## Multi-Agent AutoML Report For OTICS

Generated On: 2024-04-11 21:43:06 (EDT)

Best Model(s) Report For admin\_aesopowerdemandlogistic\_csv

## MODEL DESCRIPTION

Model Trained On: 2024/04/11 Training Start Time: 2139
Training End Time: 2143 Iraining End Time: 2143 Was Data Normalized: Yes Was Data Shuffled: Yes Deep Analysis: No Total Training Data Set: 961 Training Data Percentage: 75% Total Test Data Set: 319 Total # of Variables: 4 Adjusted for Seasonality: N Total Algorithms Run: 450 Removed Outliers: Y ROC AUC: 0.631

Roc Adc: 0.631 Precision: 0.785 (0.311 For Class=0) Recall: 0.650 (0.470 For Class=0) F1-Score: 0.711 (0.375 For Class=0)

Best Distribution FOR ACTUAL Y: RECIPINVGAUSS
Dependent Variable: AESO POWER DEMAND LABEL

Independent Variables: ['Calgary\_Weather', 'Edmonton\_Weather', 'FtMac\_Weather']

# **Receiver Operating Characteristic Curve (ROC)** Using VotingClassifier algorithm for allseason Receiver Operating Characteristic (ROC) 1.0 9.0 g 을 0.4 0.2 ROC curve for Class 1 (area = 0.63)

False Positive Rate

0.6

0.8

1.0

## IMPORTANT FILE PATHS FOR RAW AND OUTPUT DATA

0.0 0.0

0.2

NOTE: These are DOCKER CONTAINER Paths. You can view these files inside the container by using the command: docker exec -it {container id} bash If you have re-run the container, these files will be GONE but they exist on your HOST machine. The HOST MACHINE location is based on the volumes you mapped when you ran the Docker container. The Docker RUN Volume Mappings are :: (For example here is the docker run command (use multiple -v for multiple mappings):

DOCKER RUN COMMAND: docker run -d -p 5595:5595 -p 5495:5495 -p 10000:10000 -v {HOST MACHINE FOLDER}:{CONTAINER FOLDER}:z --env TRAININGPORT=5595 --env PREDICTIONPORT=5495 --env ABORTPORT=10000 --env COMPANYNAME=MYCOMPANY --env MAXRUNTIME=20 --env MAINHOST=127.0.0.1 maadsdocker/maads-batchautoml-otics

### Docker Volume Mappings:

- 1. {HOST MACHINE FOLDER}/csvuploads:/mnt/c/maads/agentfilesdocker/dist/maadsweb/csvuploads:z 2. {HOST MACHINE FOLDER}/pdfreports:/mnt/c/maads/agentfilesdocker/dist/maadsweb/pdfreports:z

- 2. (HOST MACHINE FOLDER)/putreports:/mmt/c/maads/agentfilesdocker/dist/maadsweb/autofeatures:2
  4. (HOST MACHINE FOLDER)/autilers:/mmt/c/maads/agentfilesdocker/dist/maadsweb/autofeatures:2
  5. (HOST MACHINE FOLDER)/sqlloads:/mnt/c/maads/agentfilesdocker/dist/maadsweb/gloads:2
  6. (HOST MACHINE FOLDER)/sqlloads:/mnt/c/maads/agentfilesdocker/dist/maadsweb/etworktemp:2
  7. (HOST MACHINE FOLDER)/networktemp:/mnt/c/maads/agentfilesdocker/dist/maadsweb/networktemp:2
  8. (HOST MACHINE FOLDER)/networks:/mnt/c/maads/agentfilesdocker/dist/maadsweb/networktemp:2
  8. (HOST MACHINE FOLDER)/networks:/mnt/c/maads/agentfilesdocker/networks:2
  8. (HOST MACHINE FOLDER)/networks://mnt/c/maads/agentfilesdocker/networks:2
  8. (HOST MACHINE FOLDER)/networks://mnt/c/maads/agentfilesdocker/networks:2
  8. (HOST MACHINE FOLDER)/networks:2
  8. (HOST MACHINE FOLDER)/netwo
- {HOST MACHINE FOLDER}/exception:/mnt/c/maads/agentfilesdocker/dist/maadsweb/exception:z

- 6. (HOST MACHINE FOLDER)/steepuoli./Imit/c/maads/agentfilesdocker/dist/staging:

  9. (HOST MACHINE FOLDER)/steepuoli./Imit/c/maads/agentfilesdocker/dist/staging:

  Path for Training Dataset File: /mmt/c/maads/agentfilesdocker/dist/maadsweb/csvuploads/aesopowerdemandlogistic.csv

  Path for PDF Report (i.e. this file): /mmt/c/maads/agentfilesdocker/dist/maadsweb/pdfreports/admin\_aesopowerdemandlogistic\_csv\_no\_seasons.pdf

  Path for AutoFeature File: /mmt/c/maads/agentfilesdocker/dist/maadsweb/autofeatures/admin\_aesopowerdemandlogistic\_csv\_csv
- Path for Outliers File: /mnt/c/maads/agentfilesdocker/dist/maadsweb/outliers/admin\_aesopowerdemandlogistic\_csv.csv
  Path for Algo JSON File: /mnt/c/maads/agentfilesdocker/dist/maadsweb/exception/admin\_aesopowerdemandlogistic\_csv\_trained\_algo\_no\_seasons.json
  Folder Path for MySQL Scripts: /mnt/c/maads/agentfilesdocker/dist/maadsweb/sqlloads/
- Path for Detailed Prediction File: /mnt/c/maads/agentfilesdocker/dist/maadsweb/csvuploads/admin\_aesopowerdemandlogistic\_csv\_prediction\_details.csv
  Path for Algorithm Zip File (i.e pickle files): /mnt/c/maads/agentfilesdocker/dist/maadsweb/networktemp/admin\_aesopowerdemandlogistic\_csv.zip
  Path for Algorithm Pickle Files:

- rain for Algorithm Figure Fues:

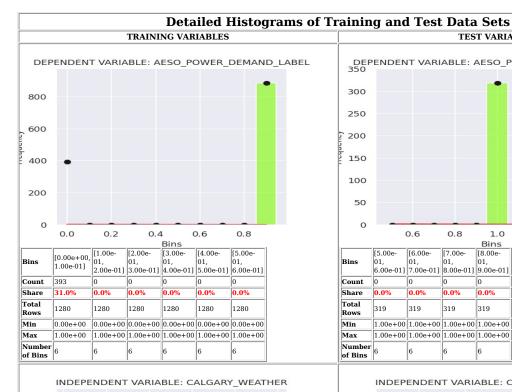
  1. /mmt/c/maads/agentfilesdocker/networks/otics\_ADMIN\_AESOPOWERDEMANDLOGISTIC\_CSVALLSEASON\_AG1\_4\_VotingClassifier\_normal\_961\_ensemble\_pkl

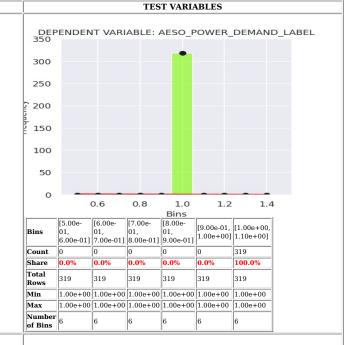
  2. /mmt/c/maads/agentfilesdocker/networks/otics\_ADMIN\_AESOPOWERDEMANDLOGISTIC\_CSVALLSEASON\_AG1\_4\_VotingClassifier\_normal\_961\_ensemble\_scalerx\_pkl

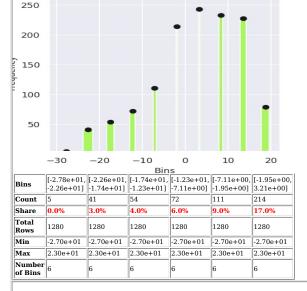
### DESCRIPTIVE STATISTICS Variables T-Statistic Count Mean STD MIN 25% 50% 75% MAX Calgary Weather -3.505 5.862 -27.75 10.113 -0.4 14.0 23.85 961.0 6.85 Edmonton\_Weather -3.116 961.0 5.916 -26.64 11.759 2.25 7.16 16.1 25.75 FtMac Weather 3562660013702567.5 961.0 2.367 -32.4 13 694 -7 65 4 56 14 8 23.85 AESO POWER DEMAND LABEL 0.591 0.492 NA 961.0 0.0 0.0 1.0 1.0 1.0

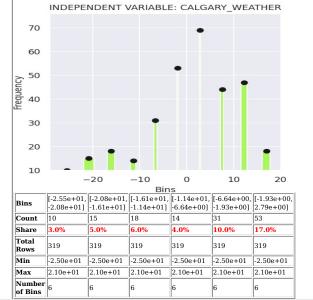
BEST ALGORITHM FOUND FOR THIS DATASET  (Note: This trained model will be used to predict AESO_POWER_DEMAND_LABEL)								
Algorithm	Description	Model Results	ROC/AUC	Precision	Recall	F1- Score	Forecast Months	Season
VotingClassifier	Voting Classifier: Combination of different classifers (DecisionTree, RandomForest,K nearest neighbour,GaussNB,Extra tree,ADA boost, etc) (Info)	VotingClassifier(estimators=[('dt', DecisionTreeClassifier(class weight='balanced', max features='log2', max leaf nodes=3, min_impurity_decrease=0.06646943636248269, min_samples_leaf=8, min_samples_split=5, min_weight_fraction_leaf=0.3194174662332445, random_state=45)), ('knn', KNeighborsClassifier(algorithm='brute', leaf_size=5, n_neighbors=1, p=6)), ('lr', Logi AdaBoostClassifier(learning_rate=2.0370480811234395, n_estimators=34, random_state=40)), ('et', ExtraTreesClassifier(class_weight='balanced_subsample', criterion='entropy', max_depth=9, min_impurity_decrease=1.0955448805815036, min_samples_leaf=4, min_samples_split=3, min_weight_fraction_leaf=0.49069151202783584, n_estimators=43, random_state=143))], voting='soft')	0.500	0.785: Class=1 (0.311: Class=0)	0.650: Class=1 (0.470: Class=0)	0.711: Class=1 (0.375: Class=0)		allseason

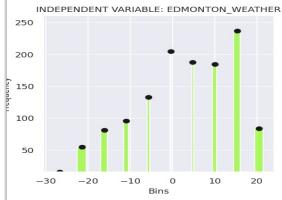
	TOP 10 ALGORITHMS FOR ALLSEASON						
Num	Algorithm	ROC/AUC	Details	Season	Description		
1	VotingClassifier	0.5000	Recall: 0.65 (class 1) Precision: 0.785 (class 1) F1 Score: 0.711 (class 1) Recall: 0.47 (class 0) Precision: 0.311 (class 0) F1 Score: 0.375 (class 0)	allseason	VOTING CLASSIFIER: Combination of different classifers (DecisionTree, RandomForest, K nearest neighbour, GaussNB, Extra tree, ADA boost, etc) URL=http://scikit-learn.org/stable/modules/generated/sklearn.ens emble.VotingClassifier.html		
			False Positive Rate: 13.3% True Negaive Rate: 13.3% False Negative Rate: 26.2% True Positive Rate: 48.7%				
2	LogisticRegression	0.4810	Recall: 0.452 (class 1) Precision: 0.94 (class 1) F1 Score: 0.611 (class 1)  Recall: 0.914 (class 0) Precision: 0.359 (class 0) F1 Score: 0.516 (class 0) False Positive Rate: 2.2% True Negaive Rate: 2.2% True Positive Rate: 33.8% False Negative Rate: 41.0%	allseason	LOGISTIC REGRESSION: URL=http://scikit-learn.org/stable/modules/generated/sklearn.lin ear_model.LogisticRegression.html#sklearn.linear_model.LogisticRegression.predict		
3	VotingClassifier_knclassifier	0.4360	Recall: 0.659 (class 1) Precision: 0.789 (class 1) F1 Score: 0.718 (class 1) Recall: 0.477 (class 0) Precision: 0.32 (class 0) F1 Score: 0.383 (class 0) False Positive Rate: 13.2% True Negaive Rate: 13.2% True Positive Rate: 25.5% True Positive Rate: 49.3%	allseason	K-NEAREST NEIGHBOUR: k-nearest neighbour URL=http://scikit-learn.org/stable/modules/neighbors.html#neares t-neighbors-classification		

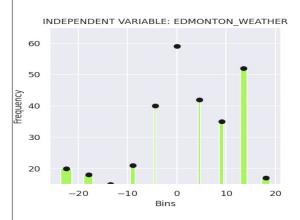






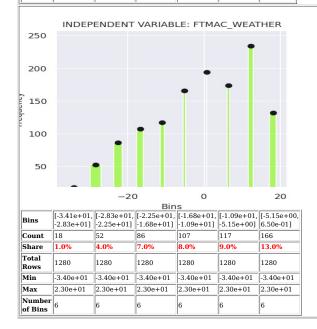


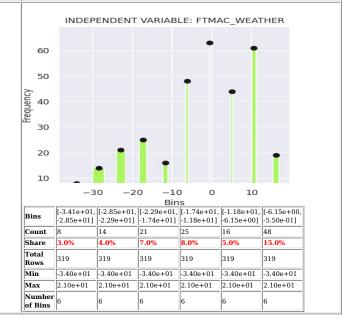


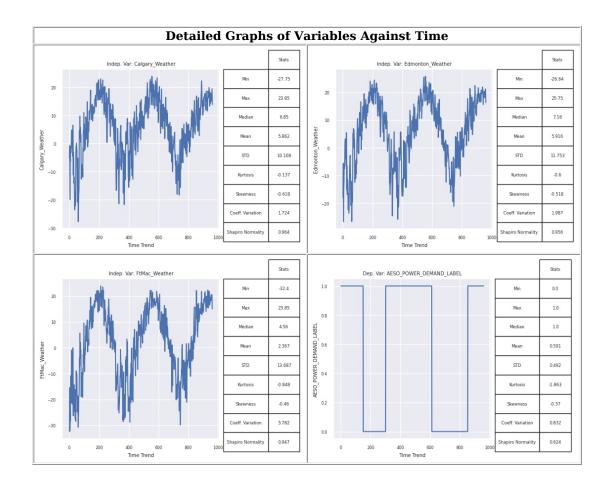


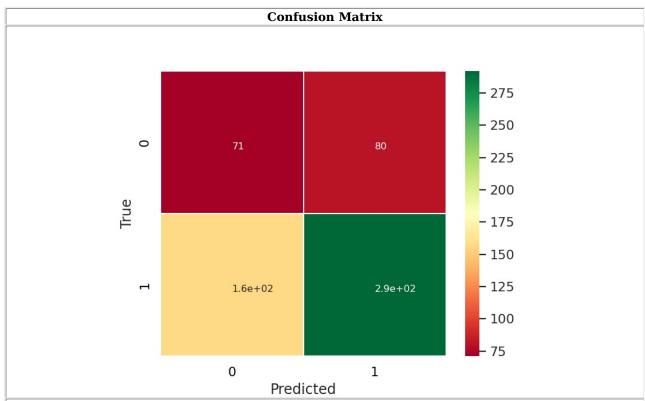
Bins	[-2.66e+01, -2.14e+01]	[-2.14e+01, -1.62e+01]	[-1.62e+01, -1.09e+01]	[-1.09e+01, -5.68e+00]		[-4.45e- 01, 4.79e+00]
Count	16	55	81	96	133	205
Share	1.0%	4.0%	6.0%	8.0%	10.0%	16.0%
Total Rows	1280	1280	1280	1280	1280	1280
Min	-2.60e+01	-2.60e+01	-2.60e+01	-2.60e+01	-2.60e+01	-2.60e+01
Max	2.50e+01	2.50e+01	2.50e+01	2.50e+01	2.50e+01	2.50e+01
Number of Bins	6	6	6	6	6	6

D:	[-2.24e+01,	[-1.79e+01,	[-1.34e+01,	[-8.91e+00,	[-4.42e+00,	[7.50e-02,
Bins	-1.79e+01]	-1.34e+01]	-8.91e+00]	-4.42e+00]	7.50e-02]	4.57e+00]
Count	20	18	15	21	40	59
Share	6.0%	6.0%	5.0%	7.0%	13.0%	18.0%
Total Rows	319	319	319	319	319	319
Min	-2.20e+01	-2.20e+01	-2.20e+01	-2.20e+01	-2.20e+01	-2.20e+01
Max	2.20e+01	2.20e+01	2.20e+01	2.20e+01	2.20e+01	2.20e+01
Number of Bins	6	6	6	6	6	6









The confusion matrix shows the True Negatives (top left)/True Positives (bottom right) on the diagonals, and False Negatives (top right) and False Positives (bottom left).

True Positives: 80 False Positives: 71 True Negatives: 157 False Negatives: 157 Total Population: 600

The False Positve Rate(FPR) is: 13.33% The False Negative Rate is: 26.17% The True Positive Rate is: 48.67% The True Negative Rate is: 11.83%

The Positive Likelihood Ratio (True Positive Rate/False Positive Rate)is: 3.65 The Negative Likelihood Ratio (False Negative Rate/True Negative Rate) is: 2.21

Accuracy: 0.5 Precision: 0.785 Recall: 0.65 F1 Score: 0.711

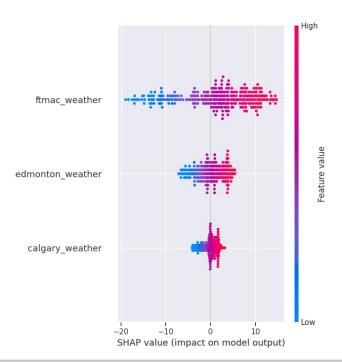
Precision Curve: [0.748, 0.749, 0.747, 0.746, 0.748, 0.750, 0.749, 0.751, 0.751, 0.752, 0.754, 0.756, 0.755, 0.760, 0.759, 0.762, 0.761, 0.762, 0.761, 0.760, 0.761, 0.762, 0.763, 0.765, 0.768, 0.769, 0.768, 0.768, 0.768, 0.767, 0.768, 0.770, 0.771, 0.771, 0.775, 0.775, 0.775, 0.775, 0.777, 0.776, 0.783, 0.785, 0.786, 0.786, 0.791, 0.790, 0.792, 0.796, 0.795, 0.797, 0.796, 0.799, 0.799, 0.798, 0.797, 0.799, 0.790, 0.799, 0.790, 0.796, 0.796, 0.796, 0.796, 0.794, 0.794, 0.793, 0.793, 0.805

Recall Curve: [1.000, 0.998, 0.989, 0.973, 0.971, 0.955, 0.935, 0.920, 0.902, 0.891, 0.886, 0.884, 0.873, 0.860, 0.851, 0.846, 0.842, 0.842, 0.835, 0.833, 0.822, 0.815, 0.808, 0.804, 0.802, 0.802, 0.797, 0.795, 0.791, 0.791, 0.791, 0.788, 0.786, 0.782, 0.782, 0.775, 0.773, 0.766, 0.757, 0.755, 0.755, 0.755, 0.753, 0.744, 0.742, 0.739, 0.739, 0.739, 0.733, 0.731, 0.728, 0.724, 0.724, 0.722, 0.719, 0.717, 0.715, 0.713, 0.713, 0.710, 0.706, 0.702, 0.697, 0.695, 0.688, 0.686, 0.682, 0.673, 0.670, 0.668, 0.664, 0.666, 0.664, 0.659, 0.657, 0.655, 0.653, 0.650, 0.648, 0.639, 0.628, 0.610, 0.599, 0.581, 0.570, 0.552, 0.541, 0.523, 0.5122, 0.497, 0.488, 0.486, 0.481, 0.472, 0.463, 0.448, 0.439, 0.430, 0.419, 0.414, 0.401, 0.396, 0.392, 0.390, 0.388, 0.376, 0.372, 0.367, 0.363, 0.361, 0.354, 0.350, 0.345, 0.345, 0.343, 0.334, 0.325, 0.321, 0.314, 0.307, 0.301, 0.294, 0.285, 0.281, 0.274, 0.267, 0.265, 0.263, 0.261, 0.254, 0.247, 0.241, 0.208, 0.234, 0.229, 0.214, 0.205, 0.192, 0.189, 0.185, 0.183, 0.180, 0.176, 0.171, 0.165, 0.163, 0.158, 0.154, 0.145, 0.136, 0.122, 0.111, 0.102, 0.087, 0.071, 0.047, 0.045, 0.029, 0.024, 0.018, 0.013, 0.009, 0.007, 0.002, 0.000]

Thresholds: [0.408, 0.409, 0.410, 0.411, 0.412, 0.413, 0.414, 0.415, 0.416, 0.417, 0.418, 0.419, 0.420, 0.421, 0.422, 0.423, 0.424, 0.425, 0.426, 0.427, 0.428, 0.429, 0.430, 0.431, 0.432, 0.433, 0.434, 0.435, 0.436, 0.437, 0.438, 0.439, 0.440, 0.441, 0.442, 0.443, 0.444, 0.445, 0.446, 0.447, 0.448, 0.449, 0.450, 0.451, 0.452, 0.455, 0.456, 0.459, 0.460, 0.461, 0.463, 0.465, 0.466, 0.467, 0.470, 0.472, 0.474, 0.475, 0.476, 0.477, 0.480, 0.481, 0.482, 0.484, 0.485, 0.487, 0.488, 0.489, 0.495, 0.496, 0.497, 0.498, 0.499, 0.503, 0.551, 0.552, 0.553, 0.554, 0.555, 0.556, 0.557, 0.558, 0.559, 0.560, 0.561, 0.562, 0.563, 0.564, 0.565, 0.566, 0.567, 0.568, 0.569, 0.570, 0.571, 0.572, 0.573, 0.574, 0.575, 0.576, 0.577, 0.578, 0.579, 0.580, 0.581, 0.582, 0.584, 0.586, 0.587, 0.588, 0.589, 0.590, 0.591, 0.592, 0.593, 0.594, 0.595, 0.596, 0.597, 0.598, 0.599, 0.600, 0.601, 0.602, 0.603, 0.604, 0.605, 0.606, 0.607, 0.608, 0.609, 0.611, 0.612, 0.613, 0.614, 0.615, 0.616, 0.617, 0.619, 0.622, 0.623, 0.626, 0.629, 0.632, 0.633, 0.634, 0.636, 0.637, 0.638, 0.639, 0.640, 0.641, 0.642, 0.6445, 0.646, 0.647, 0.652, 0.655, 0.658]

## $OUTLIERS\ REMOVED\ FROM\ TRAINING\ DATA:\ admin\_aesopowerdemandlogistic\_csv$

## MODEL EXPLANATION



- The x-axis represents the model's output values of AESO\_POWER\_DEMAND\_LABEL

  The plot is centered on the x-axis at explainer.expected value.

  All values are relative to the model's expected value like a linear model's effects are relative to the intercept.

  The y-axis lists the model's features. By default, the features are ordered by descending importance.

  The importance is calculated over the observations plotted. This is usually different than the importance ordering for the entire dataset.

  In addition to feature importance ordering, the decision plot also supports hierarchical cluster feature ordering and user-defined feature ordering.
- Each observation's prediction is represented by a colored line.
  At the top of the plot, each line strikes the x-axis at its corresponding observation's predicted value. This value determines the color of the line on a
- Moving from the bottom of the plot to the top, SHAP values for each feature are added to the model's base value.
  This shows how each feature contributes to the overall prediction.
  At the bottom of the plot, the observations converge at explainer.expected\_value.
  The points in the graph are the values of the feature in the training dataset.

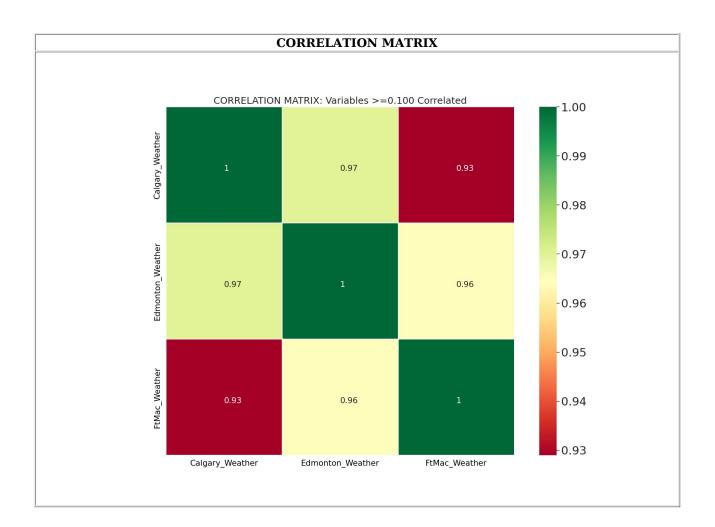
FEATURE SELECTION				
RFE Variable (Most important to Least Important)	Value			
AESO_POWER_DEMAND_LABEL	0.984			
FtMac_Weather	0.008			
Edmonton_Weather	0.005			
Calgary_Weather	0.003			
Best Variable(s) From Genetic Algorithm				
AESO_POWER_DEMAND_LABEL				
FtMac_Weather				
Calgary_Weather				
Excluded Variable(s)				
Edmonton_Weather				
PCA for Best Variable(s)	Value			
AESO_POWER_DEMAND_LABEL_pca_1	0.139			
AESO_POWER_DEMAND_LABEL_pca_2	-0.990			
AESO_POWER_DEMAND_LABEL_pca_3	0.011			
Calgary_Weather_pca_1	-0.701			
Calgary_Weather_pca_2	-0.090			
Calgary_Weather_pca_3	0.708			
FtMac_Weather_pca_1	-0.700			
FtMac_Weather_pca_2	-0.106			
FtMac_Weather_pca_3	-0.706			
PCA Explained Variance	Value			
PCA1	0.649			
PCA2	0.327			
PCA3	0.024			

- Feature selection shows which variables were more influential than other variables

   It uses two core algorithms: Recursive Feature Elimination (RFE) and Genetic Algorithm to determine influence

   It also performs PCA (principal component analysis) analysis to determine the influence of the best variables in the model

   These results should be used in conjunction with other information as well as theory to establish relevance and confidence in the chosen model formulation



CORR	CORRELATED FEATURES				
Feature(s)	Feature(s)	Correlation >= 0.100			
O Calgary_Weather	FtMac_Weather	0.929			
1 Edmonton_Weather	FtMac_Weather	0.964			
2 Calgary_Weather	Edmonton_Weather	0.970			
3 Calgary_Weather	Calgary_Weather	NaN			

SUGGESTED CORRELATED FEATURES TO DELETE						
2 Feature(s) to Delete Correlation						
O Calgary_Weather 0.929						
1 Edmonton_Weather 0.964						

END OF REPORT