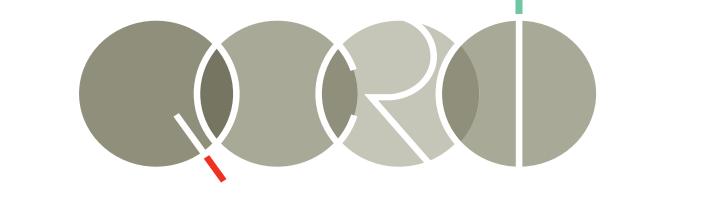
DiscoTK: Using Discourse Structure for Machine Translation Evaluation

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Discourse for MT Evaluation

- Discourse structure helps MT evaluation (Guzmán et al., 2014)
- We present two metrics that consider discourse information
 - DiscoTK_{light} only uses discourse
 - DiscoTK_{party} also uses metrics from Asiya
- DiscoTK_{party} is the best performing metric at WMT14

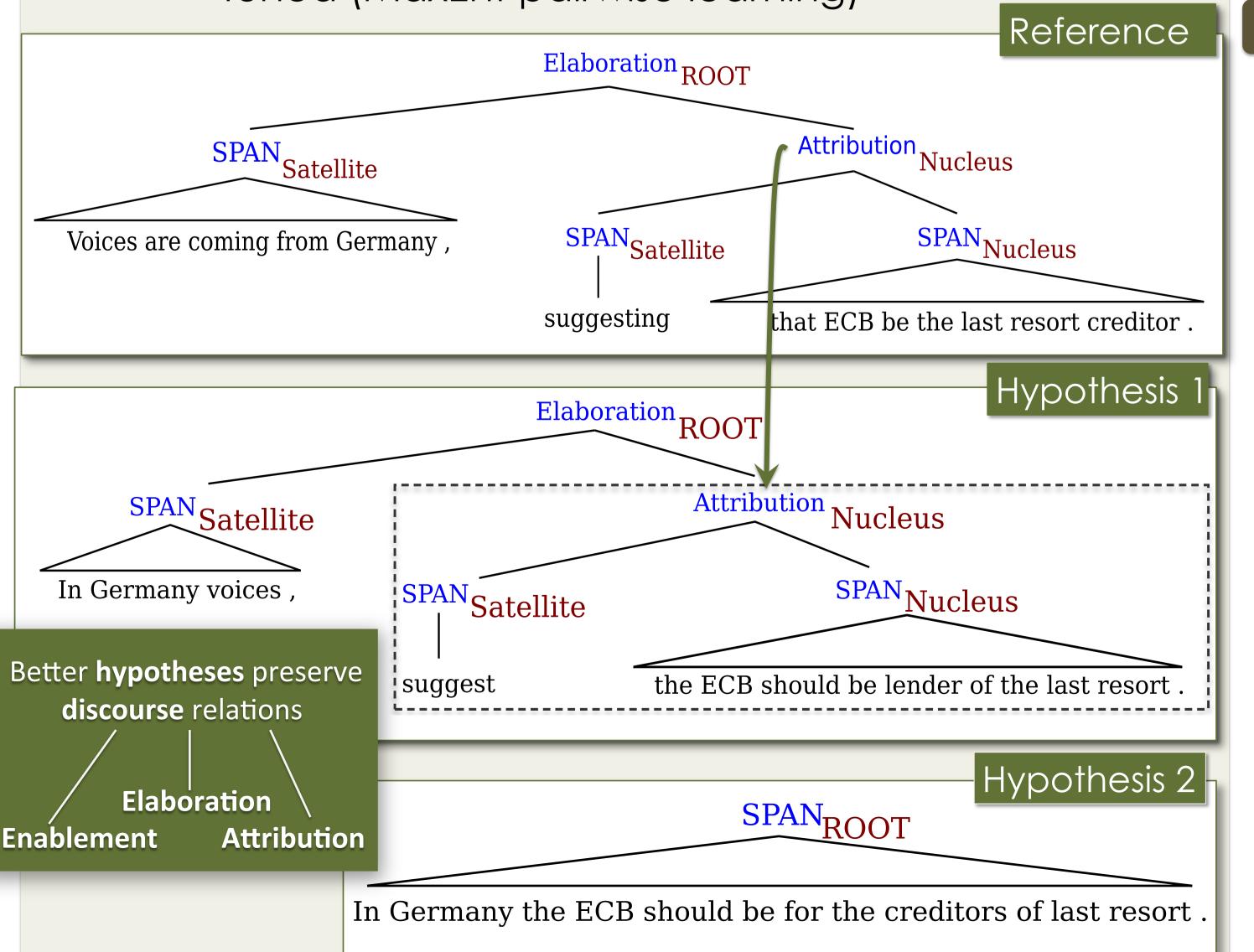
Method

Compute discourse similarity between Hyp and Ref

- RST-parse Hyp and Ref (Joty et al., 2012)
- RST trees are transformed to five different representations
- We use syntactic tree kernel (Collins & Duffy, 2002) to measure the similarity between two discourse trees
 - Use this similarity as a segment-level score
 - For system-level, average segment level scores

Combine discourse similarity with existing metrics (ASIYA)

- Uniform linear interpolation
- Tuned (MaxEnt pairwise learning)



DiscotK Discourse-based Metrics DR_{-LEX1} DiscotK_{light} ASIYA **Enablement-Root** Joint-Satellite **EDU-**Nucleus BLEU DR-NOLEX NIST **Elaboration-Nucleus** "The new .. products, **Elaboration-Nucleus** TER **Elaboration- Nucleus EDU-**Nucleus **EDU-**Satellite **EDU-**Nucleus **EDU-**Satellite TERp-A **EDU-**Nucleus **EDU-**Satellite "added Cermak and, last .. company to beter .. titles ROUGE-W METEOR-ex DR-LEX1.1 DR_{-LEX2} METEOR-pa METEOR-st **EDU-**Nucleus METEOR-sy Elaboration NUC Nucleus LEX:NUC:REL LEX:NUC LEX:REL DP-HWCM_c-4 to better .. titles to:N:ELAB better:N:ELAB published to:Nbetter:N to:**ELAB** better:**ELAB** DP-HWCM_r-4 DP-Or(*) DR_{-LEX2.1} CP-STM-4 SR-Or(*) NUC SR-Mr(*) LEX:NUC LEX:NUC:REL LEX:NUC Elaboration LEX:REL SR-Or to:Nbetter:N to:**ELAB** better:**ELAB** to:N:ELAB better:N:ELAB. to better.. DR-Or(*) DR-Orp(*)

Results	Segment Level		System Level				
VC20112		WMT12	WMT13	WMT12		WMT13	
Metric	Tuning	au	au	ρ	r	ρ	r
SEMPOS	na	_	_	0.902	0.922	_	-
SPEDE07PP	na	0.254	<u>-</u>	_	_	Y <u></u> 2	
METEOR-WMT13	na	_	0.264	_	_	0.935	0.950
$\mathrm{DISCoTK}_{light}$	Ø	0.171	0.162	0.884	0.922	0.880	0.911
	WMT11	0.207	0.201	0.860	0.872	0.890	0.909
	WMT12	_	0.200	_	_	0.889	0.910
	WMT13	0.206	_	0.865	0.871	_	_
	WMT11+12	_	0.197	_	_	0.890	0.910
	WMT11+13	0.207	_	0.865	0.871	_	-
$DiscoTK_{party}$	Ø	0.257	0.231	0.907	0.915	0.941	0.928
	WMT11	0.302	0.282	0.915	0.940	0.934	0.946
	WMT12	_	0.284		_	0.936	0.940
	WMT13	0.305	_	0.912	0.935	_	-
	WMT11+12	_	0.289	_	_	0.936	0.943
	WMT11+13	0.304	_	0.912	0.934	_	_
ASIYA	Ø	0.273	0.252	0.899	0.909	0.932	0.922
	WMT11	0.301	0.279	0.913	0.935	0.934	0.944
	WMT12	_	0.277		_	0.932	0.938
	WMT13	0.303	_	0.908	0.932	1 -	
	WMT11+12	_	0.277	_	_	0.934	0.940
	WMT11+13	0.303	_	0.908	0.933	_	_

Summary

- DiscoTK_{light} competitive at system-level
- Tuned DiscoTK_{party} improves over Asıya both at segment- and system-level
- Tuning helps consistently
- We improve over the best WMT12, WMT13 results

Tuned DiscoTK_{party} ranked 1st at WMT14

Future Work

- Learn with preference kernels from a syntacticsemantic-discourse tree representation
- Go beyond the sentence-level