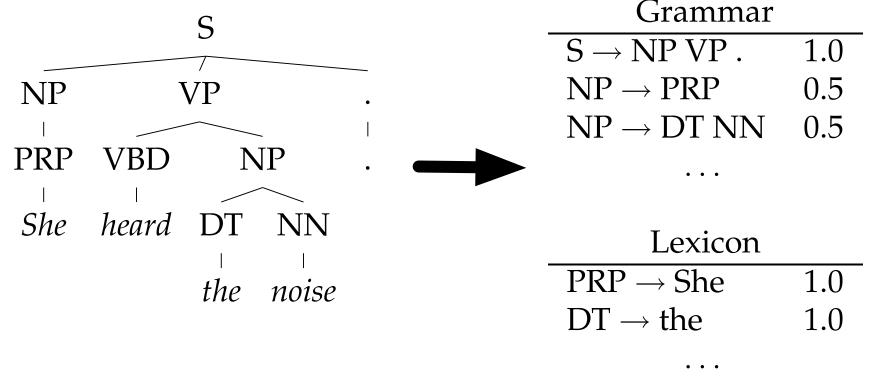
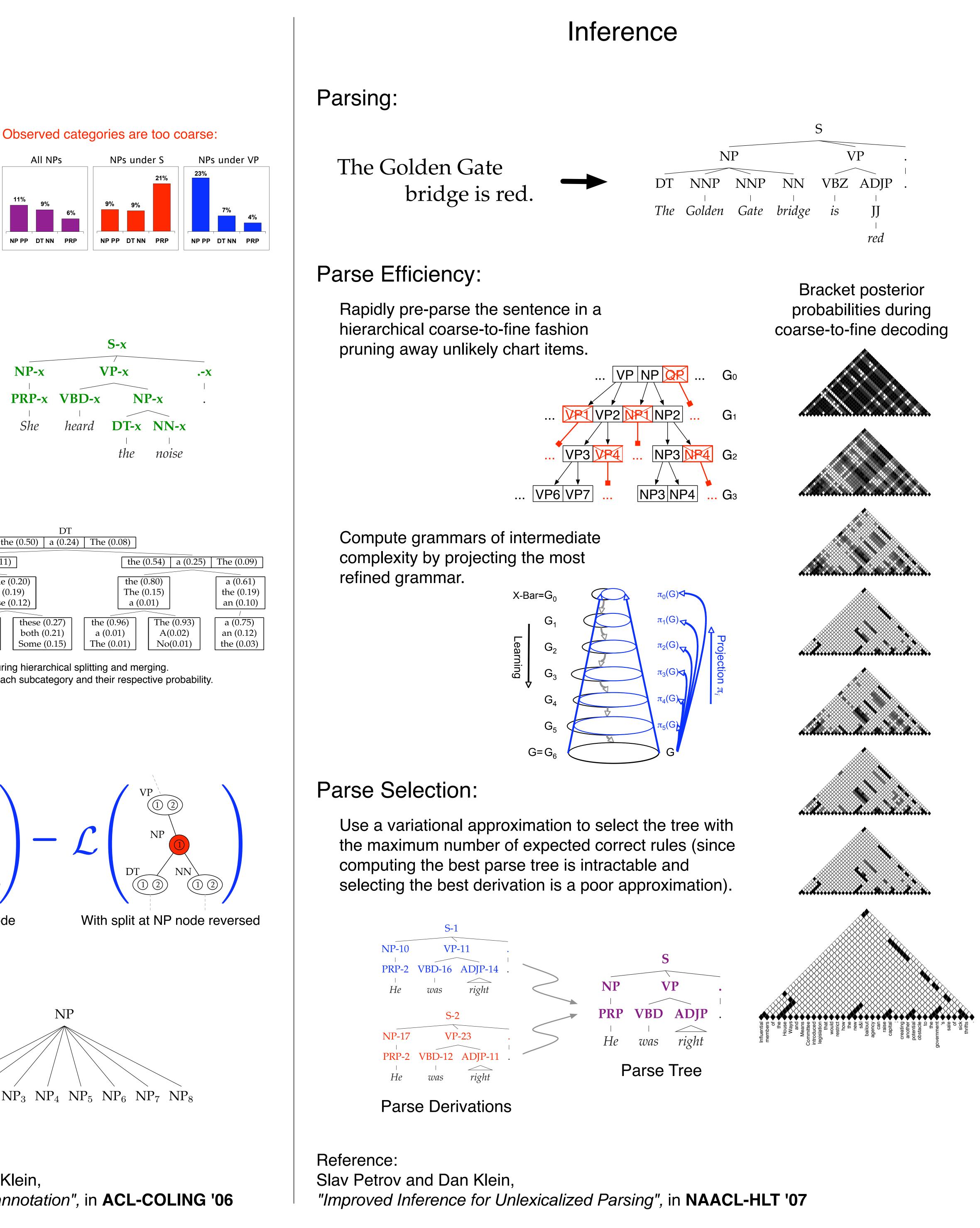


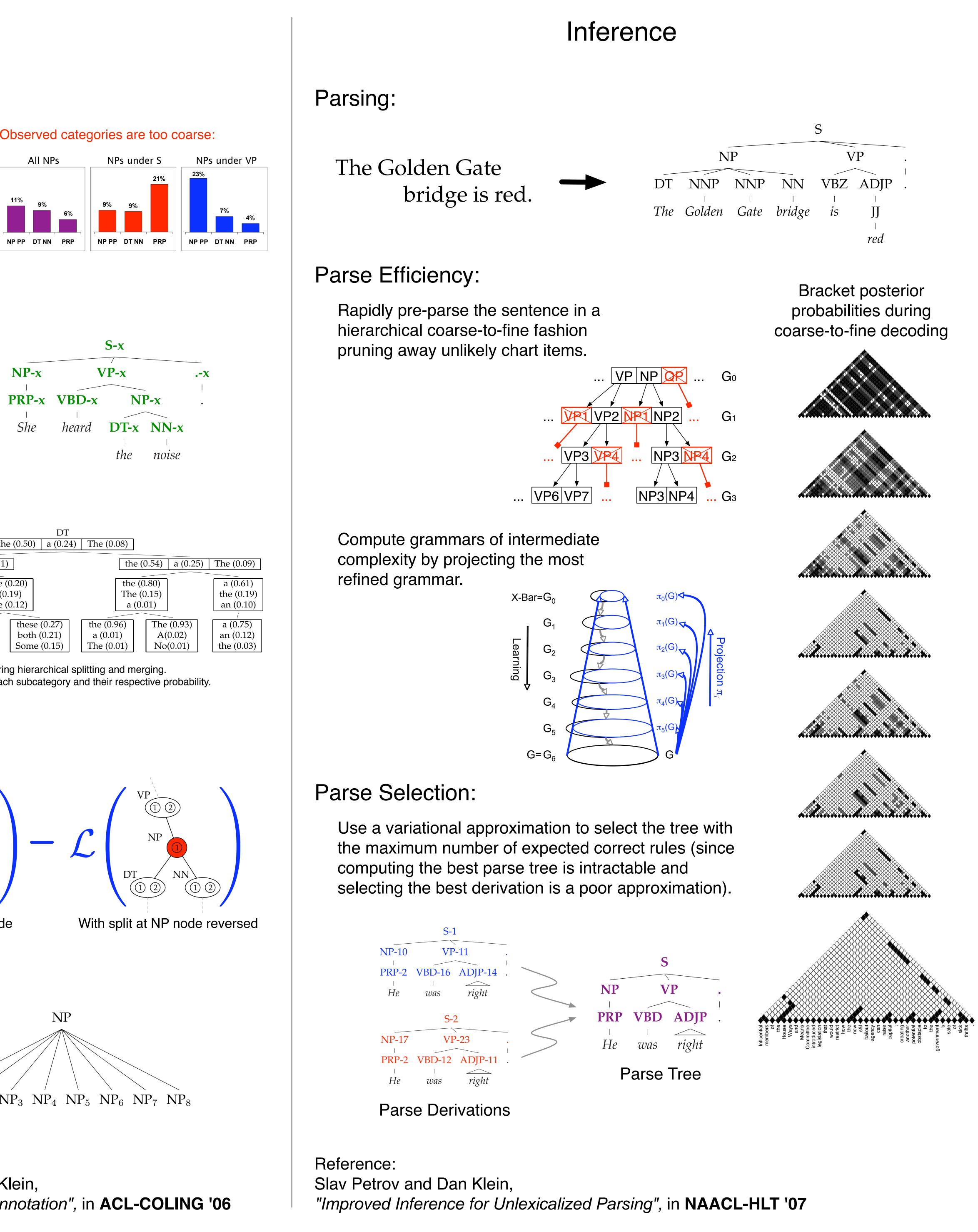


Grammar Extraction:

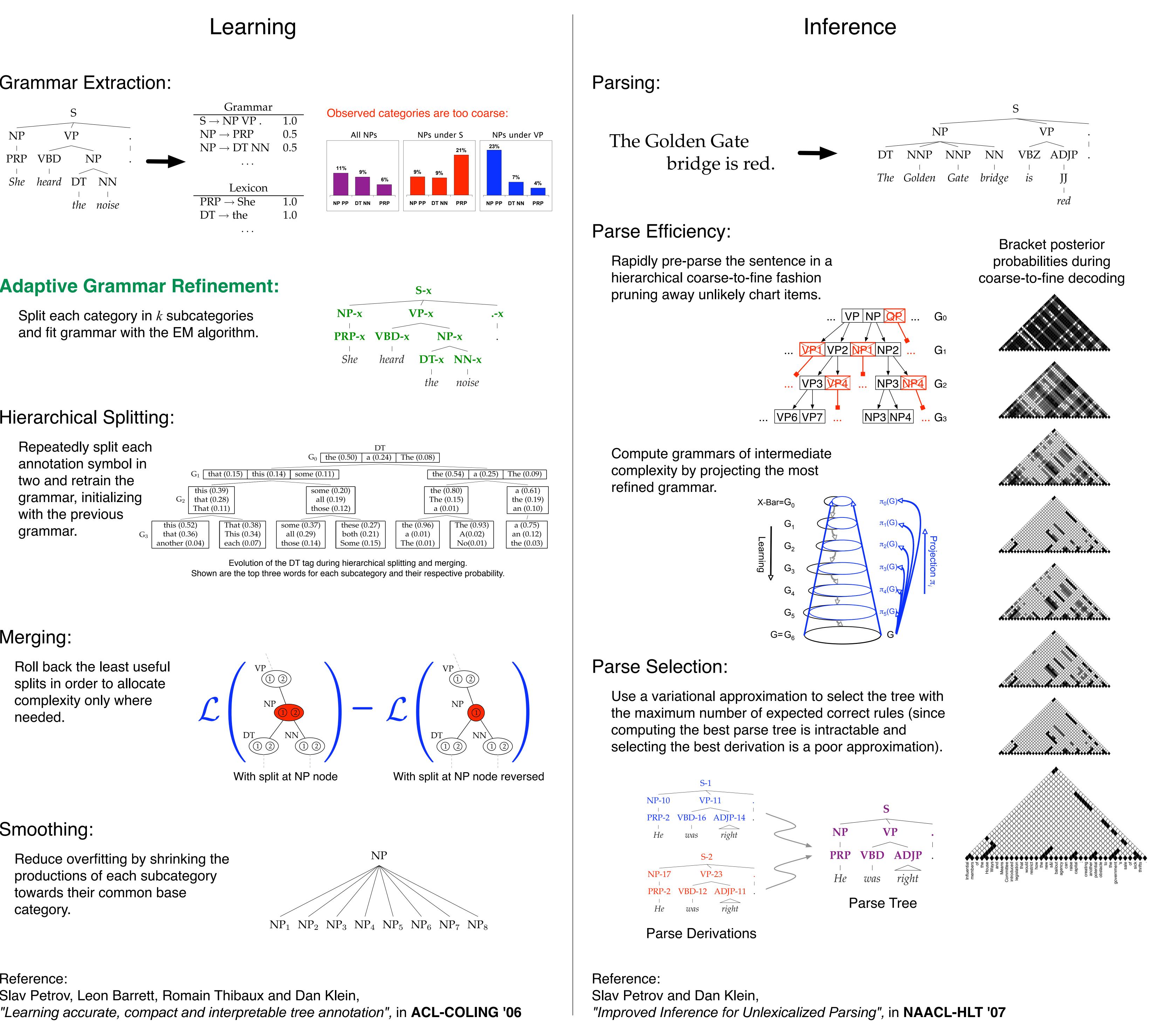




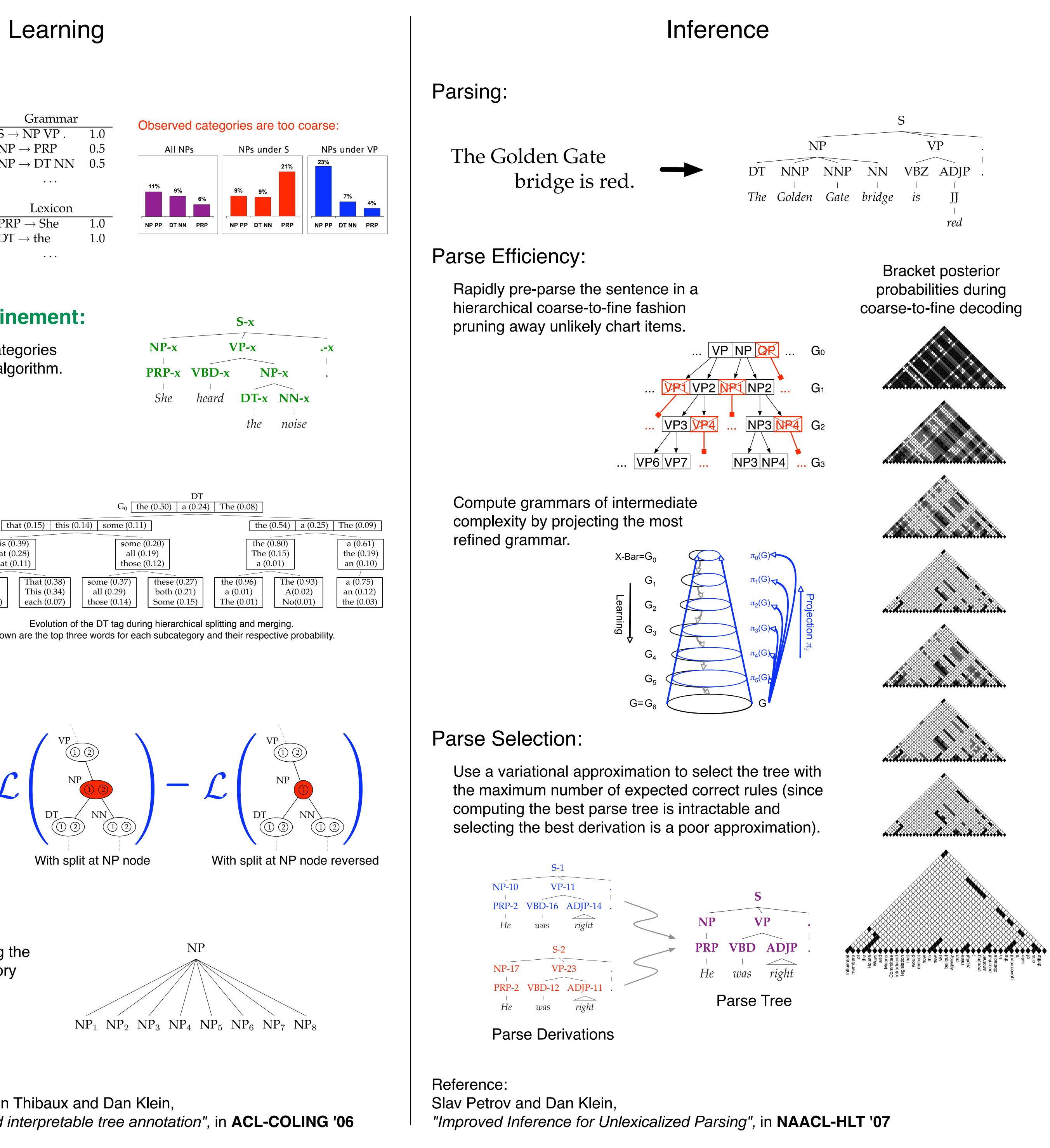
Adaptive Grammar Refinement:



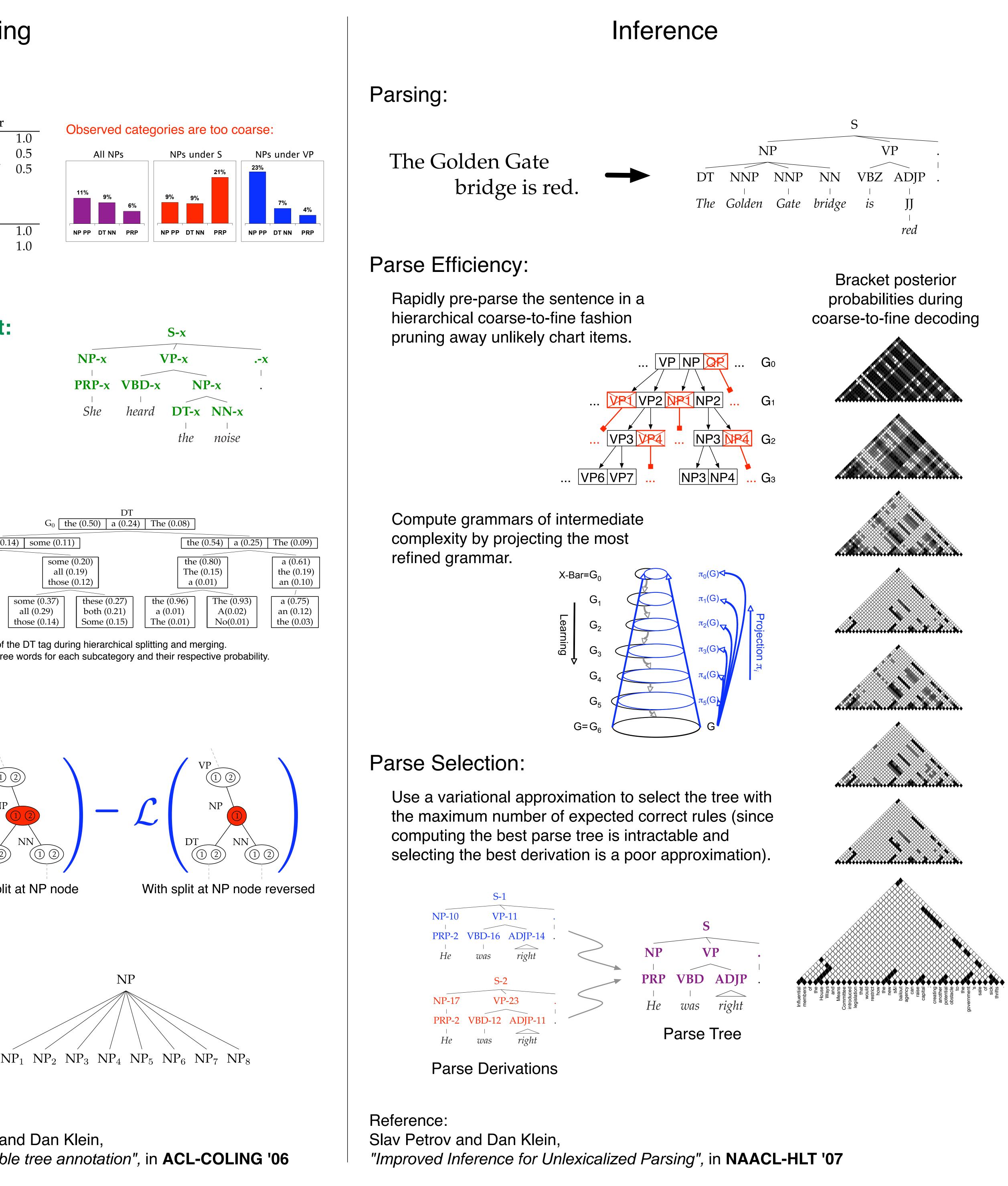
Hierarchical Splitting:



Merging:



Smoothing:



Reference:

Slav Petrov, Leon Barrett, Romain Thibaux and Dan Klein,

Learning and Inference for Hierarchically Split Probabilistic Context-Free Grammars Slav Petrov and Dan Klein University of California, Berkeley

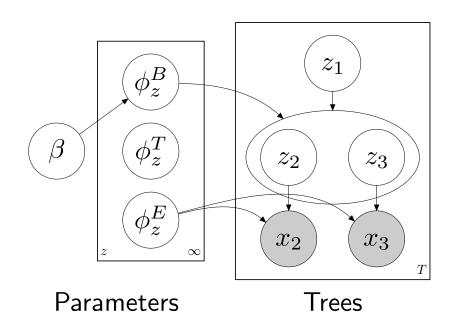
additional human input. being extremely efficient.



Learned grammars are compact and interpretable:

VBZ				DT				IN			
VBZ-0	gives	sells	takes	DT-0	the	The	а	IN-0	In	With	After
VBZ-1	comes	goes	works	DT-1	А	An	Another	IN-1	In	For	At
VBZ-2	includes	owns	is	DT-2	The	No	This	IN-2	in	for	on
VBZ-3	puts	provides	takes	DT-3	The	Some	These	IN-3	of	for	on
VBZ-4	says	adds	Says	DT-4	all	those	some	IN-4	from	on	with
VBZ-5	believes	means	thinks	DT-5	some	these	both	IN-5	at	for	by
VBZ-6	expects	makes	calls	DT-6	That	This	each	IN-6	by	in	with
VBZ-7	plans	expects	wants	DT-7	this	that	each	IN-7	for	with	on
VBZ-8	is	's	gets	DT-8	the	The	а	IN-8	If	While	As
VBZ-9	's	is	remains	DT-9	no	any	some	IN-9	because	if	while
VBZ-10	has	's	is	DT-10	an	a	the	IN-10	whether	if	That
VBZ-11	does	Is	Does	DT-11	а	this	the	IN-11	that	like	whether
	N	INP			(CD		IN-12	about	over	betweer
NNP-0	Jr.	Goldman	INC.	CD-0	1	50	100	IN-13	as	de	Up
NNP-1	Bush	Noriega	Peters	CD-1	8.50	15	1.2	IN-14	than	ago	until
NNP-2	J.	Е.	L.	CD-2	8	10	20	IN-15	out	up	down
NNP-3	York	Francisco	Street	CD-3	1	30	31		•	RB	
NNP-4	Inc	Exchange	Co	CD-4	1989	1990	1988	RB-0	recently	previously	still
NNP-5	Inc.	Corp.	Co.	CD-5	1988	1987	1990	RB-1	here	back	now
NNP-6	Stock	Exchange	York	CD-6	two	three	five	RB-2	very	highly	relativel
NNP-7	Corp.	Inc.	Group	CD-7	one	One	Three	RB-3	so	too	as
NNP-8	Congress	Japan	IBM	CD-8	12	34	14	RB-4	also	now	still
NNP-9	Friday	September	August	CD-9	78	58	34	RB-5	however	Now	Howeve
NNP-10	Shearson	D.	Ford	CD-10	one	two	three	RB-6	much	far	enough
NNP-11	U.S.	Treasury	Senate	CD-11	million	billion	trillion	RB-7	even	well	then
NNP-12	John	Robert	James		P	PRP		RB-8	as	about	nearly
NNP-13	Mr.	Ms.	President	PRP-0	It	He	Ι	RB-9	only	just	almost
NNP-14	Oct.	Nov.	Sept.	PRP-1	it	he	they	RB-10	ago	earlier	later
NNP-15	New	San	Wall	PRP-2	it	them	him	RB-11	rather	instead	because
JJS				RBR				RB-12	back	close	ahead
JJS-0	largest	latest	biggest	RBR-0	further	lower	higher	RB-13	up	down	off
JJS-1	least	best	worst	RBR-1	more	less	More	RB-14	not	Not	maybe
JJS-2	most	Most	least	RBR-2	earlier	Earlier	later	RB-15	n't	not	also

Hierarchical Dirichlet Processes as a nonparametric Bayesian alternative to split and merge:



Reference: Percy Liang, Slav Petrov, Michael Jordan and Dan Klein, "The Infinite PCFG using Hierarchical Dirichlet Processes", in EMNLP-CoNLL '07

Results

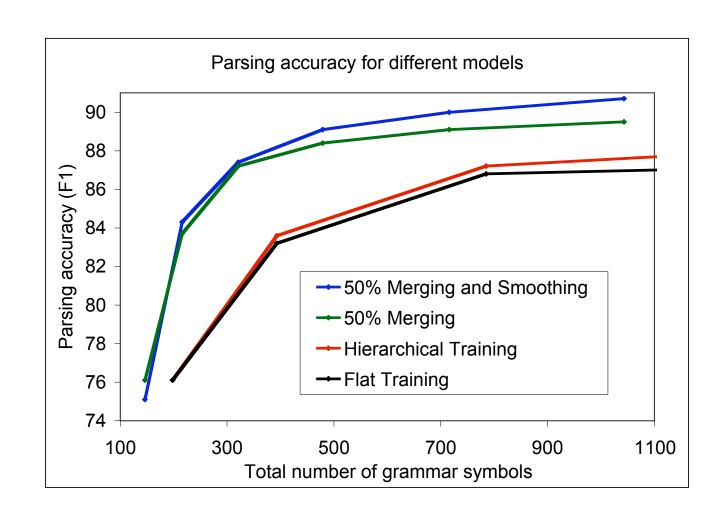
General technique for learning refined, structured models when only the trace of a complex underlying process is observed.

Learns compact and accurate grammars from a treebank without

Gives best known parsing accuracy on a variety of languages, while

Interactive demo and download at http://nlp.cs.berkeley.edu

	≤ 40 words	all				
	F_1	F_1				
JGLISH						
5	90.1	89.6				
	90.6	90.1				
ERMAN						
	76.3	-				
	80.8	80.1				
HINESE						
	80.0	76.6				
	86.3	83.4				



The most frequent three words in the subcategories of several part-of-speech tags.

Extensions

 $\begin{array}{l} \mathsf{S} \ \rightarrow \ \mathsf{NP} \ \mathsf{VP} \\ \mathsf{NP} \ \rightarrow \ \mathsf{PRP} \\ \mathsf{VP} \ \rightarrow \ \mathsf{VBD} \ \mathsf{NI} \end{array}$

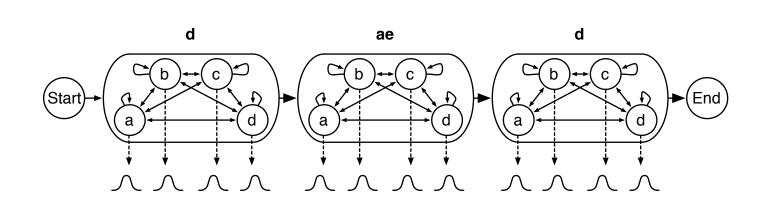
 $\mathsf{NP}\to\mathsf{DT}\,\mathsf{NN}$

 $VBD \rightarrow heard$ $\mathsf{DT} \to \mathsf{the}$

 $\begin{array}{l} \mathsf{NN} \to \mathsf{noise} \\ \mathsf{S} \to \mathsf{NP}\,\mathsf{VP} \end{array}$

 $S \rightarrow NP VP$ $NP \rightarrow PRP$ $VP \rightarrow VBD NF$ $NP \rightarrow DT NN$ $PRP \rightarrow she$ $VBD \rightarrow heard$ $DT \rightarrow the$ $NN \rightarrow noise$ $S \rightarrow NP VP$ $NP \rightarrow PRP$ $VP \rightarrow VBD NP$ $NP \rightarrow DT NN$ $PRP \rightarrow she$ $VBD \rightarrow heard$ $DT \rightarrow the$

Automatic refinement of acoustic models learns phone-internal structure as well as phone-external context:



Reference: Slav Petrov, Adam Pauls and Dan Klein, "Learning Structured Models for Phone Recognition", in EMNLP-CoNLL '07