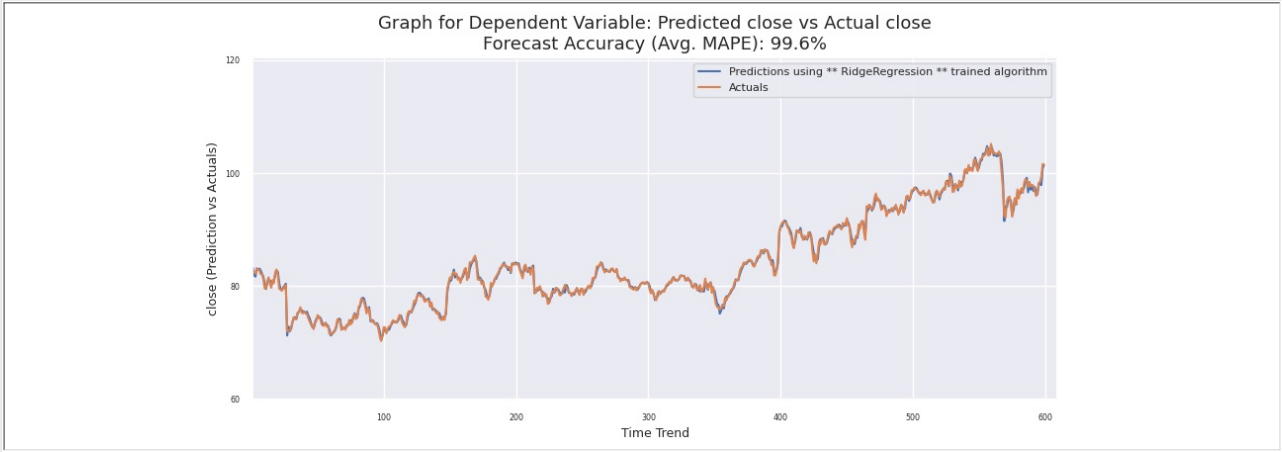


MAADSBML AutoML Report For FIERA CAPITAL

Generated On: 2024-04-24 18:21:35 (UTC)

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



MODEL DESCRIPTION	PREDICTION VARIABLE STATS	ACTUAL VARIABLE STATS
Model Trained On: 2024/04/24 Training Start Time: 1817 Training End Time: 1821 Was Data Normalized: Yes Was Data Shuffled: Yes Deep Analysis: No Total Training Data Set: 946 Training Data Percentage: 75% Total Test Data Set: 314 Total # of Variables: 5 Adjusted for Seasonality: N Total Algorithms Run: 3600 Removed Outliers: N Best Distribution FOR ACTUAL Y: <a href="#">VONMISES</a> Dependent Variable: CLOSE Independent Variables: ['Open', 'High', 'Low', 'Volume']	Mean: 84.297 STD: 8.424 Kurtosis: -0.647 Skewness: 0.611 Coef. of Variation: 0.100 Shapiro Test for Normality: 0.934 Jarque-Bera Goodness of Fit: 47.806 Anderson: 14.697 KStat: 71.074 KStatvar: 11.422 Wilcox: 0.000 Theil Slope: 0.043	Mean: 84.287 STD: 8.449 Kurtosis: -0.658 Skewness: 0.606 Coef. of Variation: 0.100 Shapiro Test for Normality: 0.934 Jarque-Bera Goodness of Fit: 47.497 Anderson: 14.514 KStat: 71.497 KStatvar: 11.462 Wilcox: 0.000 Theil Slope: 0.044
Statistics Showing Comparison Between Prediction and Actuals		
Mood(actuals,predictions): 0.157 Pearson(actuals,predictions): 0.998 Kendall Tau(actuals,predictions): 0.957 Ansari(actuals,predictions): 179951.000 Jaccard_distance(actuals,predictions): 1.000 Minkowski_distance(actuals,predictions): 222.692 Euclidean_distance(actuals,predictions): 11.637		

IMPORTANT FILE PATHS FOR RAW AND OUTPUT DATA

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-batch-automl-otics`

Docker Volume Mappings:

- 1. {HOST MACHINE FOLDER}/csvuploads:/maads/agentfilesdocker/dist/maadsweb/csvuploads:z
- 2. {HOST MACHINE FOLDER}/pdfreports:/maads/agentfilesdocker/dist/maadsweb/pdfreports:z
- 3. {HOST MACHINE FOLDER}/autofeatures:/maads/agentfilesdocker/dist/maadsweb/autofeatures:z
- 4. {HOST MACHINE FOLDER}/outliers:/maads/agentfilesdocker/dist/maadsweb/outliers:z
- 5. {HOST MACHINE FOLDER}/sqlloads:/maads/agentfilesdocker/dist/maadsweb/sqlloads:z
- 6. {HOST MACHINE FOLDER}/networktemp:/maads/agentfilesdocker/dist/maadsweb/networktemp:z
- 7. {HOST MACHINE FOLDER}/networks:/maads/agentfilesdocker/networks:z
- 8. {HOST MACHINE FOLDER}/exception:/maads/agentfilesdocker/dist/maadsweb/exception:z
- 9. {HOST MACHINE FOLDER}/staging:/maads/agentfilesdocker/dist/staging:z

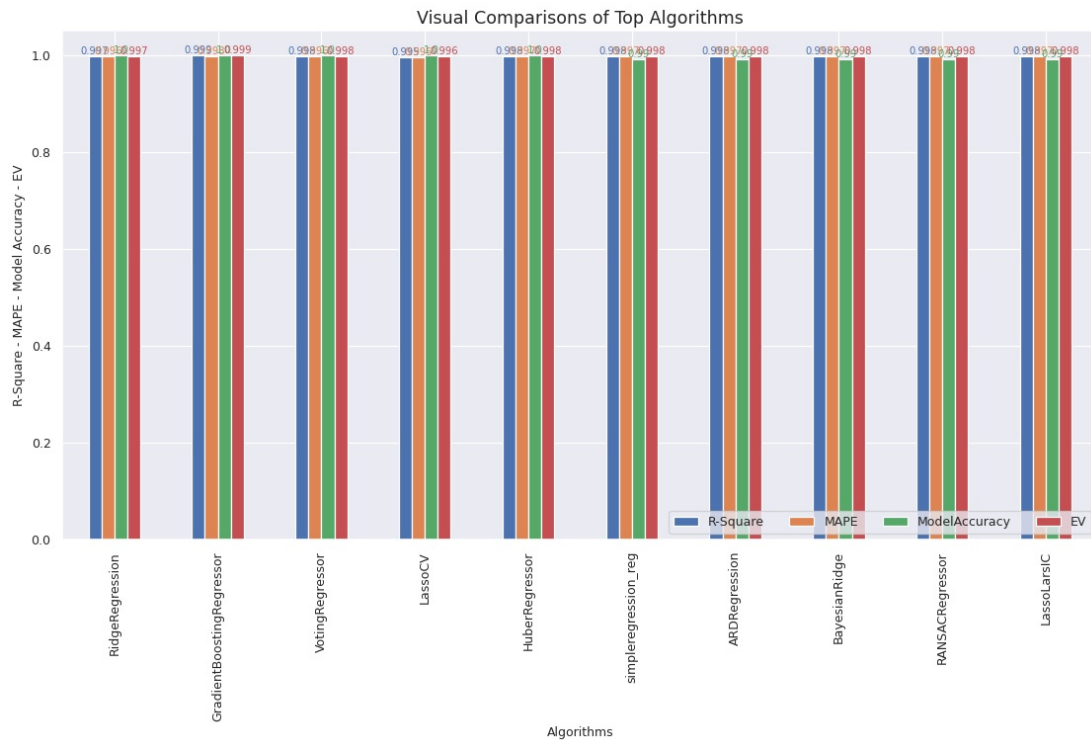
Path for Training Dataset File: `/maads/agentfilesdocker/dist/maadsweb/csvuploads/stockdata.csv`  
Path for PDF Report (i.e. this file): `/maads/agentfilesdocker/dist/maadsweb/pdfreports/admin_stockdata_csv_no_seasons.pdf`  
Path for AutoFeature File: `/maads/agentfilesdocker/dist/maadsweb/autofeatures/admin_stockdata_csv_csv`  
Path for Outliers File: `/maads/agentfilesdocker/dist/maadsweb/outliers/admin_stockdata_csv.csv`  
Path for Algo JSON File: `/maads/agentfilesdocker/dist/maadsweb/exception/admin_stockdata_csv_trained_algo_no_seasons.json`  
Folder Path for MySQL Scripts: `/maads/agentfilesdocker/dist/maadsweb/sqlloads/`  
Path for Detailed Prediction File: `/maads/agentfilesdocker/dist/maadsweb/csvuploads/admin_stockdata_csv_prediction_details.csv`  
Path for Algorithm Zip File (i.e pickle files): `/maads/agentfilesdocker/dist/maadsweb/networktemp/admin_stockdata_csv.zip`  
Path for Algorithm Pickle Files:  

- 1. `/maads/agentfilesdocker/networks/Alberta-Electric-System-Operator_AESO_ADMIN_STOCKDATA_CSVALLSEASON_AG1_4_RidgeRegression_normal_1.00000000_946.pkl`
- 2. `/maads/agentfilesdocker/networks/Alberta-Electric-System-Operator_AESO_ADMIN_STOCKDATA_CSVALLSEASON_AG1_4_RidgeRegression_normal_1.00000000_946_scalerx.pkl`
- 3. `/maads/agentfilesdocker/networks/Alberta-Electric-System-Operator_AESO_ADMIN_STOCKDATA_CSVALLSEASON_AG1_4_RidgeRegression_normal_1.00000000_946_scalery.pkl`

DESCRIPTIVE STATISTICS									
Variables	T-Statistic	Count	Mean	STD	MIN	25%	50%	75%	MAX
Open	1221.412	946.0	94.354	70.58	15.464	80.4	93.29	107.905	125.25
High	1233.139	946.0	95.083	70.75	15.58	80.892	93.925	109.128	125.72
Low	1.521615424968217e+16	946.0	93.697	69.0	15.379	79.803	92.285	107.398	123.68
Volume	-9.308	946.0	2723209.302	686200.0	1440702.602	1984825.0	2433550.0	3089250.0	28284300.0
CLOSE	NA	946.0	94.437	70.28	15.504	80.43	93.295	108.152	125.4

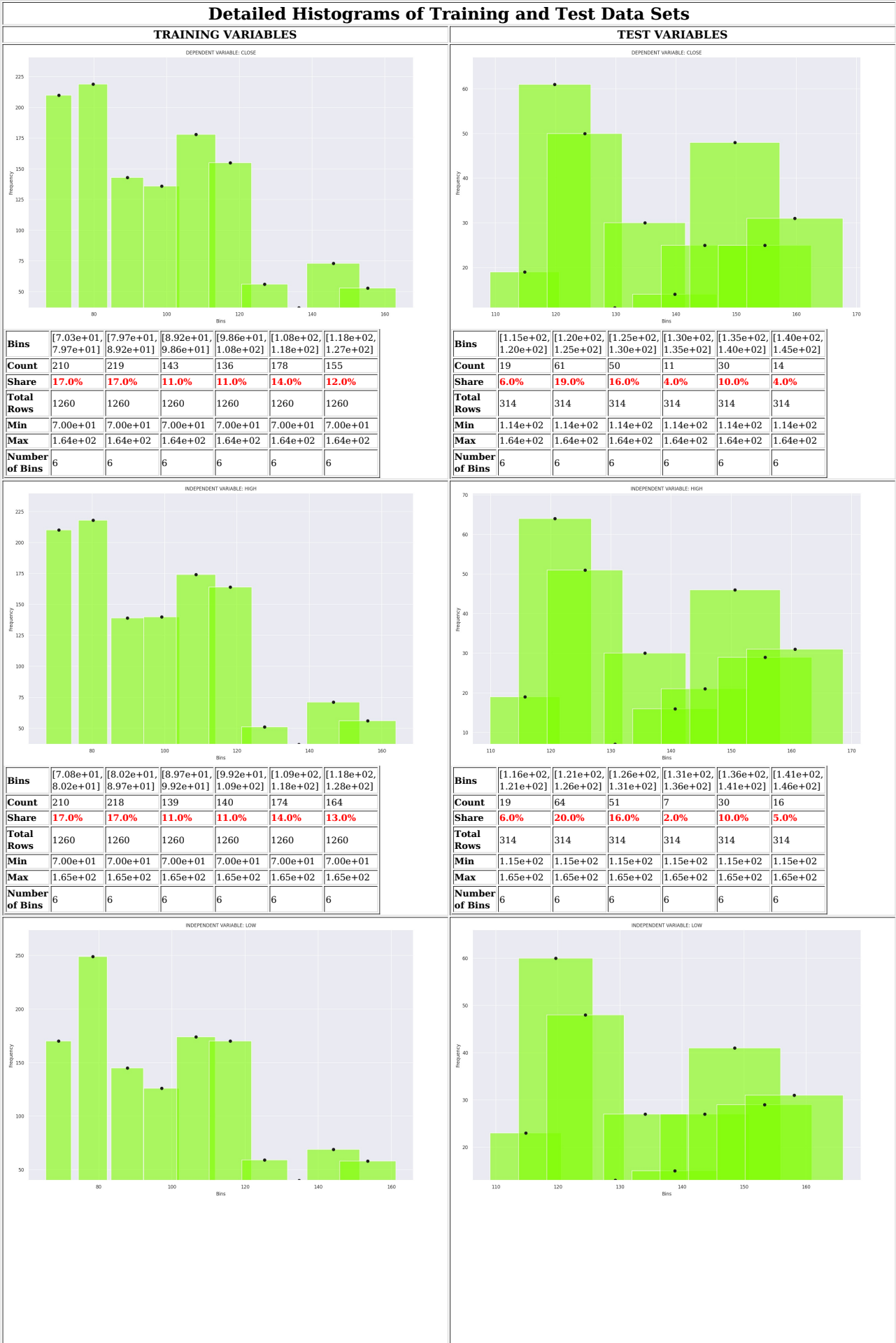
BEST ALGORITHM FOUND FOR THIS DATASET					
(Note: This trained model will be used to predict <b>CLOSE</b> )					
Algorithm	Description	Model Results	MAPE Accuracy	Forecast Months	Season
<a href="#">RidgeRegression</a>	Ridge Regression: Linear least squares with l2 regularization.	<div>Ridge(alpha=0.14549464103297374, random_state=48, solver='saga')</div> <div>Coefficient: [0.0776637186596356, 0.4807798231629039, 0.4366212011905196, 0.008674525867625061]</div> <div><b>R-square:</b> 0.997 <b>Mean Squared Error (MSE):</b> 0.226 <b>Skewness:</b> -0.355 <b>Kurtosis:</b> 3.979 <b>Mean Square Model (MSM):</b> 42573.293 <b>F-Statistic (F):</b> 188641.861 <b>Jarque-Bera (JB):</b> 36.558 <b>Explained Variance (EV):</b> 0.997</div> <div><b>Multicollinearity Test (Avg. VIF):</b> inf <b>Heteroscedasticity Test (Avg P-Value):</b> -1.000 <b>(Based on White Test, there seems to be heteroscedasticity in the model)</b> <b>Autocorrelation (Durbin-Watson) Test:</b> 1.973 <b>(Based on DW Test - there seems to be autocorrelation in your model)</b></div>	0.996	1 - 12	allseason

# TOP 10 ALGORITHMS FOR ALLSEASON

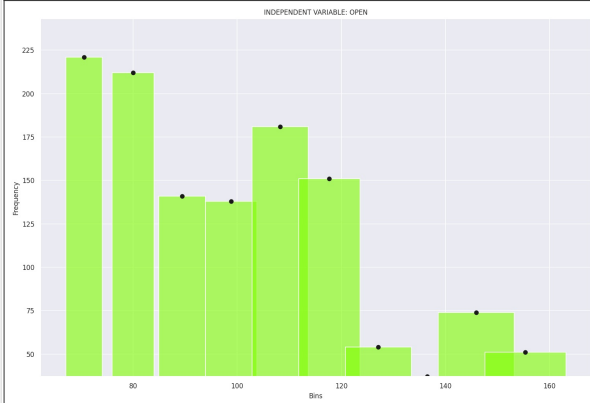


Num	Algorithm	Model Accuracy	Details	Season	Description
1	<a href="#">RidgeRegression</a>	0.9960	<b>R-square:</b> 0.997 <b>MAPE:</b> 0.996 <b>Explained Variance (EV):</b> 0.997 <b>MSE:</b> 0.226 <b>MSM:</b> 42573.293 <b>Skewness:</b> -0.355 <b>Kurtosis:</b> 3.979 <b>F:</b> 188641.861 <b>DW:</b> 1.973 <b>JB:</b> 36.558	allseason	RIDGE REGRESSION: Linear least squares with l2 regularization.
2	<a href="#">GradientBoostingRegressor</a>	0.9957	<b>R-square:</b> 0.999 <b>MAPE:</b> 0.998 <b>Explained Variance (EV):</b> 0.999 <b>MSE:</b> 0.081 <b>MSM:</b> 42498.172 <b>Skewness:</b> -2.203 <b>Kurtosis:</b> 10.673 <b>F:</b> 522015.718 <b>DW:</b> 1.723 <b>JB:</b> 1957.217	allseason	GRADIENT BOOSTING FOR REGRESSION.: GB builds an additive model in a forward stage-wise fashion; it allows for the optimization of arbitrary differentiable loss functions. In each stage a regression tree is fit on the negative gradient of the given loss function.
3	<a href="#">VotingRegressor</a>	0.9956	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.996 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.177 <b>MSM:</b> 42201.548 <b>Skewness:</b> -0.028 <b>Kurtosis:</b> 3.649 <b>F:</b> 237926.228 <b>DW:</b> 1.881 <b>JB:</b> 10.62	allseason	VOTINGREGRESSOR:
4	<a href="#">LassoCV</a>	0.9955	<b>R-square:</b> 0.995 <b>MAPE:</b> 0.995 <b>Explained Variance (EV):</b> 0.996 <b>MSE:</b> 0.362 <b>MSM:</b> 40451.365 <b>Skewness:</b> -1.673 <b>Kurtosis:</b> 4.421 <b>F:</b> 111849.886 <b>DW:</b> 1.128 <b>JB:</b> 330.376	allseason	LASSO CROSS VALIDATION: Lasso linear model with iterative fitting along a regularization path
5	<a href="#">HuberRegressor</a>	0.9950	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138 <b>MSM:</b> 42535.281 <b>Skewness:</b> -0.021 <b>Kurtosis:</b> 4.145 <b>F:</b> 308842.817 <b>DW:</b> 2.064 <b>JB:</b> 32.817	allseason	HUBER REGRESSOR: Linear regression model that is robust to outliers.
			<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138		

6	<a href="#">simpleregression_reg</a>	0.9950	<b>MSM:</b> 42521.698 <b>Skewness:</b> 0.092 <b>Kurtosis:</b> 4.302 <b>F:</b> 308608.598 <b>DW:</b> 2.07 <b>JB:</b> 43.237	allseason	LINEAR REGRESSION: Simple multivariate linear regression
7	<a href="#">ARDRegression</a>	0.9949	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138 <b>MSM:</b> 42525.09 <b>Skewness:</b> 0.044 <b>Kurtosis:</b> 4.211 <b>F:</b> 308290.472 <b>DW:</b> 2.068 <b>JB:</b> 36.864	allseason	BAYESIAN ARD: Fit the weights of a regression model, using an ARD prior. The weights of the regression model are assumed to be in Gaussian distributions. Also estimate the parameters lambda (precisions of the distributions of the weights) and alpha (precision of the distribution of the noise). The estimation is done by an iterative procedures (Evidence Maximization)
8	<a href="#">BayesianRidge</a>	0.9949	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138 <b>MSM:</b> 42524.375 <b>Skewness:</b> 0.048 <b>Kurtosis:</b> 4.218 <b>F:</b> 308418.724 <b>DW:</b> 2.069 <b>JB:</b> 37.305	allseason	BAYESIAN RIDGE REGRESSION: Fit a Bayesian ridge model and optimize the regularization parameters lambda (precision of the weights) and alpha (precision of the noise).
9	<a href="#">RANSACRegressor</a>	0.9948	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138 <b>MSM:</b> 42521.698 <b>Skewness:</b> 0.092 <b>Kurtosis:</b> 4.302 <b>F:</b> 308608.598 <b>DW:</b> 2.07 <b>JB:</b> 43.237	allseason	RANSAC (RANDOM SAMPLE CONSENSUS) ALGORITHM.: RANSAC is an iterative algorithm for the robust estimation of parameters from a subset of inliers from the complete data set. More information can be found in the general documentation of linear models.
10	<a href="#">LassoLarsIC</a>	0.9948	<b>R-square:</b> 0.998 <b>MAPE:</b> 0.997 <b>Explained Variance (EV):</b> 0.998 <b>MSE:</b> 0.138 <b>MSM:</b> 42521.698 <b>Skewness:</b> 0.092 <b>Kurtosis:</b> 4.302 <b>F:</b> 308608.598 <b>DW:</b> 2.07 <b>JB:</b> 43.237	allseason	LASSO MODEL FIT WITH LARS USING BIC OR AIC FOR MODEL SELECTION: Lasso model fit with Lars using BIC or AIC for model selection

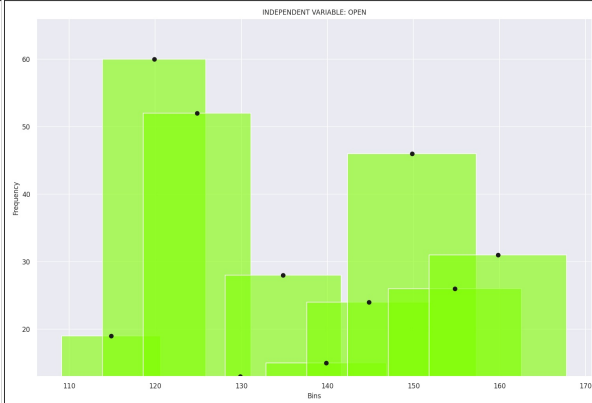


Bins	[6.90e+01, 7.84e+01]	[7.84e+01, 8.78e+01]	[8.78e+01, 9.72e+01]	[9.72e+01, 1.07e+02]	[1.07e+02, 1.16e+02]	[1.16e+02, 1.25e+02]
Count	170	249	145	126	174	170
Share	13.0%	20.0%	12.0%	10.0%	14.0%	13.0%
Total Rows	1260	1260	1260	1260	1260	1260
Min	6.90e+01	6.90e+01	6.90e+01	6.90e+01	6.90e+01	6.90e+01
Max	1.62e+02	1.62e+02	1.62e+02	1.62e+02	1.62e+02	1.62e+02
Number of Bins	6	6	6	6	6	6

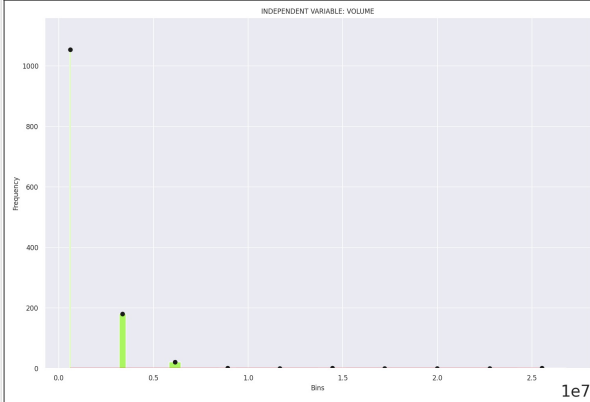


Bins	[7.06e+01, 8.00e+01]	[8.00e+01, 8.94e+01]	[8.94e+01, 9.88e+01]	[9.88e+01, 1.08e+02]	[1.08e+02, 1.18e+02]	[1.18e+02, 1.27e+02]
Count	221	212	141	138	181	151
Share	18.0%	17.0%	11.0%	11.0%	14.0%	12.0%
Total Rows	1260	1260	1260	1260	1260	1260
Min	7.00e+01	7.00e+01	7.00e+01	7.00e+01	7.00e+01	7.00e+01
Max	1.64e+02	1.64e+02	1.64e+02	1.64e+02	1.64e+02	1.64e+02
Number of Bins	6	6	6	6	6	6

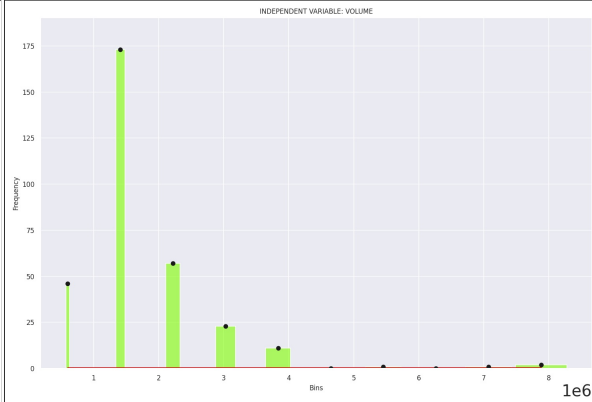
Bins	[1.15e+02, 1.20e+02]	[1.20e+02, 1.24e+02]	[1.24e+02, 1.29e+02]	[1.29e+02, 1.34e+02]	[1.34e+02, 1.39e+02]	[1.39e+02, 1.44e+02]
Count	23	60	48	13	27	15
Share	7.0%	19.0%	15.0%	4.0%	9.0%	5.0%
Total Rows	314	314	314	314	314	314
Min	1.14e+02	1.14e+02	1.14e+02	1.14e+02	1.14e+02	1.14e+02
Max	1.62e+02	1.62e+02	1.62e+02	1.62e+02	1.62e+02	1.62e+02
Number of Bins	6	6	6	6	6	6



Bins	[1.15e+02, 1.20e+02]	[1.20e+02, 1.25e+02]	[1.25e+02, 1.30e+02]	[1.30e+02, 1.35e+02]	[1.35e+02, 1.40e+02]	[1.40e+02, 1.45e+02]
Count	19	60	52	13	28	15
Share	6.0%	19.0%	17.0%	4.0%	9.0%	5.0%
Total Rows	314	314	314	314	314	314
Min	1.14e+02	1.14e+02	1.14e+02	1.14e+02	1.14e+02	1.14e+02
Max	1.64e+02	1.64e+02	1.64e+02	1.64e+02	1.64e+02	1.64e+02
Number of Bins	6	6	6	6	6	6

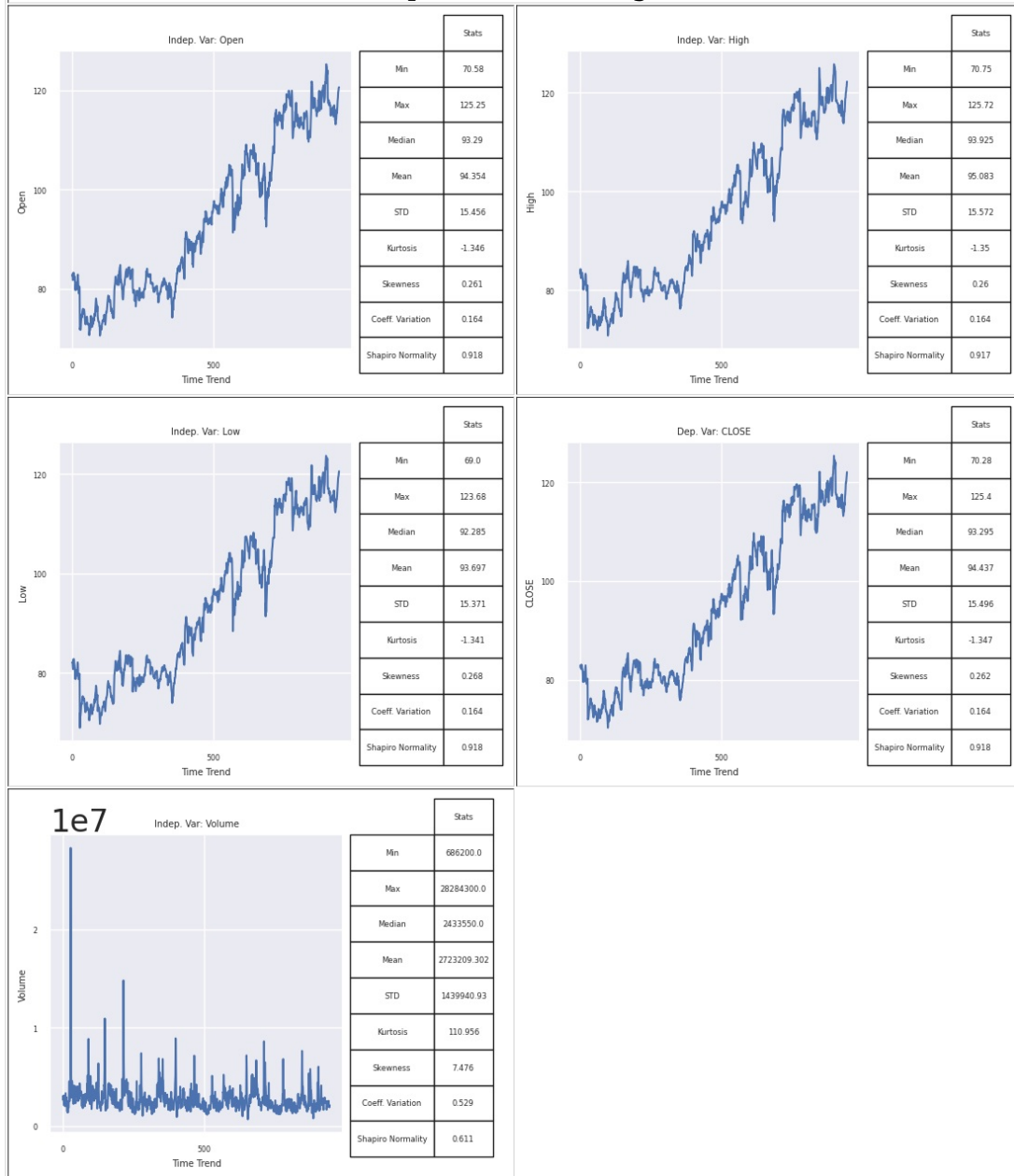


Bins	[6.02e+05, 3.37e+06]	[3.37e+06, 6.14e+06]	[6.14e+06, 8.91e+06]	[8.91e+06, 1.17e+07]	[1.17e+07, 1.44e+07]	[1.44e+07, 1.72e+07]
Count	1054	181	21	2	0	1
Share	84.0%	14.0%	2.0%	0.0%	0.0%	0.0%
Total Rows	1260	1260	1260	1260	1260	1260
Min	6.02e+05	6.02e+05	6.02e+05	6.02e+05	6.02e+05	6.02e+05
Max	2.83e+07	2.83e+07	2.83e+07	2.83e+07	2.83e+07	2.83e+07
Number of Bins	6	6	6	6	6	6

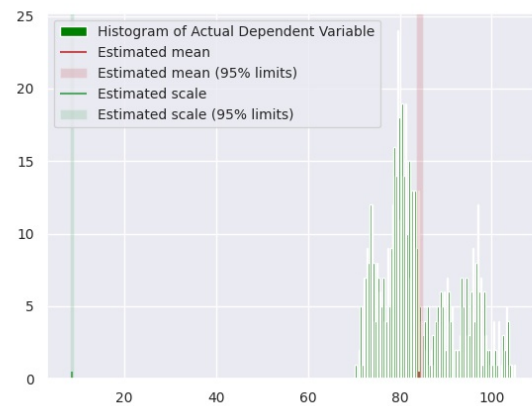
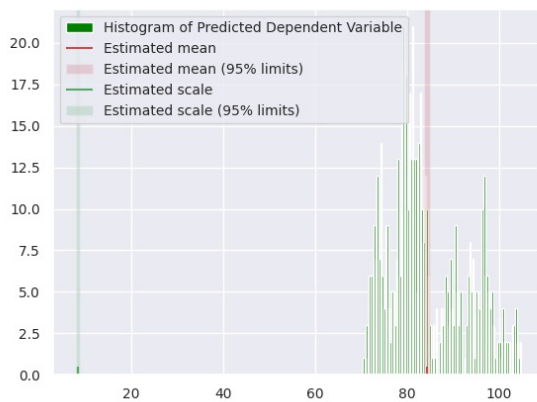


Bins	[6.02e+05, 1.41e+06]	[1.41e+06, 2.22e+06]	[2.22e+06, 3.03e+06]	[3.03e+06, 3.84e+06]	[3.84e+06, 4.65e+06]	[4.65e+06, 5.46e+06]
Count	46	173	57	23	11	0
Share	15.0%	55.0%	18.0%	7.0%	4.0%	0.0%
Total Rows	314	314	314	314	314	314
Min	6.02e+05	6.02e+05	6.02e+05	6.02e+05	6.02e+05	6.02e+05
Max	8.69e+06	8.69e+06	8.69e+06	8.69e+06	8.69e+06	8.69e+06
Number of Bins	6	6	6	6	6	6

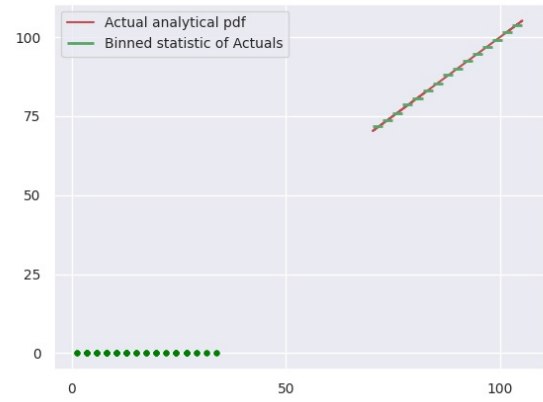
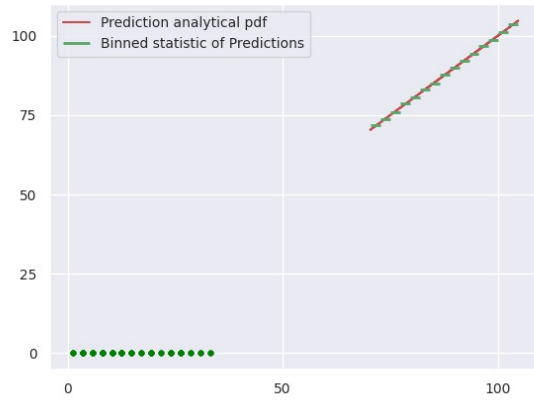
## Detailed Graphs of Variables Against Time



## Bayesian Distribution

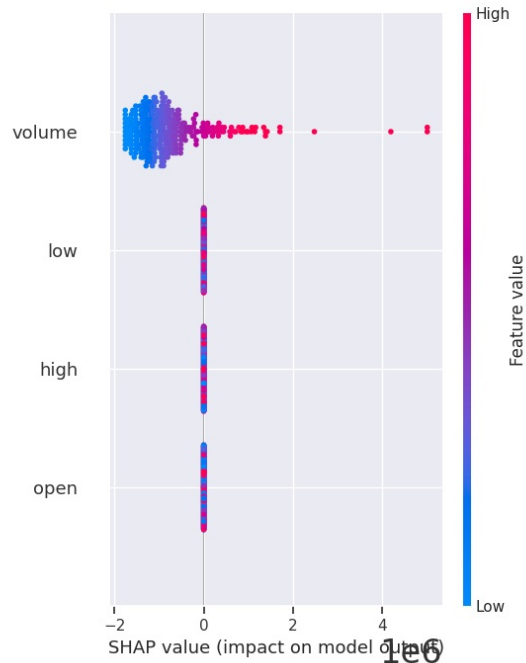


### Binned Statistic





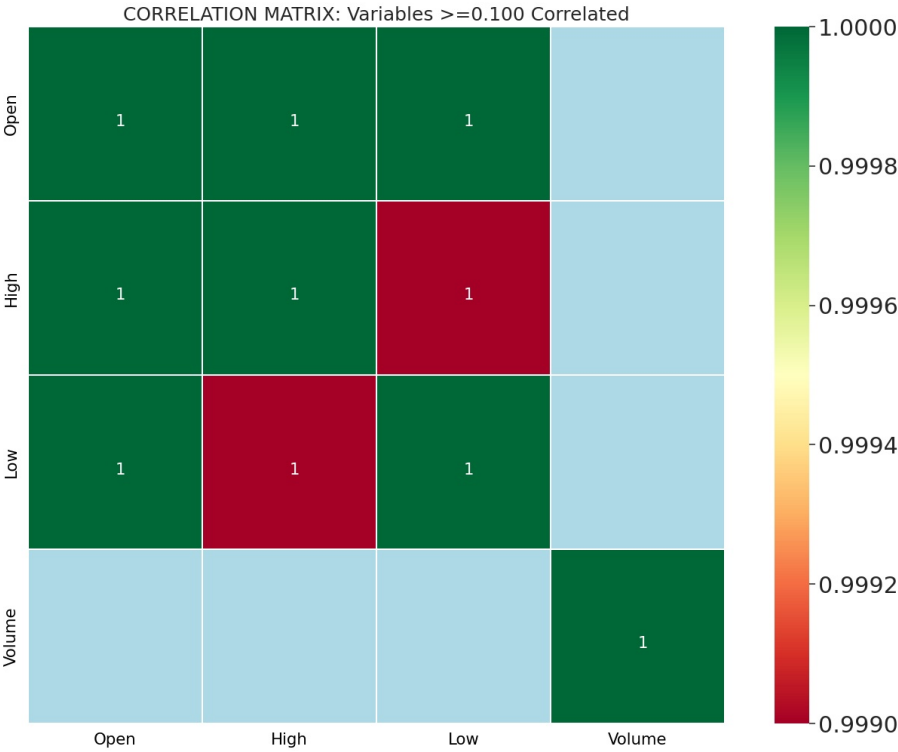
## MODEL EXPLANATION



- The x-axis represents the model's output values of **CLOSE**.
- 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 spectrum.
- 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
Close	0.414
High	0.206
Low	0.192
Open	0.128
Volume	0.059
<b>Best Variable(s) From Genetic Algorithm</b>	
Close	
Volume	
Volume	
<b>Excluded Variable(s)</b>	
High	
Low	
Open	
<b>PCA for Best Variable(s)</b>	Value
Close_pca_1	-0.707
Close_pca_2	0.707
Close_pca_3	-0.708
High_pca_1	-0.673
High_pca_2	0.218
High_pca_3	0.706
Volume_pca_1	0.707
Volume_pca_2	0.707
Volume_pca_3	-0.006
<b>PCA Explained Variance</b>	Value
PCA1	0.627
PCA2	0.373
PCA3	0.000
<ul style="list-style-type: none"> <li>• Feature selection shows which variables were more influential than other variables</li> <li>• It uses two core algorithms: Recursive Feature Elimination (RFE) and Genetic Algorithm to determine influence</li> <li>• It also performs PCA (principal component analysis) analysis to determine the influence of the best variables in the model</li> <li>• These results should be used in conjunction with other information as well as theory to establish relevance and confidence in the chosen model formulation</li> </ul>	

CORRELATION MATRIX



CORRELATED FEATURES			
		<b>Feature(s)</b>	<b>Feature(s)</b>
		<b>Correlation &gt;= 0.100</b>	
	<b>0</b>	High	Low
			0.999
	<b>1</b>	Open	Open
			NaN

SUGGESTED CORRELATED FEATURES TO DELETE		
	1 Feature(s) to Delete	Correlation
	0 High	0.999

END OF REPORT

MAADSBML Python Library: <https://pypi.org/project/maadsbml/>  
MAADSBML Docker Container For Windows: <https://hub.docker.com/r/maadsdocker/maads-batch-automl-otics>  
MAADSBML Docker Container For MAC: <https://hub.docker.com/r/maadsdocker/maads-batch-automl-otics-arm64>  
MAADSBML Sample Code and Setup: <https://github.com/smaurice101/raspberrypi/tree/main/maadsbml>

MAADSBML  
Developed and Maintained by: Otics Advanced Analytics, Inc.  
Toronto, Ontario, Canada  
<https://www.otics.ca>  
Email: support@otics.ca