

# Classification workflows conversational agent example runs

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## Introduction

This notebook gives basic demonstrations of the functionalities developed in the project “Classification workflows conversational agent”.

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## Load preliminary code

Load the packages for the ClCon monad, monad tracing, functional parsers, and data obtaining:

```
In[1]:= Import["https://raw.githubusercontent.com/antononcube/MathematicaForPrediction/master/MonadicProgramming/MonadicContextualClassification.m"]
Import["https://raw.githubusercontent.com/antononcube/MathematicaForPrediction/master/MonadicProgramming/MonadicTracing.m"]
Import["https://raw.githubusercontent.com/antononcube/MathematicaForPrediction/master/FunctionalParsers.m"]
Import["https://raw.githubusercontent.com/antononcube/MathematicaVsR/master/Projects/ProgressiveMachineLearning/Mathematica/GetMachineLearningDataset.m"]

» Importing from GitHub: MathematicaForPredictionUtilities.m
» Importing from GitHub: MosaicPlot.m
» Importing from GitHub: CrossTabulate.m
» Importing from GitHub: StateMonadCodeGenerator.m
» Importing from GitHub: ClassifierEnsembles.m
» Importing from GitHub: ROCFunctions.m
» Importing from GitHub: VariableImportanceByClassifiers.m
» Importing from GitHub: SSparseMatrix.m
» Importing from GitHub: OutlierIdentifiers.m
```

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## Load project code

Get and run the parsers specification and generation code:

```
In[14]:= Clear["ebnf*"]
Get["https://raw.githubusercontent.com/antononcube/ConversationalAgents/master/EBNF/ClassifierWorkflowsGrammar.ebnf"]
Names["ebnf*"]
LeafCount /@ res

Out[16]= {ebnfClassifierEnsembleMaking, ebnfClassifierMaking, ebnfClassifierQuery, ebnfClassifierTesting, ebnfCommand,
ebnfDataLoad, ebnfDataOutliers, ebnfDataStatistics, ebnfDataTransform, ebnfPipelineCommands, ebnfSplitting, ebnfVerification}

Out[17]= {628, 763, 897, 466, 424, 960, 531, 905, 4465, 491, 843, 24}

Load the translator package:

In[18]:= Import["https://raw.githubusercontent.com/antononcube/ConversationalAgents/master/Projects/ClassificationWorkflowsAgent/Mathematica/ClConTranslator.m"]
```

## Get data

Get Titanic data (using WL’s repository):

```
In[19]:= dsTitanic = GetMachineLearningDataset["Titanic"];
```

## Generate pipeline

Generate a ClCon pipeline from a sequence of natural language commands:

```
In[20]:= clCommands = {
  "load the titanic data",
  "split the data with 65 percent for training", "summarize data", "train a random forest classifier",
  "show classifier information",
  "display classifier training time", "show accuracy, precision, recall, and area under roc curve", "display confusion matrix plot",
  "compute the variable importance estimates";
pl = ToClConPipelineFunction[clCommands]

Out[21]= Function[{x, c}, (((((((ClConUnit[x, c] ==> ClConSplitData[0.65]) ==> ClConEchoFunctionValue[summaries:, (Multicolumn[#1, 5] &) /@RecordsSummary /@#1 &]) ==> ClConMakeClassifier[RandomForest]) ==>
  ClConEchoFunctionContext[classifier info:, If[AssociationQ[#1[classifier]], ClassifierInformation /@#1[classifier], ClassifierInformation[#1[classifier]]] &]) ==>
  ClConEchoFunctionContext[classifier property "TrainingTime" :, If[AssociationQ[#1[classifier]], (ClassifierInformation[#1, TrainingTime] &) /@#1[classifier],
  ClassifierInformation[#1[classifier], TrainingTime]] &]) ==>
  Function[{x$, c$}, ClConUnit[x$, c$] ==> ClConClassifierMeasurements[{Accuracy, Precision, Recall, AreaUnderROCCurve}] ==> ClConEchoValue]) ==>
  Function[{x$, c$}, ClConUnit[x$, c$] ==> ClConClassifierMeasurements[{ConfusionMatrixPlot}] ==> ClConEchoValue]) ==>
  Function[{x, c}, ClConUnit[x, c] ==> ClConAccuracyByVariableShuffling[] ==> ClConEchoValue]]
```

## Run the generated pipeline

Run the generated pipeline over the Titanic data:

```
In[22]:= ClConUnit[dsTitanic] ==> pl;
```

» summaries: <|trainingData →

1 id	2 passengerClass	3 passengerAge
Min 1		Min -1
1st Qu 345		1st Qu 10
Mean 667.188	3rd 470	Median 20
Median 675.5	1st 203	Mean 23.9965
3rd Qu 1000	2nd 177	3rd Qu 40
Max 1307		Max 80

, testData →

1 id	2 passengerClass	3 passengerAge	4 passengerSex	5 passengerSurvival
Min 2		Min -1		
1st Qu 308		1st Qu 0	male 545	died 525
Mean 622	3rd 239	Median 20	female 305	survived 325
Mean 632.429	1st 120	Mean 22.7233		
3rd Qu 956.75	2nd 100	3rd Qu 37.5		
Max 1309		Max 80		

>

» classifier info:

Classifier information

Input type	Mixed (number: 4)
Classes	died, survived
Method	RandomForest
Accuracy	78.2% ± 1.2%
Loss	0.494 ± 0.011
Single evaluation time	4.1 ms/example
Batch evaluation speed	37.1 examples/ms
Classifier memory	257. kB
Training examples used	850 examples
Training time	2.13 s

<

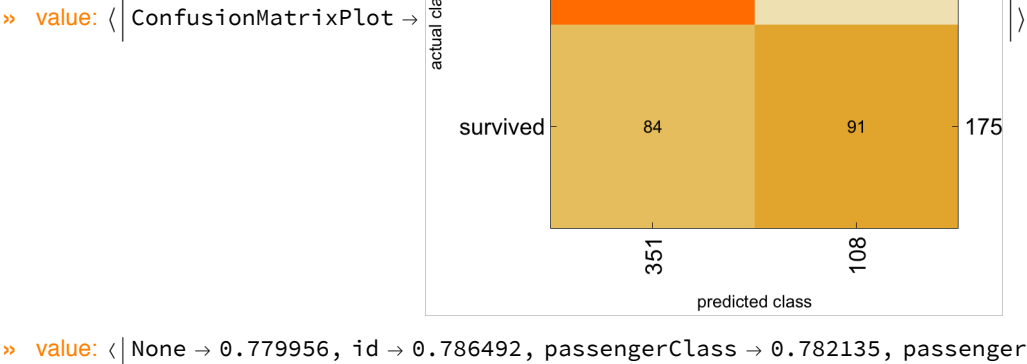
Learning curve

>

training examples used	Accuracy
10	0.67
50	0.55
100	0.52
500	0.49

» classifier property "TrainingTime": 2.13723 s

» value: <|Accuracy → 0.779956, Precision → <|died → 0.760684, survived → 0.842593|>, Recall → <|died → 0.940141, survived → 0.52|>, AreaUnderROCCurve → <|died → 0.86994, survived → 0.824004|>|>



# Trace run and command-code table

Run the generated pipeline through TraceMonad in order to obtained tabulated correspondence between (1) the generated ClCon pipeline components and (2) the natural language commands used to generate them:

```
In[23]:= (p =
  ClConUnit[dsTitanic] =>
  ToClConPipelineFunction[clCommands, "Trace" -> True]) =>
  TraceMonadTakeGrid[]
```

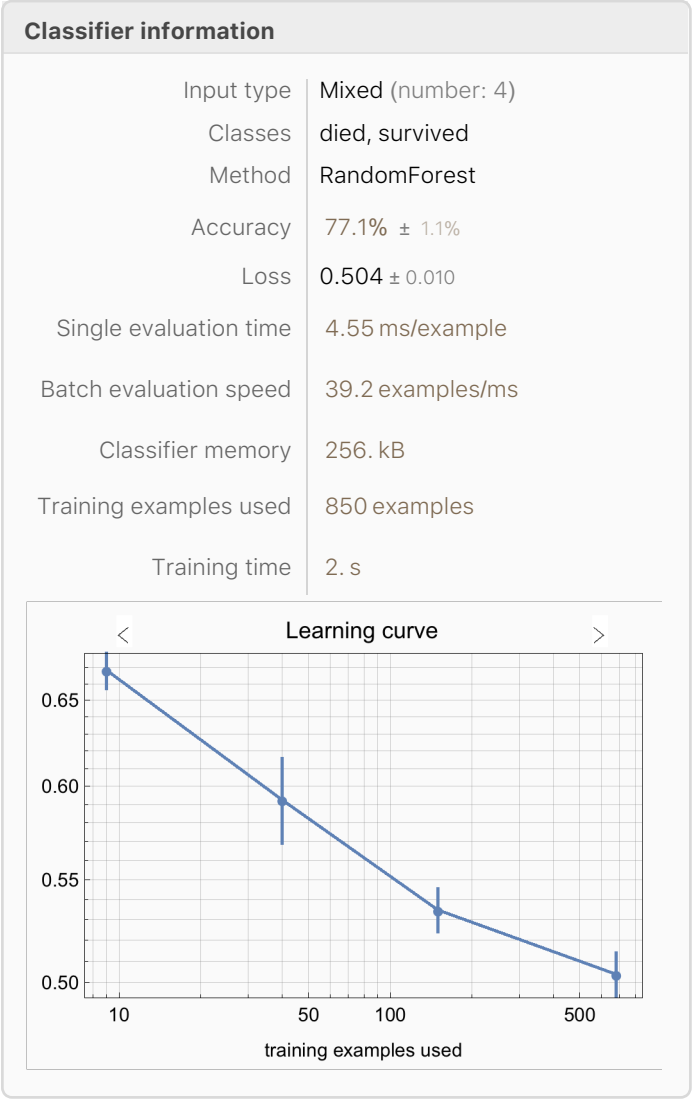
» summaries: <| trainingData ->

1 id	2 passengerClass	3 passengerAge	4 passengerSex	5 passengerSurvival
Min 3	3rd 468	Min -1	male 556	died 525
1st Qu 348	1st 193	1st Qu 10	female 294	survived 325
Median 656.5	2nd 189	Median 20		
Mean 662.324		Mean 23.1576		
3rd Qu 975		3rd Qu 40		
Max 1309		Max 70		

, testData ->

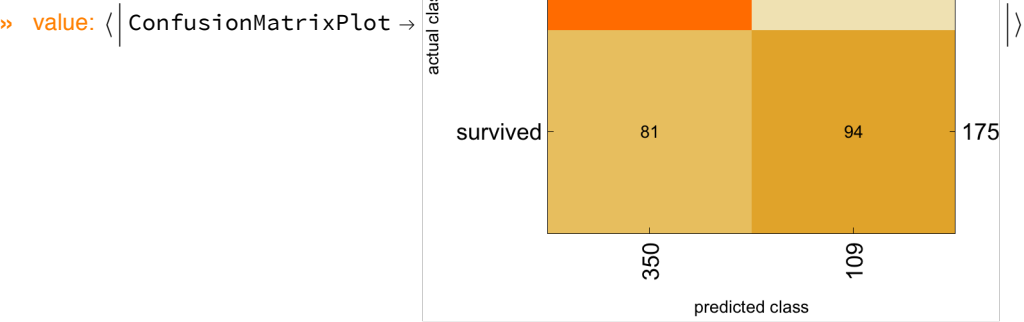
1 id	2 passengerClass	3 passengerAge	4 passengerSex	5 passengerSurvival
Min 1	3rd 241	Min -1	male 287	died 284
1st Qu 276.5	1st 130	1st Qu 10	female 172	survived 175
Median 641.438	2nd 88	Median 20		
Mean 643		Mean 24.2767		
3rd Qu 996.5		3rd Qu 40		
Max 1307		Max 80		

>



» classifier property "TrainingTime" : 2.00355 s

» value: <| Accuracy → 0.79085, Precision → <| died → 0.768571, survived → 0.862385|>, Recall → <| died → 0.947183, survived → 0.537143|>, AreaUnderROCCurve → <| died → 0.866519, survived → 0.820463|>|>



» value: <|None → 0.79085, id → 0.784314, passengerClass → 0.8061, passengerAge → 0.769063, passengerSex → 0.633987|>

Out[23]=

ClConUnit[x, c] ⇒	
ClConSplitData[0.65`] ⇒	split the data with 65 percent for training
ClConEchoFunctionValue["summaries:", (Multicolumn[#1, 5] &) /@RecordsSummary /@#1 &] ⇒	summarize data
ClConMakeClassifier["RandomForest"] ⇒	train a random forest classifier
ClConEchoFunctionContext["classifier info:", If[AssociationQ[#1["classifier"]], ClassifierInformation /@#1["classifier"], ClassifierInformation[#1["classifier"]]] &] ⇒	show classifier information
ClConEchoFunctionContext["classifier property \"TrainingTime\" :", If[AssociationQ[#1["classifier"]], (ClassifierInformation[#1, \"TrainingTime\"] &) /@#1["classifier"], ClassifierInformation[#1["classifier"], \"TrainingTime\"] &] ⇒	display classifier training time
Function[{x\$, c\$}, ClConUnit[x\$, c\$] ⇒ ClConClassifierMeasurements[{"Accuracy", "Precision", "Recall", "AreaUnderROCCurve"}] ⇒ ClConEchoValue] ⇒	show accuracy, precision, recall, and area under roc curve
Function[{x\$, c\$}, ClConUnit[x\$, c\$] ⇒ ClConClassifierMeasurements[{"ConfusionMatrixPlot"}] ⇒ ClConEchoValue] ⇒	display confusion matrix plot
Function[{x, c}, ClConUnit[x, c] ⇒ ClConAccuracyByVariableShuffling[] ⇒ ClConEchoValue]	compute the variable importance estimates