

Preference Grammars and Soft Syntactic Constraints for GHKM Syntax-based SMT

Matthias Huck, Hieu Hoang, Philipp Koehn



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Introduction

Feature-based integration of syntactic information into GHKM [string-to-tree translation](#)

- [Preference grammars](#):
soft target-side syntax
 - Target syntax as a feature
rather than via labeled non-terminals in the SCFG
- [Soft syntactic constraints](#):
non-restrictive source-syntactic enhancement
 - No hard source-syntactic constraints
(as in standard tree-to-tree translation)
imposed in extraction or decoding

Our empirical evaluation: [English](#)→[German](#) WMT task

Related Work

- GHKM string-to-tree translation: *Galley et al. (2004)*
- Open-source **Moses** implementation for GHKM translation
 - GHKM rule extraction:
Williams and Koehn (2012)
 - Decoding with CYK+ parsing and cube pruning:
Hoang et al. (2009)
 - Competitive results for European language pairs:
Nadejde et al. (2013); Williams et al. (2014)
- **Preference grammars**: beneficial as a syntactic extension of hierarchical systems (*Venugopal et al., 2009; Stein et al., 2010*)
- **Soft syntactic constraints**: related source-syntactic techniques improved hierarchical (*Marton and Resnik, 2008; Vilar et al., 2008; Hoang and Koehn, 2010*) and other syntax-based systems (*Zhang et al., 2011; Huang et al., 2013*) on Chinese→English and Arabic→English tasks

Preference Grammars

- Target-side non-terminals not decorated with syntactic labels, but with a **single generic non-terminal symbol**

baseline

$X, ADJD \rightarrow \langle \text{present, anwesend} \rangle$
 $X, ADV \rightarrow \langle \text{present, anwesend} \rangle$
 $X, AP-PD \rightarrow \langle \text{present, anwesend} \rangle$
...

preference grammar system

$X, X \rightarrow \langle \text{present, anwesend} \rangle$

- Distribution of implicit target label vectors** stored as additional information with each translation rule

$X, X \rightarrow \langle \text{present, anwesend} \rangle \# (ADJD) 0.98 (ADV) 0.001 (AP-PD) 0.01 \dots$

- Computation of a **tree-wellformedness feature** during decoding

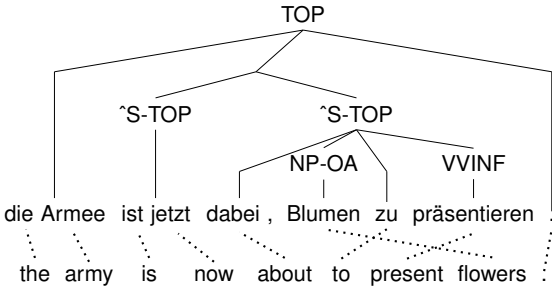
Soft Source Syntactic Constraints

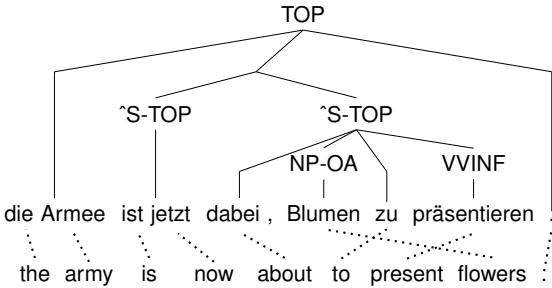
- Parse source-side data as well
- GHKM extractor stores an additional rule property:
source syntactic label vectors
- Provide parsed input data to the decoder
- During decoding, score matches and mismatches of source label vectors and input labels
- Soft features, no hard constraints

rule	source label vectors
$X, VP-OC \rightarrow \langle \text{present, zu pr\u00e4sentieren} \rangle$	$(VB), (NN)$
$X, ADJD \rightarrow \langle \text{present, anwesend} \rangle$	$(ADJP), (ADVP), \dots$
$X, TOP \rightarrow \langle X^{0} \text{ is now } X^{1} . , \text{jetzt ist } NP-SB^{0} VP-OC^{1} . \rangle$	$(TOP, NP, ADJP), \dots$
$X, ^S-TOP \rightarrow \langle \text{is } X^{0} X^{1}, \text{ist } ADV^{0} ADJD^{1} \rangle$	$(VP, ADVP, ADJP), \dots$

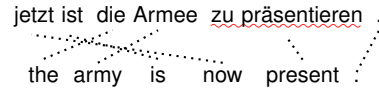
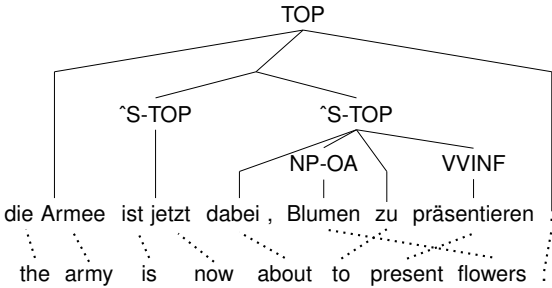
the army is now about to present flowers .

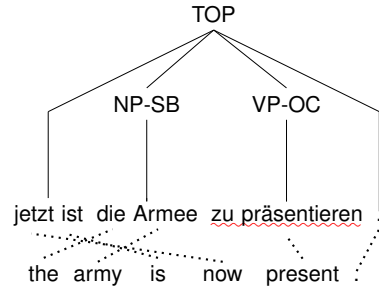
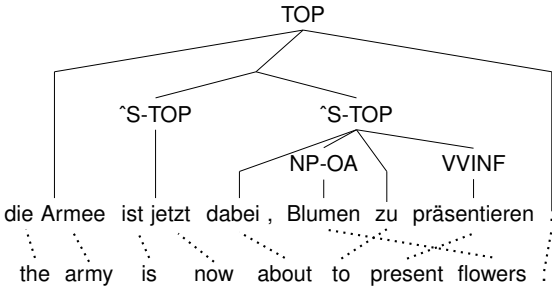
die Armee ist jetzt dabei , Blumen zu präsentieren .
the army is now about to present flowers .

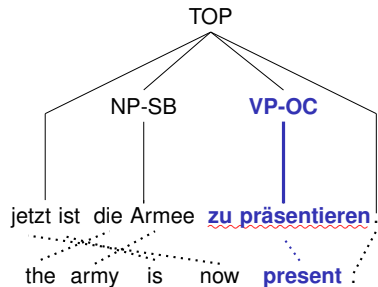
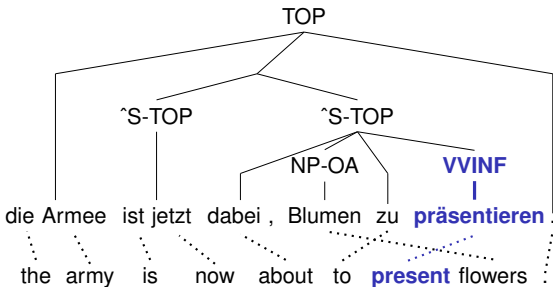


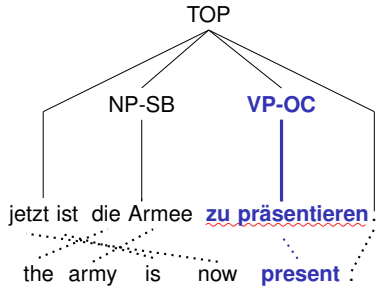
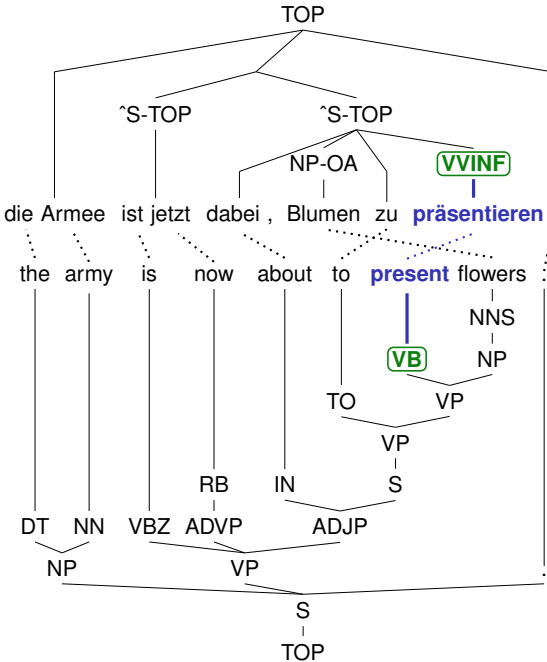


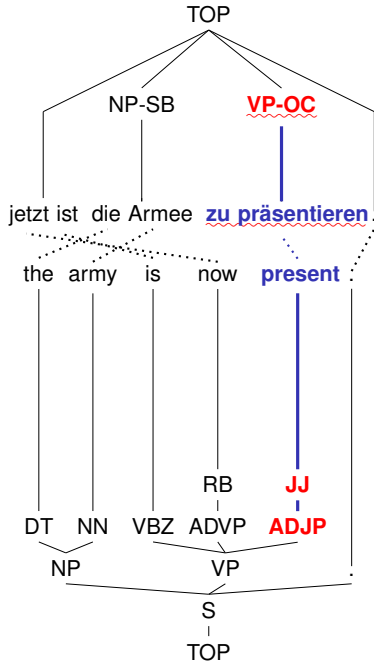
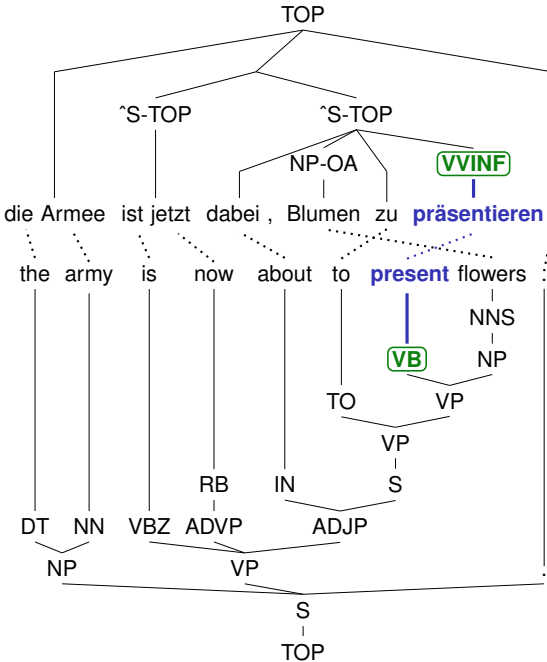
the army is now present .

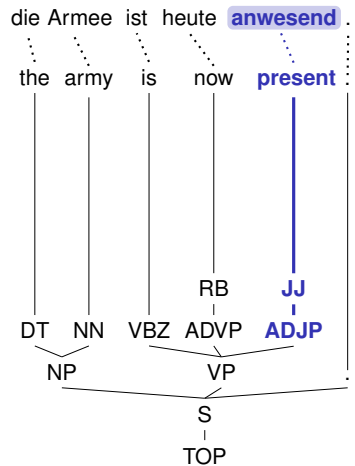
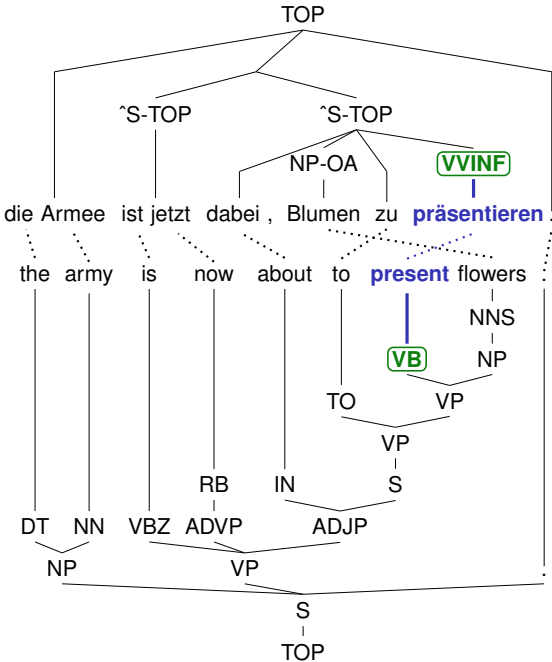


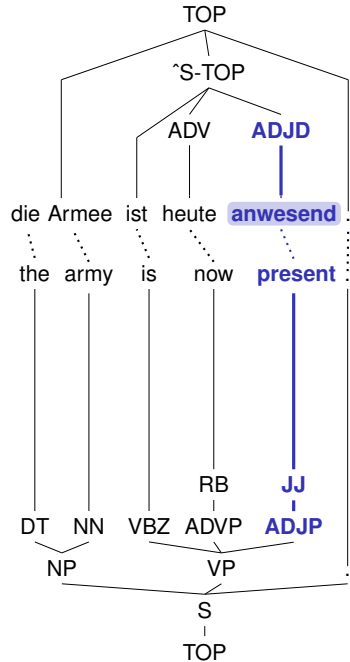
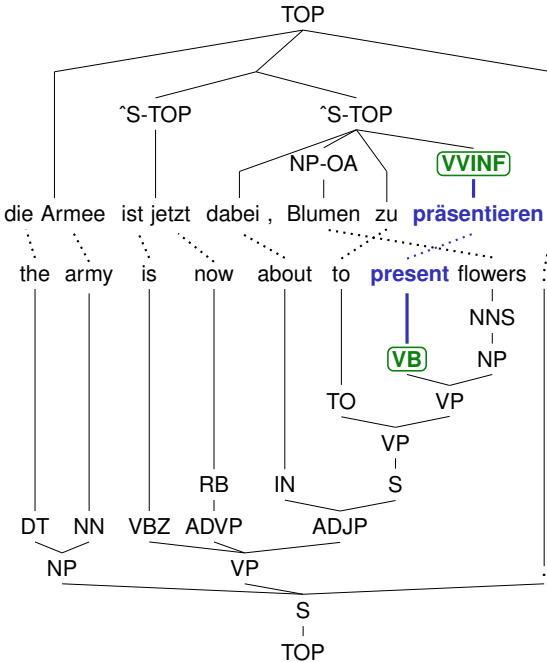


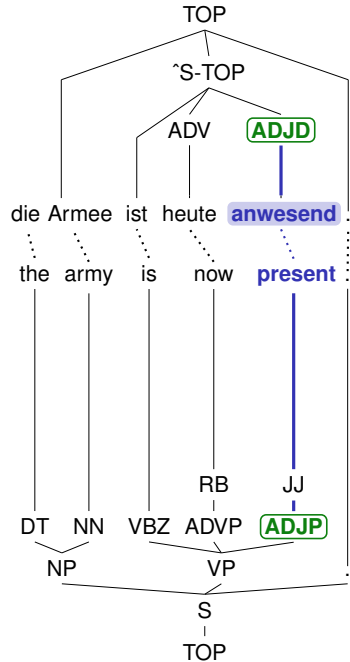
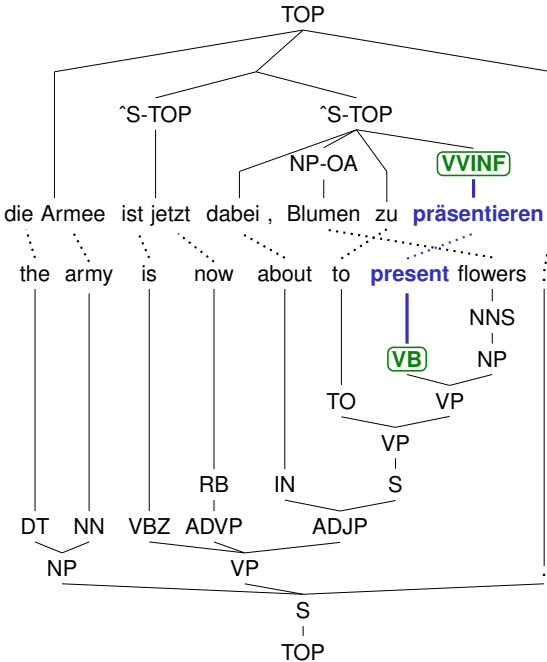












Motivation

Feature-based integration of syntactic information into GHKM string-to-tree statistical machine translation

- The hard target-side syntactic constraints that are imposed by the target non-terminal labels might be too restrictive. Should we soften them?

Preference grammars promote syntactic well-formedness on the target language side while also allowing for derivations that are not linguistically motivated (as in hierarchical translation)

- Tree-to-tree translation often underperforms. How can we effectively enhance a strong string-to-tree baseline with source-side syntactic information?

Soft syntactic constraints augment the system with additional source-side syntax features while not modifying the set of string-to-tree translation rules or the baseline feature scores

Preference Grammars

Training:

- Target-side non-terminals not decorated with syntactic labels, but with a **single generic non-terminal symbol**
- Extracted rules which differ only with respect to their non-terminal labels are collapsed to a single entry in the rule table, and their rule counts are pooled

<p>baseline</p> <p>$X, ADJ.D \rightarrow (\text{present}, \text{anwesend})$</p> <p>$X, ADV \rightarrow (\text{present}, \text{anwesend})$</p> <p>$X, AP-PD \rightarrow (\text{present}, \text{anwesend})$</p> <p>...</p>	<p>preference grammar system</p> <p>$X, X \rightarrow (\text{present}, \text{anwesend})$</p>
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- Distribution of implicit target label vectors** stored as additional information with each translation rule

$X, X \rightarrow (\text{present}, \text{anwesend}) \# (ADJ.D) 0.98 (ADV) 0.001 (AP-PD) 0.01 \dots$

Decoding:

- Computation of a **tree-wellformedness feature**

Soft Source Syntactic Constraints

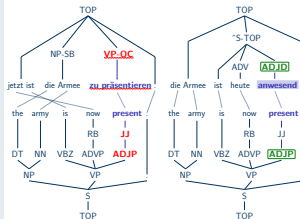
Training:

- Provide **syntactic parses of the source side** of the data
- GHKM extractor collects the **source syntactic labels that cover the source-side span of non-terminals**
- Sets of source syntactic label vectors are memorized with the rules as an **additional property**

Decoding:

- Input data parsed in a preprocessing step
- Computation of three dense **features which score matches and mismatches of input labels** and source label vectors that are associated with translation rules

rule	source label vectors
$X, VP-OC \rightarrow (\text{present}, \text{zu pr\u00e4sentieren})$	(VB), (NV)
$X, ADJ.D \rightarrow (\text{present}, \text{anwesend})$	(ADJP), (ADVP), ...
$X, TOP \rightarrow (X^* \text{ is now } X^{*-1} \dots \text{ jetzt ist } NP-SB \rightarrow VP-OC^{*-1})$	(TOP, NP, ADJP), ...
$X, S-TOP \rightarrow (\text{is } X^* \rightarrow X^{*-1} \dots \text{ ist } ADV^* \rightarrow ADJ.D^{*-1})$	(VP, ADVP, ADJP), ...



Dense features:

- a **source syntactic label vector** fully matches the input labels
- left-hand side non-terminal label mismatch**
- number of right-hand side non-terminals label mismatches**

Experimental Setup

- English→German WMT task (4.5M sentence pairs)
- Syntactic annotation: BitPar for German, Berkeley Parser for English
- Right binarization of target parse trees
- SAMT-style composite labels on source side
- Singleton hierarchical rules are discarded
- No more than 50 most frequent label vectors per rule stored
- Decoding with CYK+ and cube pruning
- Tuning with batch MIRA
- Development set: 2000 selected sentences from newstest2008-2012

Experimental Results (English→German)

system	dev		newstest2013		newstest2014	
	BLEU	TER	BLEU	TER	BLEU	TER
GHKM string-to-tree baseline	34.7	47.3	20.0	63.3	19.4	65.6
+ hard source syntactic constraints	34.6	47.4	19.9	63.4	19.4	65.6
+ soft source syntactic constraints	35.1	47.0	20.3	62.7	19.7	64.9
string-to-string (GHKM syntax-directed extraction)	33.8	48.0	19.3	63.8	18.7	66.2
+ preference grammar	33.9	47.7	19.3	63.7	18.8	66.0
+ soft source syntactic constraints	34.6	47.0	19.8	62.9	19.5	65.2

Sparse Features for Soft Syntactic Constraints

- Large number of binary features which depend on the label identity
- Separate weight tuned for each of them
- Optionally: Restrict the number of sparse features by specifying a core set of labels
- core = non-composite** – plain constituent labels as given by the syntactic parser (no SAMT-style composite labels)
- core = dev-min-occ100** – labels in the input data on the development set with minimum occurrence count threshold 100

system (tuned on newstest2012)	newstest2012		newstest2013		newstest2014	
	BLEU	TER	BLEU	TER	BLEU	TER
GHKM string-to-tree baseline	17.9	65.7	19.9	63.2	19.4	65.3
+ soft source syntactic constraints	18.2	65.3	20.3	62.6	19.7	64.7
+ sparse features	18.6	64.9	20.4	62.5	19.8	64.7
+ sparse features (core = non-composite)	18.4	65.1	20.3	62.7	19.8	64.7
+ sparse features (core = dev-min-occ100)	18.4	64.8	20.6	62.2	19.9	64.4

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