Multi Bottom-up Tree Transducers

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Multi Bottom-up Tree Transducers

Machine translation

Translation

• Input:

Official forecasts predicted just 3 percent, Bloomberg said.

• We:

die die offiziellen prognosen nur 3 prozent prognostizierten hat bloomberg gesagt.

• Google:

Offizielle Prognosen vorhergesagt nur 3 Prozent, sagte Bloomberg.



Machine translation

Translation

• Input:

The ECB wants to hold inflation to under two percent, or somewhere in that vicinity.

• We:

die ezb will unter zwei inflation prozent oder halten irgendwo damit benachbarten gebieten,.

• Google:

Die EZB will die Inflation auf unter zwei Prozent zu halten, oder irgendwo in der Nähe.



Syntax-based machine translation

Syntax-based systems





What do we have?

Input

- parallel text (English and German)
- here: EUROPARL

Example

- "We must bear in mind the Community as a whole."
- "Wir müssen uns davor hüten, alles vergemeinschaften zu wollen."

Alignments by GIZA++ [OCH, NEY '03]:



Parsing



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Better example

Example

- "We can help countries catch up, but not by putting their neighbours on hold."
- "Wir können Ländern beim Aufholen helfen, aber nicht, indem wir ihre Nachbarn in den Wartesaal schicken."

Alignments by GIZA++:





Better example



9

Small example

Input

Yugoslav President Voislav signed for Serbia. و تولى التوقيع عن صربيا الرئيس اليوغوسلافي فويسلاف <u>Transliteration:</u> w twlY AltwqyE En SrbyA Alr}ys AlywgwslAfy fwyslAf.

And then the matter was decided, and everything was put in place. ف كان ان تم الحسم و وضعت الأمور في نصاب ها

<u>Transliteration:</u> f kAn An tm AlHsm w wDEt Al>mwr fy nSAb hA.

Below are the male and female winners in the different categories. و هنا الأوائل و الأوليات في مختلف الفئات Transliteration: w hnA Al>wA}I w Al>wlyAt fy mxtlf Alf}At.



Small example

Alignment





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Multi Bottom-up Tree Transducers



Rule extraction



Select next node bottom-up

- Identify maximal subtree of aligned nodes
- Identify subtree of nodes aligned to aligned nodes, etc.
- Extract rule and leave state



Rule extraction



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 Yugoslav ^q/_Y AlywgwslAfy





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- Select next node bottom-up
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- Repeat $\text{NML}(q_Y, q_P) \xrightarrow{q_{\text{NML}}} \text{NP}(q_P, q_Y)$





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• Repeat $NML(q_Y, q_P) \xrightarrow{q_{NML}} NP(q_P, q_Y)$ $NP(q_S) \xrightarrow{q_{NP}} NP(q_S)$ $PP(q_f, q_{NP}) \xrightarrow{q_{PP}} PP(q_f, q_{NP})$



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Extended top-down tree transducer

Advantages

- simple and natural model
- easy to train (from linguistic resources) [GRAEHL et al. '08]
- ✓ symmetric

Implementation • ТІВИКОN [МАҮ, КNIGHT '06]



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Implementation

• TIBURON [MAY, KNIGHT '06]



Extended top-down tree transducer

Disadvantages (also of STSG)

- × no discontinuities
- * not binarizable [AHO, ULLMAN '72; ZHANG et al. '06]
- inefficient input/output restriction [M., SATTA '10]
- Inot composable [ARNOLD, DAUCHET '82]



Roadmap











Definition	
Extended multi bottom-up tree transducer system (Q, Σ, F, R)	(XMBOT)
 Q ranked alphabet 	(states)
 Σ ranked alphabet 	(input/output symbols)
• $F \subseteq Q_1$	(final states)
• <i>R</i> finite set of rules $\ell \rightarrow r$	(rules)
• linear $\ell \in \mathcal{T}_{\Sigma}(Q(X))$	
• $r \in Q(T_{\Sigma}(Y))$ with $Y = var(\ell)$	
Dofinition	

- linear if *r* is linear for all $\ell \rightarrow r \in R$
- nondeleting if $var(r) = var(\ell)$ for all $\ell \rightarrow r \in R$



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- nondeleting if $var(r) = var(\ell)$ for all $\ell \rightarrow r \in R$



Example

XMBOT $(Q, \Sigma, \{f\}, R)$

- $Q = \{q^{(2)}, f^{(1)}\}$
- $\Sigma = \{\sigma^{(2)}, a^{(1)}, b^{(1)}, e^{(0)}\}$
- R contains:



Note

It is linear and nondeleting



Example

XMBOT $(Q, \Sigma, \{f\}, R)$

- $Q = \{q^{(2)}, f^{(1)}\}$
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Semantics






Example (Derivation)





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Example (Derivation)







Example (Derivation)





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Example (Derivation)





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Example (Derivation)





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Example (Derivation)





Definition XMBOT $M = (Q, \Sigma, F, R)$

$$au_{M} = \{(t, u) \in \mathcal{T}_{\Sigma} imes \mathcal{T}_{\Sigma} \mid \exists q \in \mathcal{F} \colon t \Rightarrow^{*}_{M} q(u)\}$$



Definition XMBOT $M = (Q, \Sigma, F, R)$

$$\tau_{M} = \{(t, u) \in T_{\Sigma} \times T_{\Sigma} \mid \exists q \in F \colon t \Rightarrow^{*}_{M} q(u)\}$$

Example

It computes
$$\{(t, \swarrow t \in T_{\Sigma}) \mid t \in T_{\Sigma}\}$$

Its image is not recognizable



Restrictions

Definition XMBOT (Q, Σ, F, R) is • XBOT if $Q = Q_1$ • MBOT if $\ell \in \Sigma(Q(X))$ for all $\ell \to r \in R$



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Example



It is neither XBOT nor MBOT



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2) Extended Multi Bottom-up Tree Transducers



The Application



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Multi Bottom-up Tree Transducers

Theorem (ENGELFRIET et al. '09)

All linear XTOP can be simulated by linear XBOT

Proof.

Standard construction trading input-deletion for output-deletion see I-TOP \subseteq I-BOT by [ENGELFRIET '75]



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Theorem (ENGELFRIET et al. '09)

All XMBOT can be simulated by nondeleting XMBOT

- Guess subtrees that will be deleted
- Process them in nullary states (i.e. look-ahead)



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Theorem (ENGELFRIET et al. '09)

All XMBOT without recursive ε -rules can be simulated by MBOT

- Decompose large left-hand sides using "multi"-states
- Attach finite effect of ε-rules



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Definition

- XTOP *M* sensible if $|u| \in O(|t|)$ for all $(t, u) \in \tau_M$
- simple = linear and nondeleting

Theorem (MALETTI '12)

All sensible XTOP can be simulated by simple MBOT

- use (essentially) construction of [ENGELFRIET, MANETH '03]
- obtain finitely copying XTOP (without recursive ε-rules)
- apply [ENGELFRIET et al. '09] to obtain linear XMBOT
- previous theorems yield simple MBOT



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Corollary

All relevant XTOP can be simulated by simple XMBOT



Corollary

All relevant XTOP can be simulated by simple XMBOT





Corollary

All relevant XTOP can be simulated by simple XMBOT





Theorem Simple MBOT cannot be simulated by XTOP

Proof.

- even simple MBOT can copy (Example)
- see BOT ⊈ TOP by [ENGELFRIET '75]

Theorem (GILDEA '12)

Simple MBOT cannot be weakly simulated by simple XTOP



Theorem

Simple MBOT cannot be simulated by XTOP

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Summary

- ✓ generalize XTOP
- (b) discontinuities
- (c) binarizable
- (d) efficient input/output restriction
- (e) efficiently trainable
- (f) closed under composition

(even properly)



Discontinuities

Example (Derivation)



Discontinuities

- X state covers 1 input subtree
- ✓ state covers several output subtrees
- \rightarrow no input discontinuities
 - \rightarrow output discontinuities



Summary

- ✓ generalize XTOP
- discontinuities
- (c) binarizable
- (d) efficient input/output restriction
- (e) efficiently trainable
- (f) closed under composition

(even properly) (only output side)



Binarization

Definition XMBOT in 1-symbol normal form if exactly 1 (input/output) symbol occurs per rule

Theorem (ENGELFRIET et al. '09)

All XMBOT can be simulated by 1-symbol normal form XMBOT



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Proof.

Original rule:

Replacement rules:










Binarization

Definition XMBOT is fully binarized if ≤ 3 states per rule (≤ 2 in left-hand side)

Theorem (M. '11) All XMBOT can be fully binarized (in linear time



Binarization

Definition

XMBOT is fully binarized if \leq 3 states per rule (\leq 2 in left-hand side)

Theorem (M. '11)

All XMBOT can be fully binarized (in linear time)



Binarization

Definition

XMBOT is fully binarized if \leq 3 states per rule (\leq 2 in left-hand side)

Theorem (M. '11)

All XMBOT can be fully binarized (in linear time)

Proof (Binarize trees and transform into 1-symbol normal form).

Original rule:







The Theory

Binarization



Comparison

STSG cannot be binarized, but people try ...

- [ZHANG et al. '06]
- [DENERO et al. '09]



Corollary

All XMBOT can be transformed (in linear time) from joint to conditional distribution

Summary

- generalize XTOP
- discontinuities
- binarizable
- (d) efficient input/output restriction
- (e) efficiently trainable
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(even properly) (only output side)



Input/output restriction

Definition

Input restriction restricts the string language of the domain of an XMBOT to a regular language





Input/output restriction

Theorem (M., SATTA '10 & M. '11) Restricting the ... by FSA A is ...

device	input	output		
linear XMBOT M	$\mathcal{O}(M \cdot A ^3)$	$\mathcal{O}(M \cdot A ^x)$		
simple XTOP M	$\mathcal{O}(M \cdot A ^{y})$	$\mathcal{O}(\pmb{M} \cdot \pmb{A} ^{y})$		
with $x = 2 \operatorname{rk}(M) + 2$ and $y = 2 \operatorname{rk}(M) + 5$				



- ✓ generalize XTOP
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- 🗸 binarizable
- efficient input/output restriction
- (e) efficiently trainable
- (f) closed under composition

(even properly) (only output side)

(less efficient for output)



A top-down variant



[M. '11]



















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Multi Bottom-up Tree Transducers











EM training

Theorem

Derivations of XMBOT are regular (even in the weighted case)

Conclusion

program of [GRAEHL et al '08] works

- given translation pair (s_1, s_2)
- input- and output restrict to s₁ and s₂
- build derivations
- compute relative "usefulness" of each rule
- move to the next training sentence (and start anew)



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- efficiently trainable
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(even properly) (only output side)

(less efficient for output) (messy for permissive MBOT)



Composition of STSG



Conclusion

STSGs are not composable!



The Theory

Composition of STSG



Conclusion

STSGs are not composable!



The Theory

Composition of STSG



Conclusion

STSGs are not composable!



Composition of XTOP

restrictions	closed?	level of closure
simple, non-erasing, ε -free simple, non-erasing simple, ε -free simple	× × × ×	2 ∞ ∞
linear linear with regular look-ahead	× ×	≥ 2 ≥ 2
general	×	∞



Composition of XMBOT

restrictions	closed?	level of closure
simple	1	1
linear	×	1
general	×	∞ (?)



Definition

XMBOT $M = (Q, \Sigma, F, R)$ and $N = (Q', \Sigma, G, R')$ in 1-symbol normal form

$$M$$
; $N = (Q(Q'), \Sigma, F(G), R'')$

with

- Input-consuming rules from input-consuming rules of R
- 2 ε -rules from ε -rules of R'
- **③** ε -rules from ε -rule of *R* followed by input consuming rule of *R'*



Example

(1) Input-consuming rule of R and resulting rule:





Example

(2) ε -rule of R' and resulting rule:





Example

(3) ε -rule of *R* and input-consuming of *R'* and resulting rule:





Theorem (ENGELFRIET et al. '09)

The standard bottom-up tree transducer composition results hold

Summary

- ✓ generalize XTOP
- discontinuities
- 🗸 binarizable
- efficient input/output restriction
- efficiently trainable
- closed under composition

(even properly) (only output side)

(less efficient for output) (messy for permissive MBOT) (standard bottom-up results)



- ✓ generalize XTOP
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- ✓ generalize XTOP
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- ✓ preserve regularity backward



- ✓ generalize XTOP
- discontinuities
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- efficient input/output restriction
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- preserve regularity backward
- × preserve regularity forward
- × symmetric



Overview



2 Extended Multi Bottom-up Tree Transducers







XMBOT in machine translation

Moses [KOEHN et al. '07]

- framework for statistical MT
- implementations for many standard tasks (alignment, lexical scores, language model, BLEU scoring)
- supports syntax-based MT

We added

- XMBOT rule support
- XMBOT chart decoder
- adjusted language model calls



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(still broken)
















Multi Bottom-up Tree Transducers



S(NP,VP) ||| VP(VP,VP,NP) ||| S ||| VP ||| 0-2 1-0 1-1 ||| ...





S(NP,VP) ||| VP(VP,VP,NP) ||| S ||| VP ||| 0-2 1-0 1-1 ||| ...



VP(VBD(signed), PP) ||| PV(twlY) || NP-OBJ(NP(DET-NN(AltwqyE)), PP) |||

VP ||| PV NP-OBJ ||| || 0-0 ||| ...



Multi Bottom-up Tree Transducers

XMBOT decoder

FABIENNE BRAUNE

- CYK-like chart parser
- only forward application (backward planned)
- supports all standard features
- integrated cube pruning with language model

Notes

- reasonably fast
- generated the examples in Motivation (with only translation weights)
- we are still working on it



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XMBOT external tools

NINA SEEMANN

- rule extraction
- input/output restriction
- EM training
- conversion tools, pipeline scripts, ...

Notes

- in PYTHON
- computationally quite expensive
- variants for reduced POS-tags

(not inside MOSES)



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No BLEU score yet



No BLEU score yet, but we are close!



References

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