MORPHOLOGICAL ANALYSIS OF INUKTITUT

STATISTICAL NATURAL LANGUAGE PROCESSING FINAL PROJECT



Why Inuktitut?

- Official language of Nunavut
 - Government
 - Education

- Search Engines
- Spellcheckers
- Dictionaries, Thesaurus
- Grammar checkers

Inuktitut Resources



Inuktitut Computing dot CA

Français w

Mission: To facilitate the use of Inuktitut in its written form on computers and the web by providing useful tools and links to important resources.



Morning at Iqaluit, Nunavut

Inuktitut Morphological Analyser Run

Bibliographic References

Dictionary: "Inuktitut - A Multi-dialectal Outline Dictionary" by Alex Spalding

Display and Input of Inuktitut Syllabic Characters - Unicode and Legacy Fonts

Inuktitut-English Parallel Corpus

Inuktitut Linguistics for Technocrats by Mick Mallon

Linguistic Data Base

NANIVARA - Inuktitut Search Engine Run

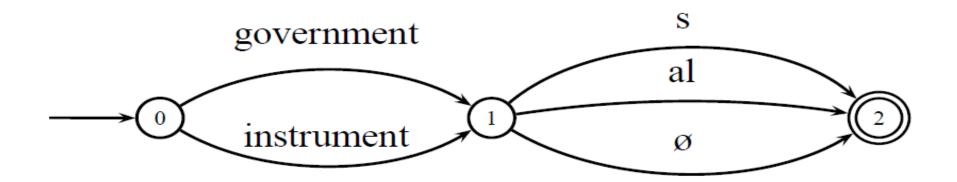
Searching the Nunavut Hansard Run

Transcoder Run

Transliteration of Web Pages

Morphology 101

- Most languages have morphology
- Most morphology consists of either suffixes or prefixes



Why Morphological Parsing?

- Information Retrieval
 - Remove morphs
- Machine Translation
- Named Entity Recognition
- Natural Language Understanding
 - Use morphs

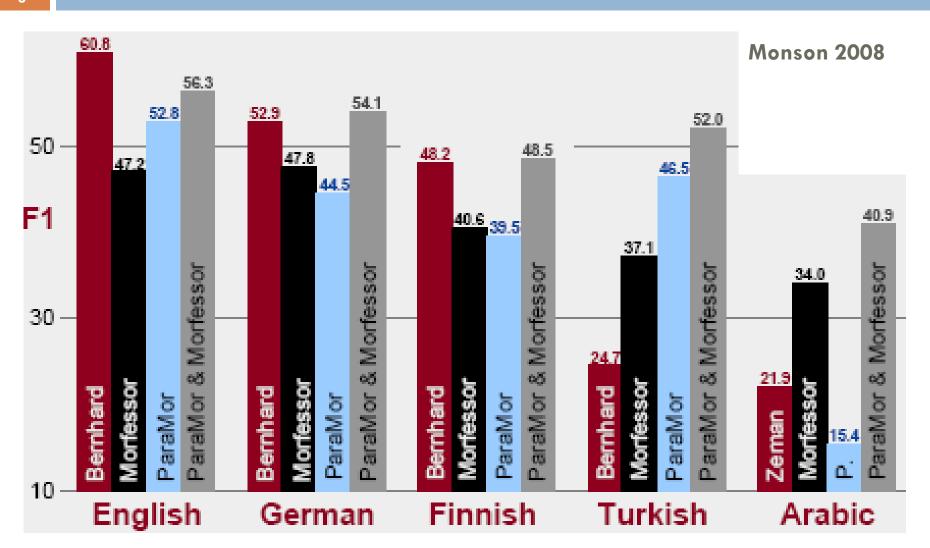
Stemming

- Useful for Information Retrieval
- □ Reduces feature space

Full Parsing

- Natural Language Generation
- Machine Translation
- Named Entity Tagging
- Text Summarization

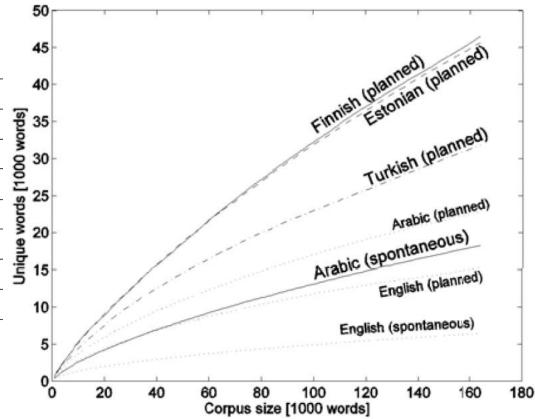
Low accuracy (F scores < 50%) Performance heavily dependant on language type



Morphology 102

- Agglutinative languages more morphemes per word (Pirkola 2001)
- And unlimited words (Kurimo 2008)

| Vietnamese | 1,06 |
|----------------|------|
| Yoruba | 1,09 |
| English | 1,68 |
| Old English | 2,12 |
| Swahili | 2,55 |
| Turkish | 2,86 |
| Russian | 3,33 |
| Inuit (Eskimo) | 3,72 |



Five Approaches

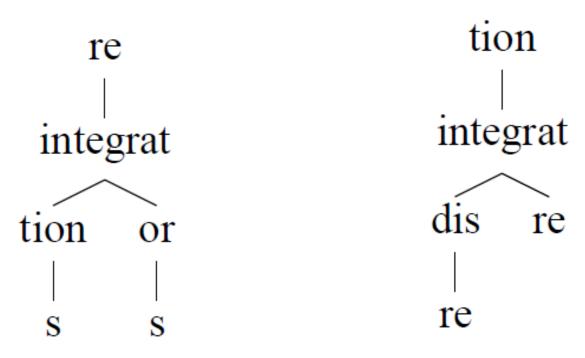
- Transition Likelihood
 - 1. Harris 1958
 - 2. Johnson & Martin 2003 (English, Innuktitut) HubMorph
 - 3. Bernhard 2007 (English, German, Finnish)
- 2. Minimum Description Length
 - Brent 1995 MBDP-1
 - De Marken 1995 Composition and Perturbation
 - 3. Goldsmith 2001,2006 Linguistica
 - 4. Creutz 2006 Morfessor
- 3. Paradigms
 - 1. Goldsmith 2001,2006 Linguistica
 - 2. Snover 2002
 - Monson 2008 ParaMor
- 4. Word Edit Distance & Latent Semantic Analysis of word context
 - 1. Yarrow & Wicentowski 2000
- Phonotactic/ Allomorphy
 - Heinz MBDP-Phon-Bigrams

Compression

- Morphological Parsing as Compression
 - Tries
 - Minimum Description Length

Harris 1955

□ Forward and Backward Tries



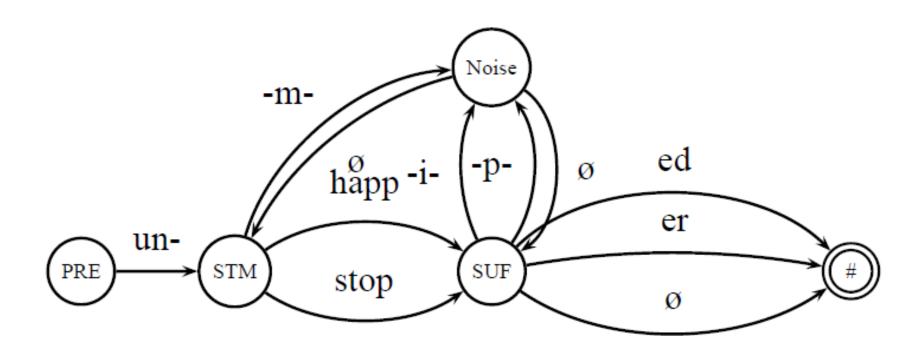
Minimum Description Length

The more morphs you find, the smaller the key

```
Key:
                                                       sima
                                                4
Data:
                                                       laur
  kati rsui sima giaqanir mik
                                                       tu
  kati maniu laur tu mik
                                                       tusaa
                                                10
  tusaa jimmaringulaur sima jug
                                                       giaganir
                                                       maniu
  tusaa jimmaringu laur sima jug
                                                16
                                                       tugut
  mali tsiariaga laur tugut
                                                1
                                                       @
  tiki laur tuqut
                                                       mali
                                                14
                                                       jimmaringu
                                                13
                                                       jimmaringulaur
                                                11
Compressed Data:
                                                       kati
                                                2
T = 32
                                                3
                                                       rsui
   3 4 5 6 1 2 7 8 9 6 1 10 11 4 12 1
                                                       mik
  13 8 4 12 1 14 15 8 16 1 17 8 16
                                                12
                                                       juq
                                                       tiki
                                                17
                                                       tsiariaga
                                                15
```

Creutz 2005: Morfessor

- A Hidden Markov Model with 4 states
- Morphemes are the strings which transition between them and the probabilities of that transition



Morfessor Performance on Inuktitut

- □ 7% Precision
- □ 7% Recall

■ Why?

```
Sample output by length(in number of sylables):
17 saimmatitsigasuarutaujaarnaturinalaugsimavuq
17 allasimajuliugpaliagutigarumaligsungattaug
17 aatutigangitsiammarittuugutigilaursimajara
16 uqalimaagaksaliarijaugattalirajartuunnik
16 tusaumautiqattautitsiarunnarnirsaugajarnirmut
16 sivumuqiallaqutiqaqsimaliraluartilluqit
16 piruqpalialaqititauqutiqinqunnatanqinnik
16 pinasuagatuarijaunngusutsimagalaaganilu
16 katiqatigiinnirsaugutigijunnarsigajarmauk
16 isumaksarsiurutigallattaagattaraluarpita
15 taimannganimmarialunitauqataummigamilli
15 pivallirtitsijummariujunnarasugillugu
15 pigunnaniqsauqigigajannguataraluaminnit
15 kiinaujagagtitaugutigigunnatanginnilu
15 kiinaujaliurasuarutiksalirinasuarluta
15 isumagigasiutigiagangikkaluarpitigut
15 ikpiqusuqiallagutiqinirsarilirakkit
15 aviktursimaugutigivalliatuinnartattinnik
```

Morphology 103

- Morphemes cannot appear in any order, the ordering is fixed
 - Within a language
 - For all human possible languages

FieldWorker 0.005

- Goals
 - Create a general system
 - Grow from an initial assumption of root+suffix (which is true for all human languages argument+head)
 - Expand to allow prefixing, multiple suffixes, compounding
 - Flexible enough to allow for allomorphs
 - Flexible enough to allow for nonlocal dependencies

Learning Grammar from morphological precedence relations

- Discover template
 - Take long words containing seed morphemes to discover full template
- Discover morphemes
 - Create dense corpora to find morphemes for each template category

Creating an Inuktitut Corpus

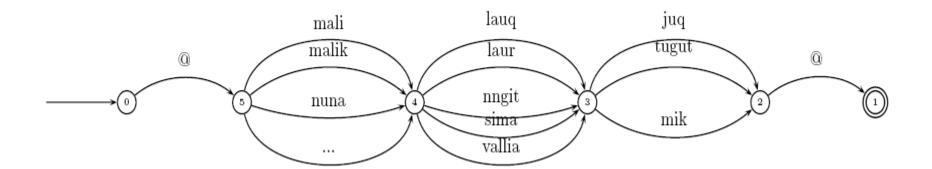
- Nunavut Hansard
 - Spelling is unsystematic, introduces too much noise for statistical learning Examples of spelling variation:

- 10 aaggaagtugangittug
- 6 aaggaaqtuqanngittuq
- 2 aanniagannangittulirijikkuni
- 5 aanniagannangittulirijikkunni
- Created a corpus from Inuktitut Magazine vol. 102-104
 - Parallel corpus in Inuktitut, English and French
 - 17,000 Inuktitut words for 32,000 English words
 - Consistent spelling

Overview

- 1. Corpus word list
- 2. Word list ranked possible morphs
- 3. Possible morphs seed list
- 4. Seed list precedence relations
- 5. Precedence relations dense corpus
- 6. Dense corpus precedence relations
- 7. Iterate

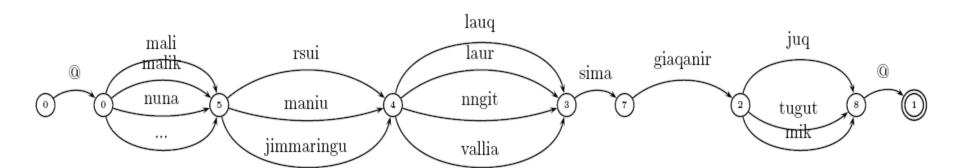
Sample Seed list



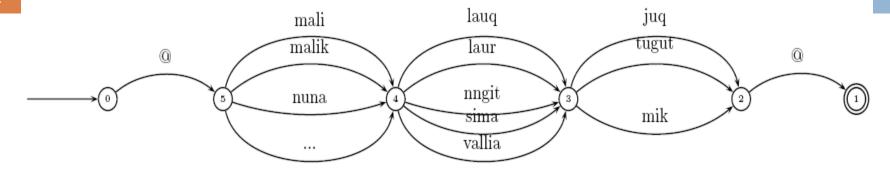
Gives a dense mini-corpus

```
1 + +kati+ +rsui+ +sima+ +giaqanir+ +mik+ +
1 + +kati+ +maniu+ +laur+ +tu+ +mik++
1 + +tusaa+ +jimmaringulaur+ +sima+ ++ +juq+ +
1 + +tusaa+ +jimmaringu+ +laur+ +sima+ +juq+ +
1 + +mali+ +tsiariaqa+ +laur+ ++ +tugut+ +
1 + +tiki+ ++ +laur+ ++ +tugut+ +
```

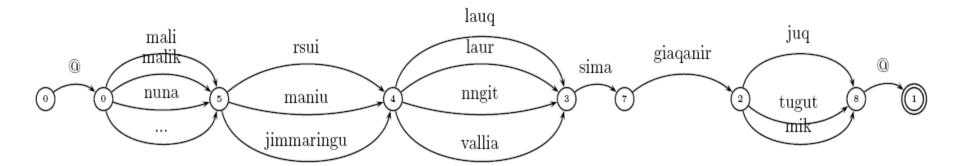
Which gives a new template



Progress



```
1 + +kati+ +rsui+ +sima+ +giaqanir+ +mik+ +
1 + +kati+ +maniu+ +laur+ +tu+ +mik++
1 + +tusaa+ +jimmaringulaur+ +sima+ ++ +juq+ +
1 + +tusaa+ +jimmaringu+ +laur+ +sima+ +juq+ +
1 + +mali+ +tsiariaqa+ +laur+ ++ +tugut+ +
1 + +tiki+ ++ +laur+ ++ +tugut+ +
```



Evaluation

```
1 + +kati+ +rsui+ +sima+ +gia!qanir+ +mik+ +
1 + +kati+ +maniu+ +laur+ +tu+ +mik++
1 + +tusaa+ +jimma!ringu!laur+ +sima+ ++ +juq+ +
1 + +tusaa+ +jimma!ringu+ +laur+ +sima+ +juq+ +
1 + +mali+ +tsia!ria!qa+ +laur+ ++ +tugut+ +
1 + +tiki+ ++ +laur+ ++ +tugut+ +
```

Recall =
$$22/32$$
 69%

Evaluation

- Recall goes up as the model iterates
- Precision goes down as the model iterates

■ Where to stop the model?

Morpho Challenge 2009 August

- Run my algorithm on English, German, Finish, Turkish and Arabic corpora of Morpho Challenge 2008
- □ If I am able to achieve respectable F-scores (~50%)
- Submit my algorithm to Morpho Challenge 2009

References

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- □ Johnson, Mark. 2008a. Unsupervised word segmentation for Sesotho using adaptor grammars. In Tenth Meeting of ACL SIGMORPHON, pages 20–27. ACL, Morristown, NJ.
- Johnson, Mark. 2008b. Using adaptor grammars to identify synergies in the unsupervised acquisition of linguistic structure. In 46th Annual Meeting of the ACL, pages 398–406. ACL, Morristown, NJ.
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