

Finite-State and the Noisy Channel

6.00.465 - Intro to NLP - J. Eisner

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Word Segmentation

theprophetsaidtothecity

- What does this say?
 - And what other words are substrings?
- Could segment with parsing (how?), but slow.
- Given L = a "lexicon" FSA that matches all English words.
- How to apply to this problem?
- What if *Lexicon* is weighted?
- From unigrams to bigrams?
- Smooth L to include unseen words?

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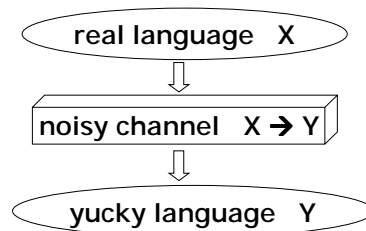
Spelling correction

- Spelling correction also needs a lexicon L
- But there is distortion ...
 - Let T be a transducer that models common typos and other spelling errors
 - $ance \rightarrow ence$ (deliverance, ...)
 - $e \rightarrow \epsilon$ (deliverance, ...)
 - $\epsilon \rightarrow e // \text{Cons_Cons}$ (athlete, ...)
 - $rr \rightarrow r$ (embarrass, occurrence, ...)
 - $ge \rightarrow dge$ (privilege, ...)
 - etc.
 - Now what can you do with L .o. T ?
 - Should T and L have probabilities?
 - Want T to include "all possible" errors ...

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Noisy Channel Model

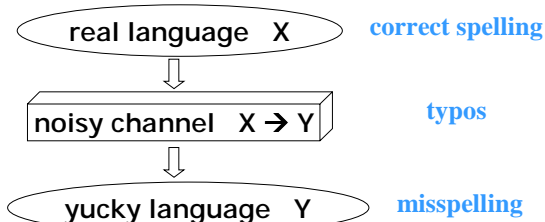


want to recover X from Y

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Noisy Channel Model

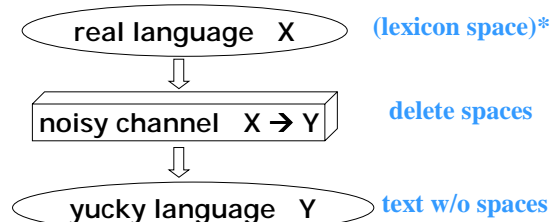


want to recover X from Y

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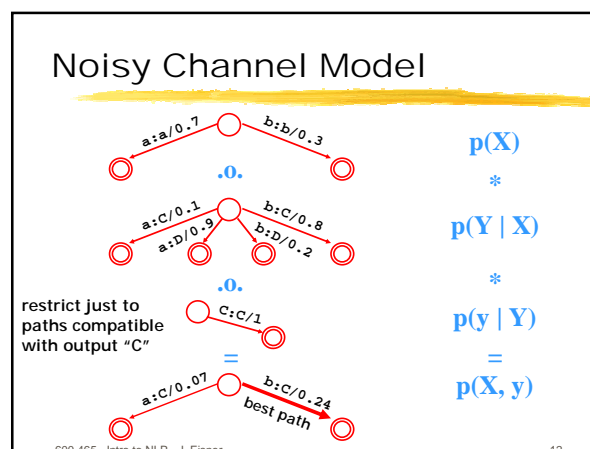
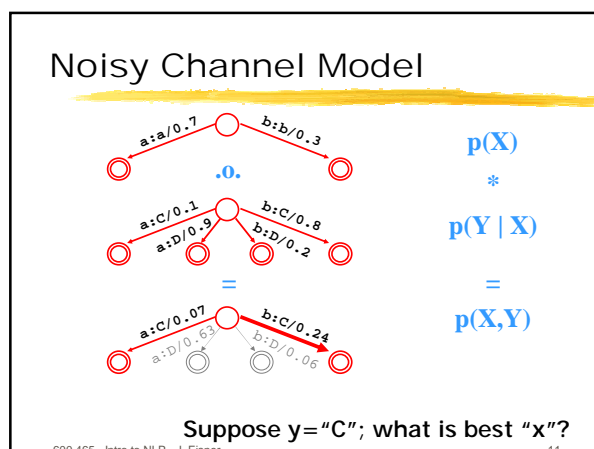
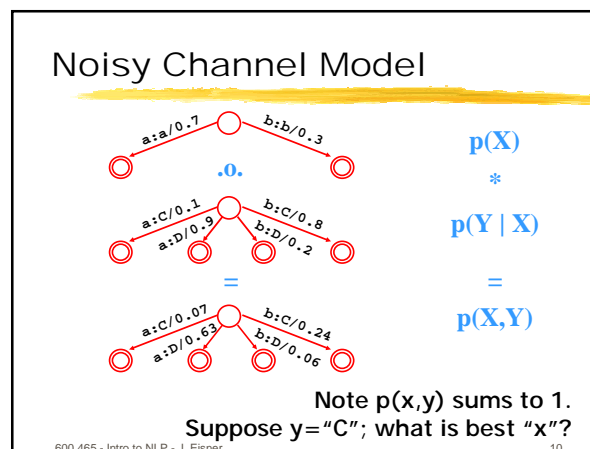
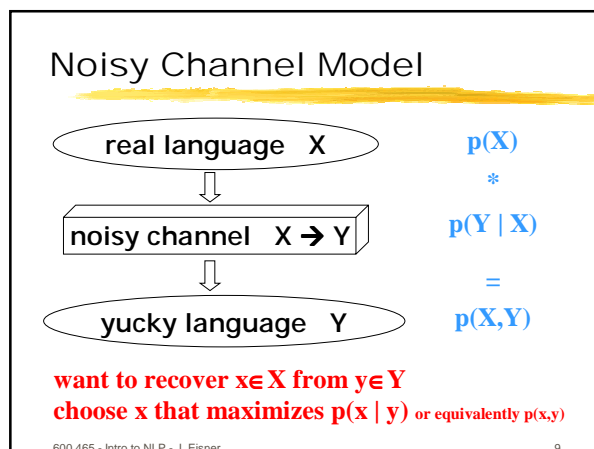
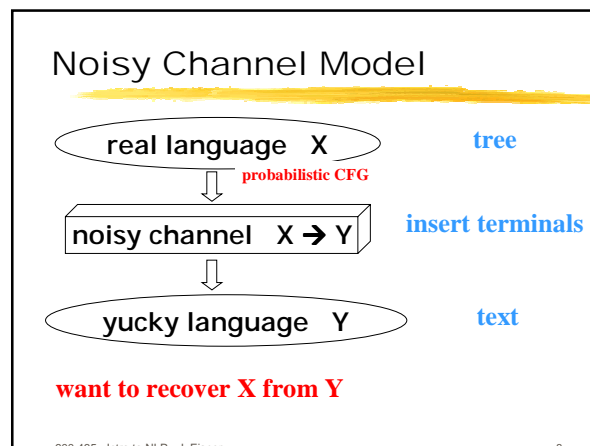
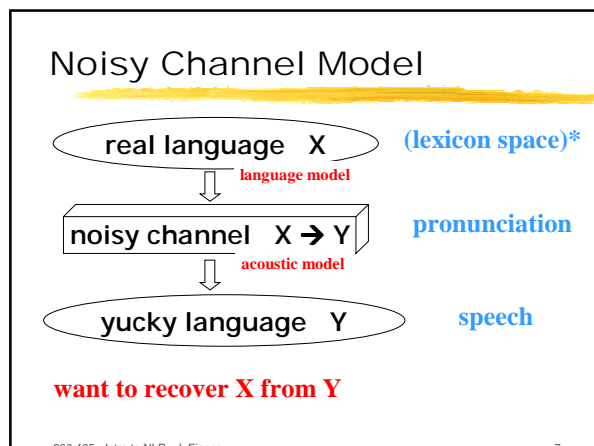
Noisy Channel Model



want to recover X from Y

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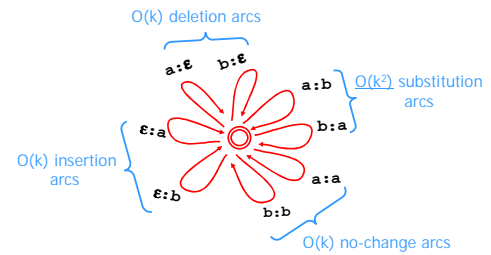
Morpheme Segmentation

- Let *Lexicon* be a machine that matches all Turkish words
 - Same problem as word segmentation
 - Just at a lower level: morpheme segmentation
 - Turkish word: *uygarlas.tiramadiklarimizdanmis.sinizcasina*
 = *uygar* + *las* + *tir* + *ama* + *dik* + *lar* + *imiz* + *dan* + *mis* + *siniz* + *casina*
 (behaving) as if you are among those whom we could not cause to become civilized
 - Some constraints on morpheme sequence: bigram probs
 - Generative model – concatenate then fix up joints
 - stop + -ing = stopping, fly + s = flies
 - Use a cascade of transducers to handle all the fixups
 - But this is just morphology!
 - Can use probabilities here too (but people often don't)

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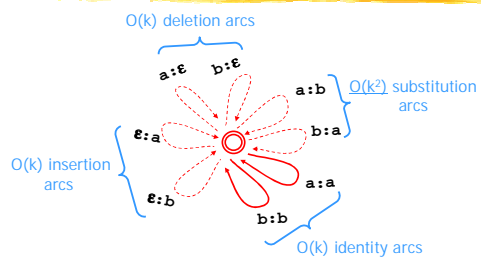
Edit Distance Transducer



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Stochastic Edit Distance Transducer

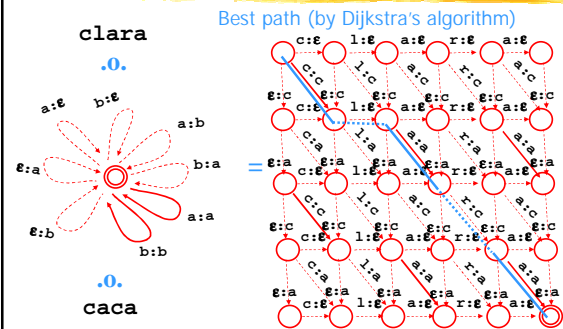


Likely edits = high-probability arcs

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Stochastic Edit Distance Transducer



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