# **CEFR** sentence classification

We will try to build a classifier for the sentences labeled by CEFR level of difficulty.

```
in[1187]:= levelNames = {"a1", "a2", "b1", "b2"};
```

## Reading the data

### **Pre-Processing**

## Modeling

```
In[1225]:= embedding = NetModel[
                    "GloVe 50-Dimensional Word Vectors Trained on Wikipedia and Gigaword-5 Data"]
Out[1225]= EmbeddingLayer
                                                                           ! "#$"#%&' ()*&+),
                                                                            /01**/+")#,
                                                                                                             2. . . . 3
 In[1226]:= decoder = NetDecoder[{"Class", levelNames}]
                                                                      /01**
Out[1226]= NetDecoder
                                                                      {138 198 738 79}
                                              %&' ()*&+)*,
 In[1227]:= encoder = NetEncoder[decoder];
 In[1228]:= encodedLabels = Thread[UnitVector[Length[levelNames], encoder[labels]]];
               Dimensions[encodedLabels]
               RandomSample[{labels, encodedLabels}<sup>T</sup>, 4] // TraditionalForm
Out[1229]= \{18758, 4\}
Out[1230]//TraditionalForm=
                \{0,0,1,0\}
                 a2 \{0, 1, 0, 0\}
                 b2 \{0, 0, 0, 1\}
                 b1 \{0, 0, 1, 0\}
 in[1231]:= encodedWords = embedding[Flatten[sentences]];
 in[1232]:= encodedSentences = Map[Mean, encodedWords];
               Dimensions[encodedSentences]
Out[1233]= \{18758, 50\}
 In[1234]:= allData = Thread[encodedSentences → encodedLabels]; allData[[1]]
Out[1234]= \{0.214019, 0.271261, 0.0504771, -0.276674, 0.502119, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.0583472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.058472, 0.05847
                    -0.400325, -0.0207804, -0.278395, 0.0111262, -0.100528, 0.618868,
                    -0.417139, -0.137525, 0.651601, 0.404062, -0.005945, 0.105084,
                    0.130525, -0.439414, -0.228674, 0.428292, 0.379462, 0.0416915, 0.528543,
                    -1.73525, -0.789228, 0.155749, 0.446609, -0.553966, 2.80366, 0.391172,
                    -0.386574, -0.00050975, -0.116356, -0.154927, 0.13208, 0.0334483,
                    0.321645, -0.272565, -0.109711, 0.119482, -0.138774, 0.242132, 0.157917,
                    0.140987, -0.210705, -0.206912, -0.0672505, 0.348317\} \rightarrow \{1, 0, 0, 0\}
 In[1235]:= trainingData = RandomSample[allData, Round[0.7 Length[allData]]];
 In[1236]:= testingData = Complement[allData, trainingData];
 In[1237]:= net = NetChain[{LinearLayer[20], ElementwiseLayer[Ramp], LinearLayer[],
                          ElementwiseLayer[Ramp]}, "Input" → 50, "Output" → decoder];
```

```
In[1238]:= net = NetTrain[net, trainingData, Method → "ADAM", ValidationSet → testingData]
```

```
; (<#+= (*%>(, -.)
                               :)$"#
                           3 6%) (1=615(= ; (<#+= (*%>(, 9.)
                          9 ?1' $ ; (<#+= (*%>(, 9.)
@ 6%) (1=615(= ; (<#+= (*%>(, 2)
Out[1238]= NetChain
                           2 ?1' $ ; (<#+= (*%)(, 2)
                                             <01 * *
                               ! "#$"#
```

#### **Evaluation**

```
In[1239]:= predict = net[Keys[testingData]];
       actual = decoder[Values[testingData]];
logic{1}{1} = countsNN = Counts[Table[Apply[Equal, z], {z, {predict, actual}^{T}}]]
Out[1241]= \langle | True \rightarrow 4232, False \rightarrow 1390 | \rangle
       countsNN[True]
Total[countsNN]
Out[1242]= 0.752757
       And compare to plain Markov chain type classifier with same GloVe embeddings
In[1243]:= classifier = Classify[Keys[trainingData] → decoder[Values[trainingData]]];
In[1244]:= cm = ClassifierMeasurements[classifier,
          Keys[testingData] → decoder[Values[testingData]], "Accuracy"]
Out[1244] = 0.699217
```