Lecture-31 Statistical Machine Translation (EM Algorithm)

Statistical Machine Translation (SMT) is a translation technique where **translation is framed as a probability problem**. Given a source sentence in one language (e.g., English), SMT searches for the most probable sentence in the target language (e.g., Hindi) based on a trained statistical model.

Key Concepts in SMT

1. Bayes' Rule:

SMT uses Bayes' theorem to find the best translation:

$$\hat{e} = rg \max_{e} P(e|f) = rg \max_{e} P(f|e) \cdot P(e)$$

- f = foreign/source sentence (e.g., Hindi)
- e = English/target sentence
- P(e): Language model (fluency of English sentence)
- P(f|e): Translation model (how likely Hindi words come from English words)

We want the computer to learn how to translate English words into Hindi words using a parallel corpus (same meaning, different language).

We're using IBM Model 1, which:

- Ignores word order (just learns word-word pairs).
- Learns translation probabilities.
- Uses an algorithm called Expectation-Maximization (EM).

Example

English	Hindi
the cat runs	बिल्ली दौड़ती है
a dog jumps	कुत्ता कूदता है
the dog runs	कुत्ता दौड़ता है
a cat jumps	बिल्ली कूदती है

Step 1: Vocabulary Setup

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• English Words: ["the", "cat", "runs", "a", "dog", "jumps"]
Hindi Words: ["बिल्ली", "दौड़ती", "है", "कुत्ता", "क्दता", "दौड़ता", "क्दती"]
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We include all Hindi words that appear in any sentence. So every English word could translate to any Hindi word.

Step 2: Initialize Uniform Probabilities

We don't know anything yet — so we assume each English word can translate to each Hindi word equally.

For example:

- P(बिल्ली | the) = 1 / 7 = 0.143
- P(ਨੈ | cat) = 1 / 7 = 0.143 And so on for all combinations.

Step 3: Expectation Step (E-step)

We go sentence by sentence. For each English-Hindi pair:

- For every Hindi word h in that sentence, and every English word e:
 - We say: "Let's give e a fair share of contributing to h, based on how likely it is."
 - We calculate expected counts of which word likely created which.

Now we calculate the fractional count assigned to each Hindi word for "cat":

- Count(cat, बिल्ली) = 0.143 / 0.429 = 0.333
- Count(cat, दौड़ती) = 0.143 / 0.429 = 0.333
- Count(cat, है) = 0.143 / 0.429 = 0.333

Same goes for "the" and "runs", because their initial probabilities are also equal.

So for each English word in the sentence, the fractional count for each Hindi word becomes:

Let's say total 6 English words and 8 Hindi words

Step 2: Initialize — Start with equal guessing

Since we don't know any translations yet, we guess equally:

Each English word can be translated to any Hindi word with equal chance.

So:

P(हिंदी|English) = 1 / 8 = 0.125 (or 12.5%)

Step 3: E-Step (Expectation Step)

Let's take one sentence pair and calculate how much "credit" each English word gets for producing each Hindi word.

Let's take Sentence 1:

English: the cat runs Hindi: बिल्ली दौड़ती है

- 3 English words → the, cat, runs
- 3 Hindi words → बिल्ली, दौड़ती, है

We now compute for each Hindi word:

Step 3.1: For each Hindi word, divide the probability among English words:

- For "बिल्ली":
 - o Candidates: the, cat, runs
 - All have equal initial chance (0.125)

$$= 0.125 + 0.125 + 0.125 = 0.375$$

Now divide each by total to get their share of credit:

• the: 0.125 / 0.375 = 0.333

• cat: 0.333

• runs: 0.333

So we say:

"the", "cat", and "runs" each get 1/3 credit for producing "बिल्ली".

Repeat the same for "दौड़ती" and "है".

Do the same process for all 4 sentences.

Step 1: Vocabulary Extraction

- English words (E): {"the", "a", "boy", "girl", "eats", "reads", "runs"}
- Hindi words (H): {"है", "लड़का", "खाता", "लड़की", "खाती", "पढ़ती", "दौड़ता"}

Step 2: Initialize Translation Probabilities

Assume uniform probabilities:

- Every English word can translate to any Hindi word equally.
- Total Hindi words = 7, so each $P(h|e) = 1/7 \approx 0.143$

Hindi word লাহকা aligning to each English word in sentence: the, boy, eats