

Using Hand-Written Rewrite Rules to Induce Underlying Morphology

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Unsupervised Morpheme Analysis – Morpho Challenge 2007

Outline

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Definitions

We consider **morphemes** to be...

- ▶ basic units of grammar with no internal structure which may be composed together to form words
- ▶ realized as sequences of linguistic symbols (phones and/or letters)

Morphemes may be rendered differently in different contexts:

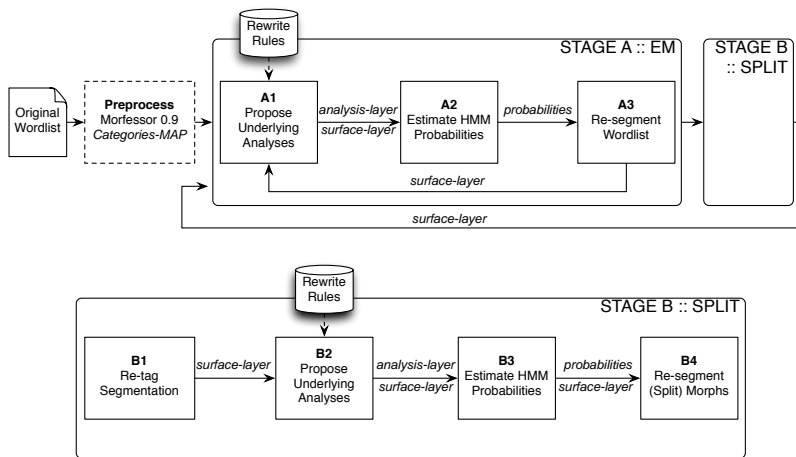
- ▶ lexical context: /s/ → en, as in *oxen*
- ▶ **phonological/orthographic context: /s/ → es, as in *dresses***

Morphological *variants* are known as **allomorphs**

Examples

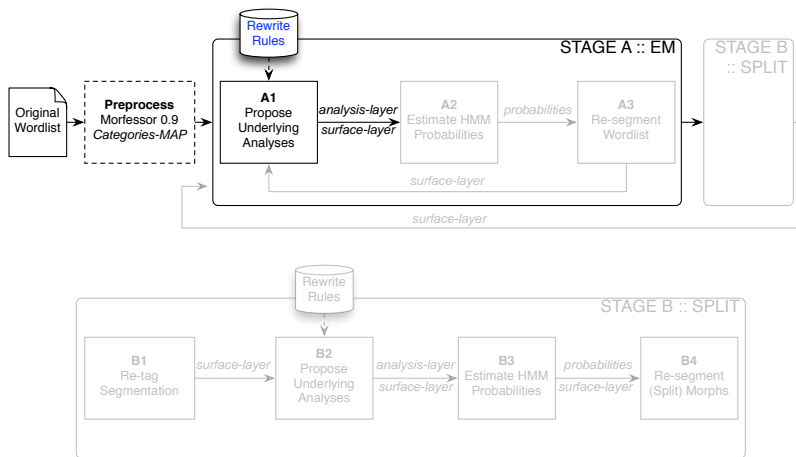
Language	Type	Morpheme	Allomorphs
English	<i>stem</i>	/wake/	wake, wak
	<i>suffix</i>	/s/	s, es
Finnish	<i>stem</i>	/katto/ roof	katto, kato
	<i>suffix</i>	/ta/ partitive	a, ä, ta, tä
Turkish	<i>stem</i>	/kanad/ wing	kanad, kanat
	<i>suffix</i>	/dik/ nominalizer	dik, dük, dık, duk tik, tük, tık, tuk diğ, düğ, dığ, duğ tiğ, tüğ, tığ, tuğ

Flowchart



Rewrite Rules

Flowchart



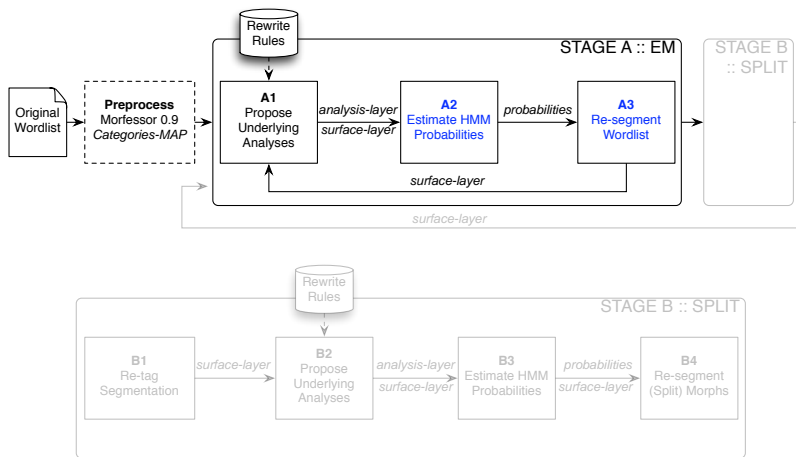
Analysis by Rewrite Rules

- ▶ Written as cascaded (ordered) rewrite rules and compiled into regular expressions.
- ▶ Rules are meant to be run in the analysis direction on a surface segmentation
- ▶ For efficiency, we only permit two types of analyses per segment s :
 - ▶ analyses where all the rules that could have applied, did. (u'')
 - ▶ analyses where no rules applied ($u' = s$)
- ▶ Example Rule capturing the fact that English suffix /s/ is written as es after sibilants (s, z, sh, ...):

$$\underset{\text{underlying}}{\emptyset} \rightarrow \underset{\text{surface}}{e} / [+SIB] + _s \quad (1)$$

Stage A :: Basic EM

Flowchart



Stage A :: Basic EM

- ▶ We **estimate** transition and emission probabilities of a morfeessor-style HMM via maximum likelihood.
- ▶ Emission probabilities are estimated by observing cooccurrences of segments s_i in the surface layer, u_i in the analysis layer, with tags t_i to estimate the probability $P(u_i|t_i)$ of emitting **underlying** morphemes:

$$P(u_i|t_i) = \sum_{s \in \text{allom.-of}(u_i)} P(u_i, s|t_i) \quad (2)$$

Where:

$$u_i = \begin{cases} u_i' & \text{if } u_i = s_i \\ u_i'' & \text{otherwise} \end{cases}$$

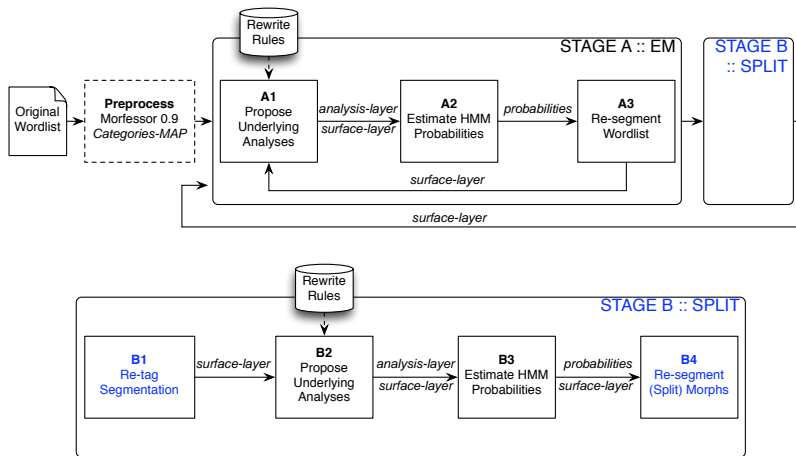
Stage A :: Basic EM

- ▶ Find the **maximum** probability segmentation of the wordlist by finding the argmax of the following equation for each word:

$$\operatorname{argmax}_{\mathbf{u}, \mathbf{t}} P(\mathbf{u}|\mathbf{t})P(\mathbf{t}) \approx \operatorname{argmax}_{\mathbf{u}, \mathbf{t}} \left[\prod_{i=1}^n P(u_i|t_i)P(t_i|t_{i-1}) \right] \quad (3)$$

Stage B :: Split Segments

Flowchart



Stage B :: Split Segments

- ▶ **Re-tag** the segmentation first, using Creutz and Lagus's 2004-2005 heuristic technique, such that only morphs exhibiting prototypical **affix-** or **stem-**distributional features are tagged as such.
- ▶ The remainder are tagged as **noise**; this makes them unavailable to be used in splitting.
- ▶ Key: Forcably **split** segments that are too frequent break under normal circumstances.

F-Measure Results

Language	Method	Precision	Recall	F-Measure
English	Morf.- <i>CatMAP</i>	82.17%	33.08%	47.17%
	Bernhard2	61.63%	60.01%	60.81%
	Tepper2-b300	75.62%	51.72%	61.43% 1% impr.
Finnish	Morf.- <i>CatMAP</i>	76.83%	27.54%	40.55%
	Bernhard2	59.65%	40.44%	48.20%
	Tepper-b600	62.01%	46.20%	52.95% 10% impr.
Turkish	Zeman	65.81%	18.79%	29.23%
	Morf.-CatMAP	76.36%	24.50%	37.10%
	Tepper-b100	61.15%	49.22%	54.54% 47% impr.

Summary

- ▶ Our approach, which utilizes a small amount of knowledge in an otherwise unsupervised framework, is successful at learning underlying morphology.
- ▶ Learning improvements over unsupervised approaches are more dramatic for languages with more allomorphic effects, like Turkish (not surprising).
- ▶ There is hope that with a technique such as ours we can pinpoint generalizations about the most effective rules, which would be useful towards developing features for templates from which to **learn** rules.

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