1/28/2018 Udacity Reviews



PROJECT

Build a Sign Language Recognizer

A part of the Artificial Intelligence Program

PROJECT REVIEW

CODE REVIEW 4

NOTES

```
▼ my_model_selectors.py
    1 import math
    2 import statistics
    3 import warnings
    5 import numpy as np
    6 from hmmlearn.hmm import GaussianHMM
    7 from sklearn.model_selection import KFold
    8 from asl_utils import combine_sequences
   10
   11 class ModelSelector(object):
   12
          base class for model selection (strategy design pattern)
   13
   14
   15
          def __init__(self, all_word_sequences: dict, all_word_Xlengths: dict, this_word: str,
   16
                       n constant=3,
   17
                       min_n_components=2, max_n_components=10,
   18
                       random_state=14, verbose=False):
   19
              self.words = all_word_sequences
   20
              self.hwords = all_word_Xlengths
   21
   22
              self.sequences = all_word_sequences[this_word]
              self.X, self.lengths = all_word_Xlengths[this_word]
   23
              self.this word = this word
   24
              self.n_constant = n_constant
   25
              self.min_n_components = min_n_components
              self.max_n_components = max_n_components
   27
   28
              self.random_state = random_state
              self.verbose = verbose
          def select(self):
   31
              raise NotImplementedError
   33
          def base_model(self, num_states):
   34
              # with warnings.catch_warnings():
   35
   36
              warnings.filterwarnings("ignore", category=DeprecationWarning)
              # warnings.filterwarnings("ignore", category=RuntimeWarning)
   37
   38
                  hmm_model = GaussianHMM(n_components=num_states, covariance_type="diag", n_iter=1000,
   39
                                          random_state=self.random_state, verbose=False).fit(self.X, self.lengths)
   40
   41
                      print("model created for {} with {} states".format(self.this_word, num_states))
   42
                  return hmm_model
   43
              except:
   44
                  if self.verbose:
   45
                      print("failure on {} with {} states".format(self.this_word, num_states))
   46
                  return None
   47
   48
   49
   50 class SelectorConstant(ModelSelector):
           """ select the model with value self.n_constant
   51
   52
   53
          def select(self):
   55
               """ select based on n_constant value
   56
              :return: GaussianHMM object
```

```
best num components = self.n constant
 59
            return self.base_model(best_num_components)
 61
 62
 63
 64 class SelectorBIC(ModelSelector):
        """ select the model with the lowest Bayesian Information Criterion(BIC) score
 65
 66
 67
        http://www2.imm.dtu.dk/courses/02433/doc/ch6 slides.pdf
 68
        Bayesian information criteria: BIC = -2 * logL + p * logN
 69
 70
        \label{eq:continuous} \begin{tabular}{ll} \tt def select(self): \\ \tt """ select the best model for self.this\_word based on \\ \end{tabular}
 71
 72
            BIC score for n between self.min\_n\_components and self.max\_n\_components
 73
 74
 75
            :return: GaussianHMM object
 76
 77
            warnings.filterwarnings("ignore", category=DeprecationWarning)
 78
            # TODO implement model selection based on BIC scores
 79
            1 = math.inf
 80
            model = None
 81
            for nc in range(self.min_n_components, self.max_n_components + 1):
 82
 83
                     hmm_model = GaussianHMM(n_components=nc, covariance_type="diag", n_iter=1000,
 84
                                               random_state=self.random_state, verbose=self.verbose)
 85
                     hmm_model.fit(self.X, self.lengths)
 86
                     d = hmm_model.n_features
 87
                     n = hmm model.n components
 88
                     N = len(self.lengths)
 89
                     p = n**2 + 2*n*d - 1
 90
 AWESOME
The no. of free parameters has been calculated correctly. This reflects your understanding of the topic and the research done 👍
Formula is perfectly implemented!
                     logL = hmm_model.score(self.X, self.lengths)
 91
                     bic = -2*logL + p*np.log(N)
92
                     if bic < 1:
 93
                         1 = bic
 94
                         model = hmm model
 95
                except:
 96
 97
                     pass
 98
            return model
99
100
101 class SelectorDIC(ModelSelector):
102
          '' select best model based on Discriminative Information Criterion
103
        Biem, Alain. "A model selection criterion for classification: Application to hmm topology optimization."
104
        Document Analysis and Recognition, 2003. Proceedings. Seventh International Conference on. IEEE, 2003.
105
        http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.58.6208&rep=rep1&type=pdf
106
        DIC = log(P(X(i)) - 1/(M-1)SUM(log(P(X(all but i))
107
108
109
110
        def select(self):
111
            warnings.filterwarnings("ignore", category=DeprecationWarning)
112
            # TODO implement model selection based on DIC scores
113
            l = -math.inf
114
            model = None
115
            for nc in range(self.min_n_components, self.max_n_components + 1):
116
117
                     hmm_model = GaussianHMM(n_components=nc, covariance_type="diag", n_iter=1000,
118
                                              random_state=self.random_state, verbose=self.verbose)
119
                     hmm_model.fit(self.X, self.lengths)
120
                     logL = hmm_model.score(self.X, self.lengths)
121
                     sc = 0
122
                     for word, (x, lengths) in self.hwords.items():
123
                         if word != self.this word:
124
                            sc += hmm_model.score(x, lengths)
125
                     mean_other_words = sc / (len(self.hwords) - 1)
126
                     dic = logL - mean other words
127
Very impressive! The score "P(X(all but i))" was a bit difficult to calculate and you nailed it
Formula for DIC has been perfectly implemented!
                     if dic > 1:
128
129
130
                         model = hmm_model
131
                except:
132
                     pass
133
            return model
```

```
134
     135
     136 class SelectorCV(ModelSelector):
                        ''' select best model based on average log Likelihood of cross-validation folds
     137
     138
     139
     140
                       # @property
     141
     142
                       def select(self):
                             warnings.filterwarnings("ignore", category=DeprecationWarning)
     143
                               # TODO implement model selection using CV
     144
     145
                               model = None
     146
                               low_score = -math.inf
                               for n_comp in range(self.min_n_components, self.max_n_components + 1):
     147
     148
                                     # get the model & its average score
     149
                                        score, hmm_model = self.avg_score(n_comp)
                                        \# find the model with highest average score
     150
       SUGGESTION
     Inline comments are used to make the code more readable
                                        if score > low_score:
     152
                                                low_score, model = score, hmm_model
     153
                               return model
     154
                       def avg_score(self, n_comp):
     155
                               \# Collect the average score of the model for its different folds.
     156
     157
                               scores = []
                               hmm_model = None
     158
                               n_split = 2 if len(self.sequences) < 3 else 3</pre>
     159
                               split_mtd = KFold(n_splits=n_split)
     160
                               for cv_train_idx, cv_test_idx in split_mtd.split(self.sequences):
     161
                                       x_train, len_train = combine_sequences(cv_train_idx, self.sequences)
     162
                                        x_test, len_test = combine_sequences(cv_test_idx, self.sequences)
     163
       AWESOME
     Good use of combine_sequences utility 👍
     164
                                                 # not using the base_model() as it uses verbose = False for all cases.
     165
                                                 \verb|hmm_model = GaussianHMM(n_components=n_comp, covariance_type="diag", n_iter=1000, linear content of the con
     166
                                                                                                     random_state=self.random_state, verbose=self.verbose)
     167
     168
                                                 hmm_model.fit(x_train, len_train)
                                                scores.append(hmm_model.score(x_test, len_test))
     169
     170
                                        except:
     171
                                               pass
                               avg = np.mean(scores) if len(scores) else -math.inf
     172
     173
                               return avg, hmm_model
     174
my_recognizer.py
▶ asl_utils.py
▶ asl_recognizer.html
```

RETURN TO PATH

Student FAQ