

# CS 224n Programming Assignment 2

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## 1 Introduction

## 2 Methods

### 2.1 Models

#### 2.1.1 Superficial Word Aligner

This model extends the `BaselineWordAligner` class, which is very simple. In the `BaselineWordAligner` class, alignment is calculated along the diagonal; if the target sentence is longer than the source sentence, the alignment vector returns -1 for each target word beyond the source sentence. Additionally, the `train()` method takes pairs of sentences, iterates through each pair, and counts all possible mappings between source and target sentence pairs.

The `SuperficialWordAligner` class goes a little bit further than that. I based my implementation off of the suggested alignment probability, expressed formally as:

$$\frac{P(f, e)}{P(f)P(e)} \tag{1}$$

I calculated  $P_{MLE}(F)$  and  $P_{MLE}(E)$ , using the `getWordProbability()` method from the `EmpiricalUnigramLanguageModel` class (modifying the `LanguageModel` and `EmpiricalUnigramLanguageModel` classes so that `getWordProbability()` accepted a string). For  $P(f, e)$ , I modified the `train()` method such that

### 2.1.2 IBM Model 1

### 2.1.3 IBM Model 2

## 2.2 EM Algorithm

## 2.3 Decoding

## 2.4 Backing off

# 3 Results

## 3.1 Model evaluation metrics

## 3.2 Model performance

test	BaselineWordAligner	SuperficialWordAligner	Model1WordAligner	Model2WordAligner
Recall	0.225854383358	0.159732540861		
Precision	0.365896980461	0.395578365574		
AER	0.68649249583	0.74935321868		

validate	BaselineWordAligner	SuperficialWordAligner	Model1WordAligner	Model2WordAligner
Recall	0.33489096573	0.02366863905		
Precision	0.19822485207	0.31460674157		
AER	0.71224489795	0.91569086651		

## 3.3 Decoding performance

## 3.4 Observations

# 4 Error analysis