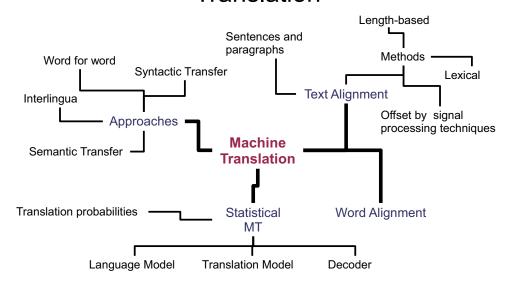
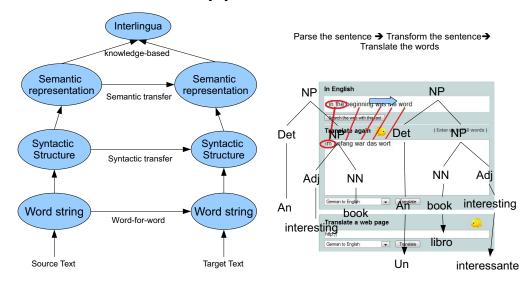
Statistical Alignment and Machine Translation

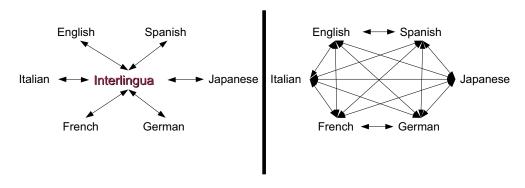


Approaches



Approaches: Interlingua

- Independent of the way particular languages express meaning
- Need only O(n) translation systems



Text Alignment

- Parallel texts (bitexts):
 - which paragraphs or sentences in one language correspond to which paragraphs or sentences in another language
 - Which words tend to be translated by which other words
- · Aligning sentences and paragraphs
 - Some group of sentences in one language corresponds in context to some group of sentences in the other language
 - 1:1 one sentence in the source document corresponds to one sentence in the target document
 - n:m n sentences in the source document correspond to m sentences in the target document
 - How much content has to overlap between sentences?
 - Problem of crossing dependecies

Methods in Text Alignment

- Length-Based Approaches: compare the lengths of units of text in the parallel corpora (short sentences will be translated as short sentences and long sentences as long sentences)
- Offset Alignment by Signal Processing Techniques: attemp to align position offsets in the two parallel texts
- Lexical Methods: Use lexical information to align beads of sentences.

Offset Alignment by Signal Processing Techniques

- Church (1993)
 - Alignment using cognates (words that are similar across languages)

supérieur (French) – superior (English)

- Find cognates at the level of character sequences
- Dot-plot construction
- Fung & McKeown(1994)
 - Seek an algorithm that will work without having found sentence boundaries, in only roughly parallel texts, and with unrelated language pairs
 - A small bilingual dictionary gives points of alignment
 - For each words, a signal is produced:

 an arrival vector of integer numbers giving the

 number of words between eachoccurence of the word at hand
 - A measure of similarity between signals is calculated using Dynamic Time Warping

Length-based approaches

- Gale & Church (1993)
 - Find the alignment A with highest probability given the two parallel texts S and T
 - To estimate the probabilities: decompose the aligned texts into a sequence of aligned beads
- Brown et al. (1991)
 - Similar to Gale & Church, but works by comparing sentence length in words (not characters)
- Wu(1994)
 - Unrelated languages (English and Cantonese)
 - Uses lexical cues

Lexical Methods

- Kay & Röscheisen(1993)
 - Assumption: two words should correspond if their distributions are the same
 - Steps:
 - Assume the first and last sentences of the texts align. These are the initial anchors.
 - Then, until most sentences are aligned:
 - Form an envelope of possible alignments.
 - Choose pairs of words that tend to co-occur in these potential partial alignments.
 - Find pairs of source and target sentences which contain many possible lexical correspondences.
- Chen (1993)
 - Sentence alignment by constructing a simple word-to-word translation model
 - The best alignment is the one that maximizes the likelihood of generating the corpus given the translation model
- Haruno & Yamazaki (1996)
 - Work with structurally different languages
 - Do lexical matching on content words only (use POS taggers)
 - Use an online dictionary to find matching word pairs

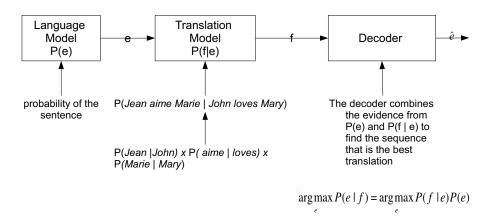
Word Alignment

- Derivation of bilingual dictionaries and terminological databases:
 - text alignment is extended to a word alignment.
 - some criterion (frequency) is used to select aligned pairs
- Word alignment based on measures of association (χ^{2})
 - works well unless one word in L1 frequently occurs with more than one word in L2. Then, it is useful to assume a one-to-one correspondence.

Translation Probabilities

- Are estimated using the EM algorithm, which solves the credit assignment problem
- Random initialization of the translation probabilities
- Compute the expected number of times we will find w_f in the French sentence given that we have w_e in the English sentence
- Reestimate the translation probabilities from the expectations

Statistical Machine Translation



Problems (Brown et al. 1990, 1993)

- Fertility is asymmetric
- · Independence assumptions
- Sensitivity to training data
- Efficiency

Lack of linguistic knowledge

- · No notion of phrases
- Non-local dependencies
- Morphology
- Sparse data problems