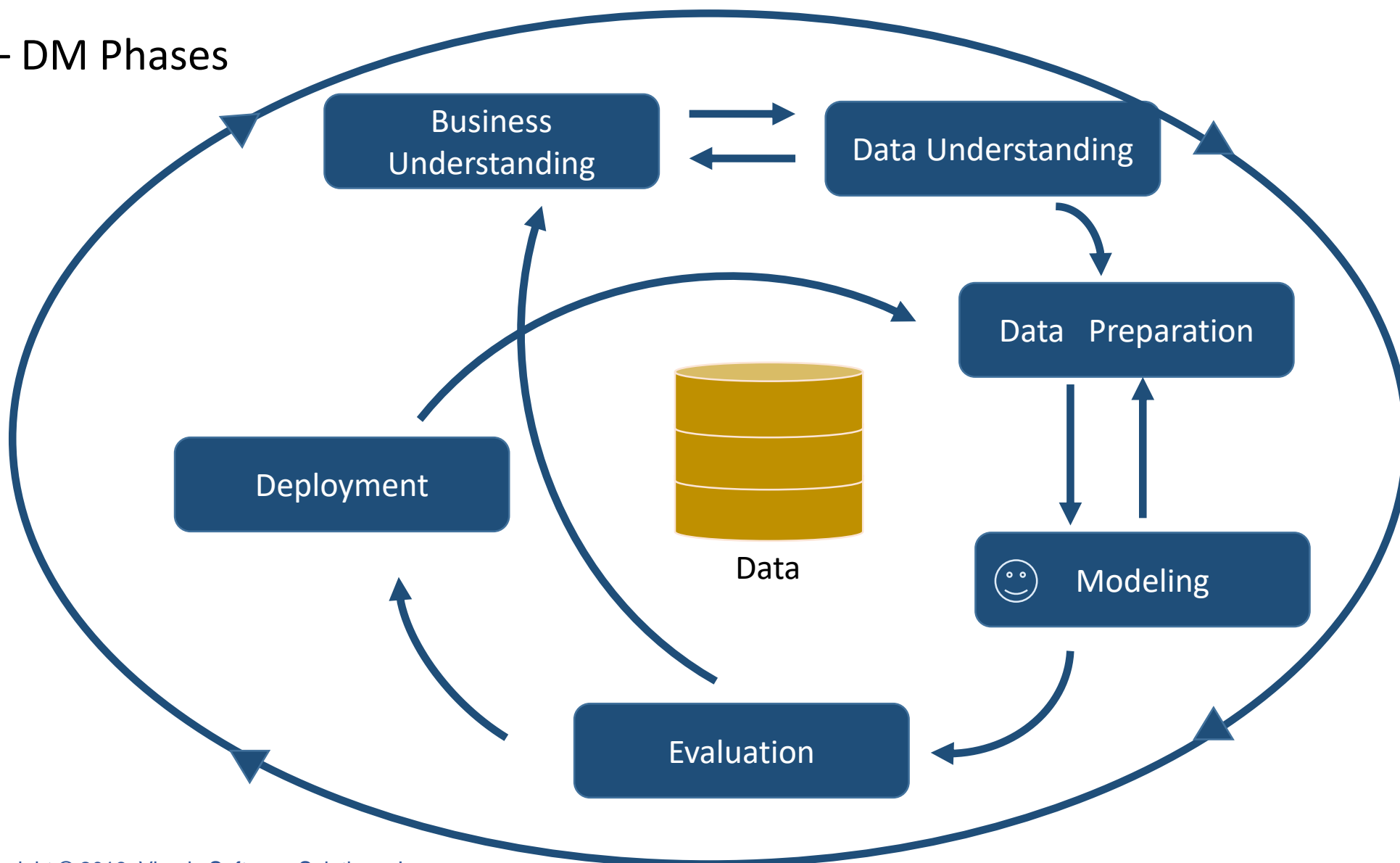
Prescriptive Analytics

Machine Learning



# Implications of Systems That “Learn”

CRISP – DM Phases





# Implications of Systems that “Learn”

- Machine Learning implies a far higher level of involvement and connectedness.
- Being in production implies being at the center of the system.
- Deployment becomes even more important.





# Four Realms of Analytics

Probability Based

**Diagnostic  
Analytics**

**Predictive  
Analytics**

Rules Based

**Descriptive  
Analytics**

**Prescriptive  
Analytics**

Past

Future



# Many Options for ML in OAC

- Work in OAC Data Visualization Project
  - One-click options
  - Use built-in scripts via My Calculations
  - Automated data enrichment in Prepare tab
  - Use Explain for data profiling and unsupervised learning
  - Upload and call custom scripts
- Train and apply models in Data Flows
- Work in “edit formula” in Criteria Tab in Answers Analysis
- Apply custom scripts in Repository
- Connect to Oracle Advanced Analytics in Oracle Database Cloud Service (High Performance or Extreme Performance)
  - Oracle Data Mining
  - Oracle R Enterprise
- Connect to Oracle Machine Learning in Autonomous Data Warehouse Cloud Service



# Many Options for ML in OAC

- 4 Different interfaces with OAC
- 4 Different places with OAC Data Visualization Projects
- Custom script in R or Python
- Integrate with Oracle Advanced Analytics, Oracle Spatial and Graph in Oracle Database Cloud Service
- Integrate with Oracle Machine Learning and Oracle Spatial and Graph in Autonomous Data Warehouse Cloud Service
- Integrate with ORAAH and Oracle Spatial and Graph in Big Data Cloud Service



# Many Options for ML in OAC

- 4 Different interfaces with OAC
  - Data Visualization Project
  - Data Flows train and apply models
  - Machine Learning Inspect Models
  - Classic interface Answers & Dashboards and .rpd evaluate function
- 4 Different Places in DV Projects
  - Visualization on Canvas right click
  - “My Calculations” custom script
  - “Explain”
  - Prepare machine learning enrichments





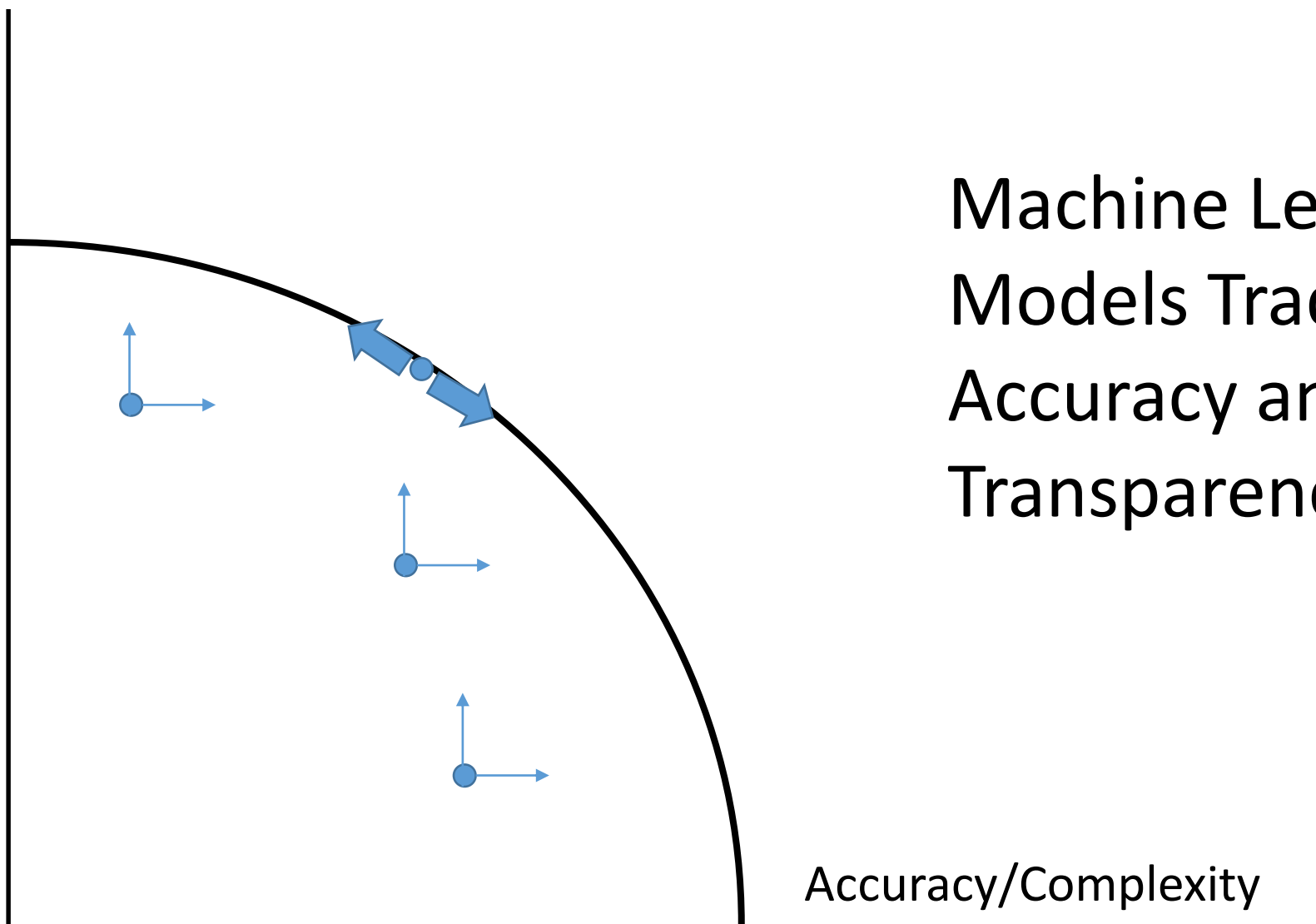
# Tradeoffs Abound

- Explanatory power and transparency vs. Accuracy
- Automated data prep vs. conscious data shaping choices
- Clear visualizations vs. multi-dimensional transforms and relationships
- Ease of use vs. computational understanding



# Horizon Function – Tradeoff Visualization

Transparency/Explanatory Power



Machine Learning  
Models Tradeoff  
Accuracy and  
Transparency.



# Demo of Single-click ML with use cases and limitations

- Clustering
- Outliers
- Trend
- Forecast



# Single-click OAC ML Don'ts

- Do not overinterpret
- Do not be afraid to use
- Do not spend a lot of time when you have more powerful tools





# Explain feature has power

- Demo of “Explain” with use cases and limitation

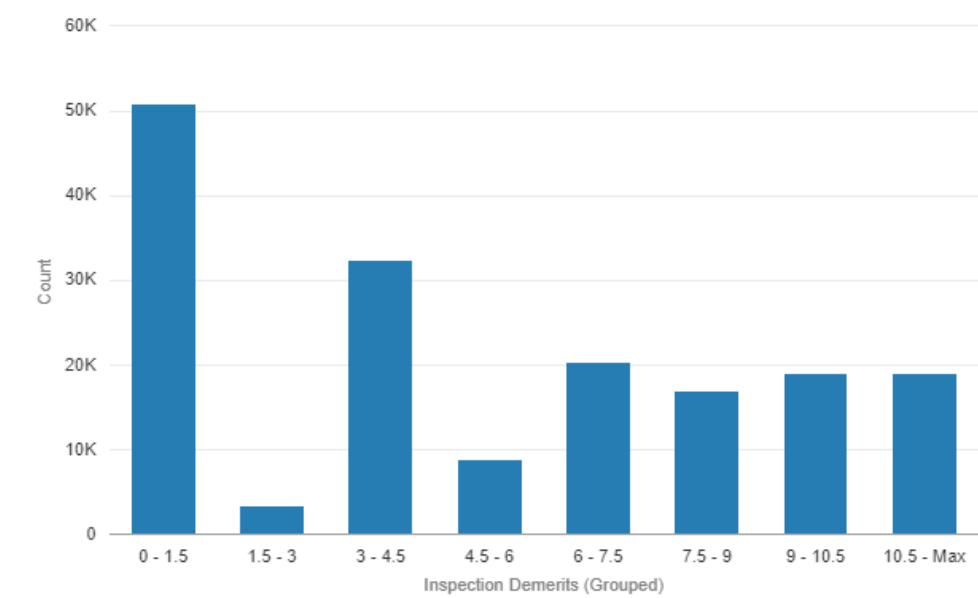
Basic Facts about Inspection Demerits

What are the values of Inspection Demerits and how do they relate to each other?

Anomalies of Inspection Demerits

What groups in the data exhibit unexpected results for Inspection Demerits?

Basic facts about Inspection Demerits



Inspection Demerits is a Numeric Measure, whose sum across 170,463 rows is 979,790.00. The values of Inspection Demerits on each row range from 0.00 to 250.00 and is 5.00 on average.

The charts below summarize the values of Inspection Demerits by the measures in this data set. Click the checkmarks above any of the visuals to add them to your project when done.



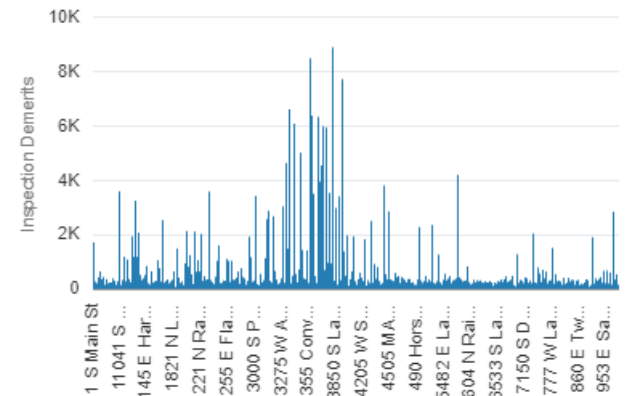
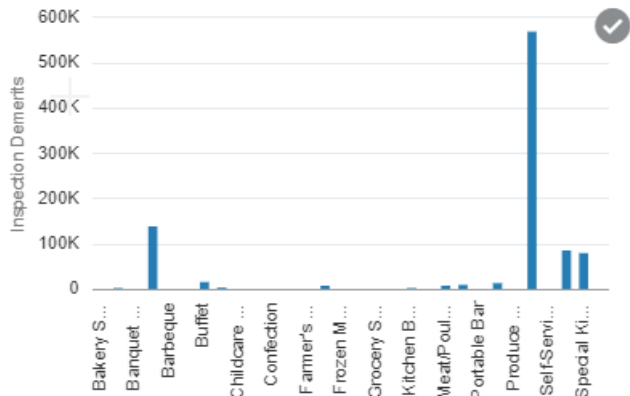
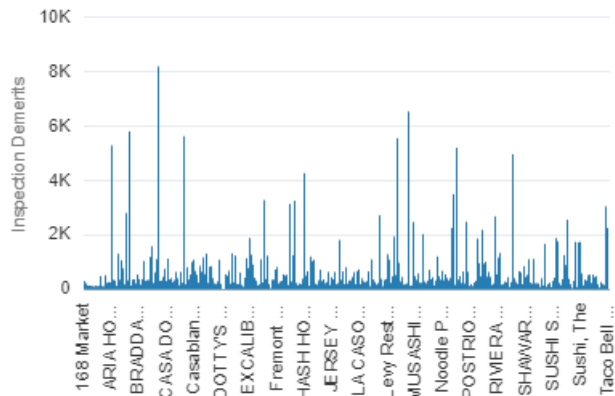
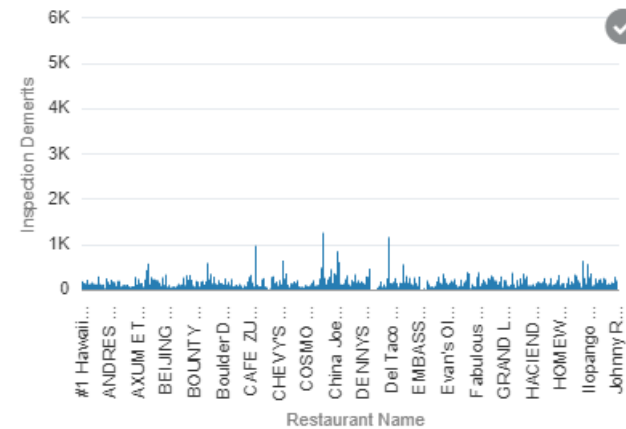
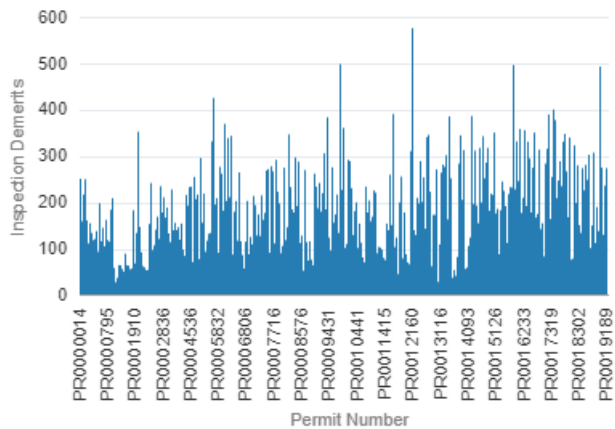
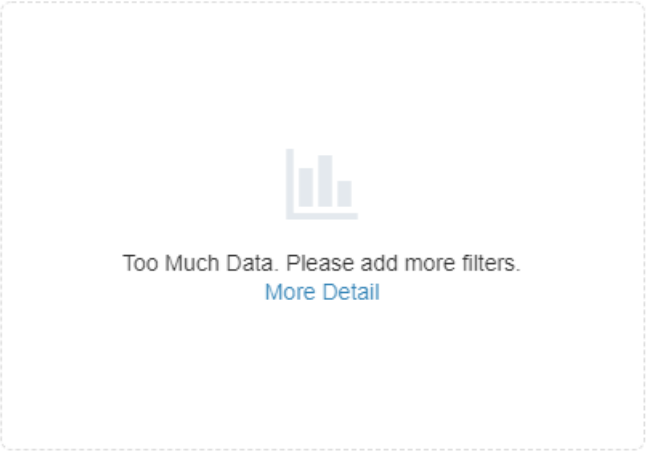
Basic Facts about Inspection Demerits

What are the values of Inspection Demerits and how do they relate to each other?

Anomalies of Inspection Demerits

What groups in the data exhibit unexpected results for Inspection Demerits?

The charts below summarize the values of Inspection Demerits by the measures in this data set. Click the checkmarks above any of the visuals to add them to your project when done.



Basic Facts about Inspection Demerits

What are the values of Inspection Demerits and how do they relate to each other?

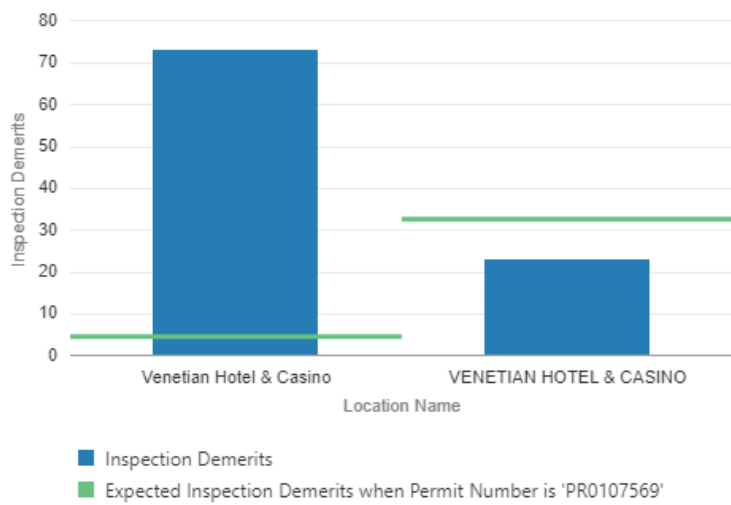
Anomalies of Inspection Demerits

What groups in the data exhibit unexpected results for Inspection Demerits?

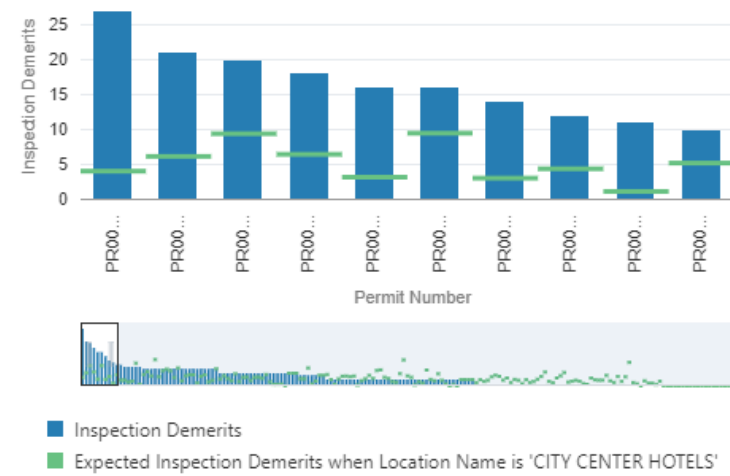
Anomalies of Inspection Demerits

231 combinations of 22 dimensions are being analyzed. Here are the top outliers for Inspection Demerits

When Permit Number is PR0107569, we expected Inspection Demerits for Location Name: Venetian Hotel & Casino to be 4.65, however, it is 73.00, representing a difference of 68.35.



When Location Name is CITY CENTER HOTELS, we expected Inspection Demerits for Permit Number: PR0023967 to be 4.08, however, it is 27.00, representing a difference of 22.92.







# OAC Data Flows

- Requires some knowledge of data mining/machine learning
- Build/train models in data flows
- Use classification for single/multi-class predictions
  - Churn models of customer loyalty
  - Predict buy/not buy specific products
- Use regression models for prediction of continuous values
  - Lifetime customer value prediction
  - Next year customer purchases
- Use clustering to segment members into groups
  - Customer segmentation based on history/buying behavior
- Use sentiment analysis to understand comments



# Building Models in Data Flows (demo)

- Training and testing supervised models
- Classification methods and algorithms in OAC
- Interpretation of results
  - Confusion matrix
  - Accuracy, precision, and recall
  - ROC curve
- Visualizing classification results
- Applying classification models
- Use cases



# Other ML Models in OAC Data Flows

- Sentiment analysis
- Regression
- Clustering



# Oracle ML in Autonomous Data Warehouse

- In-database machine learning
  - Don't move the data
  - Extreme power and scalability
  - Extreme flexibility and extendibility
- Zeppelin Notebooks
  - Great for collaboration
  - Powerful tool in the hands of knowledgeable
  - Built-in visualization capability
  - Script development and management
- Oracle Data Mining Algorithms





# Demo of Oracle Machine Learning in ADW



# Oracle Data Miner in Database Cloud Service

- GUI for building predictive analytics workflows
- Build scripts for oracle database without coding
- Powerful built-in visualizations for interpretation



# Demo of Oracle Data Miner



# Questions to ask yourself

- Do we have people currently on staff who want to execute R or Python models inside the BI system?
- Where do you want to shape data sets for machine learning/predictive analytics?
- Do we currently have clean, consistent, accurate data?
- Do we have an executive champion who understands that systems with have to be grown over time?
- Do we want to start with training or a defined use case?



# Advice

- Start simply and build
- Start (don't wait)
- Make sure data is clean and consistent
- Don't believe you need a lot of data
- Your best data is likely your internal data
- Promote solid interpretation and understanding of models and results
- Be careful of months (days and weeks often work better) with forecast
- Negative values can throw off some models
- Outliers can have very large effects



# Thank You!!

- Questions??
- Tim Vlami [tvlamis@vlamis.com](mailto:tvlamis@vlamis.com)
- Dan Vlami [dlamis@vlamis.com](mailto:dlamis@vlamis.com)



Connected

# My First Notebook

default

learning-sql-notebooks-for-the-oracle-autonomous-data-warehouse-cloud on Oracle Machine Learning blog.

Took 0 sec. Last updated by CHARLIE at October 18 2018, 2:18:52 PM.

## Show all tables

FINISHED

```
%sql
SELECT * FROM all_tables where owner = 'SH';
```

OWNER	TABLE_NAME	TABLESPACE_NAME	CLUSTER_NAME	IOT_NAME	STATUS	PCT_FREE	PCT_USED	INI_TRANS	MAX_TRANS	INITIAL_EXTENT	NEXT_EXTENT	MIN_EXTENTS	MAX_EXTENTS
SH	TIMES	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	PRODUCTS	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	PROMOTIONS	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	COUNTRIES	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	CHANNELS	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	COSTS	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	SALES	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645
SH	SUPPLEMENTARY_DEMOGRAPHICS	SAMPLESCHEMA			VALID	0		1	255	65536	1048576	1	2147483645

Took 1 sec. Last updated by USER01 at March 12 2019, 12:20:52 AM.

## Display table

FINISHED

```
%sql
/* Display SUPPLEMENTARY_DEMOGRAPHICS table */
SELECT * FROM SH.SUPPLEMENTARY_DEMOGRAPHICS;
```

CUST_ID	EDUCATION	OCCUPATION	HOUSEHOLD_SIZE	YRS_RESIDENCE	AFFINITY_CARD	BULK_PACK_DISKETTES	FLAT_PANEL_MONITOR	HOME_THEATER_PACKAGE	BOOKKEEPING_APPLICATION	PRINTER_SUPPL
101860	HS grad	Other	1	1	0	1	1	0	1	1

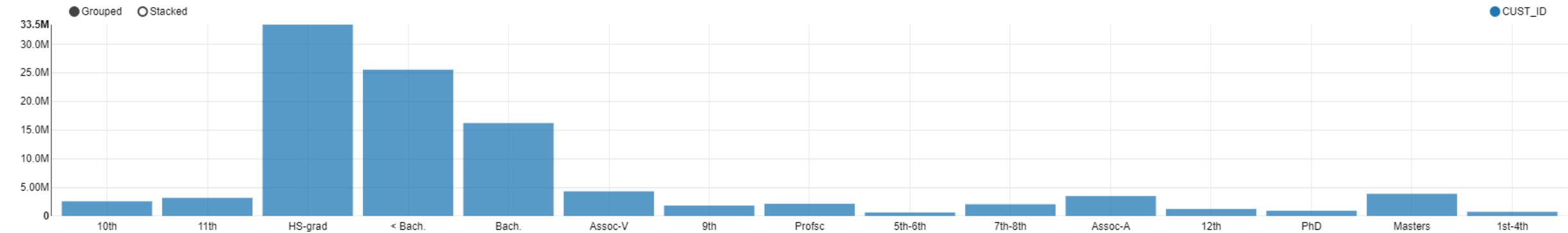


Graph the data as a bar chart of EDUCATION

FINISHED ▶ ⌵ ⌵ ⌵ ⌵

```
%sql
/* Bar chart of EDUCATION (Keys), Cust_ID Sum (Values), nothing in Groups */
SELECT * FROM SH.SUPPLEMENTARY_DEMOGRAPHICS where ROWNUM <1000;
```

Chart type icons: Bar, Line, Pie, Area, Stacked, Sunburst, Settings



Took 0 sec. Last updated by CHARLIE at October 18 2018, 2:22:22 PM. (outdated)

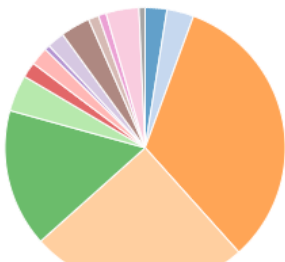
Graph EDUCATION

FINISHED ▶ ⌵ ⌵ ⌵ ⌵

```
%sql
/* Create pie chart of EDUCATION (Keys), NA (Groups), CUST_ID (SUM) (Values) */
SELECT * FROM SH.SUPPLEMENTARY_DEMOGRAPHICS where rownum < 1000
```

Chart type icons: Bar, Line, Pie, Area, Stacked, Sunburst, Settings

- 10th, 11th, HS-grad, < Bach., Bach., Assoc-V, 9th, Profsc, 5th-6th, 7th-8th, Assoc-A, 12th, PhD, Masters, 1st-4th



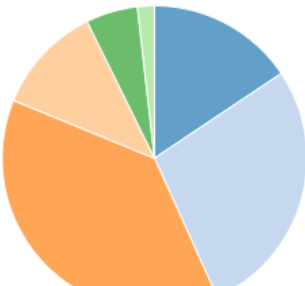
Graph HOUSEHOLD\_SIZE

FINISHED ▶ ⌵ ⌵ ⌵ ⌵

```
%sql
/* Create pie chart of HOUSEHOLD_SIZE (Keys), NA (Groups), CUST_ID (SUM) (Values) */
SELECT * FROM SH.SUPPLEMENTARY_DEMOGRAPHICS where rownum < 1000
```

Chart type icons: Bar, Line, Pie, Area, Stacked, Sunburst, Settings

- 1, 2, 3, 9+, 4-5, 6-8



ORACLE

Machine Learning

USER01 Project [USER01 Workspace]

USER01

Connected

Clustering

▶ ⌂ 📄 ✎ 🗑️ ⬇️ 🔍 🖨️ ⚙️ default

Finished

▶ ⌂ 📄 ⚙️

For more information, check the Oracle ADWC Documentation <https://docs.oracle.com/en/cloud/paas/autonomous-data-warehouse-cloud/index.html>, Oracle Machine Learning folder on Oracle on Github <https://github.com/oracle>, Oracle Advanced Analytics <http://www.oracle.com/technetwork/database/options/advanced-analytics/overview/index.html> and Oracle Machine Learning on Oracle Technology Network and Introducing Oracle Machine Learning blog post

Finished ▶ ⌂ 📄 ⚙️

Clean up previously existing tables and create Data table

Finished ▶ ⌂ 📄 ⚙️

```
%script
/* Clean up and drop any CUSTOMERS360 table if previously exists for notebook reproducibility */
BEGIN
    EXECUTE IMMEDIATE 'DROP Table CUSTOMERS360';
EXCEPTION
    WHEN OTHERS THEN NULL;
END;
/

/* JOIN selected attributes from SH.CUSTOMERS and SH.SUPPLEMENTARY_DEMOGRAPHICS tables to create better 360 view of customer */
Create table CUSTOMERS360 as SELECT a.CUST_ID, a.CUST_GENDER, a.CUST_MARITAL_STATUS, a.CUST_YEAR_OF_BIRTH, a.CUST_INCOME_LEVEL, a.CUST_CREDIT_LIMIT, b.EDUCATION, b.AFFINITY_CARD, b.HOUSEHOLD_SIZE, b.OCCUPATION, b.YRS_RESIDENCE, b.Y_BOX_GAMES
FROM SH.CUSTOMERS a, SH.SUPPLEMENTARY_DEMOGRAPHICS b
WHERE a.CUST_ID = b.CUST_ID;
```

PL/SQL procedure successfully completed.

-----

Table CUSTOMERS360 created.

-----

outdated

Display CUSTOMERS360 table

PENDING 🚫 ⌂ 📄 ⚙️

```
%sql
SELECT *FROM CUSTOMERS360;
```

outdated

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SOFTWARE SOLUTIONS

## Create a Model Settings table

```
%script
DECLARE
v_sql varchar2(100);

/* Create a Build Setting (DT) for K-Means Model Build */

BEGIN
v_sql := 'CREATE TABLE km_sh_sample_settings (setting_name VARCHAR2(30),setting_value VARCHAR2(4000))';
EXECUTE IMMEDIATE v_sql;
DBMS_OUTPUT.PUT_LINE (v_sql || ': succeeded');
EXCEPTION
WHEN OTHERS THEN
DBMS_OUTPUT.PUT_LINE (v_sql || ': drop unnecessary - no table exists');
END;
/
```

```
CREATE TABLE km_sh_sample_settings (setting_name VARCHAR2(30),setting_value
VARCHAR2(4000)): succeeded
```

PL/SQL procedure successfully completed.

-----

Took 0 sec. Last updated by USER01 at March 11 2019, 11:52:44 PM.

## Define the Model Settings

```
%script

/* Create Model Settings table */

BEGIN
INSERT INTO km_sh_sample_settings (setting_name, setting_value) VALUES
(dbms_data_mining.kmns_distance,dbms_data_mining.kmns_euclidean);

INSERT INTO km_sh_sample_settings (setting_name, setting_value) VALUES
(dbms_data_mining.prep_auto,dbms_data_mining.prep_auto_on);

-- Other examples of overrides are:
-- (dbms_data_mining.kmns_iterations,3);
-- (dbms_data_mining.kmns_block_growth,2);
-- (dbms_data_mining.kmns_conv_tolerance,0.01);
-- (dbms_data_mining.kmns_split_criterion,dbms_data_mining.kmns_variance);
-- (dbms_data_mining.kmns_min_pct_attr_support,0.1);
-- (dbms_data_mining.kmns_num_bins,10);

END;
/
```

PL/SQL procedure successfully completed.

Sales, Profit Ratio by Customer Name, Clusters





# Edit Calculation

Name

Customer Cluster KM 6

*f(x)*

```
CLUSTER((Customer Name), (Sales,  
(cast(Profit as double) / Sales), (Sales / # of  
Orders), # of Orders ), 'clusterId',  
'algorithm=k-  
means;numClusters=%1;maxIter=%2;useRandomSe  
ed=FALSE;enablePartitioning=TRUE', 6, 10)
```

Validate

Save

Cancel

## Cluster

This function groups a set of records into groups based on one or more input expressions using K-Means or Hierarchical Clustering.

*CLUSTER*((dimension\_expr1 , ...  
dimension\_exprN), (expr1, ... exprN),  
output\_column\_name, options,  
[runtime\_binded\_options])

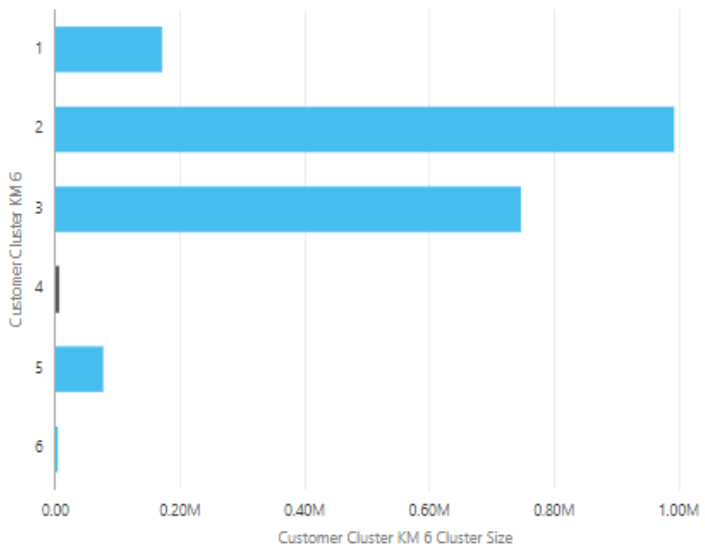
*dimension\_expr* represents a list of dimensions , e.g. (productId, companyID), to be clustered.

*expr* represents a list of dimension attributes or measures to be used to cluster the dimension\_expr.

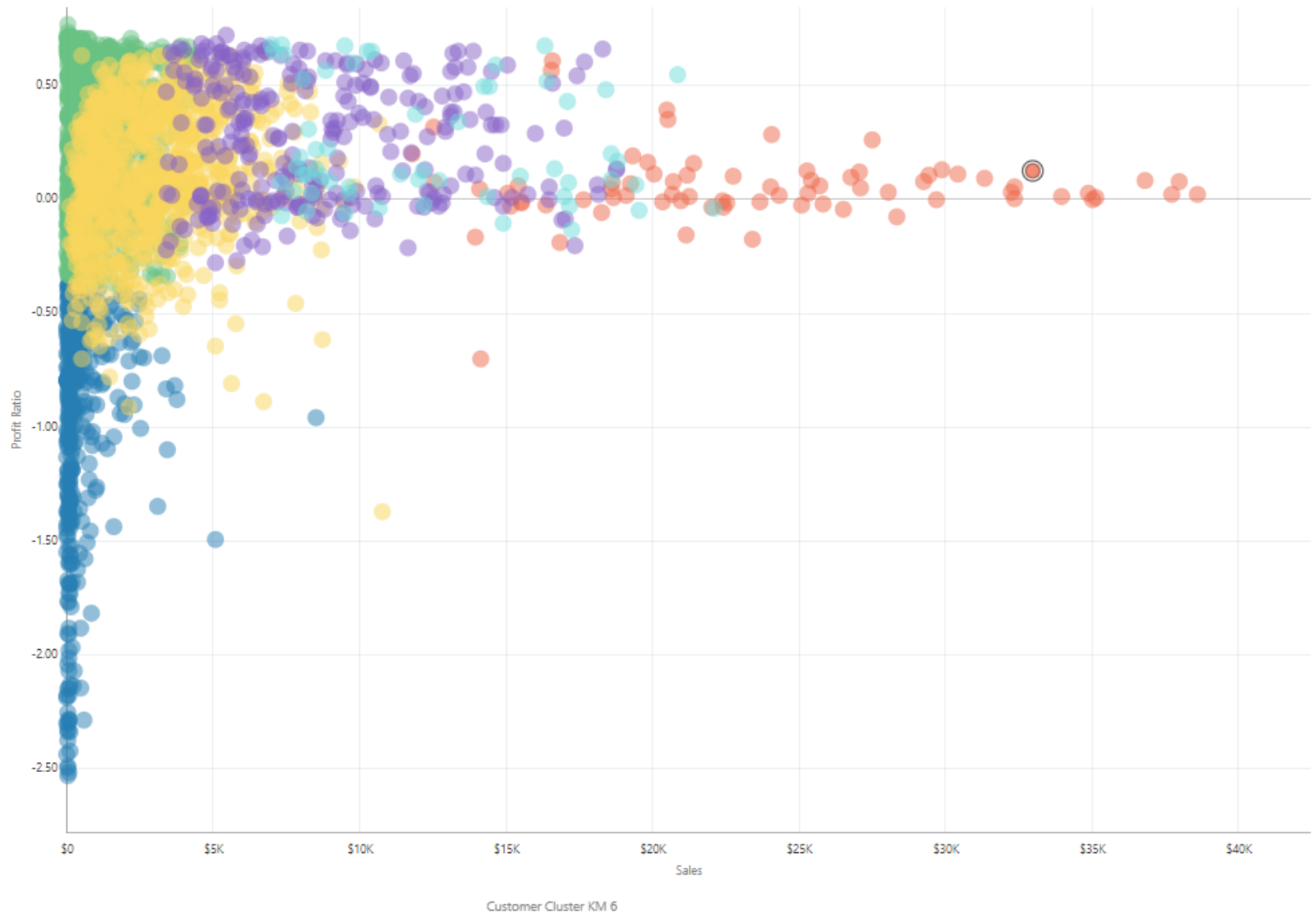
*output\_column\_name* is the output column. The valid values are "clusterId", "clusterName", "clusterDescription"

to add a filter

Customer Cluster KM 6 Cluster Size by Customer Cluster KM 6

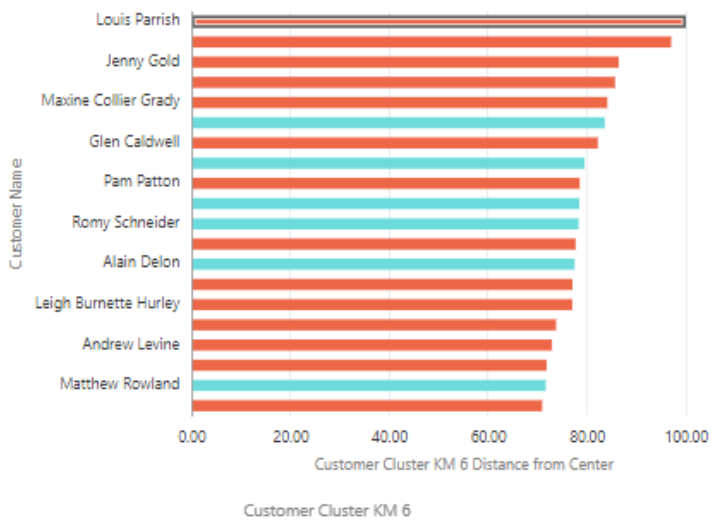


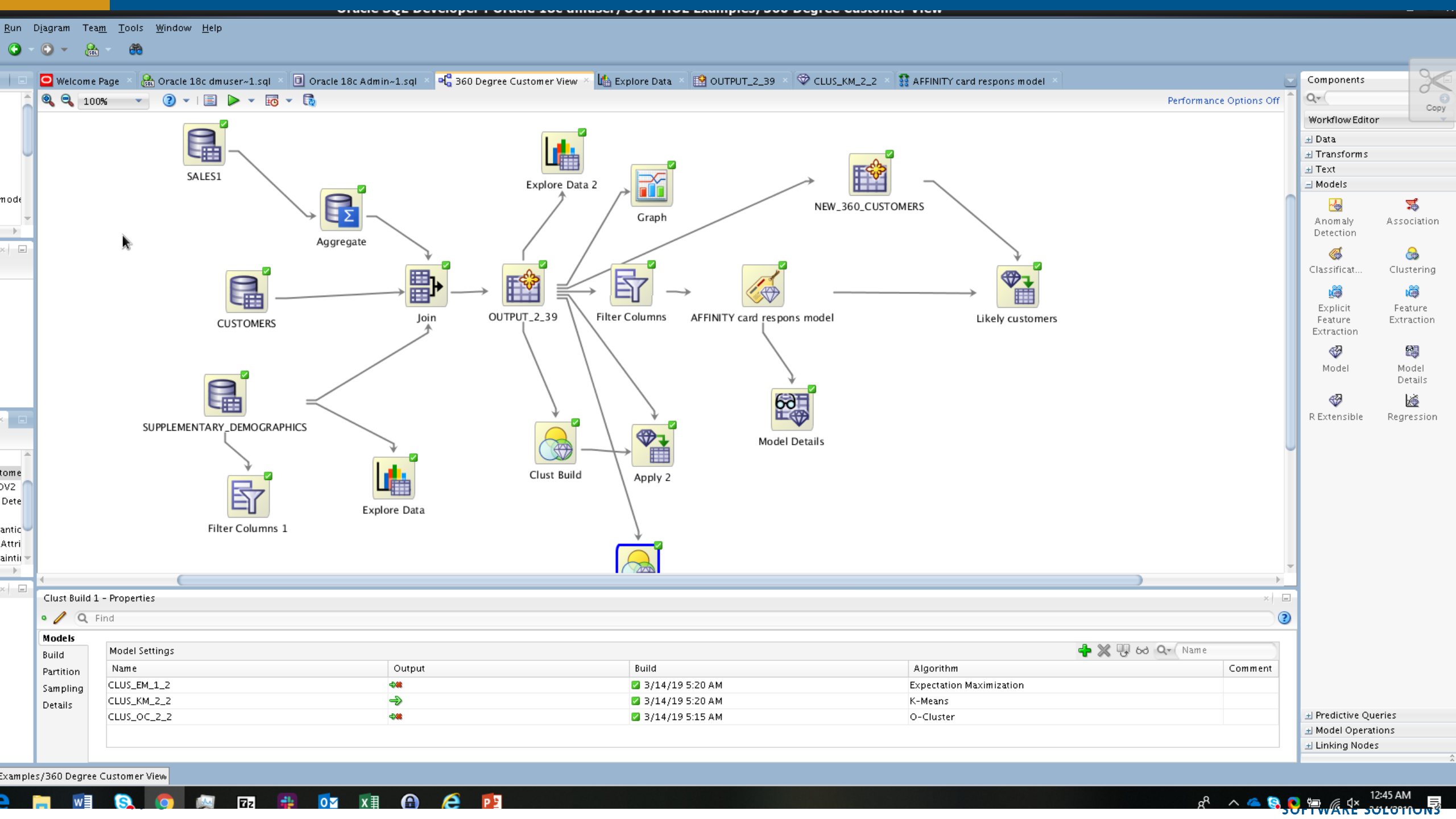
Sales, Profit Ratio by Customer Name, Customer Cluster KM 6



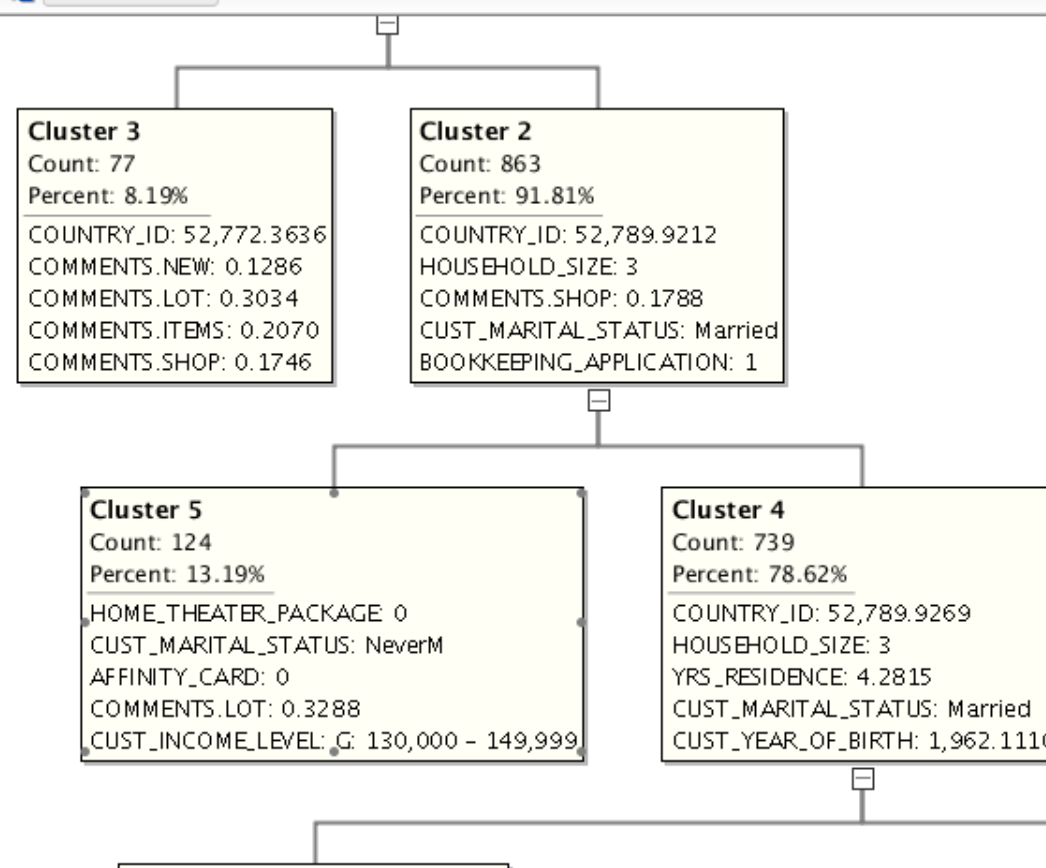
Customer Cluster KM 6 Distance from Center by Customer Name, Customer Cluster...

Top 20 Customer Cluster KM 6 Distance from Center







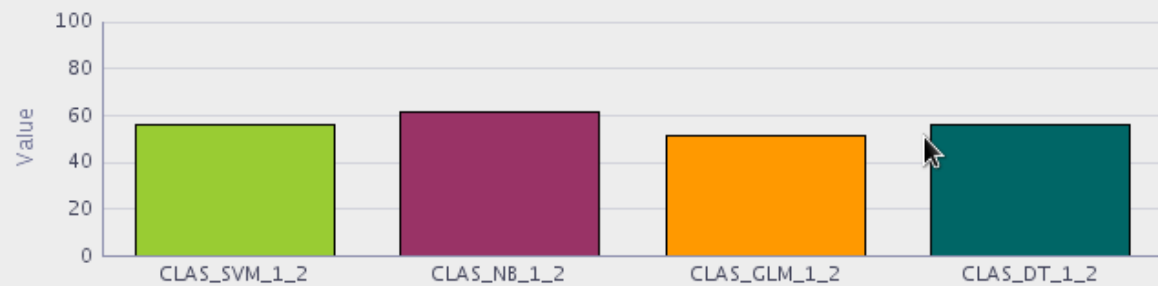


Centroid	Rule					
Name	Importance	Mode	Mean	Variance	Support	
HOME_THEATER_PACKAGE	0.3676	0				124
Y_BOX_GAMES	0.3182	1				115
FLAT_PANEL_MONITOR	0.2703	0				119
BULK_PACK_DISKETTES	0.2244	0				110
CUST_YEAR_OF_BIRTH	0.1217		1,979.06451613	16.80881196		120
YRS_RESIDENCE	0.0904		2.26612903	0.83103855		124
CUST_INCOME_LEVEL	0.0885	G: 130,000 - 149,999				117
COMMENTS.LOT	0.0874		0.32879328	0.00020443		15

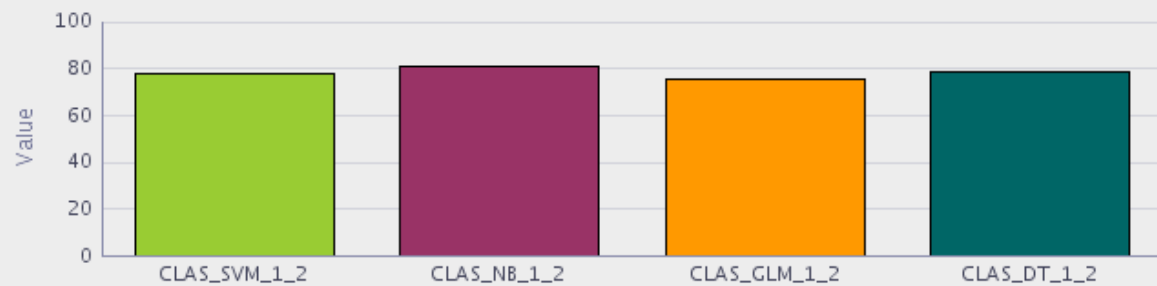
Performance | Performance Matrix | ROC | Lift | Profit

Measure: All Measures
 Sort By: Name
 Descending

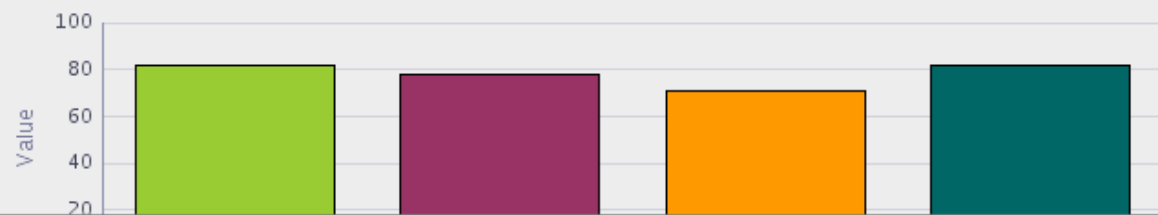
Predictive Confidence (%)



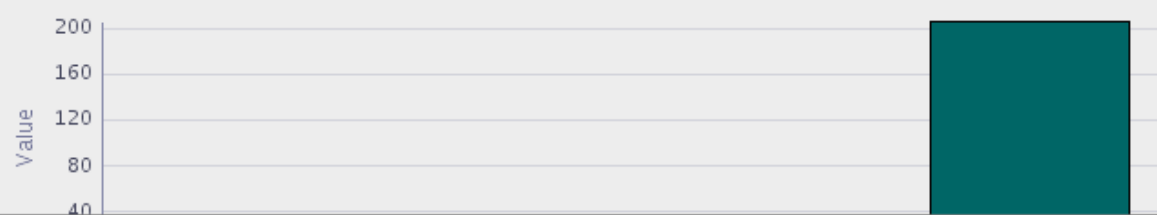
Average Accuracy (%)



Overall Accuracy (%)



Cost



Models

Name	Predictive Confidence %	Overall Accuracy %	Average Accuracy %	Cost	Algorithm	Creation Date
CLAS_DT_1_2	56.2715	81.7391	78.1357	204.96894410	Decision Tree	10/16/18 10:00 PM
CLAS_GLM_1_2	51.2016	70.4348	75.6008		Generalized Linear Mo...	10/16/18 10:00 PM
CLAS_NB_1_2	61.2716	77.6087	80.6358		Naive Bayes	10/16/18 10:00 PM
CLAS_SVM_1_2	55.9849	81.5217	77.9925		Support Vector Machine	10/16/18 10:00 PM



# Thank You!

## Questions?

Tim Vlami

[tvlamis@vlamis.com](mailto:tvlamis@vlamis.com)



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