# Use Classifiers to decide if a statement is a sentiment or other or an advice

■ What if we let the classifier decide if a statement has value or not from a sample of statements.

```
In[875]:= << dataReady.mx
In[876]:= Length[dataReady]
Out[876]= 1616
In[937]:= Export["dataReadyOut.csv", dataReady, "CSV"]
Out[937]= dataReadyOut.csv</pre>
```

# Prepare Dataset for Classification

I've extracted ingredients from all the reviews and I saved them in Enchiladalng.

Then, I decided to create a classifier. hence, I labeled sentences having ingredients 1, and others 0. My reasoning, if the reviewers talked about an ingredients, they most likely did something with it. You will notice that the frequencies between the non-informative sentences and the informative sentences are close to each other.

```
In[878]:= dataForClustering =
        Map[If[Length[Intersection[#, EnchiladaIng]] > 0, Rule[#, 1], Rule[#, 0]] &,
         dataReady];
      Counts[dataForClustering[[All, 2]]]
      \langle |0 \rightarrow 794, 1 \rightarrow 822 | \rangle
    Now, I divided my dataset into training set, testing set, and validation set
      trainingData = dataForClustering[[1;; 800]];
      testingData = dataForClustering[[801;; 1200]];
      validationData = dataForClustering[[1201;; 1616]];
    Save the dataset
     DumpSave["dataForClustering.mx", dataForClustering]
      Export["DataForClustering.xls", dataForClustering, "XLS"]
```

### Classifier

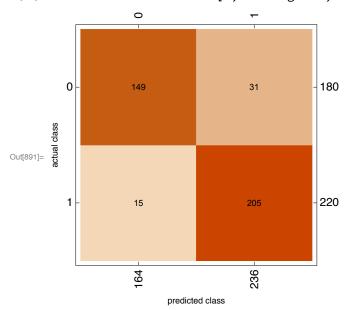
Mathematica decided on the best model. In here the best model is the Hidden Markov model. In R, you can use the package caret (which also has the model implemented)

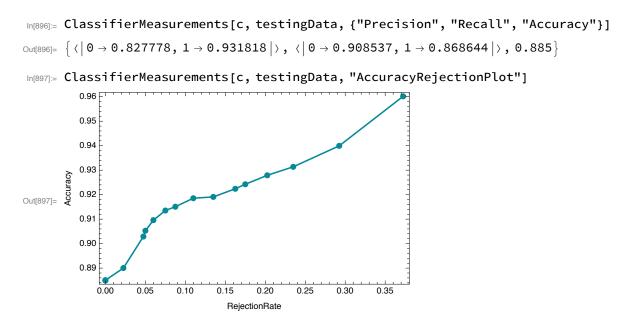
```
In[892]:= c = Classify[trainingData]
```

```
Out[892]= ClassifierFunction
```

# Performance with Testing Dataset

In[891]:= ClassifierMeasurements[c, testingData, "ConfusionMatrixPlot"]





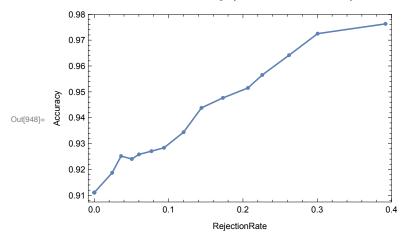
## Performance with Validation Dataset

One can see that the performance was not a flux.

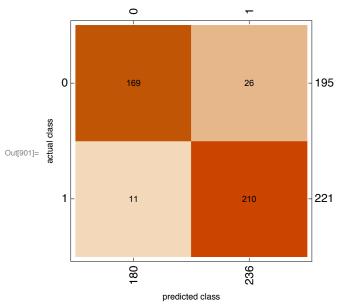
```
In[929]:= ClassifierMeasurements[c, validationData, "Accuracy"]
Out[929]= 0.911058
In[927]:= ClassifierMeasurements[c, validationData, {"Precision", "Recall"}]
Out[927]= \left\{ \langle \mid 0 \to 0.866667, 1 \to 0.950226 \mid \rangle, \langle \mid 0 \to 0.938889, 1 \to 0.889831 \mid \rangle \right\}
In[923]:= Insert[Grid[{{"Performance", "Sentiment", "Not Sentiment"},
            {"Precision", 0.87, 0.95}, {"Recall", 0.94, 0.89}}],
         {Background \rightarrow {None, {GrayLevel[0.7], {White}}}, Dividers \rightarrow {Black, {2 \rightarrow Black}},
          Frame \rightarrow True, Spacings \rightarrow {2, {2, {0.7}, 2}}}, 2]
```

	Performance	Sentiment	Not Sentiment
Out[923]=	Precision	0.87	0.95
	Recall	0.94	0.89

# ClassifierMeasurements[c, validationData, "AccuracyRejectionPlot"]



# In[901]:= ClassifierMeasurements[c, validationData, "ConfusionMatrixPlot"]



DumpSave["c.mx", c]

