

Interactive Narration Requires Interaction and Emotion

A. Pauchet⁽¹⁾, F. Rioult⁽²⁾, E. Chanoni⁽³⁾, Z. Ales⁽¹⁾ and Ovidiu Şerban⁽¹⁾
email: alexandre.pauchet@insa-rouen.fr

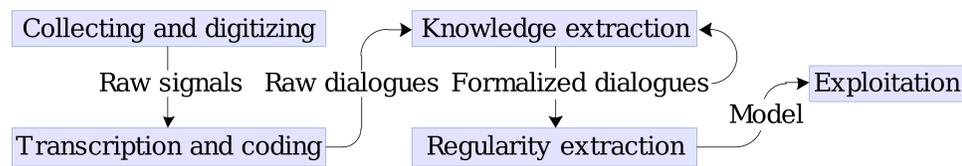
Keywords

Dialogue modelling; Knowledge extraction; Narrative ECA

Abstract

This work shows how interaction is essential for storytelling with a child. A corpus of narrative dialogues between parents and their children was coded with a mentalist grid. The results of two modelling methods were analysed by an expert in parent-child dialogue analysis. The extraction of dialogue patterns reveals regularities explaining the character's emotion. Results showed that the most efficient models contain at least one request for attention and/or emotion.

Proposed method



Corpus of narrative dialogues

In this study, thirty dialogues between children and parents (ages: 3, 4 and 5) were recorded during emotional story telling situations. These records have been transcribed and coded with a mentalist grid to capture information about the mental states (beliefs, emotions, ...) contained by the various utterances.

Parent-Children story telling



Corpus of transcribed and annotated dialogues

Line	Speaker	Utterance	Annotations
25	P	Don't worry	A P E - -
26	P	So they hide	A P B - -
27	P	They are looking for	A - F - -
28	P	who could have taken away the crown	Q - F - -
29	C	It's in ! The crown is inside the box !	A - F - -
30	P	So they are suspecting a lot of people, Cornelius, Celeste, the old lady	A P Y C J
31	P	Who could have stolen the crown?	Q - F - -
32	C	The crown, it's in!	A - F - -
33	P	Do you believe it?	Q H K - -
34	C	Yes	A - F - -
35	P	But Babar doesn't know that it's in	A P N O J
36	P	So he says that the crown is a bomb	A P N C J
37	P	or I don't know what else	A R N - -

Clustering methods applied to dialogue patterns

		Number of clusters found					
		5	20	50	80	116	150
Single-Link	(Florek et al., 1951)	41	97	183	270	320	360
CHAMELEON	(Karypis et al., 1999)	458	605	628	-	-	-
ROCK	(Guha et al., 2000)	520	600	621	626	629	630
Unnormalized spectral clustering	(Von Luxburg, 2007)	277	658	563	155	194	226
Shi and Malik spectral clustering	(Von Luxburg, 2007)	524	615	628	631	631	632
Jordan and Weiss spectral clustering	(Von Luxburg, 2007)	555	616	628	630	631	632
Affinity propagation	(Frey and Dueck, 2007)	-	-	-	-	632	-

Example of dialogue pattern

Dialogue B3 (4 years old)						Dialogue C8 (5 years old)				
	A	x	x	x	x	A	x	x	x	x
1	A	P	E	C	J	A	P	x	C	J
2	Q	x	x	x	x	A	x	x	x	x
3	A	x	x	x	x	A	x	x	x	x
4	A	x	x	x	x	A	x	x	x	x
5	A	x	x	x	x	A	x	x	x	x
6	A	P	x	C	J	A	P	x	C	J
7	A	P	E	x	x	Q	P	E	x	X
8	Q	x	x	x	x	D	x	x	x	X
9	A	x	x	x	x	A	P	E	x	x

Prediction of interaction

3 years old			4 years old			5 years old		
efficiency	model	support	efficiency	model	support	efficiency	model	support
3.2	E-Q	10,4%	2.1	D-Q	14,9%	1.9	Q	35,4%
3.4	D-Q	16,8%	2.2	E-Q	7,5%	2.2	E-E	9,1%
3.5	J-Q	9,6%	2.2	Q-Q	12,7%	2.6	J-D	8,1%
3.5	D-Q-Q	9,6%	2.6	D-E	10,4%	2.7	E-D	6,1%
3.5	E-J	8,8%	2.8	D-D	11,2%	3.1	J-E	8,1%
4.3	D-E	12,8%	3.5	J	14,9%	3.4	V	13,1%
4.3	D-J	8,0%	3.8	B	7,5%	3.7	D-E	7,1%
5.4	B	10,4%	4.0	E-E	7,5%	3.8	J-J	6,1%
5.6	V	13,6%	4.1	E-D	6,7%	4.1	E-J	7,1%
			4.3	V	6,7%			

The previous table highlights the important following facts :

- 1) regardless of the age, sequences with justifications were frequently associated with various indexes (emotion, request for attention or question). In this context, the child's interaction came after 3.1 to 4.3 sentences after the model;
- 2) the quickness of the interaction decreased with the age, from 3.2 to 1.9 sentences;
- 3) the number of sequences with emotion is merely equivalent for all ages. The older the child is, the more different the sequences of emotion are. Complex sequences (emotion and justification: J-E or E-J) appear only with the oldest children;
- 4) except with request for attention, the most efficient models (in red in Table 4) always contained emotion (E-Q or E-E).

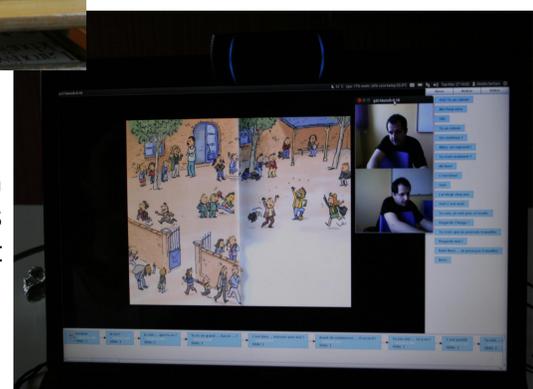
Evaluation: avatar impact on children interaction



Interaction with avatar or in a visio-conference mode is similar in term of semantics, but the modalities used are different.

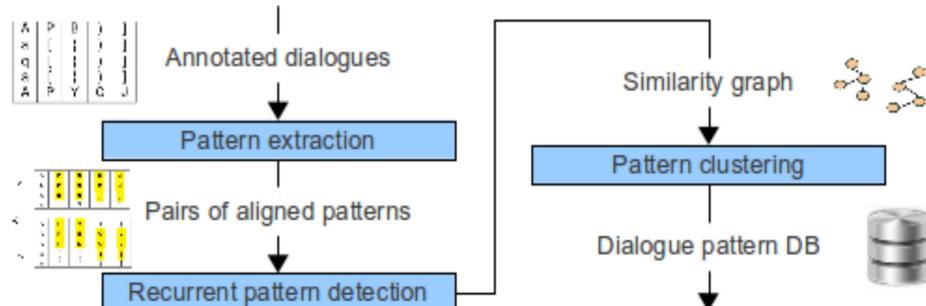
We plan to integrate the dialogue patterns we extracted into an ECA for a final evaluation.

This work benefits from a financial support from CNRS as a PEPS INS2I-INS2S project called ACAMODIA.



Regularity extraction for dialogue modeling

To collect significant dialogue patterns from the annotated dialogues, we applied a regularity extraction step based on matrix alignment using dynamic programming followed by a clustering step using machine learning heuristics to group and select the recurrent dialogue patterns. The clustering process is applied on a similarity graph computed during the matrix alignment.



⁽¹⁾Laboratoire d'Informatique, de Traitement de l'Information et des Systemes (LITIS), INSA de Rouen, Avenue de l'Universite - BP 8, 76801 Saint-Etienne-du-Rouvray, France

⁽²⁾Laboratoire GREYC, Université de Caen Basse-Normandie, Campus Côte de Nacre, Boulevard du Maréchal Juin, 14032 CAEN cedex 5, France

⁽³⁾Laboratoire PSY-NCA, UFR Psychologie, Sociologie et Sciences de l'Education, Rue Lavoisier, 76821 Mont Saint Aignan, France