

DIALOGUE-2014
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Evaluation of frame-semantic role labeling in a case-marking language

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SEMANTIC ROLE LABELING (SRL): BACKGROUND

История
не
знает
привилегий

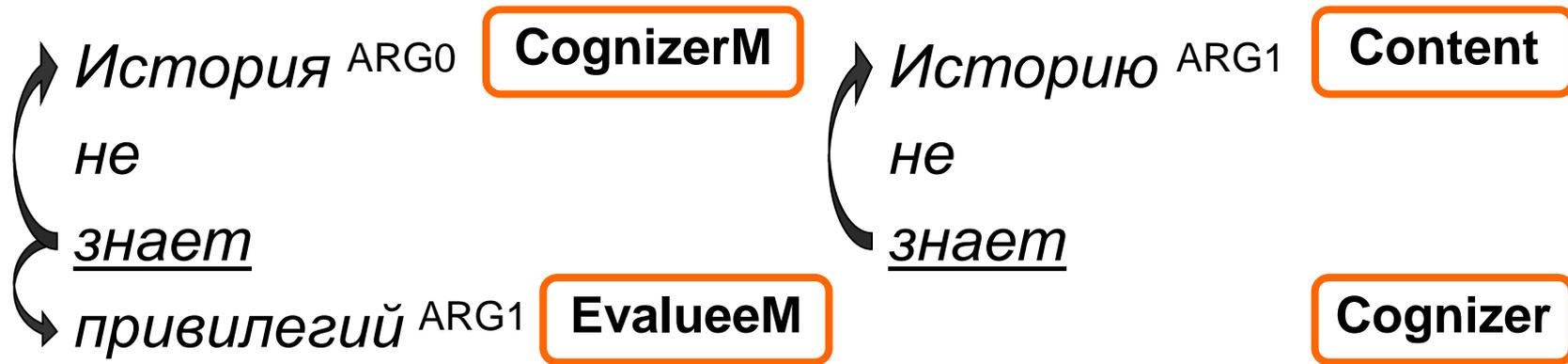
*'History does not take into
account privileges'*

Историю
не
знает

*'[She/He] does not know
history'*



SEMANTIC ROLE LABELING (SRL): BACKGROUND



'History does not take into account privileges'

'[She/He] does not know history'

CoNLL 2008 shared task (Surdeanu et al. 2008): PropBank-like SRL
SemEval 2007 (Baker et al. 2007): FrameNet-like SRL

SRL FOR RUSSIAN

- Anisimovich et al. 2012, Petrova 2013
Azarova 2008
Shelmanov & Smirnov 2014
Ermakov & Pleshko 2009
Kuznetsov 2012
... yet project in progress
- Many possible classifications of roles and frames
- Russian FrameBank as a benchmark for SRL
evaluation + class mapping



OUTLINE OF EVALUATION EXPERIMENT

- a prototype of SRL module (rule-based)
- training and test data sets from FrameBank
- how it works (cues and challenges of Russian PP SRL)
- induced roles VS gold standard roles:
 - metrics
 - the goodness of fit for non-matching pairs



DATA SOURCE: RUSSIAN FRAMEBANK

- Berkeley FrameNet : extralinguistic situations (frames) → a set of participants → lexemes.
- Russian FrameBank (www.framebank.ru): lexemes (about 2200, primarily verbs) → constructions ([Apresjan, Pall 1982] & added by annotators) → examples from RNC (about 100 for each lexeme, manually tagged)
- Theoretical framework: FrameNet and Construction Grammar + Moscow Semantic School

- *резать* ‘to cut, to carve’:
 - ✓ *Продавщица режет сыр* ‘The shop assistant is cutting cheese’; *Она режет хлеб на тонкие куски* ‘She is slicing bread (lit.: cutting bread into thin slices)’ ...
 - ✓ *Старик резал четки из кипариса* ‘The old man carved rosaries out of cypress’; *Он резал деревянные ложки ножом* ‘He carved wooden spoons with a knife’ ...
 - ✓ *У него в желудке резало* ‘He had griping pains in his stomach (lit.: It was cutting in his stomach)’

FRAMEBANK: CONSTRUCTION PATTERN

ID230. Cx name: *Pjatno vystupilo na rubaške* ['a stain appeared on the short']. Cx Pattern: Snom V na + Sloc.

Cx Item ID	Pl	Letter	Head	Phrase	Explication	Syntactic Rank	Lexico-semantic constraints	Status [obligatory / optional]
2077	1	X	Snom [Nominative case]	NPnom	substance	Subject	natural object	Oblig.
2078	2	-	vystupit' ['to appear; lit. to step forward']	-	to appear	Predicate	-	Oblig.
2079	3	Y	na + Sloc [preposition na 'on' + Locative case]	na + NPloc	location: surface	Peripheral	space and place	Oblig.

Lexical Index
of target words

Index of Morphosyntactic Items

FRAMEBANK: TAGGING

ID5914. Cx name: <i>Soberi použinat'</i> ['pick up something for supper']. Cx Pattern: Snom V Vinf.								
ID	Pl	Letter	Head	Phrase	Explication	Syntactic Rank	Lexico-semantic constraints	Status / Realization
18589	1	X	Snom	NPnom	Agent	Subject	human	Oblig.
20089	-	X	-	-	-	No	-	Omitted. Licensed by Imperative Cx
18590	2	-	<i>sobrat'</i>	-	to collect	Predicate	-	Oblig.
20090	1	-	<i>soberi.Vimper</i>	-	to collect	Predicate	-	Standard
18591	3	Y	Vinf	VPinf	what is collected	Peripheral	eat	Oblig.
20091	3	Y	<i>poest'.Vinf</i>	<i>poest' v dorogu.VPinf</i>	what is collected	Peripheral	eat	Standard
20089	2	Z	<i>nam.SPROdat</i>	<i>nam.NPdat</i>	Beneficiary	Peripheral	human	Added. Licensed by Ditransitive Dative Cx
20089	4	W	<i>v dorogu</i>	<i>v dorogu.NPdat</i>	Goal	Adjunct	abstract	Added
+								

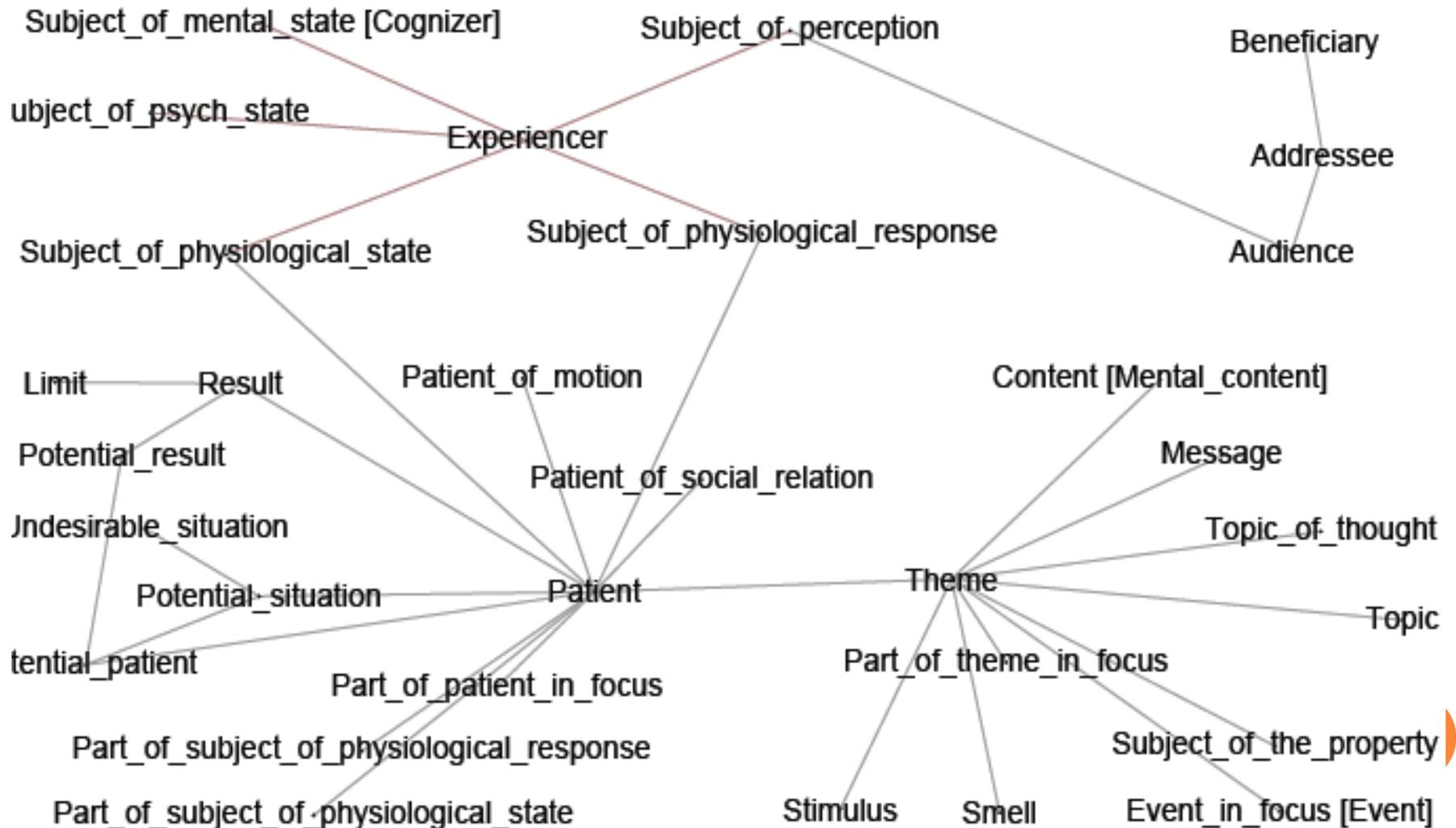


FRAMEBANK: SEMANTIC ROLES

- Different inventories of semantic roles (cf. Ch. Fillmore, Ju.D. Apresjan, E.V. Paducheva, etc.)
- FrameBank:
 - ✓ hierarchy of semantic roles → flexible search options
 - ✓ correlation between the roles and the semantic classes of verbs.
 - ✓ semantic roles graph: 96 items → 6 domains → further smaller units.



FRAMEBANK: SEMANTIC ROLES GRAPH



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SRL PROTOTYPE: PREPOSITIONAL PPs

- Four PPs:
 - za* + NPins
 - za* + NPacc
 - ot* + NPgen
 - po* + NPdat
- Very frequent (e.g. ca. 900.000 hits of “*ot* + NPgen” in the RNC).
- Highly polysemous (e.g. 14 possible roles of “*za* + NPacc”)



TRAINING AND TEST DATA

- Training data set: constructions from [Apresjan, Pall 1982]
- Test data set: constructions added by annotators

- NB type units (from dictionary), not tokens (hits from corpus)



PP	Training set: 'old' data		Test set: 'new' data	
	# constructions	# examples	# constructions	# examples
<i>za</i> + NPins	95	80	19	22
<i>za</i> + NPacc	228	223	37	51
<i>ot</i> + NPgen	266	435	70	113
<i>po</i> + NPdat	311	245	65	78
Total	900	983	191	264



SRL PROTOTYPE: PREPOSITIONAL PPs

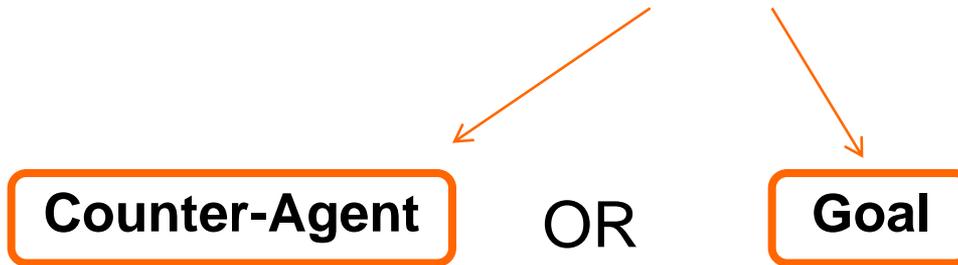
- 62 heuristics cf. [Toldova et al. 2008]
on the semantic filters in the RNC
- ✓ morphosyntactic pattern
- ✓ lexical class of the PP argument
- ✓ lexical class of the predicate
- ✓ lexical classes of other arguments
- Examples: $za + NP_{acc}$
 - ✓ $NP_{nom} V_{CHANGE POSS.} NP_{acc} za + NP_{acc} \rightarrow$ **Price**
 - ✓ $za + NP_{acc}_{TIME PERIOD} \rightarrow$ **Period**



SRL PROTOTYPE: PREPOSITIONAL PPs

- Two possible outcomes for some rules:

NP_{nom}_{ANIM} V_{MOTION} za + NP_{ins}_{ANIM}



- Counter-Agent: *Milicioner pobezhzal za prestupnikom* 'A policeman ran **after an offender**'
- Goal: *Mal'chik pobezhzal v bol'nicu za vrachom* 'A boy ran to hospital **to call the doctor**'



SRL CUES

- Semantic class of a verb
 - ✓ *Beshus'*_{EMOTION} **za doch** *moju* 'I am in a rage **because of my daughter**' → Reason for an Emotion, cf. *bojat'sja* 'to be afraid', *bespokoit'sja* 'to worry about sth.'
- Semantic class of a participant
 - ✓ **Po radio**_{COMMUNIC. FACIL.} *igrala muzyka* 'There was music broadcast (lit.: played) **by radio**' → Manner. cf. *vystupat' po televizoru* 'to speak **on TV**', *poslat' dokumenty po pochte* 'to send documents **by post**'



CHALLENGES FOR SRL

- Pragmatic factors, cf. Counter-Agent vs. Goal of motion.
- No clear semantic constraints on a verb or on its arguments.
 - ✓ “*po + NPdat_{???}*” for Reason: *Rasskaz byl prochitan po ego pros'be* ‘The story was read at his request’ → ~~Information Source~~, Reason, cf. *zhenit'sja **po ljubvi*** ‘to make a love-match (lit.: to get married **due to love**)’, *uvolit' **po sokrascheniju shtatov*** ‘to discharge sb. **on grounds of staff reduction**’, *sidet' zdes' **po drugomu delu*** ‘to stay here **on some other business**’
- Metonymies of concrete nouns
 - ✓ *Ego nevozmozhno otorvat' **ot knigi*** ‘It is impossible to divert his attention **from the book** (lit.: to tear him=it from the book)’ → ~~Patient & Location~~ Content of Action 

EVALUATION METRICS

1. Precision, Recall, F1

$$P = \text{\#Matches} / \text{\#E_Answers}$$

$$R = \text{\#Matches} / \text{\#G_Answers}$$

2. Cluster purity and collocation, F1 (Lang & Lapata 2011)

$$Pu = \frac{1}{n} \sum_{i=1}^{n_C} \max_{j=1, \dots, n_G} |C_i \cap G_j| \quad Co = \frac{1}{n} \sum_{j=1}^{n_G} \max_{i=1, \dots, n_C} |C_i \cap G_j|$$

3. Repulsion: how likely are the roles to distinguish frames

$$\text{repulsion} = \frac{\text{\#Verbs}(\text{RoleE_RoleG OR RoleE!RoleG})}{\sqrt{\text{\#Verbs}(\text{RoleE}) \times \text{\#Verbs}(\text{RoleG})}}$$

4. Distance between roles in semantic role graph



RESULTS: PRECISION

PP	Total amount of new patterns	'Strong' matching (the role is identified correctly and unambiguously)	'Weak' matching (one of the answers is correct)	P_{strong}	$P_{\text{strong+weak}}$
<i>za</i> + NPins	19	9	7	0.47	0.84
<i>za</i> + NPacc	37	22	11	0.59	0.89
<i>ot</i> + NPgen	70	41	24	0.59	0.93
<i>po</i> + NPdat	65	32	25	0.49	0.88
Total	191	104	67	0.54	0.90



THE GOODNESS OF FIT AND REPULSION

Matching Evaluation (human)	Role E	Role G	#Verbs (RoleE)	#Verbs (RoleG)	#Verbs (RoleE! RoleG)	#Verbs (RoleE! RoleG)
Good	Source	Reason	12	266	3	
[Source↑External_cause↑Age]						
<i>Lovit' kaif [ot knig]</i> 'To be in high from books'.						
Good	Path	Patient	105	712	46	
[Path↑Locat]						
<i>On breidit i mechetsja golovoj [po perekladine]</i> 'He raves, tossing his head over						
Good	Property	Reason	175	266	31	
<i>Ego zabrali [po nacional'nomu piznaku]</i> 'He was arrested on ethnic grounds'.						
Average	Term	Time_point	52	42	6	
<i>Vstrecha prodilas' [za polnoch]</i> 'The meeting lasted past midnight'.						
		Target_				

THE GOODNESS OF FIT AND REPULSION

	#Verbs (RoleE)	#Verbs (RoleG)	#Verbs (RoleE! RoleG)	#Verbs (RoleE+ RoleG)	Repulsion	Same domain
Role G						
Reason	12	266	3	0	0.05	NO
[Source↑External_cause↑Agent]↑Root↓[Setting↓Reason] high from books'.						
Patient	105	712	46	3	0.18	NO
[Path↑Location↑Setting]↑Root↓[Patient] voj [po perekladine] 'He raves, tossing his head over the crossbar'.						
Reason	175	266	31	5	0.17	(YES)
[Property↑Setting↓Reason] omu piznaku] 'He was arrested on ethnic grounds'.						
Time_point	52	42	6	2	0.17	YES
[Term↓Time_point] och] 'The meeting lasted past midnight'.						
Target_						

CONCLUSION

- Repulsion correlate quite well with the split between Average and Bad matches (repulsion threshold .20) ✓
- ... but split between Good and Average matches:: FAILED ✗
- Same-domain VS Good_Average_Bad matches:
 - all Bad matches do not share the same domain but this is not sufficient ✗
- Graph-based distances: FAILED ✗
- *Future development*
- Other SRL modules
- Token frequencies
- More (other than IS-A) types of edges in the SR graph



СПАСИБО !

