[Proceedings of a Workshop on Machine Translation, July 1990, UMIST]

Automatic Interpreting Telephony Research at ATR Tsuyoshi MORIMOTO

ATR Interpreting Telephony Research Laboratories Seika-cho, Souraku-gun, Kyoto 619-02, Japan morimoto@atr-la.atr.co.jp

An automatic telephone interpretation system will transform a spoken dialogue from the speaker's language to the listener's automatically and simultaneously. It will undoubtedly be used to overcome language barriers and facilitate communication among the people of the world.

ATR Interpreting Telephony Research project was started in 1986. The objective is to promote basic research for developing an automatic telephone interpreting system. The project period is seven-years.

(1) Continuous Speech Recognition

A grammar-driven new continuous speech recognition mechanism, which is called HMM-LR, has been developed. It combines HMM phone model and generalized LR parsing mechanism.

A generalized LR parser can handle arbitrary context free grammar. Parsing is guided by an LR table which is created from pre-defined grammar, and proceeds left-to-right without backtracking.

Input spoken sentences are those which are uttered phrase by phrase (a phrase means Japanese Bunsetsu). Then, recognition of an utterance is performed for each phrase independently, and the output result is a sequence of candidates corresponding to each phrase (this output is called a "phrase lattice"). To do this, a phrasal grammar for Japanese is defined and converted to the LR table. The lexical items are embedded in the grammar rules, that is, terminal symbols are defined as a sequence of several phonemes.

The system determines which phones should be recognized next by referring the LR table, and then verifies their existence in the input signal by comparison with the corresponding HMM phone model.

If several phones are predicted, it verifies their existence and keeps all possible parsing trees. In this process, the probabilities for each partial parsing tree are calculated and only those with a higher probability are kept using beam search technique. The recognition process proceeds until the end of the phrase, and those which are grammatically accepted and having a higher probability score are output as the final candidates.

(2) Spoken Dialogue Translation

A dialogue sentence is composed of two parts, i.e. a propositional content part and an illocutionary force part. The former corresponds to the content of the sentence and the latter to the speaker's intention or attitude to the hearer such as honorifics. The latter is chiefly expressed in a final predicate phrase by attaching several auxiliary verbs or sentential final particles. The analysis process analyzes an input sentence and extracts semantic expressions for these two parts. The propositional part is described in terms of language dependent concepts, and the illocutionary force part in terms of language independent concepts. The transfer process converts only the propositional part to the target language concepts, and the generation process merges it with the illocutionary force part and then generates a surface expression. In a sense, this method can be considered to be intermediate between the transfer approach and the interlingual approach.

(3) Integration of Speech Recognition and Language Analysis

In spoken language processing, ambiguity of input data itself becomes a crucial problem. To resolve this ambiguity, some linguistic information should be used. On that time, it is necessary to use proper information at appropriate points to avoid an unnecessary increase in processing time. The proposed method is composed of three stages. In the speech recognition stage, syntactical knowledge for phrases is used as described above. Output from this stage is several hypotheses (candidates) for each phrase. In the next stage, Kakariuke dependency between phases is checked to filter out implausible candidates. At the final stage, sentence analysis, the most plausible sentence is selected by checking strict syntactico-semantic and pragmatic appropriateness or by evaluating the preference of sentence structure.

An experimental system called SL-TRANS has been implemented. It recognizes input Japanese speech, translates it to English and outputs synthesized English voice. Some experiment results are reported in the presentation.

Automatic Interpreting Telephony Research at ATR

Tsuyoshi MORIMOTO

ATR Interpreting Telephony Research Laboratories Seika-cho, Souraku-gun, Kyoto 619-02, Japan morimoto@atr-la.atr.co.jp

ATR Interpreting Telephony Research Project

- (1) ProjectPlan
- (2) Requirement for Spoken Language Interpretation
- (3) Componential Technologies
 - (a) Continuous speech recognition
 - (b) Language translation

(c) Integration of speech recognition and language analysis

(4) Experimental Japanese-English Spoken Language Translation System (SL-TRANS)

(5) Experiment Result and Evaluation

(6) Conclusion and Future Direction

Project Plan

(1) 7 Years Project Period (1986.4~1993.3)

(2)Promote Basic Research for Developing an Automatic Telephone Interpreting System

(3) Develop a Prototypical System

between Japanese and English large vocabulary (around 1,000 or more) goal-oriented task

Requirements

(1) Accurate Speech Recognition and High Quality Language Translation

- (a) No existing of pre- or post-editing
- (b) Mono-lingual Users
- (2) Interpretation of Spoken Dialogue
 - (a) Ellipsis/anaphora resolution
 - (b)Interpreting speaker's intention

(3) Integration of Speech Recognition and Language Processing

(a)Ambiguity resolution by using proper linguistic information

(b)Optimum (efficient) interface



Figure 1 Block Diagram of the Speech Translation System

Continuous Speech Recognition

(1)HMM phone Model

(a)Generalized LR Parser is used to predict next possible phones (2)Grammar-Driven Speech Recognition

(3) Grammar for Japanese Phrase (Bunsetsu) is Defined

(4) Input is sentence uttered phrase by phrase

Recognition is performed for each phrase.

(5) Verify by comparison with HMM



Language Analysis

(1) Lexicon-based JPSG Grammar and Unificationbased Analysis Method

(a) To treat syntactic, semantic and even pragmatic constraints in uniform way

(b) To treat fragmental or complex modal expression in final predicate

intentions, honorifics, indirect-expressions etc.

(2) Ellipsis/Anaphora Resolution by Using Pragmatics



Figure 3 The Outline of the Analysis Process

Resolution of Zero-pronoun on Use of Pragmatics

Ex-1:

o-namae wo oshiete	itadake	masu	ka			
(your) name tell	could	POL	Q			
(Could <u>you</u> tell <u>me your</u> name, please ?)						

I able I Typic	al illocutionary Force Type (IFT)
Type	Explanation
PHATIC	Phatic expression such as open or close dialogue (Hello, Thank you)
INFORM	Inform a hearer of some facts
REQUEST	Request a hearer to do some action (Please tell me…)
QUESTIONIF	Yes/No question
QUESTIONREF	WH question

f Ë β 1 LLL L Table 1 T.

intic Representation in Feature Structui	Ģ
intic Representation in Feature Struct	<u>i</u>
intic Representation in Feature Stru	ษ
intic Representation in Feature S	ţ
intic Representation in Feature	S
intic Representation in Featu	ПЭ
intic Representation in Fea	ät
intic Representation in F	ë
intic Representation ii	<u>с</u>
intic Representatior	.=
intic Representati	0
intic Represent	ati
intic Represei	nti
intic Repre	sei
intic Rep	ě
intic Re	de
Intic	Ř
Ē	Ë
m	л.
Ĕ	Ĕ
Se	Se

[object [[relation MODERATE] [object [[relation DESIRE] [experiencer ?speaker] [object [[relation SURU-1] [agent ?speaker] [[relation INFORM] Ex-2:

:





Candidate Filtering Using Inter-phrase Kakariuke Dependency



K(X,Y) = F(X,Y) - wl X D(X,Y) + w2 X S(Y)

where

K(X,Y): matching score between phrase X and Y F(X,Y): frequency of appearance in the Kakariuke dictionary

D(X,Y): distance between X and Y in the input data

S(*Y*): speech recognition score

wl,w2: weights determined experimentally

	•	omochi-deshou-ka	have - POL - Q	ttion form?)
Ex-4:	Original Sentence;	tourokuyoushi-wa sudeni	registration-form-TOP already	(Do you already have a registra

	omoi-mashou-ka	think-POL-INT-Q	omochi-deshou-ka	when have POL	omochisi-mashou-ka	0BJ2 bring - POL-INT-Q	
from HMM-LR);	sudeni	already	-ga itsu-ni	rm-SBJ	-wa sen-ni	-TOP thousand-(
Input (Output	tourokusi-ta	register-PAST	tourokuyoushi	registration-fo	tourokuyoushi	registration form	

о ,

> omoti-deshou-ka <u>Output;</u> tourokuyoushi-wa sudeni

lysis
Ana
uage
ang
ring l
e Du
srenc
Prefe
ence
Sente

(a) If several sentential candidates remain after Kakariuke filtering, their syntacticosemantical legitimacy is checked by the language analyzer. (b) Nevertheless, if there are still several candidates, the sentence which has the highest preference score is selected.

 $P(X) = a1 \times S(X) - a2 \times Nt(X) - a3 \times Nu(X)$ where

P(X): preference score of sentence X

S(X): speech recognition score

Nt(X): number of nodes of syntax tree

Nu(X): number of unfilled obligatory elements

a1,a2,a3:weights determined experimentally

Nt and Nu reflect the heuristics that a simpler sentence is more plausible.

		(2-1) saremasu	(someone) do (something) / (something) is done	(2-2) sitsureisimasu	goodbye
Ex-5:	Input;	(1-1) soredewa	then	(1-2) sureba	if (someone) do (something)

Evaluation of Preference;

Of these, the combination of (1-1) and (2-2) has the fewest nodes and unfilled obligatory elements and thus, is selected.

<u>Output;</u> soredewa sitsureisimasu



Figure 5 Configuration of SL-TRANS

able 2 Experiment Results of SL-TRANS	Specific speaker Number of sentences: 37 Number of phrases: 83 Average number of phrases/ sentences: 2.2	Bunsetsu recognition rate 87% for the 1st rank (0、S Y ^{2、ン} f ハ) A 96% for the top 5 ranks Average number of output candidates: 4.6 ^{2.2} = 28.7 →number of sentential candidates: 4.6 ^{2.2} = 28.7	Average number of selected candidates/ phrases: 1.5 →number of sentential candidates: 1.52.2 = 2.4	Number of sentences selected correctly: 34 Number of sentences translated correctly: 34
Ľ	Input	HMM-LR	Kakariuke Filtering	Language Translation (System total)

TRANC

Figure 6 Translation Examples

Conclusion and Future Direction

(1) Enhancing the Efficiency of Speech Recognition (a) Introducing more precise HMM model

multiple code-book context-dependent model (b) Introducing Sentential Grammar in Speech Recognition (c) Introducing Stochastic Grammar in Speech Recognition (2) Integrating Kakariuke dependency filtering (or some semantic based candidate filtering) function with either a speech recognizer or language analyzer rather than making it as an independent process.

(3) Improving the efficiency of the language analyzer.

(4) Extending translation both in vocabulary size and sentence expression variation.

(5) Introducing a contextual processing for the sake of enhancing the ability of: (a) Ellipsis/anaphola resolution.

(b) Speech recognition ambiguity resolution.