Classification workflows conversational agent example runs

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MathematicaForPrediction at GitHub
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Introduction

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

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`

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/ClassficationWorkflowsAgent/Mathematica/ClConTranslator.m"]
```

Get data

Get the Titanic data (from WL's repository):

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

Generate a classification 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		4 passengerSex		5 passengerSurvival	
Min	1	3rd	470	Min	−1	male	545	died	525
1st Qu	345	1st	203	1st Qu	10	female	305	survived	325
Mean	667.188	2nd	177	Median	20				
Median	675.5			Mean	23.9965				
3rd Qu	1000			3rd Qu	40				
Max	1307			Max	80				

, testData →

1 id		2 passengerClass		3 passengerAge		4 passengerSex		5 passengerSurvival	
Min	2	3rd	239	Min	−1	male	298	died	284
1st Qu	308	1st	120	1st Qu	0	female	161	survived	175
Median	622	2nd	100	Median	20				
Mean	632.429			Mean	22.7233				
3rd Qu	956.75			3rd Qu	37.5				
Max	1309			Max	80				

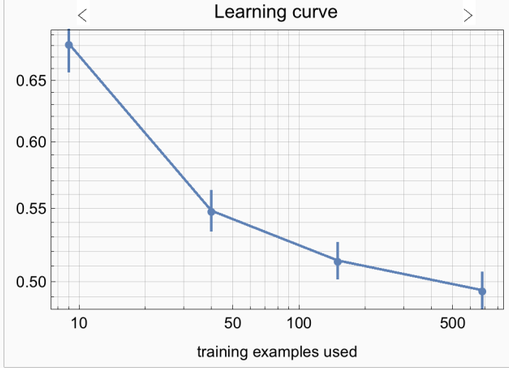
|⟩

» 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

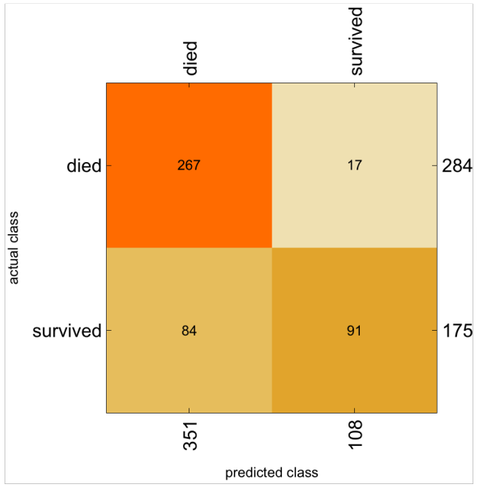
Learning curve



» 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|⟩|⟩

» value: ⟨| ConfusionMatrixPlot →



|⟩

» value: ⟨| None → 0.779956, id → 0.786492, passengerClass → 0.782135, passengerAge → 0.764706, passengerSex → 0.605664 |⟩

Trace run and ode-command table

Run the generated pipeline through TraceMonad in order to obtain 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 →

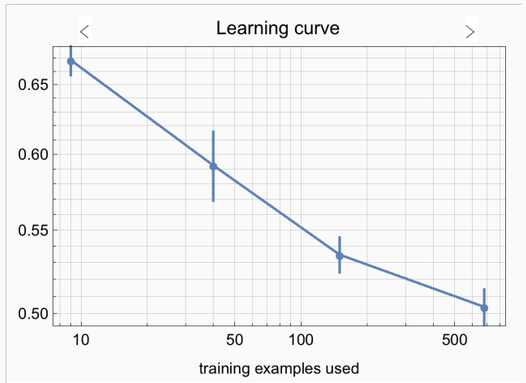
1 id		2 passengerClass	3 passengerAge		4 passengerSex		5 passengerSurvival	
Min	1	3rd 241	Min	-1	male	287	died	284
1st Qu	276.5	1st 130	1st Qu	10	female	172	survived	175
Mean	641.438	2nd 88	Median	20				
Median	643		Mean	24.2767				
3rd Qu	996.5		3rd Qu	40				
Max	1307		Max	80				

|⟩

» classifier info:

Classifier information

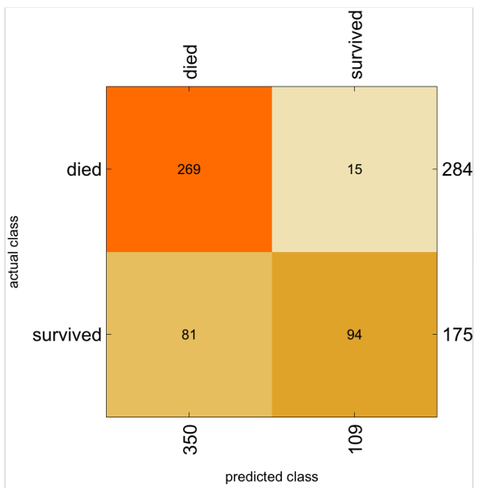
Input type	Mixed (number: 4)
Classes	died, survived
Method	RandomForest
Accuracy	77.1% ± 1.1%
Loss	0.504 ± 0.010
Single evaluation time	4.55 ms/example
Batch evaluation speed	39.2 examples/ms
Classifier memory	256. kB
Training examples used	850 examples
Training time	2. s



» 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: ⟨|ConfusionMatrixPlot →



|⟩

» 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