	Code Output: Using Machine Learning to predict Financial Crises - An Evaluation of different Learning Algorithms for Early Warning Models. Author: Chris Reimann.
In [1]:	Date: 28.02.2023. # Import Experiment Code from prepareData import Data from doExperiment import Experiment # Define Indicators and Models
	<pre>iv_macro = ["rconsbarro", "iy", "money", "xrusd", "gdp", "cpi", "ca"] iv_credit = ["tloans", "debtServ", "yieldCurve", "ltd", "debtgdp", "globaltloans", "globalyieldCurve"] iv_ca = iv_credit + ["hpnom"] iv_all = iv_macro + iv_ca models = ["Logit", "KNeighbors", "RandomForest", "ExtraTrees", "SVM", "NeuralNet"]</pre>
	# Construct Datasets df_macro = Data(iv_macro).getReady("Macro") df_credit = Data(iv_credit).getReady("Credit") df_ca = Data(iv_ca).getReady("Credit & Asset") df_all = Data(iv_all).getReady("All") Macro: The final dataset contains 1591 observations with 63 distinct crisis events. Credit: The final dataset contains 1373 observations with 60 distinct crisis events.
	Credit & Asset: The final dataset contains 1159 observations with 46 distinct crisis events. All: The final dataset contains 1101 observations with 41 distinct crisis events. # Define Experiments ex_macroIS = Experiment(df_macro, models, "InSample") ex_caIS = Experiment(df_ca, models, "InSample") ex_allIS = Experiment(df_all, models, "InSample") ex_allIS = Experiment(df_macro, models, "InSample") ex_credit = Experiment(df_credit, models, "CrossVal") ex_credit = Experiment(df_ca, models, "CrossVal") ex_ca = Experiment(df_ca, models, "CrossVal") ex_all = Experiment(df_all, models, "CrossVal")
In [5]:	Run Experiments: In-Sample # Run Experiments: Macro Indicators InSample ex_macroIs.run() ex_macroIs.auc
	Random Assignment: 0% 0/1 [00:00 , ?it/s] Logit: 0% 0/1 [00:00<?, ?it/s] KNeighbors: 0% 0/1 [00:00<?, ?it/s] RandomForest: 0% 0/1 [00:00<?, ?it/s] ExtraTrees: 0% 0/1 [00:00<?, ?it/s] SVM: 0% 0/1 [00:00<?, ?it/s] NeuralNet: 0% 0/1 [00:00<?, ?it/s]</th
Out[5]:	Model AUC 0 KNeighbors 1.000000 1 RandomForest 0.981277 2 ExtraTrees 0.833469 3 SVM 0.759257 4 NeuralNet 0.713687 5 Logit 0.707710
In [6]:	# Run Experiments: Credit Indicators InSample ex_creditIS.run() ex_creditIS.auc Random Assignment: 0% 0/1 [00:00 , ?it/s] Logit: 0% 0/1 [00:00<?, ?it/s]</th
Out[6]:	<pre>KNeighbors: 0% </pre>
	3 SVM 0.918643 4 NeuralNet 0.851912 5 Logit 0.790217 6 Random Assignment 0.500000
	# Run Experiments: Credit & Asset Indicators InSample ex_caIS.run() ex_caIS.auc Random Assignment: 0%
Out[7]:	SVM: 0% 0/1 [00:00 , ?it/s] NeuralNet: 0% 0/1 [00:00<?, ?it/s] Model AUC Neighbors 1.000000 RandomFores 0.992726 ExtraTrees 0.955384 NeuralNet 0.864574 Logit 0.808380</th
In [8]:	# Run Experiments: All Indicators InSample + Logit Coefficients ex_allIs.run() display(ex_allIs.auc) ex_allIs.logitCoef()
	Random Assignment: 0%
	New Explore 1.000000 RandomForest 0.994486 New Explore 0.963970 New Explore 0.951174 New Explore 0.932865 Logit 0.858774 Random Assignment 0.500000 Optimization terminated successfully.
Out[8]:	Current function value: 0.228989 Iterations 8 Coef. Std.Err. z P> z [0.025 0.975] rconsbarro -9.339301 2.363592 -3.951316 7.772255e-05 -13.971857 -4.706745 iy 1.889071 4.128478 0.457571 6.472609e-01 -6.202598 9.980740 money -1.333617 1.874857 -0.711316 4.768882e-01 -5.008269 2.341036
	xrusd
	debtServ -10.678297 8.783358 -1.215742 2.240832e-01 -27.893362 6.536769 yieldCurve -0.481616 0.096841 -4.973246 6.584103e-07 -0.671421 -0.291810 ltd 1.717015 0.885701 1.938595 5.255069e-02 -0.018927 3.452957 debtgdp -3.377012 0.811291 -4.162518 3.147576e-05 -4.967112 -1.786911 globaltloans 0.952151 2.056721 0.462946 6.434029e-01 -3.078948 4.983251
	hpnom 0.749860 0.234076 3.203489 1.357731e-03 0.291079 1.208640 Run Experiments: Out-of-Sample
	# Run Experiments: Macro Indicators ex_macro.run(n = 100) ex_macro.auc Random Assignment: 0%
Out[9]:	SVM: 0% 0/100 [00:00 , ?it/s] NeuralNet: 0% 0/100 [00:00<?, ?it/s]</th
	3 ExtraTrees 0.655023 4 KNeighbors 0.606674 5 SVM 0.529537 6 Random Assignment 0.500000
	# Run Experiments: Credit Indicators ex_credit.run(n = 100) ex_credit.auc Random Assignment: 0%
Out[10]:	Description1 RandomForest2 0.8258463 0.822628
	 KNeighbors 0.795530 NeuralNet 0.768945 Logit 0.757818 SVM 0.751454 Random Assignment 0.500000
	# Run Experiments: Credit & Asset Indicators ex_ca.run(n = 100) ex_ca.auc Random Assignment: 0%
Out[11]:	SVM: 0% 0/100 [00:00 , ?it/s]</th NeuralNet: 0% 0/100 [00:00 , ?it/s]</th Model AUC 0 ExtraTrees 0.840335 1 RandomForest 0.832646 2 KNeighbors 0.803151 3 NeuralNet 0.773489 4 Logi 0.761357 5 SVM 0.756435
	# Run Experiments: All Indicators ex_all.run(n = 100) ex_all.rocGraph() ex_all.auc Random Assignment: 0% 0/100 [00:00 , ?it/s]</th
Out[12]:	Logit: 0%
	3 KNeighbors 0.798073 4 Logit 0.779664 5 SVM 0.765645 6 Random Assignment 0.500000
	Ositive Rate Rate (TPR) 0.4 0.6 0.6 0.7
	0.2 Random Assignment: (AUC = 0.5) Logit: (AUC = 0.78) KNeighbors: (AUC = 0.798) RandomForest: (AUC = 0.823) ExtraTrees: (AUC = 0.833) SVM: (AUC = 0.766)
	NeuralNet: (AUC = 0.811) 0.0 0.2 0.4 0.6 0.8 1.0 False Positive Rate (FPR) Robustness Checks: Alternative Crisis Data
	<pre># Robustness Check: ESRB Crisis Data df_alt1 = Data(iv_all, crisisData = "ESRB").getReady("ESRB") ex_alt1 = Experiment(df_alt1, models, "CrossVal") ex_alt1.run(n = 100) ex_alt1.auc</pre> ESRB: The final dataset contains 426 observations with 22 distinct crisis events.
	Random Assignment: 0%
Out[13]:	Model AUC 0 RandomForest 0.896491 1 ExtraTrees 0.868925 2 KNeighbors 0.804878 3 Logit 0.759008 4 SVM 0.757480
In [14]:	<pre>5 NeuralNet 0.740183 6 Random Assignment 0.500000 # Robustness Check: Laeven & Valencia Crisis Data df_alt1 = Data(iv_all, crisisData = "LaevenValencia").getReady("LaevenValencia") ex_alt1 = Experiment(df_alt1, models, "CrossVal")</pre>
	ex_alt1.run(n = 100) ex_alt1.auc Yugoslavia, SFR not found in regex LaevenValencia: The final dataset contains 674 observations with 18 distinct crisis events. Random Assignment: 0%
Out[14]:	RandomForest: 0% 0/100 [00:00 , ?it/s] ExtraTrees: 0% 0/100 [00:00<?, ?it/s] SVM: 0% 0/100 [00:00<?, ?it/s] NeuralNet: 0% 0/100 [00:00<?, ?it/s] Model AUC 0 NeuralNet 0.886967 1 ExtraTrees 0.867728 2 KNeighbors 0.863598 3 RandomForest 0.849177 4 SVM 0.847194</td
In [15]:	<pre>5 Logit 0.836380 6 Random Assignment 0.500000 # Robustness Check: Strict Forecasting ex_macroF = Experiment(df_macro, models, "Forecast") ex_creditF = Experiment(df_credit, models, "Forecast") ex_caF = Experiment(df_ca, models, "Forecast")</pre>
In [16]:	ex_allF = Experiment(df_all, models, "Forecast") Robustness Checks: Strict Forecasting # Forecast: Macro ex_macroF.run() ex_macroF.auc
	Random Assignment: 0% 0/1 [00:00 , ?it/s] Logit: 0% 0/1 [00:00<?, ?it/s] KNeighbors: 0% 0/1 [00:00<?, ?it/s] RandomForest: 0% 0/1 [00:00<?, ?it/s] ExtraTrees: 0% 0/1 [00:00<?, ?it/s] SVM: 0% 0/1 [00:00<?, ?it/s] NeuralNet: 0% 0/1 [00:00<?, ?it/s]</td
Out[16]:	Model AUC 0 Logit 0.635322 1 NeuralNet 0.626200 2 KNeighbors 0.600879 3 RandomForest 0.534933 4 ExtraTrees 0.520512 5 Random Assignment 0.500000 6 SVM 0.427193
	ex_creditF.run() ex_creditF.auc Random Assignment: 0%
	RandomForest: 0%
	1 KNeighbors 0.670350 2 SVM 0.647153 3 RandomForest 0.642819 4 ExtraTrees 0.619348 5 Logit 0.525632 6 Random Assignment 0.500000
	# Forecast: Credit & Asset ex_caF.run() ex_caF.auc Random Assignment: 0%
Out[18]:	RandomForest: 0%
	 KNeighbors 0.705793 ExtraTrees 0.673546 NeuralNet 0.660708 SVM 0.642525 Logit 0.583442 Random Assignment 0.500000
	ex_allF.run() ex_allF.auc Random Assignment: 0%
Out[19]:	ExtraTrees: 0%
	 KNeighbors 0.651607 RandomForest 0.647569 ExtraTrees 0.594835 SVM 0.585685 Random Assignment 0.500000
In []: In []:	