

From Neural Activation to Symbolic Alignment

Alexander Mehler, Andy Lücking, Peter Menke

Dialogue Network:

Data

Classification

Conclusions

From Neural Activation to Symbolic Alignment A Network-Based Approach to the Formation of Dialog Lexica

Alexander Mehler Andy Lücking Peter Menke

Text Technology Group, Goethe-Universität Frankfurt am Main CRC 673 "Alignment in Communication", Bielefeld University

International Joint Conference on Neural Networks San Jose, California, July 31 – August 5, 2011



MercuryNews.com

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Egyptian Museum in San Jose invites public to Epagomenal festival July 16 and 17

Dress like an Egyptian and you can win a prize at the annual Egyptian Epagomenal Festival July 16 and 17 at the Rosicrucian Egyptian Museum. The two days of events feature special activities for adults, including lectures, while children can make cornhusk mummies, jewelry, senet game boards and shabtis, the small figures buried with pharaohs in their tombs.

Mary Gottschalk, Posted: 07/11/2011 07:31:43 PM



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Coherence - by folklore, and in San Jose

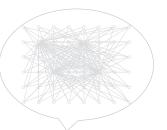


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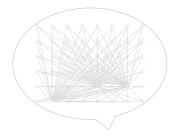
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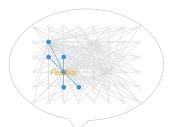


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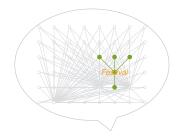
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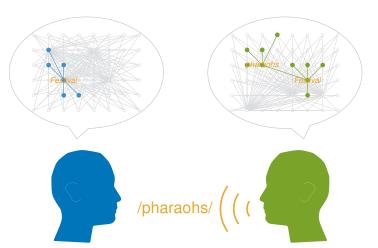


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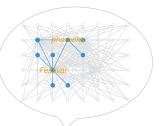


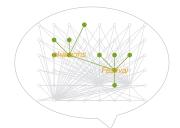
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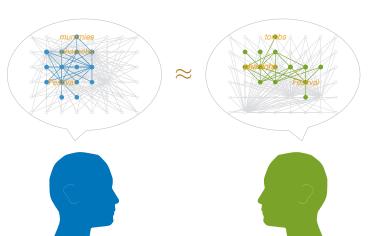


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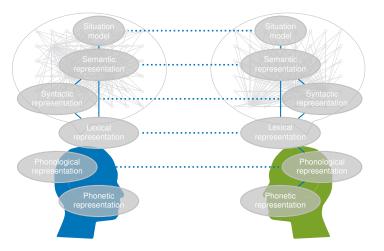
Interactive Alignment Model Pickering and Garrod (2004)

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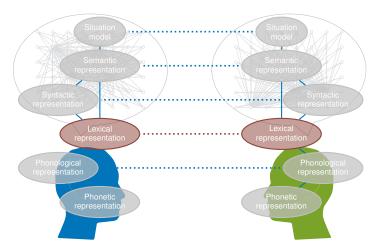
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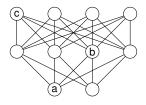
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- is non-intentional
- brings about routinized repetitive behavior
- different and differentiable from non-alignment





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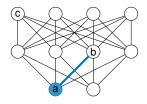
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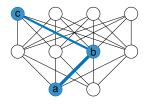
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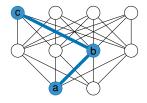
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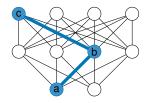
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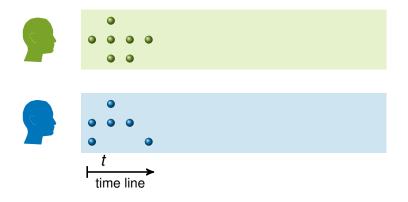
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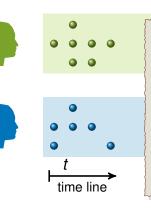
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Networking of dialogue lexica according to referential use of lexical items



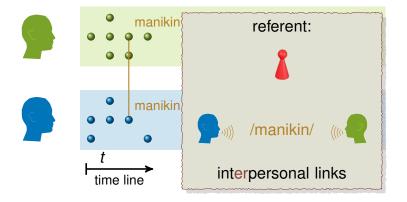
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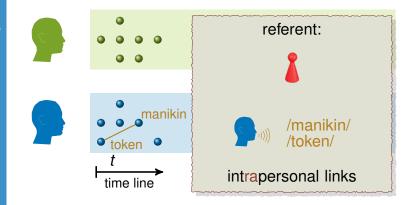
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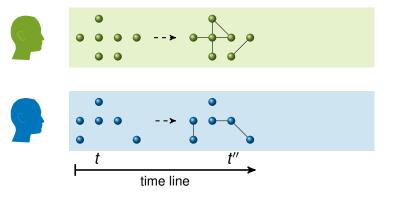
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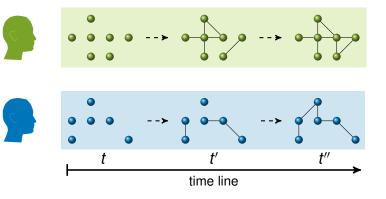
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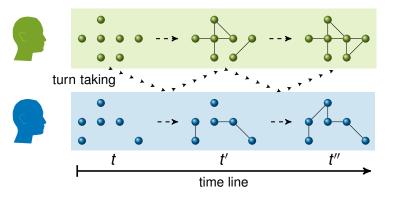
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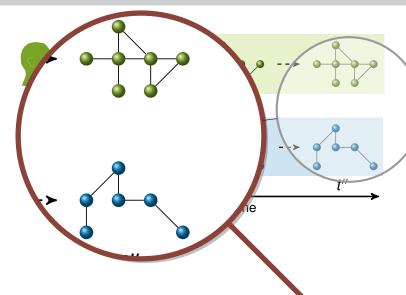
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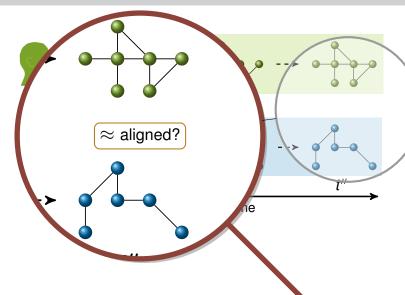
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Graph Distance Measurement Mehler, Lücking, and Weiß (2010)

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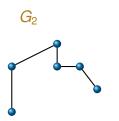
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$$D(G_1, G_2)|_{(v,w)} = 1 - \frac{I(v; w)}{\max\{H(v), H(w)\}} \in [0, 1]$$



Graph Distance Measurement Mehler, Lücking, and Weiß (2010)

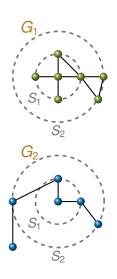
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$$D(G_1, G_2)|_{(v,w)} = 1 - \frac{I(v; w)}{\max\{H(v), H(w)\}} \in [0, 1]$$

$$\frac{I(v; w) =}{\sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \frac{|S_{i,j}(v, w)|}{n} \log_2 \frac{\frac{|S_{i,j}(v, w)|}{n}}{\frac{|S_i(v)|}{n} \cdot \frac{|S_j(w)|}{n}}$$

$$H(v) = -\sum_{i=0}^{n-1} \frac{|S_i(v)|}{n} \log_2 \frac{|S_i(v)|}{n}$$



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Experimental study on alignment of object naming

- controlled face-to-face dialogues
- task: cooperative, interchanging positioning of objects on a map



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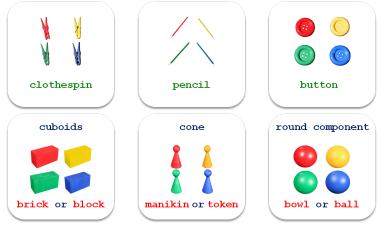
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Critical objects to be Named and Located





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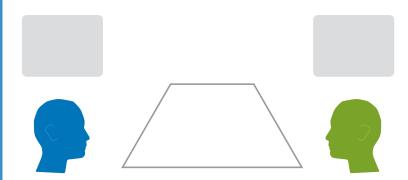
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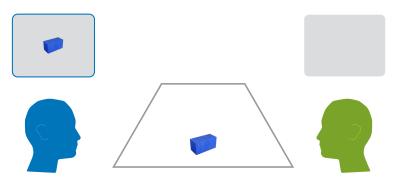
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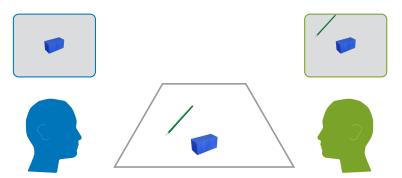
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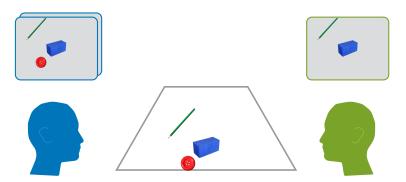
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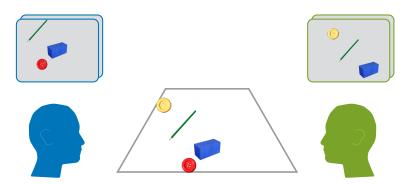
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Jigsaw Map Game Weiß et al. (2008)

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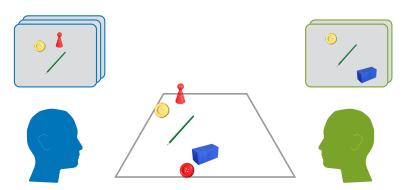
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Playing the Game!





Jigsaw Map Game Weiß et al. (2008)

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Some Facts and Figures

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level	total number	average per dialog
words	93,120	1,501.935
turns	28,380	457.742
events	1,731	27.919
event phases	5,153	83.113
lexops	4,415	71.210
repairs	3,327	53.661



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Features of Alignment

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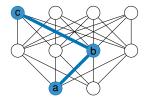
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Alignment

- is operative on mental representations
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- different and differentiable from non-alignment





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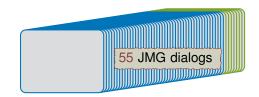
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47 instances of alignment

8 instances of non-alignment



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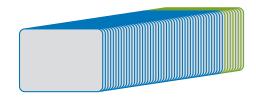
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Experimentation Quantitative Network Analysis (QNA) Mehler (2008

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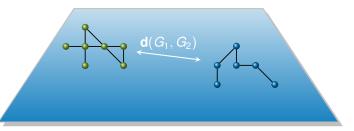
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n-dimensional space





Experimentation Quantitative Network Analysis (QNA) Mehler (2008

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Procedure	F-score	Feature Selection	
QNA[Mahalanobis,complete]	.92284	47 / 103	
QNA[Mahalanobis,complete]	.90654	46 / 103	
QNA[Mahalanobis,Ward]	.89578	49 / 103	
QNA[correlation,complete]	.861	49 / 103	
random baseline known- partition	.75073		
random baseline equi- partition	.61788		



Results

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- QNA classifications outperform the random baselines
- QNA classification based on TⁱT_AN is feasible (≥ 90%)
- No QNA classification separates the data perfectly

Observation

In any case, multiple topological indices are needed

What is the relationship among the topological indices?



Results

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Correlations

0.35

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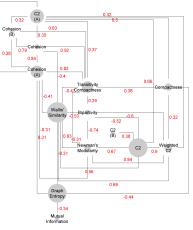
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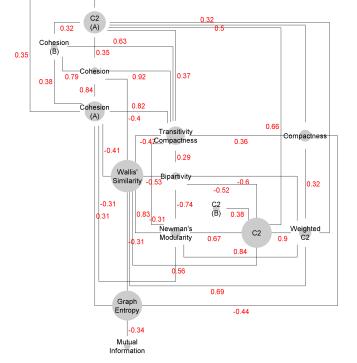
Alignment Dialogs



Non-Alignment Dialogs



non-alignment instances





From JMG to SaGA

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JMG dialogues are

- strictly task-oriented
- highly regimented
- communicatively symmetric

Is classification also feasible on more free dialogues?



Application: Route Directions Mehler, Lücking, and Menke (2011)

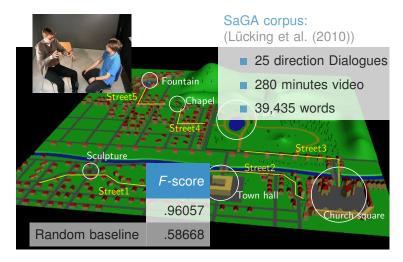
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Summary

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A network model of dialog that

- has a neural interpretation
- accounts for structure formation (alignment)
- provides a test bed for exploring relationships between topological indices



Alignment in Communicatior

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