

MORPHOLOGY, SYNTAX AND WHAT'S IN BETWEEN

Towards Paradigmatic Syntax

Reut Tsarfaty
The Department of Linguistics and Philology
Uppsala University

MACHINE TRANSLATION AND MORPHOLOGICALLY RICH LANGUAGES
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Syntax, MT and MRLs

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Some Brief Historical Overview

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Some Brief Historical Overview

- ▶ What Kevin said

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Some Brief Historical Overview

- ▶ What Kevin said
- ▶ What Yoav said
- ▶ What all linguists say all the time

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~> Morphology and Syntax Interact!

Syntax, MT and MRLs

Some Brief Historical Overview

- ▶ What Kevin said
- ▶ What Yoav said
- ▶ What all linguists say all the time

~> Morphology and Syntax Interact!

No, pre/post-processing won't work

Today

An Answer to Kevin (Sunday):

How to model a morphosyntactic tree

A Question to Jason (Thursday):

How to learn bilingual paradigms

Today

The Task:

Statistical Parsing

The Challenge:

Complex Form-Function Correspondence

The Method:

Following the footsteps of Morphology

The Proposal:

A Relational-Realizational Approach

⇒ A Stepping Stone

Towards computational typology

Part 1: The Task

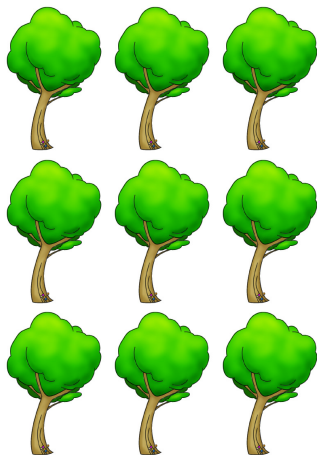
Statistical Parsing

Statistical Parsing

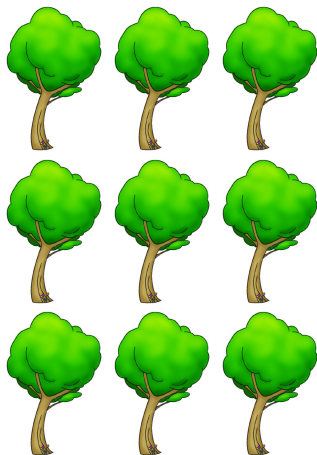
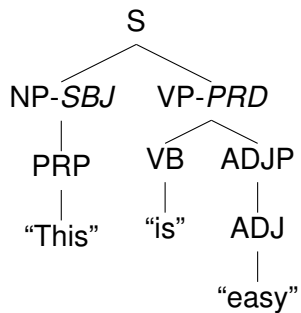
"This is easy"

Statistical Parsing

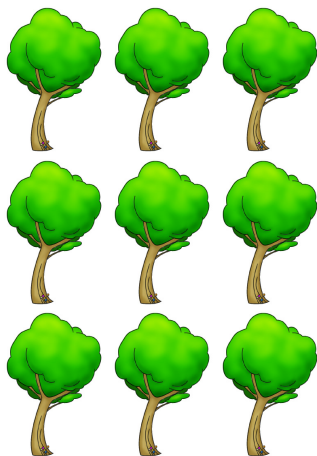
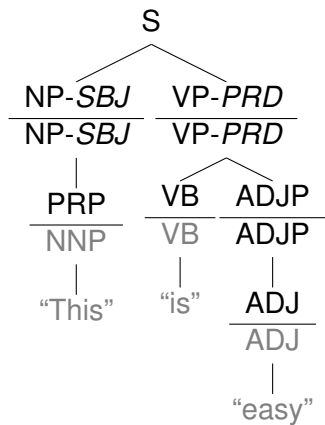
"This is easy"



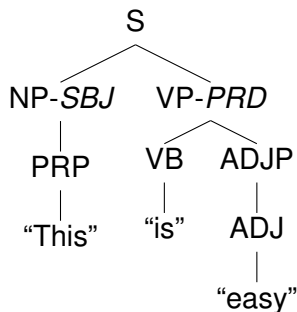
Statistical Parsing



Statistical Parsing



Works Great for English



Model	Study	F-Score
<i>Treebank Grammar</i>	Charniak 1996	75
<i>Head-Driven</i>	Collins 1997	88.6
<i>Discriminative Reranking</i>	Collins 2000	89.7
<i>Discriminative-Reranking</i>	Johnson & Charniak 2005	91.0
<i>Self-Training</i>	McClosky 2006	92.1
<i>State-Splits</i>	Petrov et al 2007	90.1
<i>Forest Reranking</i>	Liang Huang 2008	91.7

Not As Well For Other Languages

And what about this?

將水煮開後才
使用。

And this?

إغلي الماء قبل استعماله

And this?

יש להרתיח את המים
לפני השימוש.

And? ...

Language	Parser	F-Score
<i>German</i>	Rafferty & Manning 2008	79.2
<i>Czech</i>	Collins et al. 1999	79.3
<i>Chinese</i>	Levy & Manning 2003	78.8
<i>Arabic</i>	Maamouri, Bies & Kulick 2008	78.1
<i>Hebrew</i>	Tsarfaty & Sima'an 2007	74.4

Not So Well For MRLs

Why?

Homework Assignment

Read:

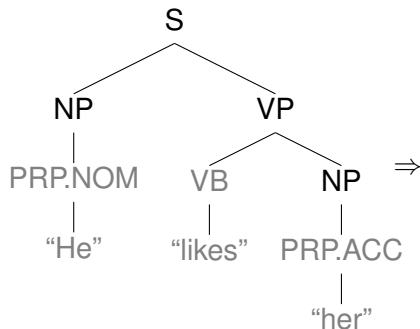
- ▶ Tsarfaty et al. “*Statistical Parsing From Morphologically Rich Languages: What, How and Whither*” Proceedings of the First Workshop on Statistical Parsing of MRLs 2010

Answer:

- ▶ What are the three main challenges in parsing of MRLs?
- ▶ How are these going to affect your MT system?

English Pronouns

English Pronouns

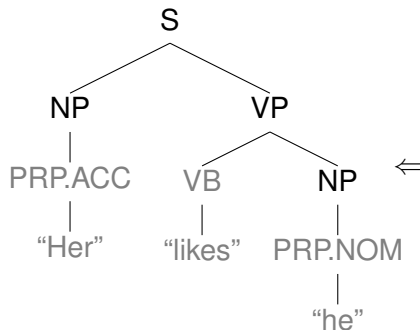


⇒

$P(\text{NP VP} \text{S})$	1
$P(\text{PRP.NOM} \text{NP})$	0.5
$P(\text{PRP.ACC} \text{NP})$	0.5
$P(\text{VB NP} \text{VP})$	1
<hr/>	
$P(\text{"He"} \text{PRP.NOM})$	1
$P(\text{"likes"} \text{VB})$	1
$P(\text{"her"} \text{PRP.ACC})$	1

$$P(\text{"He likes her"}) = P(\text{NP VP}|\text{S}) \times \dots \times P(\text{"her"}|\text{PRP.ACC}) = 0.25$$

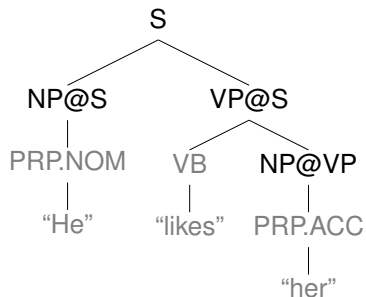
English Pronouns



$P(\text{NP VP} \text{S})$	1
$P(\text{PRP.NOM} \text{NP})$	0.5
$P(\text{PRP.ACC} \text{NP})$	0.5
$P(\text{VB NP} \text{VP})$	1
$P(\text{"He"} \text{PRP.NOM})$	1
$P(\text{"likes"} \text{VB})$	1
$P(\text{"her"} \text{PRP.ACC})$	1

$$P(\text{"Her likes he"}) = P(\text{NP VP}|\text{S}) \times \dots \times P(\text{"her"}|\text{PRP.ACC}) = 0.25$$

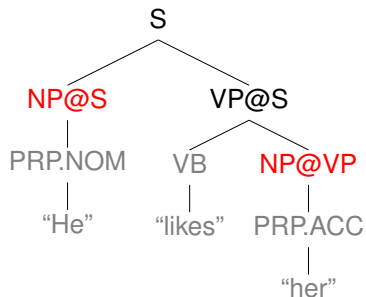
Example 1: Parent Encoding (Johnson 1998)



⇒

P(NP@S VP@S S)	1
P(PRP.NOM NP@S)	1
P(PRP.ACC NP@VP)	1
P(VB NP@VP VP@S)	1
<hr/>	
P("He" PRP.NOM)	1
P("likes" VP)	1
P("her" PRP.ACC)	1

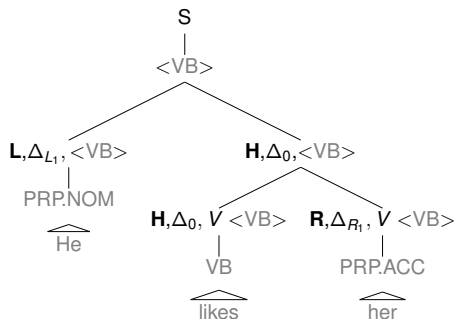
Example 1: Parent Encoding (Johnson 1998)



⇒

P(NP@S VP@S S)	1
P(PRP.NOM NP@S)	1
P(PRP.ACC NP@VP)	1
P(VB NP@VP VP@S)	1
<hr/>	
P("He" PRP.NOM)	1
P("likes" VP)	1
P("her" PRP.ACC)	1

Example 2: Head-Driven Processes (Collins 1999)



⇒

$P(\langle \text{VB} \rangle | \text{S})$ 1

$P(\mathbf{L}\Delta_{L_1}, \mathbf{H}\Delta_0 | \langle \text{VB} \rangle, \text{S})$ 1

$P(\text{PRP.NOM} | \mathbf{L}, \Delta_{L_1}, \langle \text{VB} \rangle, \text{S})$ 1

$P(\text{VP} | \mathbf{H}, \Delta_0, \langle \text{VB} \rangle, \text{S})$ 1

$P(\langle \text{VB} \rangle | \text{VP})$ 1

$P(\text{PRP.ACC} | \mathbf{R}, \Delta_{R_1}, \langle \text{VB} \rangle, \text{S})$ 1

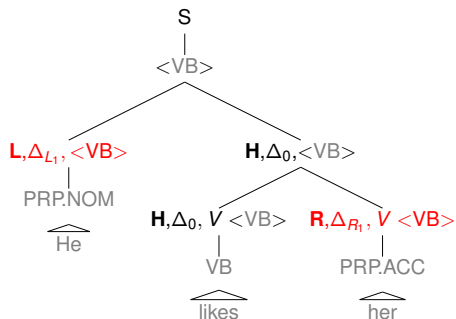
$P(\text{VB} | \mathbf{H}, \Delta_0, \langle \text{VB} \rangle, \text{S})$ 1

$P(\text{"He"} | \text{PRP.NOM})$ 1

$P(\text{"likes"} | \text{VB})$ 1

$P(\text{"her"} | \text{PRP.ACC})$ 1

Example 2: Head-Driven Processes (Collins 1999)

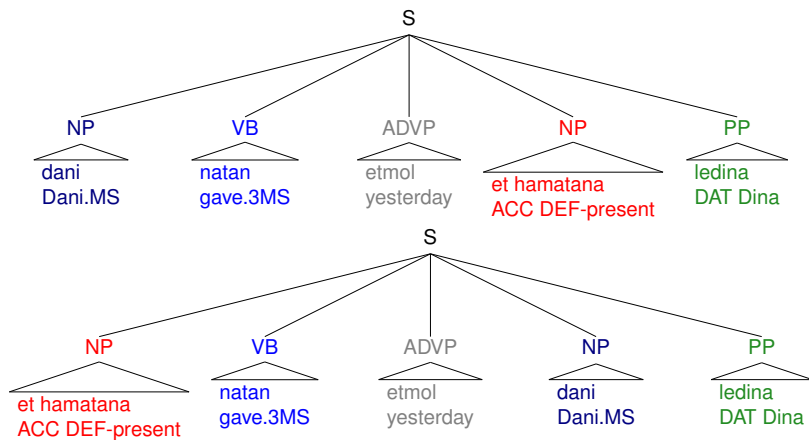


⇒

$P(\langle VB \rangle S)$	1
$P(\mathbf{L}\Delta_{L_1}, \mathbf{H}\Delta_0 \langle VB \rangle, S)$	1
$P(\text{PRP.NOM} \mathbf{L}, \Delta_{L_1}, \langle VB \rangle, S)$	1
$P(\text{VP} \mathbf{H}, \Delta_0, \langle VB \rangle, S)$	1
$P(\langle VB \rangle \text{VP})$	1
$P(\text{PRP.ACC} \mathbf{R}, \Delta_{R_1}, \langle VB \rangle, S)$	1
$P(\text{VB} \mathbf{H}, \Delta_0, \langle VB \rangle, S)$	1
<hr/>	
$P(\text{"He"} \text{PRP.NOM})$	1
$P(\text{"likes"} \text{VB})$	1
$P(\text{"her"} \text{PRP.ACC})$	1

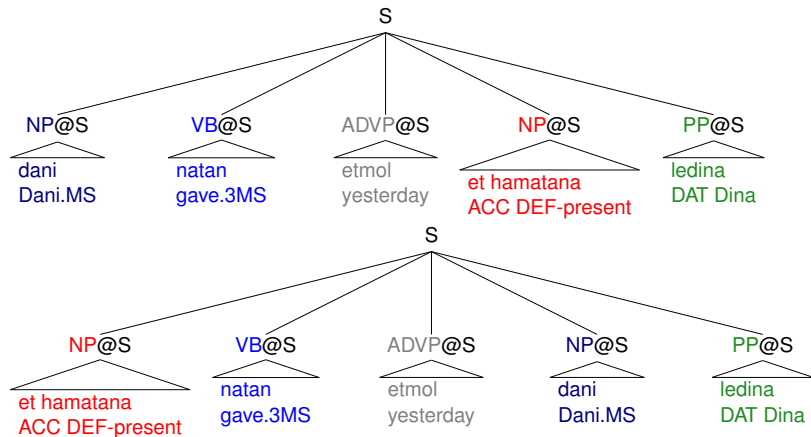
What if we take a different treebank

The Modern Hebrew Treebank



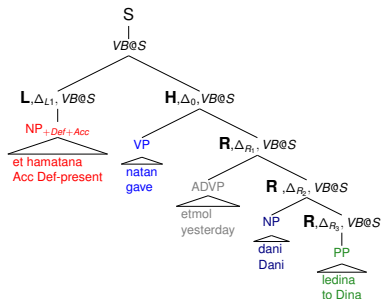
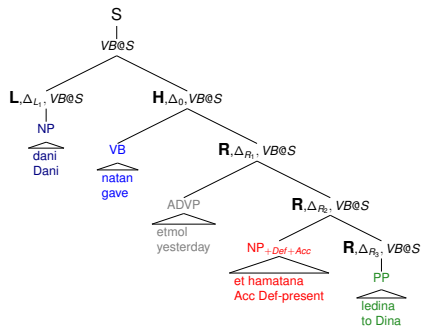
Example 1:

Parent Encoding



Example 2:

The Head-Driven (HD) Approach



So What Is Going On?

An Observation

- ▶ Configurational languages:
 \rightsquigarrow Tree configurations approximate grammatical functions
- ▶ Less-configurational languages:
 \rightsquigarrow Not.

So What Is Going On?

An Observation

- ▶ Configurational languages:
 - ↪ Tree configurations approximate grammatical functions
- ▶ Less-configurational languages:
 - ↪ Not.
 - ↪ No regularities to statistically learn from data

A Question

What kind of form-function correspondence patterns
our parser needs to learn from the data?

Part 2: The Challenge

Modeling Form-Function Correspondence

The Data

The Data

Typological Dimensions of Variation



Basic Word-Order Typology

(Greenberg 1966, Mithun 1992)



Morphological Typology

(Sapir 1921, Greenberg 1954)



Nonconfigurationality

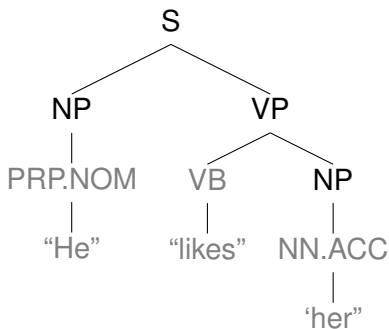
(Hale 1983, Austin and Bresnan 1996)

Nonconfigurationality as Misalignment

Predicate-Argument Relations

'SBJ' did 'PRD' to 'OBJ'

Syntactic Configuration

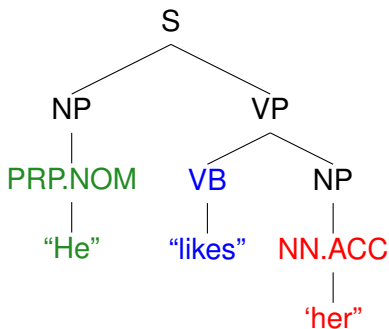


Nonconfigurationality as Misalignment

Predicate-Argument Relations

'**SBJ**' did '**PRD**' to '**OBJ**'

Configurational Languages

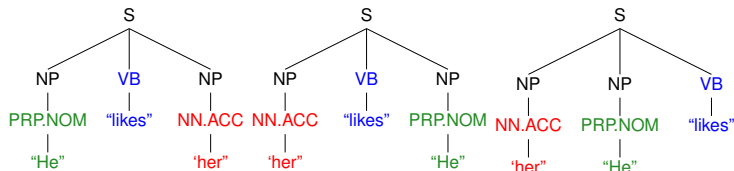


Nonconfigurationality as Misalignment

Predicate-Argument Relations

'**SBJ**' did '**PRD**' to '**OBJ**'

Less-Configurational Languages



Morphosyntactic Exponence in Hebrew



Word-Order

- (1) a. dani natan et hamatana ledina
Dani gave ACC the-present to-Dina
“Dani gave the present to Dina” (SVO)
- b. et hamatana natan dani ledina
ACC the-present gave Dani to-Dina
“Dani gave the present to Dina” (OVS)
- c. natan dani et hamatana ledina
gave Dani ACC the-present to-Dina
“Dani gave the present to Dina” (VSO)
- d. ledina natan dani et hamatana
to-dina gave Dani ACC the-present
“Dani gave the present to Dina” (VSO)

Exponence Relations in Hebrew (1:1)



Case-Assigning Prepositions

- (2) a. dani natan et hamatana ledina
Dani gave ACC DEF-present DAT-Dina
- b. et hamatana natan dani ledina
ACC DEF-present gave Dani DAT-Dina
- c. natan dani et hamatana ledina
gave Dani ACC DEF-present DAT-Dina
- d. ledina natan dani et hamatana
DAT-dina gave Dani ACC DEF-present

Exponence Relations in Hebrew (1:many)



Differential Object-Marking

- (3) a. dani natan et hamatana ledina
Dani gave ACC DEF-present to-Dina
- b. et hamatana natan dani ledina
ACC DEF-present gave Dani to-Dina
- c. natan dani et hamatana ledina
gave Dani ACC DEF-present to-Dina
- d. ledina natan dani et hamatana
to-dina gave Dani ACC DEF-present

Exponence Relations in Hebrew (1:many)



Feature Spreading (Danon, 2007)

- (4) a. dani natan [et matnat yom **hahuledet**] ledina
Dani gave [ACC present day **DEF-birth**] to-Dina
- b. [et matnat yom **hahuledet**] natan dani ledina
[ACC present day **DEF-birth**] gave Dani to-Dina
- c. natan dani [et matnat yom **hahuledet**] ledina
gave Dani [ACC present day **DEF-birth**] to-Dina
- d. ledina natan dani [et matnat yom **hahuledet**]
to-dina gave Dani [ACC present day **DEF-birth**]

Exponence Relations in Hebrew (1:many)



Agreement

- (5) a. dani natan et hamatana ledina
Dani.**MS** gave.**3MS** ACC DEF-present DAT-Dina
- b. et hamatana natan dani ledina
ACC DEF-present gave.**3MS** Dani.**MS** DAT-Dina
- c. natan dani et hamatana ledina
gave.**MS** Dani.**3MS** ACC DEF-present DAT-Dina
- d. ledina natan dani et hamatana
DAT-dina gave.**3MS** Dani.**MS** ACC DEF-present

Exponence Relations in Hebrew (many:1)



Clitics and Null Anaphors

- (6) a. dani natan et hamatana ledina
Dani.MS gave.3MS ACC DEF-present DAT-Dina
“Dani gave the present to Dina”
- b. natati et hamatana ledina
gave.1S ACC DEF-present DAT-Dina
“I gave the present to Dina”
- c. natatiha ledina
gave.1S.ACC.3FS DAT-Dina
“I gave it to Dina”

Language Types and Morphosyntactic Exponence

Recap:

CONFIGURATIONAL ————— NONCONFIGURATIONAL
1:1 ————— many : many

- ▶ Exponence relations relate grammatical functions to the formal means that realize them in the syntactic structure
- ▶ Configurationality is a special case of a 1:1 mapping between grammatical functions to configurational positions

Question:

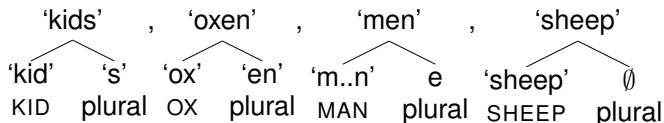
How can we model and statistically learn generally complex, many-to-many, form-function correspondence in syntax?

Part 3: The Proposal

Following the footsteps of morphology

Modeling Morphology (i): Terminology

Morpheme-Based Morphology (Bloomfield, 1933)



Morphological Exponence (Matthews 1991)

- ▶ Simple Exponence (1:1)
- ▶ Cumulative Exponence (many:1)
- ▶ Extended/Distributed Exponence (1:many)

Modeling Morphology (II): Assumptions (Stump 2001)

LEXICAL vs. INFERENCE Approaches

- ▶ LEXICAL:
morphemes are primary, properties stored in the lexicon
- ▶ INFERENCE:
properties are primary, forms are computed

INCREMENTAL vs. REALIZATIONAL Approaches

- ▶ INCREMENTAL:
morphemes/properties are accumulated incrementally
- ▶ REALIZATIONAL:
property-bundles are pre-condition for spell-out

Modeling Morphology (III): A Taxonomy

	LEXICAL	INFERENCEAL
INCREMENTAL	Item & Arrangement (Bloomfield 1933) (Lieber 1992)	Item & Processes (Hocket 1954) (Steele 1995)
REALIZATIONAL	Distributed Morphology (Halle and Marantz 1993) Lexical Phonology	(Extended) Word & Paradigm (Matthews 1972), (Anderson 1992) (Stump 2001), (Blevins 2006)

Table: A Taxonomy of Models for Morphology (Stump 2001)

The Strategy (IV): (Extended) Word-and-Paradigm

Paradigmatic Organization

/EAT/	1Sing	2Sing	3Sing	1PI	2PI	3PI
Past	1SingPast	2SingPast	3SingPast	1PIPast	2PIPast	3PIPast
Present	1SingPres	2SingPres	3SingPres	1PIPres	2PIPres	3PIPres
Perfect	1SingPerf	2SingPerf	3SingPerf	1PIPerf	2PIPerf	3PIPerf

Realization Rules

/EAT/ , /EAT/ , /EAT/ , /EAT/
+1SingPast , +3SingPast , +1SingPres , +3SingPres
| | | |
'ate' 'ate' 'eats' 'eat'

The Proposal (I): “Lifting” the Terminology

Morphological Exponence

- ▶ Simple (1:1, e.g., $PL \rightsquigarrow$ ‘s’ in ‘cats’)
- ▶ Cumulative (many:1, e.g., $3RD, SG \rightsquigarrow$ ‘s’ in ‘eats’)
- ▶ Distributed/Extended (1:many $PL \rightsquigarrow$ ‘i’, ‘ren’ in ‘children’)

The Proposal (I): “Lifting” the Terminology

Morphological Exponence : Properties \rightsquigarrow Words

- ▶ Simple (1:1, e.g., **PL** \rightsquigarrow ‘s’ in ‘cats’)
- ▶ Cumulative (many:1, e.g., **3RD,SG** \rightsquigarrow ‘s’ in ‘eats’)
- ▶ Distributed/Extended (1:many **PL** \rightsquigarrow ‘i’, ‘ren’ in ‘children’)

(Morpho)Syntactic Exponence : Relations \rightsquigarrow Positions

- ▶ Simple (1:1, e.g., **SBJ** \rightsquigarrow nominative)
- ▶ Cumulative (many:1, e.g., **PRD,OBJ** \rightsquigarrow clitics)
- ▶ Distributed/Extended (1:many, e.g., **SBJ** \rightsquigarrow agreement)

The Proposal (II): Modeling Assumptions

CONFIGURATIONAL vs. RELATIONAL Approaches

- ▶ CONFIGURATIONAL:
configurations are primary, relations are derived
- ▶ RELATIONAL:
relations are primary, configurations are computed

INCREMENTAL vs. REALIZATIONAL Approaches

- ▶ INCREMENTAL:
constructive operations,
individually/incrementally define/add relations
- ▶ REALIZATIONAL:
interpretive operations,
sets of relations are precondition to realization

The Proposal (III): A Taxonomy (Tsarfaty 2010)

	CONFIGURATIONAL	RELATIONAL
INCREMENTAL	X-Bar Theory	Dependency Grammar
REALIZATIONAL	Tree Adjoining Grammar	

Table: A Taxonomy of Generative Syntactic Frameworks

The Proposal (IV): Relational-Realizational Modeling

The *Relational* Assumption

- ▶ Paradigms organize the syntactic domain
- ▶ Cells in paradigms define sets of relations
- ▶ Sets of relations are realized in different configurations

The *Realizational* Assumption

- ▶ Sets of relations (Arg-St) are primitives
- ▶ Rules interpret sets of relations as surface forms
- ▶ Rules can refer to multiple relations and span clauses

Realization in Syntax is Recursive!

Realization of a cells refers to function cells in other paradigms

The Proposal (IV): Relational-Realizational Modeling

S(PRED)	FEATS	Affirmative	Interrogative	Imperative
ARG-ST				
intransitive		$S_{\text{affirm}+}\{\text{SBJ,PRD}\}$	$S_{\text{inter}+}\{\text{SBJ,PRD}\}$	$S_{\text{imper}+}\{\text{SBJ,PRD}\}$
transitive		$S_{\text{affirm}+}\{\text{SBJ,PRD,OBJ}\}$	$S_{\text{inter}+}\{\text{SBJ,PRD,OBJ}\}$	$S_{\text{imper}+}\{\text{SBJ,PRD,OBJ}\}$
ditransitive		$S_{\text{affirm}+}\{\text{SBJ,PRD,OBJ,COM}\}$	$S_{\text{inter}+}\{\text{SBJ,PRD,OBJ,COM}\}$	$S_{\text{imper}+}\{\text{SBJ,PRD,OBJ,COM}\}$

Figure: Paradigmatic Organization

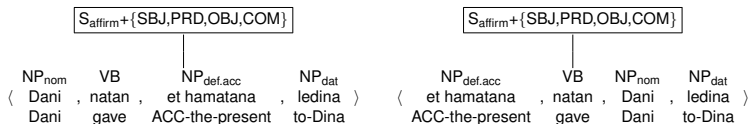
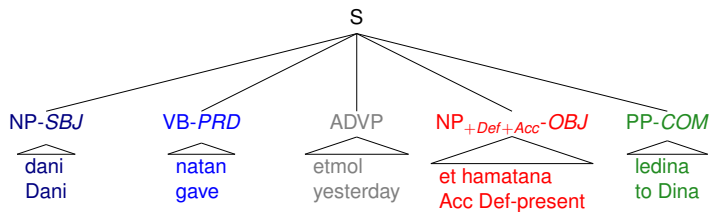


Figure: Realization Rules

Realization Rules

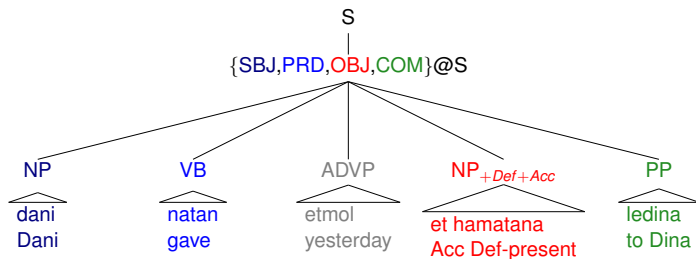
Realization Rules

A PCFG Approximation



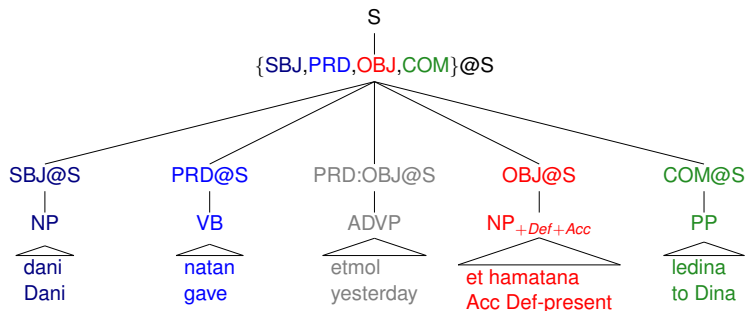
Realization Rules

Form-Function Separation

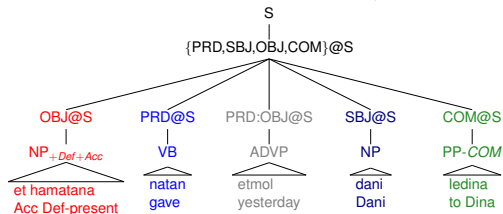
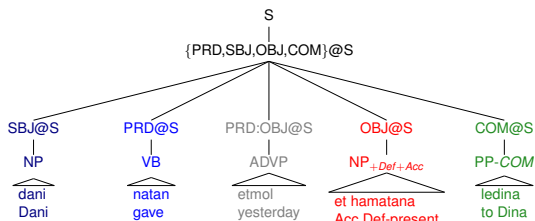


Realization Rules

Morphological and Syntactic Realization



Realization Rules: Economy and Generalization



The Generative Model

Projection:

$$P$$
$$\downarrow$$
$$\{gr_i\}_{i=1}^n @ P$$

Configuration:

$$\{gr_i\}_{i=1}^n @ P$$
$$\swarrow \quad \downarrow \quad \searrow \quad \swarrow$$
$$gr_1 @ P \quad gr_1 : gr_2 @ P \quad \dots \quad gr_n @ P$$

Realization:

$$gr_1 @ P \quad gr_1 : gr_2 @ P \quad \dots \quad gr_n @ P$$
$$\downarrow \quad \downarrow \quad \quad \downarrow$$
$$C_1 \quad \dots C_{1:2_j} \dots \quad C_n$$

The Probabilistic Model

The RR Probabilities:

$$\begin{aligned} \mathbf{P}_{\text{RR}}(r) = & \\ \text{Projection} & \mathbf{P}_{\mathbf{p}}(\{gr_i\}_{i=1}^n | P) \times \\ \text{Configuration} & \mathbf{P}_{\mathbf{c}}(\langle gr_0 : gr_1, g_1, \dots \rangle | \{gr_i\}_{i=1}^n, P) \times \\ \text{Realization} & \prod_{i=1}^n \mathbf{P}_{\mathbf{r}_1}(C_i | gr_i, P) \times \\ & \mathbf{P}_{\mathbf{r}_2}(\langle C_{0_1}, \dots, C_{0_{m_0}} \rangle | gr_0 : gr_1, P) \times \\ & \prod_{i=1}^n \mathbf{P}_{\mathbf{r}_2}(\langle C_{i_1}, \dots, C_{i_{m_i}} \rangle | gr_i : gr_{i+1}, P) \end{aligned}$$

The RR Parser:

$$\pi^* = \operatorname{argmax}_{\pi} P(\pi) = \operatorname{argmax}_{\pi} \prod_{r \in \pi} \mathbf{P}_{\text{RR}}(r)$$

Part IV: Applications

- ▶ Cross-Linguistic Parsing
- ▶ Cross-Linguistic Variation

Application I: Parsing Modern Hebrew

Data

The Modern Hebrew Treebank v2, head annotated.
6500 sentences, 500/5500/500 dev/train/test split

Models

- ▶ Grammatical Functions: PRD, SBJ, OBJ, COM, CNJ
- ▶ Morphological Splits: **PoS/Def/Acc/Gender**

Estimation

Relative Frequency + Simple Unknown Words Smoothing

Parsing

Exhaustive Viterbi Parsing (using BitPar, Schmid 2004)

Evaluation

PARSEVAL (i) Overall, and (ii) Per Category Evaluation

A Taxonomy of PCFG-based Parsers



	CONFIGURATIONAL	RELATIONAL
INCREMENTAL	Head-Driven Parsing (Collins 1999) 	
REALIZATIONAL		Relational-Realizational (Tsarfaty et al. 2009) 

Table: A Taxonomy of PCFG-Based Parsing Frameworks

Overall Results



74.66/74.35
(7385)



73.52/74.84
(21399)



76.32/76.51
(13618)

Overall Results



74.66/74.35
(7385)



73.52/74.84
(21399)



76.32/76.51
(13618)

Hebrew Parsing Results Using Gold Standard Input



	∅	83.06 (5914)	83.49 (6688)
	<i>gender</i>	82.18 (10765)	83.70 (10063)
	<i>case/def</i>	79.53 (12700)	83.66 (12386)
	<i>gender/case/def</i>	80.89 (13028)	84.13 (13618)

Swedish Parsing Results Using Gold Standard Input



	∅	78.65 (8696)	77.71 (10099)
	<i>gender</i>	73.20 (11382)	78.09 (12593)
	<i>case/def</i>	74.90 (11239)	79.09 (13912)
	<i>gender/case/def</i>	68.97 (13347)	77.89 (14991)

Application II: Cross-Linguistic Variation

Application II: Cross-Linguistic Variation

1: Apply the model to different languages, e.g.,

- ▶ **Hebrew:** a Semitic Language
- ▶ **Swedish:** a Germanic Language

2: Learn the distribution of model parameters

- ▶ RR-Projection
- ▶ RR-Configuration
- ▶ RR-Realization

3: Instantiate typological parameters for UG

Parameter 1: Basic Word-Order (Greenberg 1963)

Basic Word-Order Parameter in Hebrew:
 $P(\langle \textit{configuration} \rangle | \{\text{SBJ, PRD, OBJ}\} @ S)$

Probability	<i>Configuration</i>
35.3%	SBJ PRD OBJ □
15.6%	SBJ PRD □ OBJ □
12.3%	□ PRD SBJ OBJ □
10.3%	SBJ □ PDR OBJ □
6.5%	□ SBJ PRD OBJ □
4.1%	SBJ □ PRD □ OBJ □
3.7%	□ PRD SBJ □ OBJ □
3%	OBJ PRD SBJ □
1.7%	□ SBJ PRD □ OBJ □
1.7%	□ PRD OBJ SBJ □
1.3%	SBJ □ PRD OBJ □
1 %	□ PRD □ SBJ OBJ □

Parameter 1: Basic Word-Order (Greenberg 1963)

Basic Word-Order Parameter in Swedish:
 $P(\langle \textit{configuration} \rangle | \{\text{SBJ, PRD, OBJ}\} @ S)$

Probability	<i>Configuration</i>
35.5%	SBJ PRD OBJ □
18.9%	SBJ PRD □ OBJ □
13.9%	□ PRD SBJ PBJ □
8.1%	SBJ PRD OBJ
4.7%	□ PRD SBJ □ OBJ
3.5%	OBJ PRD SBJ
2.6%	SBJ PRD OBJ □
1.7%	OBJ PRD SBJ COM □
1.6%	PRD SBJ OBJ □
1.6%	□ PRD SBJ OBJ
1%	□ PRD SBJ □ OBJ

Parameter 2: Inflectional Systems

The Object-Marking Parameter in Hebrew:
 $P(\langle \text{morphosyntactic representation} \rangle | \text{OBJ@S})$

Probability	Realization
43.5%	NP.<NN>
14.7%	NP.DEF.ACC<NN>
8.8%	NP.<NNT>
7.4%	NP.DEF.ACC<NNP>
6.7%	NP.DEF.ACC<NN.DEF>
6.5%	NP.DEF.ACC<NNT>
5.8%	NP.DEF.ACC<PRP>

Parameter 2: Inflectional Systems

The Object-Marking Parameter in Swedish:
 $P(\langle \text{morphosyntactic representation} \rangle | \text{OBJ@S})$

Probability	Realization
46%	NP.IND.NOM
20%	NP.DEF.NOM
13.4%	S
7.3%	NP.DEF.NOM-OBJ
4.9%	VP
3.6%	NP.IND
2.8%	NP.NOM

Towards Computational Typology and Statistical UG

We can potentially use the RR parameters to...

- ▶ Quantify Intra-Language Variation
- ▶ Quantify Cross-Linguistic Variation
- ▶ Quantify Nonconfigurationality
- ▶ Learn Probabilistic P&P

Conclusion

We presented a Relational-Realizational Architecture for Specifying and Learning Morphosyntactic Descriptions

- ▶ Simple
- ▶ Formal
- ▶ Robust
- ▶ Implementable
- ▶ Interpretable
- ▶ Explanatory

Paradigms augmented with realization rules may constitute a useful and powerful modeling strategy also for **Statistical MT**

Homework Assignment II

Read:

Tsarfaty, R. *Relational-Realizational Parsing*
PhD Thesis, University of Amsterdam, 2010

Implement:

- ▶ **Transfer rules** to model *morphosyntactic variation*
- ▶ **Alignment tools** to align *paradigm cells*

Thank You!

Questions?

Thank You!

Questions?

Now a postdoc at Uppsala, looking for a tenure track position :)

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